

Information - Uncertainty

Info content $\log_2 \frac{1}{P(\text{seq})}$ in bits

Expected info content

$$H(X) = E(I(X)) = \sum_{i=1}^N p(x_i) \log_2 \left(\frac{1}{p(x_i)} \right)$$

If all p_i s = uniform, $\frac{1}{N}$
 $\log_2(N)$

Outcome reduced to M possible choices

Entropy after receipt

$$\frac{1}{M} \log_2 \left(\frac{1}{M} \right) = \log_2 M$$

Entropy in message is change

$$H_{\text{before}} = H_{\text{after}} = \log_2(N) - \log_2(M) = \log_2 \left(\frac{N}{M} \right) \text{ original # choices} \\ \text{# of choices left!}$$

Example: Have 52 cards, tell you it's a ♦

$$\text{So } \log_2 \left(\frac{52}{13} \right) = 2 \text{ bit info}$$

Additive

Fixed is easiest

Variable saves space

- max down to entropy

Huffman coding - optimal

Compute avg length of code

$\sum P_i \text{symbol}(L_{\text{symbol}})$

$$P \approx \frac{1}{2^k}$$

Log the power to which the base must be raised to do produce that #

$$\log_2 16 \rightarrow 2^x = 16$$

$$\log(x_y) = \log(x) + \log(y)$$

$$\log_b(x^p) = p \log_b x$$

$$e^{\ln(x)} = x \quad \ln(e^x) = x$$

$$\log_b \left(\frac{x}{y} \right) = \log_b(x) - \log_b(y)$$

$$\log_b(\sqrt{x}) = \frac{1}{2} \log_b(x)$$

$$\log_b(x) = \frac{\log_k(x)}{\log_k(b)}$$

6.02 #1

Clock Recovery (the constant adjust forward)

backwards

Sample middle one

Other stuff

How many samples/bit

Where does byte start?

8bit/10

1. Lots of bit transitions

2. DC balance 0s, 1s

3. Special sync symbol

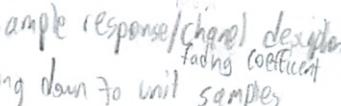
$x[n]$ = input

$y[n]$ = output

$u[n]$ = unit step 

$s[n]$ = unit step response

$\delta[n]$ = unit sample 

$h[n]$ = unit sample response (char) 

break everything down to unit samples

Time invariant $x[n-N] \rightarrow y[n-N]$

Linear $a_1 x_1[n] + b_1 x_2[n] \rightarrow a_1 y_1[n] + b_1 y_2[n]$
 weighted sums $n \rightarrow$ out
 convolution allows deconvolution

- commutative $x[n] * h[n] = h[n] * x[n]$

- associative

$$x[n] * (h_1[n] * h_2[n]) = (x[n] * h_1[n]) * h_2[n]$$

- distributive

$$x[n] * (h_1[n] + h_2[n]) = x[n] * h_1[n] + x[n] * h_2[n]$$

Parallel 

Series 

Just a product of what came before

Just clearest point - not when transition is

(causal) - depends only on current & previous values

Scalar - real #, not vector

$$U[n] = S[n] - S[n-1]$$

ISI

$$B = \left[\frac{\text{length } h[n] \text{ active}}{N} \right] + 2$$

test pattern 2^{N+1}

pick samples/bit

lift than fast/slow channel

deconvolution

$$w[n] = \frac{1}{n+1} (y[n] - w[n-1]h[1] + \dots)$$

Stability

$$\sum_{m=1}^k \left| \frac{h[m]}{h[0]} \right| \leq 1 \quad \sum_{m=1}^k w[m] \leq h[0]$$

drop first $h[0]$ if is or close to 0

Noise

$$\text{Mean } \mu_x = \frac{1}{N} \sum_{n=1}^N x[n]$$

$$P_x = \frac{1}{N} \sum_{n=1}^N x[n]^2 \quad \hat{P}_x = \frac{1}{N} \sum_{n=1}^N (x[n] - \mu_x)^2$$

$$E_x = \sum_{n=1}^N x[n]^2 \quad \hat{E}_x = \sum_{n=1}^N (x[n] - \mu_x)^2$$

$$\text{SNR} = \frac{\hat{P}_{\text{signal}}}{\hat{P}_{\text{noise}}} \quad \text{SNR(db)} = 10 \log \left(\frac{\hat{P}_{\text{signal}}}{\hat{P}_{\text{noise}}} \right)$$

stationary vs ergodic random process

$$P(x_1 \leq x \leq x_2) = \int_{x_1}^{x_2} f_x(x) dx$$

$$\mu_x = \int_{-\infty}^{\infty} x f_x(x) dx$$

$$\sigma^2 = \int_{-\infty}^{\infty} (x - \mu_x)^2 f_x(x) dx$$

$$\text{PDF (normal)} = \frac{1}{\sqrt{2\pi\sigma^2}} e^{-\frac{(x-\mu)^2}{2\sigma^2}}$$

$$\text{PDF - erf}(x) = \frac{2}{\sqrt{\pi}} \int_0^x e^{-t^2} dt$$

$$\Phi_{\mu, \sigma}(x) = \Phi \left(\frac{x-\mu}{\sigma} \right)$$

BER bit error ratio

$$M_{Y_{NF}} = \frac{1}{N} \sum_{n=1}^N Y_{NF}[n] = \frac{1}{N} \cdot \frac{N}{2} = \frac{1}{2}$$

$$\hat{P}_{Y_{NF}} = \frac{1}{N} \sum_{n=1}^N \left(Y_{NF}[n] - \frac{1}{2} \right)^2 = \frac{1}{N} \sum_{n=1}^N \left(\frac{1}{2} \right)^2 = \frac{1}{N} \cdot \frac{N}{4} = \frac{1}{4}$$

$$\begin{aligned} P(\text{error}) &= P(0) \cdot P(\text{error} | \text{transmit } 0) + P(1) P(\text{error} | 1) \\ &= 0.5 \cdot \phi(-0.5/\sigma) + 0.5 \cdot \phi(-0.5/\sigma) \\ &= \phi(-0.5/\sigma) \end{aligned}$$

$$S/NR(\text{db}) = 10 \log \left(\frac{\hat{P}_{\text{Signal}}}{\hat{P}_{\text{Noise}}} \right) = 10 \log \left(\frac{+25}{\sigma^2} \right)$$

Eye diagram

All possible voltage sequences in a certain # of bits

If have a channel that receives more 1s
make it more likely to receive a 1
Find Φ w/ test signals

$$B = \left[\frac{M}{N} \right] + 1$$

Deconvolution w is estimated \times

$$y[n] = h[0]w[n] + h[1]w[n-1] + \dots$$

$$w[n] = \frac{y[n] - (h[1]w[n-1] + h[2]w[n-2] + \dots)}{h[0]}$$

$$w[0] = \frac{y[0]}{h[0]}$$

$$w[1] = \frac{y[1] - h[1]w[0]}{h[0]}$$

$$w[2] = \frac{y[2] - (h[1]w[0] + h[2]w[1])}{h[0]}$$

6.02 Exam 1 study

3/3

Cheat sheet!

Make that instead

worse case - just do the problem at?

Huffman length: $\lceil \log_2 (\frac{N}{m}) \rceil + \# \text{ items} - 1$

Oh that was wrong in class anyway

→ Hamming Distance

BJ example $\log_2 (\frac{N}{m}) + \# \text{ choices left } (52-3=49)$

LZW I think I know

- or should I do a practice??

Just do pseudo code they give

I'm skipping JPEG

- we never reviewed

What is least significant vs most sig bit first?

Not writing convolution math formula

What is ~ again?

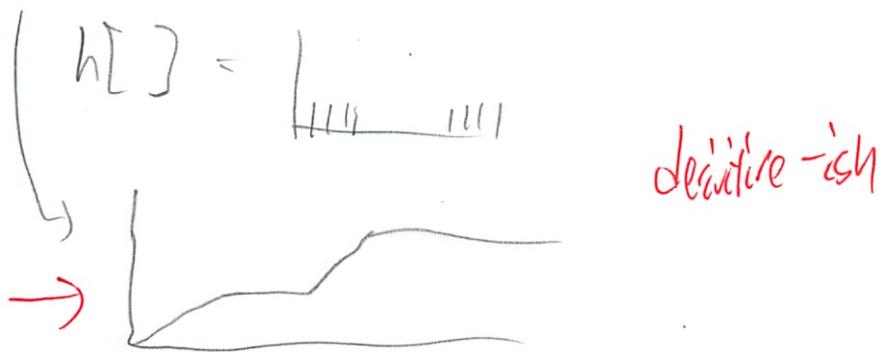
Causal?

Is prob on test?

②

② a)

Unit step response



b) Which approach gave result as $n \rightarrow \infty$

→ - need to add up all

→ - can't do by pic

(I had totally the wrong intuition)

c) Back at h

- what was the intuition here

We thought it is not it

Just below

No can do math wise

Oh forgot they gave channels

Notice is it increasing a lot

$$y[15] = h[0]x[15] + h[1]x[14] + \dots + h[15]x[0]$$

~~so~~ $x[15] - x[14]$ is 0 - rest are 1

Just add up h

③

Eye which
✓ got

13 What is $y[24]$

6011010 # 4 samples per bit

$$h[] = [1 1 1 1]$$

4 8 12 16 20 24
right after | 6 1 8
 |
 |

it never goes away;
- no is not step response

$$x[24]h[0] + x[23]h[1] + x[22]h[2] + x[21]h[3]$$

$$0 \cdot 2 + 1 \cdot 2 + 1 \cdot 3 + 1 \cdot 3$$

This was 0

messed me up

Oh start at 0, I started at 1

When counting

B Input to deconvolver

$$w_A[n] = \frac{1}{\alpha} [y[n] - (0.2w_A[n-1] + 0.3w_A[n-2] + 0.3w_A[n-3])]$$

$$w_A[n] = 0 \text{ for } n < 0$$

for what value of α will $w_A[n] = x[n]$

(4)

? How do you do this?

Use convolution sum and $h[n]$ from above

$$y[n] = x[n] \cdot 1 + x[n-1] \cdot 2 + x[n-2] \cdot 3 + x[n-3] \cdot 3$$

Reorg for deconvolver

$$y[n] = w_a[n] \cdot 1 + w_a[n-1] \cdot 2 + w_a[n-2] \cdot 3 + w_a[n-3] \cdot 3$$

If want $w_a[n] = x[n]$

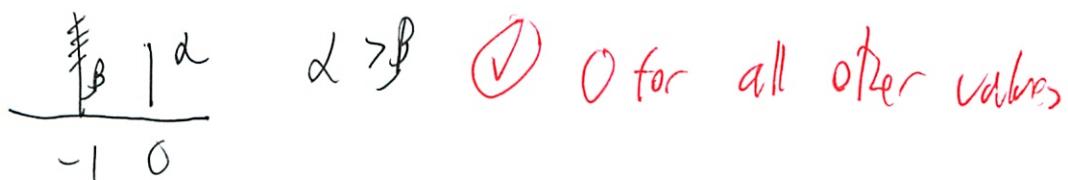
then d must = 2

2. Result in terms of unit sample

$$\delta[n+1] + 4\delta[n] + 8\delta[n-1]$$

Watch the negative time values

Qb What is h_-



5)

c) another convolution one

$$\begin{array}{ccccccccc} & 1 & 0 & 0 & 1 & 1 & 0 & 1 & 1 \\ \begin{array}{c} 3 \\ \hline 1 \end{array} & & | & & & & & & \\ & 1 & 7 & & & & & & \\ & & & & & & & & \\ & & & 3 & & & & & \\ & & & | & & & & & \\ & & & 0 & & & & & \\ & & & & 7 & & & & \\ & & & & & & & & \\ & & & & & 1 & 3 & .7 & 1 & (1111 \rightarrow \textcircled{0}) \end{array}$$

d) derive a deconvolver

$$y[n] = 1,7 \ 1 \ 1,3 \ 1,7 \ 1,3 \ 0 \ 0 \ 0 \ 0 \ 0$$

$$\begin{array}{c} 1 \\ \hline 3 \ 7 \end{array}$$

oh I can see what to do

$$1 \ 1 \ 1 \ 0 \ 1 \ 1 \ 0 \ 0 \ 0 \ 0 \rightarrow \textcircled{0}$$

$$w[n] = \frac{1}{h[0]} (y[n] - h[1]w[n-1]) \quad (\text{what does this mean?})$$
$$= \frac{y[n]}{1,7} - \frac{1,3}{1,7} w[n-1]$$

(6) w is an estimate of X

$$H \cdot W = Y$$

$$Y[n] = \sum_{m=0}^{m=n} w[n-m]h[m]$$

$$\text{(just a reverse of } Y[n] = \sum_{m=0}^{m=n} x[m]h[n-m])$$

$$Y[n] = h[0]w[n] + h[1]w[n-1] + h[2]w[n-2] + \dots$$

So just wrote w for x
where we flip is
slide

Then reverse for $w[n]$

$$w[0] = \text{previously called } w[1]$$

$$w[n] = \frac{Y[n] - (h[1]w[n-1] + h[2]w[n-2] + \dots)}{h[0]}$$

So back to our problem

$$Y[0] = h[0]x[0]$$

$$x[0] = 1$$

So backwards

$$x[0] (\text{called } w[0]) = \frac{Y[0]}{h[0]}$$

$$= \frac{1}{1} = 1$$

(7)

$$y[n] =$$

$$y[1] = 1.7 + h[1] \times [1]$$

$$= 1.7 + 1.3 \cdot 1$$

So

$$w[1] = \frac{y[1] - 1.7}{h[1]}$$

$$= \frac{1 - 1.7}{1.3} \quad \text{c: always } h[0]$$

$$= \frac{1 - 1.7}{1.3}$$

$$= 1 \quad (0)$$

$$w[2] = \frac{y[2] - 1.3 - 1.7}{h[0]}$$

$$w[2] = \frac{y[2] - (h[1]w[n-1] + h[2]w[n-2])}{h[0]}$$

$$= \frac{1 - (1.7 \cdot 1 + 1.3 \cdot 1)}{1.7} \quad \begin{matrix} h[2]=0 \\ \text{watch the formulas!} \end{matrix}$$

$$= 1$$

~~Oh I was taking wrong input so opps~~

or was it - no not write right

⑧

6. Clever question

- have $h[n]$

(

$$-.05 + .09 + .13 \quad \text{forgot the additive here}$$
$$.05(1) + .09(1) + .13(1) + .20(1) +$$

but forgot $h[0] = 0$
so $.0 + .5 + .9 = 1.4$

↑
when it changes
what should I do?

' still climbing from initial and subtract

$$+.13 + .13 + .13 \text{ or } -.05$$

$$-.05 - .09 - .13 - .20 - \cancel{.13} \cancel{.13}$$

First have looked at how many bits interfere

I implicitly did this

Sample 8 is 3rd bit

2 Bits start 0, 5, 10

$$y[n] = \sum_k h[k] \times [n-k] \text{ go third sample of bit 3}$$

But $y[12]$ & first such value where third sample of
third bit fine

' why? to get it started?

actually right

for 10

but looking for 3

(9)

| | | | | | | | | | | |
|--------|---|-----|-----|-----|-----|-----|-----|-----|-----|---|
| n | 0 | 1 | 2 | 3 | 4 | 5 | 6 | 7 | 8 | 9 |
| $h[n]$ | 0 | .04 | .12 | .24 | .40 | .60 | .72 | .84 | .96 | 1 |

Oh right!

It's additive!!!!,



that just means slower rise actually was
 not decrease I right,
 I did add

Why don't I get that?

Now consider all possible values of current + future bit

Use superposition (add on top of each other)

$$111 \quad v[n] \quad y[12] = h[12] = 1$$

$$110 \quad v[n] - v[n-10] \quad y[12] = h[12] - h[2] = 1 - .12 = .88$$

$$101 \quad v[n] - v[n-5] + v[n-10] = 1 - .84 + .12$$

↑ so it is add it up + subtract

So from 1st 2nd 3rd

Use their way in future

(8)

7. Same as previous problem, but just one line

Remember lowest curve is 101

So consider

I see

8. Just do the convolution

Hamming Code

bit error correction

'outside of scope'

Q* for unit step response look at difference for h

? for responses

back at changes

Only have to look at past responses

know which step you are at

Unit sample = Unit Step - Unit step

$$h[n] = s[n] - s[n-1]$$

∴ do they want for whole pulse?

11

II. To find $h[l]$ from deconvolve linear eq

Write in terms of $y[l]$

And then scaling factor is $h[l]$

Hamming distance

Strings \emptyset = length

of positions where strings are different

ℓ
↑
of errors

Information - Uncertainty

Info content $\log_2 \frac{1}{P(\text{seen})}$ in bits

Expected info content

$$H(X) = E(I(X)) = \sum_{i=1}^n p(x_i) \log_2 \left(\frac{1}{p(x_i)} \right)$$

If all p_i s = uniform, $\frac{1}{N}$
 $\log_2(N)$

Outcome reduced to M possible choices

Entropy after receipt

$$\frac{1}{M} \log_2 \left(\frac{1}{M} \right) = \log_2 M$$

Entropy in message is change

$$H_{\text{before}} - H_{\text{after}} = \log_2(M) - \log_2(N) \\ = \log_2 \left(\frac{M}{N} \right) \text{ original # choices} \\ \quad \# \text{ of choices left!}$$

Example: Have 52 cards, tell you its a ♦

$$\text{So } \log_2 \left(\frac{52}{13} \right) = 2 \text{ bits info}$$

Additive

Fixed is easiest

Variable saves space

- max down to entropy

Huffman coding - optimal

Compute avg length of code

$$\sum P_{\text{symbol}} (L_{\text{symbol}})$$

$$P \approx \frac{1}{2^k}$$

Log the power to which the base must be raised to do produce that #

$$\log_2 16 \rightarrow 2^x = 16$$

$$\log(xy) = \log(x) + \log(y)$$

$$\log_b(x^p) = p \log_b x$$

$$e^{\ln(x)} = x \quad \ln(e^x) = x$$

$$\log_b \left(\frac{x}{y} \right) = \log_b(x) - \log_b(y)$$

$$\log_b(\sqrt[p]{x}) = \frac{\log_b(x)}{p}$$

$$\log_b(x) = \frac{\log_b(x)}{\log_b(b)}$$

6.02 #1

Clock Recovery The (constant adj) forward/backwards

Sample middle one

Other stuff

How many samples/bit

Where does byte start?

8bit/10

1. Lots of bit transitions

2. DC balance 0s, 1s

3. Special sync symbol

$x[n]$ = input

$y[n]$ = output

$u[n]$ = unit step $\begin{matrix} 0 & n < 0 \\ 1 & n \geq 0 \end{matrix}$

$s[n]$ = unit step response

$\delta[n]$ = unit sample $\begin{matrix} 1 & n = 0 \\ 0 & n \neq 0 \end{matrix}$

$h[n]$ = unit sample response/channel fading coefficient

break everything down to unit samples

Find invariant $x[n-N] \rightarrow y[n-N]$

Linear $a x_1[n] + b x_2[n] \rightarrow a y_1[n] + b y_2[n]$
weighted sums
allows deconvolution

Convolution - Commutative $x[n] * h[n] = h[n] * x[n]$

- associative $x[n](h_1[n] * h_2[n]) = (x[n] * h_1[n]) * h_2[n]$

- distributive $x[n] * (h_1[n] + h_2[n]) = x[n] * h_1[n] + x[n] * h_2[n]$

Parallel $\begin{array}{c} \oplus \\ \oplus \end{array}$

Series $\begin{array}{c} \square \\ \square \end{array}$

Just a product of what came before

Just clearest point - not when transition is

(causal) - depends only on current & previous values

Scalar - real #, not vector

$$u[n] = s[n] - s[n-1]$$

ISI

$$B = \left[\frac{\text{length } h[n] \text{ active}}{N} \right] + 2$$

test pattern 2MxB

pick samples/bit

diff than fast/slow channel

deconvolution

$$w[n] = \frac{1}{h[n]} (y[n] - v[n-1]h[1] + \dots)$$

Stability

$$\sum_{m=1}^k \left| \frac{h[m]}{h[0]} \right| \leq 1 \quad \sum_{m=1}^k w[m] \leq h[0]$$

drop first $h[0]$ if is or close to 0

Noise

$$\text{Mean } \mu_x = \frac{1}{N} \sum_{n=1}^N x[n]$$

$$P_x = \frac{1}{N} \sum_{n=1}^N x[n]^2 \quad \hat{P}_x = \frac{1}{N} \sum_{n=1}^N (x[n] - \mu_x)^2$$

$$E_x = \sum_{n=1}^N x[n]^2 \quad \hat{E}_x = \sum_{n=1}^N (x[n] - \mu_x)^2$$

$$\text{SNR} = \frac{\hat{P}_{\text{signal}}}{\hat{P}_{\text{noise}}} \cdot \text{SNR(db)} = 10 \log \left(\frac{\hat{P}_{\text{signal}}}{\hat{P}_{\text{noise}}} \right)$$

stationary vs ergodic random process

$$P(x_1 \leq x \leq x_2) = \int_{x_1}^{x_2} f_x(x) dx$$

$$\mu_x = \int_{-\infty}^{\infty} x f_x(x) dx$$

$$\sigma^2 = \int_{-\infty}^{\infty} (x - \mu_x)^2 f_x(x) dx$$

$$\text{PDF (normal)} = \frac{1}{\sqrt{2\pi\sigma^2}} e^{-\frac{(x-\mu)^2}{2\sigma^2}}$$

$$\text{PDF - erf}(x) = \frac{2}{\sqrt{\pi}} \int_0^x e^{-t^2} dt$$

$$\Phi_{\mu, \sigma}(x) = \Phi \left(\frac{x-\mu}{\sigma} \right)$$

BER bit error ratio

$$Y_{\text{ref}} = \frac{1}{N} \sum_{n=1}^N Y_{\text{ref}}[n] = \frac{1}{N} \cdot \frac{N}{2} = \frac{1}{2}$$

$$\tilde{P}_{\text{ref}} = \frac{1}{N} \sum_{n=1}^N \left(Y_{\text{ref}}[n] - \frac{1}{2} \right)^2 = \frac{1}{N} \sum_{n=1}^N \left(\frac{1}{2} \right)^2 = \frac{1}{N} \cdot \frac{N}{4} = \frac{1}{4}$$

$$P(\text{error}) = P(0) \cdot P(\text{error} | \text{trans}0) + P(1) \cdot P(\text{error} | 1)$$

$$= .5 \cdot \phi(-.5/\sigma) + .5 \cdot \phi(-.5/\sigma)$$

$$= \phi(-.5/\sigma)$$

$$\text{SNR(dB)} = 10 \log \left(\frac{P_{\text{signal}}}{P_{\text{noise}}} \right) = 10 \log \left(\frac{.25}{\sigma^2} \right)$$

Eye diagram

All possible voltage sequences in a certain # of bits

If have a channel that receives more 1s
make it more likely to receive a 1
Find ϕ w/ test signals

$$B = \left[\frac{M}{N} \right] + 1$$

Encode LZW

initialize TABLE[0 to 255] = code for individual bytes

STRING = get input signal

while there are still input symbols:

SYMBOL = get input symbol

if STRING + SYMBOL is in TABLE

 STRING = STRING + SYMBOL

else,

 output the code for STRING

 add STRING + SYMBOL to TABLE

 STRING = SYMBOL

Output the code for STRING

Decode LZW

initialize TABLE[0 to 255] = code for individual bytes

CODE = read next code from encoder

STRING = TABLE[CODE]

Output STRING

while there are still codes to receive

CODE = read next code from encoder

if TABLE[CODE] is not defined:

 ENTRY = STRING + STRING[0]

else,

 ENTRY = TABLE[CODE]

Output ENTRY

Add STRING + ENTRY[0] to TABLE

STRING = ENTRY

Deconvolution w is estimated \times

$$y[n] = h[0]w[n] + h[1]w[n-1] + \dots$$

$$w[n] = \frac{y[n] - (h[1]w[n-1] + h[2]w[n-2] + \dots)}{h[0]}$$

$$w[0] = \frac{y[0]}{h[0]}$$

$$w[1] = \frac{y[1] - h[1]w[0]}{h[0]}$$

$$w[2] = \frac{y[2] - (h[1]w[0] + h[2]w[1])}{h[0]}$$