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Digital Communications Systems

Digital: message comprising a discrete-time sequence of symbols from source alphabet

Analog: continuous-time waveform

Measure and Information:

$$I_x = \log_2 \frac{1}{P_x}$$

Expected information:

Entropy: $H = \sum_x P_x I_x = \sum_x P_x \log_2 \frac{1}{P_x}$

Probabilistic Model



Space and outcomes

event has occurred if outcome is in the event set

$$\text{Ex: } V = \text{vowel} = \{a, e, i, o, u\}$$

$$F = \text{First five} = \{a, b, c, d, e\}$$

$$P(V) = P(a) + P(e) + P(i) + P(o) + P(u)$$

$$P(F) = P(a) + P(b) + P(c) + P(d) + P(e)$$

$$P(V \text{ or } F) = P(V) + P(F) - P(V \text{ and } F)$$

because $P(V)$ and $P(F)$ are not
mutually exclusive

Conditional Probability

$$P(V|F) = \frac{P(V \text{ and } F)}{P(F)}$$

IF $P(V|F) = P(V)$

$\Rightarrow V$ and F are independent

$$\Rightarrow P(V \text{ and } F) = P(V) \cdot P(F)$$

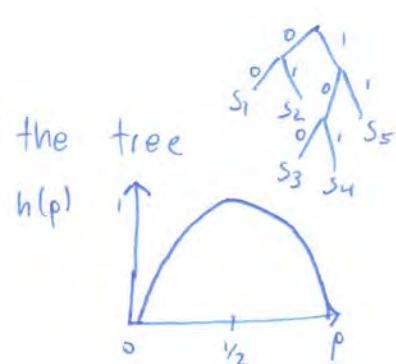
$$P(V \text{ and } F) = P(V|F) P(F)$$

$$\text{Bayes' Rule} = P(F|V) P(V) \quad \text{by symmetry}$$

$$\Rightarrow P(V|F) = \frac{P(F|V) P(V)}{P(F)}$$

Prefix free: All symbols are at the leaves of the tree

$$\text{Entropy: } H = p \log_2 \frac{1}{p} + (1-p) \log_2 \frac{1}{1-p} = h(p)$$



Prefix-free: All symbols at the leaves of the tree.

$\boxed{S} \rightarrow \dots S_8, S_{11}, S_{12}, S_7, \dots \rightarrow \boxed{E} \rightarrow \dots 011010110 \dots \rightarrow \boxed{c}$

each source symbol s_i chosen independently with probability p_i
 each s_i is drawn from the same probability distribution.

$$L = \sum p_i l_i \quad L \geq H \leftarrow \text{lower bound}$$

Efficiency: Encode information / symbols using the least amount
 of bits.
 \curvearrowright depends on prob of symbol occurring

Claim: exists a code where $H \leq L < H + 1$

Proof in notes?

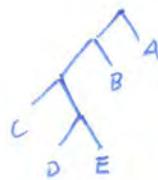
Huffman Encoding

- variable length
- shorter codeword \rightarrow more likely

Algorithm

- sort based on p_i
- merge two least probable (combine probs)
- repeat until done

Symbol	\underline{P}
A	.4
B	.2
C	.2
D	.1
E	.1



$$L = 1(.4) + 2(.2) + 3(.2) + 4(.1) + 4(.1) =$$

ZW Encoding

- need not know stats
- sender and receiver store what has been sent \Rightarrow library

6.02 Notes

Lecture 2: LZW and Huffman

Prefix Free - all symbols on leaves of code trees

$$L \geq H^{\text{entropy}}$$

Huffman Encoding

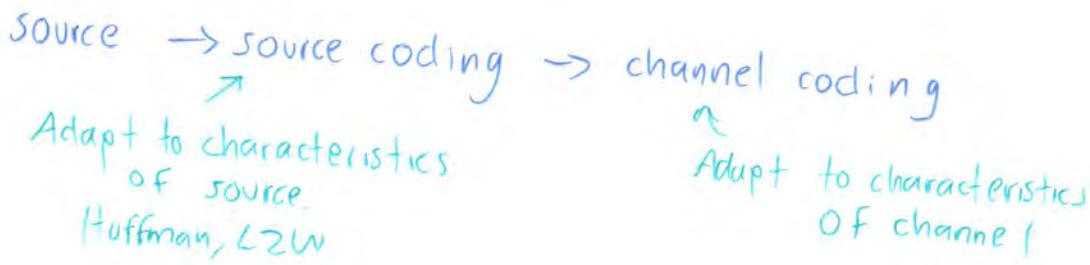
- create code with minimum expected codeword length
 - 1) Sort list by probability
 - 2) Combine ²symbols with least probability
- Good if probabilities are known ahead of time

Lempel-Ziv-Welch (LZW) Encoding/Compression

- No need to know probabilities
- Dictionary maps indeces to variable-length strings
 - initialized with all symbols

6.02 Notes

Lecture 3: Binary Symmetric Channel Hamming Distance, Repetition Codes



Binary Symmetric Channel - Abstraction, model errors
bits flipped independently with $p < .5$

Coding rate = # useful message bits /

Hamming Distance (HD) = distance b/w codewords of equal length

Easy way to calculate \Rightarrow Add all bits mod 2 with no carrying (XOR)
So:
$$\begin{array}{r} 0110100011 \\ + 1001001011 \\ \hline 1111101000 \end{array}$$
$$\begin{array}{l} 0+1=1 \\ 1+1=0 \\ 0+0=0 \end{array}$$

$HD = \# 1's$ in addition = weight of string

Minimum HD = minimum HD b/w any two valid codewords = d

Detect t errors if $d \geq t+1$

Correct t errors if $d \geq 2t+1$

can detect all patterns up to t -bit errors while simultaneously correcting all patterns up to $t-C$ bit errors ($t-C \leq t-D$) iff $d \geq t-C+t-D+1$

Parity Check

Add bit to end of message to make # 1's even $\Rightarrow \uparrow HD$

$d=2 \Rightarrow$ detect 1 bit error
correct none

Block Coding

use parity check idea
Add $n-k$ bits to message to "protect" as many bits as possible
rate = k/n

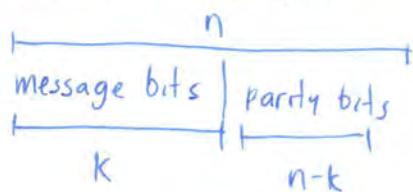
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Lecture 4: Linear Block Codes

Block Codes

message $\Rightarrow k$ bit blocks $\rightarrow n$ bit codeword $\xrightarrow{\text{send}}$
rate = $k/n \Rightarrow (n, k)$ codes or (n, k, d)
 2^n code words to choose from : 2^k to select $\uparrow \text{min HD}$

Systematic codes:



Linear block Codes

Sum of two codewords also a codeword
 $\uparrow \text{sum mod 2 aka XOR}$

Minimum HD = smallest weight among non-zero codewords

Generator Matrix G
 $c = dG$
 $\begin{matrix} \text{data } k \text{ elem row} \\ \text{codeword } n \text{ elem row} \end{matrix}$
 $\uparrow \text{kxn Generator matrix}$

We want the first k bits of c to = first k bits of d
 \hookrightarrow since systematic codes (adding parity to end)
 $G = [I | A]$

Rectangular Codes

calculate parity bit of each row and column
Rate = $r.c / (r.c + \frac{\text{rows}}{\text{columns}})$

d_1	d_2	$ p_1$
d_3	d_4	$ p_2$
p_3	p_4	

In systematic form we have
so

$$G = \left[\begin{array}{cccc|cc} 1 & 0 & 0 & 0 & 1 & 0 & 0 \\ 0 & 1 & 0 & 0 & 1 & 0 & 0 \\ 0 & 0 & 1 & 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & 1 & 0 & 0 & 1 \end{array} \right] \begin{array}{l} d_4 \\ d_2 \\ d_2 \\ d_4 \end{array}$$

\uparrow
First parity bit covers 1st 2nd data bits

\uparrow
 p_4 covers d_2 and d_4

\Rightarrow SEC = Single Error Collection

could be more efficient \Rightarrow Hamming Codes

Hamming Codes

correct single bit errors with min # parity bits

$$n = 2^{k+1} - 1$$

$$P_1 = d_1 + d_2 + d_3$$

$$P_2 = d_1 + d_3 + d_4$$

$$P_3 = d_2 + d_3 + d_4$$

more on next lecture

Bit positions with powers of 2 = parity bit
the rest is data bits

$\Rightarrow P_1$ protects all bits that have least sig dig set
 P_2 " " " second " "
 P_3 " " " third "

6.02 Notes

Lecture 5: Syndrome Decoding

Syndrome Decoding

- efficient way to decode any linear block code

Recall: encoding using G where $dG = c$ $G = [I \mid A]$

For decoding $Hr^T = S$ syndrome bits

$$H = [A^T \mid I]$$

$$He^T = S$$

error vector (which bits to flip)

1) Syndrome table:

FOR SEC populate with $\underbrace{He^T}_{\text{error vectors}}$ for each e

1	0	0	0	0	..
0	1	0	0	0	..
0	0	1	0	0	..
0	0	0	1	0	..
..

2) For each received codeword find Hr^T

TODO
FINISH

6.02 Notes

Chapter 5: Coping with bit errors using error correction codes

BSC = Binary symmetric channel $\Rightarrow p$ of bit flip independent

Repetition Code

$010100 \rightarrow \underbrace{00}_n \quad \underbrace{11}_n \quad \underbrace{00}_n \dots \dots$

$$\text{Code rate} = \frac{1}{n}$$

Can find and correct error
if pattern is off with
most likely pattern.

$$0001000 \rightarrow 000000$$

For every useful message bit, we transmit n

Hamming Distance

of bit positions in which messages m_1 and m_2 differ

If $HD = 1 \Rightarrow$ can't tell ^{single} bit errors apart

Error Detection: $HD - 1$ errors

Error Correction: $\left\lfloor \frac{HD - 1}{2} \right\rfloor$ errors

Linear Block Codes

(n, k) codes $\Rightarrow k$ message bits produce n code bits $n \geq k$

$$\text{rate} = \frac{k}{n}$$

Arithmetic modulo 2:

$$\begin{array}{lll} 0+0=0 & 0+1=1 & \text{like XOR} \\ 1+1=0 & 1+0=1 & \end{array}$$

Thm: A code is linear iff the sum of any two codewords is another codeword

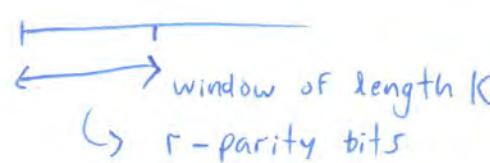
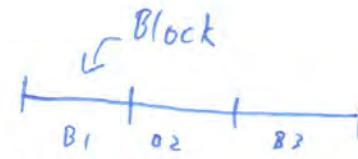
MHD = weight of code with least ^{non-zero} weight

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6.02 Recitation

Convolution Code

- linear code → compute parity only
- block vs stream
- sliding window, length K and r parity bits
- error correction, each message bit affects many parity bits



Parity Equations:

$$P_0[n] = x[n] + x[n-1] + x[n-2]$$

Window, $K=3$

$$P_1[n] = x[n] + x[n-1]$$

#parity bits $r=3$

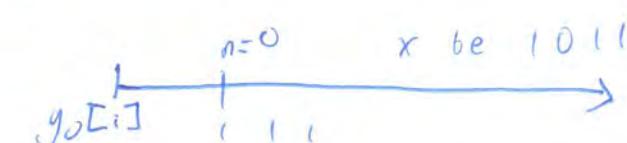
$$P_2[n] = x[n] + x[n-2]$$

rate = $1/r = 1/3$

view each parity equation w/ a generator polynomial $g_i[n]$

$g_0[n]$	are	$(1, 1, 1) \mid \begin{cases} x[n] \\ x[n-1] \\ x[n-2] \end{cases}$	\leftarrow what's actually happening
$g_1[n]$		$(1, 1, 0) \mid \begin{cases} x[n] \\ x[n-1] \\ \end{cases}$	
$g_2[n]$		$(1, 0, 1) \mid \begin{cases} x[n] \\ \\ x[n-2] \end{cases}$	

$$P_i[n] = \sum_{j=0}^{K-1} g_i[j] x[n-j] \quad \leftarrow \text{Flip in time}$$

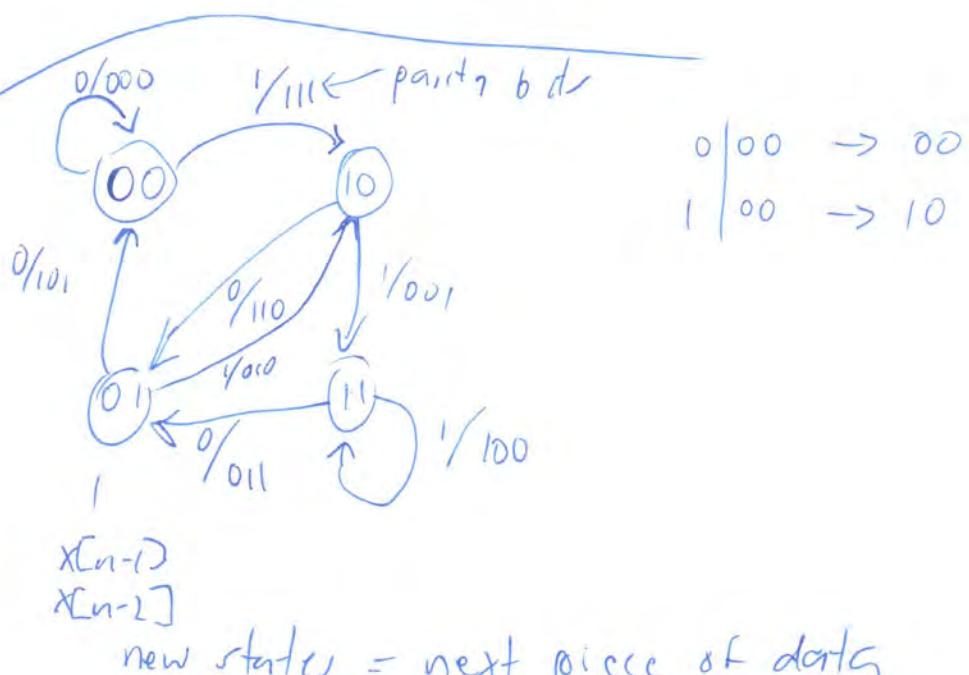


$x[-j]$	$n=0$
1 1 0 1	0 0 0 0
1 1 0 1	0 0 0 0
1 1 0 1	0
1 1 0 1	1

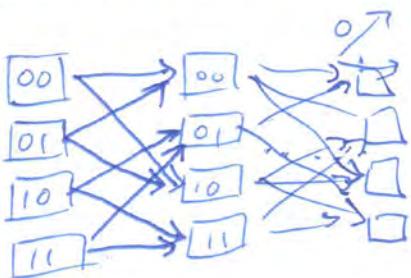
$x[1-j]$	$n=1$
$x[-j]$	$n=0$
$x[2-j]$	$n=2$
$x[3-j]$	$n=3$

1/2

$$\begin{array}{c} n=0 \ 1 \ 2 \ 3 \ 4 \ 5 \\ \text{xOR} \quad P_0 = 1 \ 1 \ 0 \ 0 \ 0 \ 1 \end{array}$$



Alternative view:



6.02 Notes

Lecture 6: Convolutional Coding

Convolutional Codes

- Very efficient
- Work on continuous stream of bits
vs linear breaks bit stream into blocks
- transmit non-systematic codes
 - ↳ only parity bit

Example code

$$\begin{aligned} p_0 &= x[n] + x[n-1] + x[n-2] \\ p_1 &= x[n] + x[n-2] \end{aligned}$$

Message: 11001101
(pad with zeroes at beginning)
0011001101

$$\begin{array}{l} n=0 \\ \begin{array}{ccccccc} x[n] = 1 & x[n-1] = 0 & x[n-2] = 0 & \Rightarrow & p_0[n_0] = 1 \\ 0011001101 & & & & p_1[n_0] = 1 \\ \uparrow & \uparrow & \uparrow & & p_0[n_1] = 0 \\ x[n-2] & : & : & & p_1[n_1] = 1 \\ & \uparrow & & & p_0[n_2] = 0 \\ & & & & p_1[n_2] = 1 \end{array} \end{array}$$

Transmit: 110101...
only parity bits!

Looking at 3 bits at a time

using a sliding window of width = 3 $\Rightarrow K = 3$

Each message bit protected by K parity bits

The greater K is \Rightarrow more redundancy
 \Rightarrow ↑ error correction possibilities

Generator: $g_0 = 111$ 1 in j th position if x
 $g_1 = 101$

$$P_i[n] = \sum_{j=0}^{K-1} g_i[j] x[n-j]$$

location of window
input bitstream as a function of j reversed in time

6.02 Notes

Lecture 7: Viterbi decoding

Encoding: Create path through trellis, sent parity bits

Decoding: Find most likely path that transmitter took

Viterbi Decoding

Branch Metric - HD b/w received parity bits and parity bits of transitions
↳ How likely it is that we did NOT make that transition

* At each timestep find most likely path to each state at that time

Path Metric - smallest sum of branch metrics, minimized over all message sequences that place the transmitter in state s at time i

Find $PM[s, i+1]$ using $PM[s, i]$ and BM for the incoming branches

Algorithm:

For each state s at time $i+1$

$a, b = \text{predecessor_states}(s)$ // states transmitter could've been at time i
// calculate branch metrics for each

$BM_a = \sim$

$BM_b = \sim$

$PM[s, i+1] = \min(PM[a, i] + BM_a, PM[b, i] + BM_b)$

// Keep track of predecessor state used to get to s in iteration $i+1$

6.02 Recitation

Quiz I Review

$$H = E = \sum p_i \log_2 \left(\frac{1}{p_i} \right)$$

LBS

$$2^{\overbrace{n-k}^{\text{\# parity bits}}} \geq \underbrace{n+1}_{\substack{\text{case where no error} \\ \text{H}}}$$

Single bit errors

possible configurations of parity bits

2010F Q1 P2

$$\cancel{2^{n-k}} \cancel{2^n} \cancel{2^k}$$

Correct for + errors

$$2^{n-k} \geq 1 + \binom{n}{1} + \binom{n}{2} + \dots + \binom{n}{d}$$

does a

$$\begin{matrix} (11, 5, 5) \\ n \quad k \quad d \end{matrix} \text{ work}$$

$$t = \left\lceil \frac{5-1}{2} \right\rceil = 2$$

Verify that

$$2^{11-5} \geq 1 + \binom{11}{1} + \binom{11}{2} ?$$

need ~~the~~ + errors

Convolutional Codes

F2011 Q1



6.02 Notes

Chapter 9: Noise

Additive white Gaussian Noise (AWGN)

$$y[k] = y_0[k] + w[k]$$

↑ ↑ ↑ Gaussian w/
 received sent noise mean = 0

PDF - Probability Density Function $f_w(w)$

$$P(w_0 \leq W \leq w_0 + dw) \approx f_w(w_0) dw$$

$$P(w_1 \leq W \leq w_2) = \int_{w_1}^{w_2} F_w(\alpha) d\alpha$$

mean \downarrow

$$N_w = \int_{-\infty}^{\infty} \alpha f_w(\alpha) d\alpha$$

$$\sigma_w^2 = \int_{-\infty}^{\infty} (\alpha - N_w)^2 f_w(\alpha) d\alpha$$

Note that

$$\int_{-\infty}^{\infty} f(\alpha) f_w(\alpha) d\alpha = 1$$

$Q(w)$: Probability of a standard Gaussian random variable taking values greater than w

$Q\left(\frac{w-\mu}{\sigma}\right)$: Probability of taking values greater than w for Gaussian distribution

Distance of w from mean μ , measured in standard deviations

Bit Error Rate (BER): probability that any given bit is in error

$$BER = Q\left(\frac{V_t - V_0}{2\sigma}\right) - Equal\ Prior\ Probabilities$$

$$BER_{\text{on-off}} = Q\left(\frac{V_p}{2\sigma}\right) \quad 0 \text{ or } V$$

$$BER_{\text{bipolar}} = Q\left(\frac{V_p}{\sigma}\right) \quad -V \text{ or } V$$

SNR - Signal to Noise Ratio

a ratio of signal power to noise variance

- Scaling wouldn't work as it would also scale the noise

Solution: Average k samples in the bit slot

↳ mean still = 0 but var = σ^2/k instead of σ^2

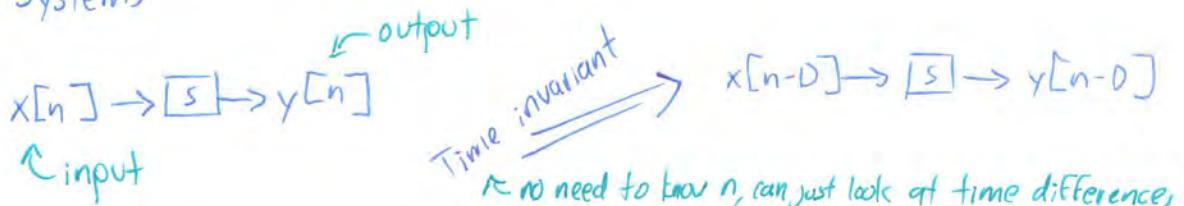
\Rightarrow SNR ↑ by Factor of k

6.02 Notes

Chapter 10

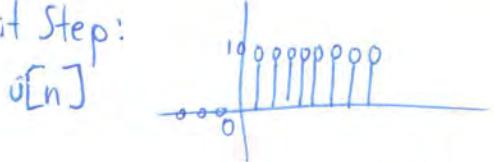
Linear Time Invariant

LTI Systems

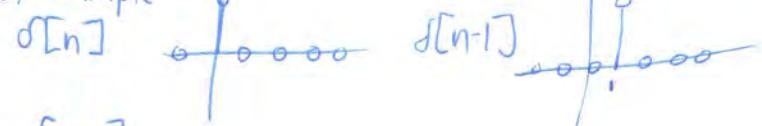


Causal Models: $y[n]$ only depends on past and present values of $x[\cdot]$

Unit Step:



Unit Sample



$$\delta[n] = u[n] - u[n-1]$$

$$u[n] \rightarrow [s] \rightarrow r[n]$$

unit step response

$$\delta[n] \rightarrow [s] \rightarrow h[n]$$

unit sample response

6.02 Notes

Chapter 11

(can characterize an LTI system by knowing $h[n]$)

- can write signal $x[n]$ as sum of scaled and shifted unit sample functions $\delta[n]$

$$x[n] = \dots + x[-1]\delta[n+1] + x[0]\delta[n] + \dots + x[k]\delta[n-k] + \dots$$

\downarrow \downarrow
-1 0 k

$$\Rightarrow x[n] = \sum_{k=-\infty}^{\infty} x[k]\delta[n-k] \rightarrow [S] \rightarrow y[n] = \sum_{k=-\infty}^{\infty} x[k]h[n-k]$$

since $\delta[n] \rightarrow [S] \rightarrow h[n]$

$$y[n] = x \star h = h \star x$$

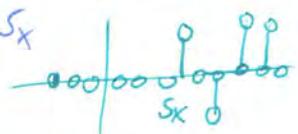
convolved with
convolution is commutative

$$y[n] = \sum_{m=0}^n h[m]x[n-m]$$

CONVOLUTION

With causal systems input is zero at beginning before a start time s_x

$$\Rightarrow y[n] = \sum_{k=s_x}^n x[k]y[n-k]$$



Convolution is also associative:

$$(h_2 \star h_1) \star x = h_2 \star (h_1 \star x)$$

Series:

$$x[n] \rightarrow [h_1] \rightarrow [h_2] \rightarrow y[n] = x[n] \rightarrow [h_2 \star h_1[\cdot]] \rightarrow y[n]$$

$$= x[n] \rightarrow [h_1 \star h_2[\cdot]] \rightarrow y[n] = x[n] \rightarrow [h_2] \rightarrow [h_1] \rightarrow y[n]$$

Parallel

$$x[n] \rightarrow [h_1[\cdot]] \rightarrow y_1[n] \quad x[n] \rightarrow [h_2[\cdot]] \rightarrow y_2[n]$$

$$y = y_1 + y_2 = (h_1 \star x) + (h_2 \star x) = (h_1 + h_2) \star x$$

Flip-Slide-Dot

$$y[n_0] = \sum_{k=-\infty}^{\infty} x[k] h[n_0 - k]$$

$\xrightarrow{h[-k+n_0]} = h \text{ reversed in time } h[-k]$
 $\text{shifted } n_0 \text{ to the right } + n_0$

Deconvolution

Process $y[n]$ through LTI Filter designed to cancel effect of channel

$$x[n] \rightarrow [h_1[\cdot]] \rightarrow y[n] \rightarrow [h_2[\cdot]] \rightarrow z[n] \quad \text{where } x[n] = z[n]$$

To do that we ~~we~~ need to satisfy:

$$(h_2 * h_1)[n] = \delta[n]$$

Fun fact since $z = h_2 * (h_1 * x) = (h_2 * h_1) * x = \delta_0 * x = x$
convolving any signal w/ unit sample yields that signal back

Find h_2 using Flip-Slide-Dot

$s[n]$ and $h[n]$

→ b/c of superposition and time invariance

$$\text{Since } s[n] = v[n] - v[n-1] \Rightarrow h[n] = s[n] - s[n-1]$$

So if causal system and $h[k] = 0$ and $s[k] = 0$

$$s[n] = \sum_{k=0}^n h[k]$$

o ↗ no need for convolution wooooooo8

Eye Diagrams

TODO



6.02 Notes

Chapter 12: Frequency Response of LTI Systems

Periodic if $x[n+p] = x[n]$ Fixed positive integer

smallest $p = \text{period}$

Periodic inputs \rightarrow periodic outputs
with frequency Ω
with frequency ω

Discrete Time Sinusoid

$$x[n] = \cos(\Omega_0 n + \theta_0)$$

angular Frequency

$\Omega_0 = 0 \Rightarrow$ lowest variation

$\Omega_0 = \pm\pi \Rightarrow (-1)^n \Rightarrow$ highest variation

* All action of interest happens at the $[-\pi, \pi]$ range

Complex Exponentials

$$e^{j\phi} = \underbrace{\cos \phi}_{\text{Real Component}} + j \underbrace{\sin \phi}_{\text{Imaginary Component}}$$

Euler

Recall:

$$c = a + jb$$

$$|c| = \sqrt{a^2 + b^2}$$

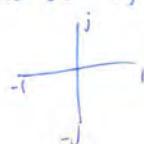
Polar form
 $c = |c| e^{j\angle c}$

$$\angle c = \arctan(b/a)$$

$$a = |c| \cos(\angle c)$$

$$b = |c| \sin(\angle c)$$

Point on the unit circle in the complex plane (radius=1) and at an angle of ϕ to positive real axis



More Identities

$$\cos \phi = \frac{1}{2}(e^{j\phi} + e^{-j\phi})$$

$$\sin \phi = \frac{1}{2j}(e^{j\phi} - e^{-j\phi}) = \frac{j}{2}(e^{-j\phi} - e^{j\phi})$$

$$\cos(A)\cos(B) = \frac{1}{2}(\cos(A+B) + \cos(A-B))$$

$$\cos(A \pm B) = \cos(A)\cos(B) \pm \sin(A)\sin(B)$$

Frequency Response

Derivation

$$y[n] = \sum_{m=-\infty}^{\infty} h[m] x[n-m]$$

From last chapter

$$= \sum_{m=-\infty}^{\infty} h[m] \cos(\omega_0(n-m) + \theta_0)$$

substitution

$$= \left(\underbrace{\left(\sum_{m=-\infty}^{\infty} h[m] \cos(\omega_0 m) \right)}_{C(\Omega)} \cos(\omega_0 n + \theta_0) + \left(\sum_{m=-\infty}^{\infty} h[m] \sin(\omega_0 m) \right) \sin(\omega_0 n + \theta_0) \right) \quad \text{Identity}$$

$$= C(\Omega) \cos(\omega_0 n + \theta_0) + S(\Omega) \sin(\omega_0 n + \theta_0)$$

$$H(\Omega) = C(\Omega) - jS(\Omega) = \dots +$$

Frequency Response

$$y[n] = |H(\Omega_0)| \cos(\omega_0 n + \theta_0 + \angle H(\Omega_0))$$

* can get $y[n]$ using only Frequency!

$$\text{Say } x[n] = A_0 e^{j(\omega_0 n + \theta_0)}$$

$$y[n] = \sum_{m=-\infty}^{\infty} h[m] x[n-m]$$

Last chapter

$$= \sum_{m=-\infty}^{\infty} h[m] A_0 e^{j(\omega_0(n-m) + \theta_0)}$$

Substitution

$$= \left(\underbrace{\left(\sum_{m=-\infty}^{\infty} h[m] e^{-j\omega_0 m} \right)}_{C(\Omega)} \right) A_0 e^{j(\omega_0 n + \theta_0)} + \left(\underbrace{\sum_{m=-\infty}^{\infty} h[m]}_{S(\Omega)} \right) A_0 e^{j(\omega_0 n + \theta_0)} j$$

Math

$$C(\Omega) - jS(\Omega) = H(\Omega)$$

$$X[n]$$

\Rightarrow If input $x[n]$ is a complex exponential $\Rightarrow y[n] = H(\Omega) e^{j\omega_0 n}$

$$A_0 e^{j\omega_0 n}$$

Properties of Frequency Response

If $h_0[n] = h[n-D]$ then $H_0(\Omega) = e^{-j\Omega D} H(\Omega)$ \Rightarrow Magnitude unchanged
 and $|H_0(\Omega)| = |H(\Omega)|$ Phase modified

Periodicity in Ω : $H(\Omega)$ repeats periodically on Ω axis with period 2π
 \Rightarrow Area of interest: $|\Omega| \leq \pi$

Lowest Frequency: $H(0) = \sum_{n=-\infty}^{\infty} h[n]$ so, if $\Omega=0$ we get constant DC input.
 ← DC gain of system

Highest Frequency: $H(\pm\pi) = \sum_{n=-\infty}^{n=\infty} (-1)^n h[n]$

Symmetry: For real $h[n]$ $C(\Omega)$ is an even function of frequency $\Rightarrow C(-\Omega) = C(\Omega)$
 $\Rightarrow |H(\Omega)|$ is even and $\angle H(\Omega)$ is odd

Real and Even: If $h[n]$ is even in time \Rightarrow Frequency response is purely real

Real and Odd: If $h[n]$ is odd in time \Rightarrow " " " immaginary

$H(\Omega)$ in Series: $H(\Omega) = H_2(\Omega)H_1(\Omega) = H_1(\Omega)H_2(\Omega)$

$$h[n] = (h_1 * h_2)[n] \Rightarrow H(\Omega) = H_1(\Omega)H_2(\Omega)$$

So, convolution in time \Rightarrow multiplication in frequency

$H(\Omega)$ in Parallel: $h[n] = (h_1 + h_2)[n] \Rightarrow H(\Omega) = H_1(\Omega) + H_2(\Omega)$

$H(\Omega) \Rightarrow h[n]$

We know $H(\Omega) = \sum_{m=-\infty}^{\infty} h[m] e^{-j\Omega m}$ Magical Math magic

$$h[n] = \frac{1}{2\pi} \int_{-\pi}^{\pi} H(\Omega) e^{j\Omega n} d\Omega$$

on

$$h[n] = \frac{1}{2\pi} \int_{2\pi n}^{2\pi(n+1)} H(\Omega) e^{j\Omega n} d\Omega$$

\nwarrow any contiguous interval of length 2π

So, $h[n]$ is represented as a weighted combination of a continuum of exponentials of the form $e^{j\Omega n}$ w/ frequencies Ω in a 2π range and associative weights $H(\Omega) d\Omega$

10/27/15

6.02 Recitation

Review:

$$x[n] \rightarrow \boxed{h[\cdot]} \rightarrow y[n] = \sum_k x[k] h[n-k]$$

$$x[n] = e^{j\omega n} \rightarrow \boxed{h[\cdot]} \rightarrow y[n] = H(\Omega) x[n] \text{ where } H(\Omega) = \sum_m h[m] e^{-jm\omega}$$

$x[n]$ is a sum of complex exponentials

$$x[n] \rightarrow \boxed{h[\cdot]} \rightarrow y[n]$$

$$X(\Omega) \quad H(\Omega) \quad Y(\Omega) \Leftarrow \text{DTFT}$$

$$X(\Omega) = \sum_m x[m] e^{-jm\omega m}$$

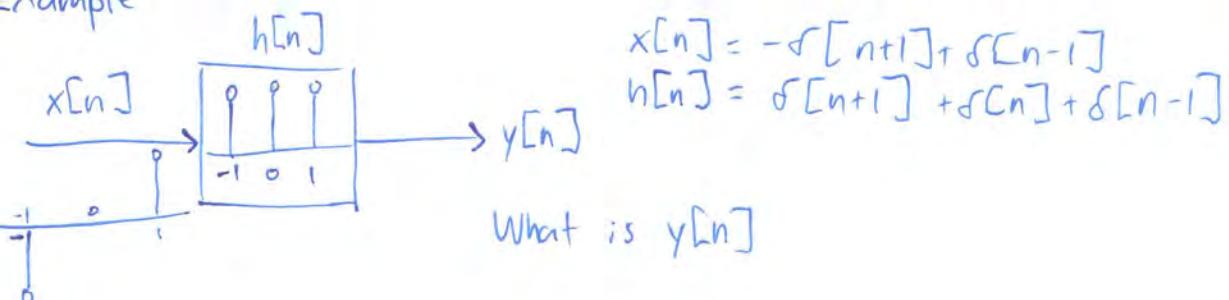
$$x[n] = \frac{1}{2\pi} \int X(\Omega) e^{jn\omega n} d\Omega$$

- Everything about $h[n] \leftrightarrow H(\Omega)$

- DTFT = another way to determine $y[n]$

$$Y(\Omega) = H(\Omega) X(\Omega) \Rightarrow y[n]$$

Example



AM One way to do it: Convolution

$$1) \text{ Flip } x[n]: x[-n] = \begin{matrix} 0 \\ 1 \\ 0 \end{matrix}$$

2) Calculate $\overset{x[n]}{h[n] - \text{shift}}$ by one and see where they align

$$h[n] = \begin{matrix} -1 & 0 & 1 \\ 0 & 0 & 0 \end{matrix} = x[n] * h[n] \Rightarrow y[n] = \delta[n+2] + \delta[n+1] + \delta[n-1] + \delta[n-2]$$

Can also do it:

$$h[n] = \delta[n+1] + \delta[n] + \delta[n-1]$$

1) Calculate Frequency Response

$$H(\Omega) = e^{-j\Omega} + \underset{m=0}{1} + e^{j\Omega}$$

$$= 2 \cos \Omega + 1 \quad \text{because of Euler}$$

$$x[n] = -\delta[n+1] + \delta[n-1]$$

$$H(\Omega) = \sum_m h[m] e^{-j\Omega m}$$

$$\sin \theta = \frac{-e^{-j\theta} + e^{j\theta}}{2j}$$

$$X(\Omega) = -e^{-j\Omega} + e^{j\Omega}$$

$$= -2j \sin \Omega$$

$$2) Y(\Omega) = H(\Omega) X(\Omega)$$

$$= (2 \cos \Omega + 1)(-2j \sin \Omega)$$

$$= -4j \cos \Omega \sin \Omega - 2j \sin \Omega$$

$$= -2j \sin 2\Omega - 2j \sin \Omega$$

double angle formula

$$Y(\Omega) = -e^{-j2\Omega} + e^{j2\Omega} + (-e^{-j\Omega} + e^{j\Omega}) \quad \text{Euler again}$$

if $\stackrel{\text{so}}{y[n]} = \frac{1}{2} \left[\dots \right] \Rightarrow$ same thing we got with convolution

NTFT Properties:

$$y[n] = x[n - n_0]$$

$$\Rightarrow y(\omega) = \sum_n x[n - n_0] \cdot e^{-j\omega n}$$

let $m = n - n_0 \Rightarrow$ change of variables

$$= \sum_m x[m] e^{-j\omega(m + n_0)}$$

$$= e^{-j\omega n_0} X(\omega)$$

\Rightarrow Shift in Time = Linear phase addition in Frequency

Initial Value:

$$X(0) = \sum_n x[n]$$

If $x[n]$ is real

$$X(\omega) = \sum_n x[n] e^{-j\omega n} = \sum_n x[n] [\cos(-\omega n) + j \sin(-\omega n)]$$

We know $\Re(X(\omega))$ is even and $\Im(X(\omega))$ is odd

$\Rightarrow |X(\omega)|$ is even and $\angle X(\omega)$ is odd

$x[n]$ is real X_{even}

$$X_{\text{even}} = \frac{1}{2} [x[n] + x[-n]]$$

$$X(\omega) = \frac{1}{2} \left(\sum_n x[n] e^{-j\omega n} + \sum_n x[-n] e^{-j\omega n} \right)$$

$$= \frac{1}{2} \left(\sum_n x[n] e^{-j\omega n} + \sum_m x[m] e^{+j\omega m} \right)$$

$$= \operatorname{Re}(X(\omega)) + j \cdot \operatorname{Im}(X(\omega)) \\ + \operatorname{Re}(X(\omega)) - j \cdot \operatorname{Im}(X(\omega))$$

\Rightarrow
 $X(\omega)$ is real

Similarly ...

$$X_{\text{odd}} \Rightarrow j \operatorname{Im}(X(\omega)) \quad (\text{all imaginary})$$

6.02 Notes

Chapter 13

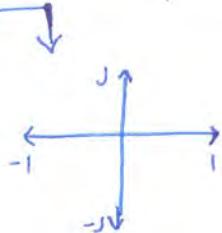
Fourier analysis and spectral representation of signals

Fourier analysis - decomposing signal into weighted sum of complex exponentials

DTFT of $x[n] = X(\omega) = \sum_{m=-\infty}^{\infty} x[m] e^{-j\omega m}$ = spectrum or spectral distribution or spectral content

DTFS of periodic $x[n]$

$$x[n] = \sum_{k=0}^{P-1} A_k e^{j\omega_k n}$$



$$\omega_k = k\Omega_1$$

$$\Omega_1 = \frac{2\pi}{P}$$

↑ Fundamental frequency

$\omega_k = k^{\text{th}}$ harmonic

$$A_k = A_{-k}^*$$

$A_{P/2} = A_{-P/2} = \text{all real}$

DTFS Analysis Equation:

$$A_k = \frac{1}{P} \sum_{n=0}^{P-1} x[n] e^{-j\omega_k n}$$

$$A_0 = \frac{1}{P} \sum_{n=0}^{P-1} x[n]$$

$$A_{P/2} = A_{-P/2} = \frac{1}{P} \sum_{n=0}^{P-1} (-1)^n x[n]$$

Even P:

$$x[n] = A_0 + \sum_{k=1}^{P/2} 2|A_k| \cos(\omega_k n + \angle A_k)$$

$$x[n] = A_0 + \sum_{k=1}^{(P-1)/2} 2|A_k| \cos(\omega_k n + \angle A_k)$$

$$y[n] = \frac{1}{P} \sum_{k=0}^{P-1} H(\omega_k) X_k e^{j\omega_k n} \Rightarrow Y_k = H(\omega_k) X_k$$

6.02 Notes

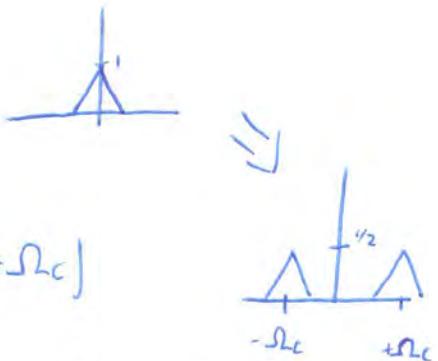
Chapter 14 Modulation and Demodulation

Modulation

$$X(\Omega) = \sum x[m] e^{-j\Omega m}$$

modulated signal $\rightarrow r[n] = x[n] \cos(\Omega_c n)$

$$R(\Omega) = \frac{1}{2} X(\Omega - \Omega_c) + \frac{1}{2} X(\Omega + \Omega_c)$$



Prevent overlapping: if X within $[-\Omega_m, +\Omega_m]$
 $\Rightarrow \Omega_c >= 2\Omega_m$

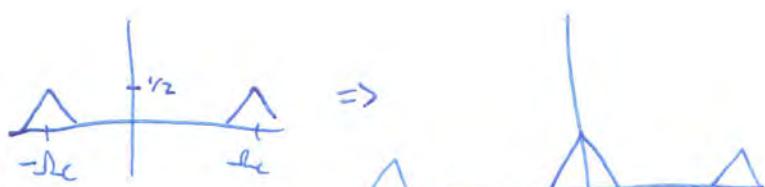
$$r_s[n] = x[n] \sin(\Omega_c n)$$

$$R_s(\Omega) = \frac{j}{2} X(\Omega + \Omega_c) - \frac{j}{2} X(\Omega - \Omega_c)$$

↑ imaginary and real components will swap

Demodulation

- Multiply by a cosine again



$$x[n] \cos(\Omega_c n) \cos(\Omega_c n)$$

$$d[n] = r[n] \cos(\Omega_c n) = x[n] \cos^2(\Omega_c n)$$

$$D(\Omega) = \frac{1}{4} X(\Omega + 2\Omega_c) + \frac{1}{2} X(\Omega) + \frac{1}{4} X(\Omega - 2\Omega_c)$$

Now take demodulated signal and filter it to get what we want.

low pass filter

$$\text{LPF}(\Omega) = \begin{cases} 2 & \text{if } \Omega \text{ in } [-\Omega_c, +\Omega_c] \\ 0 & \text{else} \end{cases}$$

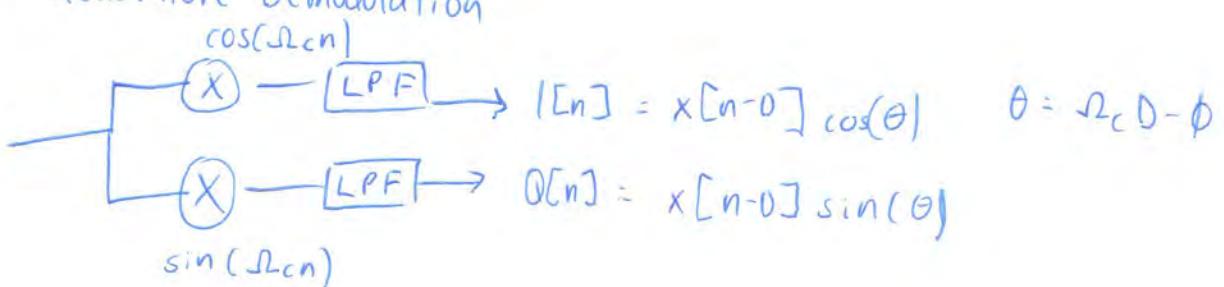
$$y(\Omega) = D(\Omega) * \text{LPF}(\Omega) = X(\Omega)$$

Phase errors

$$d[n] = x[n] \cos(\omega_c n) \cos(\omega_c n - \phi)$$

$$\Rightarrow y[n] = x[n] \underbrace{\cos(\phi)}_{\text{could be } 0}$$

Solution: Quadrature Demodulation

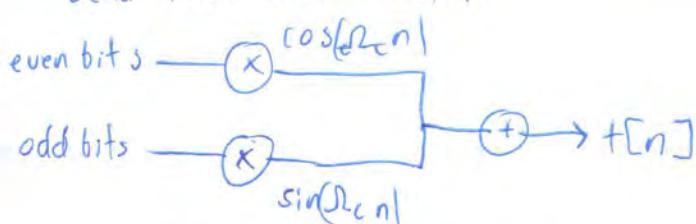


$$w[n] = I[n] + jQ[n]$$

$$|w[n]| = |x[n-0]|$$

QPSK (quadrature phase-shift keying)

- send twice much data



demodulate by cosine to get even bits
 " sin " odd "

Channel Sharing

• share signals

• modulate at different carriers frequency Ω_1, Ω_2

Pick \uparrow s.t.

$$|\Omega_2 - \Omega_1| \geq 2\Delta_m$$

o gap bw frequencies must be at least twice width of spectrum of any signal.

LPF cutoff should be $\sqrt{2}$ min separation b/w channels

Any signal can be represented as a sum of cosines and sines.

$$x(t) = \cos(2\pi f t + \phi)$$

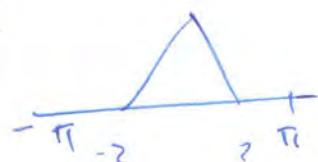
$$x[n] = x(nT) = \cos(2\pi f_n T + \phi) = \cos(\Omega \pi + \phi)$$

$$\Omega = 2\pi f_T = 2\pi f / f_s \Rightarrow f = \frac{f_s \Omega}{2\pi}$$

$x[n]$ from $v(t)$ sampled at $f_s = 1000 \text{ kHz}$



$x_1[n]$ from $x(t)$ sampled at $f_s = 500 \text{ kHz}$



$$X_1(\Omega) = \sum_n X_1[n] e^{-j\Omega n}$$

$$= \sum_n X[2n] e^{-j\Omega n}$$

$m = 2n \Rightarrow n = m/2$

$$= \sum_m X[m] e^{-j\Omega m/2}$$

$$= \sum_m X[m] e^{-j\frac{\Omega}{2}m} \Rightarrow X_1(\Omega) = X(\Omega/2)$$

Sample Question

Q. Two Tone Code

A) $f_s = 48 \text{ kHz}$

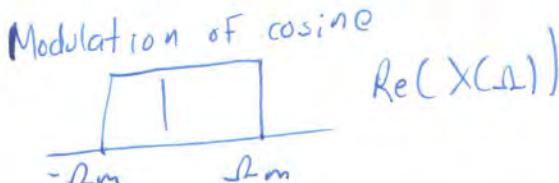
max frequency in $x(t) \Rightarrow 24 \text{ kHz}$

B) $f_1 = 1000 \text{ Hz} \Rightarrow \Omega_1$

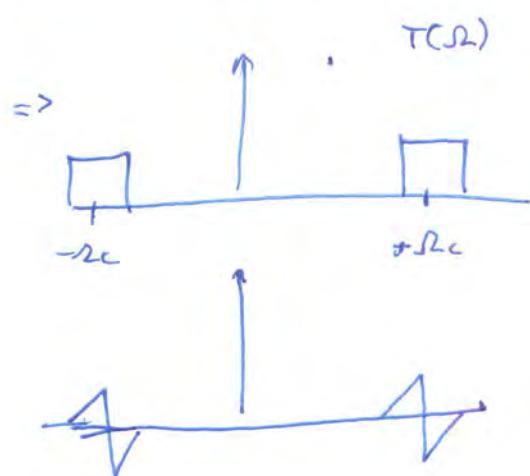
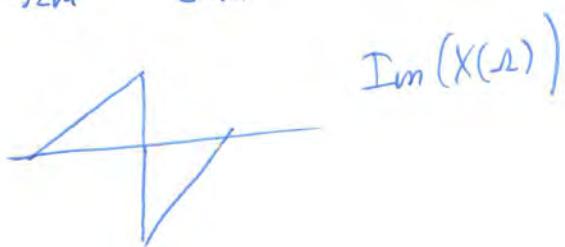
$f_2 = 1600 \text{ Hz} \Rightarrow \Omega_2$

$$P_1 = \frac{2\pi}{\Omega_1} = \frac{2\pi}{1000} = \frac{2\pi f_s}{2\pi f_1} = \frac{f_s}{f_1} = 48 \text{ samples}$$

$$P_2 = \frac{f_s}{f_2} = 3 \text{ samples}$$



$$+[n] = x[n] \cos(\Omega_c n)$$



$$x[n] \rightarrow +[n] = x[n] \sin(\Omega_c n)$$

$\sin(\Omega_c n) \text{ Euler bs}$

$$T(\Omega) = \sum_n (x[n] / 2j (e^{j\Omega_c n} - e^{-j\Omega_c n})) e^{-j\Omega n}$$

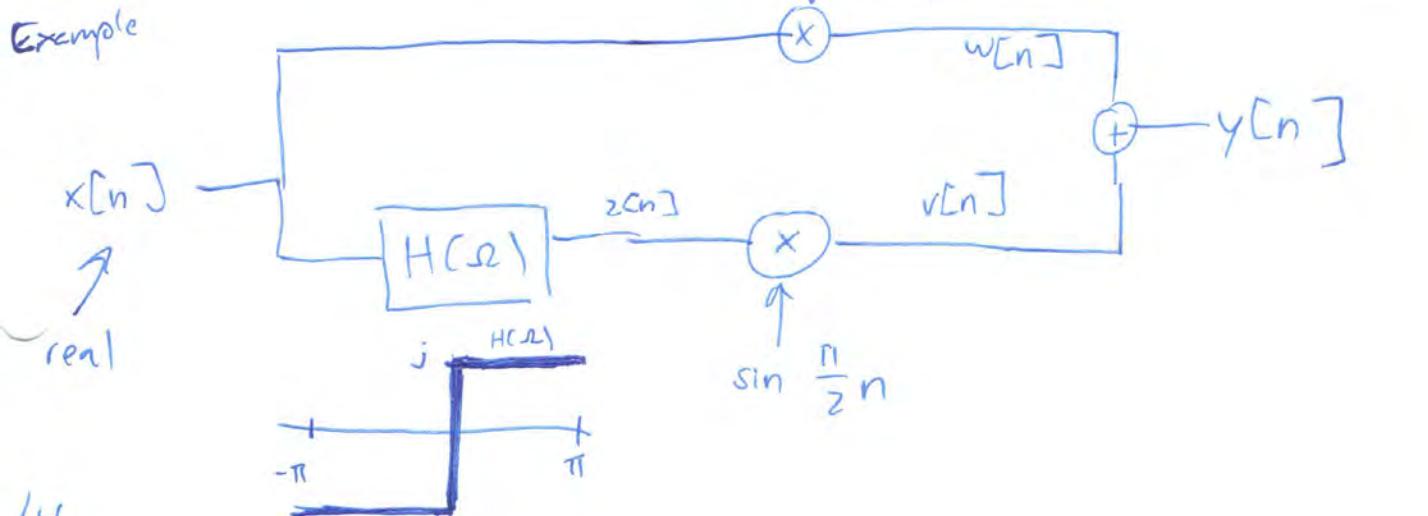
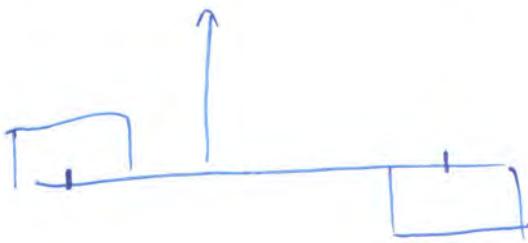
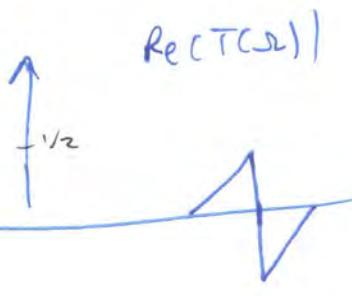
$$= \frac{j}{2} \sum x[n] [e^{-j(\omega + \omega_c)n} - e^{-j(\omega - \omega_c)n}]$$

$$= q \frac{j}{2} [x(\omega + \omega_c) - x(\omega - \omega_c)]$$

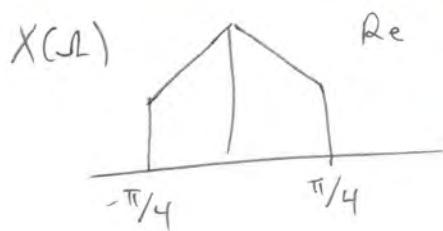
multiplied times \rightarrow

so Real becomes Imaginary \checkmark

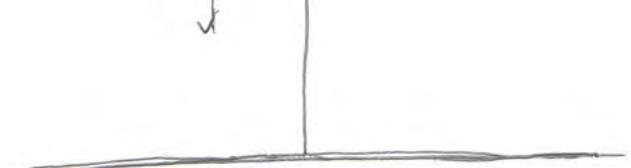
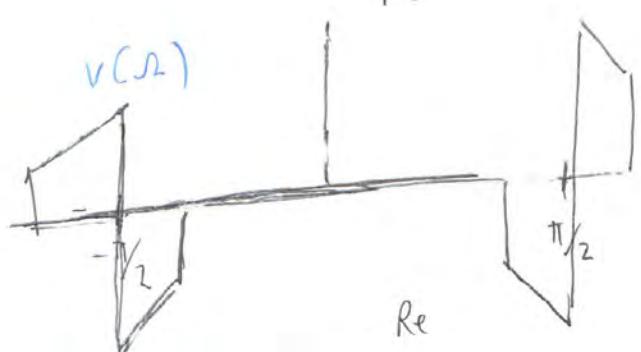
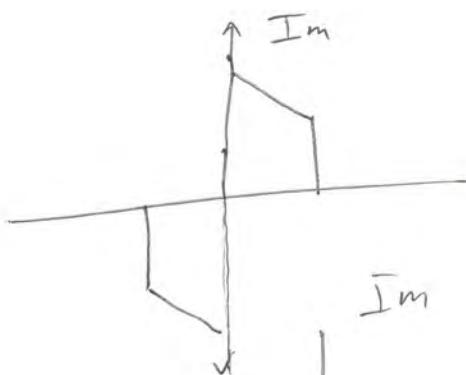
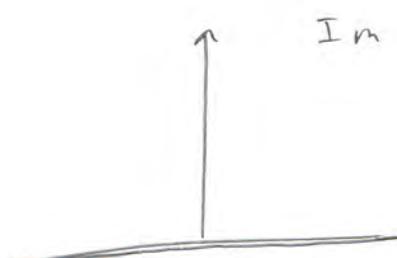
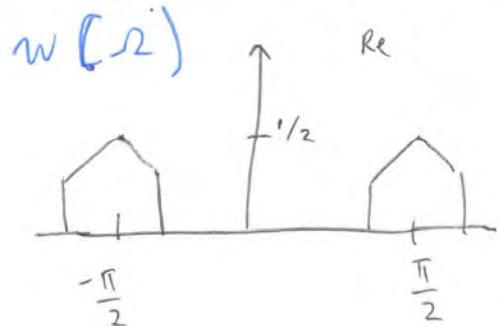
$$= \frac{j}{2} [Re(T(\omega)) + jIm(T(\omega)) - Re(x(\omega - \omega_c)) - Im(x(\omega - \omega_c))]$$



What does $X(\omega)$ look like?



What is spectrum of $w(\omega)$ modulated by $\cos \frac{\pi}{2} n$



11/4/15

6.02 Lecture

Fernando Trujano

Analyze Protocols that let nodes share a single channel

- 1) Collisions happen if two (or more) nodes transmit at once.
- 2) Nodes know when a collision occurs
- 3) Nodes might be able to hear each other (sometimes)

Goal: Design a media access protocol - Avoid collisions, good performance

- 1) utilization - use channel efficiency

$$U = \frac{\text{total throughput over all nodes}}{\text{max data rate of channel}}$$

\curvearrowright date/time
 \curvearrowright should be between $0 \rightarrow 1$

Ex:

Channel capacity: 10 Mbit/sec
 \curvearrowright throughput

N_1	1 Mbit/sec
N_2	2
N_3	2
N_4	3

$$U = \frac{(1+2+2+3) \text{ Mbit/sec}}{10 \text{ Mbit/sec}} = .8$$

- 2) Fairness: Divided evenly

X_i = throughput that node i achieves

N = # nodes in network

$$\frac{1}{N} \leq F \leq 1$$

\curvearrowright Perfect fairness

$$F = \frac{\sum\limits_{i=1}^N X_i}{N \cdot \sum\limits_{i=1}^N X_i^2}$$

- 1) Time is divided into slots of equal weight
- 2) Nodes can start transmitting at beginning of a slot
- 3) All packets are same size = an int # of slots

TDMA

- 1) Each node gets a unique index $[0, N-1]$
- 2) Node i can transmit in timeslot t iff $t \text{ and } N = i$
- Fair
- Bad if work load is skewed \rightarrow waste time

Slotted ALOHA

Assume: each packet is 1 slot long

If a node is backlogged, it sends a packet in the next timeslot w/ prob P .

If $P \uparrow \uparrow \rightarrow$ too many collisions
 $P \downarrow \downarrow \rightarrow$ no good utilization

$$P \cdot (1-P)^{N-1} \cdot \binom{N}{1} = N$$

Stabilized ALOHA.

1. On Success

$$p = \min(2p_i, 1)$$

2. On collision - decrease

$$p = \max(p_{2i}, p_{in})$$

9/17/15

6.02 Recitation

Fernando Trujano

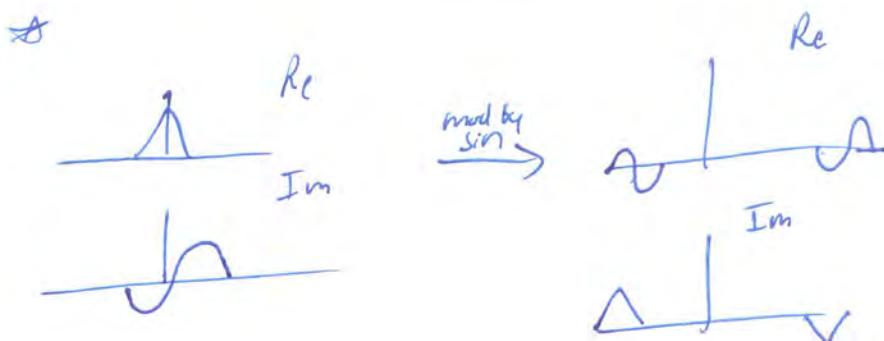
Fourier Transform DTFT

$$X(\omega) = \sum_m x[m] e^{-j\omega m}$$

$$x[n] = \frac{1}{2\pi} \int_{-\pi}^{\pi} X(\omega) e^{jn\omega} d\omega$$

ENAVS

$$\text{BER}_{\text{on-off}} = Q\left(\frac{\Delta V}{2G}\right)$$

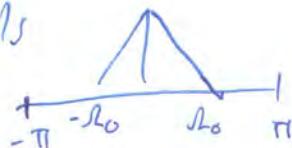


Why Gaussians?

Central limit Theorem:

averaged time signals from distribution \Rightarrow Gaussians

- Bandlimiting signals



• Not Aliasing?

Properties that relate $h[n] \rightarrow H(\omega)$

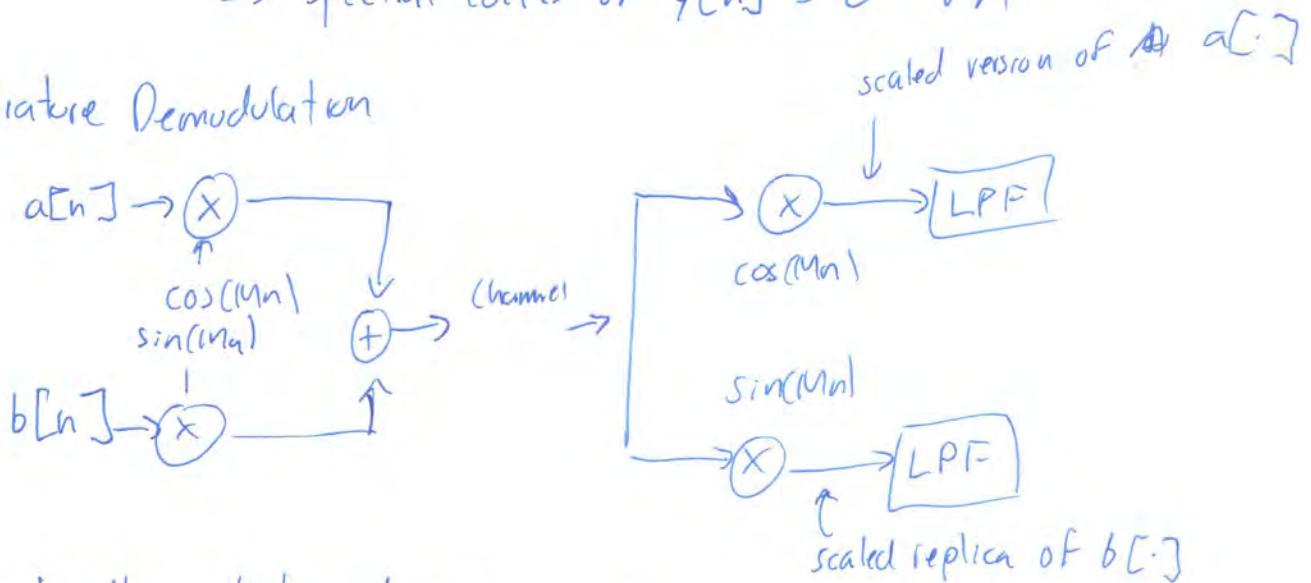
$$H(0) = \sum h[n] \quad H(\pi) \sum (-1)^n h[n]$$

Modulate with cosine \rightarrow demodulate w cosine and phase shift
Solution: Demodulate by both cosine and sin.
 \hookrightarrow Quadrature Demodulation.

Modulation

- Demodulate by something with $\pi/2$ phase difference with mod
 \Rightarrow components at origin cancel each other out
 \Rightarrow spectral coeff of $y[n] = 0 \forall n$

Quadrature Demodulation



- handle arbitrary delay lag and

- frequency: samples/second

~~Things to believe~~



$$e^{j\pi} = -1$$

$$y[n] = y(0) + y(1)e^{jn} + y(2)e^{j2n} + \dots$$

Frequency of $j = \frac{\pi}{2}$
 $-j = \frac{3\pi}{2}$

IF you see j in $x[n]$ replace e^{jn}

$$\cos \phi = \frac{e^{j\phi} + e^{-j\phi}}{2} \quad e^{j\phi} = \cos \phi + j \sin \phi$$

$$\sin \phi = \frac{e^{j\phi} - e^{-j\phi}}{2j}$$



$a + jb$

$$\text{amp} = \sqrt{a^2 + b^2} e^{j \arctan(b/a)}$$

magnitude \angle angle

$$H(\omega) = H^*(-\omega)$$

Fourier transform

$$\mathcal{F}(\omega) = \sum_{m=-\infty}^{\infty} \omega e^{-j\omega m}$$



rectify

Absolute
Values

Take
average

What is
mean &
what
is 0

Find
white signal
start,

(convert 5 samples
to single bit)

Error
counter

Answer
 $\text{avg} = 0$

6.02 Lecture #17 Notes

Networks

Analyze protocols that allow many nodes to share a single channel.

⇒ Design a media access protocol - which nodes will use to access the channel that avoids collisions and ensures good performance.

Utilization - using the channel efficiently

$$U_{\text{channel}} = \frac{\text{total throughput over all nodes}}{\text{max data rate of channel}}$$

amount of data transferred over time / data/time

$U=1 \Rightarrow$ using 100% of channel

↪ U cannot > 1 or < 0

Fairness - divide capacity equally among nodes requesting to use it.

$$F = \left(\sum_{i=1}^N x_i \right)^2 / N \cdot \sum_{i=1}^N x_i^2$$

N : total # nodes
 x_i : throughput of node i

$$\frac{1}{N} \leq F \leq 1$$

one node getting all throughput Perfectly fair

Abstractions:

- each packet is same size
- collisions happen when 2 senders send @ same time
- nodes can detect when collision occurred

Time Division Multiple Access (TDMA)

- Give each node a unique index
- Node i can transmit in time slot t iff $t \bmod N = i$
 - * Node transmits every N time slots

Analysis:

- No collisions
- Fair if all nodes have data to send
- Utilization
 - Good if every node sends all the time.
 - Otherwise \Rightarrow wasted time slots
- Not good with different loads

ALOHA

Slotted ALOHA - assume each packet takes 1 timeslot to transmit

- If node has stuff to send, send it on the next time slot with probability p .

large $p \Rightarrow$ collisions small $p \Rightarrow$ under utilization

$$P(\text{successful transmission}) = p \cdot (1-p)^{N-1}$$

Ways to have exactly one sender \Rightarrow one sender \nwarrow all others don't send

$$U_{\text{slotted-ALOHA}} = N p \cdot (1-p)^{N-1}$$

maximize U : 1) Take derivative $= 0 \Rightarrow \max \Rightarrow p = 1/N$

$$p = 1/N \Rightarrow U = (1-1/N)^{N-1}$$

Problem: N changes so optimal p changes, but nodes can't independently update p .

solution:

Stabilized ALOHA

- Let nodes dynamically adjust p .

- 1) IF transmission collides set $p = p/2$ multiplicative decrease
- 2) IF transmission works set $p = \min(2p, 1)$

Problem: If a node has several failed attempts p will be really low
Solution: and it will be hard to come back.

- 1) If fail: $p = \max(p_{\min}, p/2)$
- 2) If YES: $p = \min(2p, 1)$

Problem: Nodes can capture a channel and transmit in burst, which is unfair to others at that time. This is because p can get really high.

Solution:
1) On Fail : $p = \max(p_{\min}, p/2)$
2) On success : $p = \min(2p, p_{\max})$

CSMA - Carrier Sense Multiple Access

If nodes can hear each other, check if someone is transmitting before transmitting. \rightarrow carrier sense
More in PS7 modify stabilized ALOHA

Other Goals

- 1) Upper bound on wait to make transmission
- 2) Handle variability in node work. (How often they send)
- 3) Scalability

Introduction to Networking

Circuit Switching

Protocol

- 1) Setup-Phase: configure state with mappings in switch
- 2) Data-Transfer-Phase: Switches forward data
- 3) Teardown-Phase: Delete phase in switch

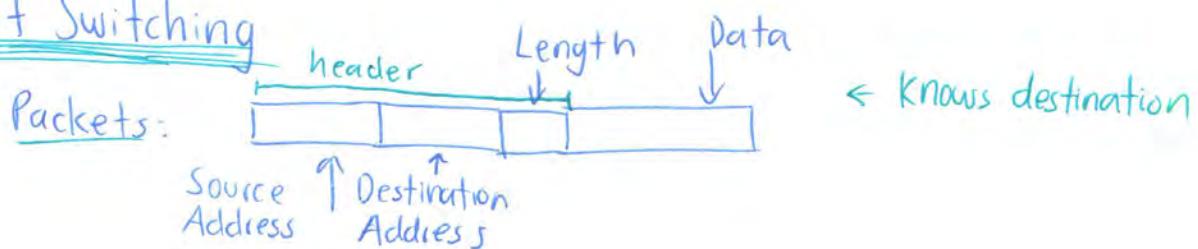
To share same link use TDM. (Time Division Multiplexing)

- Divide time into N slots "packet"
- Slot 0: send frame for conversation 0
- ...

Problems w/ TDM:

- 1) Bad when data arrives in bursts
what each connection requires
- 2) Bad when communications send more/less than R bits/sec
one have more than c/R communications.
what the link we are sharing can handle
- 3) If a link in middle of network goes down, all circuits using that link are broken.

Packet Switching



More resilient to failure, (fixes #3 in Circuit switching)
↪ packet can take longer route if node missing.

Switching: Store incoming packets in queue. Determine outgoing link of first packet using header. ↪ drop if queue full

Benefits:

- 1) Links can send any available packet \Rightarrow no wasted link capacity
- 2) Variable data rates "as needed"
- 3) No setup / teardown phase
- 4) Aggregating multiple sources of traffic smooths out bursts \rightarrow Queues absorb bursts !

Bad: 1) Packets can be dropped 2) Packets may arrive in different order

Queues

Queues add more delays:

Queuing delay: How long packets wait in queue

Processing delay: Time spent by switch to process packet

Little's Law

↪ Average time packet spends on queue

$$N = \lambda D$$

λ mean # of packets in Q.

D mean delay per packet

Circuit Switching

- Guaranteed Rate
- Wasted link capacity if bursty
- Path established before sending data
- All data in connection follows 1 path
- No reordering; constant delay, no dropped packets
- Can't route around failures

Packet Switching

- No guarantees
- More efficient
- Send data immediately
- Different packages could follow different paths
- Packets may be reordered, delayed or dropped
- Can route around failures

6.02 Lecture #19 Notes

Addressing, Forwarding, Routing w/ Failures

How do nodes determine routes to destination?

- allow each switch to know how to get to any other node
 - with minimum cost.

Addressing: How to name a node. Addresses are globally unique.

Costs: Links ^{can} have costs/weights:
1) Non-negative
2) Triangle inequality
3) Can change

$$\text{cost}(A \rightarrow B) + \text{cost}(B \rightarrow C) \leq \text{cost}(A \rightarrow C)$$

Routing Tables: Table to store information about routes.
If packet is destined for X, where to send the packet next.

Forwarding: When switch receives packet:

- 1) Check packet header for destination address, dest
- 2) Look up routing-table[dest] to get outgoing link
- 3) Add packet to queue
- 4) wait and send when in front.

Distributed Routing

Each node runs routing protocol and determines own routes

General Structure:

- 1) Nodes learn about neighbors via HELLO structure
 - Nodes send HELLO packets to each neighbor
- 2) Nodes learn about other nodes via advertisements
 - Sent by other nodes containing information about other nodes
 - Use advertisements to determine connectivity ^{and cost} to reachable nodes
- 3) Determine min-cost routes with information gathered

Difference b/w distance vector and link-state comes in ^{↳ "integrate" step}

- Types of advertisements sent and who they are sent to

- Way nodes integrate advertisements

Distance-Vector routing protocol

“Talk to neighbors about everyone”

Each node sends: $(dst, cost)$

Advertisement: List of all nodes node knows about and cost to nodes to: only its neighbors

Integrate step: Bellman Ford

advertisement from $Y \rightarrow X$

For each $(dst, cost)$:

If X is already using Y to get to dst : update cost

else if Y can provide better path to dst : update routing and cost

Link State

“Talk to everyone about neighbors”

Advertisements:

Each node sends:

List of neighbors and link cost to neighbors

to: neighbors who then forward to neighbors \Rightarrow All network

“flooding”

Integrate step: Run Dijkstra $\xleftarrow{\text{relax edges, find all shortest paths}} O(N \log N + E)$

Note that advertisement only changes if neighbor link cost changes

6.02 Notes

Lecture #20 Routing with Failures

2. What happens when parts of the network fail, and how do nodes deal w/ that.

Distance Vector: Each ad sent to neighbors \Rightarrow $2L$ ads DVR less bandwidth

Link State: Flood network. $\Rightarrow N2L$ ads

Protocol Correctness

- routing tables should reflect correct network topology

1) Route validity: $\#N$ if dst in routingTable and $\text{routingTable}[dst] = K$

\Rightarrow There is path from N to dst whose first hop is K .

\hookrightarrow invalid \Rightarrow disconnected network info not yet propagated

Route loops

2) Path visibility: Every router that has a path to a dst, learns at least one valid route to dst.

\checkmark route valid + visible

Eventual convergence: routing protocol will eventually converge^{at t}, could fail in between

link state routing failures

- If no ads are lost \Rightarrow protocol will converge

lost ads can lead to routing loops but rare b/c flooding

Link state \Rightarrow Fast convergence

Distance Vector Routing Failures

Recall in DVR nodes don't know entire topology

- Count to Infinity Problem

Attempt #1: Split Horizon

- Don't send path to dst to neighbor through which you would go

- Avoids only two-hop loops

Attempt #2: Path vector routing

- Include min cost path in ad. Node N will integrate only if N not in path

- converges and does not count to infinity

Distance Vector: uses less bandwidth

- use for very small networks
loop-free

Link-State: converges quickest, but large overhead

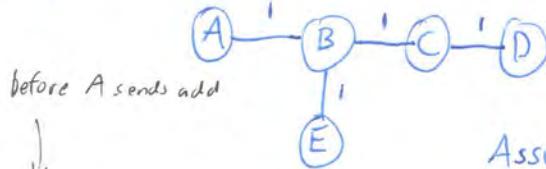
- use for small networks

Path-Vector

- use for large networks

- ① Distance Vector Routing: Count-to-infinity
- ② Link State Routing: Routing Loop
- ③ Chapter 18 Problems 5, 8, 10, 11

Count to Infinity



what E would send to neighbors

$$E: \begin{bmatrix} A, 2 \\ B, 1 \\ C, 2 \\ D, 3 \\ E, 0 \end{bmatrix}$$

Order of Ads

- i) A
 - ii) B
 - iii) C
 - iv) D
 - v) E
- $\Rightarrow A \rightarrow B$ count to inf
.....
 \nwarrow No count-to-infinity

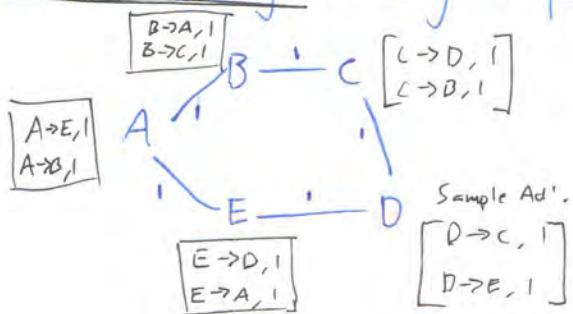
link BE fails $\Rightarrow B \leftarrow C$ count to infinity.

Assume long time passed, everyone has its routing table.

Assume new link costs get updated if old.

EVENT	ACTION
① BE Fails	
② C sends ad	$\boxed{@B} E: B \rightarrow E, \infty$ $\boxed{@C} E: C \rightarrow B, 2$ $\boxed{@A} C: A \rightarrow B, 2$ $\boxed{@B} E: B \rightarrow C, 3$ ↳ D thinks C can get to E in 2 steps, updates its path to go through C.
③ A sends ad	$\boxed{@B} E: B \rightarrow C, 3$
④ B sends ad	$\boxed{@C} E: C \rightarrow B, 4$ $\boxed{@A} E: A \rightarrow B, 4$

② Link State Routing - Routing Loops



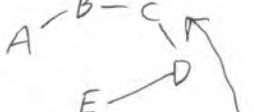
Ad's flooded through network

D receives

A->D lost

E->D lost

\Rightarrow D thinks network looks like this

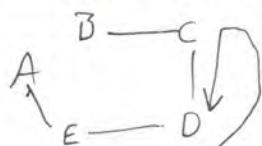


C receives

A->C lost

B->C lost

\Rightarrow C thinks network looks



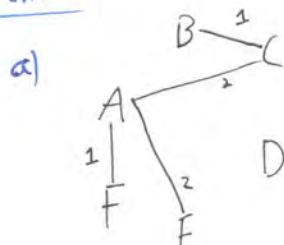
now if D wants to send to A.

sends it to C first

C sends it back to D

Chapter 18 Problems

Problem 5

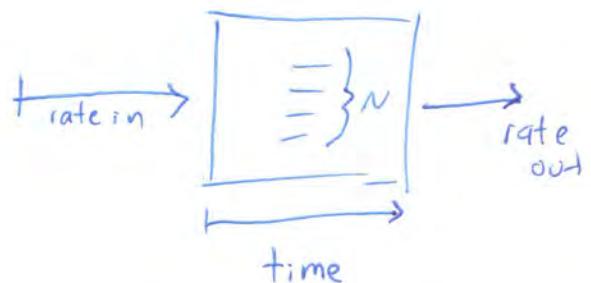


b) D->F, 3

6.02 Recitation

Little's Law

- Abstract any queueing system



- N - [packets] - average # in queue
- λ - [packets/time] - flow rate
- D - [time] - average delay time of packet in queue.

$$N = \lambda D$$

$$\sum_{t=0}^T n(t) / T = \lambda \sum_{t=0}^T D_t / \lambda$$

FTC

Reliable Data Transport Protocols

Stop and Wait

- unique IP for each packet, Sender sends K

Receiver sends ACK
Sender sends K+1

* ~~could have done~~

- Wait some time to receive ACK, otherwise send again.

Reliable =
o all data
o in order
o one copy

↳ could lead to duplicates

Examples from chapter 19

See notes

P of mutually exclusive events adds
 $P(A \text{ or } B) = P_A + P_B$
 If not mutually exclusive:
 $P(A \text{ or } B) = P_A + P_B - P(A \text{ and } B)$

P of independent events multiply
 Conditional Probability
 $P(A | B) = P(A \text{ and } B) / P(B)$
 $\Rightarrow P(A | B) = p(a)$
 $\Rightarrow A, B \text{ independent} \Rightarrow P(A \text{ and } B) = p_A p_B$

$P(A \text{ and } B) = P(A | B) P(B)$
 $= P(B | A) P(A)$
 $= P(B | A) P(A)$

Bayes Rule: $P(A | D) = \frac{P(D | A) P(A)}{P(D)}$

Information
 Expected length = $\sum p_a I_a$
 $I_a = \text{information associated with revealing output} = \log_{\frac{1}{p_a}}$

$I_a = \log_2 \left(\frac{1}{p_a} \right) \text{ bits}$
 $H = \text{entropy} = \sum p_a I_a = \sum p_a \log_2 \left(\frac{1}{p_a} \right)$

Example: A, B independent
 $I_a + I_b = \log \left(\frac{1}{p_a} \right) + \log \left(\frac{1}{p_b} \right)$
 $= \log \left(\frac{1}{p_a p_b} \right) = \log \left(\frac{1}{p_a p_b} \right)$

Example: $p = p$ of bit flip. (codeword N bits)
 $P(\text{error}) = 1 - p(\text{correct}) = p(\text{error})$

\therefore from N outcomes to $(N-1)$ outcomes
 $I_a = \log_2 \left(\frac{N-1}{N} \right) N-1$

Example: A, B independent
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Source codes - remove redundancy and compress data.
 Entropy H is a lower bound on amount of info that must be sent.
 Prefix tree \Rightarrow all symbols at the end of tree

Huffman Encoding
 Algorithm:
 - Know probabilities in advance
 - More probable = shorter codeword
 - sort nodes on p_i
 - merge two least probable
 - combine and repeat till done

LW compression
 Encoding:
 table = initialize_table() // 0 to 255 ASCII
 string = get input symbol
 while input symbols:
 symbol = get input symbol
 if string + symbol in table:
 string = string + symbol
 else:
 output code for string
 table.add(string + symbol)

Systematic Codes:
 message bits parity bits
 Linear Block Codes:
 sum of 2 codewords = another CW
 Obtain CW by doing linear trans
 of message bits with $\begin{pmatrix} 1 & 0 \\ 0 & 1 \end{pmatrix}$
 (codeword = message \times $\begin{pmatrix} 1 & 0 \\ 0 & 1 \end{pmatrix}$)
 $\therefore C = dG + K$ generator matrix

Rectangular Codes:
 parity matrix
 - calculate parity bit of each row i and column j
 Rate = $\frac{r}{n+r}$
 $G = \begin{pmatrix} 1 & 0 & 1 & 0 & 1 & 0 \\ 1 & 0 & 1 & 0 & 1 & 0 \\ 0 & 1 & 0 & 1 & 0 & 1 \\ 0 & 0 & 1 & 0 & 1 & 0 \\ 0 & 0 & 0 & 1 & 0 & 1 \\ 0 & 0 & 0 & 0 & 1 & 0 \end{pmatrix}$
 $K = \begin{pmatrix} 1 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 \\ 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & 1 \end{pmatrix}$

Not efficient so \Rightarrow
 Hamming Codes
 - correct single bit errors with min # of parity bits. $n = 2^{k-1}$
 $p_1 = d_1 + d_2 + d_3 \rightarrow$ protects all bits w/ $k=4$
 $p_2 = d_1 + d_2 + d_3 + d_4 \quad \dots$
 $p_3 = d_1 + d_2 + d_3 + d_4 \quad \dots$

Linear Block with 2^k codewords
 is K_n # bits in each codeword

If the dictionary fills up, empty it before adding new element

Just because string s was added to the table does not mean it was sent, but all prefixes of s have

Output entry
 table.add(string + entry[0])
 string = entry

Output code for string
 while (codes):
 code = read next code
 if not table[code]:
 entry = string + string[0]
 else:
 entry = table[code]
 output entry

Output String
 while (codes):
 code = read next code
 if not table[code]:
 entry = string + string[0]

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 if not table[code]:
 entry = string + string[0]

Efficient way to decode any linear block code.
 Code rate = $\frac{\# \text{ useful message bits}}{\# \text{ total bits sent}}$
 Hamming Distance
 = distance b/w codewords
 of equal length
 $H(A, B) = A + B \bmod 2$

Detect + errors if:
 $MHD \geq k+1 \Rightarrow$ corrected = $MHD-1$
 (correct + errors if:
 $MHD \geq 2k+1 \Rightarrow$ corrected = $\left\lfloor \frac{MHD-1}{2} \right\rfloor$

Algorithm:
 1) For each error pattern e:
 compute syndromes $H_e^T = S_e$
 and add to syndrome table
 2) For each received word
 compute $H_r^T = S$

3) Find k such that $S_e = S$ und
 correct error $C = R + e$
 \Leftrightarrow flip bits where 1 on e

4) Decode the codeword by taking first k bits
 be combined to give zero vector.

A has identical rows $\Rightarrow H$ has identical cols $\Rightarrow MHD = 2 \Rightarrow$ linearly dependent cols of H ie can

be combined to give zero vector.

Ex: $R = \begin{pmatrix} 1 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 \\ 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & 1 \end{pmatrix}$
 $D = \begin{pmatrix} 1 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 \\ 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & 1 \end{pmatrix}$
 $S = \begin{pmatrix} 1 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 \\ 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & 1 \end{pmatrix}$

Identical MHD. List all codewords (all combinations of 3 data bits and corresponding parity bits)

Decode 10101
 $E = \begin{pmatrix} 1 & 0 & 1 & 0 & 1 \end{pmatrix}$
 $E = D + D_2 + D_3 + D_4 = 1$
 $D = \begin{pmatrix} 1 & 0 & 0 & 0 & 0 \end{pmatrix}$
 $D_2 = \begin{pmatrix} 0 & 1 & 0 & 0 & 0 \end{pmatrix}$
 $D_3 = \begin{pmatrix} 0 & 0 & 1 & 0 & 0 \end{pmatrix}$
 $D_4 = \begin{pmatrix} 0 & 0 & 0 & 1 & 0 \end{pmatrix}$

Any path through function of j condition: $\sum_{j=0}^{k-1} g_{i,j} \times c_{i,n-j} = x_{i,n}$

Tellis View: $P_{i,n} = \sum_{j=0}^{k-1} g_{i,j} \times c_{i,n-j}$

label arrows w/ parity bit sent

Upper arrow $\Rightarrow 0$ received lower arrow $\Rightarrow 1$ received

transitions. For each message bit, make transition and send corresponding parity bits. End message w/ $K-1$ 0's to return to starting state

Free Distance: Smallest weight nonzero codeword = d-free.

(can correct $d-1$ errors if they are far enough apart)

Viterbi Decoding
 Find most likely path that transmitter took over all messages.

Branch Metric = H_D the received parity bits and parity bits of transitions

\Rightarrow how likely we did NOT make that transition

Path Metric - smallest sum of BM minimized over all messages.

Algorithm: Trace back by hand

For each state s at time $i+1$
 $a, b = \text{predecessor states}(s)$

Work on continuous stream of bits
 Transmit only parity bits

Encoding:
 $x_{i,1} = 11 \Rightarrow P_0 = x_{i,0} + x_{i,n-1} + x_{i,n-2}$
 $x_{i,2} = 101 \Rightarrow P_1 = x_{i,1} + x_{i,n-2}$

Message: $m_{i,0}m_{i,1}m_{i,2} \dots$ (bad w/ errors) $\Rightarrow 0011001100$
 $m_{i,0} = 1 \Rightarrow P_0 = 1$
 $m_{i,1} = 0 \Rightarrow P_1 = 1$
 Transmit: 110101...

Decoding:
 $H^T = \sum_{j=0}^{n-1} H_{j,j} H = [A^T | I]$

Received codeword
 $r_{i,0} = 1 \Rightarrow P_0 = 1$
 $r_{i,1} = 0 \Rightarrow P_1 = 1$

Message: $m_{i,0}m_{i,1}m_{i,2} \dots$

Algoithm: Trace back by hand

For each state s at time $i+1$
 $a, b = \text{predecessor states}(s)$

$BMA = BM(a), BMb = BM(b)$

$BMc, BMd = \min(BMA, a) + BMb, BMd = \min(BMC, b) + BMc$

Free Distance - Find by finding smallest weight path that leaves 0 and returns to 0 showing long it takes the decoder to "settle"

Reducing - Agree to only send a subset of parity bits. Increases Rate

Noise - PDF

Probability Density Function

$$P(w_1 < W \leq w_2) = \int_{w_1}^{w_2} f_W(w) dw$$

$$\text{Mean} = \mu_w = \int_{-\infty}^{\infty} w f_w(w) dw$$

$$\text{Variance} = \sigma^2 = \int_{-\infty}^{\infty} (w - \mu_w)^2 f_w(w) dw$$

$$Q(\omega) : \text{Probability of Gaussian RV taking values greater than } \omega.$$

$$Q\left(\frac{\omega - \mu}{\sigma}\right) : P \text{ of taking val} > \omega$$

Bit Error Rate (BER)

\hookrightarrow P. that any given bit is in error

$$\text{BER} = Q\left(\frac{V_t - V_o}{2\sigma}\right)$$

$$\text{BER}_{\text{bipolar}} = Q\left(\frac{V_t}{\sigma}\right)$$

$$\text{BER}_{\text{unipolar}} = Q\left(\frac{V_t}{2\sigma}\right)$$

Signal to Noise Ratio SNR

signal power/noise variance

Average k samples in bit slot

$N=0$ still but σ^2/k instead of σ^2

\hookrightarrow SNR \uparrow by factor k.

$$\text{In-off} : 0 \rightarrow 0 \text{ or } 1 \rightarrow 1 \text{ bipolar: } -V \rightarrow 0$$

$$\text{Single sample BER} = Q\left(\frac{V_t}{\sigma}\right)$$

$$\text{M samples BER} = Q\left(\frac{MV_t}{\sigma}\right)$$

bit energy in bit slot = sum of squares of M values sent. Energy $\propto V^2$

Energy bipolar = MV^2

$$\text{Energy} = MV^2$$

LTI Models

(\hookrightarrow Linear-Time Invariant)

$$\text{Periodic: } x[n+\rho] = x[n]$$

$$\text{smallest } \rho = \text{period}$$

$$\text{Unit Step: } u[n]$$

$$x[n] \rightarrow [S] \rightarrow y[n]$$

$$\text{Causal Model: } y[n] \text{ only depends on past and present values of } x[-]$$

$$\text{Unit Sample: } d[n]$$

$$\delta[n] = u[n] - u[n-1]$$

$$v[n] \rightarrow [D] \rightarrow s[n]$$

$$\delta[n] \rightarrow [D] \rightarrow h[n]$$

$$\text{can characterize LTI system with } \mathcal{T}$$

$$x[n] = \sum_{k=-\infty}^n x[k] \delta[n-k]$$

$$y[n] = \sum_{k=-\infty}^n x[k] h[n-k]$$

$$\Rightarrow y[n] = x[n] * h[n]$$

$$y[n] = x[n] \xrightarrow{[h]} y[n]$$

$$x[n] \xrightarrow{[h]} y[n]$$

$$x[n] \xrightarrow{[h_1 + h_2]} y[n]$$

$$= x[n] \xrightarrow{[h_1]} \xrightarrow{[h_2]} y[n]$$

$$x[n] \xrightarrow{[h]} y[n]$$

Frequency Response

Modulation

Demodulation

$$x[\Delta t] = \sum x[n] e^{-j\omega n \Delta t}$$

$$r[n] = x[n] \cos(\Omega n)$$

$$R[\Delta t] = \frac{1}{2} X(\Omega + \Delta \omega) + \frac{1}{2} X(\Omega - \Delta \omega)$$

$$R[\Delta t] = \frac{1}{2} X(\Omega + \Delta \omega) - \frac{j}{2} X(\Omega - \Delta \omega)$$

$$d[\Delta t] = \cos(\Delta \omega n)$$

$$x[n] \cos(\Omega n) \cos(\Delta \omega n)$$

$$d[\Delta t] = r[n] \cos(\Omega n)$$

$$D[\Delta t] = \frac{1}{4} X(\Omega + 2\Delta \omega) + \frac{1}{2} X(\Omega)$$

$$+ \frac{1}{4} X(\Omega - 2\Delta \omega)$$

$$X(n) \text{ is a sum of complex exp.}$$

$$e^{j\phi} = \cos \phi + j \sin \phi$$

$$e^{j\phi} = \cos(\Omega n + \theta_0)$$

$$\cos \phi = \frac{j}{2} (e^{j\phi} - e^{-j\phi})$$

$$\sin \phi = \frac{1}{2} (e^{j\phi} - e^{-j\phi})$$

$$\cos(A) \cos(B) = \frac{1}{2} (\cos(A+B) + \cos(A-B))$$

$$\cos(A \pm B) = \cos(A) \cos(B) \pm \sin(A) \sin(B)$$

$$H(\Omega) = \sum_{n=-\infty}^{\infty} h[n] e^{-jn\Omega \Delta t}$$

$$H(\Omega) = \sum_{n=-\infty}^{\infty} h[n] e^{-jn\Omega \Delta t}$$

$$H(\Omega) = H(\Omega) e^{j\Delta \omega n \Delta t}$$

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$$\text{Unit Step: } u[n]$$

$$x[n] \rightarrow [S] \rightarrow y[n]$$

$$\text{Causal Model: } y[n] \text{ only depends on past and present values of } x[-]$$

$$\text{Unit Sample: } d[n]$$

$$\delta[n] = u[n] - u[n-1]$$

$$v[n] \rightarrow [D] \rightarrow s[n]$$

$$\delta[n] \rightarrow [D] \rightarrow h[n]$$

$$\text{can characterize LTI system with } \mathcal{T}$$

$$x[n] = \sum_{k=-\infty}^n x[k] \delta[n-k]$$

$$y[n] = \sum_{k=-\infty}^n x[k] h[n-k]$$

$$\Rightarrow y[n] = x[n] * h[n]$$

$$x[n] \xrightarrow{[h]} y[n]$$

Frequency Response

Modulation

Demodulation

$$x[\Delta t] = \sum x[n] e^{-j\omega n \Delta t}$$

$$r[n] = x[n] \cos(\Omega n)$$

$$R[\Delta t] = \frac{1}{2} X(\Omega + \Delta \omega) + \frac{1}{2} X(\Omega - \Delta \omega)$$

$$R[\Delta t] = \frac{1}{2} X(\Omega + \Delta \omega) - \frac{j}{2} X(\Omega - \Delta \omega)$$

$$R[\Delta t] = \frac{1}{2} X(\Omega + \Delta \omega) + \frac{j}{2} X(\Omega - \Delta \omega)$$

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DTFT Mathing

Frequency division Multiplexing (FDM)

$$x(t) = \sum x[n] \xrightarrow{\text{IF } x(t)=0 \text{ } x[n]}$$

$$\Rightarrow \text{all signals cancelling out.}$$

$$x[n] = \sum x[n] \delta[n]$$

$$\xrightarrow{\text{IF } x[n]=0}$$

$$\Rightarrow \text{all signals cancelling out.}$$

$$\xrightarrow{\text{IF } x[n]=0}$$

$$\xrightarrow{\text{IF } x[n]=0}$$

$$\xrightarrow{\text{IF } x[n]=0}$$

Given
work
explanation

$$x[n] = 2e^{jn\pi}$$

$$y[n] = 6[n-2] + 2e^{j(n-4)} + e^{jn-6}$$

Find $h[n]$: Note that we can get $\delta[n]$ by scaling $x[n]$ by $\frac{1}{2}$ and shifting 1 timestep. We also know that $\delta[n] \rightarrow [] \rightarrow h[n]$ so,

$$h[n] = .5\delta[n-1] + \delta[n-3] + .5\delta[n-5]$$

*Find $H(\Omega)$ by taking DFT of $h[n]$:

We know that a shift in time \Rightarrow Linear phase addition in frequency.

So if $y_0[n] = h[n-0] \Rightarrow H_0(\Omega) = e^{-j0\Omega} H(\Omega)$

$$H(\Omega) = .5e^{-j\Omega} + e^{-j3\Omega} + .5e^{-j5\Omega}$$

*Find $H(\Omega)$ by taking DFT of $y[n]$ and $x[n]$

$$X(\Omega) = 2e^{-j\Omega} \quad Y(\Omega) = e^{-j2\Omega} + 2e^{-j4\Omega} + e^{-j6\Omega}$$

$$Y(\Omega) = X(\Omega)H(\Omega) \Rightarrow H(\Omega) = Y(\Omega)/X(\Omega)$$

$$H(\Omega) = .5e^{-j\Omega} + e^{-j3\Omega} + .5e^{-j5\Omega}$$

① Rewrite $H(\Omega)$ in Form: $H(\Omega) = e^{j\Delta\Omega} A(\Omega)$

② Factor out an $e^{-j2\Omega}$ that two remaining have the same $\angle \neq 2\pi$ in this case

$$H(\Omega) = e^{-j3\Omega} \left(\frac{1}{2}e^{j2\Omega} + j\frac{1}{2}e^{-j2\Omega} \right)$$

$$= e^{-j3\Omega} \left(1 + \frac{1}{2}(e^{j2\Omega} + e^{-j2\Omega}) \right)$$

③ "wiggly" equation to substitute a sin or cosine. $\cos\Phi = \frac{1}{2}e^{j\Phi} + e^{-j\Phi}$

$$H(\Omega) = e^{-j3\Omega} (1 + \cos(2\Omega))$$

$$\text{So, } \alpha(\Omega) = -3\Omega \quad A(\Omega) = 1 + \cos(2\Omega)$$

Note that $A(\Omega) = |H(\Omega)|$ since $A(\Omega) \geq 0 \forall \Omega$

a) when is $H(\Omega) = 0$? when $\cos(2\Omega) = -1$
which happens at $\Omega = \pm \frac{\pi}{2}$

Given
work
explanation

$$H(\Omega) = 2 \quad \alpha(\Omega) = 1 + (-1)^{\Omega+5}$$

$$H(\pi) = -3 \quad \text{Find } y[n]$$

We know: $y(\Omega) = H(\Omega)x(\Omega)$

Step 1: Find $y(\Omega)$

$$y(\Omega) = H(\Omega)x(0)$$

Step 1.1: Find $X(0)$

- look at component in $x[n]$ with $\Omega=0$

$$x[n] = [] + (-1)^{n+5}$$

④ Rewrite in complex exp.

$$[] = e^{j0} \quad \text{Get amplitude}$$

$$|e^{j0}| = 1 = X(0) \quad \text{Get phase}$$

Step 2: $y(0) \Rightarrow y[n] \leftarrow$

Recall that:

$$y[0] = Y(0) + Y(1)e^{j\pi} + Y(2)e^{j2\pi} + \dots$$

$$y[1] = Y(0)e^{j\pi} + \dots$$

$$= 2 + \dots$$

Step 3: Get $y(\pi)$

$$y(\pi) = \underbrace{H(\pi)x(\pi)}_{-3}$$

Step 3.1: Find $X(\pi)$

- look at component in $x[n]$ with $\Omega=\pi$

$$x[n] = [] + (-1)^{n+5}$$

④ Rewrite in complex exp

$$e^{j\pi} \Rightarrow [e^{j\pi}]^{\frac{1}{n+5}} = e^{j\pi n} e^{j\pi 5}$$

$$e^{j5\pi} = -1 \quad \text{Amplitude}$$

$$Y(\pi) = H(\pi)x(\pi) = (-3)(-1) = 3$$

Step 4: Inverse DFT

$$y[n] = Y(0)e^{j0\pi} + Y(1)e^{j\pi\pi}$$

$$\Rightarrow y[n] = 2 + 3e^{j\pi\pi}$$

Networks

Protocols

Switching

Routing

Reliable Transport

Quiz III

Utilization = $\frac{\text{total throughput}}{\text{max data rate}}$
 $U=1 \Rightarrow 100\% \text{ utilization}$

$$\text{Fairness} = \left(\sum_{i=1}^N x_i \right)^2 / N \sum_{i=1}^N x_i^2$$

Workload
of node i / Total # nodes

Total # nodes

- No collisions
- Fair if all nodes have data to send.
- Utilization bad w/ skewed workload

to send, send on next + with prob P.

Successful transmission = $P \cdot \frac{\# \text{backlogged nodes}}{N-1} \cdot (1-P)$

One node sends > others don't

$$U = Np \cdot (1-p)^{N-1} \quad p \neq \frac{1}{N} \Rightarrow U = \left(1 - \frac{1}{N}\right)^{N-1}$$

Stabilized ALOHA: let nodes dynamically adjust p.

On Fail $\Rightarrow p = \max(p_{\min}, p/2)$

On Success $\Rightarrow p = \min(2p, p_{\max})$

(CSMA: Carrier Sense Multiple Access)

- Check if someone is transmitting before sending

Utilization of TDMA = 1 if all backlogged

Throughput of TDMA node = $\frac{1}{N}$

" slotted ALOHA = $p(1-p)^{N-1}$

CSMA vs Stabilized ALOHA:

CSMA has q utilization and more

this is because CSMA nodes maintain

high p \Rightarrow so there are more

sending attempts.

Carrier Sense: Allows nodes to check if

others starting in current timeslot

Contention window: Nodes guaranteed

to attempt transmission in bounded

time \Rightarrow more fair.

TDMA packets removed from queue

at rate $\frac{1}{N}$

TDMA utilization with non uniform

usage $= \frac{\sum k}{N} + \frac{N \sum r_i}{N-k+1} \cdot p_{\text{node i sends}}$

First to transmit $\Rightarrow U=1$

all others aren't backlogged all the time

Circuit Switching:

- Guaranteed Rate

- Wanted link capacity, if bursty

- Path established before sending

- All data follows 1 path

- No reordering, constant delay

- Can't route around failures

Packet Switching

packet: 

source \rightarrow address \rightarrow data

header \rightarrow dest address

length \rightarrow drop if full

queue incoming packets in

or first packet using header.

Queues - Absorb bursts

Add queuing delay and processing

delay. Little's Law

Avg time

packet spreads

on queue.

mean # of average

packets in queue

switch rate \rightarrow per packet

which packets

No guarantees.

- More efficient

- Send data immediately

- Different packages could different paths

- Packets can be reordered, delayed or dropped

- Can route around failures.

Total time on link = propagation time

+ transmission time

Total time = propagation + trans + queue + processing

In Link State: nodes forward messages even if they are last in chain

Dijkstra:

W = all nodes; routingTable[Self] = self

costTable[Self] = 0; costTable[Others] = ∞

while W not empty:

U = node in W min cost to so far

W remove(U)

for each neighbor V of U:

d = costTable[U] + cost(U,V)

if d < costTable[V]: # found shorter

costTable[V] = d; costTable[Others] = ∞

routingTable[V] = routingTable[U]

with in network

Time to notice failure

DU = Z_HELLO + ADDREQ + L_L

L_SA_Acks: 2 over each link?

longest path

with m users: Wmax = $(RTT_{min} \cdot \beta + K)/m$

N = m * RTTmin

L \leftarrow L bits/burst

D \leftarrow W/L

packets in queue

queuing delay

Wmax = $RTT_{min} \cdot \beta + K$

greater would experience packet loss

R \leftarrow bottleneck bandwidth

with m users: Wmax = $(RTT_{min} \cdot \beta + K)/m$

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R \leftarrow bottleneck bandwidth

with m users: Wmax = $(RTT_{min} \cdot \beta + K)/m$

N = m * RTTmin

L <