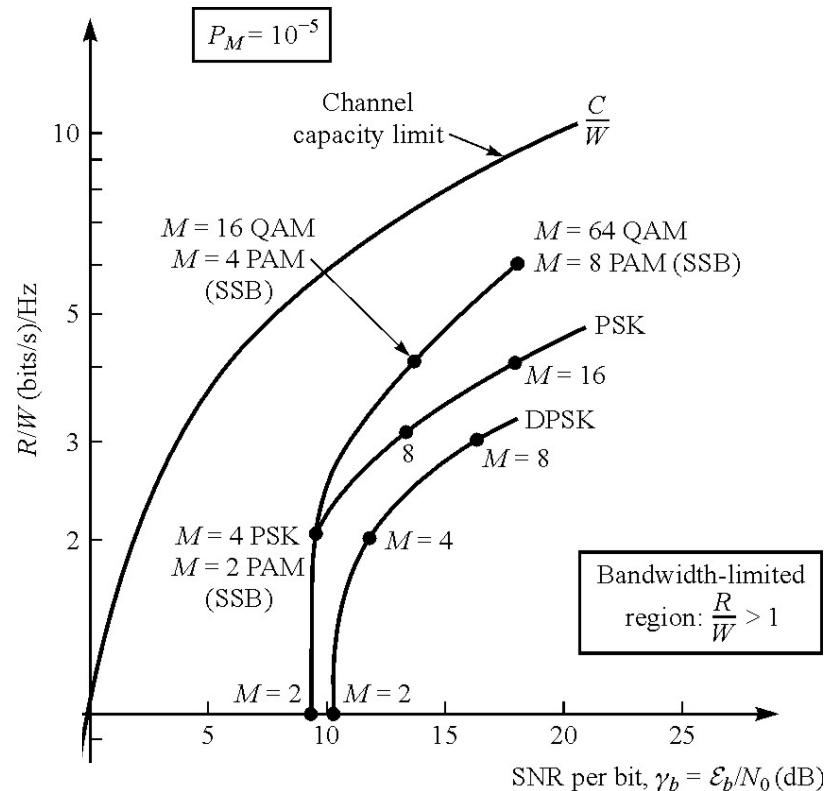


# Channel Capacity

- Discrete Memoryless Channels
- Random Codes
- Block Codes
- Trellis Codes



# Channel Models

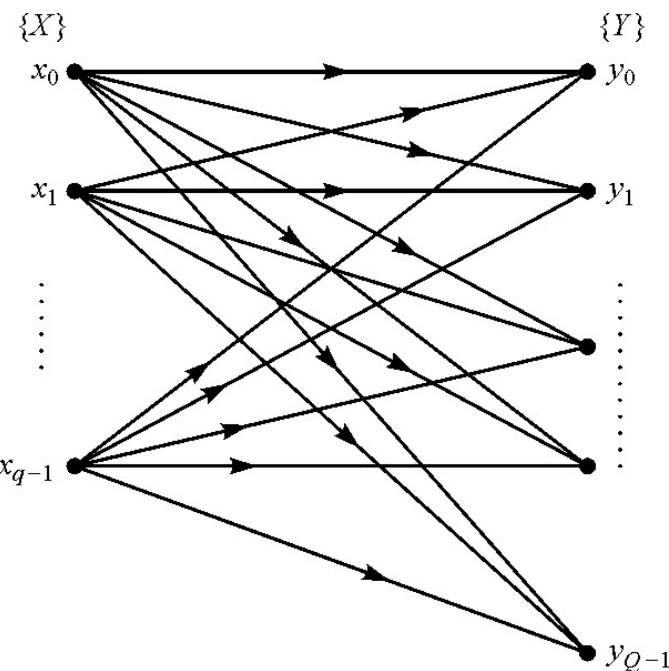
- Discrete Memoryless Channel
  - Discrete-discrete
    - Binary channel, M-ary channel
  - Discrete-continuous
    - M-ary channel with soft-decision (analog)
  - Continuous-continuous
    - Modulated waveform channels (QAM)

# Discrete Memoryless Channel

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

Probability transition matrix

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



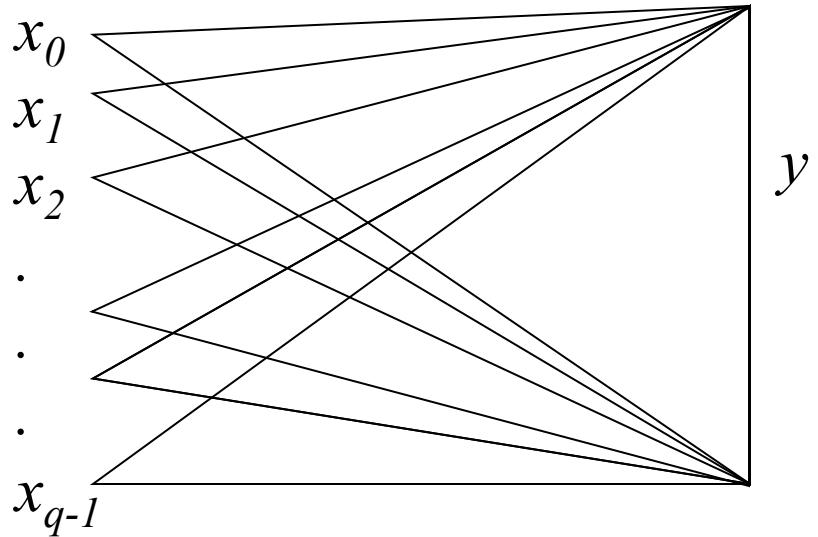
# Discrete Memoryless Channel

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

AWGN

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

$$\mathbf{P} = \begin{bmatrix} p(y | X = x_1) \\ \vdots \\ p(y | X = x_{q-1}) \end{bmatrix}$$



# 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:

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

# Discrete Memoryless Channel

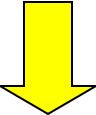
- Continuous-continuous
  - Using orthogonal function expansion:

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

# Discrete Memoryless Channel

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

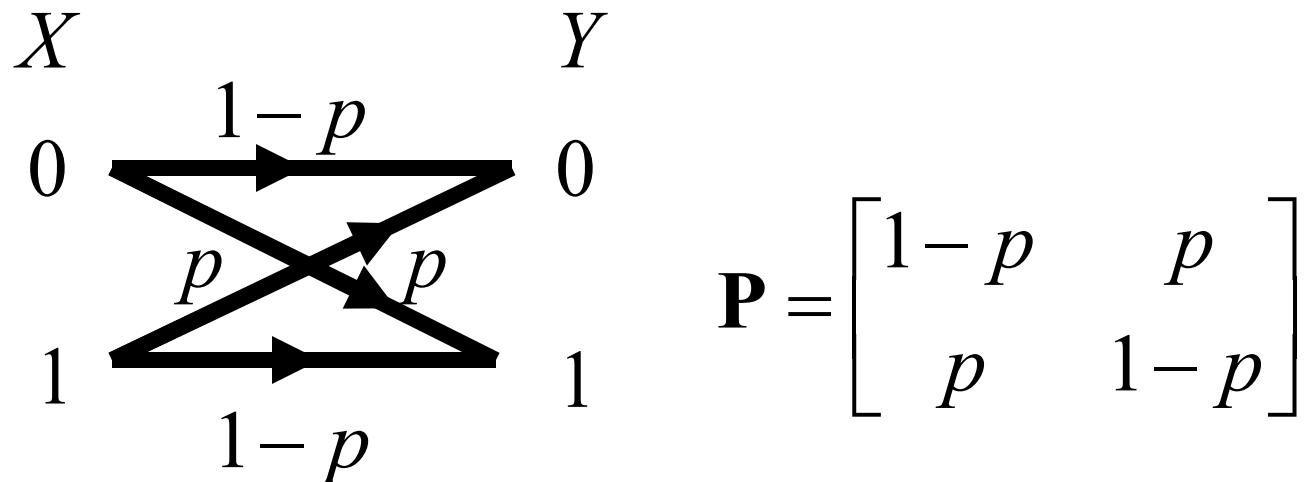
$$y_1 = x_1 + n_1 \quad \text{Gaussian noise}$$



$$x_1 \qquad \qquad \qquad y_1$$
$$\cdot \qquad \qquad \qquad \cdot$$
$$p(y_i | x_i) = \frac{1}{\sqrt{2\pi\sigma_i^2}} e^{-(y_i - x_i)^2 / 2\sigma_i^2} \qquad y_2$$
$$\cdot \qquad \qquad \qquad \cdot$$
$$\cdot \qquad \qquad \qquad \cdot$$
$$x_N \qquad \qquad \qquad y_N$$

# Capacity of binary symmetric channel

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



# Capacity of binary symmetric channel

- Average Mutual Information

$$I(X;Y) = P(X=0)P(Y=0|X=0)\log \frac{P(Y=0|X=0)}{P(Y=0)} + \dots$$

$$P(X=0)P(Y=1|X=0)\log \frac{P(Y=1|X=0)}{P(Y=1)} + \dots$$

$$P(X=1)P(Y=0|X=1)\log \frac{P(Y=0|X=1)}{P(Y=0)} + \dots$$

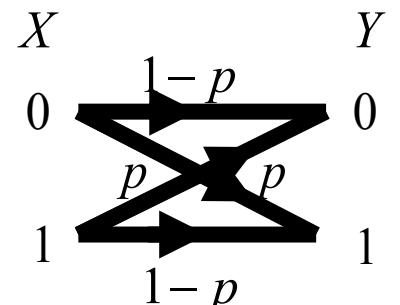
$$P(X=1)P(Y=1|X=1)\log \frac{P(Y=1|X=1)}{P(Y=1)}$$

$$= P(X=0)(1-p)\log \frac{1-p}{(1-p)P(X=0)+pP(X=1)} + \dots$$

$$P(X=0)p\log \frac{p}{pP(X=0)+(1-p)P(X=1)} + \dots$$

$$P(X=1)p\log \frac{p}{(1-p)P(X=0)+pP(X=1)} + \dots$$

$$P(X=1)(1-p)\log \frac{1-p}{pP(X=0)+(1-p)P(X=1)}$$

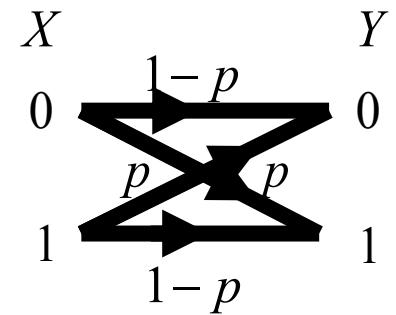


# Capacity of binary symmetric channel

- Channel Capacity is Maximum Information
  - earlier showed:  $\max(I(X;Y)) \Leftrightarrow P(X=1) = P(X=0) = \frac{1}{2}$

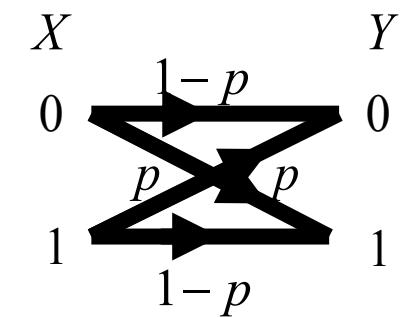
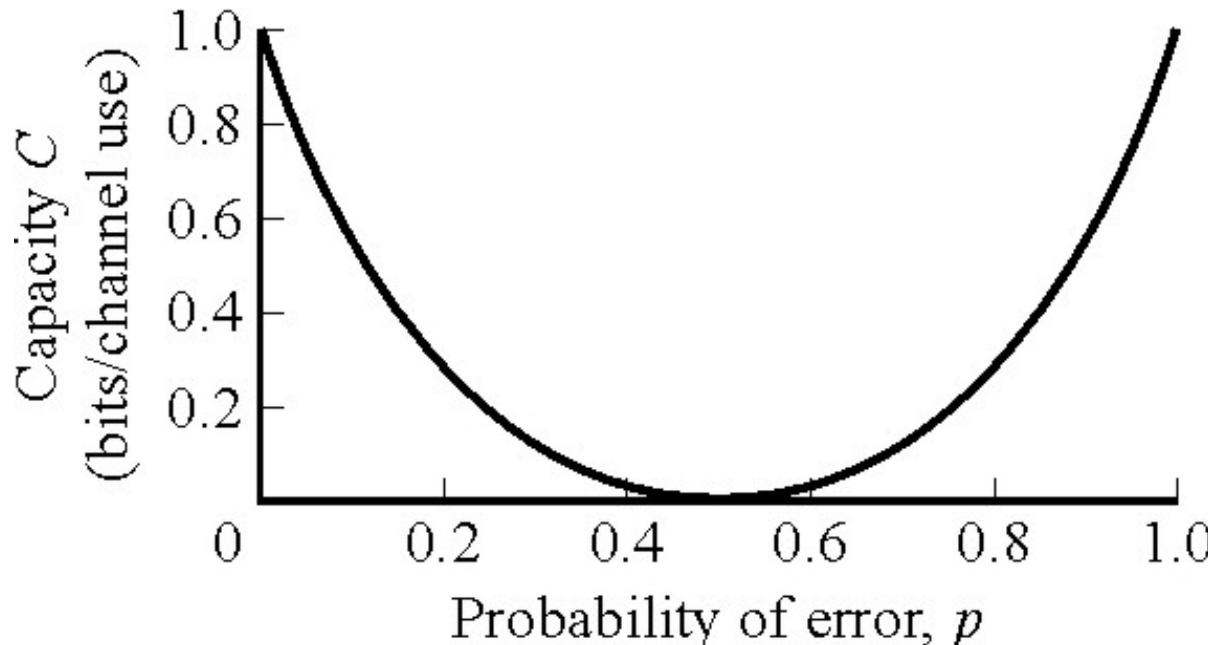
$$\begin{aligned} C = \max(I(X;Y)) &= \left[ P(X=0)(1-p) \log \frac{1-p}{(1-p)P(X=0) + pP(X=1)} + \dots \right. \\ &\quad P(X=0)p \log \frac{p}{pP(X=0) + (1-p)P(X=1)} + \dots \\ &\quad P(X=1)p \log \frac{p}{(1-p)P(X=0) + pP(X=1)} + \dots \\ &\quad \left. P(X=1)(1-p) \log \frac{1-p}{pP(X=0) + (1-p)P(X=1)}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

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



# Capacity of binary symmetric channel

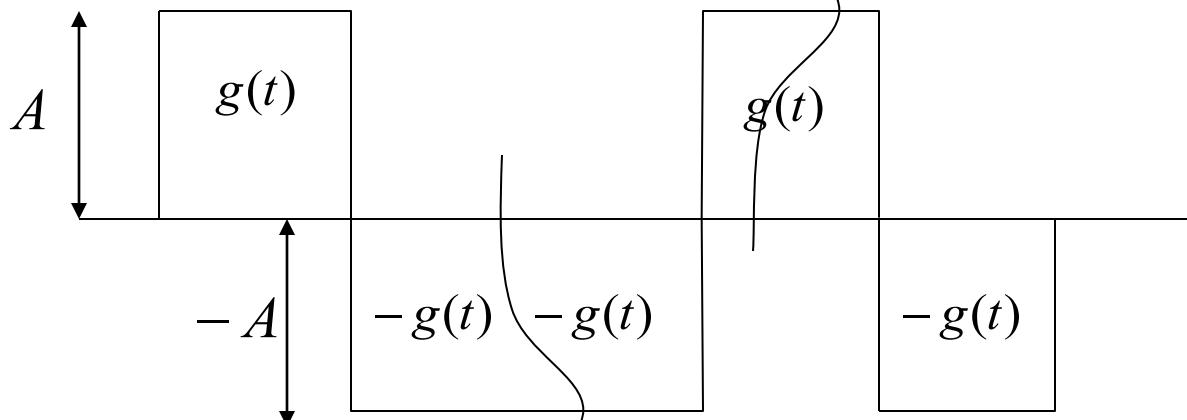
- Channel Capacity
  - When  $p=1$  bits are inverted but information is perfect if invert them back!



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

# Capacity of binary symmetric channel

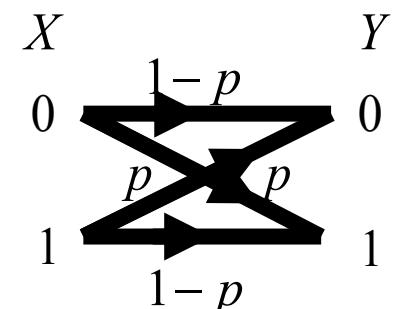
- Effect of SNR on Capacity
  - Binary PAM signal (digital signal amplitude  $2A$ )



AGWN

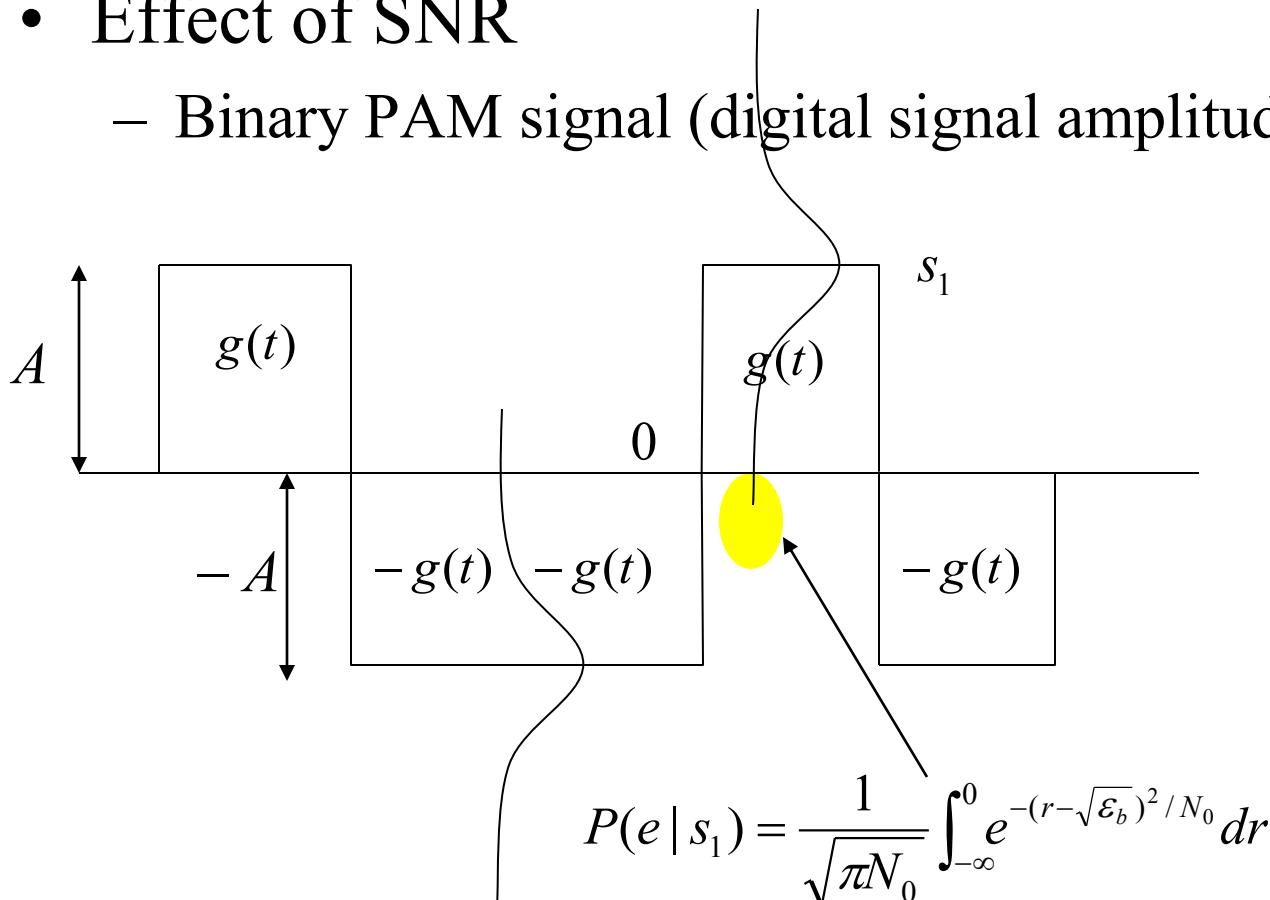
$$p(r | s_1) = \frac{1}{\sqrt{\pi N_0}} e^{-(r - \sqrt{\varepsilon_b})^2 / N_0}$$

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

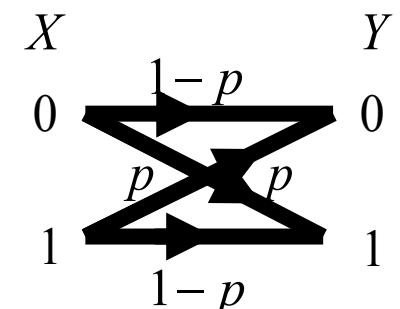


# Capacity of binary symmetric channel

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



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



# Capacity of binary symmetric channel

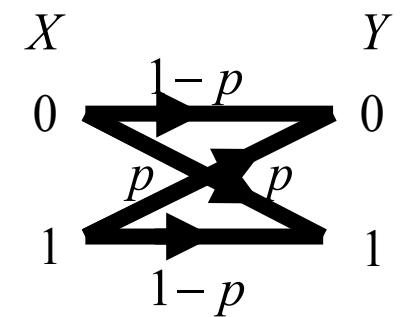
- Effect of SNR
  - Binary PAM signal (digital signal amplitude  $2A$ )

$$P_b = \frac{1}{2} P(e | s_1) + \frac{1}{2} P(e | s_2)$$

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

$$= Q\left(\sqrt{2SNR_b}\right) = Q\left(\sqrt{2\gamma_b}\right)$$

$$= Q\left(2A\sqrt{\frac{1}{2N_0}}\right)$$

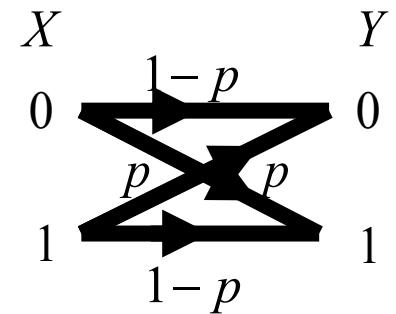


# Capacity of binary symmetric channel

- 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?  
 $\Rightarrow$

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

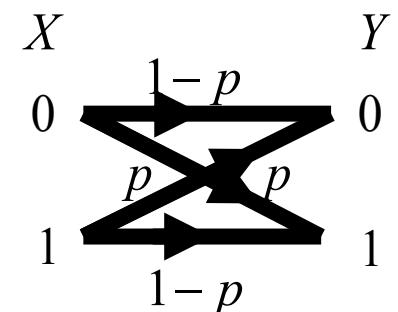


# Capacity of binary symmetric channel

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

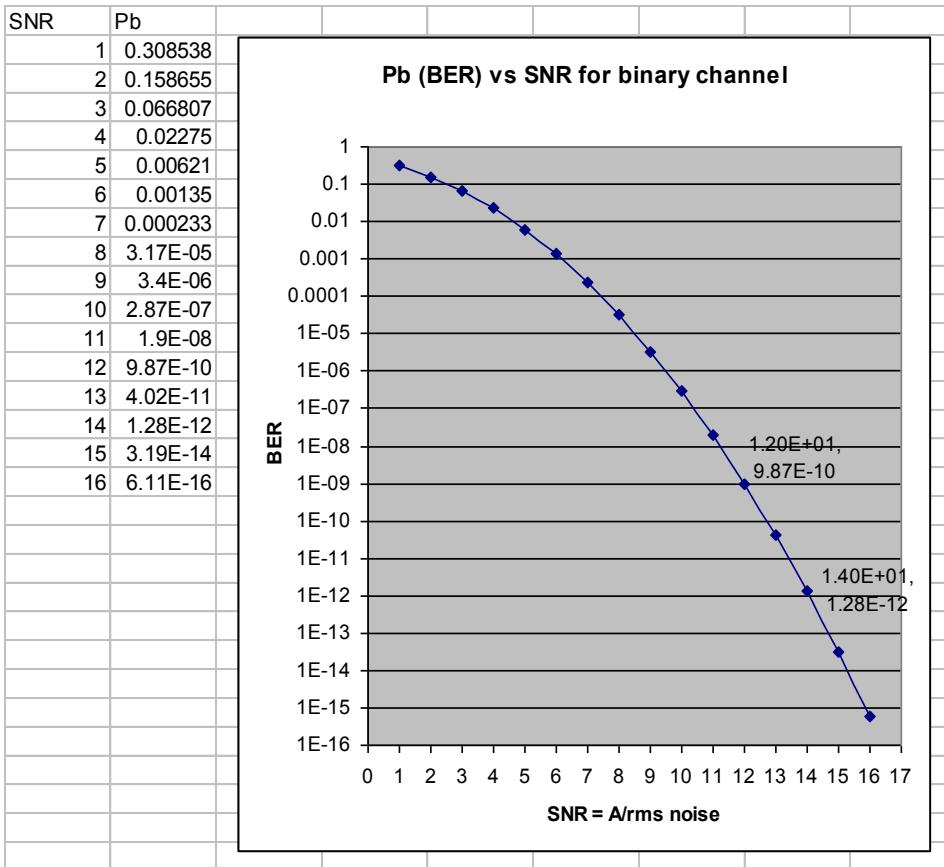
$$\begin{aligned}P_b = p &= Q\left(\frac{\frac{1}{2} \text{Amplitude}}{\text{rms noise}}\right) \\&= \frac{1}{2} \operatorname{erfc}\left(\frac{\frac{1}{2} \text{Amplitude}}{\sqrt{2} \text{ rms noise}}\right)\end{aligned}$$

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

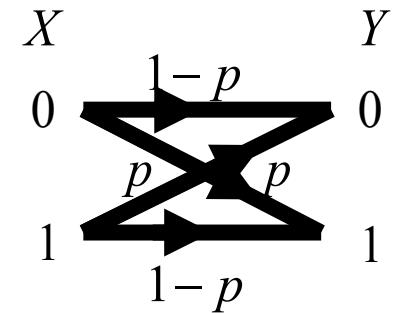


# Capacity of binary symmetric channel

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



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



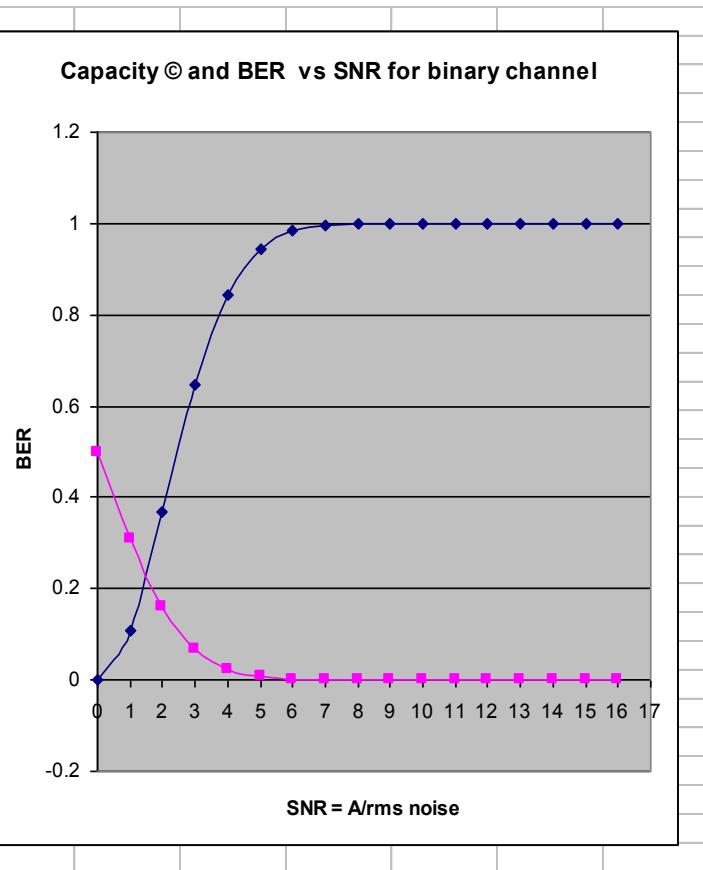
# Capacity of binary symmetric channel

- Effect of SNR

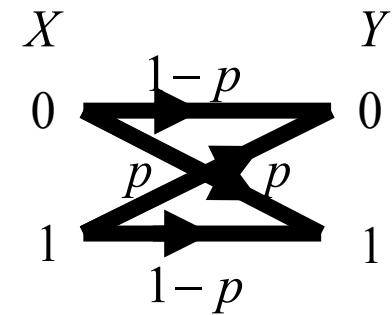
– Binary PAM signal (digital signal amplitude  $2A$ )

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

SNR	P <sub>b</sub>	C
0	0.5	0
1	0.308538	0.108522
2	0.158655	0.368917
3	0.066807	0.646106
4	0.02275	0.843385
5	0.00621	0.945544
6	0.00135	0.985185
7	0.000233	0.996857
8	3.17E-05	0.999481
9	3.4E-06	0.999933
10	2.87E-07	0.999993
11	1.9E-08	0.999999
12	9.87E-10	1
13	4.02E-11	1
14	1.28E-12	1
15	3.19E-14	1
16	6.11E-16	1



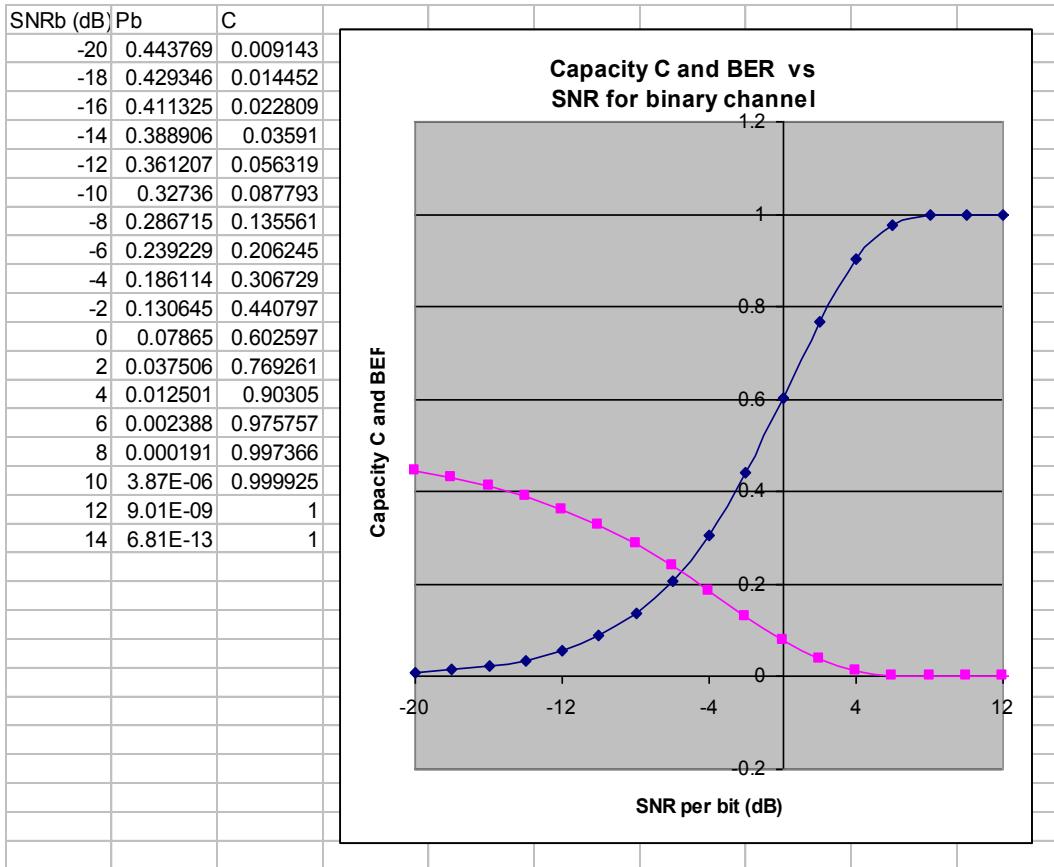
$$P_b = \frac{1}{2} \operatorname{erfc} \left( \frac{1}{\sqrt{2}} \frac{\text{Amplitude}}{\sqrt{2} \text{ rms noise}} \right)$$



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

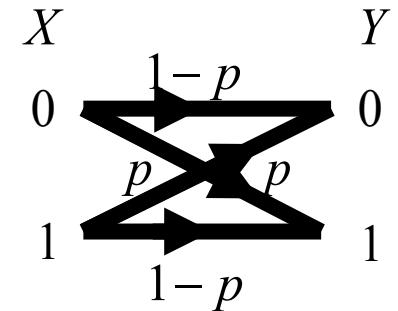
# Capacity of binary symmetric channel

- Effect of SNR<sub>b</sub>
  - Binary PAM signal (digital signal amplitude  $2A$ )
$$C = (1 - p) \log_2 2(1 - p) + p \log_2 2p$$



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

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

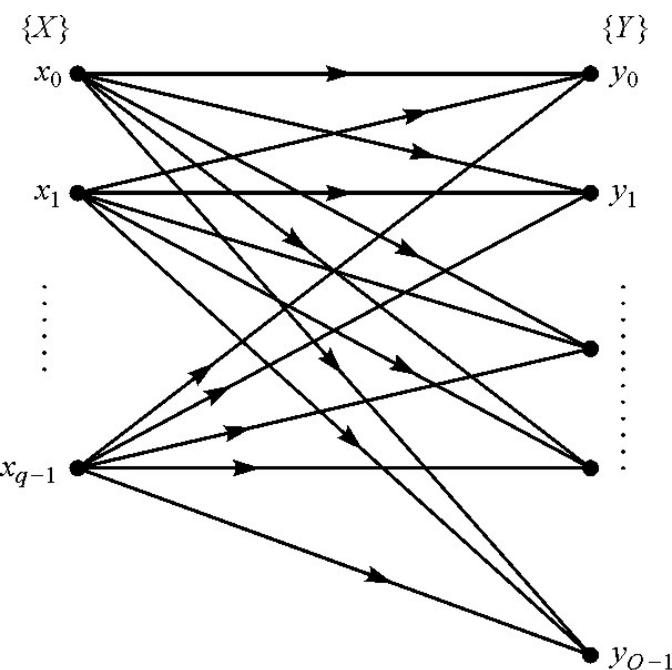


# Channel Capacity of Discrete Memoryless Channel

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

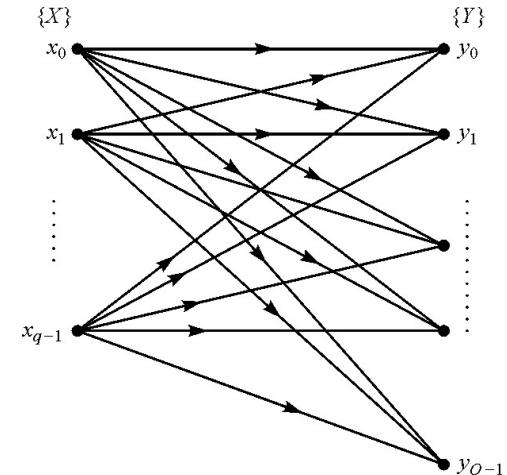
Probability transition matrix

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



# Channel Capacity of Discrete Memoryless Channel

## Average Mutual Information



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

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

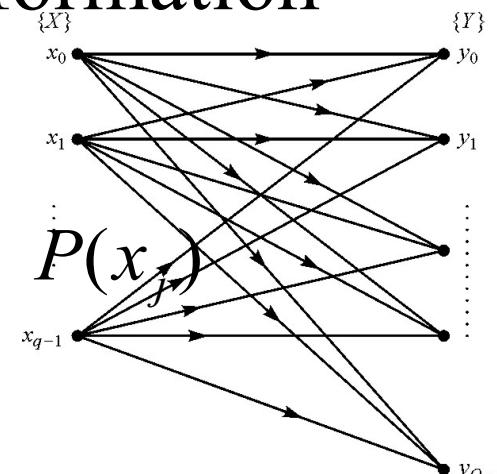
# Channel Capacity of Discrete Memoryless Channel

Channel Capacity is Maximum Information

Occurs for  $P(x_j) = p$ , for all  $j$

only if  $\mathbf{P} = \text{symmetric}$

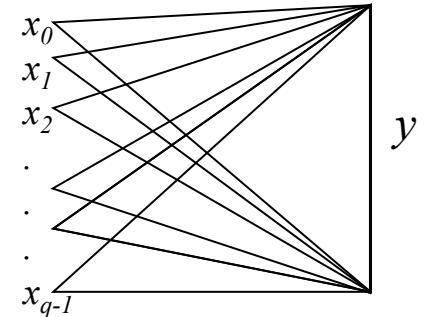
Otherwise must work out max



$$\begin{aligned}
 C &= \max_{P(x_j)} (I(X;Y)) = \max_{P(x_j)} \left( \sum_{j=0}^{q-1} \sum_{i=1}^{Q-1} P(X=x_j) P(Y=y_i | X=x_j) \log \frac{P(Y=y_i | X=x_j)}{P(Y=y_j)} \right) \\
 &\equiv \max_{P(x_j)} \left( \sum_{j=0}^{q-1} \sum_{i=1}^{Q-1} P(x_j) P(y_i | x_j) \log \frac{P(y_i | x_j)}{P(y_j)} \right) \Big|_{\substack{P(x_j) \geq 0, \\ \sum_{j=0}^{q-1} P(x_j) = 1}}
 \end{aligned}$$

# Channel Capacity Discrete Memoryless Channel

- Discrete-continuous
- Channel Capacity



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

where

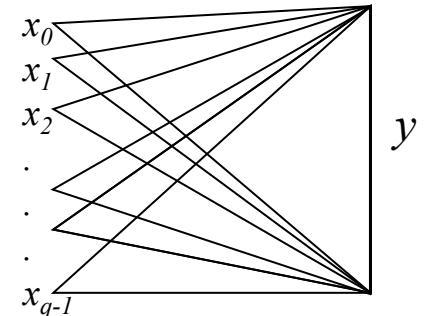
$$p(Y=y) = \sum_{i=0}^{q-1} P(X=x_i) p(Y=y | X=x_i)$$

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

# Channel Capacity Discrete Memoryless Channel

- Discrete-continuous
- Channel Capacity with AWGN

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

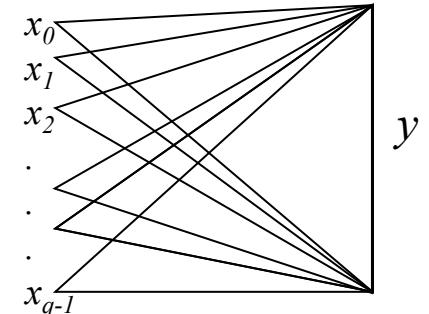


$$C = \max_{P(x_i)} \left( \sum_{i=0}^{q-1} \int_{-\infty}^{\infty} P(X=x_i) \frac{1}{\sqrt{2\pi}\sigma} e^{-(y-x_i)^2/2\sigma^2} \log \frac{\frac{1}{\sqrt{2\pi}\sigma} e^{-(y-x_i)^2/2\sigma^2}}{\sum_{i=0}^{q-1} P(X=x_i) \frac{1}{\sqrt{2\pi}\sigma} e^{-(y-x_i)^2/2\sigma^2}} dy \right)$$

# Channel Capacity Discrete Memoryless Channel

- Binary Symmetric PAM-continuous
- Maximum Information when:

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

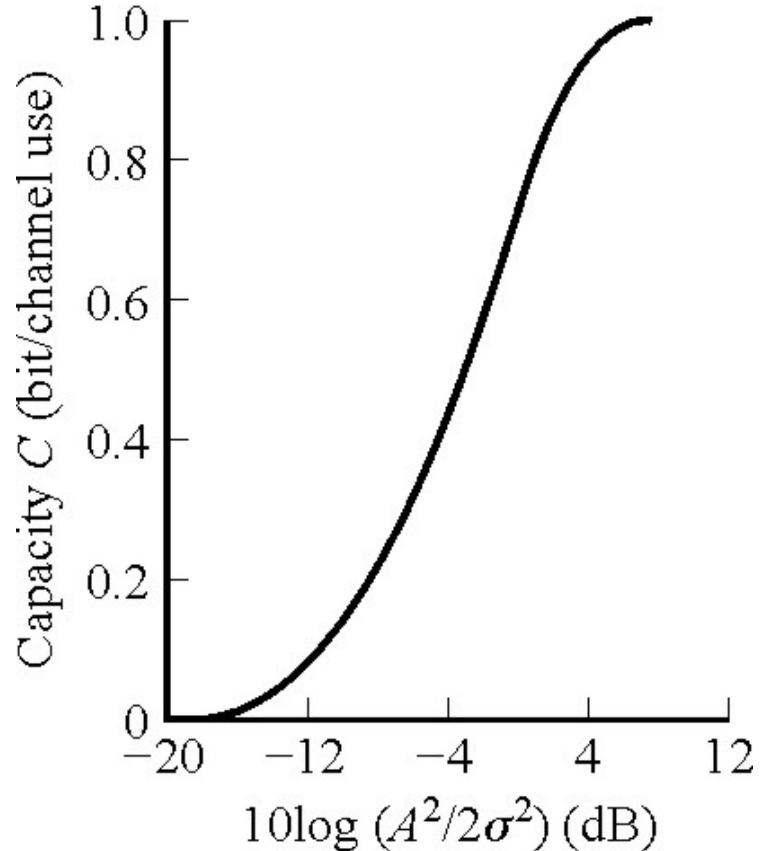


$$\begin{aligned} C = & \frac{1}{2} \frac{1}{\sqrt{2\pi}\sigma} \left[ \log \frac{2e^{2A^2/2\sigma^2}}{e^{A^2/2\sigma^2} + e^{-A^2/2\sigma^2}} \int_{-\infty}^{\infty} e^{-y^2/2\sigma^2} dy \right. \\ & \left. + \log \frac{2e^{-2A^2/2\sigma^2}}{e^{A^2/2\sigma^2} + e^{-A^2/2\sigma^2}} \int_{-\infty}^{\infty}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

# Channel Capacity Discrete Memoryless Channel

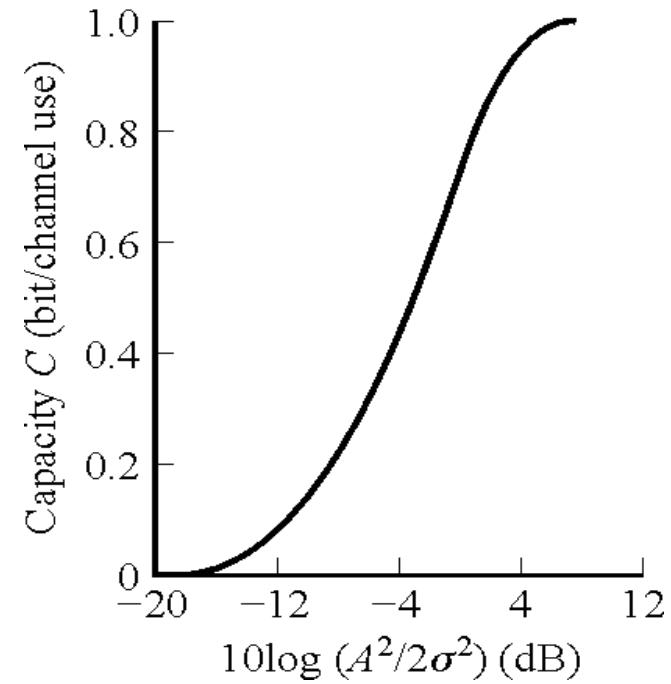
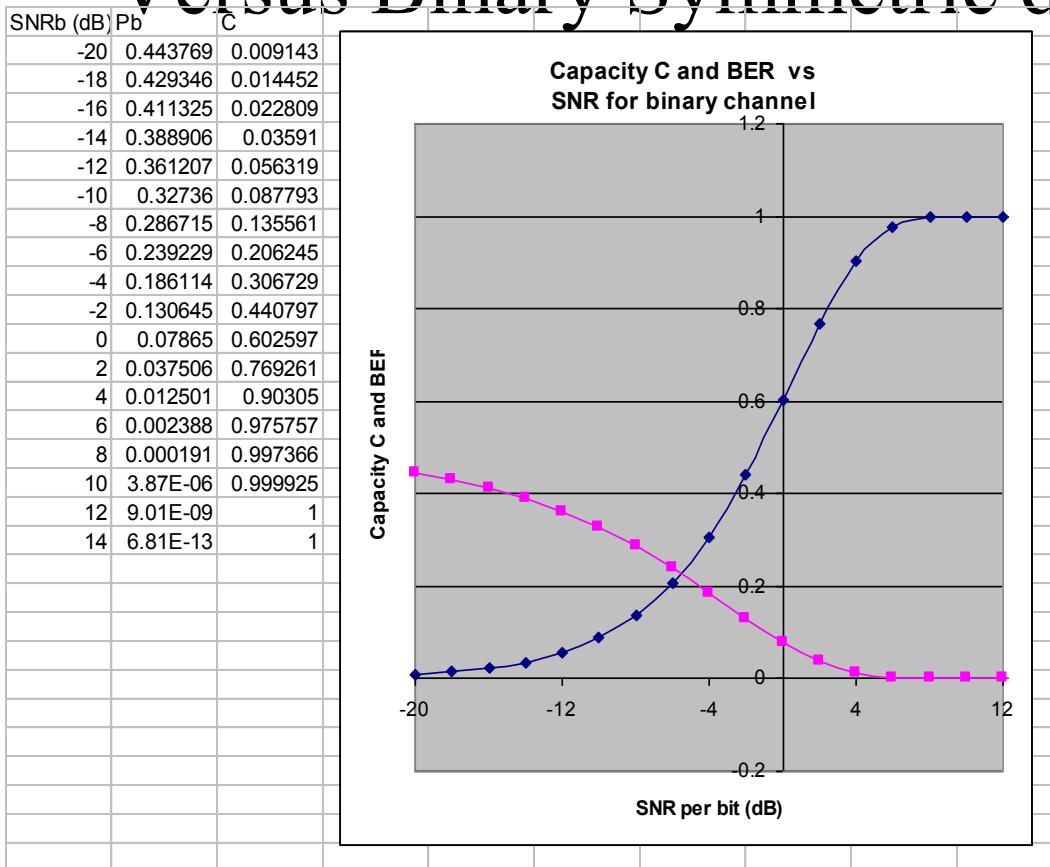
- Binary Symmetric PAM-continuous
- Maximum Information when:

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



# Channel Capacity Discrete Memoryless Channel

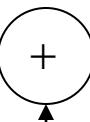
- Binary Symmetric PAM-continuous
- Versus Binary Symmetric discrete



# 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:

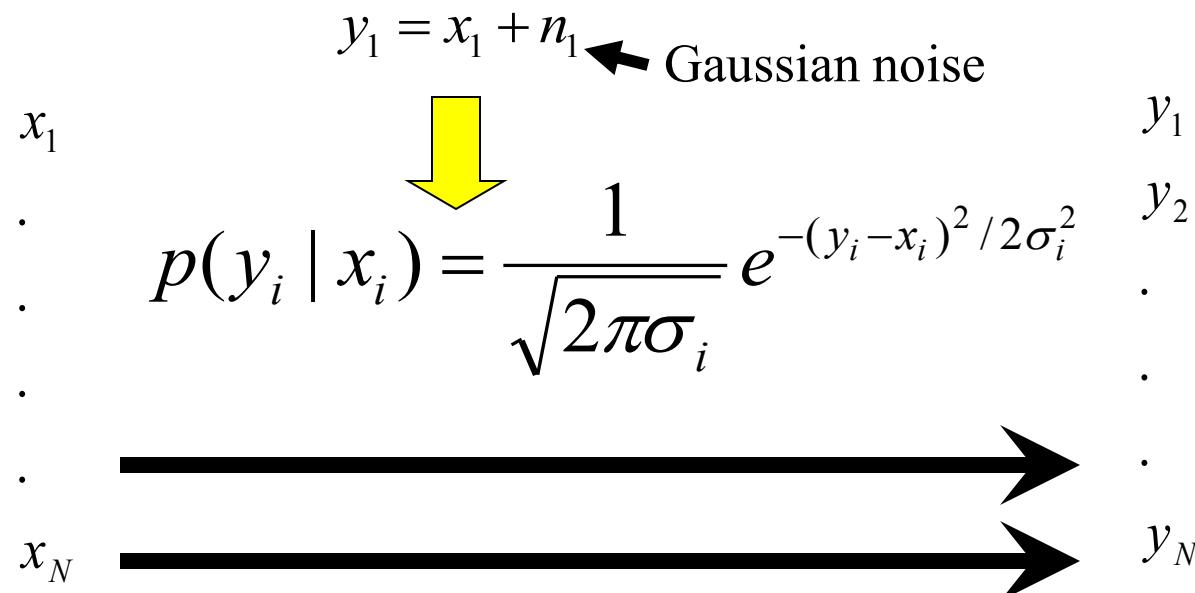
$$\begin{array}{ccc} x(t) & \xrightarrow{\hspace{1cm}} & y(t) \\ & & \end{array}$$



$$= \sum_{i=1}^N x_i f_i(t) \qquad \qquad \qquad = \sum_{i=1}^N y_i f_i(t)$$
$$= \sum_{i=1}^N n_i f_i(t)$$

# Discrete Memoryless Channel

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

$$y_1 = x_1 + n_1 \quad \text{Gaussian noise}$$
$$\begin{matrix} x_1 \\ \cdot \\ \cdot \\ \cdot \\ \cdot \\ \cdot \\ x_N \end{matrix} \xrightarrow{\hspace{1cm}} p(y_i | x_i) = \frac{1}{\sqrt{2\pi\sigma_i^2}} e^{-(y_i - x_i)^2 / 2\sigma_i^2} \xrightarrow{\hspace{1cm}} \begin{matrix} y_1 \\ y_2 \\ \cdot \\ \cdot \\ \cdot \\ y_N \end{matrix}$$


# Discrete Memoryless Channel

- Continuous-continuous
- Capacity is (Shannon)

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

$$N = 2WT$$

$$I(\mathbf{X}_N; \mathbf{Y}_N) = \int_{\mathbf{X}_N} \cdots \int_{\mathbf{Y}_N} p(\mathbf{y}_N | \mathbf{x}_N) p(\mathbf{x}_N) \log \frac{p(\mathbf{y}_N | \mathbf{x}_N)}{p(\mathbf{y}_N)} d\mathbf{x}_N d\mathbf{y}_N$$

$$= \sum_{i=1}^N \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} p(y_i | x_i) p(x_j) \log \frac{p(y_i | x_i)}{p(y_i)} dy_i dx_i$$

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

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

# Discrete Memoryless Channel

- Continuous-continuous
- Maximum Information when:

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

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

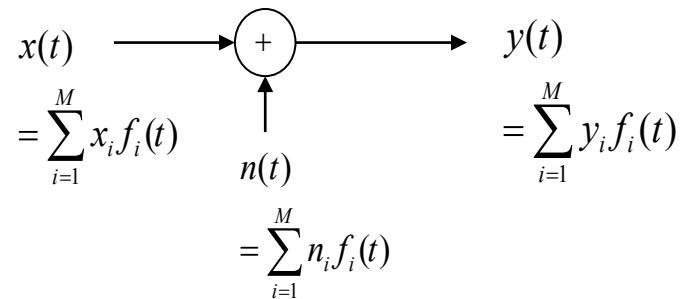
- Continuous-continuous
- Thus Capacity is:

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

$$= \lim_{T \rightarrow \infty} W \log \left( 1 + \frac{2\sigma_x^2}{N_0} \right)$$

$$= W \log \left( 1 + \frac{P_{av}}{WN_0} \right)$$

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



# Discrete Memoryless Channel

- Continuous-continuous

- Thus Normalized Capacity is:

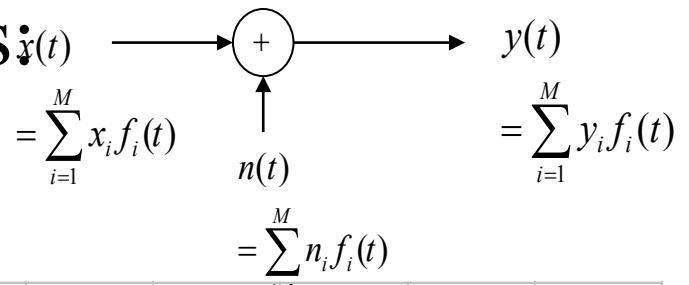
$$\frac{C}{W} = \log_2 \left( 1 + \frac{P_{av}}{WN_0} \right), \text{ but } P_{av} = C\mathcal{E}_b$$

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

$\Rightarrow$

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

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



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

