

Channel Models

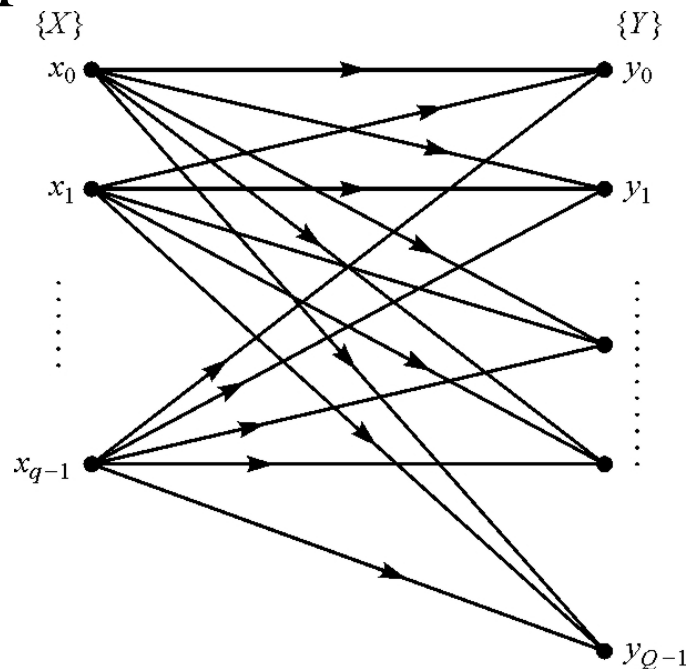
- Discrete Memoryless Channel
 - Discrete-discrete
 - Binary channel, M-ary channel
 - Discrete-continuous
 - M-ary channel with soft-decision (analog)
 - Continuous-continuous
 - Modulated waveform channels (QAM)

Discrete Memoryless Channel

- Discrete-discrete
 - Binary channel, M-ary channel

Probability transition matrix

$$\mathbf{P} = \begin{bmatrix} P(Y = y_1 | X = x_1) & \cdot & \cdot & \cdot \\ \cdot & \cdot & P(y_i | x_j) = p_{ji} & \cdot \\ \cdot & \cdot & \cdot & \cdot \\ \cdot & \cdot & \cdot & P_{q-1Q-1} \end{bmatrix}$$



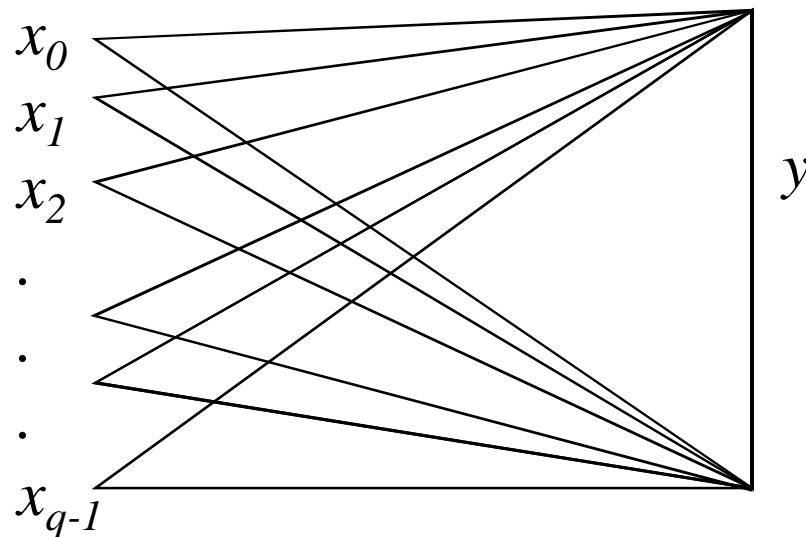
Discrete Memoryless Channel

- Discrete-continuous
 - M-ary channel with soft-decision (analog) output

AWGN

$$p(y | x_k) = \frac{1}{\sqrt{2\pi\sigma}} e^{-(y-x_k)^2/2\sigma^2}$$

$$\mathbf{P} = \begin{bmatrix} p(y | X = x_1) \\ \text{graph of } p(y | X = x_1) \\ p(y | X = x_k) \\ \text{graph of } p(y | X = x_k) \end{bmatrix}$$



Discrete Memoryless Channel

- Continuous-continuous
 - Modulated waveform channels (QAM)
 - Assume Band limited waveforms, bandwidth = W
 - Sampling at Nyquist = $2W$ sample/s
 - Then over interval of $N = 2WT$ samples use an orthogonal function expansion:

$$\begin{array}{ccc} x(t) & \longrightarrow & \textcircled{+} & \longrightarrow & y(t) \\ = \sum_{i=1}^N x_i f_i(t) & & \uparrow & & = \sum_{i=1}^N y_i f_i(t) \\ & & n(t) & & \\ & & = \sum_{i=1}^N n_i f_i(t) & & \end{array}$$

Discrete Memoryless Channel

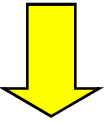
- Continuous-continuous
 - Using orthogonal function expansion:

The diagram shows a block representing a discrete memoryless channel. It consists of a circular adder with a plus sign inside. An input signal $x(t)$ enters the adder from the left. A noise signal $n(t)$ enters the adder from the bottom. The output signal $y(t)$ exits the adder to the right.

$$\begin{aligned} x(t) & \longrightarrow \text{Adder} \longrightarrow y(t) \\ & \uparrow \\ & n(t) \end{aligned}$$
$$\begin{aligned} x(t) &= \sum_{i=1}^N x_i f_i(t) \\ n(t) &= \sum_{i=1}^N n_i f_i(t) \\ y(t) &= \sum_{i=1}^N y_i f_i(t) \\ &= \sum_{i=1}^N \left\{ \int_0^T y(t) f_i^*(t) dt \right\} f_i(t) \\ &= \sum_{i=1}^N \left\{ \int_0^T [x(t) + n(t)] f_i^*(t) dt \right\} f_i(t) \\ &= \sum_{i=1}^N [x_i + n_i] f_i(t) \end{aligned}$$

Discrete Memoryless Channel

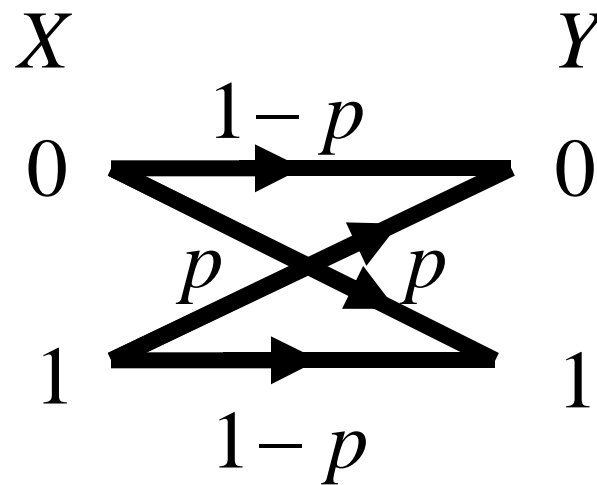
- Continuous-continuous
 - Using orthogonal function expansion get an equivalent discrete time channel:

$$y_1 = x_1 + n_1 \leftarrow \text{Gaussian noise}$$

$$p(y_i | x_i) = \frac{1}{\sqrt{2\pi\sigma_i}} e^{-(y_i - x_i)^2 / 2\sigma_i^2}$$

x_1 y_1
.
.
.
.
 x_N y_N

Capacity of binary symmetric channel

- BSC $X = \{0,1\}$ $Y = \{0,1\}$

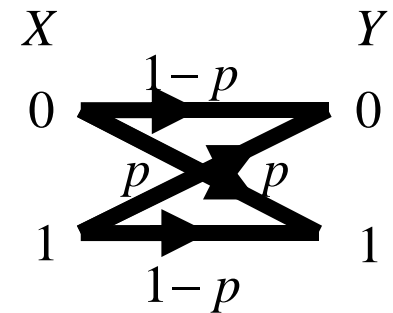


$$\mathbf{P} = \begin{bmatrix} 1-p & p \\ p & 1-p \end{bmatrix}$$

Capacity of binary symmetric channel

- Average Mutual Information

$$\begin{aligned}
 I(X;Y) &= P(X=0)P(Y=0|X=0)\log\frac{P(Y=0|X=0)}{P(Y=0)} + \dots \\
 &\quad P(X=0)P(Y=1|X=0)\log\frac{P(Y=1|X=0)}{P(Y=1)} + \dots \\
 &\quad P(X=1)P(Y=0|X=1)\log\frac{P(Y=0|X=1)}{P(Y=0)} + \dots \\
 &\quad P(X=1)P(Y=1|X=1)\log\frac{P(Y=1|X=1)}{P(Y=1)} \\
 &= P(X=0)(1-p)\log\frac{1-p}{(1-p)P(X=0)+pP(X=1)} + \dots \\
 &\quad P(X=0)p\log\frac{p}{pP(X=0)+(1-p)P(X=1)} + \dots \\
 &\quad P(X=1)p\log\frac{p}{(1-p)P(X=0)+pP(X=1)} + \dots \\
 &\quad P(X=1)(1-p)\log\frac{1-p}{pP(X=0)+(1-p)P(X=1)}
 \end{aligned}$$



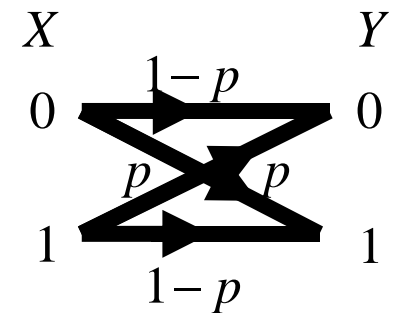
Capacity of binary symmetric channel

- Channel Capacity is Maximum Information

– earlier showed: $\max(I(X;Y)) \Leftrightarrow P(X=1) = P(X=0) = \frac{1}{2}$

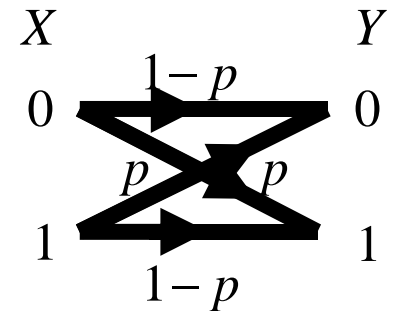
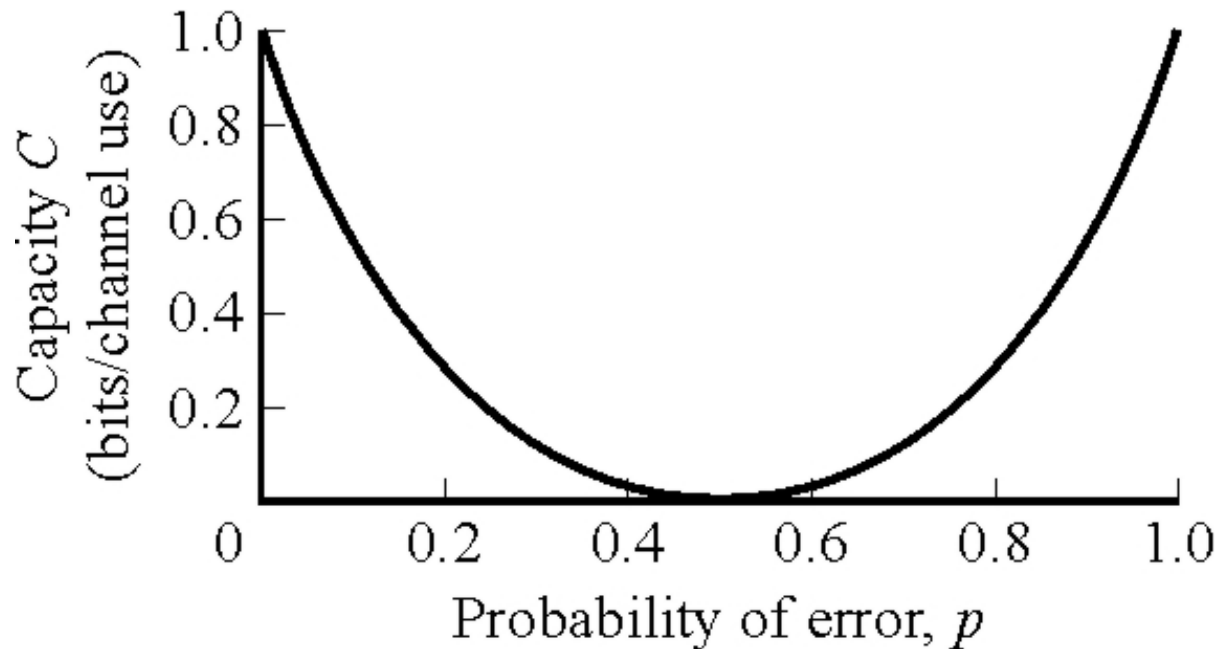
$$C = \max(I(X;Y)) = \left[\begin{aligned} &P(X=0)(1-p) \log \frac{1-p}{(1-p)P(X=0) + pP(X=1)} + \dots \\ &P(X=0)p \log \frac{p}{pP(X=0) + (1-p)P(X=1)} + \dots \\ &P(X=1)p \log \frac{p}{(1-p)P(X=0) + pP(X=1)} + \dots \\ &P(X=1)(1-p) \log \frac{1-p}{pP(X=0) + (1-p)P(X=1)} \end{aligned} \right]_{P(X=1)=P(X=0)=\frac{1}{2}}$$

$$= (1-p) \log 2(1-p) + p \log 2p$$



Capacity of binary symmetric channel

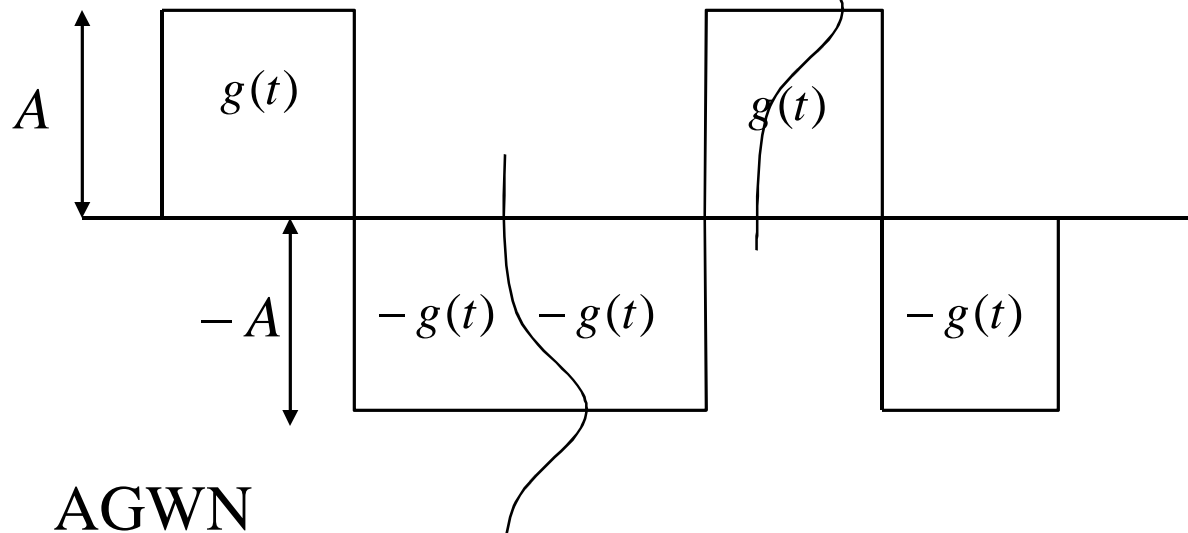
- Channel Capacity
 - When $p=1$ bits are inverted but information is perfect if invert them back!



$$C = (1-p) \log_2 2(1-p) + p \log_2 2p$$

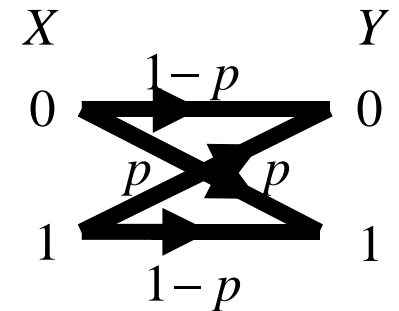
Capacity of binary symmetric channel

- Effect of SNR on Capacity
 - Binary PAM signal (digital signal amplitude $2A$)



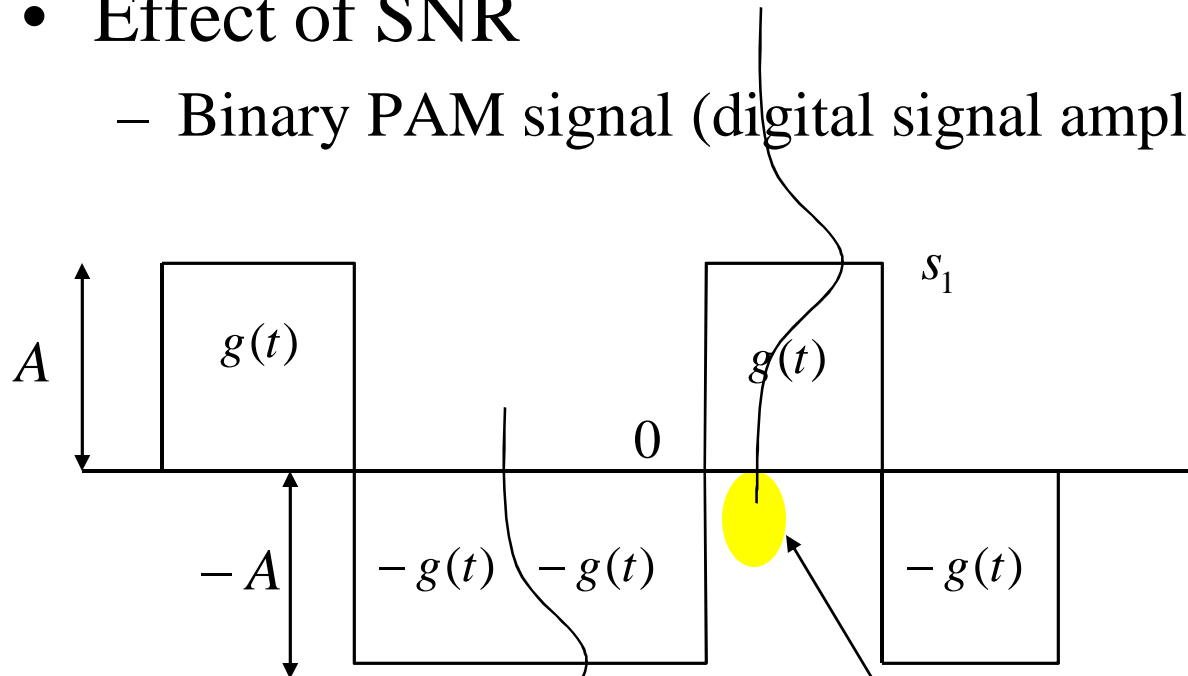
$$p(r | s_1) = \frac{1}{\sqrt{\pi N_0}} e^{-(r - \sqrt{\epsilon_b})^2 / N_0}$$

$$p(r | s_2) = \frac{1}{\sqrt{\pi N_0}} e^{-(r + \sqrt{\epsilon_b})^2 / N_0}$$



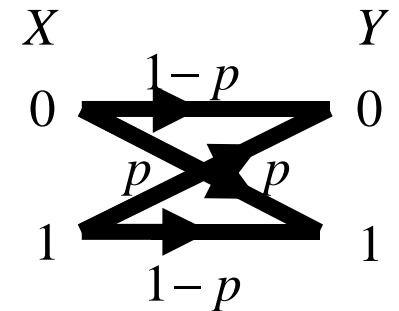
Capacity of binary symmetric channel

- Effect of SNR
 - Binary PAM signal (digital signal amplitude $2A$)



$$P(e | s_1) = \frac{1}{\sqrt{\pi N_0}} \int_{-\infty}^0 e^{-(r - \sqrt{\mathcal{E}_b})^2 / N_0} dr$$

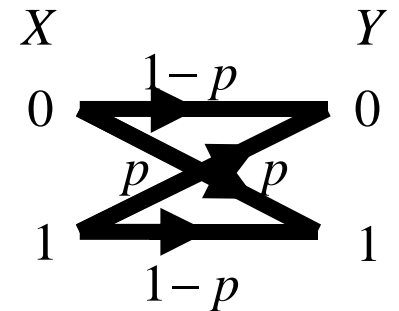
$$= Q\left(\sqrt{\frac{2\mathcal{E}_b}{N_0}}\right) = P(e | s_2)$$



Capacity of binary symmetric channel

- Effect of SNR
 - Binary PAM signal (digital signal amplitude $2A$)

$$\begin{aligned} P_b &= \frac{1}{2} P(e | s_1) + \frac{1}{2} P(e | s_2) \\ &= Q\left(\sqrt{\frac{2\mathcal{E}_b}{N_0}}\right) \\ &= Q\left(\sqrt{2SNR_b}\right) = Q\left(\sqrt{2\gamma_b}\right) \\ &= Q\left(2A\sqrt{\frac{1}{2N_0}}\right) \end{aligned}$$



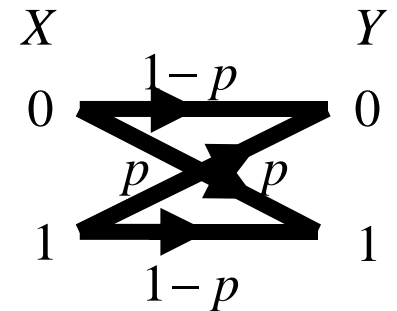
Capacity of binary symmetric channel

- Effect of SNR
 - Binary PAM signal (digital signal amplitude $2A$)

rms noise = $\sigma = \sqrt{\frac{N_0}{2}}$ Not sure about this
Does it depend on bandwidth?

\Rightarrow

$$\begin{aligned} P_b &= Q\left(2A\sqrt{\frac{1}{2N_0}}\right) \\ &= Q\left(2A\sqrt{\frac{2}{4N_0}}\right) \\ &= Q\left(\frac{1}{2}\frac{2A}{\sigma}\right) = Q\left(\frac{1}{2}\frac{2A}{\text{rms noise}}\right) \end{aligned}$$

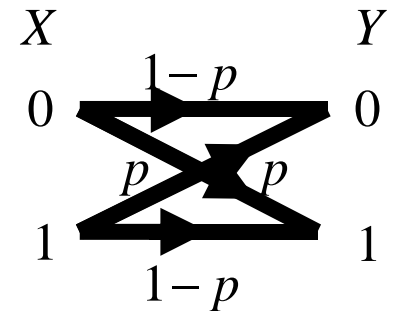


Capacity of binary symmetric channel

- Effect of SNR
 - Binary PAM signal (digital signal amplitude $2A$)

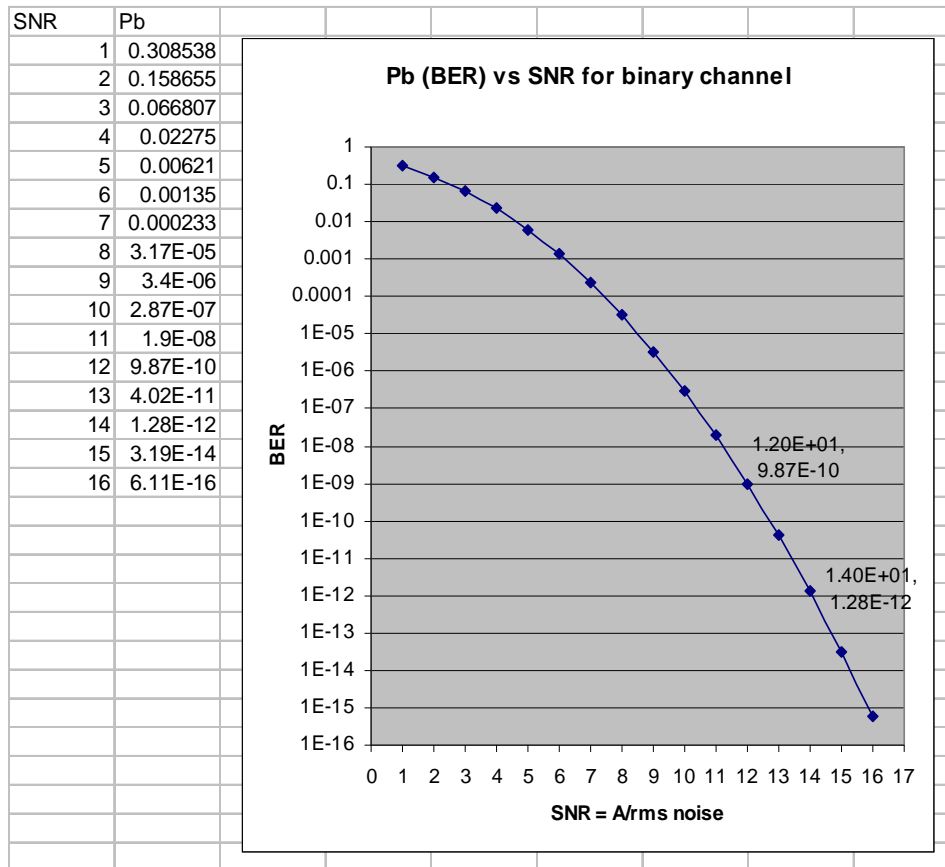
$$P_b = p = Q\left(\frac{1}{2} \frac{\text{Amplitude}}{\text{rms noise}}\right)$$
$$= \frac{1}{2} \operatorname{erfc}\left(\frac{1}{2} \frac{\text{Amplitude}}{\sqrt{2} \text{ rms noise}}\right)$$

$$C = (1-p) \log 2(1-p) + p \log 2p$$

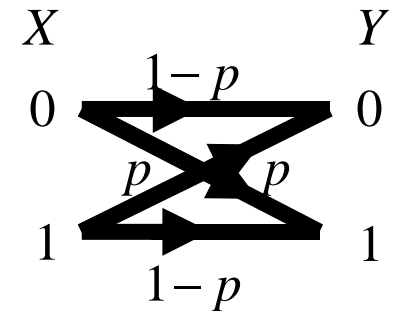


Capacity of binary symmetric channel

- Effect of SNR
 - Binary PAM signal (digital signal amplitude $2A$)



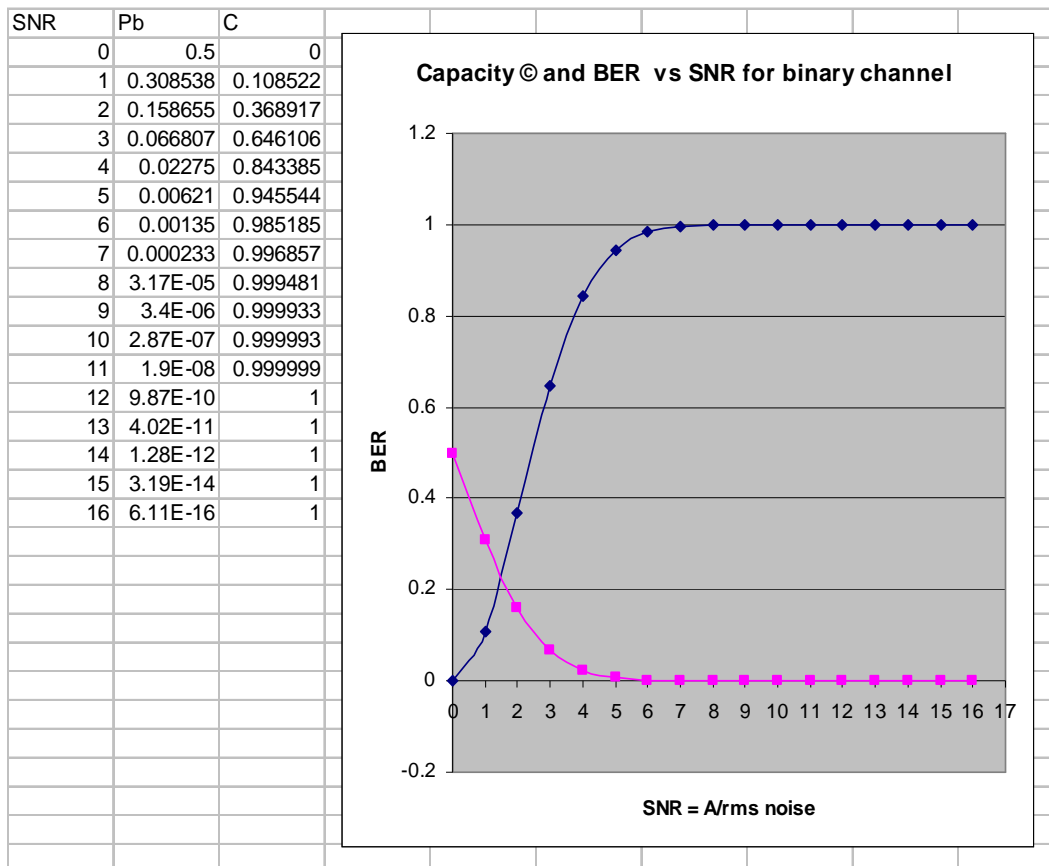
$$P_b = \frac{1}{2} \operatorname{erfc} \left(\frac{1}{2} \frac{\text{Amplitude}}{\sqrt{2} \text{ rms noise}} \right)$$



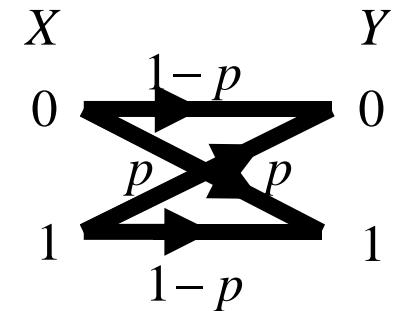
Capacity of binary symmetric channel

- Effect of SNR
 - Binary PAM signal (digital signal amplitude $2A$)

$$C = (1 - p) \log_2 2(1 - p) + p \log_2 2p$$



$$P_b = \frac{1}{2} \operatorname{erfc} \left(\frac{1}{2} \frac{\text{Amplitude}}{\sqrt{2} \text{ rms noise}} \right)$$

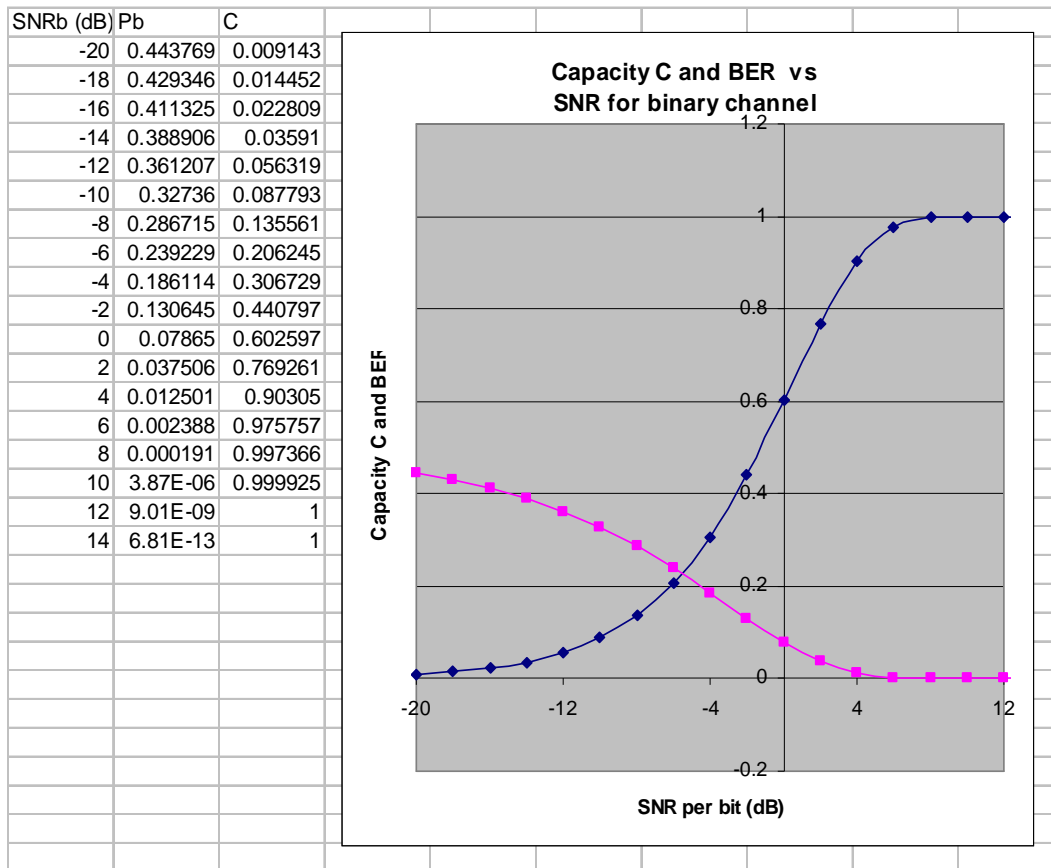


At capacity SNR = 7,
so waste lots of SNR
to get low BER!!!

Capacity of binary symmetric channel

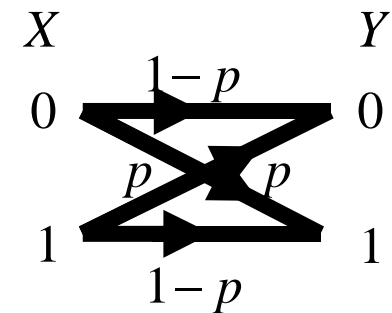
- Effect of SNR_b
 - Binary PAM signal (digital signal amplitude $2A$)

$$C = (1 - p) \log_2 2(1 - p) + p \log_2 2p$$



$$P_b = p = Q(\sqrt{2\gamma_b})$$

$$= \frac{1}{2} \operatorname{erfc}(\sqrt{\gamma_b})$$

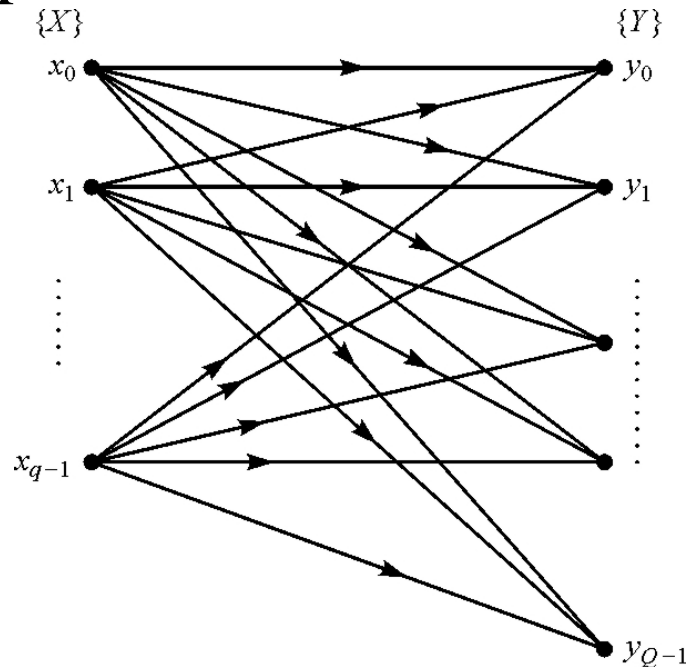


Channel Capacity of Discrete Memoryless Channel

- Discrete-discrete
 - Binary channel, M-ary channel

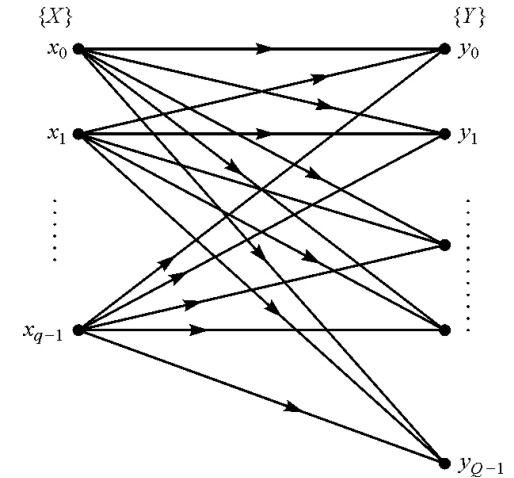
Probability transition matrix

$$\mathbf{P} = \begin{bmatrix} P(Y = y_1 | X = x_1) & \cdot & \cdot & \cdot \\ \cdot & \cdot & P(y_i | x_j) = p_{ji} & \cdot \\ \cdot & \cdot & \cdot & \cdot \\ \cdot & \cdot & \cdot & P_{q-1Q-1} \end{bmatrix}$$



Channel Capacity of Discrete Memoryless Channel

Average Mutual Information

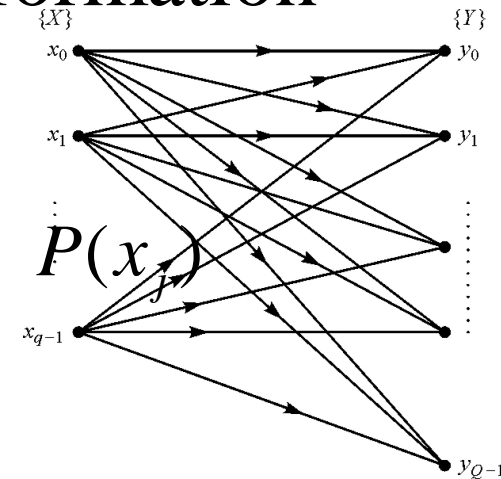


$$I(X;Y) = \sum_{j=0}^{q-1} \sum_{i=1}^{Q-1} P(X = x_j) P(Y = y_i | X = x_j) \log \frac{P(Y = y_i | X = x_j)}{P(Y = y_j)}$$
$$\equiv \sum_{j=0}^{q-1} \sum_{i=1}^{Q-1} P(x_j) P(y_i | x_j) \log \frac{P(y_i | x_j)}{P(y_j)}$$

Channel Capacity of Discrete Memoryless Channel

Channel Capacity is Maximum Information

Occurs for $P(x_j) = p$, for all j
 only if $\mathbf{P} = \text{symmetric}$
 Otherwise must work out max

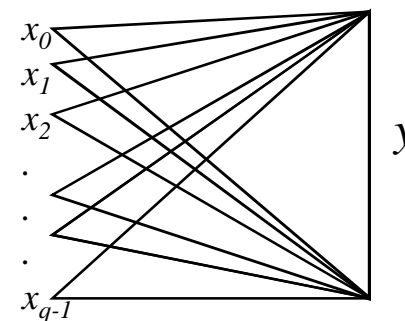


$$C = \max_{P(x_j)} (I(X;Y)) = \max_{P(x_j)} \left(\sum_{j=0}^{q-1} \sum_{i=1}^{Q-1} P(X = x_j) P(Y = y_i | X = x_j) \log \frac{P(Y = y_i | X = x_j)}{P(Y = y_j)} \right)$$

$$\equiv \max_{P(x_j)} \left(\sum_{j=0}^{q-1} \sum_{i=1}^{Q-1} P(x_j) P(y_i | x_j) \log \frac{P(y_i | x_j)}{P(y_j)} \right) \Bigg|_{P(x_j) \geq 0, \sum_{j=0}^{q-1} P(x_j) = 1}$$

Channel Capacity Discrete Memoryless Channel

- Discrete-continuous
- Channel Capacity



$$C = \max_{P(x_i)} (I(X;Y)) = \max_{P(x_i)} \left(\sum_{i=0}^{q-1} P(X = x_i) \int_{-\infty}^{\infty} p(Y = y | X = x_i) \log \frac{p(Y = y | X = x_i)}{p(Y = y)} dy \right)$$

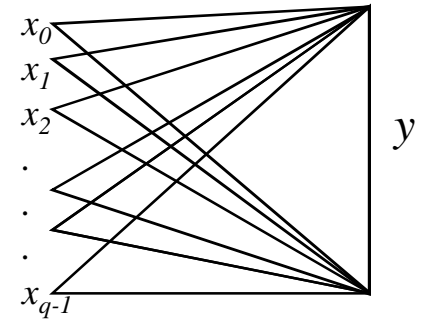
where

$$p(Y = y) = \sum_{i=0}^{q-1} P(X = x_i) p(Y = y | X = x_i)$$

$$\mathbf{P} = \begin{bmatrix} p(y | X = x_1) \\ \text{graph of } p(y | X = x_1) \\ p(y | X = x_k) \\ \text{graph of } p(y | X = x_k) \end{bmatrix}$$

Channel Capacity Discrete Memoryless Channel

- Discrete-continuous
- Channel Capacity with AWGN

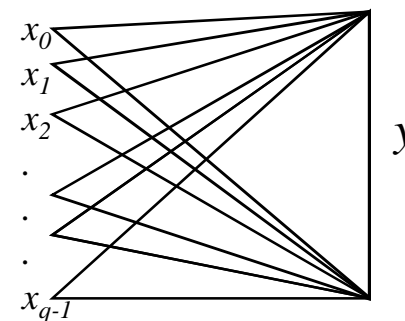


$$p(y | x_k) = \frac{1}{\sqrt{2\pi\sigma}} e^{-(y-x_k)^2/2\sigma^2}$$

$$C = \max_{P(x_i)} \left(\sum_{i=0}^{q-1} \int_{-\infty}^{\infty} P(X = x_i) \frac{1}{\sqrt{2\pi\sigma}} e^{-(y-x_i)^2/2\sigma^2} \log \frac{\frac{1}{\sqrt{2\pi\sigma}} e^{-(y-x_i)^2/2\sigma^2}}{\sum_{i=0}^{q-1} P(X = x_i) \frac{1}{\sqrt{2\pi\sigma}} e^{-(y-x_i)^2/2\sigma^2}} dy \right)$$

Channel Capacity Discrete Memoryless Channel

- Binary Symmetric PAM-continuous
- Maximum Information when:



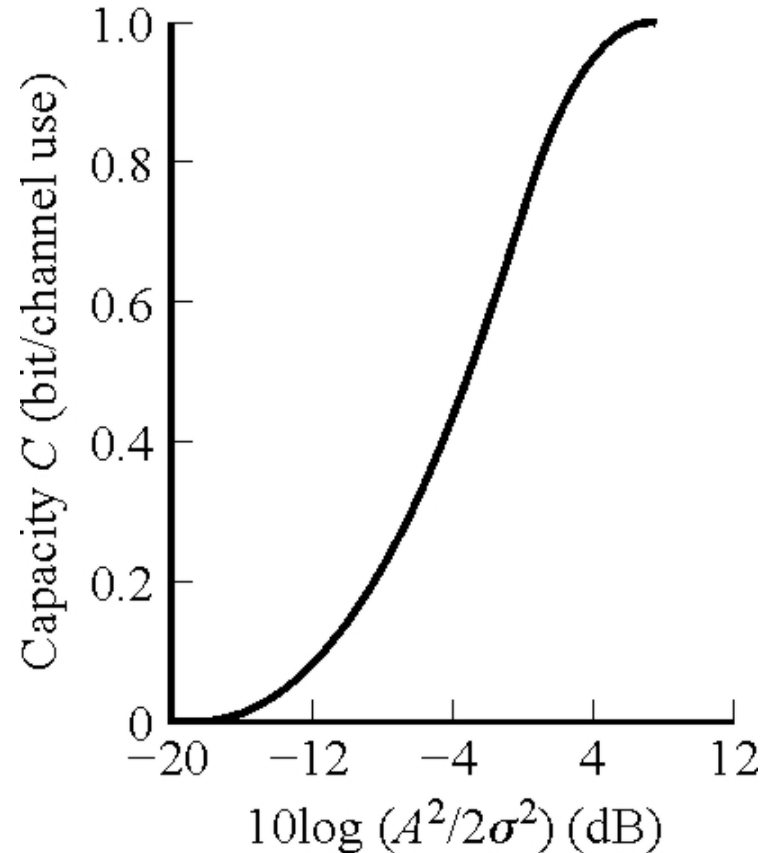
$$P(X = A) = P(X = -A) = \frac{1}{2}$$

$$C = \frac{1}{2} \frac{1}{\sqrt{2\pi\sigma}} \left[\log \frac{2e^{2A^2/2\sigma^2}}{e^{A^2/2\sigma^2} + e^{-A^2/2\sigma^2}} \int_{-\infty}^{\infty} e^{-y^2/2\sigma^2} dy \right. \\ \left. + \log \frac{2e^{-2A^2/2\sigma^2}}{e^{A^2/2\sigma^2} + e^{-A^2/2\sigma^2}} \int_{-\infty}^{\infty} e^{-y^2/2\sigma^2} dy \right]$$

Channel Capacity Discrete Memoryless Channel

- Binary Symmetric PAM-continuous
- Maximum Information when:

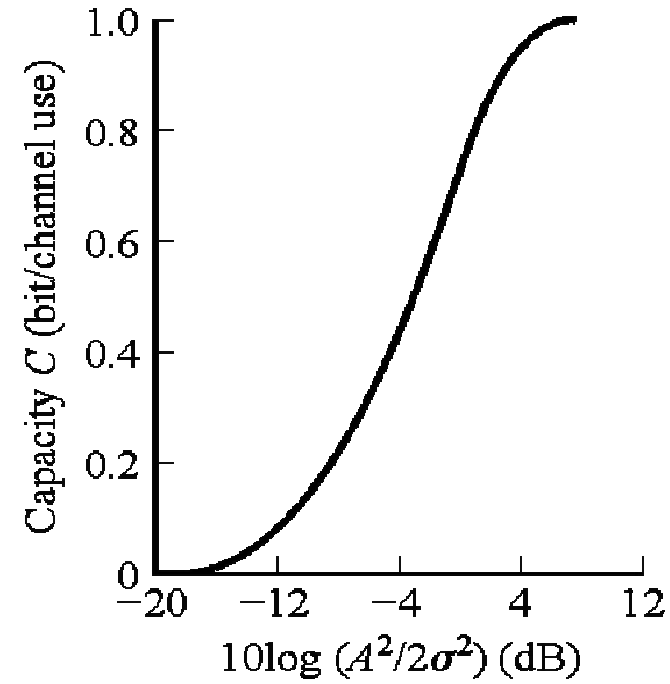
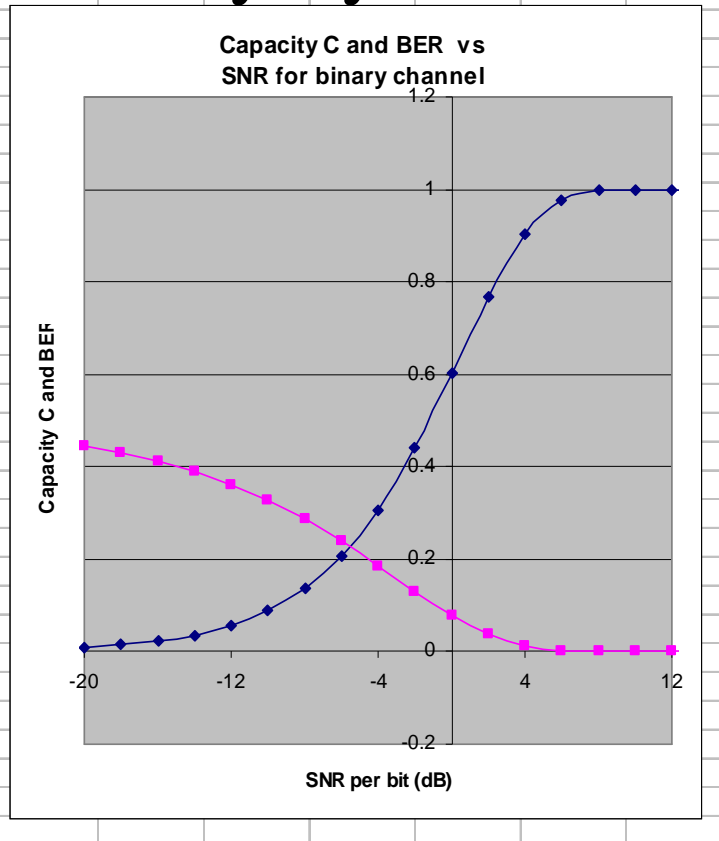
$$C = \frac{1}{2} \frac{1}{\sqrt{2\pi}\sigma} \left[\log \frac{2e^{2A^2/2\sigma^2}}{e^{A^2/2\sigma^2} + e^{-A^2/2\sigma^2}} \int_{-\infty}^{\infty} e^{-y^2/2\sigma^2} dy \right. \\ \left. + \log \frac{2e^{-2A^2/2\sigma^2}}{e^{A^2/2\sigma^2} + e^{-A^2/2\sigma^2}} \int_{-\infty}^{\infty} e^{-y^2/2\sigma^2} dy \right]$$



Channel Capacity Discrete Memoryless Channel

- Binary Symmetric PAM-continuous
- Versus Binary Symmetric discrete

SNRb (dB)	Pb	C
-20	0.443769	0.009143
-18	0.429346	0.014452
-16	0.411325	0.022809
-14	0.388906	0.03591
-12	0.361207	0.056319
-10	0.32736	0.087793
-8	0.286715	0.135561
-6	0.239229	0.206245
-4	0.186114	0.306729
-2	0.130645	0.440797
0	0.07865	0.602597
2	0.037506	0.769261
4	0.012501	0.90305
6	0.002388	0.975757
8	0.000191	0.997366
10	3.87E-06	0.999925
12	9.01E-09	1
14	6.81E-13	1



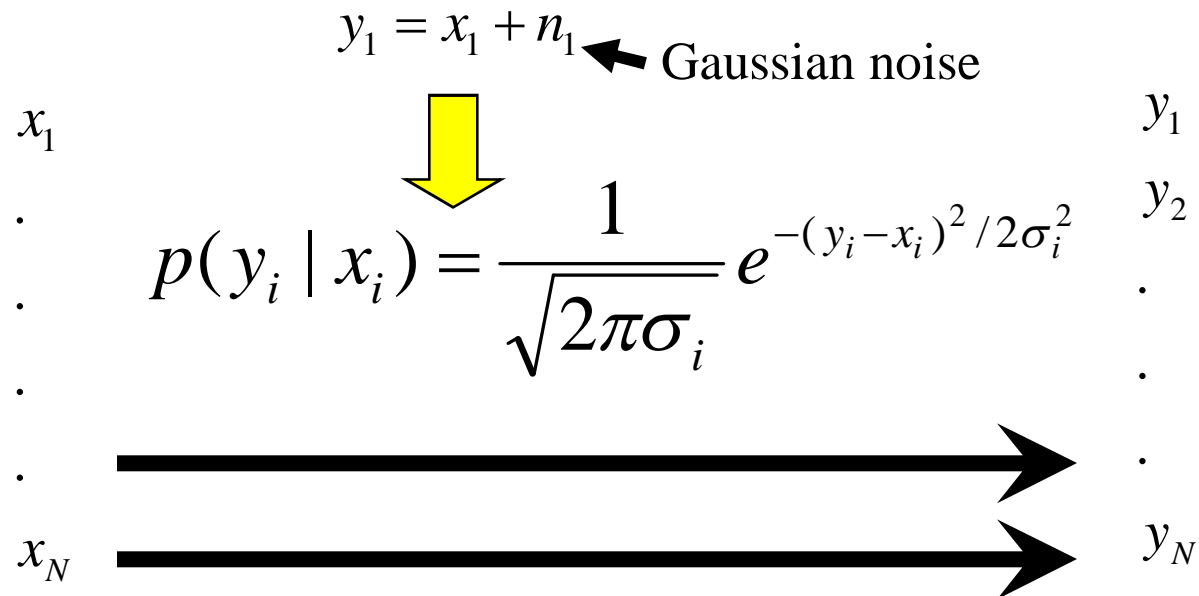
Discrete Memoryless Channel

- Continuous-continuous
 - Modulated waveform channels (QAM)
 - Assume Band limited waveforms, bandwidth = W
 - Sampling at Nyquist = $2W$ sample/s
 - Then over interval of $N = 2WT$ samples use an orthogonal function expansion:

$$\begin{array}{ccc} x(t) & \longrightarrow & \textcircled{+} & \longrightarrow & y(t) \\ = \sum_{i=1}^N x_i f_i(t) & & \uparrow & & = \sum_{i=1}^N y_i f_i(t) \\ & & n(t) & & \\ & & = \sum_{i=1}^N n_i f_i(t) & & \end{array}$$

Discrete Memoryless Channel

- Continuous-continuous
 - Using orthogonal function expansion get an equivalent discrete time channel:

$$y_1 = x_1 + n_1 \leftarrow \text{Gaussian noise}$$

$$p(y_i | x_i) = \frac{1}{\sqrt{2\pi\sigma_i}} e^{-(y_i - x_i)^2 / 2\sigma_i^2}$$

Discrete Memoryless Channel

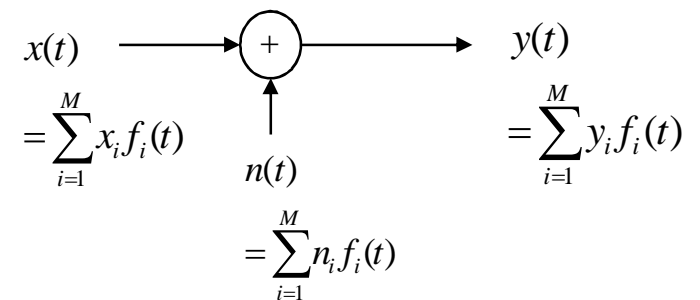
- Continuous-continuous
- Capacity is (Shannon)

$$C = \lim_{T \rightarrow \infty} \max_{p(x)} \frac{1}{T} I(X; Y)$$

$$N = 2WT$$

$$\begin{aligned}
 I(\mathbf{X}_N; \mathbf{Y}_N) &= \int_{\mathbf{x}_N} \cdots \int \int_{\mathbf{Y}_N} \cdots \int p(\mathbf{y}_N | \mathbf{x}_N) p(\mathbf{x}_N) \log \frac{p(\mathbf{y}_N | \mathbf{x}_N)}{p(\mathbf{y}_N)} d\mathbf{x}_N d\mathbf{y}_N \\
 &= \sum_{i=1}^N \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} p(y_i | x_i) p(x_j) \log \frac{p(y_i | x_i)}{p(y_i)} dy_i dx_i
 \end{aligned}$$

$$p(y_i | x_i) = \frac{1}{\sqrt{2\pi\sigma_i^2}} e^{-(y_i - x_i)^2 / 2\sigma_i^2}$$



Discrete Memoryless Channel

- Continuous-continuous
- Maximum Information when:

$$p(x_i) = \frac{1}{\sqrt{2\pi\sigma_x}} e^{-x_i^2/2\sigma_x^2} \quad \text{Statistically independent zero mean Gaussian inputs}$$

then

$$\begin{aligned} \max_{p(x)} I(\mathbf{X}_N; \mathbf{Y}_N) &= \sum_{i=1}^N \frac{1}{2} \log \left(1 + \frac{2\sigma_x^2}{N_0} \right) \\ &= \frac{1}{2} N \log \left(1 + \frac{2\sigma_x^2}{N_0} \right) \\ &= WT \log \left(1 + \frac{2\sigma_x^2}{N_0} \right) \end{aligned}$$

Discrete Memoryless Channel

- Continuous-continuous
- Constrain average power in $x(t)$:

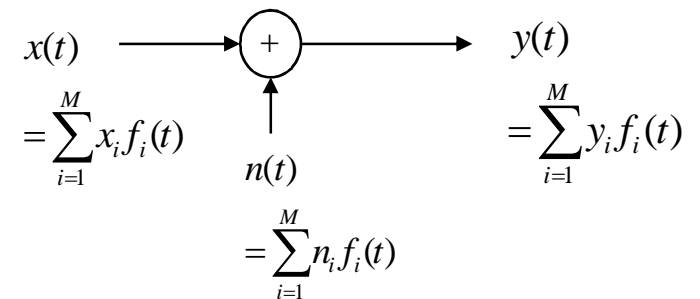
$$\begin{aligned} P_{av} &= \frac{1}{T} \int_0^T E[x^2(t)] dt \\ &= \frac{1}{2} \sum_{i=1}^N E(x_i^2) \\ &= \frac{N\sigma_x^2}{T} = 2W\sigma_x^2 \end{aligned}$$

Discrete Memoryless Channel

- Continuous-continuous
- Thus Capacity is:

$$\begin{aligned}
 C &= \lim_{T \rightarrow \infty} \max_{p(x)} \frac{1}{T} I(\mathbf{X}_N; \mathbf{Y}_N) \\
 &= \lim_{T \rightarrow \infty} W \log \left(1 + \frac{2\sigma_x^2}{N_0} \right) \\
 &= W \log \left(1 + \frac{P_{av}}{WN_0} \right)
 \end{aligned}$$

$$p(y_i | x_i) = \frac{1}{\sqrt{2\pi\sigma_i}} e^{-(y_i - x_i)^2 / 2\sigma_i^2}$$

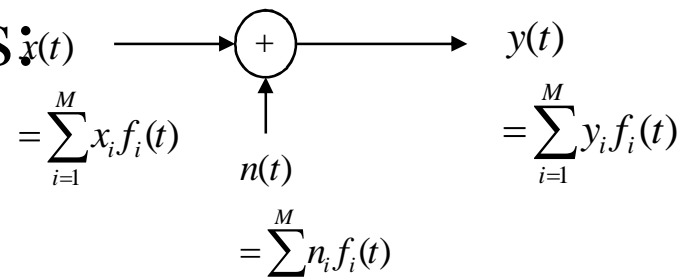


Discrete Memoryless Channel

- Continuous-continuous

$$p(y_i | x_i) = \frac{1}{\sqrt{2\pi\sigma_i}} e^{-(y_i-x_i)^2/2\sigma_i^2}$$

- Thus Normalized Capacity is:



$$\frac{C}{W} = \log_2 \left(1 + \frac{P_{av}}{WN_0} \right), \text{ but } P_{av} = C\mathcal{E}_b$$

$$= \log_2 \left(1 + \frac{C\mathcal{E}_b}{WN_0} \right)$$

\Rightarrow

$$\frac{\mathcal{E}_b}{N_0} = \frac{2^{C/W} - 1}{C/W}$$

etab/No (d	C/W
-1.44036	0.1
-1.36402	0.15
-1.24869	0.225
-1.07386	0.3375
-0.8075	0.50625
-0.39875	0.759375
0.234937	1.139063
1.230848	1.708594
2.822545	2.562891
5.41099	3.844336
9.669259	5.766504
16.65749	8.649756
27.92605	12.97463
45.69444	19.46195
73.22669	29.19293
115.4055	43.78939
179.5542	65.68408

