

18.06

4/13

Exam 2

avg = 75 harder than he meant

$\sigma = 18$ $n = 56$

Diff eq $\frac{du}{dt} = Au$ $u(t) = e^{At} u(0)$

$e^{At} = I + At + \frac{(At)^2}{2!} + \dots$ has eigenvalues

$1 + \lambda t + \frac{(\lambda t)^2}{2!} + \dots$

(same vectors as A)

$$Ax_k = \lambda_k x_k \Rightarrow AS = SA$$

$$S = \begin{bmatrix} x_1 \\ \vdots \\ x_n \end{bmatrix} \quad A = S \Lambda S^{-1}$$

Diagonalization - the building block of everything

2

$$A^k = S \Lambda^k S^{-1}$$

$$A^k u_0 = S \Lambda^k S^{-1} u_0 \\ = c_1 \lambda_1^k x_1 + \dots + c_n \lambda_n^k x_n$$

See the point of eigenvalues + eigenvectors

No + A^k ← discrete steps

But e^{At} ← continuous

↳ the exponential

(wrote down on page 1)

$$e^{At} x_1 = e^{\lambda_1 t} x_1$$

$$e^{At} x_n = e^{\lambda_n t} x_n$$

$$\sum_0^{\infty} \frac{x^n}{n!} = e^x$$

"exponential series"

$$\sum_0^{\infty} x^n = 1 + x + x^2 + x^3 + \dots = \frac{1}{1-x}$$

"geometric series"

3

$$X = i\theta \rightarrow (\cos\theta + i \sin\theta) \quad \leftarrow \text{Euler}$$

Can let x be a matrix
 ↳ Series still converges!

e^A has a zero eigenvalue when
 ↳ never! - its always invertible

$$(e^{At})^{-1} = e^{-At}$$

$$\left[\frac{1}{e^x} = e^{-x} \right]$$

as long as only one matrix A

if 2 matrices \rightarrow not same eigenvectors

$$A = \begin{bmatrix} 0 & 1 \\ -1 & 0 \end{bmatrix}$$

need its eigenvalues

anti symmetric \leftarrow note

$$A^T = -A \quad \downarrow \text{So } \lambda \text{ are pure imaginary}$$

(4)

Symmetric
 $A^T = A \rightarrow \lambda$ are all real

So eigenvalues are

$$A - \lambda I = \begin{bmatrix} -\lambda & 1 \\ -1 & -\lambda \end{bmatrix}$$

$$= \lambda^2 + 1 = 0$$

$$\lambda = i, -i$$

Then $x_1 = \begin{bmatrix} 1 \\ i \end{bmatrix}$

$$x_2 = \begin{bmatrix} 1 \\ -i \end{bmatrix}$$

(I can't do this
stuff in my head!)

$$\frac{du}{dt} = Au$$

$$\frac{du_1}{dt} = 1 \cdot u_2$$

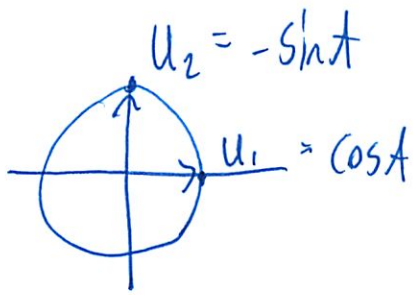
$$\frac{du_2}{dt} = -1 \cdot u_1$$

$$u_0 = \begin{bmatrix} 1 \\ 0 \end{bmatrix}$$

\uparrow start out

Physics' fav
example -
simple motion

5



$$u_1(t) = \cos t$$

$$u_2(t) = -\sin t$$

What is e^{At}

$$A = \begin{bmatrix} 0 & 1 \\ -1 & 0 \end{bmatrix}$$

$$A^2 = \begin{bmatrix} -1 & 0 \\ 0 & -1 \end{bmatrix}$$

$$A^3 = \begin{bmatrix} 0 & -1 \\ 1 & 0 \end{bmatrix}$$

$$A^4 = \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix} = A^2 \cdot A^2$$

every 4 terms back to the identity

$$e^{At} = \left[1 - \frac{1t^2}{2!} + \frac{1t^4}{4!} - \frac{1t^6}{6!} \dots \right]$$

↑ straight factorial
↑ see a pattern
↑ 2x2 matrix

$$\left[\begin{array}{c} \text{---} \\ t - \frac{t^3}{3!} + \text{---} \\ \text{---} \end{array} \right]$$

$$So = \begin{bmatrix} \cos t & \sin t \\ -\sin t & \cos t \end{bmatrix} \begin{bmatrix} 1 \\ 0 \end{bmatrix}$$

(looks familiar)

↳ Rare case when we can put in directly

OR Rotation matrix

Orthogonal

not sym or anti sym

If $A = \text{anti sym} \rightarrow e^{At} = \text{orthogonal}$

Since if $A^T = -A$ anti sym then e^{At} is orthogonal

$$e^{-At} = (e^{At})^T$$

important fact for a lot of purposes

②

$$\begin{aligned}\text{Look at } (e^{At})^T &= \left(I + At + \frac{A^2 t^2}{2} + \dots \right)^T \\ &= \left(I - At + \frac{A^2 t^2}{2} + \dots \right) \\ &= e^{-At}\end{aligned}$$

checked ' is orth

since transpose = inverse

For other diff eqn - prob don't want to use power series

$$e^{At} u(0) =$$

↓

combn of eigenvectors

$$c_1 x_1 + \dots + c_n x_n$$

$$u(t) = e^{At} u(0) = c_1 e^{\lambda_1 t} x_1 + \dots + c_n e^{\lambda_n t} x_n$$

↓

$$c_1 x_1 + \dots + c_n x_n$$

8

If 2nd order

$$u'' - 2u' + u = 0$$

Want $\frac{du}{dt} = Au$

$$u = \begin{bmatrix} y \\ y' \end{bmatrix}$$

$$\frac{du}{dt} = \begin{bmatrix} y' \\ y'' \end{bmatrix} = \begin{bmatrix} 0 & 1 \\ -1 & 2 \end{bmatrix} \begin{bmatrix} y \\ y' \end{bmatrix}$$

phase plane

↙ function and its velocity

What are the ds?

$$\begin{bmatrix} 0-d & 1 \\ -1 & 2-d \end{bmatrix}$$

$$\lambda^2 - 2\lambda + 1 = 0$$

↑ notice same as original problem!

double root / eigenvalue $\lambda = 1, 1$

9

Now eigenvector

$$\begin{bmatrix} -1 & 1 \\ -1 & 1 \end{bmatrix}$$

What's in Nullspace?

Only 1D

So only 1 eigenvector

$$C_1 e^{\lambda t} x + C_2 t e^{\lambda t}$$

(since repeat)

Diff eq by linear algebra

Prof - not a whole lot of brilliant ideas

Next weeks hw: Diff eq + Markov

Shorter than usual

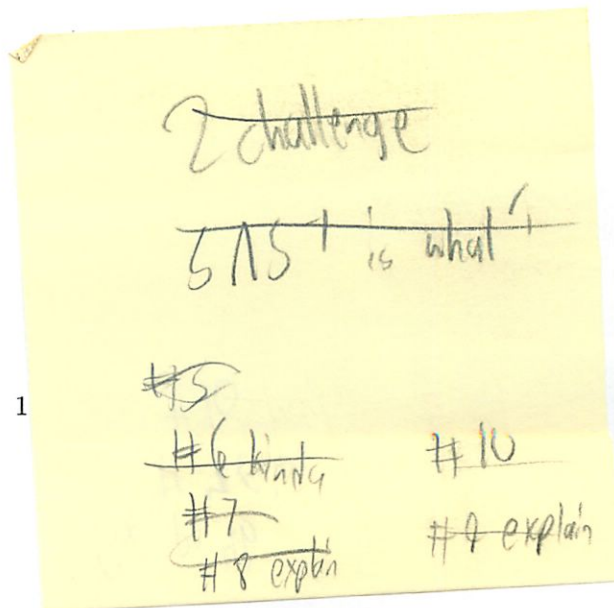
18.06 Spring 2012 – Problem Set 7

This problem set is due Thursday, April 19th, 2012 at 4pm (hand in to Room 2-106). The textbook problems are out of the 4th edition. For computational problems, please include a printout of the code with the problem set (for MATLAB in particular, `diary('filename')` will start a transcript session, `diary off` will end one.)

Every problem is worth 10 points.

1. Do Problem 2 from Section 8.3.
2. Do Problem 7 from Section 8.3 (do also the "challenge problem" part).
3. Do Problems 9 from Section 8.3.
4. Do Problems 12 from Section 8.3.
5. Do Problems 4 from Section 6.3.
6. Do Problems 5 from Section 6.3.
7. Do Problem 12 from Section 6.3.
8. Do Problem 24 from Section 6.3.
9. Do Problem 26 from Section 6.3.
10. Do Problem 30 from Section 6.3.

18.06 Wisdom. No 18.06 recitations on Tue April 17th - enjoy your Patriot's Day vacation!



8.3#2 Diagonalize Markov to $A = SAS^{-1}$
 by finding its other eigenvector
 $\frac{10}{10}$

$$A = \begin{bmatrix} .9 & .15 \\ .1 & .85 \end{bmatrix}$$

So first find λ

~~$$(.9 - \lambda)(.85 - \lambda) - .15 \cdot .1$$~~

~~$$.765 + \lambda^2 - .85\lambda - .9\lambda - .015$$~~

~~$$\lambda^2 - 1.75\lambda - .75 = 0$$~~

~~$$\lambda = 2.1, -.35$$~~

I think I did that wrong

Matlab \rightarrow $U = 1, .75$ and was given
 just use \rightarrow

Eigenvectors

$$\lambda \neq 1 \quad \begin{bmatrix} -.1 & .15 \\ .1 & -.15 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \end{bmatrix}$$

(2)

$$\begin{aligned} -1x_1 + .15x_2 &= 0 \\ 1x_1 + -.15x_2 &= 0 \end{aligned}$$

Again Matlab

$$S = \begin{pmatrix} .8321 & -.7071 \\ .5547 & .7071 \end{pmatrix}$$

$$S^{-1} = \begin{pmatrix} .7211 & .7211 \\ -.5657 & .8485 \end{pmatrix}$$

$$A = S \Lambda S^{-1}$$

$$= \begin{pmatrix} .1 & .15 \\ .1 & .85 \end{pmatrix} \quad \text{works}$$

$$\text{Limit } A^k = S \Lambda^k S^{-1} \quad \text{as } \begin{bmatrix} 1 & \\ & .75^k \end{bmatrix} \text{ approaches } \begin{bmatrix} 1 & 0 \\ 0 & 0 \end{bmatrix}$$

$$\text{So } S \begin{bmatrix} 1 & 0 \\ 0 & 0 \end{bmatrix} S^{-1} \rightarrow \begin{bmatrix} .6 & .6 \\ .4 & .4 \end{bmatrix} \quad \text{"steady state"}$$

(3)

R.3#7 Find eigenvalues, vectors of A

9/10

$$\lambda = 1, 5$$

$$S = \begin{bmatrix} .8231 & -.7071 \\ .5547 & .7071 \end{bmatrix}$$

(same as last problem)

$$A^k \text{ is } S \Lambda^k S^{-1}$$

$$S \begin{bmatrix} 1^k & 0 \\ 0 & 5^k \end{bmatrix} S^{-1}$$

$$S \begin{bmatrix} 1 & 0 \\ 0 & 0 \end{bmatrix} S^{-1}$$

(same exact as last problem)

$$\begin{bmatrix} .16 & .16 \\ .4 & .4 \end{bmatrix} = \lim_{k \rightarrow \infty} A^k = A^\infty$$

(4) Challenge

What Markov matrices produce this steady state
(Add ^{cols} to 1)

~~Any one with~~

So $\lambda = 1, .75$ were this
 $1, .5$

But I don't see the pattern

Back

$$\begin{bmatrix} .6 + .4a & .6 - .6a \\ .4 - .4a & .4 + .6a \end{bmatrix} \quad \begin{array}{l} a \leq 1 \\ .4 + .6a \geq 0 \end{array}$$

\therefore But what is this?

OHS Here we know $A^\infty = \begin{bmatrix} .6 & .6 \\ .4 & .4 \end{bmatrix}$

and $\lambda = 1, < 1$

(unknown also call a)

know eigenvector for 1
but not other eigenvector

Find vector in terms of a

also restrictions that $\sum c_i = 1$

Write 2nd vector in terms of B, C

Need to use the restrictions

4b

$$\text{So } \begin{bmatrix} 18-a & 13 \\ i2 & 17-a \end{bmatrix} \begin{bmatrix} b \\ c \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \end{bmatrix}$$

$$(18-a)b + 13c = 0$$

$$i2b + (17-a)c = 0$$

but $18-a + i2 = 1 \leftarrow \text{seems final}$

$$13 + 17-a = 1$$

want A in general

$$\begin{pmatrix} 16 & b \\ 14 & c \end{pmatrix} \begin{pmatrix} 1 \\ a \end{pmatrix} \begin{pmatrix} 16 & b \\ 14 & c \end{pmatrix}^{-1} = \begin{pmatrix} 16 & 14 \\ 16 & 14 \end{pmatrix}$$

but very complex to find

$$\begin{matrix} Ax = dx \\ \begin{pmatrix} a & b \\ c & d \end{pmatrix} \begin{pmatrix} 16 \\ 14 \end{pmatrix} = \begin{pmatrix} 16 \\ 14 \end{pmatrix} \end{matrix}$$

$$a+c = 1$$

$$b+d = 1$$

$$16a + 14b = 16$$

$$16c + 14d = 14$$

Solve from there. Also all ans $\oplus a, b, c, d \geq 0$

9c

Own

$$c = 1 - a$$

$$d = 1 - b$$

$$1.6(1 - a) + 1.4(1 - b) = 1.4$$

$$1.6 - 1.6a + 1.4 - 1.4b = 1.4$$

$$\frac{1.6a}{1.6} = \frac{1.6 - 1.4b}{1.6}$$

$$a = 1 - \frac{2}{3}b$$

$$1.6 - 1.6\left(1 - \frac{2}{3}b\right) + 1.4 - 1.4b = 1.4$$

$$1.6 - 1.6 - 1.4b + 1.4b - 1.4b = 1.4$$

$$-1.4b = 1.4$$

$$b = -1$$

$$a = 1 - \frac{2}{3}(-1)$$

$$= \frac{5}{3}$$

$$c = -\frac{2}{3}$$

$$d = 2$$

b, c, d should be in terms of a.

But that is not a function of something ...

(4d)

And violating $a, b, c, d > 0$

(but how would that work)

Doesn't seem to be latitude

And should be a function of ~~both~~ a, b

- not an exact ans

- I screwed this up

(5)

8.3 #9

10/10

Prove that the square of a Markov is still a Markov

Use $S\Lambda^2S^{-1}$

like last time $\begin{bmatrix} 1^2 & .75^2 \end{bmatrix} = \begin{bmatrix} 1 & .5625 \end{bmatrix}$

So $S\Lambda^2S^{-1} = \begin{bmatrix} .8250 & .5625 \\ .1750 & .7375 \end{bmatrix}$

S_2 confirmed, but not proved

We know $\alpha_1 = 1$
 $\alpha_2 < 1$

So $\begin{bmatrix} 1 & 0 \\ 0 & \alpha_1 \end{bmatrix}$

When multiplying this

$$\begin{bmatrix} S_1 & S_3 \\ S_2 & S_4 \end{bmatrix} \begin{bmatrix} 1 & 0 \\ 0 & \alpha_1 \end{bmatrix} = \begin{bmatrix} S_1 & S_3 \cdot \alpha_1 \\ S_2 & S_4 \cdot \alpha_1 \end{bmatrix}$$

S_1 and

S_2 are as they are

6

Then times S^{-1} more of the same

$$\begin{bmatrix} s_1 & s_3^{-1} \\ s_2 & s_4^{-1} \end{bmatrix} \begin{bmatrix} s_1^{-1} & s_3^{-1} \\ s_2^{-1} & s_4^{-1} \end{bmatrix}$$

Maintains the Markov property

How do you know that the columns add to 1?

Book M^2 is still non neg

$$[1 \dots 1] M = [1 \dots 1]$$

So multiply on the right by M to find

$$[1 \dots 1] M^2 = [1 \dots 1]$$

So basically multiplying by a Markov does nothing
A bit clearer that what I showed...

~~Question 4?~~

(7)

8.3 # 12 A Markov diff eq is not

8/10

$$\frac{du}{dt} = Au$$

but

is

$$\frac{du}{dt} = (A - I)u$$

diagonal \ominus

rest $A - I$ is \oplus

$$\sum \text{cols} = 0$$

Find the eigenvalues of $B = A - I = \begin{bmatrix} -2 & 3 \\ 2 & -3 \end{bmatrix}$

$$\text{hence } \lambda = 0, -1.5$$

Why does $A - I$ have $\lambda = 0$?

$$\text{So } (-2 - \lambda)(-3 - \lambda) - 2 \cdot 3$$

$$106 + 12\lambda + 3\lambda + \lambda^2 - 106$$

$$\lambda^2 + 1.5\lambda = 0$$

$$\lambda = 0, -1.5 \quad \checkmark$$

8

So it balances out?

Since this was $\begin{bmatrix} .8 & .3 \\ .7 & .7 \end{bmatrix}$

The -1 on the diagonal made it

$$\begin{bmatrix} -.2 & .3 \\ .2 & -.3 \end{bmatrix}$$

Basically $x, -x$ $y, -y$

Then multiplying together has $(-y)(-x) - (x)(y) = 0$

When $e^{\lambda_1 t}$ and $e^{\lambda_2 t}$ multiply x_1 and x_2
what is steady state vs $t \rightarrow \infty$?

So $e^{\lambda_1 t}$ are diff eq solns?

x_1 are eigen vectors

which are $\begin{bmatrix} .8321 & -.7071 \\ .5547 & .7071 \end{bmatrix}$ E/cob family!

$$e^{0t} \begin{bmatrix} .8321 \\ .5547 \end{bmatrix} = 1$$

$$e^{-.5t} \begin{bmatrix} -.7071 \\ .7071 \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \end{bmatrix}$$



So what is the steady state sol'n?

④

6.3#4 A door is opened

10/10

$$v(0) = 30 \text{ people}$$

$$w(0) = 10 \text{ people}$$

$$\frac{dv}{dt} = w - v$$

$$\frac{dw}{dt} = v - w$$

$$v + w = \text{fixed} = 40$$

$$\frac{du}{dt} = Au$$

So eigens help find steady state

$$u = e^{\lambda t} x$$

So A is $\begin{matrix} v \\ w \end{matrix} \begin{bmatrix} -1 & 1 \\ 1 & -1 \end{bmatrix}$ ① matches Book

But where was this described in the textbook?

$$\text{Ah } \frac{d}{dt} \begin{bmatrix} v \\ w \end{bmatrix} = \begin{bmatrix} -1 & 1 \\ 1 & -1 \end{bmatrix} \begin{bmatrix} v \\ w \end{bmatrix} = \begin{matrix} -v + w \\ v - w \end{matrix}$$

10

$$\lambda = -2, 0$$

$$S = \begin{pmatrix} .7071 & .7071 \\ -.7071 & .7071 \end{pmatrix}$$

What is the significance of this $\frac{1}{\sqrt{2}}$?

So $t=1$ is initial time? A^1 is just A

0_c is initial $t=0$ - usually this way

So book shows starting from 0

$$e^{-2(1)} = .135$$

$$e^{0(1)} = 1$$

How do we use that
plug into original

$$v) \quad 20 + .135(10) = 21.35$$

$$w) \quad 20 - .135(10) = 18.65$$

(1)

So how does SAS^{-1} stack up?

$$\begin{bmatrix} -1 & 1 \\ 1 & -1 \end{bmatrix}$$

Just gives back A'

but wasn't this supposed to be the state?

Or only the steady state?

Or only Markov? -no, don't think so

$$A = SAS^{-1}$$

$$A^2 = SA^2S^{-1}$$

$$A^k = SA^kS^{-1}$$

$$A^\infty = SA^\infty S^{-1} \text{ steady state}$$

← what does A^2 even mean?
Here, you're just solving Diff. eqs in matrix form. See sol'n.

So try again w/ $A = \infty$

$$e^{0(\infty)} = 1$$

$$e^{-2(\infty)} = \infty \quad \text{'wait what'}$$

Try other way $A^\infty = \begin{bmatrix} \infty & 0 \\ 0 & 0 \end{bmatrix}$

$$SA^\infty S = \begin{bmatrix} \infty & -\infty \\ -\infty & \infty \end{bmatrix}$$

⊗ Not right either

(2)

Its A^* is not steady state'

Only w/ $\lambda=1$ I believe

(chap 8.3)

So that was my issue

So stick to $v = e^{\lambda t} x$

$$\text{But } e^{-2(\infty)} = 0$$

$$e^{0(\text{inf})} = 1$$

$$u(t) = e^{\lambda_1 t} x_1 + e^{\lambda_2 t} x_2$$

no need to orthonormalize, although this is fine.

$$= (e^{-2t} \begin{bmatrix} .7071 \\ -.7071 \end{bmatrix}) + D e^{0 \cdot t} \begin{bmatrix} .7071 \\ .7071 \end{bmatrix}$$

Ah, right need to calibrate on start

$$\begin{bmatrix} 30 \\ 10 \end{bmatrix} = C \begin{bmatrix} .7071 \\ -.7071 \end{bmatrix} + D \begin{bmatrix} .7071 \\ .7071 \end{bmatrix}$$

$$C = 14.14 = 10\sqrt{2}$$

$$D = 28.28 = 20\sqrt{2}$$

exact ans better

So then time ∞

$$\begin{bmatrix} \hat{x} \\ \hat{y} \end{bmatrix} = 10\sqrt{2} \cdot 0 \cdot \begin{bmatrix} 1/\sqrt{2} \\ -1/\sqrt{2} \end{bmatrix} + 20\sqrt{2} \cdot 1 \cdot \begin{bmatrix} 1/\sqrt{2} \\ 1/\sqrt{2} \end{bmatrix}$$

$$= \begin{bmatrix} 20 \\ 20 \end{bmatrix} \quad \checkmark$$

(13)

Q3#5 - Reverse diffusion

10/10

$$\frac{dv}{dt} = v - w$$

$$\frac{dw}{dt} = w - v$$

$$\text{So } A = \begin{bmatrix} 1 & -1 \\ -1 & 1 \end{bmatrix}$$

$$\lambda = 0, 2$$

$$S = \begin{bmatrix} -1/\sqrt{2} & -1/\sqrt{2} \\ -1/\sqrt{2} & 1/\sqrt{2} \end{bmatrix}$$

$$\begin{bmatrix} 30 \\ 10 \end{bmatrix} = C \begin{bmatrix} -1/\sqrt{2} \\ -1/\sqrt{2} \end{bmatrix} + D \begin{bmatrix} -1/\sqrt{2} \\ 1/\sqrt{2} \end{bmatrix}$$

$$C = -20\sqrt{2}$$

$$D = -10\sqrt{2}$$

$$u = -20\sqrt{2} \cdot 1 \cdot \begin{bmatrix} -1/\sqrt{2} \\ +1/\sqrt{2} \end{bmatrix} + -10\sqrt{2} \begin{bmatrix} -1/\sqrt{2} \\ -1/\sqrt{2} \end{bmatrix} e^{2t}$$

$$= \begin{bmatrix} \text{infinity} \\ -\text{infinity} \end{bmatrix} + \begin{bmatrix} -10 \\ -10 \end{bmatrix} = \begin{bmatrix} \text{inf} \\ -\text{inf} \end{bmatrix} \rightarrow \begin{bmatrix} 40 \\ 0 \end{bmatrix}$$

but how does that work w/ limits

(136)

(11) Write general sol as fn of t

$$\begin{bmatrix} v(t) \\ w(t) \end{bmatrix} = -20\sqrt{2} e^{0t} \begin{bmatrix} -1/\sqrt{2} \\ +1/\sqrt{2} \end{bmatrix} - 10\sqrt{2} e^{2t} \begin{bmatrix} -1/\sqrt{2} \\ -1/\sqrt{2} \end{bmatrix}$$

$v+w$ constant

$$\lim_{t \rightarrow \infty} = \begin{bmatrix} \infty \\ -\infty \end{bmatrix}$$

Then apply limit

$$\begin{bmatrix} 40 \\ 0 \end{bmatrix}$$

✓ So I had understood it 😊

At some t reach 0, 40

Past that point, the interpretation of v, w as people in room breaks down
(like w/ partial fraction of people)

(14)

Q.3 #12 Substitute $y = e^{\lambda t}$ into

10/10

$$y'' = 6y' - 9y$$

to show $\lambda = 3$ is a repeated root.

$$(e^{\lambda t})'' = 6(e^{\lambda t})' - 9(e^{\lambda t})$$

It is that what they meant!

OHJ

Do it

$$\lambda^2 e^{\lambda t} = 6\lambda e^{\lambda t} - 9e^{\lambda t}$$

non-zero #

can divide all by $e^{\lambda t}$

$$\lambda^2 = 6\lambda - 9$$

quadratic eq as λ

$$\lambda^2 - 6\lambda + 9 = 0$$

$$\lambda = +3, +3$$

Answered let q_v

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For new matrix

$$\frac{d}{dt} \begin{bmatrix} y \\ y' \end{bmatrix} = \begin{bmatrix} 0 & 1 \\ -9 & 6 \end{bmatrix} \begin{bmatrix} y \\ y' \end{bmatrix}$$

Show $\lambda = 3, 3$

So $\det(A - \lambda I) =$ what we found earlier

So it's the same

Only 1 line of eigenvectors

Since put in same λ , will get the same result

or say nullspace of $A - 3I$ is 1-dimensional

Show 2nd sol is $y = t e^{3t}$

- saw in 18.03 we did this

- to avoid repeated root

Plug in again + show it works

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$$(te^{3t})'' = 6(te^{3t})' - 9(te^{3t})$$

~~$$9te^{3t} = 18te^{3t} - 9te^{3t}$$~~

$\frac{dte^{3t}}{dt}$ is product rule

$$3(3t+2)e^{3t} = 6(3t+1)e^{3t} - 9te^{3t}$$

$$(9t+6)e^{3t} = (18t+6)e^{3t} - 9te^{3t}$$

divide out e^{3t}

$$9t+6 = 18t+6 - 9t$$

$$9t+6 = 9t+6$$

✓ matches

(17)

6.3 # 24 Write $A = \begin{bmatrix} 1 & 1 \\ 0 & 3 \end{bmatrix}$ as SAS^{-1}

10/10

$$\lambda = 1, 3$$

$$S = \begin{bmatrix} 1 & .4472 \\ 0 & .8944 \end{bmatrix} \quad S^{-1} = \begin{bmatrix} 1 & -.5 \\ 0 & 1.11 \end{bmatrix}$$

$$\begin{bmatrix} 1 & .4472 \\ 0 & .8944 \end{bmatrix} \begin{bmatrix} 1 & \\ & 3 \end{bmatrix} \begin{bmatrix} 1 & -.5 \\ 0 & 1.11 \end{bmatrix}$$

$$e^{At} = \begin{bmatrix} e^t & \\ & e^{3t} \end{bmatrix}$$

$$e^{At} = \begin{bmatrix} e^t & .5e^{3t} - .5e^t \\ 0 & e^{3t} \end{bmatrix}$$

So at $t=0$

$$\begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix}$$

$$\frac{d e^{At}}{dt} = \begin{bmatrix} e^t & 1.5e^{3t} - .5e^t \\ 0 & 3e^{3t} \end{bmatrix}$$

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At $t=0$


$$= \begin{bmatrix} 1 & 1 \\ 0 & 3 \end{bmatrix}$$

What is that supposed to match w/ again?
LOHS Change is A itself

Book/ Much simpler $S = \begin{bmatrix} 1 & 1 \\ 0 & 2 \end{bmatrix}$ $S^{-1} = \begin{bmatrix} 1 & -1/2 \\ 0 & 1/2 \end{bmatrix}$

Why was mine so complex?

And no explanation

OHs Remember eigenvector is everything along the line
Matlab wants to have it normalized 
but $\begin{bmatrix} 1 \\ 2 \end{bmatrix}$ is just as good

OHs $\frac{de^{At}}{dt} \rightarrow \frac{de^{A0}}{dt} = A$
we showed this

P319 $\frac{de^{At}}{dt} = A e^{At}$
so w/ $t=0$, e^{At} is I, so get A back
Problem is weirding this for specific instance
© Understand

(19)

Q3 #26 Give 2 reasons why e^{At} is never singular

10/10 a) Write its inverse

How supposed to find?

Gauss-Jordan

book $(e^{At})^{-1} = e^{-At}$

How get that

OH Proof of this is outside scope of course

$$e^{x+y} = e^x e^y \text{ for } \#s$$

$$(e^x)^{-1} = e^{-x}$$

But does this generalize to matrices?

Not completely, but

$$e^{A+B} = e^A e^B \text{, No, but if } AB=BA, \text{ then yes}$$

commute

So $tA, -tA$ commute

$$\text{So } e^{tA} + e^{-tA} = e^{tA - tA} = e^0 = \mathbf{1}$$

Use power series to prove relation

(20)

b) write down its eigenvalues

$$\text{If } Ax = \lambda x \text{ Then } e^{At} x = \underline{\quad} x$$

? $e^{\lambda t}$

o Badh

Now the main q: Why is e^{At} never singular

Singular = non invertible

It is invertible \rightarrow so not singular

Since $e^{At} = e^{-At}$

But still need to show that

\hookrightarrow did in OH, see prev page

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Quiz #30 Another good idea for $y'' = -y$ is
7/10 the trapezoidal method (half forward/half back)
(i.w.t.f.)

Best way to keep (y_n, z_n) on a circle

$$\begin{bmatrix} 1 & -\Delta t/2 \\ \Delta t/2 & 1 \end{bmatrix} \begin{bmatrix} y_{n+1} \\ z_{n+1} \end{bmatrix} =$$

$$\begin{bmatrix} 1 & \Delta t/2 \\ -\Delta t/2 & 1 \end{bmatrix} \begin{bmatrix} y_n \\ z_n \end{bmatrix}$$

a) Invert to write as $V_{n+1} = AV_n$

OH/ Gives you row info

Rules for trapezoidal method → outside scope of course

Go sentence by sentence

Find inverse Gauss Jordan or det

Symbol is parameter
Should be simple

$$\frac{1}{\det} \begin{bmatrix} d & -b \\ -c & a \end{bmatrix}$$

22

own

$$\det = (1) - \left(\frac{\Delta t}{2}\right)\left(-\frac{\Delta t}{2}\right)$$

$$= 1 + \frac{\Delta t^2}{4}$$

$$\text{left}^{-1} = \frac{1}{1 + \frac{\Delta t^2}{4}} \begin{bmatrix} 1 & \Delta t/2 \\ -\Delta t/2 & 1 \end{bmatrix}$$

Write as $U_{n+1} = A U_n$ Some w/ $\frac{1}{\det}$

$$U_{n+1} = \underbrace{\begin{bmatrix} 1 + \Delta t/2 & \\ -\Delta t/2 & 1 \end{bmatrix}}_{PA} \begin{bmatrix} Y_n \\ Z_n \end{bmatrix}$$

'wait wasn't that done for us?
How does inverse fit in here?



(23)

OH) Then show A is orthogonal

$A^T A = I$ ← means orthogonal

(Just ignore the extra info + do the steps)

if input = norm 1 → then output is norm 1
norm $A \cdot V$ = norm ~~V~~ V

Down) So $A = \begin{bmatrix} 1 & \Delta t/2 \\ -\Delta t/2 & 1 \end{bmatrix}$

transpose of this is

$A^T = \begin{bmatrix} 1 & -\Delta t/2 \\ \Delta t/2 & 1 \end{bmatrix}$

Multiply $A^T A = I$

$\begin{bmatrix} 1 & -\Delta t/2 \\ \Delta t/2 & 1 \end{bmatrix} \begin{bmatrix} 1 & \Delta t/2 \\ -\Delta t/2 & 1 \end{bmatrix} = \begin{bmatrix} 1 - \frac{\Delta t^2}{4} & \frac{\Delta t}{2} - \frac{\Delta t}{2} \\ \frac{\Delta t}{2} - \frac{\Delta t}{2} & \frac{\Delta t^2}{4} + 1 \end{bmatrix}$

$\frac{d}{dt} e^{At}$

= not really I (I at t=0).
- did I multiply wrong?

(24)

These points U_n never leave the circle

$$A = (I - B)^{-1} (I + B) \text{ is always orthogonal} \\ \text{if } B^T = -B$$

b) MATLAB

$$U_0 = (1, 0) \text{ to } U_{32} \text{ w/ } \Delta t = \frac{2\pi}{32}$$

So how do we do
enter values

$$U_1 = \begin{bmatrix} 1 \\ .0982 \end{bmatrix}$$

$$U_2 = \begin{bmatrix} .9904 \\ -.1963 \end{bmatrix}$$

$$U_3 = \begin{bmatrix} .9711 \\ -.2936 \end{bmatrix}$$

etc on point at

does not seem to be a circle ...

$$\text{back to } \begin{bmatrix} -.116 \\ -.10117 \end{bmatrix} = U_{32}$$


```
>> A = [1 (2*pi)/64; -(2*pi)/64 1]
```

```
A =
```

```
1.0000    0.0982
-0.0982    1.0000
```

```
>> U = [1; 0]
```

```
U =
```

```
1
0
```

u_0

```
>> U = A*U
```

```
U =
```

```
1.0000
-0.0982
```

u_1

```
>> U = A*U
```

```
U =
```

```
0.9904
-0.1963
```

2

```
>> U = A*U
```

```
U =
```

```
0.9711
-0.2936
```

3

```
>> U = A*U
```

```
U =
```

```
0.9423
-0.3889
```

4

```
>> U = A*U
```

```
U =
```

```
0.9041
-0.4814
```

5

```
>> U = A*U
```

```
U =
```

0.8568
-0.5702

6

>> U = A*U

U =

7

0.8008
-0.6543

>> U = A*U

U =

8

0.7366
-0.7329

>> U = A*U

U =

9

0.6647
-0.8052

>> U = A*U

U =

10

0.5856
-0.8705

>> U = A*U

U =

11

0.5001
-0.9280

>> U = A*U

U =

12

0.4090
-0.9771

>> U = A*U

U =

13

0.3131
-1.0172

>> U = A*U

U =

0.2132
-1.0480

14

>> U = A*U

U =

0.1104
-1.0689

15

>> U = A*U

U =

0.0054
-1.0797

16

>> U = A*U

U =

-0.1006
-1.0803

17

>> U = A*U

U =

-0.2066
-1.0704

18

>> U = A*U

U =

-0.3117
-1.0501

19

>> U = A*U

U =

-0.4148
-1.0195

20

>> U = A*U

U =
-0.5149
-0.9788

21

>> U = A*U

U =
-0.6110
-0.9282

22

>> U = A*U

U =
-0.7021
-0.8682

23

>> U = A*U

U =
-0.7874
-0.7993

24

>> U = A*U

U =
-0.8658
-0.7220

25

>> U = A*U

U =
-0.9367
-0.6370

26

>> U = A*U

U =
-0.9993
-0.5450

27

>> U = A*U

U =
-1.0528

28

-0.4469

>> U = A*U

U =

-1.0967
-0.3436

29

>> U = A*U

U =

-1.1304
-0.2359

30

>> U = A*U

U =

-1.1536
-0.1249

31

>> U = A*U

U =

-1.1658
-0.0117

32

>>

18.06 Spring 2012 – Problem Set 7

This problem set is due Thursday, April 19th, 2012 at 4pm (hand in to Room 2-106). The textbook problems are out of the 4th edition. For computational problems, please include a printout of the code with the problem set (for MATLAB in particular, `diary('filename')` will start a transcript session, `diary off` will end one.)

Every problem is worth 10 points.

1. Do Problem 2 from Section 8.3.

Solution. Since $\begin{bmatrix} 0.6 \\ 0.4 \end{bmatrix}$ and $\begin{bmatrix} -1 \\ 1 \end{bmatrix}$ are the eigenvector vectors for the eigenvalues 1 and 0.75, respectively,

$$S = \begin{bmatrix} 0.6 & -1 \\ 0.4 & 1 \end{bmatrix}.$$

A^k approaches to

$$S \begin{bmatrix} 1 & 0 \\ 0 & 0 \end{bmatrix} S^{-1} = \begin{bmatrix} 0.6 & 0.6 \\ 0.4 & 0.4 \end{bmatrix}.$$

□

2. Do Problem 7 from Section 8.3 (do also the "challenge problem" part).

Solution. The eigenvalues are 1 and 0.5, and the eigenvectors are

$$\begin{bmatrix} 0.6 \\ 0.4 \end{bmatrix} \text{ and } \begin{bmatrix} 1 \\ -1 \end{bmatrix}.$$

Since

$$A^k = S \begin{bmatrix} 1 & 0 \\ 0 & 0.5 \end{bmatrix}^k S^{-1},$$

for

$$S = \begin{bmatrix} 0.6 & 1 \\ 0.4 & -1 \end{bmatrix},$$

$$A^\infty = S \begin{bmatrix} 1 & 0 \\ 0 & 0 \end{bmatrix} S^{-1} = \begin{bmatrix} 0.6 & 0.6 \\ 0.4 & 0.4 \end{bmatrix}.$$

Challenge problem Let $A = \begin{bmatrix} a & b \\ 1-a & 1-b \end{bmatrix}$, $0 \leq a, b \leq 1$, be a Markov Matrix with steady state $\begin{bmatrix} 0.6 \\ 0.4 \end{bmatrix}$. Then

$$A \begin{bmatrix} 0.6 \\ 0.4 \end{bmatrix} = \begin{bmatrix} 0.6 \\ 0.4 \end{bmatrix}.$$

Hence $0.6a + 0.4b = 0.6$. In other words,

$$A = \begin{bmatrix} 0.6 + 0.4x & 0.4 - 0.4x \\ 0.6 - 0.6x & 0.4 + 0.6x \end{bmatrix}$$

for some $-\frac{2}{3} \leq x \leq 1$.

□

3. Do Problem 9 from Section 8.3.

Solution. If every entry of A is nonnegative, every entry of A^2 is also nonnegative. Since, for any $j = 1, \dots, n$, $\sum_i (A)_{ij} = 1$,

$$\sum_i (A^2)_{ij} = \sum_{i,k} a_{ik}a_{kj} = \sum_k \sum_i a_{ik}a_{kj} = \left(\sum_k \left(\sum_i a_{ik} \right) a_{kj} \right) = \sum_k a_{kj} = 1.$$

□

4. Do Problem 12 from Section 8.3.

Solution. The eigenvalues are $\lambda_1 = 0$ and $\lambda_2 = -0.5$. We have a steady state for the Markov matrix A . For the steady state v , $(A - I)v = 0 = 0v$. So $A - I$ have $\lambda = 0$. If $u_t = e^{\lambda_1 t} c_1 x_1 + e^{\lambda_2 t} c_2 x_2$ for the initial value $u_0 = c_1 x_1 + c_2 x_2$, u_t converges to $c_1 x_1$ as $t \rightarrow \infty$.

□

5. Do Problem 4 from Section 6.3.

Solution. $v + w$ is constant if and only if $\frac{d(v+w)}{dt} = 0$.

$$\frac{d(v+w)}{dt} = \frac{dv}{dt} + \frac{dw}{dt} = (w - v) + (v - w) = 0,$$

so $v + w$ is constant.

Let $u = \begin{bmatrix} v \\ w \end{bmatrix}$. Then

$$\frac{du}{dt} = \begin{bmatrix} \frac{dv}{dt} \\ \frac{dw}{dt} \end{bmatrix} = \begin{bmatrix} w - v \\ v - w \end{bmatrix} = \begin{bmatrix} -1 & 1 \\ 1 & -1 \end{bmatrix} \begin{bmatrix} v \\ w \end{bmatrix}.$$

The eigenvalue of $A = \begin{bmatrix} -1 & 1 \\ 1 & -1 \end{bmatrix}$ is given by solving $\det(A - \lambda I) = 0$. $\det(A - \lambda I) = (-1 - \lambda)^2 - 1 = 1 + 2\lambda + \lambda^2 - 1 = \lambda(\lambda + 2)$ so the eigenvalues are $\lambda_1 = 0$, $\lambda_2 = -2$. We then observe that the corresponding eigenvectors are $x_1 = \begin{bmatrix} 1 \\ 1 \end{bmatrix}$ and $x_2 = \begin{bmatrix} 1 \\ -1 \end{bmatrix}$, respectively.

Then the pure exponential solutions are given by

$$u_1(t) = e^{\lambda_1 t} x_1 = \begin{bmatrix} 1 \\ 1 \end{bmatrix}$$
$$u_2(t) = e^{\lambda_2 t} x_2 = e^{-2t} \begin{bmatrix} 1 \\ -1 \end{bmatrix}$$

So the complete solutions are given by

$$u(t) = C \begin{bmatrix} 1 \\ 1 \end{bmatrix} + De^{-2t} \begin{bmatrix} 1 \\ -1 \end{bmatrix} = \begin{bmatrix} C + De^{-2t} \\ C - De^{-2t} \end{bmatrix}.$$

From the initial condition that $u(0) = \begin{bmatrix} v(0) \\ w(0) \end{bmatrix} = \begin{bmatrix} 30 \\ 10 \end{bmatrix}$, we get $C = 20$, $D = 10$.

That is,

$$v(t) = 20 + 10e^{-2t}$$
$$w(t) = 20 - 10e^{-2t}.$$

So $v(1) = 20 + 10e^{-2}$, $w(1) = 20 - 10e^{-2}$, $v(\infty) = w(\infty) = 20$.

□

6. Do Problem 5 from Section 6.3.

Solution. Now we have

$$\frac{du}{dt} = -Au = \begin{bmatrix} 1 & -1 \\ -1 & 1 \end{bmatrix} u.$$

The eigenvalues of $-A$ are given by -1 times the eigenvalues of A , so now we have $\lambda_1 = 0$, $\lambda_2 = 2$. The corresponding eigenvectors are the same as those of A , namely $x_1 = \begin{bmatrix} 1 \\ 1 \end{bmatrix}$ and $x_2 = \begin{bmatrix} 1 \\ -1 \end{bmatrix}$.

Then the pure exponential solutions are given by

$$u_1(t) = e^{\lambda_1 t} x_1 = \begin{bmatrix} 1 \\ 1 \end{bmatrix}$$
$$u_2(t) = e^{\lambda_2 t} x_2 = e^{2t} \begin{bmatrix} 1 \\ -1 \end{bmatrix}$$

So the complete solutions are given by

$$u(t) = C \begin{bmatrix} 1 \\ 1 \end{bmatrix} + De^{2t} \begin{bmatrix} 1 \\ -1 \end{bmatrix} = \begin{bmatrix} C + De^{2t} \\ C - De^{2t} \end{bmatrix}.$$

From the initial conditions $u(0) = \begin{bmatrix} v(0) \\ w(0) \end{bmatrix} = \begin{bmatrix} 30 \\ 10 \end{bmatrix}$, we get $C = 20$, $D = 10$, and $v(t) = 20 + 10e^{2t}$. So as $t \rightarrow \infty$, $v \rightarrow \infty$.

□

7. Do Problem 12 from Section 6.3.

Solution. Substituting $y = e^{\lambda t}$ into $y'' = 6y' - 9y$ gives

$$\lambda^2 e^{\lambda t} = 6\lambda e^{\lambda t} - 9e^{\lambda t},$$

so $e^{\lambda t}(\lambda - 3)^2 = 0$, which means $\lambda = 3$ is a repeated root.

In terms of the matrix equation, since the matrix has trace 6 and determinant 9, its only eigenvalue is 3, with one independent eigenvector $\begin{bmatrix} 1 \\ 1 \end{bmatrix}$.

To show that $y = te^{3t}$ is the second solution, just substitute this into the original differential equation. Since we have:

$$\begin{aligned} y' &= e^{3t} + 3te^{3t} \\ y'' &= 3e^{3t} + (3e^{3t} + 9te^{3t}) = 6e^{3t} + 9te^{3t}. \end{aligned}$$

Also,

$$6y' - 9y = 6e^{3t} + 18te^{3t} - 9te^{3t} = 6e^{3t} + 9te^{3t},$$

so we see that $y'' = 6y' - 9y$ when $y = te^{3t}$.

□

8. Do Problem 24 from Section 6.3.

Solution. A is an upper-triangular matrix, so we can read off its eigenvalues as the diagonal entries: 1, 3. By inspection we see that $(1, 0)$ is an eigenvector with eigenvalue 1. To find an eigenvector with eigenvalue 3 we observe

$$A - 3I = \begin{pmatrix} -2 & 1 \\ 0 & 0 \end{pmatrix},$$

and so $(1, 2)$ is in its nullspace. Thus

$$S = \begin{pmatrix} 1 & 1 \\ 0 & 2 \end{pmatrix}$$

and

$$A = S\Lambda S^{-1} = \frac{1}{2} \begin{pmatrix} 1 & 1 \\ 0 & 2 \end{pmatrix} \begin{pmatrix} 1 & 0 \\ 0 & 3 \end{pmatrix} \begin{pmatrix} 2 & -1 \\ 0 & 1 \end{pmatrix}.$$

Thus

$$e^{At} = S e^{\Lambda t} S^{-1} = \frac{1}{2} \begin{pmatrix} 1 & 1 \\ 0 & 2 \end{pmatrix} \begin{pmatrix} e^t & 0 \\ 0 & e^{3t} \end{pmatrix} \begin{pmatrix} 2 & -1 \\ 0 & 1 \end{pmatrix} = \begin{pmatrix} e^t & \frac{1}{2}(e^{3t} - e^t) \\ 0 & e^{3t} \end{pmatrix}.$$

When $t = 0$ this is $\begin{pmatrix} 1 & 0 \\ 0 & 1 \end{pmatrix}$, as expected. Differentiating with respect to t , we get

$$\begin{pmatrix} e^t & \frac{1}{2}(3e^{3t} - e^t) \\ 0 & 3e^{3t} \end{pmatrix};$$

at $t = 0$ this is $\begin{pmatrix} 1 & 1 \\ 0 & 3 \end{pmatrix} = A$.

□

9. Do Problem 26 from Section 6.3.

Solution. e^{At} is nonsingular because

- (a) its inverse is given by e^{-At} ,
- (b) its eigenvalues are $e^{\lambda t}$ where λ is an eigenvalue of A — thus 0 is never an eigenvalue of e^{At} .

□

10. Do Problem 30 from Section 6.3.

Solution. (a) $\begin{pmatrix} 1 & -\Delta t/2 \\ \Delta t/2 & 1 \end{pmatrix}^{-1} = \frac{1}{1+(\Delta t)^2/4} \begin{pmatrix} 1 & \Delta t/2 \\ -\Delta t/2 & 1 \end{pmatrix}$, so if $\mathbf{U}_n = (Y_n, Z_n)$ we have

$$\mathbf{U}_{n+1} = \frac{1}{1+(\Delta t)^2/4} \begin{pmatrix} 1 & \Delta t/2 \\ -\Delta t/2 & 1 \end{pmatrix} \begin{pmatrix} 1 & \Delta t/2 \\ -\Delta t/2 & 1 \end{pmatrix} \mathbf{U}_n = A\mathbf{U}_n$$

where

$$A = \frac{1}{1+(\Delta t)^2/4} \begin{pmatrix} 1 - (\Delta t)^2/4 & \Delta t \\ -\Delta t & 1 - (\Delta t)^2/4 \end{pmatrix}.$$

Then

$$\begin{aligned} A^T A &= \frac{1}{(1+(\Delta t)^2/4)^2} \begin{pmatrix} 1 - (\Delta t)^2/4 & -\Delta t \\ \Delta t & 1 - (\Delta t)^2/4 \end{pmatrix} \begin{pmatrix} 1 - (\Delta t)^2/4 & \Delta t \\ -\Delta t & 1 - (\Delta t)^2/4 \end{pmatrix} \\ &= \frac{1}{(1+(\Delta t)^2/4)^2} \begin{pmatrix} (1 - (\Delta t)^2/4)^2 + (\Delta t)^2 & 0 \\ 0 & (1 - (\Delta t)^2/4)^2 + (\Delta t)^2 \end{pmatrix} \\ &= \begin{pmatrix} 1 & 0 \\ 0 & 1 \end{pmatrix}. \end{aligned}$$

If $B^T = -B$ and $A = (I - B)^{-1}(I + B)$ then $A^T A = (I + B^T)(I - B^T)^{-1}(I - B)^{-1}(I + B) = (I - B)(I + B)^{-1}(I - B)^{-1}(I + B)$. But notice that $(I + B)(I - B) = I - B^2 = (I - B)(I + B)$, hence this equals $(I - B)(I - B)^{-1}(I + B)^{-1}(I + B) = I$. Similarly $AA^T = (I - B)^{-1}(I + B)(I - B)(I + B)^{-1} = (I - B)^{-1}(I - B)(I + B)(I + B)^{-1} = I$, so A is indeed orthogonal.

- (b) If $\Delta t = 2\pi/32$ then using Matlab to compute A^{32} gives

$$\begin{pmatrix} 0.9998 & -0.0201 \\ 0.0201 & 0.9998 \end{pmatrix},$$

which is close to the identity, but there is clearly a potentially significant error.

□

OH

4/13

Only when $\alpha = 1$ is A^k steady state

Symmetric matrices

$$A = A^T = A^* = A'$$

↙ Symbols for transpose
in other subjects

Key facts

1. All eigenvalues are REAL

2. Eigenvectors are orthogonal

Can choose $S=Q$ eigenmatrix ~~vector~~
vector matrix

$$A = S \Lambda S^{-1}$$

$$\downarrow$$

$$A = Q \Lambda Q^T$$

Examples

$$A = \begin{bmatrix} 0 & 1 \\ 1 & 0 \end{bmatrix} = A^T \quad \lambda = 1, -1$$

$$A = \begin{bmatrix} 0 & 1 \\ -1 & 0 \end{bmatrix} = -A^T \quad \lambda = i, -i \quad S = \begin{bmatrix} 1 & 1 \\ i & -i \end{bmatrix} \quad \text{not sym}$$

$$A = \begin{bmatrix} 2 & 1 \\ -1 & 2 \end{bmatrix} \quad \lambda = 2+i, 2-i \quad \text{not sym}$$

(2)

Dealing w/ complex #s

L will do later

← in Chap 10 but discussing later today

(How does this match ???)

Fact

$$Q^T Q = I$$

$$Q^T = Q^{-1}$$

↓ so

$$A = Q \Lambda Q^T$$

← one of the great theorems of lin algebra

$Q = \text{orthonormal}$

If $A^T \Rightarrow Q \Lambda Q^T$ (I can hear correctly)

Since $\Lambda^T = \Lambda$

3

Examples (explained)

Simplest possible \rightarrow $A = \begin{bmatrix} 0 & 1 \\ 1 & 0 \end{bmatrix} \rightarrow$ symmetric so $\lambda = 1, -1$
 \uparrow real

Then $Q = \begin{bmatrix} 1/\sqrt{2} & 1/\sqrt{2} \\ 1/\sqrt{2} & -1/\sqrt{2} \end{bmatrix} \in \text{orthogonal}$

\uparrow
 Since orthonormal

$\begin{bmatrix} 0 & 1 \\ -1 & 0 \end{bmatrix}$ not sym
 $+2I \hookrightarrow \begin{bmatrix} 2 & 1 \\ -1 & 2 \end{bmatrix}$ so add 2 as well

$\lambda = i, -i$
 \downarrow
 $2+i, 2-i$ same S

$S = \begin{bmatrix} 1 & 1 \\ i & -i \end{bmatrix}$
 ~~$S = \begin{bmatrix} 1 & 1 \\ i & -i \end{bmatrix}$~~
 \downarrow orthogonal

\uparrow note take dot product differently when complex

False reasoning for real A

$Ax = \lambda x$ EOM

take dot product (regular) w/ A^T

$x^T Ax = \lambda x^T x$ EOM

(4)

False concludes eigenvalues are always real

Divide by one side / rearrange

$$\lambda = \frac{x^T A x}{x^T x}$$

$\hookrightarrow \|x\|^2$

Everything looks real

x looks real

↗ but this is false - λ does not have to be real

$$S = \begin{bmatrix} 1 & 1 \\ i & -i \end{bmatrix}$$

appears dot product

$$\|x\|^2 = x^T x = \begin{bmatrix} 1 & i \end{bmatrix} \begin{bmatrix} 1 \\ i \end{bmatrix} = 0 \quad \leftarrow \text{bit wrong way}$$

Must do something more to set length

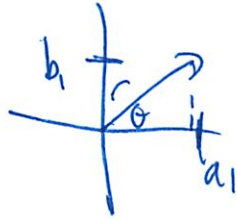
$$\|x\| \text{ is } \sqrt{2}$$

5

Complex x in $\mathbb{C}^n = \begin{pmatrix} a_1 + ib_1 \\ a_2 + ib_2 \\ \vdots \\ a_n + ib_n \end{pmatrix}$

So if $n=1$

Size of $a_1 + ib_1$ is
(length)



length of this vector

$$C = |a_1 + ib_1|$$

$$= \sqrt{a_1^2 + b_1^2}$$

So $\|x\|$ is not $x^T x$

$$\text{but } \sqrt{a_1^2 + b_1^2 + a_2^2 + b_2^2 + \dots + a_n^2 + b_n^2}$$

Can only be 0 if everything is 0

So want a complex dot product

$$r^2 = a^2 + b^2$$
$$(a+ib)(a-ib) \leftarrow \text{complex conjugate}$$

So $\overline{X}^T X$

\uparrow
X conjugate \rightarrow in book uses X^H $\|X\|^2 = X^H X$

$$\begin{bmatrix} a_1 - ib_1 & \dots \end{bmatrix} \begin{bmatrix} a_1 + ib_1 \\ \vdots \end{bmatrix} = \begin{bmatrix} a_1^2 + b_1^2 + \\ a_2^2 + b_2^2 + \\ \vdots \end{bmatrix}$$

* Don't forget to take the conjugate

~~If following A complex~~

$$A = A^H = A^* = A' = \overline{A^T}$$

\uparrow \uparrow
Symbols don't change

MATLAB auto uses A^H instead of A^T

①

↓ key facts
to remember

So can't just have $S=Q$

So instead for symmetric complex

$$A = A^H = \bar{A}^T$$

called "Hermitian"

if orthogonal for symmetric complex

~~$$Q^T Q = I$$~~

$$Q^{-1} \rightarrow Q^H$$

$$Q^H Q = I$$

Complex conj
and transpos

↑ Hermit right

called "unitary"

So all λ are real and eigenvectors orthogonal
Since still symmetric

(w/ Q^H not Q^T)

$$A = Q \Lambda Q^H$$

Then since λ is real, can go backwards

$$A = A^H \Leftrightarrow A = Q \Lambda Q^H$$

8

Take conjugate transpose

⋮ (missed)

Spectral theorem

aka the principal axis theorem

Connects w/ an ellipsoid



3×3 symmetric matrix A

$$\begin{bmatrix} x & y & z \end{bmatrix} A \begin{pmatrix} x \\ y \\ z \end{pmatrix} = 1$$

all lie on an ellipsoid

3 principal directions

—every ellipsoid can be lined up (by Q)
to the principal axes

(9)

perp axis
(e vector)

lengths (d)

$$\cancel{ax^2} + 2bxy + cy^2 = 1$$

$$A = \begin{bmatrix} a & b \\ b & c \end{bmatrix} \cancel{ax^2}$$

$$\begin{bmatrix} x & y \end{bmatrix} \begin{bmatrix} a & b \\ b & c \end{bmatrix} \begin{bmatrix} x \\ y \end{bmatrix} =$$

$$= ax^2 + 2bxy + cy^2 = 1$$

Linear algebra is giving the key facts of geometry
n dim \rightarrow no problem

Done everything except proved the key facts

10

Fact 1 | Given $A = A^H \rightarrow$ real λ 's
↳ can be complex

Proof | $Ax = \lambda x$

$x^H A x = \lambda \underbrace{x^H x}_1$
? x^H not x^T since it might be complex

$\underbrace{x^H A x}$

is also real

↳ big step

is a #

[] [] []

~~hard~~ to prove real \rightarrow take complex conj (aka Hermitian)

$(x^H A x)^H \rightarrow$ get same thing

$= x^H A x$

↳ now

Fri About positive, definite matrices and examples thereof

(11)

Can verify

$$\bar{u}(x) = \begin{bmatrix} u_1(x) \\ u_2(x) \\ \vdots \\ u_n(x) \end{bmatrix}$$

$$\bar{u}(x) = A u(x)$$

$$u(x) = e^{Ax} \cdot \bar{u}(0)$$

(Think back to basic cases)

$$A = S \Lambda S^{-1}$$

$$e^{Ax} = I + (Ax) + \frac{(Ax)^2}{2!} + \dots$$

$$= I + S(\Lambda x)S^{-1} + \frac{S(\Lambda^2 x)S^{-1}}{2!} + \dots$$

factor out S, S^{-1}

$$I = S S^{-1}$$

(2)

$$= S \left(I + \lambda A + \frac{(\lambda t)^2}{2!} + \dots \right) S^{-1}$$

$$= S \begin{pmatrix} 1 + \lambda_1 t + \frac{(\lambda_1 t)^2}{2!} + \dots & & 0 & & 0 \\ 0 & & & & \\ \vdots & & & & \\ 0 & & & & \end{pmatrix}$$

$$1 + \lambda_1 t + \frac{(\lambda_1 t)^2}{2!} + \dots$$

$$= \begin{pmatrix} e^{\lambda_1 t} & & & \\ & e^{\lambda_2 t} & & \\ & & \ddots & \\ & & & e^{\lambda_n t} \end{pmatrix}$$

Billings on Symmetric $A=A^T$

Today Positive Definite

1. ^{all} pivots > 0
2. ^{all} eigenvalues $\lambda > 0$

5 diff ways to identify

3. All "upper left" det are $(+)$

$\begin{bmatrix} 5 & 4 \\ 4 & 3 \end{bmatrix}$ all pos - good, but not $(+)$ def.

det=5	$\begin{bmatrix} 5 & \\ & 4 \end{bmatrix}$	$\oplus \oplus$
det=-1	$\begin{bmatrix} 4 & \\ & 3 \end{bmatrix}$	$\oplus \oplus$

$\textcircled{\text{No}}$ - not $(+)$ def

$\begin{bmatrix} 5 & 4 \\ 0 & -1/5 \end{bmatrix}$ $\textcircled{\text{No}}$

$\begin{bmatrix} 2 & 1 \\ 1 & 2 \end{bmatrix}$

det=2	$\begin{bmatrix} 2 & \\ & 1 \end{bmatrix}$
det=3	$\begin{bmatrix} 1 & \\ & 2 \end{bmatrix}$

$\textcircled{\text{Yes}}$

also w/ elim $\begin{bmatrix} 2 & 1 \\ 0 & 1.5 \end{bmatrix}$ $\textcircled{\text{Yes}}$

②

Also $\lambda = 3, 1$ (yes)

$$\begin{bmatrix} a & b \\ c & d \end{bmatrix} \quad \begin{array}{l} a > 0 \\ ac \geq b^2 \end{array}$$

(is it all 3 must be true, or any 1)

The more \oplus H's are \rightarrow the more \oplus it passes test
??

But note b^2

On diagonal \rightarrow big + pos

Off diagonal \rightarrow small

$$\begin{bmatrix} 5 & -4 \\ -4 & 3 \end{bmatrix} \quad \text{still no}$$

\oplus det by itself is not enough

$$\begin{bmatrix} -1 & 0 \\ 0 & -1 \end{bmatrix} \quad \text{det } \oplus \quad \text{but } |\lambda| \text{ det} = \ominus \quad \text{(No)}$$

\leftarrow is actually neg definite

3

$\begin{bmatrix} 2 & 1 \\ 1 & c \end{bmatrix}$ For which c is this \oplus definite

$$\oplus \begin{bmatrix} 2 & 1 \\ 1 & c \end{bmatrix}$$

$c > \frac{1}{2}$ for \oplus def

$$\det = 2c - 1$$

but what if $c = \frac{1}{2}$

$\det = 0$ \otimes Not pos definite

Singular

$$\lambda = 0, 2\frac{1}{2}$$

But it is \oplus semi definite

1. all $\lambda \geq 0$

2. all pivots ≥ 0

This has all the basic ideas of every chap

(4)

⊕ def rules cont → 4, A can be factored into $A = LL^T$

$$LL^T = \begin{matrix} \text{L} \\ \text{L}^T \end{matrix} \begin{pmatrix} 1 & 0 \\ 1 & 2 \end{pmatrix} \begin{pmatrix} 1 & 1 \\ 0 & 2 \end{pmatrix} =$$

Claim → will get ⊕ def matrix

How science uses it

$$= \begin{pmatrix} 1 & 1 \\ 1 & 5 \end{pmatrix} \text{ (yes) is } \oplus \text{ definite}$$

L can't be singular

↳ would be semi-definite

L cols must be ind

(i was that a rick)

5, Def: $A = A^T$ is pos def if

Verification →

$$x^T A x > 0 \text{ for all } x \neq 0$$

| x_n | $n \times n$ | $n \times 1$ = number

(5)

L can represent the energy in a system
The natural quantity

$$E = \begin{bmatrix} a & b \\ b & c \end{bmatrix}$$

using #5

$$\begin{bmatrix} x_1 & x_2 \end{bmatrix} \begin{bmatrix} a & b \\ b & c \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} = \begin{bmatrix} x_1 & x_2 \end{bmatrix} \begin{bmatrix} ax_1 + bx_2 \\ bx_1 + cx_2 \end{bmatrix}$$

~~Full~~

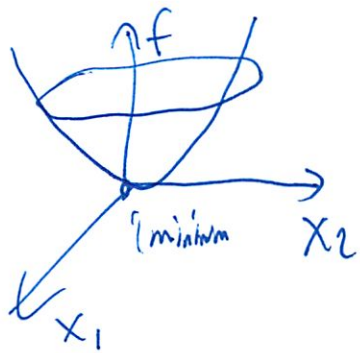
$$= ax_1^2 + 2b \cdot x_1 x_2 + cx_2^2$$

↑
cross term
dangerous the one that
can go \ominus

$$= f(x_1, x_2)$$

Should be $>$ for all $x, x \neq 0$

6



$f > 0$
→ A is pos def
→ graph is a bowl

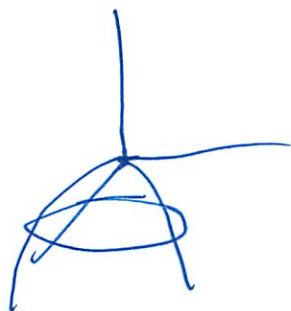
ex $A = \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix}$

$$f = x_1^2 + x_2^2$$

What does the picture look like?

$$A = \begin{bmatrix} -1 & 0 \\ 0 & -1 \end{bmatrix}$$

$$f = -x_1^2 - x_2^2$$



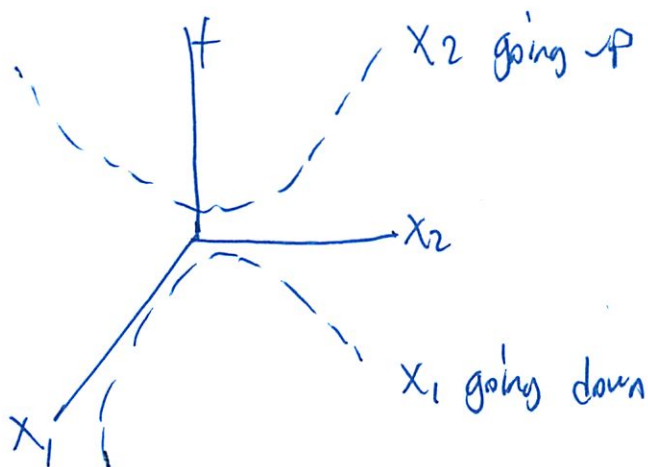
negative

7


Saddle point


$$A = \begin{pmatrix} -1 & 0 \\ 0 & 1 \end{pmatrix}$$

$$f = -x_1^2 + x_2^2$$



↑ like Rocky Mts pass

 x_2 is mts range
min for mts

 x_1 is your path
max for you

$$\begin{pmatrix} 5 & -4 \\ -4 & 3 \end{pmatrix}$$

Not \oplus def
but saddle point

Connects w/ minimum

Lin calculus \rightarrow next pg

8

$$f(x) \begin{matrix} \text{min} & f' = 0 & f'' > 0 \\ \text{max} & f' = 0 & f'' < 0 \end{matrix}$$

$$f(x, y) \begin{matrix} \text{min} \rightarrow \\ \text{max} \rightarrow \end{matrix} \left(\begin{matrix} df/dx = 0 \\ df/dy = 0 \end{matrix} \right) \left(\begin{matrix} \frac{\partial^2 f}{\partial x^2} & \frac{\partial^2 f}{\partial x \partial y} \\ \frac{\partial^2 f}{\partial x \partial y} & \frac{\partial^2 f}{\partial y^2} \end{matrix} \right) > \left(\frac{\partial^2 f}{\partial x \partial y} \right)^2$$

saddle

(controlled on what's what)

$$A = \begin{pmatrix} \frac{\partial^2 f}{\partial x^2} & \frac{\partial^2 f}{\partial x \partial y} \\ \frac{\partial^2 f}{\partial x \partial y} & \frac{\partial^2 f}{\partial y^2} \end{pmatrix} \leftarrow \text{has to be } \oplus \text{ definite}$$

↑ test is same as $ac - b^2$

$$\frac{\partial^2 f}{\partial x^2} > 0$$

← 2 part test

(9)

Proving

if 5 \rightarrow then 2 ($\lambda > 0$)

~~Solve~~

Suppose $Ax = \lambda x$

then $\underbrace{x^T A x}_{\substack{\text{condition} \\ 5}} = \lambda \underbrace{x^T x}_{\substack{\text{always} \\ +}}$

so $\lambda \oplus$

$$\lambda = \frac{x^T A x}{x^T x}$$

But 2 \rightarrow 5

Start w/ $\lambda \oplus$

Prove $x^T A x > 0$

$$\underbrace{\lambda}_{\oplus} \underbrace{x^T x}_{\oplus} = \underbrace{x^T A x}$$

So must be \oplus
but only for the eigenvectors!
have not gotten there all!

18.06 Lecture
Matrices in Engineering

4/23

(watching online)

Matrix in Engineering

↳ but I don't see that online!

Oh well - guess it's not online ...

Chap 8.1 book

$k = \text{symmetric}$ (often \oplus definite)

$$k = A^T A$$

$$k = A^T C A$$

Since $\frac{1}{2} u^T k u$ is energy (never negative)

$k = \text{stiffness matrix}$

$k^{-1} f$ is response to forces

Direct linear \rightarrow matrix eqns

Indirect continuous \rightarrow diff eqn

②

$$ku = f$$

$$M \frac{d^2 u}{dt^2} + k u = f$$

Then $kx = \lambda Mx$

(What's the point of this chap??)

A Line of Springs

fixed-fixed $\rightarrow k_0 \quad A_0^T \quad C_0 \quad A_0$

fixed-free $\rightarrow k_1 \quad A_1^T \quad C_1 \quad A_1$

free-free $\rightarrow k_{\text{sinusoidal}}$



$u = (u_1, u_2, u_3)$ = movement of the masses
up + down

$Y = (Y_1, Y_2, Y_3, Y_4)$ or (Y_1, Y_2, Y_3) = tension in
the springs

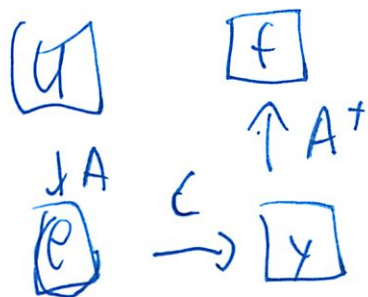
(3)

u = Moments of n masses

e = Elongations of m springs

Y = Internal forces in m springs

f = External forces



(a bunch more eq)

Remaining eqns - matrices smaller

Then when 2×2 one spring is free

(can be circular)

Can convert to continuous

$$A^T (A u(x)) = f(x) \text{ is}$$

$$-\frac{d}{dx} \left(c(x) \frac{du}{dx} \right) = f(x)$$

(I hope we don't actually have to know)

18.06

4/203

Last time notes from Day

$$A^2 = A \rightarrow R = Q\sqrt{\Lambda}Q^T$$

$$A = \begin{bmatrix} 3 & 1 \\ 1 & 3 \end{bmatrix}$$

$$S=Q = \frac{1}{\sqrt{2}} \begin{bmatrix} 1 & 1 \\ 1 & -1 \end{bmatrix}$$

$$R = Q \begin{bmatrix} 2 & 0 \\ 0 & \sqrt{2} \end{bmatrix} Q^T$$

$$= \frac{1}{2} \begin{bmatrix} 1 & 1 \\ 1 & -1 \end{bmatrix} \begin{bmatrix} 2 & 0 \\ 0 & \sqrt{2} \end{bmatrix} \begin{bmatrix} 1 & 1 \\ 1 & -1 \end{bmatrix}$$

Big Fact If $A = R^T R$ and R has ind

cols show $\bar{x}^T A \bar{x} > 0$ for $\bar{x} \neq 0$
(ie A pos def.)

$$\begin{aligned} x^T A x &= x^T R^T R x = (x^T R^T) (R x) \\ &= (R x^T) (R x) > 0 \end{aligned}$$

($R x = 0$ b/c R has ind cols)

②

Recall

$$A^H \text{ (hermitian) } = A^T$$

$$x^H A^H x > 0 \rightarrow \text{Pos Def}$$

OH

4/25

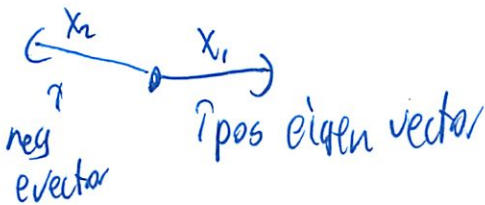
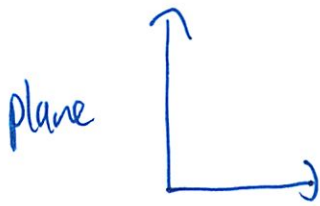
Wed - Symmetric (8.4)

Fri - Pos definite (6.5)

Mon - App to Engineering (8.1)

Today - Similar

Ex # 2 $\rightarrow A_1$ has one \oplus , one \ominus λ



$\lambda \approx 0$, little
 $|\lambda|$, little

~~the~~ ^{pos} eigen vector is w/ eig. value that is pos

 A symmetric S vectors are orthogonal

②

$$\lambda_n^T (A x_n)$$



Dot prod $x_1 \cdot Ax_1$ is \oplus since point in same dir
 $x_2 \cdot Ax_2$ \ominus diff
? Eigenvector

So x_2 is a good candidate for a vector

Such that $x A x^T < 0$

Since x_2 is an eigenvector for a \ominus
Eigenvale

18.06 Spring 2012 – Problem Set 8

This problem set is due Thursday, April 26th, 2012 at 4pm (hand in to Room 2-106). The textbook problems are out of the 4th edition. For computational problems, please include a printout of the code with the problem set (for MATLAB in particular, `diary('filename')` will start a transcript session, `diary off` will end one.)

Every problem is worth 10 points.

1. Do Problem 3 from Section 6.4.
2. Do Problem 22 from Section 6.4 (now a paradox for you).
3. Do Problem 23 from Section 6.4.
4. Do Problem 30 from Section 6.4.
5. Do Problem 2 & 14 from Section 6.5.
6. Do Problem 28 from Section 6.5.
7. Do Problem 31 from Section 6.5.
8. Do Problem 35 from Section 6.5.
9. Do Problem 7 from Section 8.1.
10. Do Problem 11 from Section 8.1.

18.06 Wisdom.

- Your exams so far make up 30% of your final grade. If you wish to maximize your performance in 18.06, it is all about practice (without this, you won't know what you don't know either). Make a weekly study plan involving doing as many old exams as you can fit in your schedule (find them on the 18.06 web). Organize your work. Be systematic. Learn to check your results. Keep a list of all questions (and confusing concepts) you encounter on your journey, and ask your TAs to explain all the hows and whys. We are here for the same!
- Also try out the great Math Learning Center, that you can visit and ask the instructors your questions in a friendly atmosphere, and at evening hours too: *Room 2-102, Mon-Thu, 3:00–5:00pm and 7:30–9:30pm.*

Michael Plasmie

P-Set 8 86/100

4/23

1, 6, 4 #3 Find the eigenvalues and the unit
eigenvectors of

$$\begin{bmatrix} 2 & 2 & 2 \\ 2 & 0 & 0 \\ 2 & 0 & 0 \end{bmatrix}$$

Matlab

$$\lambda = -2, 0, 4$$

← you should do this by hand!
(-2)

$$V = \begin{bmatrix} -1/\sqrt{3} & 0 & \sqrt{2/3} \\ 1/\sqrt{3} & -1/\sqrt{2} & 1/\sqrt{6} \\ 1/\sqrt{3} & 1/\sqrt{2} & 1/\sqrt{6} \end{bmatrix}$$

Book / Unit vectors

$$\begin{bmatrix} 0 \\ 1 \\ -1 \end{bmatrix} / \sqrt{2}$$

$$\begin{bmatrix} 2 \\ 1 \\ 1 \end{bmatrix} / \sqrt{6}$$

$$\begin{bmatrix} 1 \\ -1 \\ -1 \end{bmatrix} / \sqrt{3}$$

Same as what we had...

②

6.4 # 22 ① If $AA^T = A^T A$ then A and A^T

10/10 share the same eigenvectors. (true)

② A and A^T always share the same eigenvalues,

③ Find the flags they have the same S and Λ

$$\text{So } A = A^T$$

6.4 # 22

OH | Presents w/ false argument \rightarrow asks you to find mistake

① Remember if commute \rightarrow share same eigenvector

\hookrightarrow from end of 6.2

② From $\det(A - \lambda I) = 0$

$$\text{for } A^T \det(A^T - \lambda I) = 0$$

Since how doing det does not change if you transpose it

$$\text{Since } \det(A) = \det(A^T)$$

③ Have same eigenvalues

But only same eigenvalues in that form! Same

Hint: Order of eigenvalues and eigenvectors diff

for A and A^T - can't sub one in for the other

(3)

6.3 # 23 Which of these classes of Matrices does this belong to

9/10

$$A = \begin{bmatrix} 0 & 0 & 1 \\ 0 & 1 & 0 \\ 1 & 0 & 0 \end{bmatrix}$$

- invertible ✓
- Orthogonal ✓
- projection ✗
- permutable ✓
- diagonalizable ✓
- Markov ✓

$$B = \frac{1}{5} \begin{bmatrix} 1 & 1 & 1 \\ 1 & 1 & 1 \\ 1 & 1 & 1 \end{bmatrix}$$

- invertible ✗
- Orthogonal ✗
- projection ✓ $P^2 = P$ and $P = P^T$
- permutable ✗
- diagonalizable ✓
- Markov ✓

What factorizations are possible?

- LU ✗
- QR ✓
- SAS⁻¹ ✓
- Q Λ Q⁻¹ ✓

- LU ✗ lower-upper
 - QR ✗ Gram-Schmidt
 - SAS⁻¹ ✓ has an inverse
 - Q Λ Q⁻¹ ✓ symmetric
- is this wrong in the ans
~~prob not~~

(36)

(OH) Can do SAS^{-1} whenever you have a complete set
of eigenvalues and eigenvectors \rightarrow can do
Does not matter if invertible

4

6.4 # 30

If λ_{\max} has the largest eigenvalue of symmetric matrix A , no diagonal entry can be larger than λ_{\max} . What is the 1st entry a_{11} of $A = Q \Lambda Q^T$. Show why $a_{11} \leq \lambda_{\max}$

10/10

a_{11} of $A = Q \Lambda Q^T$. Show why $a_{11} \leq \lambda_{\max}$

In book

$$\begin{bmatrix} q_{11} & \dots & q_{1n} \\ \vdots & & \vdots \\ q_{n1} & \dots & q_{nn} \end{bmatrix} \begin{bmatrix} \lambda_1 q_{11} \\ \vdots \\ \lambda_n q_{1n} \end{bmatrix} \leq \left(|q_{11}|^2 + \dots + |q_{1n}|^2 \right) = \lambda_{\max}$$

Why is this λ_{\max} ?

OH! If diagonal matrix, diagonals are λ_s
But not necessarily diagonal matrix here
Any symmetric matrix A

46

Q = orthogonal

↳ essentially eigenvectors orthogonal

Have matrix as product of 3 matrices, so have an expression for it

$$\begin{array}{c|c} & \\ \hline & \end{array} \times \begin{array}{c|c} & \\ \hline & \end{array} = \begin{array}{c|c} & \\ \hline & \end{array} \times \begin{array}{c|c} & \\ \hline & \end{array} = \begin{array}{c|c} & \\ \hline & \end{array}$$

$$\begin{array}{c|c} & \\ \hline & \\ \hline & \end{array} \times \begin{array}{c|c} & \\ \hline & \\ \hline & \end{array} = \begin{array}{ccc|c} \textcircled{1} & \textcircled{2} & \textcircled{3} & \\ \hline 0 & 0 & 0 & q_{11} \\ 0 & 0 & 0 & q_{12} \\ 0 & 0 & 0 & q_{13} \end{array} \begin{array}{c|c} & \\ \hline & \end{array} = \begin{array}{ccc} & 0 & 0 \\ & 0 & 0 \\ & 0 & 0 \end{array}$$

1st group $q_{11} \cdot \lambda_{11} + q_{12} \cdot \lambda_{21} + q_{13} \cdot \lambda_{31}$

② $q_{12} \cdot \lambda_{22}$

③ $q_{13} \cdot \lambda_{33}$

$$a_{11} = q_{11}(q_{11} \lambda_{11}) + q_{12}(q_{12} \lambda_{22}) + q_{13}(q_{13} \lambda_{33})$$

$$= q_{11}^2 \lambda_{11} + q_{12}^2 \lambda_{22} + q_{13}^2 \lambda_{33} + \dots$$

(16)

Q is orthogonal — so rows + cols are orthonormal

↑ 1 word for matrices

↑ 1 word applies to rows + cols

$$\text{So } q_{ij} < 1$$

$$\text{Also } \sum q_{ij}^2 = 1$$

$$\text{So } a_{ii} \leq \lambda_{\max}$$

↑ if replace all coeff w/ largest one (λ_{\max})

So $\rho \leq \lambda_{\max}$ at most

5

6.5 #2. Which of A_1, A_2, A_3, A_4 has 2 positive eigenvalues?
10/10

$$A_1 = \begin{bmatrix} 5 & 6 \\ 6 & 7 \end{bmatrix} \quad A_2 = \begin{bmatrix} -1 & -2 \\ -2 & -5 \end{bmatrix}$$

$$A_3 = \begin{bmatrix} 1 & 10 \\ 10 & 100 \end{bmatrix} \quad A_4 = \begin{bmatrix} 1 & 10 \\ 10 & 101 \end{bmatrix}$$

So that means which is positive definite

$$a > 0 \quad ac - b^2 > 0$$

- $A_1 \rightarrow 35 - 36 \otimes$
- ~~A_2~~
- $A_3 \rightarrow 100 - 100 \otimes$ Semi
- $A_4 \rightarrow 101 - 100 \odot$

11 and A_4

Find an x such that $x^T A_i x < 0$ so a fails the test
(which test is this)

Oh $x^T A x > 0$ for every non 0 x

$$ax^2 + 2bxy + cy^2 < 0$$

$$x = \begin{bmatrix} x \\ y \end{bmatrix}$$

to fail

6

$$5x^2 + 2(6)xy + 7y^2 \leq 0$$

$$5x^2 + 12xy + 7y^2 \leq 0$$

if both -, the value will be neg

Wolfram

$$x < 0$$

Alpha

$$-\frac{5x}{17} < y < -x$$

So one is $\begin{bmatrix} 1 \\ -1 \end{bmatrix}$

$$5(1)^2 + 12(1)(-1) + 7(-1)^2$$

$$5 - 12 + 7 = 0$$

which is not > 0

↳ like semi definite or something like that

7

6.5 #14 If A is pos def \rightarrow then A^{-1} is pos definite

Best proof: eigenvalues A^{-1} are pos since

\hookrightarrow Saw on last qv A, A^T are same eigenvalues

Here A, A^T eigenvalues are pos iff pivots are pos

But A^{-1} and A^T ?

We know $Q^{-1} = Q^T$ w/ orthonormal

Book)

$$\lambda(A^{-1}) = \frac{1}{\lambda(A)}$$

\hookrightarrow Oh deh!

2nd proof (for 2×2)

$$\text{entries of } A^{-1} = \frac{1}{ac-b^2} \begin{bmatrix} c & -b \\ -b & a \end{bmatrix}$$

passes the det test

\hookrightarrow So eigenvalues pos if $a > 0$
 $ac - b^2 > 0$ \square

Q.5 # 28

10/10

Without multiplying

$$A = \begin{bmatrix} \cos \theta & -\sin \theta \\ \sin \theta & \cos \theta \end{bmatrix} \begin{bmatrix} 2 & 0 \\ 0 & 5 \end{bmatrix}$$

$$\hookrightarrow \begin{bmatrix} \cos \theta & \sin \theta \\ -\sin \theta & \cos \theta \end{bmatrix}$$

Find tr

A) \det of A

is this $Q \Lambda Q^T$?

Q is Q^T

but how do you reconstruct fact?

(? and what does this do w/ \oplus \det ?)

$$\text{So } \det(A - \lambda I) = 0$$

~~or~~

Help | \det product of eigenvalues

$$2 \cdot 5 = 10$$

(remember these rules)

9

b) eigenvalues

$$\lambda = 2, 5$$

← easy

c) eigenvectors

$$x = \begin{bmatrix} \cos \theta \\ \sin \theta \end{bmatrix}, \begin{bmatrix} -\sin \theta \\ \cos \theta \end{bmatrix}$$

d) why its pos def

↳ all ~~the~~ eigenvalues are \oplus

a) So could use these for det?

↳ but you find it 1st ...

⑩

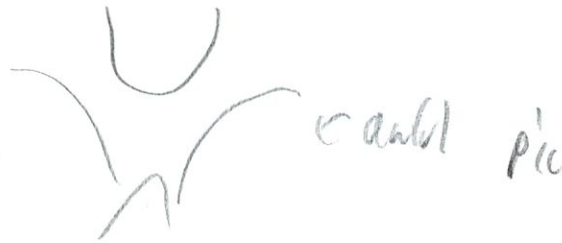
6.5 # 3) Which values of c give a bowl and

10/10

which a Saddle pt for

$$z = 4x^2 + 12xy + cy^2$$

So Saddle pt



transitions at 9

< 9 Saddle pt

> 9 bowl

$= 9$ trough

How do you tell?

Help!

Oh do pos def

det > 0 means local min \rightarrow ie bowl

So positive def if > 9 / bowl

(11)

6.5 #35 Suppose C is pos definite (so $y^T C y > 0$

10/6 when $y \neq 0$) and A has 'ind cols'
(so $Ax \neq 0$ when $x \neq 0$)

Apply $x^T A^T C A x$ to show $A^T C A$
is \oplus definite

So $x^T A x > 0$ for every non zero vector x

What is here to show?

Also need the other facts

- all pivots \oplus

We know this applies to C

But how to apply to A ?

Help

Just plug in

Get the definition

$A^T C A$ is pos def \checkmark

(12)

Pr. # 7

For 5 springs and 4 masses w/ both ends fixed, what are A, C, k ?

7/10

$$\begin{aligned}
 e_1 &= u_1 \\
 e_2 &= u_2 - u_1 \\
 e_3 &= u_3 - u_2 \\
 e_4 &= u_4 - u_3 \\
 e_5 &= -u_4
 \end{aligned}$$



$$e = Au$$

$$A = \begin{pmatrix} 1 & 0 & 0 & 0 \\ -1 & 1 & 0 & 0 \\ 0 & -1 & 1 & 0 \\ 0 & 0 & -1 & 1 \\ 0 & 0 & 0 & -1 \end{pmatrix}$$

from book

$$e = Au$$

$$x C = \begin{pmatrix} c_1 & & & \\ & c_2 & & \\ & & 1 & \\ & & & c_5 \end{pmatrix}$$

$$y = Ce$$

(13)

force balance

$$f_1 = y_1 - y_2$$

$$f_2 = y_2 - y_3$$

$$f_3 = y_3 - y_4$$

$$f_4 = y_4 - y_5$$

Don't need

$$k = A^T C A$$

$$k_0 = \begin{pmatrix} 2 & -1 & 0 & 0 \\ -1 & 2 & -1 & 0 \\ 0 & -1 & 2 & -1 \\ 0 & 0 & -1 & 2 \end{pmatrix}$$

(14)

8.1 # 11 (MATLAB) Chem e has 1st deriv

2/10

$\frac{du}{dx}$ from fluid velocity

$\frac{d^2u}{dx^2}$ from diffusion

Replace $\frac{du}{dx}$ by a forward diff

then a centered diff

then a backwards diff

$$\Delta x = \frac{1}{8}$$

Graph sols of $-\frac{d^2u}{dx^2} + 10 \frac{du}{dx} = 1$
 $u(0) = u(1) = 0$

∴ What is a forward difference

Mathworld $\Delta a_n = a_{n+1} - a_n$

Central $\delta(f_n) = \delta_n = f_{n+1/2} - f_{n-1/2}$

backward $\nabla_p = f_p - f_{p-1}$

(5)

I don't get this problem at all ...

Book

$$E = \text{diag}(\text{ones}(6,1) \ 1)$$

$$k = 64 \cdot 2(\text{eye}(7) - E - E')$$

$$D = 80 \cdot (E - \text{eye}(7))$$

$$(k+D) \setminus \text{ones}(7,1) \quad \leftarrow \text{forward}$$

$$(k-D) \setminus \text{ones}(7,1) \quad \leftarrow \text{backward}$$

$$(k+D)/2 - D'/2 \setminus \text{ones}(7,1) \quad \leftarrow \text{centered}$$

What are E, k, D ? Seem so random!

What is \setminus ?

Yeah I don't get this at all

```
>> E = diag(ones(6,1),1)
```

```
E =
```

```

0    1    0    0    0    0    0
0    0    1    0    0    0    0
0    0    0    1    0    0    0
0    0    0    0    1    0    0
0    0    0    0    0    1    0
0    0    0    0    0    0    1
0    0    0    0    0    0    0

```

```
>> K = 64*2*(eye(7)-E-E')
```

```
K =
```

```

128  -128   0    0    0    0    0
-128  128  -128   0    0    0    0
  0  -128  128  -128   0    0    0
  0    0  -128  128  -128   0    0
  0    0    0  -128  128  -128   0
  0    0    0    0  -128  128  -128
  0    0    0    0    0  -128  128

```

```
>> D=80*(E-eye(7))
```

```
D =
```

```

-80   80    0    0    0    0    0
  0  -80   80    0    0    0    0
  0    0  -80   80    0    0    0
  0    0    0  -80   80    0    0
  0    0    0    0  -80   80    0
  0    0    0    0    0  -80   80
  0    0    0    0    0    0  -80

```

```
>> (K+D)\ones(7,1)
```

```
ans =
```

```

-0.0202
-0.0410
-0.0080
 0.0806
 0.0811
-0.1547
-0.3918

```

```
>> (K-D)\ones(7,1)
```

```
ans =
```



```
0.0391
0.0343
0.0054
-0.0205
-0.0287
-0.0208
-0.0080
```

```
>> (K+D/2-D'/2)\ones(7,1)
```

```
ans =
```

```
0.0012
-0.0096
-0.0276
-0.0332
-0.0069
0.0420
0.0629
```

18.06 Spring 2012 – Problem Set 8

This problem set is due Thursday, April 26th, 2012 at 4pm (hand in to Room 2-106). The textbook problems are out of the 4th edition. For computational problems, please include a printout of the code with the problem set (for MATLAB in particular, `diary('filename')` will start a transcript session, `diary off` will end one.)

Every problem is worth 10 points.

1. Do Problem 3 from Section 6.4.

Solution. Solve $\det(A - \lambda I) = 0$ to get the eigenvalues 0, -2, 4. Then find unit vectors in the nullspaces of the matrices $A - 0I$, $A + 2I$, $A - 4I$ to get the unit eigenvectors $\frac{1}{\sqrt{2}}[0, 1, -1]^T$, $\frac{1}{\sqrt{3}}[1, -1, -1]^T$, and $\frac{1}{\sqrt{6}}[2, 1, 1]^T$. \square

2. Do Problem 22 from Section 6.4 (now a paradox for you).

Solution. The flaw is that the correspondence between eigenvalues and eigenvectors need not be the same for A and A^T . For example, the matrix

$$\begin{bmatrix} 0 & 1 \\ -1 & 0 \end{bmatrix}$$

has eigenvalues i and $-i$ with corresponding eigenvectors $[1, i]^T$ and $[i, 1]^T$, while the transpose

$$\begin{bmatrix} 0 & -1 \\ 1 & 0 \end{bmatrix}$$

has eigenvalues i and $-i$ with corresponding eigenvectors $[i, 1]^T$ and $[1, i]^T$. \square

3. Do Problem 23 from Section 6.4.

Solution. A is invertible, orthogonal, permutation, diagonalizable, Markov. A has possible factorizations QR , $S\Lambda S^{-1}$ and $Q\Lambda Q^T$.

B is projection, diagonalizable and Markov. B has possible factorizations LU , QR , $S\Lambda S^{-1}$ and $Q\Lambda Q^T$; note that the matrices U and R have only one non-zero diagonal entry. \square

4. Do Problem 30 from Section 6.4.

Solution. From the decomposition $A = Q\Lambda Q^T$ we have $a_{11} = \sum_i \lambda_i q_{1i}^2$. From here, we get $a_{11} \leq \sum_i \lambda_{\max} q_{1i}^2 = \lambda_{\max} \sum_i q_{1i}^2 = \lambda_{\max}$. \square

5. Do Problem 2 & 14 from Section 6.5.

Solution. **Problem 2** Only A_4 has two positive eigenvalues. The vector $x = [-7, 6]^T$ has $x^T Ax = -7 < 0$.

Problem 14 The eigenvalues of A^{-1} are positive because they are inverses of the eigenvalues of A . For the 2 by 2 case, one can also check directly that the entries of A^{-1} pass the determinant tests. \square

6. Do Problem 28 from Section 6.5.

Solution. Use the product formula for determinants to get:

$$\det(A) = 1 \times 10 \times 1 = 10.$$

We recognize that A is factorised in the form $Q\Lambda Q^T$. Thus it has eigenvalues $\lambda_1 = 2$, $\lambda_2 = 5$, and corresponding eigenvectors $x_1 = (\cos \theta, \sin \theta)^T$ and $x_2 = (-\sin \theta, \cos \theta)^T$. Both eigenvalues are positive, so the matrix is positive definite. \square

7. Do Problem 31 from Section 6.5.

Solution. Complete the square to get

$$z = 4x^2 + 12xy + cy^2 = (2x + 3y)^2 + (c - 9)y^2.$$

Thus for $c > 9$ the graph of z is a bowl, and for $c < 9$ the graph has a saddle point. When $c = 9$, the graph, $z = (2x + 3y)^2$, is a "trough" staying at zero along the line $2x + 3y = 0$. \square

8. Do Problem 35 from Section 6.5.

Solution. "Factor" the tranposes to get

$$x^T A^T C A x = (Ax)^T C (Ax).$$

Since C is assumed positive definite, this energy can drop to zero only when $Ax = 0$. Since A is assumed to have independent columns, $Ax = 0$ only happens when $x = 0$. Thus $A^T C A$ has positive energy and is positive definite. \square

9. Do Problem 7 from Section 8.1.

Solution. (To first of the above) For 5 springs and 4 masses, the 5 by 4 matrix A has two non-zero diagonals: all $a_{ii} = 1$ and $a_{a+1,i} = -1$. With $C = \text{diag}(c_1, c_2, c_3, c_4, c_5)$, we get $K = A^T C A$, symmetric tridiagonal with diagonal entries $K_{ii} = c_i + c_{i+1}$ and off-diagonals $K_{i+1,i} = -c_{i+1}$. With $C = I$ this K is the $-1, 2, -1$ matrix, and $K(2, 3, 3, 2) = (1, 1, 1, 1)$ solves $Ku = \text{ones}(4, 1)$. (K^{-1} will solve $Ku = \text{ones}(4)$.) \square

10. Do Problem 11 from Section 8.1. There was a book version/numbering issue. Here are the two "problems 8.1.11" in question:

- 11 (MATLAB) Find the displacements $u(1), \dots, u(100)$ of 100 masses connected by springs all with $c = 1$. Each force is $f(i) = .01$. Print graphs of u with fixed-fixed and fixed-free ends. Note that $\text{diag}(\text{ones}(n, 1), d)$ is a matrix with n ones along diagonal d . This print command will graph a vector u :

```
plot(u, '+'); xlabel('mass number'); ylabel('movement'); print
```

- 12 (MATLAB) Chemical engineering has a first derivative du/dx from fluid velocity as well as d^2u/dx^2 from diffusion. Replace du/dx by a forward difference, then a centered difference, then a backward difference, with $\Delta x = \frac{1}{8}$. Graph your three numerical solutions of

$$-\frac{d^2u}{dx^2} + 10 \frac{du}{dx} = 1 \text{ with } u(0) = u(1) = 0.$$

This *convection-diffusion equation* appears everywhere. It transforms to the Black-Scholes equation for option prices in mathematical finance.

Problem 12 is developed into the first MATLAB homework in my 18.085 course on Computational Science and Engineering at MIT. Videos on ocw.mit.edu.

Solution. To "11" in the above snippet: The two graphs of 100 points are "discrete parabolas" starting at $(0, 0)$: symmetric around 50 in the fixed-fixed case, ending with slope zero in the fixed-free case.

To "12" in the above snippet: Forward/backward/centered for du/dx has a big effect because that term has the large coefficient. MATLAB:

```
E = diag(ones(6, 1);
K = 64 * (2 * eye(7) - E - E');
D = 80 * (E - eye(7));
K + D\ones(7, 1); % forward
(K - D')\ones(7, 1); % backward
(K + D/2 - D'/2)\ones(7, 1); % centered is usually the best : more accurate
```

□

18.06 Wisdom.

- Your exams so far make up 30% of your final grade. If you wish to maximize your performance in 18.06, it is all about practice (without this, you won't know what you don't know either). Make a weekly study plan involving doing as many old exams as you can fit in your schedule (find them on the 18.06 web). Organize your work. Be systematic. Learn to check your results. Keep a list of all questions (and confusing concepts) you encounter on your journey, and ask your TAs to explain all the hows and whys. We are here for the same!
- Also try out the great Math Learning Center, that you can visit and ask the instructors your questions in a friendly atmosphere, and at evening hours too: *Room 2-102, Mon-Thu, 3:00-5:00pm and 7:30-9:30pm.*

Avg 6 pts later

Do your own HW - don't just copy from book

Diagonalization

Input: Square matrix A

Output: Factorization $S \Lambda S^{-1}$

\uparrow invertable \uparrow diagonal $[\lambda_1 \dots \lambda_n]$

But not always possible

a) Diagonalize $\begin{bmatrix} 3 & 0 & -2 \\ 0 & 8 & 2 \\ 0 & -3 & 1 \end{bmatrix}$

1. Find λ_s

2. Find eigenvectors

②

1. λ] $\det(A - \lambda I) = 0$ solve for λ

$$A - \lambda I = \begin{bmatrix} 3-\lambda & 0 & -2 \\ 0 & 8-\lambda & 2 \\ 0 & -3 & 1-\lambda \end{bmatrix}$$

find det

$$\begin{bmatrix} 3-\lambda & 0 & -2 & 3-\lambda & 0 \\ 0 & 8-\lambda & 2 & 0 & 8-\lambda \\ 0 & -3 & 1-\lambda & 0 & -3 \\ \ominus & \ominus & \ominus & \oplus & \oplus & \oplus \end{bmatrix}$$

$$= (3-\lambda)(8-\lambda)(1-\lambda) - (3-\lambda)(2)(-3)$$

$$= (3-\lambda) \left[(8-\lambda)(1-\lambda) - (2)(-3) \right]$$

$$= (3-\lambda) (8 - 9\lambda + \lambda^2 + 6)$$

$$= (3-\lambda) (14 - 9\lambda + \lambda^2)$$

③

$$= (3-\lambda)(\lambda-2)(\lambda-7)$$

$$\begin{aligned} \lambda_1 &= 2 \\ \lambda_2 &= 3 \\ \lambda_3 &= 7 \end{aligned} \quad \leftarrow \text{all distinct} \\ \text{so able to find } S$$

Q: How to be sure not linear comb of

$$Ax_1 = \lambda_1 x_1$$

$$? \text{ if } x_1 = x_2$$

$$\text{then } \lambda_1 \stackrel{\text{must}}{=} \lambda_2$$

Contradiction ■

Eigenvector x_1 corresponding to $\lambda_1 = 2$

$$x_1 \text{ satisfies } Ax_1 = \lambda_1 x_1$$

$$\Downarrow \\ (A - \lambda_1 I)x_1 = 0$$

4

$$A - \lambda_1 I = \begin{bmatrix} 1 & 0 & -2 \\ 0 & 6 & 2 \\ 0 & -3 & -1 \end{bmatrix}$$

To find $N(A - \lambda_1 I)$ row ops!

$$\begin{bmatrix} 1 & 0 & -2 \\ 0 & 6 & 2 \\ 0 & 0 & 0 \end{bmatrix}$$

So nullspace $\Rightarrow x_1 = \begin{bmatrix} a \\ b \\ c \end{bmatrix}$

$$a - 2c = 0$$

$$6b + 2c = 0$$

$$b = 1$$

$$c = -3$$

$$a = -6$$

5
Do the rest

$$\lambda_2 = \begin{bmatrix} 1 \\ 8 \end{bmatrix}$$

$$\lambda_3 = \begin{bmatrix} -1 \\ -4 \\ 2 \end{bmatrix}$$

$$A = S \Lambda S^{-1}$$

\uparrow \uparrow
 $n \times n$ $n \times n$

~~[scribble]~~

$$S = \begin{bmatrix} -6 & 1 & -1 \\ 1 & 0 & -4 \\ -3 & 0 & 2 \end{bmatrix}$$

$$\Lambda = \begin{bmatrix} 2 & 0 & 0 \\ 0 & 3 & 0 \\ 0 & 0 & 7 \end{bmatrix}$$

$$S^{-1} = \begin{bmatrix} 0 & -1/5 & -2/5 \\ 1 & -3/2 & -5/2 \\ 0 & -3/10 & -1/10 \end{bmatrix}$$

← I missed how we got that be hand

Achad's Gauss Jordan
[S | I]

6)
b) Diagonalize $A = \begin{bmatrix} 1 & 1 \\ 0 & 1 \end{bmatrix}$

$$\det(A - \lambda I) = \det \begin{bmatrix} 1-\lambda & 1 \\ 0 & 1-\lambda \end{bmatrix}$$
$$= (1-\lambda)^2$$
$$\lambda = 1$$

eigenvector

vector x s.t.,

$$Ax = \lambda x \Leftrightarrow (A - I)x = 0$$

$$A - I = \begin{bmatrix} 0 & 1 \\ 0 & 0 \end{bmatrix}$$

$$x = \begin{pmatrix} 1 \\ 0 \end{pmatrix}$$

Not diagonalizable

Generally not true that repeated eigenvalues means
not diagonalizable \downarrow

⑦

$$A = \begin{bmatrix} 2 & 0 \\ 0 & 2 \end{bmatrix}$$

$$\det(A - \lambda I) = \det \begin{pmatrix} 2 - \lambda & 0 \\ 0 & 2 - \lambda \end{pmatrix}$$

$$\begin{pmatrix} 1 \\ 0 \end{pmatrix}, \begin{pmatrix} 0 \\ 1 \end{pmatrix} \in N(A - 2I)$$

↑ are in nullspace

$$A = \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix} \begin{bmatrix} 2 & 0 \\ 0 & 2 \end{bmatrix} \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix}$$

↑ is diagonalizable

Why do we care about diagonalizable matrices?

c) Compute A^2, A^3, A^4, \dots

$$A = S \Lambda S^{-1}$$

$$A^2 = S \Lambda S^{-1} S \Lambda S^{-1}$$

$$= S \Lambda I \Lambda S^{-1}$$

$$= S \Lambda^2 S^{-1}$$

8

$$A^3 = S \Lambda^3 S^{-1}$$

$$\Lambda^n = \begin{bmatrix} 2^n & 0 & 0 \\ 0 & 3^n & 0 \\ 0 & 0 & 7^n \end{bmatrix}$$

So only 3 matrix multiplications

(This was all over the last p-set)
(will be on next exam)

d) Markov matrices

$$\sum \text{cols} = 1$$

all entries > 0

$$A = \begin{bmatrix} 1/2 & 1/5 \\ 1/1 & 1/5 \end{bmatrix}$$

Fun fact $\lambda = 1$ is always an eigenvalue
for a Markov matrix

$$\text{Trace } A = \lambda_1 + \lambda_2$$

$$1.4 = 1 + \lambda_2$$

$$\lambda_2 = \cancel{1.4}, 4$$

⑨

$$\lambda_1 = 1 \quad x_1 = \begin{pmatrix} 5 \\ 1 \end{pmatrix}$$

$$A - I = \begin{pmatrix} -0.1 & 1.5 \\ 1 & -1.5 \end{pmatrix}$$

$$\lambda = .4$$

$$A - .4I = \begin{pmatrix} .5 & 1.5 \\ 1 & 1 \end{pmatrix}$$

$$x_2 = \begin{pmatrix} -1 \\ 1 \end{pmatrix}$$

$$S = \begin{bmatrix} 5 & -1 \\ 1 & 1 \end{bmatrix}$$

$$\lambda = \begin{bmatrix} 1 & 0 \\ 0 & .4 \end{bmatrix}$$

Steady State

$$\begin{aligned} u_\infty &= A u_\infty = \\ &= A^2 u_\infty \\ &= \dots \\ &= A^\infty u_\infty \end{aligned}$$

(10)

Steady state vector is the eigenvector corresponding to $\underline{1}$

$$u_{\infty} = \begin{pmatrix} 5 \\ 1 \end{pmatrix}$$

Solving Diff Eq

Think back to 18.03

What is the sol'n to $x'(t) = a x(t)$?

guess $x = e^{\lambda t}$

$$\lambda e^{\lambda t} = a e^{\lambda t}$$

$$\lambda = a$$

$$x(t) = x(0) e^{at}$$

Generalized: What is the sol to

$$\bar{u}'(t) = a \bar{u}(t)$$

$$\bar{u}(t) = \bar{u}(0) \cdot e^{at}$$

18.06
Similar Matrices

"Similar" matrices

↳ same eigenvalues

$$\underline{B} = \underline{M}^{-1} \underline{A} \underline{M}$$

example

A is similar to Λ ; choose $M = S =$ e vectors

(missed first few min of talk)

More detail needed if repeated λ s

Was once climax of linear algebra

↳ Jordan Form

↳ what to do to diagonalize if not full set
 eigenvectors

If n ind eigenvectors then $S^{-1}AS = \Lambda$

↳ Jordan form

But what if repeated λ , 1 eigenvector

$$A = \begin{bmatrix} 2 & 1 \\ 4 & -2 \end{bmatrix}$$

(2)

$$\lambda = 0, 0$$

↳ Not similar to $A = \begin{bmatrix} 0 & 0 \\ 0 & 0 \end{bmatrix}$

But how close can we get?

$$J = \begin{bmatrix} 0 & 1 \\ 0 & 0 \end{bmatrix} \leftarrow \text{put a 1 here}$$

It will be triangular

So this is similar

$$J = M^{-1} A M$$

for some M

3x3 Jordan

suppose $\lambda = 0, 0, 0$

he found a family of matrices

↳ everyone in the family similar

$$J = \begin{bmatrix} 0 & 0 & 0 \\ 0 & 0 & 0 \\ 0 & 0 & 0 \end{bmatrix} \text{ or } \begin{bmatrix} 0 & 1 & 0 \\ 0 & 0 & 0 \\ 0 & 0 & 0 \end{bmatrix} \text{ or } \begin{bmatrix} 0 & 1 & 0 \\ 0 & 0 & 1 \\ 0 & 0 & 0 \end{bmatrix}$$

↑ lower - no one similar rank 1

↑ rank 2

↑ diff families

③

To be similar, A, B have same rank

↳ when you multiply by an invertible matrix
you don't change the rank

So transfer it over

$$MBM^{-1} = A$$

$$\text{rank } A \leq \text{rank } B$$

↳ when multiple ~~the~~ might get smaller

↳ but when multiplying by invertible \Rightarrow no change

Why do B and A have the same eigenvalue?

Suppose $Bx = \lambda x$

Then $M^{-1}AMx = \lambda x$

But show A has eigenvalue same

Note eigenvalues diff

Multiply both sides by M

$$A(Mx) = \lambda(Mx)$$

(what happened for eigenvector)

(4)

Proof 2

know $\det(B - \lambda I) = 0$

know $\det(M^{-1}AM - \lambda I) = 0$

Show $\det(A - \lambda I) = 0$

So

$$\det(M^{-1}AM - \lambda I) = 0$$

$$\det(M^{-1}AM - \lambda M^{-1}M) = 0$$

$$\det(M^{-1}(A - \lambda I)M) = 0$$

$$\det(M^{-1}) \overset{\text{product rule}}{\cdot} \det(A - \lambda I) \cdot \det(M) = 0$$

1

$$\det(A - \lambda I) = 0$$

He likes the previous proof - since talks about eigenvectors as well

5

$$\begin{bmatrix} 0 & 1 & & \\ 1 & 0 & 1 & \\ & 1 & 0 & 1 \\ & & 1 & 0 \end{bmatrix} \text{ Sym + real}$$

$$A^T = A$$

$$\begin{bmatrix} 0 & i & & \\ -i & 0 & i & \\ & -i & 0 & i \\ & & -i & 0 \end{bmatrix}$$

$$\overline{A}^T = A^H = A$$

↑ Hermitian + complex

$$= i \begin{bmatrix} 0 & 1 & & \\ -1 & 0 & 1 & \\ & -1 & 0 & 1 \\ & & -1 & 0 \end{bmatrix}$$

Same eigenvalues

Discussion on MITx

- I suggested Tutor
- Kahn Academy videos
 - more student i more brief concepts
- Student i Diff levels of difficulty

6

40 or 50 x 15 min lectures

Or a 5 min review of key ideas

1 of concept, 1 of solving problem

Prof: Kahn academy guy is friendly

The writing isn't great

Should I get out of the picture?

18.06
Singular Value Decomposition

4/27

(from 2005 MIT OCW - where it was at of order)
Video

~~$A = U \Sigma V^T$~~

$$A = U \Sigma V^T$$

Σ diagonal

U, V orthogonal

"final and best factorization of a matrix"

diagonal \rightarrow only diagonal is non zero

$$\begin{bmatrix} a & 0 & 0 \\ 0 & b & 0 \\ 0 & 0 & c \end{bmatrix}$$

will include sym pos def

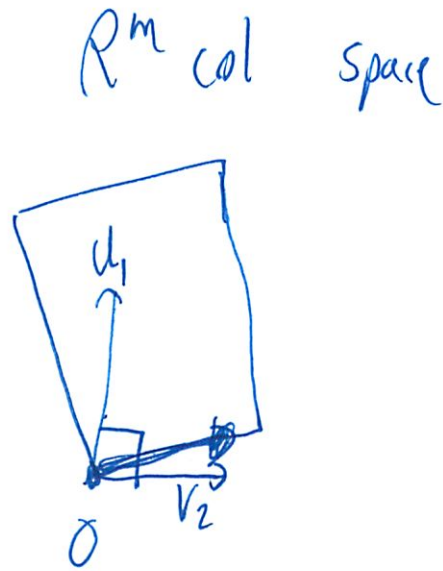
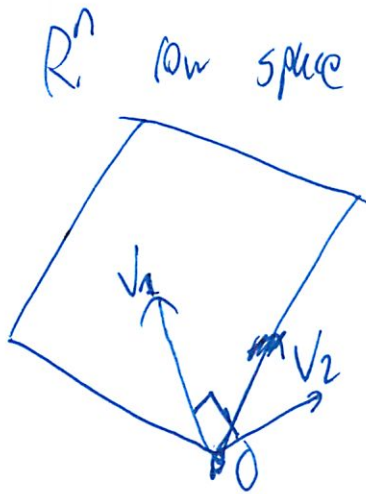
L ~~A~~ are eig vectors are orthogonal

$S \rightarrow Q$
 $\lambda \rightarrow \text{eig}$

Then only 1 U and V
sense

②

But usually not orthogonal



$$u_1 = A v_1$$

Looking for orthogonal basis in row space
that gets mapped into orthogonal basis in col space

Gram Schmidt says how to get orthogonal basis in row space

$$u_1 = v_1$$

$$u_2 = v_2 - \text{proj}(v_2)$$

etc

3

If I take any orth basis and multiply by A
won't be orth on other side

But we want to find those here that are orth on other side

$$A \begin{bmatrix} v_1 & v_2 & \dots & v_r \end{bmatrix} = \begin{bmatrix} u_1 & \dots & u_r \end{bmatrix} \begin{bmatrix} \sigma_1 & & & \\ & \sigma_2 & & \\ & & \dots & \\ & & & \dots \end{bmatrix}$$

↑ basis vector in row space
↑ basis vector col space
↑ multiplicative factor

Goal:

$$\sigma_1 u_1 = A v_1$$

$$\sigma_2 u_2 = A v_2$$

(think of what you want
and express it as a matrix
multiplication)

$$AV = U\Sigma$$

Goal

find orthonormal basis in row and col space

So like diagonalize matrix (for Σ)

take $A = \begin{bmatrix} 4 & 4 \\ -3 & 3 \end{bmatrix}$ invertible
rank 2

← this stuff should be natural!

(9)

want) V_1, V_2 in row space \mathbb{R}^2

U_1, U_2 in col space \mathbb{R}^2

$$\sigma_1 > 0 \quad \sigma_2 > 0$$

words about null space

have a basis for null space

$$V_{(r+1)} \rightarrow V_m$$

take orthonormal basis

$\begin{pmatrix} \sigma_1 \\ \vdots \\ \sigma_r \\ 0 \end{pmatrix}$
(continued)

just get 0 on the diag matrix

$$A \begin{bmatrix} V_1 & V_2 & \dots & V_r & V_{r+n} & V_n \end{bmatrix} = \begin{bmatrix} U_1 & U_2 & \dots & U_r & U_n \end{bmatrix} \begin{bmatrix} \sigma_1 \\ \sigma_2 \\ \vdots \\ \sigma_r \\ 0 \\ \vdots \\ 0 \end{bmatrix}$$

5

We need to get the matrix ~~sym~~ in shape

↳ not sym so cant use e-vectors (not orth)

$$\begin{bmatrix} 4 & 4 \\ -3 & 3 \end{bmatrix}$$

Some how must get the orthonormal V, V_2 U_1, U_2 to get it to work

$$A V_1 = \sigma_1 U_1$$

$$A V_2 = \sigma_2 U_2$$

$AV = U\Sigma$ ← matrices that get us there

Multiply by V^{-1}

$$A = U \Sigma V^{-1}$$

↳ square orthogonal so

$$V^{-1} = V^T$$

$$= U \Sigma V^T$$

⑥

We have 2 orthogonal matrices here

Don't want to use both at once

Want to make the U s disappear \rightarrow only left w/
 V s

So it's the same combo as usual

$A^T A$ \swarrow
- pos def or at least pos semi-def

$$A^T A = V \Sigma^T U^T U \Sigma V^T$$

looks worse - but collapses

$$= V \Sigma^T \underset{\substack{\uparrow \\ \text{diagonal} \\ \uparrow}}{I} \Sigma V^T$$

$$= V \Sigma^2 V^T$$

So this tells us the V s

Now find the U s!

↳ opposite order

②

$$V_s = \text{e vectors} \quad A^T A$$

$$U_s = I \quad A A^T$$

$\Sigma = +$ square roots of $\sqrt{\sigma^2}$ so σ

Step 1, Compute $A^T A$

Want the e vectors

$$A^T A = \begin{bmatrix} 4 & -3 \\ 4 & 3 \end{bmatrix} \begin{bmatrix} 4 & 4 \\ -3 & 3 \end{bmatrix} = \begin{bmatrix} 25 & 7 \\ 7 & 25 \end{bmatrix}$$

So what are the e vectors of this

$$\lambda = 32, 18$$

\leftarrow ~~the~~ $\sqrt{\quad}$ is sign

$$x = \begin{bmatrix} 1 \\ 1 \end{bmatrix}, \begin{bmatrix} 1 \\ -1 \end{bmatrix}$$

Normalize

$$\begin{bmatrix} 1/\sqrt{2} \\ 1/\sqrt{2} \end{bmatrix}, \begin{bmatrix} 1/\sqrt{2} \\ -1/\sqrt{2} \end{bmatrix}$$

8

$$\begin{bmatrix} 4 & 4 \\ -3 & 3 \end{bmatrix} = U \begin{bmatrix} \sqrt{32} & \\ & \sqrt{18} \end{bmatrix} \begin{bmatrix} \frac{1}{\sqrt{2}} & \frac{1}{\sqrt{2}} \\ \frac{1}{\sqrt{2}} & -\frac{1}{\sqrt{2}} \end{bmatrix} V^T$$

?
 don't know yet
 $\sum_{||} \begin{bmatrix} u_1 \\ u_2 \end{bmatrix}$

So could figure out U from pieces
 something w/ inverse?

$$U = \cancel{A \Sigma^{-1} V^T} \\ A \Sigma^{-1} V^T \quad \text{correct}$$

but finding inverse is hard

Find u_1, u_2

$$\begin{aligned} AA^T &= A \Sigma V^T V \Sigma^T U^T \\ &= A \Sigma I \Sigma^T U^T \\ &= A \Sigma^2 U^T \end{aligned}$$

(sym semi def matrix)

9

Its' e vectors go into U

$$A A^T = \begin{bmatrix} 4 & 4 \\ -3 & 3 \end{bmatrix} \begin{bmatrix} 4 & -3 \\ 4 & 3 \end{bmatrix} = \begin{bmatrix} 32 & 0 \\ 0 & 18 \end{bmatrix}$$

↑ lucky
Came out diagonal

$$\lambda = \del{\del{\del{32, 18}}} 32, 18$$

$$x = \begin{bmatrix} 1 \\ 0 \end{bmatrix}, \begin{bmatrix} 0 \\ 1 \end{bmatrix}$$

Not an accident that e-values same
 $AB \neq BA$ same e-values
But e-vectors different

So

$$A = \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix} \begin{bmatrix} \sqrt{32} & \\ & \sqrt{18} \end{bmatrix} \begin{bmatrix} \frac{1}{\sqrt{2}} & \frac{1}{\sqrt{2}} \\ \frac{1}{\sqrt{2}} & -\frac{1}{\sqrt{2}} \end{bmatrix}$$

↑ best orthogonal matrix = I

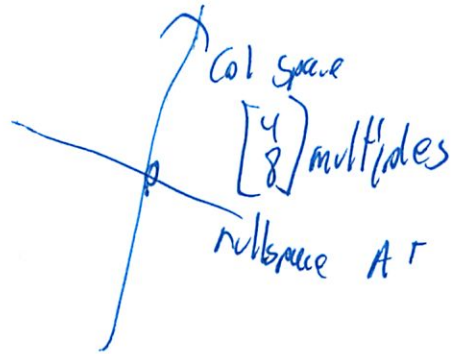
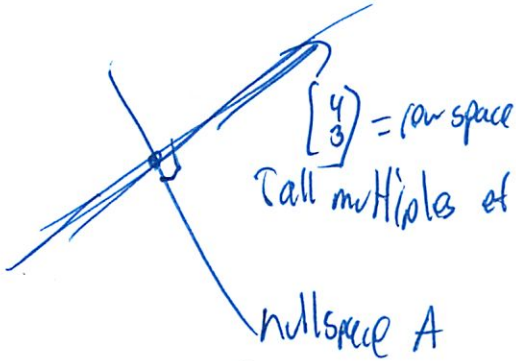
If multiply it back - it should check out ^{except he missed a sign}

10

ex 2

$$\begin{bmatrix} 4 & 3 \\ 8 & 6 \end{bmatrix} \leftarrow \text{rank } 1$$

1D row and col space



So what $\begin{pmatrix} x \\ y \end{pmatrix}$ does it = $\begin{pmatrix} 0 \\ 0 \end{pmatrix}$

Choosing orthogonal bases are no problem
 Make orthonormal

Only one V_1 to choose

$$V_1 = \begin{bmatrix} .8 \\ .6 \end{bmatrix}$$

$$U_1 = \begin{bmatrix} 4/\sqrt{5} \\ 2/\sqrt{5} \end{bmatrix}$$

$$V_2 = \begin{bmatrix} .6 \\ -.8 \end{bmatrix}$$

$$U_2 = \begin{bmatrix} 2/\sqrt{5} \\ -1/\sqrt{5} \end{bmatrix}$$

So

$$\begin{bmatrix} 4 & 3 \\ 8 & 6 \end{bmatrix} = \begin{bmatrix} 1/\sqrt{5} & 2/\sqrt{5} \\ 2/\sqrt{5} & -1/\sqrt{5} \end{bmatrix} \begin{bmatrix} \sqrt{125} & 0 \\ 0 & 0 \end{bmatrix} \begin{bmatrix} .8 & .6 \\ .6 & -.8 \end{bmatrix}$$

$U \quad \Sigma \quad V^T$

$$A^T A = \begin{bmatrix} 4 & 3 \\ 8 & 6 \end{bmatrix} \begin{bmatrix} 4 & 8 \\ 3 & 6 \end{bmatrix} = \begin{bmatrix} 80 & 60 \\ 60 & 45 \end{bmatrix}$$

$\lambda = 0, 125$

④ We are choosing the right basis for the 4 subspaces of linear algebra

V_1, \dots, V_p	is an the an right basis	orthonormal basis for row space
U_1, \dots, U_r	" "	Col space
V_{r+1}, \dots, V_n	" "	nullspace
U_{r+1}, \dots, U_n		Left nullspace

So these are the right bases
make the matrix diagonal

$$A V_i = \sigma_i U_i$$

(Still need to watch last five)
(3 min late)

Orthogonal bases

Fourier series (continuous)

$$F(x) = \sum_0^{\infty} a_k \cos kx + \sum_{k=1}^{\infty} b_k \sin kx$$

Discrete Fourier transform (DFT)

FAST Fourier transform (FFT)

$$= \sum_{k=-\infty}^{\infty} c_k e^{ikx}$$

e^{ikx} (all int k)

are orth funcs on $(0, 2\pi)$

So this lin algebra carries over to diff eq,
(continuous Fourier transform)

②

Orthogonal

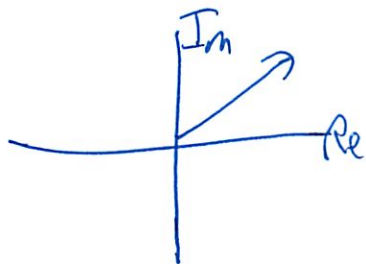
$$\bar{X}^T Y = 0$$

$$X^H Y = 0$$

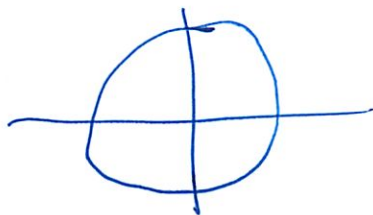
Complex function

Test for orthogonality $X_1^T Y_1 + X_2^T Y_2 + \dots = 0$

$$e^{ikx} = \cos kx + i \sin kx \quad \leftarrow \text{Euler}$$



is on unit circle



~~lim~~

Whole continuum

$$\int_0^{2\pi} e^{ikx} e^{imx} dx$$

$= 0 \rightarrow$ orthogonal

③

But didn't take the complex conjugate of one of the factors

$$\int_0^{2\pi} e^{i(m-k)x} dx = \frac{e^{i(m-k)x}}{i(m-k)} \Big|_0^{2\pi} = 0 \quad \underline{\text{Good}}$$

Hard to visualize function space

But we have a basis

Same pattern as in algebra - but ~~only~~ many functions

Finite Cases

Suppose have

$$F = c_1 v_1 + c_2 v_2 + \dots + c_n v_n$$

Suppose ^{Chap 4 - projections} v_s are orthogonal

c_i : Component in the v_i direction

(4)

$$V_1^T F = C_1 V_1^T V_1$$

$$C_1 = \frac{\overline{V_1^T F}}{\overline{V_1^T V_1}}$$

↑ (complex conj)

What if job is find c_1 ?

Dot both sides w/ 1st function
↳ integral

~~$V_1 = e^{ix}$~~

$$\int_0^{2\pi} e^{-ix} F(x) dx = \int_0^{2\pi} c_1 e^{-ix} e^{ix}$$
$$= 2\pi c_1$$

$$\frac{1}{2\pi} \int_0^{2\pi} e^{-ix} F(x) dx = c_1$$

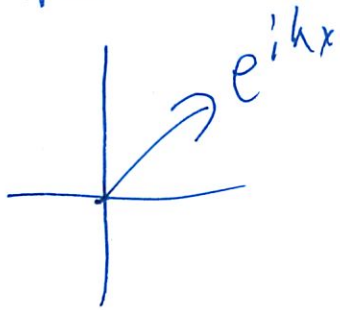
$$C_k = \frac{\int_0^{2\pi} e^{-ikx} F(x) dx}{2\pi}$$

(looks familiar)

Q5

At this point: most classes integration practice

For space



For all matrices \rightarrow can do functions

Like Gram Schmidt

$1, x^2, x^3, \dots$

$$\int_{-1}^1 1 \cdot x \cdot dx = 0$$

area to left balances right = 0

$$\int_{-1}^1 1 \cdot x^2 \cdot dx$$

$\neq 0$ not orthogonal

but a basis

so gram schmidt comes in

⑥ Industries use diff eq
- like oil industry

So have functions

But to compute \rightarrow go to vectors

FFT

Always expand vector F
like e^{ikx} , but vectors

Compute coeffs C_1, \dots, C_n

For all C_s

Old way is n^2 multiplies

FFT $n \log n$ " \in "Holy Grail of algorithms"

$$n = 2^{10} = 1024$$

$$n^2 = 2^{20} \approx 1 \text{ mill}$$

$$n \log_2 n = 1024 \cdot 10 = 10,000 \quad \left\{ \begin{array}{l} 100 \text{ to } 1 \text{ for a} \\ \text{simple math idea} \end{array} \right.$$

(7)

Why are sin, cos so important?

Why is Fourier important?

e^{ikx} is an eigenvector for derivative

$$\frac{d}{dx} e^{ikx} = \underbrace{ik}_{\lambda} \underbrace{e^{ikx}}_{\text{eigenfunction}}$$

Operator

but care at imaginary
So skew sym. matrix

$$\frac{d^2}{dx^2} e^{ikx} = \underbrace{-k^2}_{\lambda} \underbrace{e^{ikx}}_{\text{eigenvector}}$$

Real eigenvalues

Orthogonal eigenvectors

Or switch sign

$$-\frac{d^2}{dx^2} e^{ikx} = k^2 e^{ikx}$$

pos semi def

① More eigenfunctions

Why Fourier important?

↳ since eigenvectors, eigenfunctions of $\frac{d}{dx}$

Can work on any linear eqs

Can't solve non-linear eqns

DFT

basis vectors

$$\begin{matrix} k=0 & 1 & 2 & 3 \\ \begin{bmatrix} 1 & 1 & 1 & 1 \\ 1 & i & i^2 & i^3 \\ 1 & i^2 & i^4 & i^6 \\ 1 & i^3 & i^6 & i^9 \end{bmatrix} \end{matrix}$$

↳ the 4 basis vectors = F_4
Complex
-orthogonal

$$1, e^{ix}, e^{2ix}, e^{3ix}$$

$$\text{at } x = 0, \frac{\pi}{2}, \pi, \frac{3\pi}{2}$$



equal spacing \rightarrow sampling

9

What would be the 6 Fourier vectors in 6 dimensions

$$N=6$$

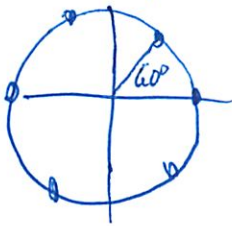
$$F_6 = \begin{pmatrix} 1 & 1 & 1 & 1 & 1 & 1 \\ 1 & w & w^2 & w^3 & w^4 & w^5 \\ 1 & w^2 & w^4 & w^6 & w^8 & w^{10} \\ 1 & w^3 & w^6 & 1 & 1 & 1 \\ 1 & w^4 & w^8 & 1 & 1 & 1 \\ 1 & w^5 & w^{10} & 1 & 1 & 1 \end{pmatrix}$$

← cols are orthogonal

$e^{i\theta}$ ~~sample~~ \in sample

space at

$\leftarrow 60^\circ$



but want in radians $\frac{\pi}{3}$

$$\frac{\pi}{3}, \frac{2\pi}{3}, \frac{3\pi}{3}$$

$$w = e^{i\frac{\pi}{3}}, e^{i\frac{2\pi}{3}}, e^{i\pi}$$

(complex, but nice ~~the~~ pattern

(2 min late)

Exam Review

Sun 1pm

18.06 Recitation

5/1

Pos Def Matrices

When is a sq matrix A pos def
↓
symmetric

A is invertible sym \rightarrow so can diagonalize

1. All pos e-values

2. All pos pivots

3. Pos det on upper left



4. For any vector $x \neq 0$

$$x^T A x > 0$$

5. $A = R^T R$ for some sq, invertible matrix R

$$R = QD$$

(see next)

②

$$A = Q \Lambda Q^T \text{ if } A \text{ sym}$$

Eigenvalues are all real

Various facts

$$\begin{bmatrix} \lambda_1 & & \\ & \ddots & \\ & & \lambda_n \end{bmatrix} = \begin{bmatrix} \sqrt{\lambda_1} & & \\ & \ddots & \\ & & \sqrt{\lambda_n} \end{bmatrix}^2 = D^2$$

$$\begin{aligned} A &= Q D^2 Q^T \\ &= Q D (Q D)^T \end{aligned}$$

ex which are pos def?

a) $\begin{bmatrix} 3 & 4 \\ 4 & 6 \end{bmatrix}$ Yes by criteria 3

b) $\begin{bmatrix} 2 & 1 \\ 3 & 2 \end{bmatrix}$ No not sym

c) $\begin{bmatrix} -1 & 2 & 7 \\ 2 & 3 & 5 \\ 7 & 5 & 2 \end{bmatrix}$ No - by criteria 3
-1 \notin positive

3

$$\begin{bmatrix} 2 & 0 \\ 0 & 7 \end{bmatrix} \text{ yes criteria 1}$$

e values (+)

So small matrix + criteria 3

large matrix do the elim

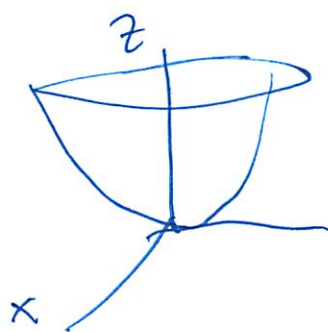
ex Describe graph of $z = ax^2 + bxy + cy^2$

$$\begin{pmatrix} x & y \end{pmatrix} \underbrace{\begin{pmatrix} a & b/2 \\ b/2 & c \end{pmatrix}}_A \begin{pmatrix} x \\ y \end{pmatrix} = ax^2 + bxy + cy^2$$

Why is this useful?

If A is pos def, criteria 4 tells us

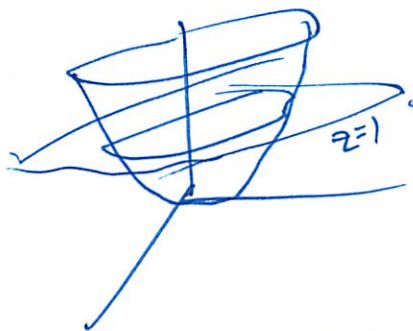
$z \geq 0$ for all $\begin{pmatrix} x \\ y \end{pmatrix} \neq \begin{pmatrix} 0 \\ 0 \end{pmatrix}$



will curve upwards
look like a bowl

(4)

If look at a slice



is a circle

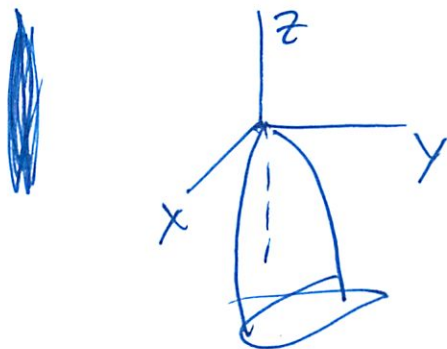
$$ax^2 + bxy + cy^2 = 1$$

If A is pos def then this
is an ellipse

A is neg def if $-A$ is pos def

she thinks this is true

Energy test is pretty clear \rightarrow just reverses inequality



5

Could have mixed e-values - one \oplus
- one \ominus

then A is "indefinite"

Ya get a "saddle"



← worst saddle ever!

If ya horiz slice

↳ get a hyperbole

Jordan Forms

2 matrices A and B are similar if there
exists some invertible matrix M such
that $A = M^{-1} B M$

(we see this all the time
conj operation)

Note: If A, B are similar, they have the same
e-value

But: having same e-value does not guarantee
they are similar

6

If λ -values same \rightarrow e vectors are not necessarily the same!

If $Ax = \lambda x$, eigenvector of B is $M^{-1}x$

$$\begin{aligned} \text{Check } B(M^{-1}x) &= (MA M^{-1})(Mx) \\ &= \underbrace{MAx}_{\lambda x} \end{aligned}$$

Still does not tell us if similar

Put into Jordan Form

Will tell us if similar

A matrix is in Jordan form if we can write

$$A = \begin{bmatrix} J_1 & & \\ & J_2 & 0 \\ & 0 & \ddots \\ & & & J_n \end{bmatrix}$$

$$J_i = \begin{bmatrix} \lambda & 1 & & 0 \\ & \lambda & \ddots & \\ & & \ddots & 1 \\ 0 & & & \lambda \end{bmatrix} \leftarrow \text{Jordan blocks}$$

λ s are eigenvalues of A

7

ex/ $A = \begin{bmatrix} 5 & 4 & 2 & 1 \\ 0 & 1 & -1 & -1 \\ -1 & -1 & 3 & 0 \\ 1 & 1 & -1 & 2 \end{bmatrix}$

What is the Jordan form?

1. Find eigenvalues

$$\det(A - \lambda I) = 0$$

$$(\lambda - 1)(\lambda - 2)(\lambda - 4)^2$$

So we can kinda start Jordan form

Jordan form is 4×4

E values on diagonal

Jordan block

$$\begin{bmatrix} 1 & & & \\ & 2 & & \\ & & 4 & \\ & & & 4 \end{bmatrix}$$

(could be 1 or 2 blocks)

Look at $S^{-1} S$

Only if enough e-vectors
no duplicates

⑧

Since $4, 4$ same e-vector so 1 block since $\text{nullsp}(A - \lambda I) = 0$

In large matrices, same e-value can have diff e-vectors
↳ since $\dim = 2$

$$\begin{bmatrix} 1 & & & \\ & 2 & & \\ & & 4 & 4 \end{bmatrix}$$

We have a mystery matrix A
1 square

$$\det(A - \lambda I) = (\lambda - 2)^3 (\lambda - 3)$$

What are the possible Jordan forms?

We know $\dim = 4$ since 4 e values

$$A = 4 \times 4$$

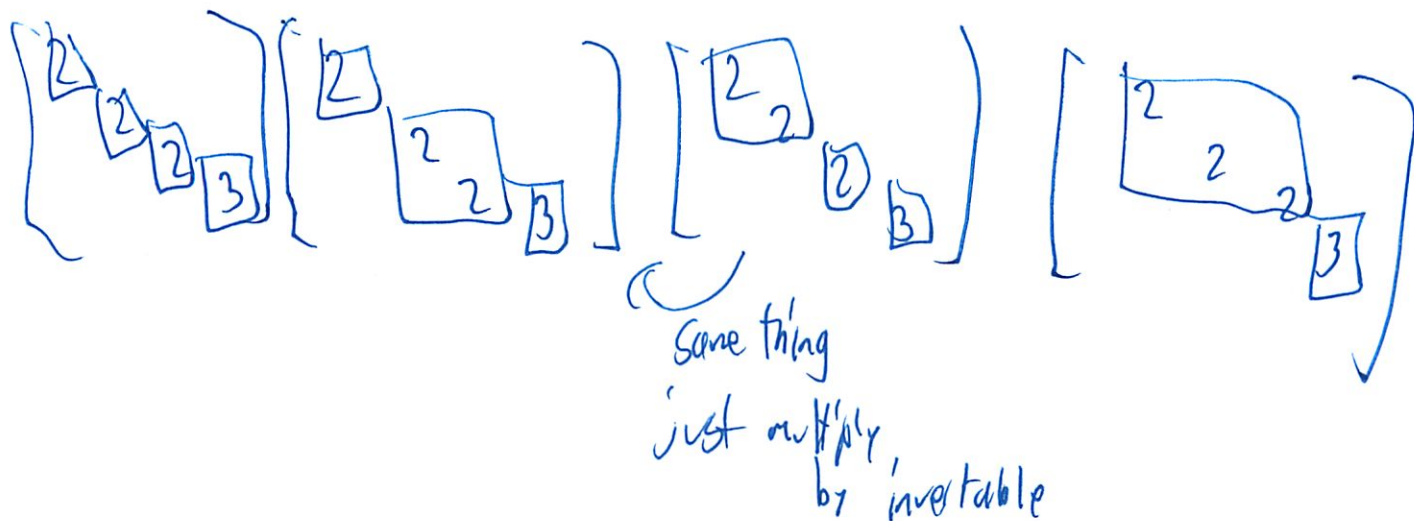
So Jordan form is 4×4

$$\begin{bmatrix} 2 & & & \\ & 2 & & \\ & & 2 & \\ & & & 3 \end{bmatrix}$$

not as useful to write $\begin{bmatrix} 2 & 2 & 2 \\ & 2 & 2 \end{bmatrix}$ but still works

9

What are the Jordan blocks?



Singular Value Decomposition

$A =$ ~~$m \times n$~~ $m \times n$ matrix

We can make square matrices out of A

$A^T A$ or $A A^T$

~~$n \times n$~~
 $n \times n$

~~$m \times m$~~
 $m \times m$

$U_i =$ eigenvectors

$V_i =$ eigenvectors

(10)

$$A v_i = \sigma_i u_i$$

So what effect does this make?

$$A \begin{bmatrix} | & & | \\ v_1 & \dots & v_r \\ | & & | \end{bmatrix} = \begin{bmatrix} | & & | \\ u_1 & \dots & u_r \\ | & & | \end{bmatrix} \begin{bmatrix} \sigma_1 & & \\ & \dots & \\ & & \sigma_r \end{bmatrix}$$

$$r = \text{rank } A$$

We want A in terms of

Then want to move a part

What can't

So complete a matrix

$$\overbrace{A}^{r \times n} \begin{bmatrix} | & & | & & | \\ v_1 & \dots & v_r & v_{r+1} & \dots & v_n \\ | & & | & & | & | \end{bmatrix} = \begin{bmatrix} | & & | & & | \\ u_1 & \dots & u_r & u_{r+1} & \dots & u_m \\ | & & | & & | & | \end{bmatrix} \begin{bmatrix} \sigma_1 & & & & & \\ & \dots & & & & \\ & & \sigma_r & & & \\ & & & 0 & & \\ & & & & \dots & \\ & & & & & 0 \end{bmatrix}$$

$m \times m$ Σ $m \times n$

(11)

$V_{r+1} \dots V_n$ is an orthonormal basis for $N(A)$

$U_{r+1} \dots U_m$ is " " " " $N(A^T)$

Since cols orthonormal \rightarrow can move w/
just multiplying by V^T
(gives identity on LHS)

$$A = U \Sigma V^T$$

Ex) Find SVD of

$$A = \begin{bmatrix} 3 & -1 & 1 \\ 1 & 3 & 1 \end{bmatrix} = \begin{matrix} 2 \times 3 \\ m \ n \end{matrix}$$

1. V is 3×3

U is 2×2

Σ is 2×3

(12)

$$A^T A = \begin{pmatrix} 10 & -6 & 2 \\ -6 & 10 & 2 \\ 2 & 2 & 2 \end{pmatrix}$$

Eigen values = 0, 6, 16

e-vectors = $\begin{bmatrix} 1 \\ 1 \\ -2 \end{bmatrix}, \begin{bmatrix} 1 \\ 1 \\ 1 \end{bmatrix}, \begin{bmatrix} 1 \\ -1 \\ 0 \end{bmatrix}$

These are not v_i
Need to normalize!

$$\frac{1}{\sqrt{6}} \begin{bmatrix} 1 \\ 1 \\ -2 \end{bmatrix}, \frac{1}{\sqrt{3}} \begin{bmatrix} 1 \\ 1 \\ 1 \end{bmatrix}, \frac{1}{\sqrt{2}} \begin{bmatrix} 1 \\ -1 \\ 0 \end{bmatrix}$$

So

$$V = \begin{pmatrix} 1/\sqrt{6} & 1/\sqrt{3} & 1/\sqrt{2} \\ 1/\sqrt{6} & 1/\sqrt{3} & -1/\sqrt{2} \\ -1/\sqrt{6} & 1/\sqrt{3} & 0 \end{pmatrix}$$

③

$$Av_1 = \sigma_1 u_1$$

$|8| = Av_1$ so throw out one
(will always happen somewhere)

$$\frac{1}{\sqrt{3}} \begin{pmatrix} 3 \\ 3 \end{pmatrix} = Av_2 = \sigma_2 u_2$$

$$u_2 = \begin{pmatrix} 1/\sqrt{2} \\ 1/\sqrt{2} \end{pmatrix} \leftarrow \text{length 1 along } \frac{1}{\sqrt{3}} \begin{pmatrix} 3 \\ 3 \end{pmatrix}$$

~~But~~ $3/\sqrt{3}$ is $\sigma_2 \frac{1}{\sqrt{2}}$

$$\text{so } \sigma_2 = \sqrt{6}$$

And finally

$$\frac{1}{\sqrt{2}} \begin{pmatrix} 4 \\ -4 \end{pmatrix} = Av_3 = \sigma_3 u_3$$

$$u_3 = \begin{pmatrix} 1/\sqrt{2} \\ -1/\sqrt{2} \end{pmatrix} \leftarrow \text{unit vector along } \begin{pmatrix} 4 \\ -4 \end{pmatrix}$$

$$\sigma_3 = 4$$

(16)

So can now write the SVD

SVD

$$A = \begin{pmatrix} 1/\sqrt{2} & 1/\sqrt{2} \\ 1/\sqrt{2} & -1/\sqrt{2} \end{pmatrix} \begin{pmatrix} \sqrt{6} & 0 & 0 \\ 0 & 4 & 0 \end{pmatrix} V$$

18.06

5/2

(20 min fire alarm)

Today FFT
Linear Transforms

Fri Exam review

Diagonalization $S\Lambda S^{-1}$

$A^n = e^{At}$

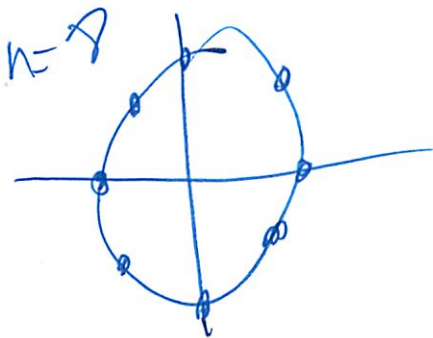
A^{-1}

$A^T = A = Q\Lambda Q^T$

Pos def

Similar

Everything is a power of $\omega = e^{2\pi i/8}$



$$F_8 = \begin{pmatrix} 1 & 1 & \dots & 1 \\ 1 & \omega & & \omega^7 \\ 1 & \omega^2 & & \omega^{14} \\ 1 & \omega^3 & & \vdots \\ 1 & \omega^4 & & \vdots \\ 1 & \omega^5 & & \vdots \\ 1 & \omega^6 & & \vdots \\ 1 & \omega^7 & & \omega^{49} \end{pmatrix}$$

① If add the 8 #s it cancels

$$1 + w + w^2 + \dots + w^7 = 0$$

$$\frac{w^8 - 1}{w - 1}$$

$$w^8 = w^0 = 1$$

~~twice~~

$$w^{(i-1)(k-1)}$$

8x8 is 64 multiplications

FFT cuts this way down

Key to the FFT

The 8x8 connects closely to 2 copies of the 4x4

$$F_8 = \begin{matrix} \text{Simple} \\ \begin{bmatrix} 1 & & & \\ & \ddots & & \\ & & 1 & \\ & & & w^3 \end{bmatrix} \end{matrix} \begin{matrix} \text{2 copies} \\ \begin{bmatrix} F_4 & 0 \\ 0 & F_4 \end{bmatrix} \end{matrix} \begin{matrix} P \\ \begin{bmatrix} 1 & & & \\ & 1 & & \\ & & 1 & \\ & & & 1 \end{bmatrix} \\ \text{permutation} \end{matrix}$$

(3)

The 64 entries has been reduced to 32 non-zero entries

$$P_V = \begin{bmatrix} V_{\text{even}} \\ V_{\text{odd}} \end{bmatrix} = \begin{bmatrix} v_0 \\ v_2 \\ v_4 \\ v_6 \\ \hline v_1 \\ v_3 \\ v_5 \\ v_7 \end{bmatrix}$$

Now recurse on the 4 into 2 and 2

So the Fs go to 1

So $n \log_2 n$ steps n times = $n \log_2 n$

Book has the algebra to make it work

MIT is releasing an even faster FFT
La fast, sparse FFT

4

Number theory people like π
↳ prime #'s stuff

Linear transformation

Used to be on day 1

Matrix secondary for pure linear algebraists

$$y_{\text{out}} = T(x_{\text{in}})$$

Linear functions - very restricted

$$\cancel{T(ax+by)} = cT(x) + dT(y)$$

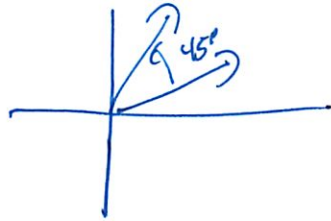
Best example $T(x) = Ax$

$$y = Ax$$

Is this a linear transformation?

⑤
1. $T =$ rotate all vectors by 45°
 (x, y) plane

linear: $T(x) = y$



Yes

~~What~~ what matrix does it?

- called a rotational matrix

$$A = \begin{pmatrix} \cos 45 & \sin 45 \\ -\sin 45 & \cos 45 \end{pmatrix}$$

2. Shift $T(x) = x + x_{\text{fixed}}$

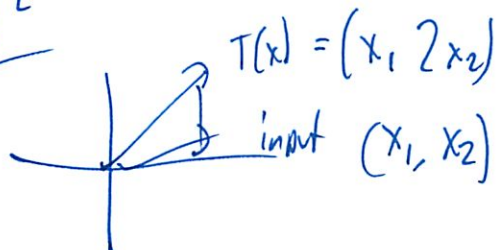
Not linear

$T(0)$ has to come out 0

would kill the shift idea
its always 0

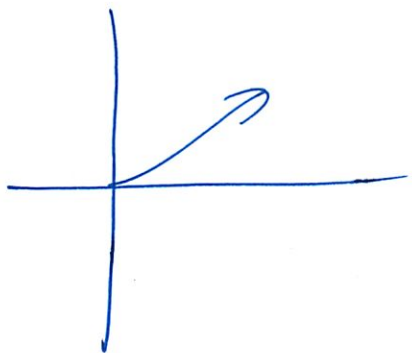
⑥ In linear algebra: ^{linear} means no constant term

3. \mathbb{R}^2



yes $A = \begin{bmatrix} 1 & 0 \\ 0 & 2 \end{bmatrix}$

4. Shear



$$A = \begin{bmatrix} 1 & 5 \\ 0 & 1 \end{bmatrix}$$

Linear transformations come from matrices

$$Y = T(x)$$

\mathbb{R}^m \mathbb{R}^n

is a $m \times n$ matrix that does it

only example

⑦

To find A from T

↳ must choose coords

↳ basically choosing a basis

Vectors you write are a combo of the bases

$$\begin{bmatrix} 3 \\ 4 \end{bmatrix} \text{ is } 3v_1 + 1v_2 + 4v_3$$

↑
basis
vector

$$T(cx + dy) = cT(x) + dT(y)$$

$T \rightarrow A_{m \times n}$ depending on basis

- If choose ^{ind} e vectors as basis

↳ would come out diagonal

"the right basis"

Similar matrices

Review

Wed - Similar

Fri - SVD

M - Fourier, FFT, Complex

ConYean

remember
what day is what

This month in 18.06 went
so fast

5/2

Diagonalization is not possible for every A

↳ if too few e-vectors

S is still best choice

but M is any invertible matrix

So $M^{-1}AM$ - might be diagonal

λ -values same - no matter the M

$$A \underset{\text{similar}}{\approx} M^{-1}AM$$

(made up that symbol)

$$\text{If } B = M^{-1}AM \text{ then } A = MBM^{-1}$$

↳ if A similar to B

B is similar to A

②

If $B = \text{diagonalizable}$ its Λ
and M is S

But again here M is all invertible

* All matrices w/ same e-values are similar to each other *

Jordan Form

(poorly explained in the book)

aka Jordan normal / Jordan Canonical form

- upper triangular matrix representing the operator on some basis

- any ~~non~~ non-zero non-diagonal entries must = 1
be above main diagonal
and be identical diagonal entries on left + below

$$J = \begin{bmatrix} J_1 & & \\ & J_2 & \\ & & \ddots \\ & & & J_n \end{bmatrix} \quad J_i = \begin{bmatrix} \lambda_i & 1 & & \\ & \lambda_i & & \\ & & \ddots & \\ & & & \lambda_i \end{bmatrix}$$

Each Jordan block

3

Jordan form is as close as we can get diagonalizing the matrices when only 1 vector

(this is a special case (should be clear))

<u>Not changed by m</u>	<u>Changed by m</u>
e values	e vectors
trace + det	null space
rank	col space
# ind e vectors	row space
Jordan form	left nullspace
	Singular values

Every similar matrix $B = M^{-1}JM$

J^T is similar to J

(4)

Jordan Form

For every A , choose M so that $M^{-1}AM$ is as diagonal as possible

When A has full set evecors
↳ put into col of M
↳ i.e. $M = S$
 SAS^{-1}

But otherwise \rightarrow only s ind e vectors
↳ s blocks of J
The block handles are evecors of A
Lagoin 1×1 blocks is Λ

This is the big theorem on matrix similarity

For each J solve $\frac{du}{dt} = Ju$

$$M J M^{-1} M J M^{-1} = M J I J M^{-1} = M J^2 M^{-1}$$

5

(So this is like an estimation? - ask in Oth...)

Similar Matrices

Similarity

is transitive

$$A \Leftrightarrow B$$

$$A \Leftrightarrow B \Leftrightarrow C$$



"equivalence relation"

Can start w/ single matrix

Ask what are all the matrices that are similar

Get a whole class of them that are similar



Or can be ~~seen~~ separate



not in common

②

So we can divide into classes

- describe by its elements which are diagonal

$$\begin{pmatrix} 1 & 0 \\ 0 & 2 \end{pmatrix} \text{ same } e\text{-values}$$

- can always switch the order

So diagonal matrix represent similar groups

Jordan Form

if just diagonal matrix \rightarrow not equiv class
not all will be included

by including the Jordan Form blocks, we
include all of them

(3)

6.7 # 3

How would you describe the rank 1

In middle matrix - only one non zero

σ_1 • 1st col u

Now if multiply by v^T

Cols are σ_1^2

①

O/H 2

5/2

Jordan Form Finding

~~It~~ Too theoretical - but can't do in 18.06

- know diagonals

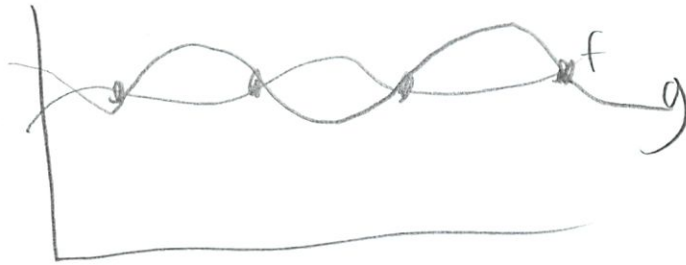
- can guess blocks

- but can't define for sure in 18.06

10.1 - 15 - E is just sym

②

So trying to find a g that meets the same pts



If know fourier matrix, you can solve for c
The coefficients of the unknown

SVD Reading

5/2

Solves 3 things

1. λ vectors are not orthogonal
2. Might not be enough e-vectors
3. Might not be square

* So singular vector fixes this

↳ this of term U, V

Review FFT

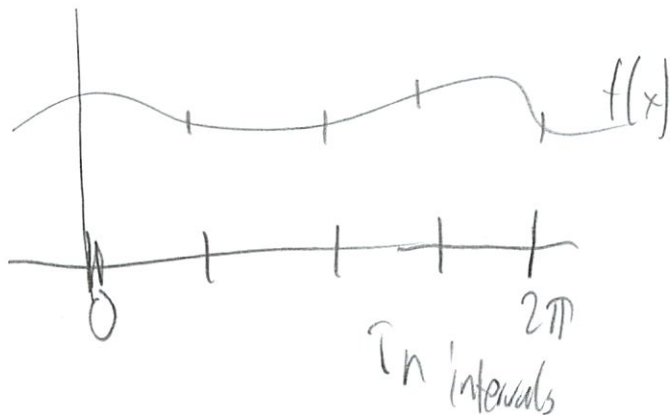
5/2

$F^H = F$ conjugate transpose

Have f_n

Evaluate at intervals

know the value of $f(x)$ at $0, \frac{2\pi}{n}, 2\left(\frac{2\pi}{n}\right), 3\left(\frac{2\pi}{n}\right), \dots, (n-1)\left(\frac{2\pi}{n}\right)$



Look at these n values

Find some series

$$\sum_{k=1}^{n-1} C_k e^{ikx}$$

$$\text{s.t. } f\left(\frac{2\pi}{n}\right) = q\left(\frac{2\pi}{n}\right)$$

Fourier Series

From finite dimensions to infinite
 ∞ many components

1. V could be $(1, \frac{1}{2}, \frac{1}{4}, \dots)$

2. V ~~also~~ could be a function $(\sin(x))$

Fourier series connects them

Dot product of ∞ vectors is ∞ series

Does the series converge?

∞ length vectors \rightarrow no

So only finite length vectors allowed

$$\|V\|^2 = V \cdot V = V_1^2 + \dots \text{ must be finite}$$

$\hat{=}$ called 'in Hilbert space'

②

So $v = (1, \frac{1}{2}, \frac{1}{4}, \dots)$ is included

Since length is

$$v \cdot v = 1 + \frac{1}{4} + \frac{1}{16} + \dots$$

$$= \frac{1}{1 - \frac{1}{4}} = \frac{4}{3}$$

I was never good at this!

Schwarz inequality $|v \cdot w| \leq \|v\| \|w\|$

$$\text{ratio} = \cos \theta$$

functions

$f(x)$ defined for $0 \leq x \leq 2\pi$

$$(f, g) = \int_0^{2\pi} f(x) g(x) dx$$

$$\|f\|^2 = \int_0^{2\pi} (f(x))^2 dx$$

So $f(x) = \sin x$ comes w/ inner product of itself

$$\int_0^{2\pi} (\sin x)^2 dx = \pi$$

$\sin x, \cos x$ are orthogonal

(3)

Fourier Series

$$y(x) = a_0 + a_1 \cos x + b_1 \sin x + a_2 \cos 2x + b_2 \sin 2x + \dots$$

Orthogonal basis

periodic $\rightarrow 2\pi$

so just look $0 \rightarrow 2\pi$
even though ∞ series

δ is a small spike 

if divide by length \rightarrow orthonormal

(this is very familiar from last semester

↳ do I remember it though...)

④

Fourier Coefficients

$$f(x) = a + a_1 \cos x + b_1 \sin x + a_2 \cos 2x + \dots$$

$$\text{Coeff } a_1 \int_0^{2\pi} f(x) \cos x dx =$$

$$\int_0^{2\pi} a_1 \cos^2 x dx = \pi a_1$$

$$a_k = \frac{1}{\pi} \int_0^{2\pi} f(x) \cos kx dx$$

$$b_k = \frac{1}{\pi} \int_0^{2\pi} f(x) \sin kx dx$$

Linear algebra

∞ dim similar to n dim case

$$b = c_1 v_1 + c_2 v_2 + \dots + c_n v_n$$

Multiply both sides by v_1^T

Use orthogonality so $v_1^T v_2 = 0$

$$c_1 \text{ term } v_1^T b = c_1 v_1^T v_1 + 0 + \dots + 0$$

5

$$\text{So } c_1 = \frac{V_1^T b}{V_1^T V_1}$$

* Coeffs are easy to find for orthogonal basis

Eq for c_s

$$c_1 V_1 + \dots + c_n V_n = b$$

$$\text{or } \begin{bmatrix} V_1 & \dots & V_n \end{bmatrix} \begin{bmatrix} c_1 \\ \vdots \\ c_n \end{bmatrix} = b$$

Reading
10.3
FFT

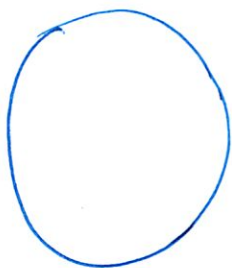
5/2

Multiply quickly by F F^{-1}
2 Fourier

Represent f as a sum of harmonics
 $c_k e^{ikh}$

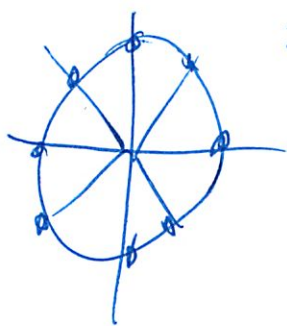
Look at through freq space as coeffs c_k
instead of physical space $f(x)$

Roots of unity



$$z^n = 1$$

Sols z are the n th roots of unity



$$z^8 = 1$$

$$\frac{1}{8}(360) = 45$$

2

Complex # $w = e^{i\theta} = e^{j2\pi/8}$
still confuses me somewhat...

Fourier matrix

$n=4$

$$F = \begin{bmatrix} 1 & 1 & 1 & 1 \\ 1 & w & w^2 & w^3 \\ 1 & w^2 & w^4 & w^6 \\ 1 & w^3 & w^6 & w^9 \end{bmatrix}$$

Annotations: A vertical arrow labeled '1s' points to the first column. A horizontal arrow labeled '1s' points to the first row. A curved arrow labeled 'add 1' points from the first row to the second row. A bracket on the right side of the matrix is labeled 'double'.

Symmetric not hermitian

Unitary $\left(\frac{1}{2} F^H\right) \left(\frac{1}{2} F\right) = I$

Cols of F give $F^H F = 4I$

inverse is $\frac{1}{4} F^H$ which is $F^{-1} = \frac{1}{4} \overline{F}$

3

Then

$$\begin{bmatrix} Y_0 \\ Y_1 \\ Y_2 \\ Y_3 \end{bmatrix} = F_C = \begin{bmatrix} 1 & 1 & 1 & 1 \\ 1 & W & W^2 & W^3 \\ 1 & W^2 & W^4 & W^6 \\ 1 & W^3 & W^6 & W^9 \end{bmatrix} \begin{bmatrix} C_0 \\ C_1 \\ C_2 \\ C_3 \end{bmatrix}$$

↑ outputs ↑ inputs

$$Y_0 = C_0 + C_1 + C_2 + C_3$$

$$Y_1 = C_0 + C_1 e^{i2\pi/4} + C_2 e^{i4\pi/4} + C_3 e^{i6\pi/4}$$

$$= C_0 + C_1 W + C_2 W^2 + C_3 W^3$$

↑ raise notation

finite fourier series

4 terms, evaluated at 4 points

FFT Part 1

Matrix multiplication goes faster w/ zeros

Connect F_n w/ the half sized $F_{n/2}$

Or two copies of this

(4)

$$F_4 = \begin{bmatrix} F_2 & \\ & F_2 \end{bmatrix} \neq \begin{bmatrix} 1 & 1 & 0 & 0 \\ 1 & i^2 & 0 & 0 \\ 0 & 0 & 1 & 1 \\ 0 & 0 & 1 & i^2 \end{bmatrix}$$

Need some factors

$$F_4 = \begin{bmatrix} 1 & & & \\ & 1 & & \\ & & i & \\ & & & -i \end{bmatrix} \begin{bmatrix} 1 & & & \\ & 1 & & \\ & & 1 & \\ & & & i^2 \end{bmatrix} \begin{bmatrix} 1 & & & \\ & 1 & & \\ & & 1 & \\ & & & 1 \end{bmatrix}$$

(random...)

$$F_{1024} = \begin{bmatrix} I_{512} & D_{512} \\ I_{512} & -D_{512} \end{bmatrix} \begin{bmatrix} F_{512} & \\ & F_{512} \end{bmatrix} \begin{bmatrix} \text{even} & \text{odd} \\ \text{permutation} & \end{bmatrix}$$

(identity) (diagonal)

Full FFT by Recursion

keep going to $F_{n/4}$
etc (recursion)

18.06 Spring 2012 – Problem Set 9

This problem set is due Thursday, May 3rd, 2012 at 4pm (hand in to Room 2-106). The textbook problems are out of the 4th edition. For computational problems, please include a printout of the code with the problem set (for MATLAB in particular, `diary('filename')` will start a transcript session, `diary off` will end one.)

Every problem is worth 10 points.

1. Do Problems 5 & 11 from Section 6.6.
2. Do Problems 17 & 19 from Section 6.6.
3. Do Problem 22 from Section 6.6.
4. Do Problems 3 & 6 from Section 6.7.
5. Do Problems 9 & 10 from Section 6.7.
6. Do Problem 14 from Section 6.7.
7. Do Problems 1 & 4 from Section 8.5.
8. Do Problems 3 & 19 from Section 10.2.
9. Do Problem 7 from Section 10.3.
10. Do Problem 15 from Section 10.3.

18.06 Wisdom. To reiterate: You should get as ready as you can for Exam 3, and for the finals, by doing as many old exams as you have time for (found on the 18.06 website under "Past Courses"). Get into a good rhythm with these rehearsals to stay fully tuned in on Linear Algebra, and to master the concepts - you will need it, both here and beyond.

Q.6 #5 Similar Matrices

10/10

Which of the 6 are similar

↳ same e-value in any order

Find λ in Matlab for speed + correctness

$$\begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix} \quad \lambda = 1, 1$$

$$\begin{bmatrix} 0 & 1 \\ 1 & 0 \end{bmatrix} \quad \lambda = -1, 1$$

$$\begin{bmatrix} 1 & 1 \\ 0 & 0 \end{bmatrix} \quad \lambda = 1, 0$$

$$\begin{bmatrix} 0 & 0 \\ 1 & 1 \end{bmatrix} \quad \lambda = 1, 0$$

$$\begin{bmatrix} 1 & 0 \\ 1 & 0 \end{bmatrix} \quad \lambda = 0, 1$$

$$\begin{bmatrix} 0 & 1 \\ 0 & 1 \end{bmatrix} \quad \lambda = 0, 1$$

} similar

} similar

} similar order
does not matter

Prob also some pattern I forgot

②

6.6 #11 Solve $\frac{du}{dt} = Ju$ for J in #10

$$J = \begin{pmatrix} \lambda & 1 \\ 0 & \lambda \end{pmatrix}$$

Starting from $u(0) = (5, 2)$

Remember $t e^{\lambda t}$

About Jordan Form ...

So J is as close to diagonal as can get
Can't solve from changing variables

$$\frac{du}{dt} = Ju \quad \text{is} \quad \frac{dx}{dt} = \lambda x + y$$

$$\frac{dy}{dt} = \lambda y$$

Then work backward - triangular

$$y = y(0) e^{\lambda t}$$

$$x = (x(0) + t y(0)) e^{\lambda t}$$

So I see the pattern \rightarrow by why?

(2b)

Then fill in start point

$$y = 2e^{2x}$$

$$x = (5 + 2t)e^{2x}$$

is that it is
↓ this form

$$u = \begin{bmatrix} 5 \\ 2 \end{bmatrix} e^{2x} + \begin{bmatrix} 2 \\ 0 \end{bmatrix} t e^{2x}$$

3)

6.6 #17 True or False

10/10 a) A sym matrix can't be similar to a non sym matrix

I think need full set of evalues for sym matrix - so can't be similar

Books Diagonalizing a non sym $\Rightarrow A = SAS^{-1}$ True

Then A is non sym and similar **False**
(confused by what is what when its symmetric)

So can diag w/out sym - but if sym its $Q \Lambda Q^T$ - and can handle repeated evalues

b) An invertible matrix can't be similar to a singular matrix.

True. Singular means one of the eigenvalues is 0

No e-value can be 0 in an invertible

So can't be similar \checkmark True

4

c) A can't be similar to $-A$ unless $A=0$

'Is this something about pos or neg definite?'

Eigenvalues should be $-$, would not be the

same true

Book

$\begin{bmatrix} 0 & 1 \\ -1 & 0 \end{bmatrix}, \begin{bmatrix} 0 & -1 \\ 1 & 0 \end{bmatrix}$ are similar so False

↳ So only needed to find 1 counter example...

d) A can't be similar to $A+I$

So what are λ of I

↳ 1, 1, 1, ...

But how do λ change when add

Add cellwise - so everything one more

So will be different

True

5

Q19 If A is 6×4 and B is 4×6
 AB and BA have diff sizes
but w/ blocks

$$M^{-1}FM = \begin{bmatrix} I & -A \\ 0 & I \end{bmatrix} \begin{bmatrix} AB & 0 \\ A & 0 \end{bmatrix} \begin{bmatrix} I & A \\ 0 & I \end{bmatrix}$$
$$= \begin{bmatrix} 0 & 0 \\ B & BA \end{bmatrix} = 6$$

a) What are the size of the 4 blocks
zeros?
What blocks \rightarrow Jordan blocks

Note long
divergence

$$\text{So } M^{-1}AM = \begin{bmatrix} J_1 & & \\ & \ddots & \\ & & J_s \end{bmatrix}$$

Jordan block

One Jordan block per unique λ -value/e-value

So do we know which blocks are here?
kinda seen in recitation

6

Key: Note same λ does not mean 'similar'
Fact

We don't have specific #s

Otherwise could draw blocks

Then look at $S D S^{-1}$

↳ I don't get how this is working w/ splitting up into blocks

If same e-vector λ

Back to qV

Still no clue what its asking

Block Diagonal blocks 6×6
 4×6

AB has same E values as BA plus $6-4$ zeros

(6)

What is F ? - just letter given to something

OK)

$$M = \begin{bmatrix} I & A \\ 0 & I \end{bmatrix}$$

$$F = \begin{bmatrix} AB & 0 \\ B & 0 \end{bmatrix}$$

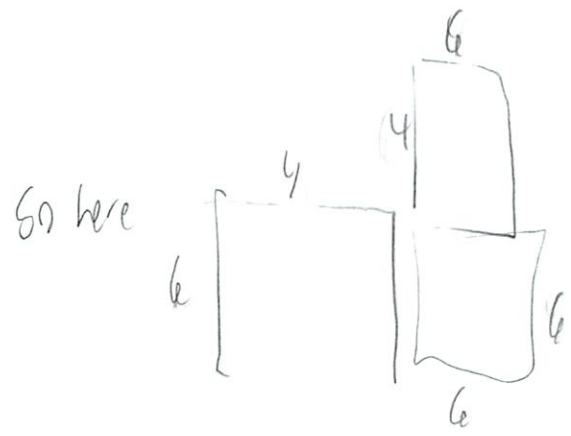
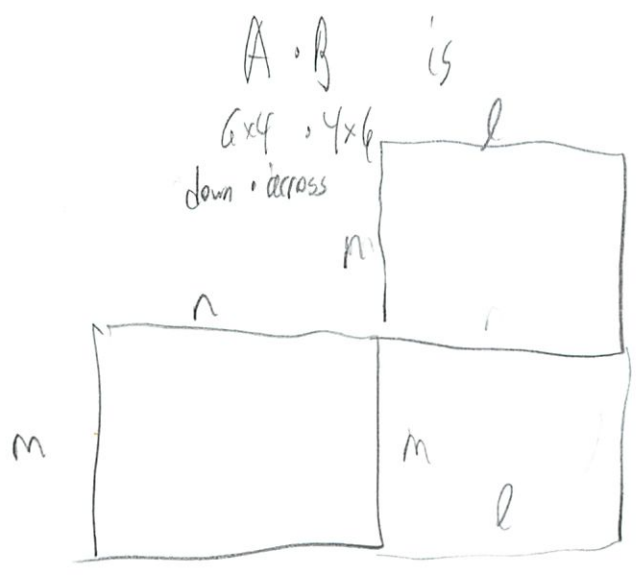
O is not just 0

L is a matrix O

So asks what is size of each

So upper right is 6×4

lower left is 4×6



60

6

6	6	4
6	6	4
4	6	4

7

b) This eq is $MF = G$ so F, G have same 10

e-values

Since similar

F has 6 evalues of AB plus 4 zeros

G has 4 evalues of BA plus 6 zeros

AB has same evalues as BA plus — 0s

So AB is size 6×6

BA is 4×4

So $6 - 4 = 2$ (D)

Qib #22 If an $n \times n$ matrix A has all eigenvalues
10/10 $\lambda = 0$ prove that $A^n = \text{zero matrix}$,
(Maybe prove that $J^n = \text{zero matrix}$, by direct manipulation,
or use Cayley Hamilton theory)

$$\text{So } A = \begin{bmatrix} 0 & 0 & 0 \\ 0 & 0 & 0 \\ 0 & 0 & 0 \end{bmatrix}$$

$$\lambda = 0, 0, 0$$

We want $B = M^{-1}AM$ to get B
which is similar to A
which is nearly as diag as possible

↳ Similar is same evals

9

$$M^{-1}AM = \begin{bmatrix} J_1 & & \\ & \ddots & \\ & & J_s \end{bmatrix} = J$$

$$^? \begin{bmatrix} \sigma & 1 & 0 \\ 0 & \sigma & 0 \\ & & \ddots \end{bmatrix} \text{ repeated } = 1 \text{ block}$$

$$J^n = \begin{bmatrix} \text{not just} & 0^n & 1^n & 1^n \\ & 0^n & & 0^n \\ & & & \end{bmatrix}$$

but it shows its the same

And since $J = B$ and is similar to A

A^n is the same

[? seems too easy ...

Book $A = MJM^{-1}$

$$A^n = MJ^nM^{-1} = 0$$

Each J^k has 1s on the k th diagonal

$\det(A - \lambda I) = \lambda^n$ so $J^n = 0$ by C-H Theorem

kinda what I did - but I don't get

10

hw 7 # 3

10/10

For this $A = \begin{bmatrix} 1 & 2 \\ 3 & 6 \end{bmatrix}$ and all rank 1 matrices, why is σ_1^2 the sum of all a_{ij}^2 ?

So SVD is

$$A = U \Sigma V^T = u_1 \sigma_1 v_1^T + \dots + u_r \sigma_r v_r^T$$

A is diagonalized

$$A v_1 = \sigma_1 u_1$$

$$A v_2 = \sigma_2 u_2$$

OH] $A = \begin{bmatrix} u_1 & u_2 \\ 1 & 1 \end{bmatrix} \begin{bmatrix} \sigma_1 & 0 \\ 0 & 0 \end{bmatrix} \begin{bmatrix} -v_1 \\ -v_2 \end{bmatrix}$

u, v not orthonormal
when multiply get 1
can't just multiply

Multiply both sides by A^T

$$A A^T = \begin{bmatrix} u_1 & u_2 \\ 1 & 1 \end{bmatrix} \begin{bmatrix} \sigma_1 & 0 \\ 0 & 0 \end{bmatrix} \underbrace{\begin{bmatrix} -v_1 \\ -v_2 \end{bmatrix} A^T}_{A \begin{pmatrix} v_1 \\ v_2 \end{pmatrix}^T}$$

Since $A^T B^T = (BA)^T$

(1)

$$\left(\begin{bmatrix} \sigma & 0 \\ 0 & 0 \end{bmatrix} \begin{bmatrix} | & | \\ u_1 & u_2 \\ | & | \end{bmatrix} \right)^T$$

the important line
By def SVD
 $\sigma_1 u_1 = A v_1$
 $\sigma_2 u_2 = A v_2$

$$= \begin{bmatrix} | & | \\ u_1 & u_2 \\ | & | \end{bmatrix} \begin{bmatrix} \sigma_1^2 & 0 \\ 0 & 0 \end{bmatrix} \begin{bmatrix} | & | \\ u_1 & u_2 \\ | & | \end{bmatrix}^T$$

$(\sigma_1 \ \sigma_2) \begin{bmatrix} | & | \\ u_1 & u_2 \\ | & | \end{bmatrix} = A \begin{bmatrix} | & | \\ v_1 & v_2 \\ | & | \end{bmatrix}$
vectors by convention are long

orthonormal so $Q^T = Q^{-1}$
 $AA^{-1} = I$

$$= \begin{bmatrix} \sigma_1^2 & 0 \\ 0 & 0 \end{bmatrix}$$

radically vanishes

Since

$$\begin{bmatrix} a_{11} & a_{12} \\ a_{21} & a_{22} \end{bmatrix} \begin{bmatrix} a_{11} & a_{12} \\ a_{21} & a_{22} \end{bmatrix}$$

$$= \begin{bmatrix} a_{11}^2 + a_{12}^2 & * \\ * & a_{21}^2 + a_{22}^2 \end{bmatrix}$$

$$= \begin{bmatrix} \sigma_1^2 & 0 \\ 0 & 0 \end{bmatrix} \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix}$$

$$= \begin{bmatrix} \sigma_1^2 & 0 \\ 0 & 0 \end{bmatrix}$$

Can just add

(TA confused) !!

12

6.7 #6 Compute $A^T A$ and AA^T for this A

$$A = \begin{bmatrix} 1 & 1 & 0 \\ 0 & 1 & 1 \end{bmatrix}$$

Multiply the matrices $U \Sigma V^T$ to recover A
 Σ has same shape as A .

Step 1. Compute $A^T A$

$$\begin{bmatrix} 1 & 0 \\ 1 & 1 \\ 0 & 1 \end{bmatrix} \begin{bmatrix} 1 & 1 & 0 \\ 0 & 1 & 1 \end{bmatrix} = \begin{bmatrix} 1 & 1 & 0 \\ 1 & 2 & 1 \\ 0 & 1 & 1 \end{bmatrix}$$

2. Find $\lambda = 0, 1, 3$

3. Find eig vectors = $\begin{bmatrix} 1/\sqrt{3} \\ -1/\sqrt{3} \\ 1/\sqrt{3} \end{bmatrix} \begin{bmatrix} 1/\sqrt{2} \\ 0 \\ 1/\sqrt{2} \end{bmatrix} \begin{bmatrix} 1/\sqrt{6} \\ \sqrt{2}/3 \\ 1/\sqrt{6} \end{bmatrix}$

4. Fill in what we know

$$\begin{bmatrix} 1 & 1 & 0 \\ 0 & 1 & 1 \end{bmatrix} = \bar{U} \Sigma V^T = \begin{bmatrix} \sqrt{3} & \sqrt{1} & \sqrt{3} \end{bmatrix} \begin{bmatrix} 1/\sqrt{3} & -1/\sqrt{2} & 1/\sqrt{6} \\ -1/\sqrt{3} & 0 & \sqrt{2}/3 \\ 1/\sqrt{3} & 1/\sqrt{2} & 1/\sqrt{6} \end{bmatrix}$$

13

5. Calculate AA^T

$$\begin{bmatrix} 1 & 1 & 0 \\ 0 & 1 & 1 \end{bmatrix} \begin{bmatrix} 1 & 0 \\ 1 & 1 \\ 0 & 1 \end{bmatrix} = \begin{bmatrix} 2 & 1 \\ 1 & 2 \end{bmatrix}$$

6. $\lambda = 1, 3$ will be the same, but dim diff ...

$$7. X = \begin{bmatrix} 1/\sqrt{2} & 1/\sqrt{2} \\ 1/\sqrt{2} & 1/\sqrt{2} \end{bmatrix}$$

8. Fill it in

$$A = \begin{bmatrix} -1/\sqrt{2} & 1/\sqrt{2} \\ 1/\sqrt{2} & 1/\sqrt{2} \end{bmatrix} \begin{bmatrix} \sqrt{0} & \sqrt{1} \\ & \sqrt{3} \end{bmatrix} \begin{pmatrix} 1/\sqrt{3} & -1/\sqrt{2} & 1/\sqrt{6} \\ -1/\sqrt{3} & 0 & \sqrt{2}/3 \\ 1/\sqrt{3} & 1/\sqrt{2} & 1/\sqrt{6} \end{pmatrix}^T$$

dim error

hmm - how to avoid

$$\begin{aligned} \dim & (x \times y) \cdot (y \cdot z) \\ & \quad \quad \quad \underbrace{\hspace{2cm}}_{\text{same}} \\ & = (x \cdot z) \end{aligned}$$

(14)

So elim the 0 eig vector

$$A = \begin{bmatrix} -1/\sqrt{2} & \sqrt{3}/2 \\ 1/\sqrt{2} & \sqrt{3}/2 \end{bmatrix} \begin{bmatrix} V^T \end{bmatrix}$$

but now 2x2 3x3 ...
 Will the $\lambda=0$ col?

$$V^T = \begin{bmatrix} -1/\sqrt{2} & 0 & 1/\sqrt{2} \\ 1/\sqrt{6} & \sqrt{2}/3 & 1/\sqrt{6} \end{bmatrix}$$

$$A = \begin{bmatrix} 1 & 1 & 0 \\ 0 & 1 & 1 \end{bmatrix} \text{ worked at}$$

basically don't include $\lambda=0$, right?

$$\Sigma \text{ I got was } \begin{bmatrix} \sqrt{1} & & \\ & \sqrt{3} & \\ & & \end{bmatrix}$$

Why is it the same shape as A ?

$$\text{Or } \begin{bmatrix} 0 & \sqrt{1} & 0 \\ 0 & 0 & \sqrt{3} \end{bmatrix}$$

There were no non-square examples in the book
 ↓ diff from example I found online

(15)

Oh add 0s at the end

$$\begin{bmatrix} \sigma_1 & 0 & 0 \\ 0 & \sigma_3 & 0 \end{bmatrix}$$

Example on WP

$$A = \begin{bmatrix} 4 & 5 \end{bmatrix}$$

$$U = \begin{bmatrix} 4 & 4 \\ 4 & 4 \end{bmatrix} \quad \Sigma = \begin{bmatrix} 4 & 5 \end{bmatrix} \quad V^T = \begin{bmatrix} 5 & 5 \end{bmatrix}$$

So that's how it is supposed to work out

Order matters!

(15)

6.7 #9

10/10

Suppose u_1, \dots, u_n and v_1, \dots, v_n are orthonormal bases for \mathbb{R}^n

Construct the matrix A that transforms each v_j into u_j to give

$$A v_1 = u_1$$

\vdots

$$A v_n = u_n$$

So u, v are orthonormal bases

normally

$$\left[\begin{array}{l} A v_1 = \sigma_1 u_1 \\ A v_2 = \sigma_2 u_2 \\ \vdots \\ A v_r = \sigma_r u_r \end{array} \right.$$

So here A is something that involves σ and real A

$$A = \text{real } A \cdot \sigma^{-1}$$

(6)

Book $A = UV^T$ since all $\sigma = 1$

$$\text{So } \Sigma = I$$

So why $\sigma = 1$

Oh the orthonormal is not always there?

Orthogonal gives

$$V^T V = I$$
$$U^T U = I$$

I kinda had it

Still not good w/ all these rules...

Fixed

A is usually $U \Sigma V^T$

in this case Σ is identity

Since all multipliers are given as 1

I didn't realize this was given and part of problem

(17)

↳ What's so special about rank 2?

Q. 7 # 10 Construct a matrix w/ rank 2 that has

$$Av = 12u$$

$$\text{for } v = \frac{1}{2}(1, 1, 1, 1) \text{ and}$$

$$u = \frac{1}{3}(2, 2, 1)$$

Only 1 singular

So find A

$$A \begin{bmatrix} \frac{1}{2} & \frac{1}{2} & \frac{1}{2} & \frac{1}{2} \\ & & & \end{bmatrix} = 12 \begin{bmatrix} \frac{2}{3} & \frac{2}{3} & \frac{1}{3} \\ & & \end{bmatrix}$$

σ_1

$$\begin{bmatrix} 24 \\ 0 \\ 0 \\ 0 \end{bmatrix} \begin{matrix} \frac{1}{2} = 12 \cdot \frac{2}{3} \\ \frac{1}{2} = 0 \cdot \frac{2}{3} \end{matrix}$$

↳ is that the right approach?

or
are multiple

$$\begin{bmatrix} 24 \\ 12 \\ 6 \\ 3 \end{bmatrix}$$

This feels weirdly wrong...

(76)

Friend \searrow Similar to 9

$$A = \begin{pmatrix} 2/3 \\ 2/3 \\ 1/3 \end{pmatrix} \begin{pmatrix} 12 \end{pmatrix} \begin{pmatrix} 1/2 & 1/2 & 1/2 & 1/2 \end{pmatrix} =$$

must deal w/ the 12

$$= \begin{pmatrix} 4 & 4 & 4 & 4 \\ 4 & 4 & 4 & 4 \\ 2 & 2 & 2 & 2 \end{pmatrix}$$

(18)

6.7 #14 Suppose A is invertible w/ $\sigma_1 > \sigma_2 > 0$

10%

Change A by as small a matrix as possible to produce a singular matrix A_0

Hint: U and V don't change

$$\text{From } A = [u_1 \ u_2] \begin{bmatrix} \sigma_1 & \\ & \sigma_2 \end{bmatrix} [v_1 \ v_2]^T$$

So make A from invertible to singular

Can't change U or V

So have to change σ_s

So change one of them

$$\begin{bmatrix} 1 & \\ & 2 \end{bmatrix} \begin{bmatrix} 2 & \\ & 2 \end{bmatrix} = \begin{bmatrix} 1 & \\ & \sigma_1 \end{bmatrix} \begin{bmatrix} v_1 \\ v_2 \end{bmatrix} = \begin{bmatrix} u_1 \sigma_1 & u_2 \sigma_2 \end{bmatrix}$$

So ~~the~~ change either $u_1 \sigma_1 v_1 + u_2 \sigma_2 v_2 = \text{final}$

(14)

Book | Change σ_2

(Why is σ_2 better than σ_1 ?)

Friend

Since were told $\sigma_1 > \sigma_2$

Not changing matrix by as much

(stupid q)

(20)

Fourier series

8.5 #1 Integrate

10/10

$$2 \cos jx \cos kx = \cos(j+k)x + \cos(j-k)x$$

to show $\cos jx \perp \cos kx$

Book what is this trying to show

$$\begin{aligned} \int_0^{2\pi} \cos((j+k)x) dx &= \\ &= \left[\frac{\sin((j+k)x)}{j+k} \right]_0^{2\pi} \\ &= 0 \end{aligned}$$

$$\int_0^{2\pi} \cos((j-k)x) dx = 0$$

Notice $j-k \neq 0$ in denom

$$\text{If } j=k \rightarrow \int_0^{2\pi} \cos^2 jx dx = \pi$$

↑ resonance

(216)

Find)

Given identity

Integrate it

See 0

Since have 2 sin functions added
together

eval at $0, 2\pi$ so will be 0

b) Should be factor of 2

But get π

$\int 2 \cos^2 jx \rightarrow$ comes out 2π

(21)

8.5 #4 1st 3 Lagrange polynomials are $1, x, x^2 - \frac{1}{3}$

Choose c so $x^3 - cx$ is orthogonal

$$\int_{-1}^1 (x) (x^3 - cx) dx =$$

$$= \int_{-1}^1 (x^4 - cx^2) dx$$

$$= \left. \frac{1}{5} x^5 - \frac{c}{3} x^3 \right|_{-1}^1$$

$$= 2 \left(\frac{1}{5} - \frac{c}{3} \right) = 0$$

$$\frac{1}{5} = \frac{c}{3}$$

$$c = 3/5$$

Odd fun is just 0

Look for even function

That's the even fun

28

10, 2 # 3 Solve $Az=0$ to find a vector
in nullspace of A in \mathbb{H}^2

10/10

$$A = \begin{bmatrix} i & 1 & i \\ 1 & i & \bar{i} \end{bmatrix}$$

Show z is \perp to cols of A^H

$$\begin{bmatrix} i & 1 & i \\ 1 & i & \bar{i} \end{bmatrix} z = \begin{bmatrix} 0 \\ 0 \\ 0 \end{bmatrix}$$

$$iz_1 + z_2 + iz_3 = 0$$

$$z_1 + iz_2 + \bar{i}z_3 = 0$$

$$\text{So } z_3 = -i$$

$$z_1 = -1$$

$$z_2 = i$$

$$-i + i + i = -i$$

$$-1 + -1 + i = -2 + i$$

Not \perp - is anything possible?

23

Book 1

Ah if you do mix

P should have solved more robustly

$$z = \begin{bmatrix} 1+i \\ 1+i \\ -2 \end{bmatrix}$$

$$i(1+i) + (1+i) + i(-2) = 0$$

$$i-1+1+i-2i = 0 \quad \ominus$$

$$(i+1) + i(1+i) + i(-2) = 0$$

$$i+1+i-1-2i = 0 \quad \ominus$$

Show z is \perp to col of A^H

$$\text{So } A = \begin{bmatrix} i & 1 & i \\ 1 & i & i \end{bmatrix} \quad z = \begin{bmatrix} 1+i \\ 1+i \\ -2 \end{bmatrix}$$

Show orthogonal by dot product is 0

Just basically did that

(28)

Show z is not \perp to cols of A^+

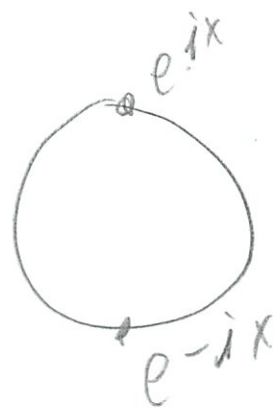
$$A^+ = \begin{bmatrix} i & 1 \\ 1 & i \\ i & i \end{bmatrix} \quad z = \begin{bmatrix} 1+i \\ 1+i \\ -2 \end{bmatrix}$$

Can't multiply nicely \rightarrow dimension

So goal row space is not $C(A^+)$ but $C(A^+)$

(25)

10.2 #19 The function e^{-ix} and e^{ix} are orthogonal on interval $0 \leq x \leq 2\pi$
 since inner product is $\int_0^{2\pi} \dots = 0$



? access from circle
 but how is that orthogonal

well start at 0



? equal distances away

? how would you write this

26

Book

v_s are cols of a unitary matrix U

$$\text{So } U^H = U^{-1}$$

$$\begin{aligned} \text{Then } z &= U U^H z = \text{multiply by cols} \\ &= v_1 (v_1^H z) + \dots + v_n (v_n^H z) \end{aligned}$$

a typical orthonormal expansion

But how is this $\int_0^{2\pi} f(x) dx$

Fixed

$$\begin{aligned} \text{inner product} &= \int_0^{2\pi} e^{-ix} \cdot e^{ix} dx \\ &= \int_0^{2\pi} e^{2ix} dx \\ &= \frac{1}{2i} \left[e^{2ix} \right]_0^{2\pi} \\ &= 0 \quad \text{QED} \end{aligned}$$

(29)

$W_3 \neq 7$ Put vector $c = [1, 0, 1, 0]$ to find

$$Y = Fc$$

$1/10$

$$\begin{bmatrix} 1 & 1 & 1 & 1 \\ 1 & w & w^2 & w^3 \\ 1 & w^2 & w^4 & w^6 \\ 1 & w^3 & w^6 & w^9 \end{bmatrix} \begin{bmatrix} 1 \\ 0 \\ 1 \\ 0 \end{bmatrix}$$

$$Y_0 = 1 + 1$$

$$Y_1 = 1 + w^2$$

$$Y_2 = 1 + w^4$$

$$Y_3 = 1 + w^6$$

$$c = (0, 1, 0, 1)$$

$$Y_0 = 1 + 1$$

$$Y_1 = w + w^3$$

$$Y_2 = w^2 + w^6$$

$$Y_3 = w^3 + w^9$$

like that sin/cos thing

?

10.3 # 15

10/10

Fast convolution To multiply $C \circ x$

Can multiply $F(E(F^{-1}x))$ instead

Direct way uses n^2 sep multiplications

Knowing E and F , the 2nd way uses
 $n \lg_2 n + n$ multiplications

How many from E, F, F^{-1} ?
(6.006 qu)

↳ not really

Don't understand process well enough

Book

Diagonal $E = h$

F, F^{-1} need $\frac{1}{2}n \lg_2 n$ for each FFT



18.06 Spring 2012 – Problem Set 9

This problem set is due Thursday, May 3rd, 2012 at 4pm (hand in to Room 2-106). The textbook problems are out of the 4th edition. For computational problems, please include a printout of the code with the problem set (for MATLAB in particular, `diary('filename')` will start a transcript session, `diary off` will end one.)

Every problem is worth 10 points.

1. Do Problems 5 & 11 from Section 6.6.

Solution. Problem 5. $\begin{bmatrix} 1 & 1 \\ 0 & 0 \end{bmatrix}$, $\begin{bmatrix} 0 & 0 \\ 1 & 1 \end{bmatrix}$, $\begin{bmatrix} 1 & 0 \\ 1 & 0 \end{bmatrix}$ and $\begin{bmatrix} 0 & 1 \\ 0 & 1 \end{bmatrix}$ are similar (they all have eigenvalues 1 and 0).

$\begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix}$ is by itself (eigenvalue 1), and $\begin{bmatrix} 0 & 1 \\ 1 & 0 \end{bmatrix}$ is also by itself, with eigenvalues 1 and -1.

Problem 11. $\mathbf{u}(0) = \begin{bmatrix} 5 \\ 2 \end{bmatrix} = \begin{bmatrix} v(0) \\ w(0) \end{bmatrix}$. The equation

$$\frac{d\mathbf{u}}{dt} = \begin{bmatrix} \lambda & 1 \\ 0 & \lambda \end{bmatrix} \mathbf{u}$$

has

$$\frac{dv}{dt} = \lambda v + w \quad \text{and} \quad \frac{dw}{dt} = \lambda w.$$

Then $w(t) = 2e^{\lambda t}$ and $v(t)$ must include $2te^{\lambda t}$ (this comes from the repeated λ). To match $v(0) = 5$, the solution is $v(t) = 2te^{\lambda t} + 5e^{\lambda t}$. \square

2. Do Problems 17 & 19 from Section 6.6.

Solution. Problem 17

- (a) *False:* Diagonalize a nonsymmetric matrix $A = SAS^{-1}$. Then Λ is symmetric and similar.
(b) *True:* A singular matrix has $\lambda = 0$.
(c) *False:* $\begin{bmatrix} 0 & 1 \\ -1 & 0 \end{bmatrix}$ and $\begin{bmatrix} 0 & -1 \\ 1 & 0 \end{bmatrix}$ are similar (they have $\lambda = \pm 1$).
(d) *True:* Adding I increases all eigenvalues by 1.

Problem 19 Diagonal blocks 6 by 6, 4 by 4; AB has the same eigenvalues as BA plus 6 - 4 zeroes. \square

3. Do Problem 22 from Section 6.6.

Solution. Let $J = M^{-1}AM$ be the Jordan form of A . Since J is a strictly upper triangular matrix, $J^n = 0$. Hence $A^n = (M^{-1}JM)^n = M^{-1}J^nM = 0$. \square

4. Do Problems 3 & 6 from Section 6.7.

Solution. Problem 3 There is a orthogonal matrix V such that $VA^TAV^T = (VA^T) \cdot (VA^T)^T = \begin{bmatrix} \sigma^2 & 0 \\ 0 & 0 \end{bmatrix}$. Let $a_1 = [a_{11} a_{12}]$ and $a_2 = [a_{21} a_{22}]$. Since orthogonal matrices preserve length of vectors,

$$\sigma^2 + 0^2 = \|Va_1^T\|^2 = \|a_1^T\|^2 = a_{11}^2 + a_{12}^2$$

and

$$0^2 + 0^2 = \|Va_2^T\|^2 = \|a_2^T\|^2 = a_{21}^2 + a_{22}^2.$$

Hence $\sum a_{ij}^2 = \sigma^2$.

Problem 6

The eigenvalues of $AA^T = \begin{bmatrix} 2 & 1 \\ 1 & 2 \end{bmatrix}$ are 3 and 1. The corresponding normal eigenvectors are $\begin{bmatrix} \frac{1}{\sqrt{2}} \\ \frac{1}{\sqrt{2}} \end{bmatrix}$ and $\begin{bmatrix} \frac{1}{\sqrt{2}} \\ -\frac{1}{\sqrt{2}} \end{bmatrix}$. Hence we have $U = \begin{bmatrix} \frac{1}{\sqrt{2}} & \frac{1}{\sqrt{2}} \\ \frac{1}{\sqrt{2}} & -\frac{1}{\sqrt{2}} \end{bmatrix}$.

The eigenvalues of $A^T A = \begin{bmatrix} 1 & 1 & 0 \\ 1 & 2 & 1 \\ 0 & 1 & 1 \end{bmatrix}$ are 3, 1 and 0. The corresponding normal

eigenvectors are $\begin{bmatrix} \frac{1}{\sqrt{6}} \\ \frac{2}{\sqrt{6}} \\ \frac{1}{\sqrt{6}} \end{bmatrix}$, $\begin{bmatrix} \frac{1}{\sqrt{2}} \\ 0 \\ -\frac{1}{\sqrt{2}} \end{bmatrix}$, and $\begin{bmatrix} \frac{1}{\sqrt{3}} \\ -\frac{1}{\sqrt{3}} \\ \frac{1}{\sqrt{3}} \end{bmatrix}$. Hence we have $V = \begin{bmatrix} \frac{1}{\sqrt{6}} & \frac{1}{\sqrt{2}} & \frac{1}{\sqrt{3}} \\ \frac{2}{\sqrt{6}} & 0 & -\frac{1}{\sqrt{3}} \\ \frac{1}{\sqrt{6}} & -\frac{1}{\sqrt{2}} & \frac{1}{\sqrt{3}} \end{bmatrix}$.

Hence

$$A = U \begin{bmatrix} \sqrt{3} & 0 & 0 \\ 0 & 1 & 0 \end{bmatrix} V^T.$$

□

5. Do Problems 9 & 10 from Section 6.7.

Solution. Problem 9 Since A is a orthogonal matrix, $AA^T = A^T A = I$. The only eigenvalue of I is 1, so $A = UIV^T = UV^T$.

Problem 10 Note that $A = U\Sigma V^T$ where U and V are orthogonal matrices with first columns u and v , respectively, and $\Sigma = \begin{bmatrix} 12 & \\ & \end{bmatrix}$. Our matrix $A = 12uv^T$ and the only singular value if $\sigma_1 = 12$.

□

6. Do Problem 14 from Section 6.7.

Solution. Let $A = U\Sigma V^T = U \text{diag}(\sigma_1, \dots, \sigma_n) V^T$ with $\sigma_1 \geq \sigma_2 \geq \dots \geq \sigma_n > 0$. The closet singular matrix is $A_0 = U\Sigma_0 V^T = U \text{diag}(\sigma_1, \dots, \sigma_{n-1}, 0) V^T$. □

7. Do Problems 1 & 4 from Section 8.5.

Solution. Problem 1 If $j \neq k$, the integral of RHS is

$$\int_0^{2\pi} \cos(j+k)x + \cos(j-k)x dx = \frac{\sin(j+k)x}{j+k} + \frac{\sin(j-k)x}{j-k} \Big|_{x=0}^{2\pi} = 0.$$

Hence $(\cos jx, \cos kx) = 0$. If $j = k$,

$$\int_0^{2\pi} \cos(j+k)x + \cos(j-k)x dx = \frac{\sin(j+k)x}{j+k} + 1 \Big|_{x=0}^{2\pi} = 2\pi.$$

So $(\cos jx, \cos kx) = \pi$.

Problem 4 For any c , $x^3 - cx$ is orthogonal to 1 and $x^2 - \frac{1}{3}$, and

$$\int_{-1}^1 1x \cdot (x^3 - cx) dx = \frac{2}{5} - 2\frac{c}{3}.$$

If we let $c = \frac{3}{5}$, then $x^3 - cx$ is orthogonal to every other functions. □

8. Do Problems 3 & 19 from Section 10.2.

Solution. Problem 3 $z = C[-1-i, -1-i, 2]^T$ for some constant C . Since $z^H A^H = 0$, $z \bar{a}_i = 0$ for any column a_i of A . However, $z \cdot [-i, 1, -i] = -1 - i$. Hence columns of A^T are not orthogonal to z .

Problem 19 Note that $e^{-ix} = \cos x - i \sin x$ and $e^{ix} = \cos x + i \sin x$. Their inner product is

$$\int_0^{2\pi} e^{-ix} \overline{e^{ix}} dx = \int_0^{2\pi} \cos^2 x - \sin^2 x - 2i \cos x \sin x dx = \int_0^{2\pi} \cos 2x - i \sin 2x dx = 0. \quad \square$$

9. Do Problem 7 from Section 10.3.

Solution.

$$c = \begin{bmatrix} 1 \\ 0 \\ 1 \\ 0 \end{bmatrix} \rightarrow \begin{bmatrix} 1 \\ 1 \\ 0 \\ 0 \end{bmatrix} \rightarrow \begin{bmatrix} 2 \\ 0 \\ 0 \\ 0 \end{bmatrix} \rightarrow \begin{bmatrix} 2 \\ 0 \\ 2 \\ 0 \end{bmatrix} = Fc.$$

Also,

$$C = \begin{bmatrix} 0 \\ 1 \\ 0 \\ 1 \end{bmatrix} \rightarrow \begin{bmatrix} 0 \\ 0 \\ 1 \\ 1 \end{bmatrix} \rightarrow \begin{bmatrix} 0 \\ 0 \\ 2 \\ 0 \end{bmatrix} \rightarrow \begin{bmatrix} 2 \\ 0 \\ -2 \\ 0 \end{bmatrix} = FC. \quad \square$$

10. Do Problem 15 from Section 10.3.

Solution. Diagonal E needs n multiplications, while the Fourier matrix F and F^{-1} need $\frac{1}{2}n \log_2 n$ multiplications each by the FFT. The total is much less than the ordinary n^2 for C times x . \square

18.06 Wisdom. To reiterate: You should get as ready as you can for Exam 3, and for the finals, by doing as many old exams as you have time for (found on the 18.06 website under "Past Courses"). Get into a good rhythm with these rehearsals to stay fully tuned in on Linear Algebra, and to master the concepts - you will need it, both here and beyond.

18.06

5/4/12

(1 min late)

Find M if ~~possible~~ Possible

NOT possible

$$M \begin{bmatrix} 1 & 1 & 1 \\ 1 & 2 & 2 \\ 1 & 2 & 2 \end{bmatrix} M^{-1} = \begin{bmatrix} 1 & & \\ & 1 & \\ & & 1 \end{bmatrix}$$

If same λ 's ok I have to find M :

No repeated λ 's, \rightarrow they will be similar

Means will be diagonalizable
for sure A will be ok

Only if not \rightarrow need to do Jordan Form
details of Jordan form not on there

(2)

det $(A - \lambda I)$ and then find roots

↳ NOT good for finding λ on large matrices

is repeated MAM^{-1}

λ get better + better
more + more triangular

$$A = QR$$

↳ Look at $A_2 = RQ$

Note QR is similar to RQ

? more nearly triangular

$$M QR M^{-1} = RQ$$

$$Q^{-1} QR Q = RQ$$

3

$$A = \begin{bmatrix} 4 & 3 \\ 1 & 2 \end{bmatrix}$$

$$A^{100} = ?$$

Find λ

$$\begin{vmatrix} 4-\lambda & 3 \\ 1 & 2-\lambda \end{vmatrix} = \lambda^2 - 6\lambda + 5 = 0$$

$$(\lambda - 5)(\lambda - 1)$$

$$x = \begin{bmatrix} 3 \\ 1 \end{bmatrix} \quad \begin{bmatrix} 1 \\ -1 \end{bmatrix}$$

Are x orth? No, since original matrix is not symmetric

$$\frac{du}{dt} = Au$$

$$u(0) = \begin{bmatrix} 4 \\ 0 \end{bmatrix}$$

$$u = c_1 e^{5t} \begin{bmatrix} 3 \\ 1 \end{bmatrix} + c_2 e^t \begin{bmatrix} 1 \\ -1 \end{bmatrix}$$

(4)

At $t=0$

$$\begin{bmatrix} 4 \\ 1 \\ 0 \end{bmatrix} = C_1 \begin{bmatrix} 3 \\ 1 \\ 1 \end{bmatrix} + C_2 \begin{bmatrix} 1 \\ 1 \\ -1 \end{bmatrix}$$

\downarrow \downarrow
1 1

Repeated λ

Look for e vectors

ie $\begin{bmatrix} 1 & 1 & 1 \\ 1 & 1 & 1 \\ 1 & 1 & 1 \end{bmatrix} \lambda = 0, 0, 3$
? since singular

$x = \begin{bmatrix} \\ \\ \end{bmatrix} \begin{bmatrix} 1 \\ 1 \\ 1 \end{bmatrix}$
since symmetric (missed)

$\begin{bmatrix} 1 & 1 & 1 \\ 0 & 0 & 0 \\ 0 & 0 & 0 \end{bmatrix} \lambda = 1, 1, 1$
matrix singular

$A = 1I = \begin{bmatrix} 0 & 1 & 1 \\ 0 & 0 & 0 \\ 0 & 0 & 0 \end{bmatrix}$

(5)

$$\frac{du}{dt} = Au$$

$$\frac{d}{dt} \begin{bmatrix} u_1 \\ u_2 \\ u_3 \end{bmatrix} = \begin{bmatrix} 1 & 1 & 1 \\ 0 & 1 & 1 \\ 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} u_1 \\ u_2 \\ u_3 \end{bmatrix}$$

$$\lambda_1 = 0 \quad \leftarrow \text{know its singular } (\lambda=0)$$

$$\lambda_2 = \frac{1}{2}$$

$$\lambda_3 = 1$$

$$X = \begin{matrix} \lambda_1 & \lambda_2 & \lambda_3 \\ \begin{bmatrix} 1 \\ 1 \\ 1 \end{bmatrix} & \begin{bmatrix} -1 \\ 1 \\ 0 \end{bmatrix} & \begin{bmatrix} 1 \\ 1 \\ -2 \end{bmatrix} \end{matrix}$$

(could you reconstruct the matrix from this? Yes)

$$A = SAS^{-1} \leftarrow$$

Symmetric
Since the 3^x are orthogonal

6
Markov matrix? No

The largest $\lambda = 1$ is a good sign

But has \ominus in the X_3

λ for orthogonal Q is that when you multiply by a #, length does not change

$$\|Qx\|^2 = \|x\|^2$$

so all $\lambda = 1$

$$(Qx)^T(Qx) = x^T x$$

$$\begin{aligned} \text{Since } x^T Q^T Q x &= \\ &= x^T I x \\ &= x^T x \end{aligned}$$

Q6

Projection matrix? No

λ can only be 0 or 1
Perp to plane? projection onto

7

Pos def

$$A = \begin{pmatrix} 1 & b & 0 \\ b & 4 & 2 \\ 0 & 2 & 4 \end{pmatrix}$$

If that is \oplus def, what #s can b be?

If $b=0$, ~~or work~~ Yes

$b=10$ No

Largest b :

do $\begin{bmatrix} 1 & b \\ b & 4 \end{bmatrix}$ test

if $b=2$, this is on border line (semidef)

But does this cause the $\underline{3 \times 3}$ test to fail?

Yes, \ominus det = -4

So not semidef

invertible or singular? (did we answer?)

⑧ Requirements

$$4 - b^2 > 0$$

$$16 - 4 - 4b^2 > 0$$

$$12 > 4b^2$$

Fovier is not an exam

Complex is not an exam, or final

This exam is basically all the ~~exam~~ stuff

inc SVD

SVD

$$AV = U \Sigma$$

$$A = U \Sigma V^T$$

Try computing

$$A = \begin{bmatrix} 1 & 1 & 1 \\ 1 & 1 & 1 \end{bmatrix}$$

rank = 1

does not need to be
square

9

$$A^T A = \begin{bmatrix} 1 & 1 & 1 & 1 \\ 1 & 1 & 1 & 1 \\ 1 & 1 & 1 & 1 \\ 1 & 1 & 1 & 1 \end{bmatrix} \quad A A^T = \begin{bmatrix} 4 \\ 4 \\ 4 \\ 4 \end{bmatrix}$$

Want orthogonal x , but first need λ

$$\lambda = [4, 0, 0, 0] \quad \lambda = 4$$

note similar λ is on purpose

$$\begin{bmatrix} 1 & 1 & 1 & 1 \end{bmatrix} = \begin{bmatrix} u \\ 1 \end{bmatrix} \begin{bmatrix} 2 & 0 & 0 & 0 \end{bmatrix} \begin{bmatrix} v^T \\ \text{---} \\ \text{---} \\ \text{---} \end{bmatrix}$$

$\begin{matrix} \leftarrow \text{Col space} \\ \text{dim} = 1 \end{matrix}$
 $\begin{matrix} \leftarrow \text{Row space} \\ \text{dim} = 1 \end{matrix}$
 $\begin{matrix} \leftarrow \text{Null space} \\ \text{dim} = 3 \end{matrix}$

Singular value = $\sqrt{4} = 2$

$A v = \sigma u$ ~ just 1

Orthogonal bases for all the spaces

Prof: Be able to do SVD for a 2×2

(10)

$$A = \begin{bmatrix} 1/2 & 3/4 & 4 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix}$$

$$A^{k \rightarrow \infty}$$

What happens?

Diagonalize

$$\lambda = 1/2, 1, 1$$

$$\begin{bmatrix} (1/2)^k & & \\ & 1^k & \\ & & 1^k \end{bmatrix} \stackrel{\text{in limit}}{=} \begin{bmatrix} 0 & & \\ & 1 & \\ & & 1 \end{bmatrix}$$

Exam 11 AM Walker Man

18.06
Review Session

5/6

1-6 8.1 8.2 8.3 8.5

Fourier - Prof said no

But the early part is on there

Syllabus

Diff Eq

Symmetric Matrix

Pos Def Matrix

Engineering

Similar Matrices

SVD

Fourier Series, ~~FFT~~

(Not FFT)

Linear Transformations

②

Practice exam

$$1. A = \begin{bmatrix} .7 & .4 \\ .3 & .6 \end{bmatrix}$$

a) Find Λ and S

$$S^{-1}AS = \Lambda$$

b) Find ~~the~~ A^k

c) Find \lim as $k \rightarrow \infty$ of $u_k = A^k u_0$
if $u_0 = \begin{bmatrix} 1 \\ 1 \end{bmatrix}$

So I remember this qu

Can I find λ, x by hand?

$$\det(A - \lambda I) = 0$$

$$(0.7 - \lambda)(0.6 - \lambda) - 0.3 \cdot 0.4$$

$$0.42 - 0.7\lambda - 0.6\lambda + \lambda^2 - 0.12 = 0$$

$$\lambda^2 - 1.3\lambda + 0.30 = 0$$

(3)

$$b^2 \pm \sqrt{4ac}$$

$$\frac{-b \pm \sqrt{b^2 - 4ac}}{2a} \text{ Remember!}$$

$$\frac{1.3 \pm \sqrt{1.3^2 - 4 \cdot 1 \cdot .30}}{2 \cdot 1}$$

(small #'s!)

$$\begin{array}{r} 1.3 \\ 1.3 \\ \hline 3.9 \\ 1.30 \\ \hline 1.69 \end{array}$$

$$\frac{1.3 \pm \sqrt{1.69 - 1.2}}{2}$$

$$1.3 \pm \sqrt{.49}$$

$$\frac{1.3 \pm .07}{2}$$

$$\frac{1.37}{2}$$

$$\frac{1.23}{2}$$

$$X = \begin{bmatrix} .7 - \frac{1.37}{2} & .4 \\ .3 & .6 - \frac{1.37}{2} \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \end{bmatrix}$$

④

Am I doing this right \rightarrow too much algebra

then
$$\left[\begin{array}{cc|c} \cancel{A} & x_1 & x_1 \\ & x_2 & x_2 \end{array} \right] \left[\begin{array}{c} \lambda_1 \\ \lambda_2 \end{array} \right] \left[\begin{array}{c} \\ \\ \end{array} \right]$$

? need to find inv

What are the clues here?

If sym \rightarrow then $Q^{-1} = Q^T$?

But not sym, so find w/ Gauss Jordan?
Need to review

Multiply $[A \ I]$ by A^{-1} to get $[I \ A^{-1}]$

basically
$$\left[\begin{array}{cc|c} A & & I \end{array} \right] \leftarrow \text{do to both sides}$$

elim, then Rref

get
$$\left[\begin{array}{cc|c} I & & A^{-1} \end{array} \right]$$

6

A^k is just multiply A^k

Take the \lim to ∞

I don't know know what that $u_0 = \begin{bmatrix} 1 \\ 1 \end{bmatrix}$ means

~~the~~

Something diff eq possibly?

Answer

via 'lets find A

$$\begin{bmatrix} \lambda_1 & 0 \\ 0 & \lambda_2 \end{bmatrix}$$

Notice $A =$ markov matrix ohh...

$$\lambda = 1$$

$$a = .3$$

$$\begin{aligned} \text{Since } \sum \lambda &= \text{trace}(A) \\ &= \text{sum of diagonal} \\ 1 + \lambda_2 &= .7 + .6 \end{aligned}$$

5b

$$\prod \lambda = \det(A)$$

To find S

$$\lambda = 1 \mid N(A - \lambda I) = N(A - I)$$

$$= N \begin{pmatrix} -1 & 4 \\ 3 & -4 \end{pmatrix}$$

$$x_1 = \begin{bmatrix} 4 \\ 3 \end{bmatrix}$$

$$\leftarrow \text{ie } \begin{bmatrix} -1 & 4 \\ 3 & -4 \end{bmatrix} \begin{bmatrix} 4 \\ 3 \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \end{bmatrix}$$

$$\lambda = -3 \mid N(A - \lambda I)$$

$$= N \begin{pmatrix} 4 & 4 \\ 3 & 3 \end{pmatrix}$$

$$x_2 = \begin{bmatrix} 1 \\ -1 \end{bmatrix}$$

(56)

Pt in order of e-values

$$S = \begin{bmatrix} 4 & 1 \\ 3 & -1 \end{bmatrix} \quad a = \begin{bmatrix} 1 & 0 \\ 0 & 3 \end{bmatrix}$$

Now

If wanted S^{-1}

- could do Gauss Jordan

- or the short way

$$\frac{1}{ac-bd} \begin{bmatrix} d & -b \\ -c & a \end{bmatrix}$$

But don't have to find here

b) $A = S \Lambda S^{-1}$

$$A^k = (S \Lambda S^{-1})(S \Lambda S^{-1})$$

$$= S \Lambda^k S^{-1}$$

still need to multiply by $S S^{-1}$

(5d)

$$\begin{bmatrix} 4 + 3(0.3)^k & 4 - 4(0.3)^k \\ 3 - 3(0.3)^k & 3 + 4(0.3)^k \end{bmatrix}$$

c) $\lim_{k \rightarrow \infty} A^k u_0 = \frac{1}{7} \begin{bmatrix} 4 & 4 \\ 3 & 3 \end{bmatrix} \begin{pmatrix} 1 \\ 1 \end{pmatrix}$

\uparrow
starting point

since $(.3)^k \rightarrow 0$

$$= \frac{1}{7} \begin{pmatrix} 8 \\ 6 \end{pmatrix}$$

~~5d~~

6

2. Suppose A is an invertible 3×3 matrix

a) Show how to prove that $x^T A^T A x$ is always ≥ 0 if x is not 0

Why fail if A not \rightarrow ?

So this chap/section I'm particularly bad at.

So $A^T A = I$?

Well $A^{-1} A$

If $A =$ orthonormal cols then $A^T A = I$
 \uparrow Q called

[]

each col is length 1
 $\sqrt{\sum c_i^2}$

When dot product any 2 cols = 0

7

Nothing special w/ A sym

~~$AA^T = A^T A$~~

~~$AB^{-1} = B^{-1}A$~~

$(AB)^{-1} = B^{-1}A^{-1}$ always true

$AA^{-1} = A^T A$ if diagonal

$I \quad I \leftarrow$ in all cases

Or if $A = \begin{bmatrix} & & \\ & & \\ & & 3 \end{bmatrix}$

$\begin{bmatrix} & & \\ & & \\ & & 3 \end{bmatrix} \begin{bmatrix} & & \\ & & \\ & & 3 \end{bmatrix} = 3 \times 3$

Oh A is 3×3

So $\underbrace{A^T}_{3 \times 3} \underbrace{A}_{3 \times 3} = 3 \times 3$

8

b) Show prove $A^T A$ is similar to AA^T

Does it follow same e-value + e-vector?

This is what I was confused on

Similar $B = M^{-1} A M$ is similar to M
^ Any invertible matrix

Don't have to find Jordan Form
↳ if same, then similar

Can guess M - no robust way to find

⊗ In order to be similar must have same λ
but having same $\lambda \rightarrow$ does not mean similar
but will be if both have full set e-vectors
↳ i.e. diagonalizable

⑨ Similar def

A, B are $n \times n$ matrices (must be square)

A, B are similar if there is some $n \times n$ matrix M
such that $A = M^{-1}BM$

If A and B are similar, they have same evales.

Having the same evales does not guarantee similarity

eg $\begin{bmatrix} 2 & 0 \\ 0 & 2 \end{bmatrix}$ $\begin{bmatrix} 2 & 1 \\ 0 & 2 \end{bmatrix}$ are not similar

Guessing M is usually pretty hard

↳ won't likely need to find

A, B are similar if same Jordan form

10

So basically show

$$A^T A = M^{-1} A A^T M$$

∴ what next then?

Reversible

$$M A M^{-1} = \cancel{M} \cancel{M}^{-1} B \cancel{M} \cancel{M}^{-1}$$

Answer

$$(A B)^T = B^T A^T$$

$$\begin{aligned} \text{So } x^T A^T A x &= (A x)^T A x \\ &= \|A x\|^2 \geq 0 \end{aligned}$$

Only when $Ax = 0$

A is invertible

∴ $Ax = 0$ only when $x = 0$

Otherwise this quant is always positive

①

So ~~$A^T A = \text{dot product of vector}$~~

$\text{Vector}^T \text{Vector} = \text{dot product of vector}$
 $= \text{length of vector}$

b) $A^T A = A^{-1} A A^T A$

\uparrow
Can multiply by those because
~~they~~ def of similar

$\underbrace{\hspace{2cm}}_I$

$$A^T A = I A^T A$$

So similar

So then it follows (since similar) same

λ -value but not same e -vectors

(12)

c) If SVD is written $A = U \Sigma V^T$

What is $A^T A$ reduced to simplest form?
(didn't think about)

$$A^T A = (U \Sigma V^T)^T (U \Sigma V^T)$$

Using $(AB)^T = B^T A^T$

$$(ABC)^T = C^T B^T A^T$$

$$= (V^T)^T \Sigma^T \underbrace{U^T U}_I \Sigma V^T$$

$$= V \Sigma^T \Sigma V^T$$

not square so

~~not~~ Σ^2

Since not $n \times n$

also not orthonormal cols

That's it

13

13. Suppose a 3×3 real sym matrix A has
e-values $4, 4, 0$

a) Find the det from evalues $(2A - I)^{-1}$
?? why $2A$

$$\text{So } A = S \Lambda S^{-1}$$

$$\begin{bmatrix} 4 & & \\ & 4 & \\ & & 0 \end{bmatrix}$$

This means its likely that
other form I forget: the name of
SVD

$$A = U \Sigma V^T$$

$$\begin{aligned} \text{But really } \det(A) &= \prod \lambda \\ &= 4 \cdot 4 \cdot 0 = 0 \end{aligned}$$

But they want $2A$?

(14)

b) T or F A has 3 independent vectors
and can be diagonalizable

false - it repeats
 $\lambda = 0$ \rightarrow which is it?
or both

c) ~~T~~ or F $x^T A x$ is never negative

didn't we just show that's always true?

d) T or F $\frac{1}{4}A$ is ~~the~~ Markov

False \rightarrow Markov is diagonalizable - don't think so...

it has only 1 and < 1

Just guessing F - feel pretty sure

(15)

e) If A has orthonormal e vectors

q_1, q_2, q_3 w/ $\lambda = 4, 4, 0$

Find a formula for A in terms of q_1, q_2, q_3 using diagonalization

What does diagonalization mean again?

$$S^T A S = \Lambda = \begin{pmatrix} \lambda_1 & & \\ & \lambda_2 & \\ & & \dots \end{pmatrix}$$

That's what I thought...

But does this have to be SVD

$$\begin{bmatrix} q_1 & q_2 & q_3 \end{bmatrix} \begin{bmatrix} 4 & & \\ & 4 & \\ & & 0 \end{bmatrix} \begin{bmatrix} v_1 & v_2 & v_3 \end{bmatrix}^T$$

That's σ not λ

$$\Sigma = \begin{bmatrix} \sigma_1^2 & & \\ & \sigma_2^2 & \\ & & \dots \end{bmatrix}$$

(15b)

Answer

a) Need to find the e-values of $(2A-I)^{-1}$
 M has e-value λ
 So M^{-1} has e-value $\frac{1}{\lambda}$
 So it can find e-values of $2A-I$

Since $Mx = \lambda x$
 $Ix = x$

$(M-I)x = (\lambda-1)x$

$M-I$ has e-value $\lambda-1$

A has e-values $4, 4, 10$

$2A$ has e-values $8, 8, 0$

$2A-I$ has $7, 7, -1$

$(2A-I)^{-1}$ has $\frac{1}{7}, \frac{1}{7}, -1$

det is \prod so $\frac{1}{7} \cdot \frac{1}{7} \cdot -1 = -\frac{1}{49}$

(150)

b) We do have enough info
If 3×3 real sym matrix \rightarrow can always
diag
remember
were told

(True)

c) (True) A is pos semi det
This is one of the tests

If diag. then all e-vect ind
Try to find e-vectors to see if diag.
0 e-value means not inv

d) ~~Is~~ Is there an A such that $\frac{1}{4}A$ is Markov
since if there is \rightarrow not enough info

Is not enough info since

True if $\frac{1}{4}A = \begin{bmatrix} 1 & 0 & 0 \\ 0 & .5 & .5 \\ 0 & .5 & .5 \end{bmatrix}$

(50)

False $\frac{1}{4}A = \begin{pmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 0 \end{pmatrix}$

So diagonalizable to check

TA: kinda guess

Markov is a strong condition

e) How to diag A ?

But it gives the vectors

So just write the formula

$$A = \begin{bmatrix} | & Q & | \\ a_1 & q_2 & q_3 \\ | & | & | \end{bmatrix} \begin{bmatrix} 4 & & \\ & 4 & \\ & & 0 \end{bmatrix} \begin{bmatrix} Q^T \\ - & a_1 & - \\ - & q_2 & - \\ & q_3 & - \end{bmatrix}$$

↑ were told orthonormal
↑ order must match

Multiply it out

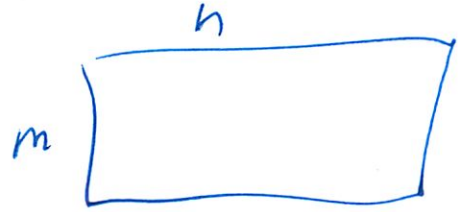
$$= 4q_1q_1^T + 4q_2q_2^T$$

Might have had to orthonormalize w/ Gram Schmidt

(16)

SVD Primer

$A = m \times n$ matrix
rank r



Look at AA^T $m \times m$ orthogonal vectors u_1, \dots, u_r
 $A^T A$ $n \times n$ " " v_1, \dots, v_r } $AV_i = \sigma_i u_i$
Core square so can find σ_i
same rank r

To find σ_i , solve $Av_i = \sigma_i u_i$ } Then at those that correspond to $\sigma = 0$
Must normalize or σ screwed up

So use Gram-Schmidt

Already orthogonal unless the λ same \rightarrow must orthonormalize
if diff

$V = [v_1 \dots v_r]$ \leftarrow want to be square
~~So add $v_{r+1} \dots v_n$~~
So add $v_{r+1} \dots v_n$ orthogonal basis for $N(A)$
with $\sigma = 0$

16b

$$U = \left[\begin{array}{c|c|c} \begin{array}{c} | \\ \underline{u_1} \\ | \end{array} & \dots & \begin{array}{c} | \\ \underline{u_r} \\ | \end{array} & \begin{array}{c} | \\ \underline{u_{r+1}} \\ | \end{array} & \dots & \begin{array}{c} | \\ \underline{u_n} \\ | \end{array} \end{array} \right]$$

Orthogonal basis
for $N(A^T)$

Enlarge w/

(17)

Ya Find the evalues of $A = \begin{pmatrix} 0 & 1 \\ 1 & 0 \end{pmatrix}$ and $R = \begin{pmatrix} 0 & -1 \\ 1 & 0 \end{pmatrix}$

Then evalues $A = \begin{pmatrix} a & 1 \\ 1 & a \end{pmatrix}$ $R = \begin{pmatrix} b & -1 \\ 1 & b \end{pmatrix}$

Looks like pos def ...

a, b real

a) Under what condition ~~is~~ on " a " do all
sols of $\frac{du}{dt} = Au$ approach 0 as $t \rightarrow \infty$

Answers

A has e-values $1, -1$ ($a+1, a-1$)

R : $i, -i$
($b+1, b-i$)

b) $a < -1$

c) $b < 0$

d) $a > 1$

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This is a diff eq problem

When solving diff eq w/ bounds \rightarrow look at λ values

So λ is \ominus when want sol stable
 λ \oplus instable

g) Since complex λ - want real parts negative

Q2 Diagonalizing

$$S^{-1}AS = \Lambda$$

So $\Lambda = e$ values λ

$S = e$ vectors

$$\prod \lambda = \det$$

$$\sum \lambda = \text{trace} = \sum \text{diagonal entries}$$

$$\begin{aligned} \Omega &= SAS^{-1}SAS^{-1} = \\ &= S\Lambda^2S^{-1} \end{aligned}$$

$$U_{k+1} = \begin{bmatrix} F_{k+1} \\ \hat{F}_k \end{bmatrix} U_k$$

$AB = BA \rightarrow$ means share same e -vector

\because always share same e -vector

don't think so since not auto true

②

6.3 Application to Diff eq

$$\frac{du}{dt} = Au \quad \text{w/ } u(0) \text{ at } t=0$$

$$v = e^{At} x \quad \text{when } Ax = v$$

~~Then solve for~~

put together Σ solutions

$$u(t) = C e^{t \begin{bmatrix} 1 \\ 1 \end{bmatrix}} + D e^{-t \begin{bmatrix} 1 \\ -1 \end{bmatrix}} =$$

? specific sol

Solve for C, D w/ $u(0)$

Should practice this

$$\frac{dy}{dt} = Au = \begin{bmatrix} 0 & 1 \\ 1 & 0 \end{bmatrix} u \quad u(0) = \begin{pmatrix} y \\ z \end{pmatrix}$$

So this is

$$\frac{d}{dt} \begin{bmatrix} y \\ z \end{bmatrix} = \begin{bmatrix} 0 & 1 \\ 1 & 0 \end{bmatrix} \begin{bmatrix} y \\ z \end{bmatrix}$$

3

Find λ

~~AE~~

$$\det(A - \lambda I)_{ax} = 0$$

\ominus ~~no~~ _{yes} $\lambda = -1$??

well want eigenvectors

$$\lambda = -1$$

$$\begin{bmatrix} -1 & 1 \\ 0 & -1 \end{bmatrix} \begin{bmatrix} x \\ y \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \end{bmatrix}$$

$$-x + y = 0$$

$$x + -y = 0$$

$$x = y \quad \begin{bmatrix} 1 \\ 1 \end{bmatrix}$$

Somehow

$\lambda = 1$ imarkov

$$\begin{bmatrix} 1 & 1 \\ 1 & 1 \end{bmatrix} \begin{bmatrix} x \\ y \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \end{bmatrix}$$

$$x + y = 0$$

$$\begin{bmatrix} 1 \\ -1 \end{bmatrix}$$

9

6

$$e^{1x} \begin{bmatrix} 1 \\ 1 \end{bmatrix} + e^{-1x} \begin{bmatrix} 1 \\ -1 \end{bmatrix}$$

$$\begin{bmatrix} 4 \\ 2 \end{bmatrix} = C e^{x} \begin{bmatrix} 1 \\ 1 \end{bmatrix} + D e^{-x} \begin{bmatrix} 1 \\ -1 \end{bmatrix}$$

$$e^0 = 1$$

$$\begin{bmatrix} 4 \\ 2 \end{bmatrix} = C \begin{bmatrix} 1 \\ 1 \end{bmatrix} + D \begin{bmatrix} 1 \\ -1 \end{bmatrix}$$

This was given

$$C + D = 4$$

$$C - D = 2$$

$$C = 2 + D$$

$$2 + D + D = 4$$

$$2D = 2$$

$$D = 1$$

$$C = 3$$

$$u(x) = 3e^{1x} \begin{bmatrix} 1 \\ 1 \end{bmatrix} + 1e^{-x} \begin{bmatrix} 1 \\ -1 \end{bmatrix}$$

Ok final should be a
had a

5

Second order

$$d \begin{bmatrix} y \\ y' \end{bmatrix} = \begin{bmatrix} \quad \\ \quad \end{bmatrix} \begin{bmatrix} y \\ y' \end{bmatrix}$$

Matrix exponential
skipping

lec 4 Symmetric | $P = P^T$

e-vectors of P and P^T are \perp unit vectors
*most important matrices

~~$A \neq A^T$~~

~~$S S^T = I$~~ ~~$S S^T = I$~~ $S = S^T$ ~~right~~ yes only when orthogonal
 $Q^T = Q^{-1}$

Only real e-values
e-vectors can be chosen orthonormal

Since symmetric

6

$$A = Q \Lambda Q^{-1} = Q \Lambda Q^T$$

(Remember what stuff is what!)

E values can be complex
↳ but not if sym!

Prod of pivots = determine

$$\det A = \det A^T$$

pos ~~or~~ eig values = # pos pivots

All sym matrices are diag.

Always enough eig vectors

(I never look in worked examples section...)

(Despite me relying on this...)

①

6.5 Pos Definite Matrices

Sym matrices that have \oplus evals
makes it truly special (but not rare)

1. all pivots > 0
2. all $\Delta > 0$ } but these statements are
equivalent

3. All upper left det are \oplus

$$\begin{array}{l} \det=5 \quad \begin{array}{|c|c|} \hline 5 & 4 \\ \hline 4 & 3 \\ \hline \end{array} \quad \begin{array}{l} \oplus \\ \oplus \end{array} \\ \det=-1 \quad \begin{array}{|c|c|} \hline 5 & 4 \\ \hline 4 & 3 \\ \hline \end{array} \quad \text{No} \end{array}$$

So for $\begin{pmatrix} a & b \\ c & d \end{pmatrix}$ $a \geq 0$ and $ac - b^2 > 0$

4. A can be factored into $A = \underbrace{L}_{\text{lower triangular}} L^T$

Can't be singular

Cols must be ind

①

this is sym

5. $A = A^T$ if $x^T A x > 0$
for all $x \neq 0$
(the normal def)

(pos def must be sym)
Don't think so - WP

But in this chap they need to be!

also note

$$x = \begin{bmatrix} x \\ y \end{bmatrix}$$

$$x^T A x =$$

$$\begin{bmatrix} x & y \end{bmatrix} \begin{bmatrix} a & b \\ c & d \end{bmatrix} \begin{bmatrix} x \\ y \end{bmatrix} = ax^2 + 2bxy + cy^2 > 0$$

~~A~~

Just look at $x^T A x > 0$
means pos def

$x^T A x$ is the energy in a system
where the other rules are from

If A, B sym def, $A + B$ will be

(9)

When it has one of the properties it has them all

Also $A = R^T R$ for matrix R w/ ind cols

↳ If the cols of R are ind, $A = R^T R$ is \oplus def

Pos Semi Def

When $x^T A x = 0$

When $\lambda = 0$

$$Ax = \lambda x$$

$$x^T A x = \lambda x^T x$$

$$x^T A x = \lambda |x|^2$$

$$\text{So } \oplus \times \oplus = \oplus$$

A factors in $R^T R$ w/ dep cols in R

like Similar Matrices

$B = M^{-1} A M$ is similar to A

↳ Any invertible matrix

(must be square I'm pretty sure)

Can't actually find, but can guess...

10

What is $x^T x$

So for sym/orthonormal $x^T = x^{-1}$
So $x^{-1} x = I$

but normally ...

(TA question - but don't have ...)

$$\|v\| = \text{length} = \sqrt{v^2}$$

$$\text{dot product } v \cdot v = v_1 u_1 + v_2 u_2$$

So is $x^T x$ the length?

$$x = \begin{bmatrix} \vdots \\ \vdots \\ \vdots \\ \vdots \\ \vdots \end{bmatrix}^5$$

$$5 \begin{bmatrix} \vdots \\ \vdots \\ \vdots \\ \vdots \\ \vdots \end{bmatrix}^T \begin{bmatrix} \vdots \\ \vdots \\ \vdots \\ \vdots \\ \vdots \end{bmatrix}^5 = 5 \times 5 \begin{bmatrix} \quad \end{bmatrix}$$

length squared?

$A^T A$ is square, sym, pos semi def

$$(AB)^T = B^T A^T$$

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And last time we say $(AB)^{-1} = B^{-1}A^{-1}$ ☺

$$\|v\| = \sqrt{v_1^2 + v_2^2 + v_3^2}$$

The dot product of a vector is the length squared

$$v = [1, 2, 3]$$

$$\|v\|^2 = \begin{bmatrix} 1 \\ 2 \\ 3 \end{bmatrix} \cdot \begin{bmatrix} 1 \\ 2 \\ 3 \end{bmatrix} = 1 + 4 + 9 = 14$$

So it must be $x = x^T$ is $\|x\|^2$

Since

Plus here $A = A^T$ Sym

Since $x = \text{vector!}$

(12)

Similar

Eigenvalues don't change for similar
eigenvectors do

det, trace, rank, Jordan form same/unchanged

eigenvectors, col, row, null space changed

Ex 7 SVD

$$A v_1 = \sigma_1 u_1$$

$$A v_2 = \sigma_2 u_2$$

$$A v_3 = \sigma_3 u_3$$

⋮

$$A = U \Sigma V^T$$

~~matrix~~

$$w/ \quad A A^T = U \Sigma U^T$$

$$A^T A = V \Sigma V^T$$

Be sure to actually try one...

$$\begin{bmatrix} 2 & 2 \\ 1 & 1 \end{bmatrix}$$

So



$$AA^T$$

$$\begin{bmatrix} 2 & 2 \\ 1 & 1 \end{bmatrix} \begin{bmatrix} 2 & 1 \\ 2 & 1 \end{bmatrix} = \begin{bmatrix} 8 & 4 \\ 4 & 2 \end{bmatrix} \quad \textcircled{1}$$

$$\det = (8-\lambda)(2-\lambda) - 16 = 0$$

$$16 - 10\lambda + \lambda^2 - 16 = 0$$

$$\lambda^2 - 10\lambda = 0$$

$$\frac{+10 \pm \sqrt{100 - 4 \cdot 1 \cdot 0}}{2}$$

$$\frac{10 \pm 10}{2} \quad \lambda = 0, 10 \quad \textcircled{2}$$

$$\lambda = 0$$

$$\begin{bmatrix} 8 & 4 \\ 4 & 2 \end{bmatrix} \begin{bmatrix} x \\ y \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \end{bmatrix}$$

$$8x + 4y = 0$$

$$\cancel{4x + 2y = 0}$$

unspecified or multiple...

$$8x = -4y$$

$$2x = -y$$

$$\begin{bmatrix} -1 \\ 2 \end{bmatrix}$$

(14)

$$\lambda = 10$$

$$\begin{bmatrix} -2 & 4 \\ 4 & -8 \end{bmatrix} \begin{bmatrix} x \\ y \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \end{bmatrix}$$

$$-2x + 4y = 0$$

$$4y = 2x$$

$$2y = x$$

$$4x = +8y$$

$$x = +2y$$

~~Contradict~~

$$\begin{bmatrix} 2 \\ 1 \end{bmatrix}$$


$$WA = \begin{bmatrix} 1/\sqrt{5} \\ -2/\sqrt{5} \end{bmatrix} \begin{bmatrix} -2/\sqrt{5} \\ -1/\sqrt{5} \end{bmatrix}$$

So ? what is length of $\begin{bmatrix} -1 \\ 2 \end{bmatrix}$

$$\sqrt{1^2 + 2^2} = \sqrt{5} \text{ so they did it right}$$

but my signs are diff

~~And why did they conflict above?~~

But they ~~should~~ work $2(-1) = (-2)$ 

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$$\text{So } u = \begin{bmatrix} 1 & 2 \\ -2 & 1 \end{bmatrix}$$

Now v is $A^T A$

$$\begin{bmatrix} 2 & 1 \\ 2 & 1 \end{bmatrix} \begin{bmatrix} 2 & 2 \\ 1 & 1 \end{bmatrix} = \begin{bmatrix} 3 & 5 & 3 & 5 \\ 3 & 5 & 3 & 5 \end{bmatrix} \text{ Can't add}$$

$$\det (5\lambda - \lambda)(3\lambda - \lambda) - 25$$
$$25\lambda^2 - 10\lambda + \lambda^2 - 25 = 0$$
$$\lambda^2 - 10\lambda = 0$$

(doing this the long way)

~~$$\frac{a \pm \sqrt{b^2 - 4ac}}{2}$$~~

$$\lambda = 0, 10$$

$$\lambda = 0 \left| \begin{bmatrix} 3 & 5 \\ 3 & 5 \end{bmatrix} \begin{bmatrix} x \\ y \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \end{bmatrix} \right.$$
$$\begin{bmatrix} 1 \\ -1 \end{bmatrix}$$

(16)

$$\lambda = 7$$

$$\begin{bmatrix} -3 & 3 \\ 3 & -3 \end{bmatrix} \begin{bmatrix} x \\ y \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \end{bmatrix}$$

$$-3x + 3y = 0$$

$$3x = 3y$$

$$x = y$$

$$\begin{bmatrix} 1 \\ 1 \end{bmatrix}$$

$$V = \begin{bmatrix} 1 & 1 \\ -1 & 1 \end{bmatrix}$$

So σ is
Oh wait it should be same -

$$0, 10$$

~~$$\sigma = 0, \sqrt{10}$$~~

$$\sigma = 0, \sqrt{10}$$

But Σ is σ^2 so $\begin{bmatrix} 0 & \\ & 10 \end{bmatrix}$

Don't forget to normalize!!
Orthogonal

(17)

$$A = U \Sigma V^T$$

$$\begin{bmatrix} 2 & 2 \\ 1 & 1 \end{bmatrix} = \frac{1}{\sqrt{5}} \begin{bmatrix} 2 & 1 \\ 1 & -2 \end{bmatrix} \begin{bmatrix} \sqrt{10} & \\ & 0 \end{bmatrix} \frac{1}{\sqrt{2}} \begin{bmatrix} 1 & 1 \\ 1 & -1 \end{bmatrix}$$

↑ note order

I think you can also swap order

but ~~not~~ non zero values usually first

And A does not have to be square
had some trouble on p-set
(which I now fixed in)
(but should try -)

$$A = \begin{bmatrix} 2 & 1 \end{bmatrix}$$

$$\text{So } AA^T = \begin{bmatrix} 2 & 1 \end{bmatrix} \begin{bmatrix} 2 \\ 1 \end{bmatrix} = \begin{bmatrix} 5 \end{bmatrix}$$

$\lambda = 5$

~~5~~ but how

$$\begin{bmatrix} 5 & -5 \end{bmatrix} \begin{bmatrix} x \\ x \end{bmatrix} = 0 ?$$

Wolfram alpha

$$x = 1$$

? can be any value?

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$$A^T A$$

$$\begin{bmatrix} 2 \\ 1 \end{bmatrix} \begin{bmatrix} 2 & 1 \end{bmatrix} = \begin{bmatrix} 4 & 2 \\ 2 & 1 \end{bmatrix}$$

$$(4-\lambda)(1-\lambda) - 4$$

$$4 - 5\lambda + \lambda^2 - 4 = 0$$

$$\lambda^2 - 5\lambda$$

$$\frac{5 \pm 5}{2}$$

$$\lambda = 0, 5$$

∴ yeah adds 0

λ=0

$$\begin{bmatrix} 4 & 2 \\ 2 & 1 \end{bmatrix} \begin{bmatrix} x \\ y \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \end{bmatrix}$$

$$4x + 2y = 0$$

$$2x = -y$$

$$\begin{bmatrix} 1 \\ -2 \end{bmatrix}$$

λ=5

$$\begin{bmatrix} -1 & 2 \\ 2 & -4 \end{bmatrix} \begin{bmatrix} x \\ y \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \end{bmatrix}$$

$$-x + 2y = 0$$

$$2y = x$$

$$\begin{bmatrix} 2 \\ 1 \end{bmatrix}$$

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So then what

$$\begin{bmatrix} 1 \end{bmatrix} \begin{bmatrix} \sqrt{5} & 0 \\ 0 & 0 \end{bmatrix} \begin{bmatrix} 2 & 1 \\ 1 & -2 \end{bmatrix}$$

Try it

$$\begin{bmatrix} 1 & 1 & 2 & 2 \end{bmatrix} \otimes \text{error}$$

(Wish I had my p-set where I figured this out...)

Or add 0s:

$$\begin{bmatrix} 1 & 0 \\ 0 & 0 \end{bmatrix}_{2 \times 2} \begin{bmatrix} \sqrt{5} & 0 \\ 0 & 0 \end{bmatrix}_{2 \times 2} \text{ works} = \begin{bmatrix} \sqrt{5} & 0 \\ 0 & 0 \end{bmatrix}$$

Then $2 \times 2 \times 2 \times 2$ works

$$\begin{bmatrix} 2\sqrt{5} & \sqrt{5} \\ 0 & 0 \end{bmatrix}$$

Oh never normalized!

$$\sqrt{2^2 + 1^2} = \sqrt{5}$$

$$\frac{1}{\sqrt{5}} \begin{bmatrix} 2\sqrt{5} & \sqrt{5} \\ 0 & 0 \end{bmatrix} = \begin{bmatrix} 2 & 1 \\ 0 & 0 \end{bmatrix} \text{ works}$$

(20)

Basically it appears just add 0s

8.1 Matrices in Eng

I hope he doesn't ask anything here...
Absent that day and no video online...

8.2 Graphs in Networks

(practice exam didn't cover this much)

incidence matrix

note every vector in nullspace is \perp to every vector
in row space

8.3 Markov Matrices

cols add to 1
all values pos or 0

Converges to same U_{∞} no matter U_0
↳ That's why I was confused in
the review ~~class~~ session

(20)

So λ is 1 and < 1
(look at trace)

$$u_0 = \lambda^k \cdot a \text{ (coeff)} \begin{bmatrix} 1 \\ 1 \end{bmatrix} + \text{other}$$

$$u_k = x_1 + c_2 (\lambda_2)^k x_2 + \dots + c_n (\lambda_n)^k x_n$$

Where does coeff come from?

Oh from start u_0 - like $(+D)$ in diff eq

I believe

Can use w/ consumption matrix

8.5 Fourier Series

From finite dimensions to ∞ dim

Vector becomes a function $f(x)$

$\|v\|$ must be finite

Is hermitian on \mathbb{R}^n
not in index!

- 10.2 not on sheet

(22)

Schwartz in eq $|v \cdot w| \leq \|v\| \|w\|$

inner product

$$(f, g) = \int_0^{2\pi} f(x) g(x) dx$$

length squared

$$\|f\|^2 = \int_0^{2\pi} (f(x))^2 dx$$

Fourier series

$$x(x) = a_0 + a_1 \cos x + b_1 \sin x + a_2 \cos 2x + b_2 \sin 2x + \dots$$

Fourier coeffs

$$a_n = \frac{1}{2\pi} \int_0^{2\pi} f(x) \cos kx dx$$

$$b_n = \frac{1}{2\pi} \int_0^{2\pi} f(x) \sin kx dx$$

Will we have to do anything here?

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Inverse Practice Gauss Jordan

$$A = \begin{bmatrix} 2 & -1 & 0 \\ -1 & 2 & -1 \\ 0 & -1 & 2 \end{bmatrix}$$

Convert AI to IA^{-1}

$$l = -\frac{1}{2} \begin{bmatrix} 2 & -1 & 0 & 1 & 0 & 0 \\ -1 & 2 & -1 & 0 & 1 & 0 \\ 0 & -1 & 2 & 0 & 0 & 1 \end{bmatrix}$$

$$\begin{matrix} -1 - (-\frac{1}{2})2 \\ 2 - (-\frac{1}{2})-1 \end{matrix} \begin{bmatrix} 2 & -1 & 0 & 1 & 0 & 0 \\ 0 & 1\frac{1}{2} & -1 & \frac{1}{2} & 1 & 0 \\ 0 & -1 & 2 & 0 & 0 & 1 \end{bmatrix}$$

$$l = \frac{1}{2}$$

$$l = -\frac{1}{1.5} = -\frac{2}{3} \begin{bmatrix} 2 & -1 & 0 & 1 & 0 & 0 \\ 0 & 1\frac{1}{2} & -1 & \frac{1}{2} & 1 & 0 \\ 0 & 0 & \frac{4}{3} & \frac{1}{3} & \frac{2}{3} & 1 \end{bmatrix}$$

$$-1 - (-\frac{2}{3})(1.5)$$

$$2 - (-\frac{2}{3})(-1)$$

$$0 - (\frac{1}{2})(-\frac{2}{3})$$

↑ Why did I screw up this side?

$$0 - (-\frac{1}{2})(1)$$

Edk only 20

Then other one is wrong

(24)

Now REF

Want 0s above pivots

rows added to rows above them

So row 2 = ~~$\frac{3}{4}$ row 2~~ row 2 + $\frac{3}{4}$ row 3

$$-1 + \frac{3}{4} \left(\frac{4}{3} \right) \begin{pmatrix} 2 & -1 & 0 & 1 & 0 & 0 \\ 0 & \frac{3}{2} & 0 & \frac{3}{4} & \frac{3}{2} & \frac{3}{4} \\ 0 & 0 & \frac{4}{3} & \frac{1}{3} & \frac{2}{3} & 1 \end{pmatrix} \quad \textcircled{0}$$

$\frac{1}{2} + \frac{3}{4} \left(\frac{1}{3} \right)$

$\frac{1}{2} + \frac{3}{12} - \frac{1}{12} = \frac{3}{4}$

$1 + \frac{3}{4} \left(\frac{2}{3} \right)$

$= 1 + \frac{6}{12}$

Now row 1 = row 1 + $\frac{2}{3}$ row 2 ⓪ figured it out

$$1 + \frac{2}{3} \left(\frac{3}{4} \right) \begin{pmatrix} 2 & 0 & 0 & \frac{3}{2} & 1 & \frac{1}{2} \\ 0 & \frac{3}{2} & 0 & \frac{3}{4} & \frac{3}{2} & \frac{3}{4} \\ 0 & 0 & \frac{4}{3} & \frac{1}{3} & \frac{2}{3} & 1 \end{pmatrix} \quad \text{Close}$$

$1 + \frac{6}{12}$

$\frac{2}{3} \left(\frac{3}{4} \right) = \frac{6}{12}$

25

Now Gram Schmidt

to make orthonormal vectors

$$\text{for } A = QR$$

messed this up on prev exam -
R is weird/stress

$$\begin{bmatrix} a \\ b \\ c \end{bmatrix} = \begin{bmatrix} q_1 & q_2 & q_3 \\ | & | & | \\ | & | & | \end{bmatrix} \begin{bmatrix} q_1^T a & q_1^T b & q_1^T c \\ q_2^T b & q_2^T c \\ q_3^T c \end{bmatrix}$$

orthogonal
original

$$A = a$$

$$B = b - \frac{A^T b}{A^T A} A$$

lots of painful multiplying

$$C = c - \frac{A^T c}{A^T A} A - \frac{A^T B}{A^T A} B$$

care looking at
look at old ones

26

Try it

$$a = \begin{bmatrix} 1 \\ -1 \\ 0 \end{bmatrix} \quad b = \begin{bmatrix} 2 \\ 0 \\ -2 \end{bmatrix} \quad c = \begin{bmatrix} 3 \\ -3 \\ 3 \end{bmatrix}$$

$$A = a = \begin{bmatrix} 1 \\ -1 \\ 0 \end{bmatrix}$$

$$B = b - \frac{A^T b}{A^T A} A$$

$$= \begin{bmatrix} 2 \\ 0 \\ -2 \end{bmatrix} - \frac{[1 \ -1 \ 0]_3 \begin{bmatrix} 2 \\ 0 \\ -2 \end{bmatrix}}{[1 \ -1 \ 0] \begin{bmatrix} 1 \\ -1 \\ 0 \end{bmatrix}} \begin{bmatrix} 1 \\ -1 \\ 0 \end{bmatrix}$$

$$= \begin{bmatrix} 2 \\ 0 \\ -2 \end{bmatrix} - \frac{[2 + 0 + 0]}{[1 + 1 + 0]} \begin{bmatrix} 1 \\ -1 \\ 0 \end{bmatrix}$$

$$= \begin{bmatrix} 2 \\ 0 \\ -2 \end{bmatrix} - \begin{bmatrix} 1 \\ -1 \\ 0 \end{bmatrix} = \begin{bmatrix} 1 \\ +1 \\ -2 \end{bmatrix} \quad \begin{bmatrix} 1 \\ -1 \\ -2 \end{bmatrix}$$

~~WTF~~ Fugred' it out
copy error, algebra error

(27)

$$C = C - \frac{B^T C}{B^T A} B - \frac{A^T C}{A^T A} A \quad \text{D}$$

$$= \begin{bmatrix} 3 \\ -3 \\ 0 \end{bmatrix} - \frac{\begin{bmatrix} 1 & 1 & -2 \end{bmatrix} \begin{bmatrix} 3 \\ -3 \\ 3 \end{bmatrix}}{\begin{bmatrix} 1 & 1 & 2 \end{bmatrix} \begin{bmatrix} 1 \\ 1 \\ 2 \end{bmatrix}} \begin{bmatrix} 1 \\ 1 \\ 2 \end{bmatrix} - \frac{A^T C}{A^T A} A$$

$$= \begin{bmatrix} 3 \\ -3 \\ 0 \end{bmatrix} - \frac{\begin{bmatrix} 3 & -3 & -6 \end{bmatrix}}{\begin{bmatrix} 1 & 1 & 4 \end{bmatrix}} \begin{bmatrix} 1 \\ 1 \\ 2 \end{bmatrix} - \frac{A^T C}{A^T A} A$$

$$= \begin{bmatrix} 3 \\ -3 \\ +3 \end{bmatrix} - \frac{-6}{6} \begin{bmatrix} 1 \\ 1 \\ 2 \end{bmatrix} = \begin{bmatrix} 4 \\ -2 \\ 5 \end{bmatrix} - \text{etc}$$

(28)

$$= \begin{bmatrix} 4 \\ -2 \\ 5 \end{bmatrix} - \frac{\begin{bmatrix} 1 & -1 & 0 \\ 3 \\ -3 \\ 3 \end{bmatrix}}{\begin{bmatrix} 4+2+2 \\ 1 & -1 & 0 \\ -1 \\ 0 \end{bmatrix}} \begin{bmatrix} 1 \\ -1 \\ 0 \end{bmatrix}$$

$$= \begin{bmatrix} 4 \\ -2 \\ 5 \end{bmatrix} - \frac{\begin{bmatrix} 3+3+0 \\ 1+1+0 \end{bmatrix}}{\begin{bmatrix} 1 \\ -1 \\ 0 \end{bmatrix}}$$

$$= \begin{bmatrix} 4 \\ -2 \\ 5 \end{bmatrix} - \frac{6}{2} \begin{bmatrix} 1 \\ -1 \\ 0 \end{bmatrix}$$

$$= \begin{bmatrix} 1 \\ -1 \\ 5 \end{bmatrix} \quad \begin{bmatrix} 1 \\ -1 \\ 1 \end{bmatrix} \quad \text{Some algebra error somewhere}$$

Don't forget to normalize

$$a_1 = \frac{1}{\sqrt{2}} \begin{bmatrix} 1 \\ -1 \\ 0 \end{bmatrix} \quad a_2 = \frac{1}{\sqrt{2}} \begin{bmatrix} 1 \\ 1 \\ -2 \end{bmatrix} \quad a_3 = \frac{1}{\sqrt{3}} \begin{bmatrix} 1 \\ -1 \\ 1 \end{bmatrix}$$

29

5/7

$$(AB)^{-1} = B^{-1} A^{-1}$$

$$AA^{-1} = A^{-1}A \quad \text{if diagonalizable?}$$

Sym

Sym always means diag...

M has e value λ

M^{-1} $\frac{1}{\lambda}$

$2M$ 2λ

$M - I$ $\lambda - 1$

Don't ~~enlarge~~ enlarge w/ 0s for SVD U and V

instead insert orthonormal basis

for $N(A)$	for V	$\leftarrow n$ -cols
$N(A^T)$	U	$\leftarrow m$ -cols

Quick Flip Thigh

18.06 Quiz 3

Professor Strang

May 6, 2011

Your PRINTED name is _____ 1.

Your Recitation Instructor (and time) is _____ 2.

Instructors: (Hezari)(Pires)(Sheridan)(Yoo) _____ 3.

Please show enough work so we can see your method and give due credit.

1. (a) Find two eigenvalues and eigenvectors of

$$A = \begin{bmatrix} 2 & 3 \\ 0 & 5 \end{bmatrix}$$

$$\lambda = 2, 5$$

$$(2-\lambda)(5-\lambda) - 3 \cdot 0 = 0$$

$$\begin{pmatrix} 1 \\ 0 \end{pmatrix} \begin{pmatrix} 1 \\ 1 \end{pmatrix}$$

$$10 - 7\lambda + \lambda^2 = 0$$

(too small space)

(b) Express any vector $u_0 = \begin{bmatrix} a \\ b \end{bmatrix}$ as a combination of the eigenvectors.

? exponential

$$\begin{pmatrix} c_1 \\ c_2 \end{pmatrix} = S^{-1} u_0 = \begin{pmatrix} 1 & 1 \\ 0 & 1 \end{pmatrix}^{-1} \begin{pmatrix} a \\ b \end{pmatrix} = \begin{pmatrix} a-b \\ b \end{pmatrix}$$

did we ever learn

(c) What is the solution $u(t)$ to $\frac{du}{dt} = Au$ starting from $u(0) = u_0$?

$$e^{\lambda t} \begin{pmatrix} x \end{pmatrix}$$

vector

yes

$$u(t) = c_1 e^{\lambda_1 t} x_1 + c_2 e^{\lambda_2 t} x_2 =$$

(d) Find a formula $u_k =$ _____ for the solution to $u_{k+1} = Au_k$ which starts from that vector u_0 . Set $k = -1$ to find $A^{-1}u_0$.

time is be sure

despite is inverse

2. This problem is about the matrix

$$A = \begin{bmatrix} \sqrt{2} & 1 \\ 0 & \sqrt{2} \end{bmatrix}$$

(a) Find all eigenvectors of A . Exactly why is it impossible to diagonalize A in the form

$$A = SAS^{-1}?$$

ζ (repeat \rightarrow no ζ yes, e -vectors are the same
non real?

(b) Find the matrices U, Σ, V^T in the Singular Value Decomposition $A = U \Sigma V^T$.

Tell me *two orthogonal vectors* v_1, v_2 in the plane so that Av_1 and Av_2 are also orthogonal.

Did before --

(c) Find a matrix B that is similar to A (but different from A).

Show that A and B meet the requirement to be similar (*what is it?*).

ζ guess show $B = M^{-1}AM$

$$M = \begin{bmatrix} 0 & 1 \\ 1 & 0 \end{bmatrix}$$

3. Suppose A is a real m by n matrix.

(a) Prove that the symmetric matrix $A^T A$ has the property $x^T(A^T A)x \geq 0$ for every vector x in R^n . Explain each step in your reason.

Pos de

$$(x^T A^T) (Ax)$$

$$(Ax)^T (Ax)$$

will be ≥ 0

but not for same

reason
as before?

(b) According to part (a), the matrix $A^T A$ is positive semidefinite at least — and possibly positive definite. Under what condition on A is $A^T A$ positive definite?

(c) If $m < n$ prove that $A^T A$ is *not* positive definite.

- Your PRINTED name is _____
 Your Recitation Instructor (and time) is _____
 Instructors: (Hezari)(Pires)(Sheridan)(Yoo)

Please show enough work so we can see your method and give due credit.

1. (a) Find two eigenvalues and eigenvectors of

$$A = \begin{bmatrix} 2 & 3 \\ 0 & 5 \end{bmatrix}$$

$$P(\lambda) = \det(A - \lambda I) = \lambda^2 - 7\lambda + 10 = (\lambda - 2)(\lambda - 5) = 0 \Rightarrow \lambda_1 = 2, \lambda_2 = 5$$

$$N(A - 2I) = N \left(\begin{bmatrix} -1 & 3 \\ 0 & 3 \end{bmatrix} \right) = \text{span} \left\{ \begin{bmatrix} 1 \\ 0 \end{bmatrix} \right\}$$

$$N(A - 5I) = N \left(\begin{bmatrix} -3 & 3 \\ 0 & 0 \end{bmatrix} \right) = \text{span} \left\{ \begin{bmatrix} 1 \\ 1 \end{bmatrix} \right\}$$

- (b) Express any vector $u_0 = \begin{bmatrix} a \\ b \end{bmatrix}$ as a combination of the eigenvectors.

$$\begin{bmatrix} c_1 \\ c_2 \end{bmatrix} = S^{-1} u_0 = \begin{bmatrix} 1 & 1 \\ 0 & 1 \end{bmatrix}^{-1} \begin{bmatrix} a \\ b \end{bmatrix} = \begin{bmatrix} 1 & -1 \\ 0 & 1 \end{bmatrix} \begin{bmatrix} a \\ b \end{bmatrix} = \begin{bmatrix} a-b \\ b \end{bmatrix}$$

So $u_0 = c_1 x_1 + c_2 x_2 = (a-b) \begin{bmatrix} 1 \\ 0 \end{bmatrix} + b \begin{bmatrix} 1 \\ 1 \end{bmatrix}$

- (c) What is the solution $u(t)$ to $\frac{du}{dt} = Au$ starting from $u(0) = u_0$?

$$u(t) = c_1 e^{\lambda_1 t} x_1 + c_2 e^{\lambda_2 t} x_2 = (a-b) e^{2t} \begin{bmatrix} 1 \\ 0 \end{bmatrix} + b e^{5t} \begin{bmatrix} 1 \\ 1 \end{bmatrix}$$

- (d) Find a formula $u_k =$ _____ for the solution to $u_{k+1} = Au_k$ which starts from that vector u_0 . Set $k = -1$ to find $A^{-1} u_0$.

$$u_k = A^k u_0 = c_1 \lambda_1^k x_1 + c_2 \lambda_2^k x_2 = (a-b) 2^k \begin{bmatrix} 1 \\ 0 \end{bmatrix} + b 5^k \begin{bmatrix} 1 \\ 1 \end{bmatrix}$$

$$u_{-1} = (a-b) 2^{-1} \begin{bmatrix} 1 \\ 0 \end{bmatrix} + b 5^{-1} \begin{bmatrix} 1 \\ 1 \end{bmatrix} = \begin{bmatrix} \frac{a-b}{2} & \frac{b}{5} \end{bmatrix}$$

2. This problem is about the matrix

$$A = \begin{bmatrix} \sqrt{2} & 1 \\ 0 & \sqrt{2} \end{bmatrix}$$

- (a) Find all eigenvectors of A . Exactly why is it impossible to diagonalize A in the form $A = SAS^{-1}$? $P(\lambda) = (\lambda - \sqrt{2})(\lambda - \sqrt{2}) = 0 \Rightarrow \lambda_1 = \lambda_2 = \sqrt{2}$ repeated eigenvalues

$$N(A - \sqrt{2}I) = N \left(\begin{bmatrix} 0 & 1 \\ 0 & 0 \end{bmatrix} \right) = \text{span} \left\{ \begin{bmatrix} 1 \\ 0 \end{bmatrix} \right\}$$

There are not enough independent eigenvectors to form an invertible matrix S with eigenvectors as its columns.

- (b) Find the matrices U, Σ, V^T in the Singular Value Decomposition $A = U \Sigma V^T$.

Tell me two orthogonal vectors v_1, v_2 in the plane so that Av_1 and Av_2 are also orthogonal. $\beta = A^T A = \begin{bmatrix} \sqrt{2} & 0 \\ 0 & \sqrt{2} \end{bmatrix} \begin{bmatrix} \sqrt{2} & 1 \\ 0 & \sqrt{2} \end{bmatrix} = \begin{bmatrix} 2 & \sqrt{2} \\ \sqrt{2} & 2 \end{bmatrix}$

$\beta(\lambda) = \lambda^2 - 5\lambda + 4 = (\lambda - 1)(\lambda - 4) = 0 \Rightarrow \lambda_1 = 4, \lambda_2 = 1$ for β .

$\Rightarrow \sigma_1 = \sqrt{\lambda_1} = 2, \sigma_2 = \sqrt{\lambda_2} = 1 \Rightarrow \Sigma = \begin{bmatrix} 2 & 0 \\ 0 & 1 \end{bmatrix}$

$N(\beta - 4I) = N \left(\begin{bmatrix} -2 & \sqrt{2} \\ \sqrt{2} & -2 \end{bmatrix} \right) = \text{span} \left\{ \begin{bmatrix} 1 \\ \sqrt{2} \end{bmatrix} \right\} \Rightarrow v_1 = \frac{1}{\sqrt{3}} \begin{bmatrix} 1 \\ \sqrt{2} \end{bmatrix}$ normalized

$N(\beta - I) = N \left(\begin{bmatrix} 3 & \sqrt{2} \\ \sqrt{2} & 1 \end{bmatrix} \right) = \text{span} \left\{ \begin{bmatrix} \sqrt{2} \\ -1 \end{bmatrix} \right\} \Rightarrow v_2 = \frac{1}{\sqrt{5}} \begin{bmatrix} \sqrt{2} \\ -1 \end{bmatrix}$

$U = \frac{Av_1}{\sigma_1} = \frac{1}{2} \begin{bmatrix} \sqrt{2} \\ 0 \end{bmatrix} = \begin{bmatrix} 1 \\ 0 \end{bmatrix}$

$U_2 = \frac{Av_2}{\sigma_2} = \frac{1}{1} \begin{bmatrix} \sqrt{2} \\ -1 \end{bmatrix} = \begin{bmatrix} \sqrt{2} \\ -1 \end{bmatrix}$

$A = \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix} \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix} \begin{bmatrix} 1 & \sqrt{2} \\ 0 & -1 \end{bmatrix}^T$ because $v_1 \perp v_2$

- (c) Find a matrix B that is similar to A (but different from A).

Show that A and B meet the requirement to be similar (what is it?).
 we say $B \sim A$ if $B = MAM^{-1}$ for some invertible M .
 choose for example $M = \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix}$. (You can choose any M)
 Then $B = \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix} \begin{bmatrix} \sqrt{2} & 1 \\ 0 & \sqrt{2} \end{bmatrix} \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix}^{-1} = \dots = \begin{bmatrix} \sqrt{2} & 0 \\ 0 & \sqrt{2} \end{bmatrix}$
 and $B \neq A$ but $B \sim A$!

3. Suppose A is a real m by n matrix.

(a) Prove that the symmetric matrix $A^T A$ has the property $x^T (A^T A) x \geq 0$ for every vector x in \mathbb{R}^n . Explain each step in your reason.

$$x^T (A^T A) x = (x^T A^T) A x = (Ax)^T A x = (Ax) \cdot (Ax) \geq 0.$$

(b) According to part (a), the matrix $A^T A$ is positive semidefinite at least — and possibly positive definite. Under what condition on A is $A^T A$ positive definite? we want to see under what condition

$$x^T (A^T A) x = 0 \text{ implies } x = 0.$$

So let $x^T A^T A x = 0$. By (a) we get $(Ax) \cdot (Ax) = 0$.

So $Ax = 0$. Hence to get $x = 0$ from $Ax = 0$,

we need $N(A) = \{0\}$ or A must have independent columns.

(c) If $m < n$ prove that $A^T A$ is not positive definite. we use (b) and show that if $m < n$ then $N(A) \neq \{0\}$:

OK! we know that $\dim N(A) = n - r$ where $r = \text{rank}(A)$. But $r \leq m$. So

$\dim N(A) = n - r \geq n - m$. Since $m < n$, $n - m > 0$ therefore: $N(A) \neq \{0\}$.

Your PRINTED name is: _____ 1.

Your recitation number is _____ 2.

3.

1. (40 points) Suppose u is a unit vector in \mathbb{R}^n , so $u^T u = 1$. This problem is about the n by n symmetric matrix $H = I - 2uu^T$.

(a) Show directly that $H^2 = I$. Since $H = H^T$, we now know that H is not only symmetric but also _____.

(b) One eigenvector of H is u itself. Find the corresponding eigenvalue.

(c) If v is any vector perpendicular to u , show that v is an eigenvector of H and **find the eigenvalue**. With all these eigenvectors v , that eigenvalue must be repeated how many times? Is H **diagonalizable**? Why or why not?

(d) Find the diagonal entries H_{11} and H_{ii} in terms of u_1, \dots, u_n . Add up $H_{11} + \dots + H_{nn}$ and separately add up the eigenvalues of H .

2. (30 points) Suppose A is a positive definite symmetric n by n matrix.

(a) How do you know that A^{-1} is also positive definite? (We know A^{-1} is symmetric. I just had an e-mail from the International Monetary Fund with this question.)

(b) Suppose Q is any **orthogonal** n by n matrix. How do you know that $Q A Q^T = Q A Q^{-1}$ is positive definite? Write down which test you are using.

(c) Show that the block matrix

$$B = \begin{bmatrix} A & A \\ A & A \end{bmatrix}$$

is positive **semidefinite**. How do you know B is not positive definite?

3. (30 points) This question is about the matrix

$$A = \begin{bmatrix} 0 & -1 \\ 4 & 0 \end{bmatrix}$$

(a) Find its eigenvalues and eigenvectors.

Write the vector $u(0) = \begin{bmatrix} 2 \\ 0 \end{bmatrix}$ as a combination of those eigenvectors.

(b) Solve the equation $\frac{du}{dt} = Au$ starting with the same vector $u(0)$ at time $t = 0$.

In other words: the solution $u(t)$ is what combination of the eigenvectors of A ?

(c) Find the 3 matrices in the Singular Value Decomposition $A = U \Sigma V^T$ in two steps.

–First, compute V and Σ using the matrix $A^T A$.

–Second, find the (orthonormal) columns of U .

- Your PRINTED name is: _____ 1.
 Your recitation number is _____ 2.
 _____ 3.

1. (40 points) Suppose u is a unit vector in \mathbb{R}^n , so $u^T u = 1$. This problem is about the n by n symmetric matrix $H = I - 2u u^T$.

(a) Show directly that $H^2 = I$. Since $H = H^T$, we now know that H is not only symmetric but also _____
 [Solution] Explicitly, we find $H^2 = (I - 2uu^T)^2 = I^2 - 4uu^T + 4uu^T uu^T$ (2 points):
 since $u^T u = 1$, $H^2 = I$ (3 points). Since $H = H^T$, we also have $H^T H = I$, implying that H is an orthogonal (or unitary) matrix.

(b) One eigenvector of H is u itself. Find the corresponding eigenvalue.
 [Solution] Since $Hu = (I - 2uu^T)u = u - 2uu^T u = u - 2u = -u$, $\lambda = -1$.

(c) If v is any vector perpendicular to u , show that v is an eigenvector of H and find the eigenvalue. With all these eigenvectors v , that eigenvalue must be repeated how many times? Is H diagonalizable? Why or why not?

[Solution] For any vector v orthogonal to u (i.e. $u^T v = 0$), we have $Hv = (I - 2uu^T)v = v - 2uu^T v = v$, so the associated λ is 1. The orthogonal complement to the space spanned by u has dimension $n-1$, so there is a basis of $(n-1)$ orthonormal eigenvectors with this eigenvalue. Adding in the eigenvector u , we find that H is diagonalizable.

(d) Find the diagonal entries H_{11} and H_{ii} in terms of u_1, \dots, u_n . Add up $H_{11} + \dots + H_{nn}$ and separately add up the eigenvalues of H .

[Solution] Since the i th diagonal entry of uu^T is u_i^2 , the i th diagonal entry of H is $H_{ii} = 1 - 2u_i^2$ (3 points). Summing these together gives $\sum_{i=1}^n H_{ii} = n - 2 \sum_{i=1}^n u_i^2 = n - 2$ (3 points). Adding up the eigenvalues of H also gives $\sum \lambda_i = (1) - 1 + (n-1)(1) = n - 2$ (4 points).

2. (30 points) Suppose A is a positive definite symmetric n by n matrix.

(a) How do you know that A^{-1} is also positive definite? (We know A^{-1} is symmetric. I just had an e-mail from the International Monetary Fund with this question.)

[Solution] Since a matrix is positive-definite if and only if all its eigenvalues are positive (5 points), and since the eigenvalues of A^{-1} are simply the inverses of the eigenvalues of A , A^{-1} is also positive definite (the inverse of a positive number is positive) (5 points).

(b) Suppose Q is any orthogonal n by n matrix. How do you know that $QAQ^T = QAQ^{-1}$ is positive definite? Write down which test you are using.

[Solution] Using the energy text ($x^T Ax > 0$ for nonzero x), we find that $x^T QAQ^T x = (Q^T x)^T A(Q^T x) > 0$ for all nonzero x as well (since Q is invertible). Using the positive eigenvalue test, since A is similar to QAQ^{-1} and similar matrices have the same eigenvalues, QAQ^{-1} also has all positive eigenvalues. (5 points for test, 5 points for application)

(c) Show that the block matrix

$$B = \begin{bmatrix} A & A \\ A & A \end{bmatrix}$$

is positive semidefinite. How do you know B is not positive definite?

[Solution] First, since B is singular, it cannot be positive definite (it has eigenvalues of 0). However, the pivots of B are the pivots of A in the first n rows followed by 0s in the remaining rows, so by the pivot test, B is still semi-definite. Similarly, the first n upper-left determinants of B are the same as those of A , while the remaining ones are 0s, giving another proof. Finally, given a nonzero vector

$$u = \begin{bmatrix} x \\ y \end{bmatrix}$$

where x and y are vectors in \mathbb{R}^n , one has $u^T B u = (x+y)^T A(x+y)$ which is nonnegative (and zero when $x + y = 0$).

3. (30 points) This question is about the matrix

$$A = \begin{bmatrix} 0 & -1 \\ 4 & 0 \end{bmatrix}$$

(a) Find its eigenvalues and eigenvectors.

Write the vector $u(0) = \begin{bmatrix} 2 \\ 0 \end{bmatrix}$ as a combination of those eigenvectors.

[Solution] Since $\det(A - \lambda I) = \lambda^2 + 4$, the eigenvalues are $2i, -2i$ (4 points). Two associated eigenvectors are $[1 - 2i]^T, [1 2i]^T$, though there are many other choices (4 points). $u(0)$ is just the sum of these two vectors (2 points).

(b) Solve the equation $\frac{du}{dt} = Au$ starting with the same vector $u(0)$ at time $t = 0$.

In other words: the solution $u(t)$ is what combination of the eigenvectors of A ?

[Solution] One simply adds in factors of $e^{\lambda t}$ to each term, giving

$$u(t) = e^{2it} \begin{bmatrix} 1 \\ -2i \end{bmatrix} + e^{-2it} \begin{bmatrix} 1 \\ 2i \end{bmatrix}$$

(c) Find the 3 matrices in the Singular Value Decomposition $A = U\Sigma V^T$ in two steps.

-First, compute V and Σ using the matrix $A^T A$.

-Second, find the (orthonormal) columns of U .

[Solution] Note that $A^T A = V\Sigma^T U^T U \Sigma V^T = V\Sigma^2 V^T$, so the diagonal entries of Σ are simply the positive roots of the eigenvalues of

$$A^T A = \begin{bmatrix} 0 & 4 \\ -1 & 0 \end{bmatrix} \begin{bmatrix} 0 & -1 \\ 4 & 0 \end{bmatrix} = \begin{bmatrix} 16 & 0 \\ 0 & 1 \end{bmatrix}$$

i.e. $\sigma_1 = 4, \sigma_2 = 1$. Since $A^T A$ is already diagonal, V is the identity matrix. The columns of U should satisfy $Au_1 = \sigma_1 v_1, Au_2 = \sigma_2 v_2$: by inspection, one obtains

$$u_1 = \begin{bmatrix} 0 \\ 1 \end{bmatrix}, u_2 = \begin{bmatrix} -1 \\ 0 \end{bmatrix}, U = \begin{bmatrix} 0 & -1 \\ 1 & 0 \end{bmatrix}$$

18.06

Professor Johnson

Quiz 3

May 1, 2009

Your PRINTED name is: _____

Please circle your recitation:

Grading

(R01)	M2	2-314	Qian Lin	_____
(R02)	M3	2-314	Qian Lin	1
(R03)	T11	2-251	Martina Balagovic	_____
(R04)	T11	2-229	Inna Zakharevich	2
(R05)	T12	2-251	Martina Balagovic	_____
(R06)	T12	2-090	Ben Harris	3
(R07)	T1	2-284	Roman Bezrukavnikov	_____
(R08)	T1	2-310	Nick Rozenblyum	4
(R09)	T2	2-284	Roman Bezrukavnikov	_____
				Total:

- 1 (20 pts.) For each part, give as much information as possible about the eigenvalues of the matrix A described in that part. (Each part describes a *different* matrix A . A may be complex.)
- (a) The recurrence $\mathbf{u}_{k+1} = A\mathbf{u}_k$ has a solution where $\|\mathbf{u}_k\| \rightarrow 0$ as $k \rightarrow \infty$ for one initial vector \mathbf{u}_0 , but also has a solution with $\|\mathbf{u}_k\| \rightarrow \infty$ as $k \rightarrow \infty$ for a *different* choice of the initial vector \mathbf{u}_0 .
 - (b) The equation $(A^2 - 4I)\mathbf{x} = \mathbf{b}$ has no solution for some right-hand side \mathbf{b} .
 - (c) $A = e^{B^T B}$ for some real matrix B with full column rank.
 - (d) $A = B^T B$ for a 4×3 real matrix B , and the matrix BB^T has eigenvalues $\lambda = 3, 2, 1, 0$. (Hint: think about the SVD of B .)

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2 (20 pts.) You are given the matrix

$$A = \begin{pmatrix} 0.5 & 0.2 & 0.2 \\ 0.1 & 0.5 & 0.5 \\ 0.4 & 0.3 & 0.3 \end{pmatrix}.$$

- (i) What are the eigenvalues of A ? [*Hint*: Very little calculation required! You should be able to see two eigenvalues by inspection of the form of A , and the third by an easy calculation. You *shouldn't* need to compute $\det(A - \lambda I)$ unless you really want to do it the hard way.]
- (ii) The vector $\mathbf{u}(t)$ solves the system

$$\frac{d\mathbf{u}}{dt} = A\mathbf{u}$$

for some initial condition $\mathbf{u}(0)$. If you are told that $\mathbf{u}(t)$ approaches some constant vector as $t \rightarrow \infty$, give as much true information as possible regarding the initial condition $\mathbf{u}(0)$.

[*Note*: be sure you understand that this is *not the same thing* as solving the recurrence $\mathbf{u}_{k+1} = A\mathbf{u}_k$! Imagine how you would find $\mathbf{u}(t)$ if you knew what $\mathbf{u}(0)$ was.]

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- 3 (10 pts.) The 3×3 matrix A has three independent eigenvectors \mathbf{v}_1 , \mathbf{v}_2 , and \mathbf{v}_3 with corresponding eigenvalues λ_1 , λ_2 , and λ_3 (that is, $A\mathbf{v}_i = \lambda_i\mathbf{v}_i$ for $i = 1, 2, 3$).

If

$$\mathbf{b} = c_1\mathbf{v}_1 + c_2\mathbf{v}_2 + c_3\mathbf{v}_3$$

for some coefficients c_1 , c_2 , and c_3 , then write (in terms of λ_i , c_i , and \mathbf{v}_i) a formula for the solution \mathbf{x} of

$$A^2\mathbf{x} + 2A\mathbf{x} - 3I\mathbf{x} = \mathbf{b}$$

(you can assume that a solution exists for any \mathbf{b}).

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4 (15 pts.) A is a 3×3 real-symmetric matrix. Two of its eigenvalues are $\lambda_1 = 1$ and $\lambda_2 = -1$ with eigenvectors $\mathbf{v}_1 = (1, 1, 1)$ and $\mathbf{v}_2 = (1, -1, 0)$, respectively. The third eigenvalue is $\lambda_3 = 0$.

(I) Give an eigenvector \mathbf{v}_3 for the eigenvalue λ_3 . (*Hint: what must be true of \mathbf{v}_1 , \mathbf{v}_2 , and \mathbf{v}_3 ?*)

(II) Using your result from (I), write the matrix e^A as the product of three matrices, and explicitly give the three matrices. (You need not work out the arithmetic, but your answer should contain no matrix inverses or matrix exponentials. *If you find yourself doing a lot of arithmetic, you are forgetting a useful property of this matrix!*)

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18.06 Spring 2009 Exam 3 Practice

General comments

Exam 2 covers the first 31 lectures of 18.06, mainly focusing on lectures 19–31 (eigenproblems). The topics covered are (very briefly summarized):

1. All of the topics from exams 1 and 2, although of course these are not the focus of the exam.
2. Determinants: their properties, how to compute them (simple formulas for 2×2 and 3×3 , usually by elimination for matrices $> 3 \times 3$), their relationship to linear equations (zero determinant = singular), their use for eigenvalue problems.
3. Eigenvalues and eigenvectors: their definition $A\vec{x} = \lambda\vec{x}$, their properties, the fact that for an eigenvector the matrix (or *any function of the matrix*) acts just like a number. Computing λ from the characteristic polynomial $\det(A - \lambda I)$ and \vec{x} from $N(A - \lambda I)$; zero eigenvalues $\lambda = 0$ just correspond to $N(A)$. Understand (from the definition) why, if A has an eigenvalue λ , then A^k has an eigenvalue λ^k , αA has an eigenvalue $\alpha\lambda$, and $A + \beta I$ has an eigenvalue $\lambda + \beta$, all with the *same* eigenvector.
4. Diagonalization $A = SAS^{-1}$: where it comes from, its use in understanding properties of matrices and eigenvalues. **The basic idea that, to solve a problem involving A , you first expand your vector in the basis of the eigenvectors (S), then for each eigenvector you treat A as just a number λ , then at the end you add up the solutions.**
5. Similar matrices: A and $B = MAM^{-1}$ have the same eigenvalues for any invertible matrix M , and if $A\vec{x} = \lambda\vec{x}$ then $B\vec{y} = \lambda\vec{y}$ for $\vec{y} = M\vec{x}$. Similar matrices have the same trace (sum of the eigenvalues) and determinant (product of the eigenvalues).
6. Using eigenvalues/eigenvectors to solve problems involving matrix powers, such as linear recurrences (e.g. Fibonacci). Multiplying by A many times tends towards the eigenvector for the largest $|\lambda|$. Markov matrices: what the defining properties are, and the consequences (a steady state with $\lambda = 1$, all other solutions decay away, the sum of the components of the vector is conserved, a unique steady state if all entries of the matrix are > 0). $A^n = SA^nS^{-1}$.
7. Using eigenvalues/eigenvectors to solve linear systems of differential equations $\frac{d\vec{u}}{dt} = A\vec{u}$ with initial conditions $\vec{u}(0)$. Practical scheme: expand $\vec{u}(0)$ in eigenvector basis and multiply each term by $e^{\lambda t}$. Formal solution: $e^{At}\vec{u}(0)$, and the meaning of the matrix exponential $e^A = Se^\Lambda S^{-1}$ and how to compute it and manipulate it.
8. If $A = A^T$ (real-symmetric), then the eigenvalues are real and the eigenvectors are orthogonal (or can be chosen orthogonal), and A is diagonalizable as $A = Q\Lambda Q^T$ for an orthogonal Q . If $A = B^T B$ where B has full column rank, then A is positive definite: all $\lambda > 0$ and all pivots > 0 and $\vec{y}^T A \vec{y} > 0$ for any $\vec{y} \neq 0$; connection to minimization problems (like least-squares).
9. Complex matrices, for which we replace \vec{x}^T and A^T by $\vec{x}^H = \overline{\vec{x}^T}$ and $A^H = \overline{A^T}$ (and why). What to do if you get a complex λ : consequences for matrix powers (recurrence relations) and differential equations are oscillating solutions, using $e^{i\theta} = \cos \theta + i \sin \theta$.
10. Defective matrices and generalized eigenvectors: what to do if A is not diagonalizable, especially for a practical problem like $A^k \vec{u}$ or $e^{At} \vec{u}$. (Note that real-symmetric, real-orthogonal, Hermitian, and unitary matrices are never defective, nor are $n \times n$ matrices with n distinct eigenvalues.)

11. Singular value decomposition $A = U\Sigma V^T$ and their relationship to eigenvectors/eigenvalues of $A^T A$ and AA^T .

The central concept from this part of the course is highlighted in boldface above. Once you have an eigenvector, *any* operation involving the matrix just becomes that operation with the single number λ . And single numbers are easy to handle. So, we try to find the eigenvectors and then express every vector in that basis (aside from rare defective cases), at which point problems become easy (ideally)! Also, you should be able to recognize and reason about how and why special forms of the matrix A (symmetric, Markov, singular, etcetera) give you additional information about the eigenvectors and eigenvalues.

Defective matrices and SVDs will see at most limited coverage on the exam, perhaps one part of a problem each, at most.

Some practice problems

The 18.06 web site has exams from previous terms that you can download, with solutions. I've listed a few practice exam problems that I like below, but there are plenty more to choose from. The exam will consist of 3 or 4 questions (perhaps with several parts each), and you will have one hour. You can find the solutions to these problems on the 18.06 web site (in the section for old exams/psets).

1. (Fall 2002 exam 3.) (a) What are the eigenvalues of the 5×5 matrix $A = \begin{pmatrix} 1 & 1 & 1 & 1 & 1 \\ 1 & 1 & 1 & 1 & 1 \\ 1 & 1 & 1 & 1 & 1 \\ 1 & 1 & 1 & 1 & 1 \\ 1 & 1 & 1 & 1 & 1 \end{pmatrix}$? Please look

at A , not at $\det(A - \lambda I)$. (b) Solve $\frac{d\vec{u}}{dt} = A\vec{u}$ starting from $\vec{u}(0) = (0, 1, 1, 1, 2)^T$. (First split $\vec{u}(0)$ as the sum of two eigenvectors of A .) (c) Using part (a), what are the *eigenvalues* and *trace* and *determinant* of the matrix B which is the same as A except that it has zeros on its diagonal?

2. (Fall 2002 exam 3.) (a) if A is similar to B show that e^A is similar to e^B . (Hint: first write down the definitions of "similar" and e^A .) (b) If A has 3 eigenvalues $\lambda = 0, 2, 4$, find the eigenvalues of e^A . (c) Explain this connection with determinants: $\det(e^A) = e^{\text{trace of } A}$.
3. (Fall 2002 exam 3.) Companies in the US, Asia, and Europe have assets of \$12 trillion. At the start, \$6 trillion are in the US and \$6 trillion are in Europe. Each year, half the US money stays home, 1/4 each goes to Asia and Europe. For Asia and Europe, half stays home and half is sent to the US, hence

$$\begin{pmatrix} \text{US} \\ \text{Asia} \\ \text{Europe} \end{pmatrix}_{\text{year } k+1} = \begin{pmatrix} .5 & .5 & .5 \\ .25 & .5 & 0 \\ .25 & 0 & .5 \end{pmatrix} \begin{pmatrix} \text{US} \\ \text{Asia} \\ \text{Europe} \end{pmatrix}_{\text{year } k}$$

(a) The eigenvalues and eigenvectors of this *singular* matrix A are what? (b) The limiting distribution of the \$12 trillion after many many years is US=?, Asia=?, Europe=?

4. (Fall 2002 exam 2.) If you know that $\det A = 6$, what is $\det B$ for B given by:

$$A = \begin{pmatrix} \text{row 1} \\ \text{row 2} \\ \text{row 3} \end{pmatrix} \quad B = \begin{pmatrix} \text{row 3} + \text{row 2} + \text{row 1} \\ \text{row 2} + \text{row 1} \\ \text{row 1} \end{pmatrix}$$

5. (Spring 2004 exam 3.) For the symmetric matrix $A = \begin{pmatrix} 1 & 0 & 1 \\ 0 & 1 & -1 \\ 1 & -1 & 0 \end{pmatrix}$, you are given that one of the eigenvalues is $\lambda = 1$ with a line of eigenvectors $\vec{x} = (c, c, 0)$. (a) That line is the nullspace of what matrix constructed from A ? (b) Find (in any way) the other two eigenvalues of A and two corresponding eigenvectors. (c) The diagonalization $A = SAS^{-1}$ has an especially nice form because $A = A^T$. Write all entries in the nice symmetric diagonalization of A . (d) Give a reason why e^A is or is not a symmetric positive-definite matrix.

6. (Spring 2004 exam 3.) (a) Find the eigenvalues and eigenvectors (depending on c) of $A = \begin{pmatrix} 0.3 & c \\ 0.7 & 1-c \end{pmatrix}$. For which c is the matrix A *not diagonalizable*?¹ (b) What is the largest range of (real) values of c so that A^n approaches a limiting matrix A^∞ as $n \rightarrow \infty$? (c) What is that limit of A^n (still depending on c)?
7. (Spring 2005 exam 3.) (a) Find all the eigenvalues and all the eigenvectors of the following A . *It is a symmetric Markov matrix with a repeated eigenvalue.*

$$A = \begin{pmatrix} 2/4 & 1/4 & 1/4 \\ 1/4 & 2/4 & 1/4 \\ 1/4 & 1/4 & 2/4 \end{pmatrix}.$$

- (b) Find the limit of A^k as $k \rightarrow \infty$. (c) Choose any positive numbers r , s , and t so that $A - rI$ is positive-definite, $A - sI$ is indefinite, and $A - tI$ is negative definite. (d) Suppose that this $A = B^T B$. What are the singular values σ_i in the SVD of B ?
8. (Spring 2005 exam 3.) (a) Complete this 2×2 matrix A , depending on the real number a , so that its eigenvalues are $\lambda = 1$ and $\lambda = -1$. $A = \begin{pmatrix} a & 1 \\ ? & ? \end{pmatrix}$. (b) How do you know that A has two independent eigenvectors? (c) Which choices of a give orthogonal eigenvectors and which don't?
9. (Spring 2005 exam 3.) Suppose that the 3×3 matrix A has 3 independent eigenvectors $\vec{x}_{1,2,3}$ and corresponding eigenvalues $\lambda_{1,2,3}$. (The λ 's might not be different.) (a) Describe the general form of every solution $\vec{u}(t)$ to the differential equation $\frac{d\vec{u}}{dt} = A\vec{u}$ in terms of the λ 's and \vec{x} 's. (The answer $e^{At}\vec{u}(0)$ is not sufficient.) (b) Starting from any vector \vec{u}_0 , suppose $\vec{u}_{k+1} = A\vec{u}_k$. What are the conditions on the \vec{x} 's and λ 's to guarantee that $\vec{u}_k \rightarrow 0$ as $k \rightarrow \infty$? Why?
10. (Fall 2005 exam 3.) This 4×4 matrix H is a special matrix called a "Hadamard matrix:"

$$H = \begin{pmatrix} 1 & 1 & 1 & 1 \\ 1 & -1 & 1 & -1 \\ 1 & 1 & -1 & -1 \\ 1 & -1 & -1 & 1 \end{pmatrix}.$$

It has two key properties: $H^T = H$, and $H^2 = 4I$. (a) Figure out the eigenvalues of H and explain your reasoning. (b) Figure out H^{-1} and $\det H$. Explain your reasoning. (c) This matrix S contains three eigenvectors of H . Find a 4-th eigenvector \vec{x}_4 and explain your reasoning.

$$S = \begin{pmatrix} 1 & 1 & 0 & ? \\ 1 & 0 & -1 & ? \\ 1 & 0 & 1 & ? \\ -1 & 1 & 0 & ? \end{pmatrix}.$$

- (d) Find the solution to $d\vec{u}/dt = H\vec{u}$ given that $\vec{u}(0)$ is the 3rd column of S .
11. (Fall 2005 exam 3.) Suppose A is a 3×3 symmetric matrix with eigenvalues 2, 5, 7 and corresponding eigenvectors \vec{x}_1 , \vec{x}_2 , and \vec{x}_3 . (a) Suppose \vec{x} is a linear combination $\vec{x} = c_1\vec{x}_1 + c_2\vec{x}_2 + c_3\vec{x}_3$. Find $A\vec{x}$. Now find $\vec{x}^T A\vec{x}$ using the symmetry of A . Explain why $\vec{x}^T A\vec{x} > 0$ unless $\vec{x} = 0$.

12. (Fall 2006 exam 3.) (a) Find all three eigenvalues of A , and an eigenvector matrix S . $A = \begin{pmatrix} -1 & 2 & 4 \\ 0 & 0 & 5 \\ 0 & 0 & 1 \end{pmatrix}$. (b)

Explain why $A^{1001} = A$. Is $A^{1000} = I$? (c) The matrix $A^T A$ for this A is $A^T A = \begin{pmatrix} 1 & -2 & -4 \\ -2 & 4 & 8 \\ -4 & 8 & 42 \end{pmatrix}$. How many

¹The solutions are a little tricky. A is *not* a Markov matrix because c may be < 0 . However, its columns sum to 1, and that was enough to give us an eigenvalue $\lambda = 1$ in our analysis of Markov matrices. In the non-diagonalizable case, the solution's formula for part (b) is incorrect. As we know from lecture 30, for a repeated eigenvalue $\lambda = 1$ that is defective, there is a term in A^n that goes as 1^n and another term that goes as $n1^{n-1}$. Since the latter blows up, the defective case does not have a finite A^∞ limit.

eigenvalues of $A^T A$ are positive? zero? negative? Does $A^T A$ have the same eigenvectors as A ? (Don't compute anything, but explain your answers.)

13. (Fall 2006 exam 3.) Suppose the $n \times n$ matrix A has n orthonormal eigenvectors $\vec{q}_1, \dots, \vec{q}_n$ and n positive eigenvalues $\lambda_1, \dots, \lambda_n$. That is, $A\vec{q}_j = \lambda_j\vec{q}_j$. (a) What are the eigenvalues and eigenvectors of A^{-1} ? (b) Any vector \vec{b} can be written as a combination of the eigenvectors $\vec{b} = c_1\vec{q}_1 + c_2\vec{q}_2 + \dots + c_n\vec{q}_n$ for some coefficients c_j . What is a quick formula for c_1 using the orthogonality of the \vec{q} 's? (c) The solution to $A\vec{x} = \vec{b}$ is also a combination of the eigenvectors $A^{-1}\vec{b} = d_1\vec{q}_1 + d_2\vec{q}_2 + \dots + d_n\vec{q}_n$. What is a quick formula for d_1 . (You can write it in terms of the c 's even if you didn't answer part b.)
14. (Fall 2007 exam 3.) Suppose that we form a sequence of real numbers f_k defined by the recurrence relation $f_{k+1} = f_k - f_{k-1} + f_{k-2}$, starting with the initial numbers $f_0 = 2$, $f_1 = 1$, and $f_2 = 2$. (a) Define a 3-component vector $\vec{g}_k = (f_k, f_{k-1}, f_{k-2})$ and a 3×3 matrix A so that $\vec{g}_{k+1} = A\vec{g}_k$. (b) If you constructed A correctly, the three eigenvalues should be 1 and $\pm i$, and the latter two eigenvectors should be $(-1, \pm i, 1)$. Check that you have these $\pm i$ eigenvalues and eigenvectors, and find the $\lambda = 1$ eigenvector. (c) Give an explicit formula for f_k for any k (formulas involving A^k are not acceptable; elementary arithmetic and powers of complex numbers only). (d) Is there any choice of initial conditions (f_0 , f_1 , and f_2) that will make $|f_k|$ diverge as $k \rightarrow \infty$? Explain.
15. (Fall 2007 exam 3.) Some 3×3 real matrix A has eigenvalues $\lambda_1 = 0$, $\lambda_2 = 1$, and $\lambda_3 = 2$, with corresponding eigenvectors $\vec{x}_1 = (1, 0, 0)$, $\vec{x}_2 = (0, 1, 2)$, and $\vec{x}_3 = (0, 1, 1)$. (a) Give a basis for the nullspace $N(A)$, the column space $C(A)$, and the row space $C(A^T)$. (b) Find *all* solutions (the complete solution) \vec{x} to $A\vec{x} = \vec{x}_2 - 3\vec{x}_3$. (c) Is A real-symmetric, orthogonal, Markov, or none of the above?

18.06 Quiz 3 Solution

Hold on Friday, 1 May 2009 at 11am in Walker Gym.
Total: 65 points.

Problem 1:

For each part, give as much information as possible about the eigenvalues of the matrix A described in that part. (Each part describes a *different* matrix A . A may be complex.)

- (a) The recurrence $\mathbf{u}_{k+1} = A\mathbf{u}_k$ has a solution where $\|\mathbf{u}_k\| \rightarrow 0$ as $k \rightarrow \infty$ for one initial vector \mathbf{u}_0 , but also has a solution with $\|\mathbf{u}_k\| \rightarrow \infty$ as $k \rightarrow \infty$ for a *different* choice of the initial vector \mathbf{u}_0 .
- (b) The equation $(A^2 - 4I)\mathbf{x} = \mathbf{b}$ has no solution for some right-hand side \mathbf{b} .
- (c) $A = e^{B^T B}$ for some real matrix B with full column rank.
- (d) $A = B^T B$ for a 4×3 real matrix B , and the matrix BB^T has eigenvalues $\lambda = 3, 2, 1, 0$. (Hint: think about the SVD of B .)

Solution (20 points = 5+5+5+5)

(a) (There was a bug in this problem: in the first condition, we should have required the initial vector \mathbf{u}_0 to be nonzero.) The first condition implies that A has an eigenvalue with absolute value $|\lambda| < 1$. The second condition implies that either A has an eigenvalue with absolute value $|\lambda| > 1$, or A is defective for 2 eigenvalues λ with $|\lambda| = 1$.

(b) The condition says that $A^2 - 4I$ is singular. But we know that, if $\lambda_1, \dots, \lambda_n$ are eigenvalues of A , then the eigenvalues of $A^2 - 4I$ are $\lambda_1^2 - 4, \dots, \lambda_n^2 - 4$. The condition $A^2 - 4I$ being singular says that one of $\lambda_i^2 - 4$ is zero, and hence $\lambda_i = 2$ or -2 . That is to say A has an eigenvalue 2 or -2 .

(c) Since B has full column rank, the eigenvalues of $B^T B$ are positive real numbers λ_i . Hence, we know $A = e^{B^T B}$ has eigenvalues e^{λ_i} ; they are real numbers bigger than 1.

(d) Since BB^T and $B^T B$ have the same set of nonzero eigenvalues. So $B^T B$ must have eigenvalues 3, 2, 1. Moreover, since B is a 4×3 matrix, $B^T B$ is a 3×3 matrix. Hence, 3, 2, 1 are exactly all the eigenvalues.

Problem 2: You are given the matrix

$$A = \begin{pmatrix} 0.5 & 0.2 & 0.2 \\ 0.1 & 0.5 & 0.5 \\ 0.4 & 0.3 & 0.3 \end{pmatrix}.$$

- (i) What are the eigenvalues of A ? [Hint: Very little calculation required! You should be able to see two eigenvalues by inspection of the form of A , and the third by an easy calculation. You *shouldn't* need to compute $\det(A - \lambda I)$ unless you really want to do it the hard way.]

- (ii) The vector $\mathbf{u}(t)$ solves the system

$$\frac{d\mathbf{u}}{dt} = A\mathbf{u}$$

for some initial condition $\mathbf{u}(0)$. If you are told that $\mathbf{u}(t)$ approaches some constant vector as $t \rightarrow \infty$, give as much true information as possible regarding the initial condition $\mathbf{u}(0)$.

[Note: be sure you understand that this is *not the same thing* as solving the recurrence $\mathbf{u}_{k+1} = A\mathbf{u}_k$! Imagine how you would find $\mathbf{u}(t)$ if you knew what $\mathbf{u}(0)$ was.]

Solution (20 points = 10+10)

(i) First, the last two columns of A are the same. Hence A is singular and it must have an eigenvalue $\lambda_1 = 0$. Also, we observe that A is a Markov matrix. This means that $\lambda_2 = 1$ is an eigenvalue of A . Finally, we know the trace of A is the sum of its three eigenvalues. So, $\text{Tr}(A) = 0.5 + 0.5 + 0.3 = 1.3$ and the last eigenvalue is $\lambda_3 = 1.3 - 1 - 0 = 0.3$.

(ii) We can write $\mathbf{u}(0) = c_1 \mathbf{v}_1 + c_2 \mathbf{v}_2 + c_3 \mathbf{v}_3$ using three eigenvectors $\mathbf{v}_1, \mathbf{v}_2, \mathbf{v}_3$, which correspond to $\lambda_1 = 0, \lambda_2 = 1, \text{ and } \lambda_3 = 0.3$, respectively. We know that this system has solution $\mathbf{u}(t) = c_1 \mathbf{v}_1 + c_2 e^t \mathbf{v}_2 + c_3 e^{0.3t} \mathbf{v}_3$. So, if either one of c_2 and c_3 is nonzero, the system would blow up as $t \rightarrow \infty$. Therefore, the only possibility for $\mathbf{u}(t)$ to approach some constant is to have $c_2 = c_3 = 0$, that is to say that $\mathbf{u}(0)$ is a multiple of the eigenvector $\mathbf{v}_1 = (0, -1, 1)^T$. In this case, $\mathbf{u}(t) = \mathbf{u}(0) = c_1 \mathbf{v}_1$ is a constant.

Problem 3:

The 3×3 matrix A has three independent eigenvectors $v_1, v_2,$ and v_3 with corresponding eigenvalues $\lambda_1, \lambda_2,$ and λ_3 (that is, $Av_i = \lambda_i v_i$ for $i = 1, 2, 3$).

If

$$b = c_1 v_1 + c_2 v_2 + c_3 v_3$$

for some coefficients $c_1, c_2,$ and $c_3,$ then write (in terms of $\lambda_i, c_i,$ and v_i) a formula for the solution x of

$$A^2 x + 2Ax - 3Ix = b$$

(you can assume that a solution exists for any b).

Solution (10 points)

Using the eigenvalues $v_1, v_2, v_3,$ we have

$$\begin{aligned} x &= (A^2 + 2A - 3I)^{-1}b \\ &= (A^2 + 2A - 3I)^{-1}(c_1 v_1 + c_2 v_2 + c_3 v_3) \\ &= \frac{c_1}{\lambda_1^2 + 2\lambda_1 - 3} v_1 + \frac{c_2}{\lambda_2^2 + 2\lambda_2 - 3} v_2 + \frac{c_3}{\lambda_3^2 + 2\lambda_3 - 3} v_3. \end{aligned}$$

Problem 4: A is a 3×3 real-symmetric matrix. Two of its eigenvalues are $\lambda_1 = 1$ and $\lambda_2 = -1$ with eigenvectors $v_1 = (1, 1, 1)^T$ and $v_2 = (1, -1, 0)^T$, respectively. The third eigenvalue is $\lambda_3 = 0$.

(I) Give an eigenvector v_3 for the eigenvalue λ_3 . (*Hint:* what must be true of $v_1, v_2,$ and v_3 ?)

(II) Using your result from (I), write the matrix e^A as the product of three matrices, and explicitly give the three matrices. (You need not work out the arithmetic, but your answer should contain no matrix inverses or matrix exponentials. If you find yourself doing a lot of arithmetic, you are forgetting a useful property of this matrix!)

Solution (15 points = 7+8)

(I) For a real-symmetric matrix, its eigenvectors are orthogonal to each other. So, by inspection, in order for v_3 to be perpendicular to v_2 , we need its first two components same. Hence, we should take v_3 to be $(1, 1, -2)^T$. To easy the second part, we can normalize the eigenvectors

$$q_1 = v_1 / \|v_1\| = (1, 1, 1)^T / \sqrt{3},$$

$$q_2 = v_2 / \|v_2\| = (1, -1, 0)^T / \sqrt{2},$$

$$q_3 = v_3 / \|v_3\| = (1, 1, -2)^T / \sqrt{6}.$$

Alternatively, we can use Gram-Schmidt to find (a multiple of) v_3 as follows. We start with $v = (1, 0, 0)$,

$$v_3 = v - (v \cdot q_1)q_1 - (v \cdot q_2)q_2 = (1, 0, 0)^T - \frac{1}{3}(1, 1, 1)^T - \frac{1}{2}(1, -1, 0)^T = \left(\frac{1}{6}, \frac{1}{6}, -\frac{1}{3}\right)^T.$$

(II) We can write

$$\begin{aligned} A &= SAS^{-1} = Q\Lambda Q^T \\ &= \begin{pmatrix} 1/\sqrt{3} & 1/\sqrt{2} & 1/\sqrt{6} \\ 1/\sqrt{3} & -1/\sqrt{2} & 1/\sqrt{6} \\ 1/\sqrt{3} & 0 & -2/\sqrt{6} \end{pmatrix} \begin{pmatrix} 1 & 0 & 0 \\ 0 & -1 & 0 \\ 0 & 0 & 0 \end{pmatrix} \begin{pmatrix} 1/\sqrt{3} & 1/\sqrt{3} & 1/\sqrt{3} \\ 1/\sqrt{2} & -1/\sqrt{2} & 1/\sqrt{6} \\ -2/\sqrt{6} & 1/\sqrt{6} & -2/\sqrt{6} \end{pmatrix}. \end{aligned}$$

Hence,

$$\begin{aligned} e^A &= Qe^\Lambda Q^T \\ &= \begin{pmatrix} 1/\sqrt{3} & 1/\sqrt{2} & 1/\sqrt{6} \\ 1/\sqrt{3} & -1/\sqrt{2} & 1/\sqrt{6} \\ 1/\sqrt{3} & 0 & -2/\sqrt{6} \end{pmatrix} \begin{pmatrix} e & 0 & 0 \\ 0 & 1/e & 0 \\ 0 & 0 & 1 \end{pmatrix} \begin{pmatrix} 1/\sqrt{3} & 1/\sqrt{3} & 1/\sqrt{3} \\ 1/\sqrt{2} & -1/\sqrt{2} & 1/\sqrt{6} \\ -2/\sqrt{6} & 1/\sqrt{6} & -2/\sqrt{6} \end{pmatrix}. \end{aligned}$$

18.06

Professor Strang

Quiz 3

May 7th, 2012

Your PRINTED name is: Michael Plasmeier

Grading

1 20

2 18

3 27

Please circle your recitation:

65

r01	T 11	4-159	Ailsa Keating	ailsa
r02	T 11	36-153	Rune Haugseng	haugseng
r03	T 12	4-159	Jennifer Park	jmypark
r04	T 12	36-153	Rune Haugseng	haugseng
r05	T 1	4-153	Dimiter Ostrev	ostrev
r06	T 1	4-159	Uhi Rinn Suh	ursuh
r07	T 1	66-144	Ailsa Keating	ailsa
r08	T 2	66-144	Niels Martin Moller	moller
r09	T 2	4-153	Dimiter Ostrev	ostrev
r10	ESG		Gabrielle Stoy	gstoy

1 (33 pts.)

Suppose an $n \times n$ matrix A has n independent eigenvectors x_1, \dots, x_n . Then you could write the solution to $\frac{du}{dt} = Au$ in three ways: *Diff eq*

$$u(t) = e^{At}u(0), \text{ or}$$

$$u(t) = Se^{At}S^{-1}u(0), \text{ or}$$

$$u(t) = c_1e^{\lambda_1 t}x_1 + \dots + c_n e^{\lambda_n t}x_n.$$

Here, $S = [x_1 | x_2 | \dots | x_n]$.

eigenvectors

(a) From the definition of the exponential of a matrix, show why e^{At} is the same as $Se^{At}S^{-1}$.

2 Well $A = S\Lambda S^{-1}$ *explain this step.*
 $e^{At} = S e^{\Lambda t} S^{-1}$

Since $A \rightarrow e^{At}$
 $\Lambda \rightarrow e^{\Lambda t}$ through some process

(b) How do you find c_1, \dots, c_n from $u(0)$ and S ?

5 You set $u(0) = c_1 e^{\lambda_1 \cdot 0} [x_1] + c_2 e^{\lambda_2 \cdot 0} [x_2] + \dots$
 $u(0) = c_1 x_1 + c_2 x_2 + \dots$
given find given find given
 n eq, n unknowns (since x_i is $n \times 1$)

explicit formula?

a) didn't look at

(c) For this specific equation, write $u(t)$ in any one of the three forms, using numbers not symbols: You can choose which form.

13

$$\frac{du}{dt} = \begin{bmatrix} 1 & 2 \\ -1 & 4 \end{bmatrix} \begin{bmatrix} u_1 \\ u_2 \end{bmatrix} \text{ starting from } u(0) = \begin{bmatrix} 4 \\ 3 \end{bmatrix}.$$

$$(1-\lambda)(4-\lambda) + 2$$

$$4 - 5\lambda + \lambda^2 + 2 = 0$$

$$\lambda^2 - 5\lambda + 6$$

$$\frac{5 \pm \sqrt{25 - 4 \cdot 6}}{2}$$

$$\lambda = \frac{5 \pm 1}{2} = 3, 2$$

$$\lambda = 3 \begin{bmatrix} -2 & 2 \\ -1 & 1 \end{bmatrix}$$

$$-2x + 2y = 0$$

$$x = y$$

$$\begin{bmatrix} 1 \\ 1 \end{bmatrix}$$

$$\lambda = 2 \begin{bmatrix} -1 & 2 \\ -1 & 2 \end{bmatrix}$$

$$-x + 2y = 0$$

$$2y = x$$

$$\begin{bmatrix} 2 \\ 1 \end{bmatrix}$$

$$\begin{aligned} \frac{du_1}{dt} &= u_1 + 2u_2 \\ \frac{du_2}{dt} &= -u_1 + 4u_2 \end{aligned}$$

$$u = c_1 e^{3t} \begin{bmatrix} 1 \\ 1 \end{bmatrix} + c_2 e^{2t} \begin{bmatrix} 2 \\ 1 \end{bmatrix}$$

$$\begin{bmatrix} 4 \\ 3 \end{bmatrix} = c_1 \begin{bmatrix} 1 \\ 1 \end{bmatrix} + c_2 \begin{bmatrix} 2 \\ 1 \end{bmatrix}$$

$$4 = c_1 + 2c_2$$

$$3 = c_1 + c_2$$

$$c_1 = 3 - c_2$$

$$4 = (3 - c_2) + 2c_2$$

$$= 3 + c_2$$

$$1 = c_2$$

$$c_1 = 2$$

$$u = 2e^{3t} \begin{bmatrix} 1 \\ 1 \end{bmatrix} + 1e^{2t} \begin{bmatrix} 2 \\ 1 \end{bmatrix}$$

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2 (30 pts.)

This question is about the real matrix

$$A = \begin{bmatrix} 1 & c \\ 1 & -1 \end{bmatrix}, \text{ for } c \in \mathbb{R}.$$

(a) - Find the eigenvalues of A , depending on c .

- For which values of c does A have real eigenvalues?

Sym means real e values
c = -1 certainly

$$(1-\lambda)(-1-\lambda) - c$$
$$-1 + 0\lambda + \lambda^2 - c = 0$$

$$\lambda^2 - (1+c) = 0 \quad c \text{ is constant, real \#}$$

$$\lambda = \frac{0 \pm \sqrt{0^2 - 4 \cdot 1 \cdot (1+c)}}{2} = \frac{\pm \sqrt{4(1+c)}}{2}$$

So when is $-4(1+c) \geq 0$

$$1+c \geq 0$$
$$c \geq -1$$

When $c = -1$ $\lambda = 0$ which is real

8/8

$$I^{-1} \left[\begin{array}{cc|cc} 1 & & & \\ & 1 & & \\ & & 1 & \\ & & & 1 \end{array} \right]$$

$$I = I^{-1}$$

$$R = \frac{1}{0} \left[\begin{array}{cc|cc} 0 & 1 & 1 & \\ 1 & 0 & & 1 \end{array} \right]$$

Swap

$$\left[\begin{array}{cc|cc} 1 & 0 & 0 & 1 \\ 0 & 1 & 1 & 0 \end{array} \right]$$

↑ sum

2/3

(b) - For one particular value of c, convince me that A is similar to both the matrix

$$B = \begin{bmatrix} 2 & 0 \\ 0 & -2 \end{bmatrix},$$

and to the matrix

$$C = \begin{bmatrix} 2 & 2 \\ 0 & -2 \end{bmatrix}.$$

$$A = \begin{bmatrix} 1 & c \\ 1 & -1 \end{bmatrix}$$

- Don't forget to say which value c this happens for.

Similar $B = M^{-1} A M$ is similar to A

Were not taught a robust way to compare

What if $M = I$

~~$$\begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix} \begin{bmatrix} 2 & 0 \\ 0 & -2 \end{bmatrix} \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix} \neq \begin{bmatrix} 1 & c \\ 1 & -1 \end{bmatrix}$$~~

~~$$\begin{bmatrix} 0 & 1 \\ 1 & 0 \end{bmatrix} \begin{bmatrix} 2 & 0 \\ 0 & -2 \end{bmatrix} \begin{bmatrix} 0 & 1 \\ 1 & 0 \end{bmatrix} \neq \begin{bmatrix} 1 & c \\ 1 & -1 \end{bmatrix}$$~~

Distinct e.v.s
 \Rightarrow can diagonalize

You can find a M

Since λ values must be the same *not sufficient*

$$\begin{aligned} \text{So } \lambda (\lambda - 2)(\lambda + 2) &= 0 \\ -4 + 0\lambda + \lambda^2 & \\ \lambda^2 - 4 & \end{aligned}$$

$$\frac{-b \pm \sqrt{b^2 - 4ac}}{2} = \pm 2$$

Page 6 of 12

$$\text{So when } \pm 2 = \frac{\pm \sqrt{4(1+c)}}{2}$$



+)]

$$y = \sqrt{4(1+c)}$$

$$16 = 4(1+c)$$

$$4 = 1+c$$

$$3 = c$$

-]]

$$-y = -\sqrt{4(1+c)}$$

← same

Other

$$(2-\lambda)(-2-\lambda) - 2 \cdot 0 = \text{same } \lambda \text{ as before}$$

When $c=3$

5/11

(c) For one particular value of c , convince me that A cannot be diagonalized. It is not similar to a diagonal matrix Λ , when c has that value.

- Which value c ?

- Why not?

A can't be diag when repeated eig values - repeated rows cols

Like $c = -1$

$$\begin{bmatrix} 1 & -1 \\ 1 & -1 \end{bmatrix}$$

$$(1-\lambda)(-1-\lambda) + 1$$

$$-1 + 0\lambda + \lambda^2 + 1$$

$$\lambda^2 = 0$$

$$\lambda = 0, 0$$

repeated λ

Not similar to Λ Why?

5/11

18/30

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3 (37 pts.)

oh A is not this directly $\begin{pmatrix} \lambda_1 \\ \lambda_2 \\ \vdots \\ \lambda_n \end{pmatrix}$ small to large

(a) Suppose A is an $n \times n$ symmetric matrix with eigenvalues $\lambda_1 \leq \lambda_2 \leq \dots \leq \lambda_n$.

10

- What is the largest number real number c that can be subtracted from the diagonal entries of A , so that $A - cI$ is positive semidefinite?
- Why?

All eig^{values} must be > 0 for pos def
 ≥ 0 for pos semidef

So largest c is

$$\lambda_1 - c = 0$$

\uparrow smallest λ

$$\lambda_1 = c$$

$A - cI$ symm. ? -3

(b) Suppose B is a matrix with independent columns.

$B = \begin{bmatrix} \\ \\ \end{bmatrix}$
indep cols

- What is the nullspace $N(B)$?

- Show that $A = B^T B$ is positive definite. Start by saying what that means about $x^T A x$.

Nullspace of B is

$$\begin{bmatrix} B \end{bmatrix} \begin{bmatrix} x \\ y \end{bmatrix} = \begin{bmatrix} 0 \end{bmatrix}$$

(The x, y such that $Bx = 0$)

$A = B^T B$ is positive definite

We know $x^T A x > 0$ for pos def

$$A x = \lambda x^T$$

$$A x^T A x = \lambda x^T A x$$

positive on both sides $\lambda > 0$

$$x^T x = \|x\|^2 \text{ which is pos}$$

$\lambda =$ is pos, since pos def

So $B^T B$

2

(c) This matrix A has rank $r = 1$:

$$A = \begin{bmatrix} 1 & 1 \\ 2 & 2 \end{bmatrix}$$

SVD

similar to book!
almost exactly same!

15

- Find its largest singular value σ from $A^T A$.
- From its column space and row space, respectively, find unit vectors u and v so that

$$Av = \sigma u, \quad \text{and} \quad A = u\sigma v^T.$$

- From the nullspaces of A and A^T put numbers into the full SVD (Singular Value Decomposition) of A :

$$A = \begin{bmatrix} | & | \\ u & \dots \\ | & | \end{bmatrix} \begin{bmatrix} \sigma & 0 \\ 0 & \dots \end{bmatrix} \begin{bmatrix} | & | \\ v & \dots \\ | & | \end{bmatrix}^T$$

$$\underline{AA^T} \quad \begin{bmatrix} 1 & 1 \\ 2 & 2 \end{bmatrix} \begin{bmatrix} 1 & 2 \\ 1 & 2 \end{bmatrix} = \begin{bmatrix} 1 \cdot 1 + 1 \cdot 1 & 1 \cdot 2 + 1 \cdot 2 \\ 2 \cdot 1 + 2 \cdot 1 & 2 \cdot 2 + 2 \cdot 2 \end{bmatrix} = \begin{bmatrix} 2 & 4 \\ 4 & 8 \end{bmatrix}$$

$$\begin{aligned} (2-\lambda)(8-\lambda) - 16 &= 0 \\ 16 - 10\lambda + \lambda^2 - 16 &= 0 \\ \lambda^2 - 10\lambda &= 0 \\ \lambda &= 0, 10 \end{aligned}$$

$$\underline{\lambda=0} \quad \begin{bmatrix} 2 & 4 \\ 4 & 8 \end{bmatrix} \begin{bmatrix} x \\ y \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \end{bmatrix}$$

$$2x + 4y = 0$$

$$x = -2y$$

$$\begin{bmatrix} 2 \\ 1 \end{bmatrix}$$

$$\underline{A^T A} \quad \begin{bmatrix} 1 & 2 \\ 1 & 2 \end{bmatrix} \begin{bmatrix} 1 & 1 \\ 2 & 2 \end{bmatrix} = \begin{bmatrix} 1 \cdot 1 + 2 \cdot 2 & 1 \cdot 1 + 2 \cdot 2 \\ 1 \cdot 1 + 2 \cdot 2 & 1 \cdot 1 + 2 \cdot 2 \end{bmatrix} = \begin{bmatrix} 5 & 5 \\ 5 & 5 \end{bmatrix}$$

$$(5-\lambda)(5-\lambda) - 25 = 0$$

$$25 - 10\lambda + \lambda^2 - 25 = 0$$

$$\lambda^2 - 10\lambda = 0$$

$$\lambda = 0, 10$$

$$\underline{\lambda=0} \quad \begin{bmatrix} 5 & 5 \\ 5 & 5 \end{bmatrix} \begin{bmatrix} x \\ y \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \end{bmatrix}$$

$$5x + 5y = 0$$

$$x = -y$$

$$\begin{bmatrix} -1 \\ 1 \end{bmatrix}$$

$$\lambda = 10$$

$$\begin{bmatrix} -8 & 4 \\ 4 & -2 \end{bmatrix} \begin{bmatrix} x \\ y \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \end{bmatrix}$$

$$4x = 2y$$

$$2x = y$$

$$\begin{bmatrix} 1 \\ 2 \end{bmatrix}$$

$$\frac{1}{\sqrt{5}} \begin{bmatrix} 1 & -2 \\ 2 & 1 \end{bmatrix}$$

$$\lambda = 10$$

$$\begin{bmatrix} -5 & 5 \\ 5 & -5 \end{bmatrix} \begin{bmatrix} x \\ y \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \end{bmatrix}$$

$$5x = 5y$$

$$x = y$$

$$\begin{bmatrix} 1 \\ 1 \end{bmatrix}$$

$$\frac{1}{\sqrt{2}} \begin{bmatrix} 1 & -1 \\ 1 & 1 \end{bmatrix}$$

$$A = \frac{1}{\sqrt{5}} \begin{bmatrix} 1 & -2 \\ 2 & 1 \end{bmatrix} \begin{bmatrix} \sqrt{10} & 0 \\ 0 & 0 \end{bmatrix} \frac{1}{\sqrt{2}} \begin{bmatrix} 1 & -1 \\ 1 & 1 \end{bmatrix}^T$$

largest singular value $\sigma = \sqrt{10}$

~~remark~~

no null space needed - smart or we did

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18.06

5/7

Debrief

Think figured out all but the 2 proof qs
Should be better at those
If had more time to study
Well for final...

Figured out a mistake on one of the practical qs fairly quickly
at end
Very stressful
but think I pulled it together rather well

18.06 Professor Strang Quiz 3 – Solutions May 7th, 2012

Your PRINTED name is: _____

Grading

1

2

3

Please circle your recitation: _____

r01	T 11	4-159	Ailsa Keating	ailsa
r02	T 11	36-153	Rune Haugseng	haugseng
r03	T 12	4-159	Jennifer Park	jmypark
r04	T 12	36-153	Rune Haugseng	haugseng
r05	T 1	4-153	Dimiter Ostrev	ostrev
r06	T 1	4-159	Uhi Rinn Suh	ursuh
r07	T 1	66-144	Ailsa Keating	ailsa
r08	T 2	66-144	Niels Martin Moller	moller
r09	T 2	4-153	Dimiter Ostrev	ostrev
r10	ESG		Gabrielle Stoy	gstoy

1 (33 pts.)

Suppose an $n \times n$ matrix A has n independent eigenvectors x_1, \dots, x_n . Then you could write the solution to $\frac{du}{dt} = Au$ in three ways:

$$u(t) = e^{At}u(0), \quad \text{or}$$

$$u(t) = Se^{At}S^{-1}u(0), \quad \text{or}$$

$$u(t) = c_1e^{\lambda_1 t}x_1 + \dots + c_n e^{\lambda_n t}x_n.$$

Here, $S = [x_1 \mid x_2 \mid \dots \mid x_n]$.

(a) From the definition of the exponential of a matrix, show why e^{At} is the same as $Se^{At}S^{-1}$.

Solution. Recall that $A = SAS^{-1}$, and $A^k t^k = SA^k t^k S^{-1}$. Then, definition of the exponential:

$$\exp(At) = \sum_{k=0}^{\infty} \frac{A^k t^k}{k!} = S \left(\sum_{k=0}^{\infty} \frac{\Lambda^k t^k}{k!} \right) S^{-1} = Se^{At}S^{-1}.$$

□

(b) How do you find c_1, \dots, c_n from $u(0)$ and S ?

Solution. Since $e^0 = 1$, we see that

$$u(0) = c_1x_1 + \dots + c_nx_n = S \begin{bmatrix} c_1 \\ \vdots \\ c_n \end{bmatrix},$$

where we used the definition of the matrix product. Thus the answer is:

$$\begin{bmatrix} c_1 \\ \vdots \\ c_n \end{bmatrix} = S^{-1}u(0).$$

□

(c) For this specific equation, write $u(t)$ in any one of the (added: latter two of the) three forms, using *numbers* not symbols: You can choose which form.

$$\frac{du}{dt} = \begin{bmatrix} 1 & 2 \\ -1 & 4 \end{bmatrix} u, \quad \text{starting from } u(0) = \begin{bmatrix} 4 \\ 3 \end{bmatrix}.$$

Solution. We diagonalize A and get:

$$A = \begin{bmatrix} 2 & 1 \\ 1 & 1 \end{bmatrix} \begin{bmatrix} 2 & 0 \\ 0 & 3 \end{bmatrix} \begin{bmatrix} 1 & -1 \\ -1 & 2 \end{bmatrix}.$$

Thus $c = \begin{bmatrix} 1 \\ 2 \end{bmatrix}$, so for the second form

$$u(t) = \begin{bmatrix} 2 & 1 \\ 1 & 1 \end{bmatrix} \begin{bmatrix} 2 & 0 \\ 0 & 3 \end{bmatrix} \begin{bmatrix} 1 & -1 \\ -1 & 2 \end{bmatrix} \begin{bmatrix} 4 \\ 3 \end{bmatrix} = \begin{bmatrix} 2 & 1 \\ 1 & 1 \end{bmatrix} \begin{bmatrix} e^{2t} & 0 \\ 0 & e^{3t} \end{bmatrix} \begin{bmatrix} 1 \\ 2 \end{bmatrix},$$

while in the third form:

$$u(t) = e^{2t} \begin{bmatrix} 2 \\ 1 \end{bmatrix} + 2e^{3t} \begin{bmatrix} 1 \\ 1 \end{bmatrix}.$$

□

2 (30 pts.)

This question is about the real matrix

$$A = \begin{bmatrix} 1 & c \\ 1 & -1 \end{bmatrix}, \quad \text{for } c \in \mathbb{R}.$$

- (a) - Find the eigenvalues of A , depending on c .
- For which values of c does A have real eigenvalues?

Solution. Since $0 = \text{tr}A = \lambda_1 + \lambda_2$, we see that $\lambda_2 = -\lambda_1$.

Also, $-1 - c = \det A = -\lambda_1^2$. Thus,

$$\lambda = \pm\sqrt{1+c}.$$

Therefore,

the eigenvalues are real precisely when $c \geq -1$.

□

(b) - For one particular value of c , convince me that A is similar to both the matrix

$$B = \begin{bmatrix} 2 & 0 \\ 0 & -2 \end{bmatrix},$$

and to the matrix

$$C = \begin{bmatrix} 2 & 2 \\ 0 & -2 \end{bmatrix}.$$

- Don't forget to say which value c this happens for.

Solution. If two matrices are similar, then they do have the same eigenvalues (those are $2, -2$ for both B and C). Here we must therefore have $0 = \text{tr}A$ and $-1 - c = \det A = -4$. We see that this happens precisely when $c = 3$, where we check that indeed the eigenvalues are $2, -2$. However, this does not guarantee that they are similar - and hence is not convincing.

Convincing: The eigenvalues $2, -2$ are different, so both A, B and C are diagonalizable,

with the same diagonal matrix (for example to $\Lambda = B!$). Therefore A, B and C are all similar when $c = 3$. □

(c) For one particular value of c , convince me that A cannot be diagonalized. It is not similar to a diagonal matrix Λ , when c has that value.

- Which value c ?

- Why not?

Solution. As we saw above, $\text{tr}A = 0$, so regardless of c the eigenvalues come in pairs $\lambda_2 = -\lambda_1$. This means that whenever $\lambda_1 \neq 0$, we have two different eigenvalues, and hence A is diagonalizable (not what we're after).

Thus we need $\lambda_1 = \lambda_2 = 0$, a repeated eigenvalue, which happens when $c = -1$ (so $\det A = 0$) as the only suspect – does it work?

Convincing: For $c = -1$, we have $N(A - 0 \cdot I) = \text{span} \left\{ \begin{bmatrix} 1 \\ 1 \end{bmatrix} \right\}$

With only a 1-dimensional space of eigenvectors for the matrix, we are convinced that A is not diagonalizable for $c = -1$. □

3 (37 pts.)

(a) Suppose A is an $n \times n$ symmetric matrix with eigenvalues $\lambda_1 \leq \lambda_2 \leq \dots \leq \lambda_n$.

- What is the largest number real number c that can be subtracted from the diagonal entries of A , so that $A - cI$ is positive semidefinite?

- Why?

Solution. - We first realize that: If A is symmetric, then $A - cI$ is also symmetric, since in general $(A + B)^T = A^T + B^T$ (simple, but very important to check!).

- Then we realize that the eigenvalues of $A - cI$ are $\lambda_1 - c \leq \lambda_2 - c \leq \dots \leq \lambda_n - c$.

Therefore:

$c = \lambda_1$ is the largest that can ensure positive semidefiniteness (and it does).

□

(b) Suppose B is a matrix with independent columns.

- What is the nullspace $N(B)$?

- Show that $A = B^T B$ is positive definite. Start by saying what that means about $x^T A x$.

Solution. - Then $Bx = 0$ only has the zero solution, so $N(B) = \{0\}$.

- Again, we start by observing that $A^T = A$ is symmetric. Then we recall what positive definite means (the "energy" test):

$$\boxed{x^T A x > 0 \text{ whenever } x \neq 0.}$$

Thus, we see here (by definition the inner product property of the transpose of a matrix):

$$\boxed{x^T A x = x^T (B^T (Bx)) = (Bx)^T (Bx) = \|Bx\|^2 \geq 0.}$$

So $A = B^T B$ is positive semidefinite. But finally, the equality $\|Bx\|^2 = 0$, only happens when $Bx = 0$ which by $N(B) = \{0\}$ means $x = 0$. □

(c) This matrix A has rank $r = 1$:

$$A = \begin{bmatrix} 1 & 1 \\ 2 & 2 \end{bmatrix}.$$

- Find its largest singular value σ from $A^T A$.
- From its column space and row space, respectively, find unit vectors u and v so that

$$Av = \sigma u, \quad \text{and} \quad A = u\sigma v^T.$$

- From the nullspaces of A and A^T put numbers into the full SVD (Singular Value Decomposition) of A :

$$A = \begin{bmatrix} | & | \\ u & \dots \\ | & | \end{bmatrix} \begin{bmatrix} \sigma & 0 \\ 0 & \dots \end{bmatrix} \begin{bmatrix} | & | \\ v & \dots \\ | & | \end{bmatrix}^T.$$

Solution. We compute:

$$A^T A = \begin{bmatrix} 5 & 5 \\ 5 & 5 \end{bmatrix}.$$

Thus the two eigenvalues are $\lambda_1 = 0$ and $\lambda_2 = 10$, and $\sigma = \sqrt{10}$. For v , we find a vector in $N(A^T A - 10I)$, and normalize to unit length:

$$v = \begin{bmatrix} 1/\sqrt{2} \\ 1/\sqrt{2} \end{bmatrix}.$$

Then we find u using

$$u = \frac{Av}{\sigma} = \begin{bmatrix} 1/\sqrt{5} \\ 2/\sqrt{5} \end{bmatrix}.$$

Since we have the orthogonal sums of subspaces $\mathbb{R}^2 = \mathbb{R}^m = c(A) \oplus N(A^T)$ and also $\mathbb{R}^2 = \mathbb{R}^n = c(A^T) \oplus N(A)$, we need to find one unit vector from each of $N(A)$ and $N(A^T)$ and augment to v and u , respectively:

$$v_2 = \begin{bmatrix} 1/\sqrt{2} \\ -1/\sqrt{2} \end{bmatrix} \in N(A),$$

$$u_2 = \begin{bmatrix} -2/\sqrt{5} \\ 1/\sqrt{5} \end{bmatrix} \in N(A^T),$$

Thus, we finally see the full SVD:

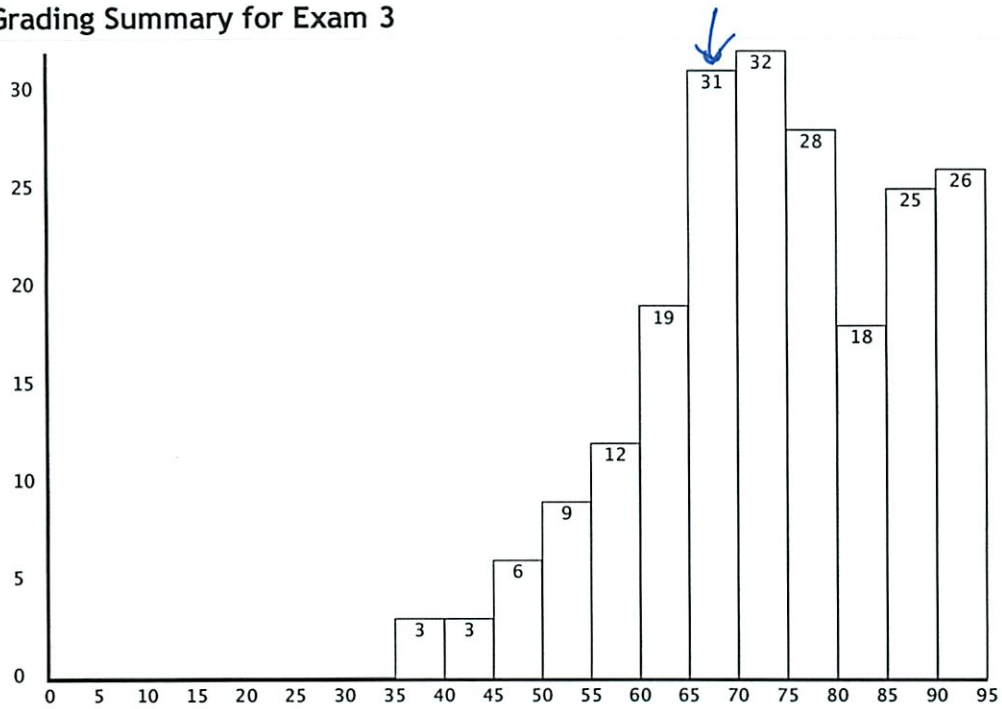
$$A = U\Sigma V^T = \begin{bmatrix} 1/\sqrt{5} & -2/\sqrt{5} \\ 2/\sqrt{5} & 1/\sqrt{5} \end{bmatrix} \begin{bmatrix} \sqrt{10} & 0 \\ 0 & 0 \end{bmatrix} \begin{bmatrix} 1/\sqrt{2} & 1/\sqrt{2} \\ 1/\sqrt{2} & -1/\sqrt{2} \end{bmatrix}^T.$$

We remember, as a final check, to verify that the square matrices U and V both contain orthonormal bases of \mathbb{R}^2 as they should:

$$UU^T = I_2,$$













$$VV^T = I_2.$$

□

18.06 Linear Algebra[Dashboard](#) [Students](#) [Assignments](#)**Grading Summary for Exam 3**

Number of Scores: 212
Average: 72.96
Standard Deviation: 13.43

18.06 Linear Algebra**Grade Report****Grade Report for Michael E. Plasmeier**

Assignment/Exam Name	Graph	Due Date	Points	Max Pts	Weight
Homework 1		02.16.2012	83.00	100.00	1.50%
Homework 2		02.23.2012	65.00	100.00	1.50%
Homework 3		03.01.2012	77.00	100.00	1.50%
Homework 4		03.08.2012	77.00	100.00	1.50%
Exam 1		03.09.2012	69.00	100.00	15.00%
Homework 5		03.22.2012	83.00	100.00	1.50%
Homework 6		04.05.2012	85.00	100.00	1.50%
Exam 2		04.11.2012	56.00	100.00	15.00%
Homework 7		04.19.2012	94.00	100.00	1.50%
Homework 8		04.26.2012	86.00	100.00	1.50%
Homework 9		05.03.2012	91.00	100.00	1.50%
Exam 3		05.08.2012	65.00	100.00	15.00%
CUMULATIVE SCORE			39.62 / 58.50 = 67.72%		

Instructor's Comments