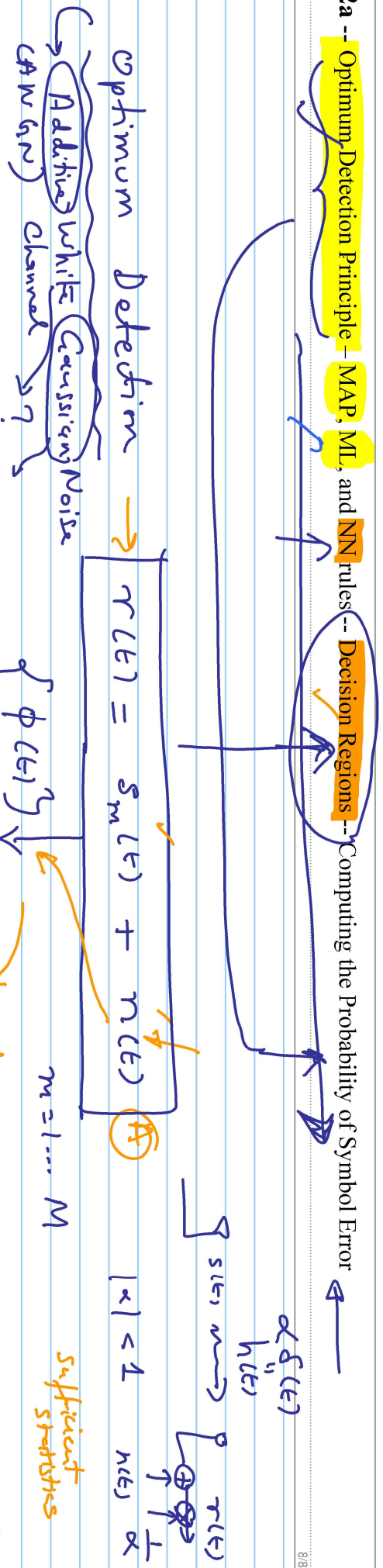


Lesson 2a -- Optimum Detection Principle -- MAP, ML, and NN rules -- Decision Regions -- Computing the Probability of Symbol Error



Probability of Correct Decision

$$\begin{aligned}
 P(C) &= \sum_{m=1}^2 P(C|s_m) P(s_m) \\
 &= \sum_{m=1}^2 P(s_m) \int_{\mathcal{R}_m} p(r|s_m) dr \\
 &= \sum_{m=1}^2 \int_{\mathcal{R}_m} p(r|s_m) p(s_m) dr
 \end{aligned}$$

$\left. \begin{array}{l} \text{exclusive} \\ \text{correct decision} \\ \text{transmitted} \end{array} \right\} s_m \in \{ +d\sqrt{\frac{E_s}{2}}, -d\sqrt{\frac{E_s}{2}} \}$

$s_m \in \{ +d, -d \}$

\mathcal{R}_1
 \mathcal{R}_2
 s_1
 s_2
 $-d$
 d
 ϕ
 $\text{TH} \rightarrow \text{threshold}$
 Decision Boundary

To maximize

$$P(c)$$

\Rightarrow

maximize

$$P(r|s_m) \cdot P(s_m)$$

$\forall s_m$

$$P(a|b)P(b) = P(b|a)P(a)$$

\Rightarrow

maximize

$$P(s_m|r) P(r)$$

$\forall s_m$

\Rightarrow

maximize

$$P(s_m|r)$$

$\forall s_m$

since prior is common

MAP
Maximum A Posteriori Probability
probabilities that the symbol s_m was sent having "observed" r

Maximum A Posteriori (MAP) Rule

Pick

$$s = s_i \quad (i=1, 2, \dots, M)$$

Symbol decision

iff

$$P(s_i|r) \geq P(s_j|r) \quad \forall j \neq i \rightarrow \textcircled{1}$$

$$P(r|s_i) P(c_i)$$

$$\geq P(r|s_j) P(c_j)$$

Bayes' Rule

and

$$P(s_i) = \frac{1}{M} \quad \forall i$$

\rightarrow equal priors

Symbol probabilities

Maximum Likelihood (ML) Rule

Pick $\hat{s} = s_i$ iff

$$p(r|s_i) \geq p(r|s_j)$$

$\forall j \neq i \rightarrow \textcircled{2}$

Recall $r = s_i + n$ $\xrightarrow{\text{Gaussian pdf}}$ $f(n) = \frac{1}{\sqrt{2\pi} \frac{N_0}{2}}$

$$e^{-\frac{n^2}{2 \times \frac{N_0}{2}}}$$

$$\sigma^2 = \frac{N_0}{2}$$

$$p(r|s_i) = P(\text{Noise takes a value } r - s_i)$$

$$= \frac{1}{\sqrt{\pi N_0}} e^{-\frac{(r-s_i)^2}{N_0}}$$

$$\text{Pick } \hat{s} = s_i \text{ iff } \frac{1}{\sqrt{\pi N_0}} e^{-\frac{(r-s_i)^2}{N_0}}$$

$$\geq \frac{1}{\sqrt{\pi N_0}} e^{-\frac{(r-s_j)^2}{N_0}}$$

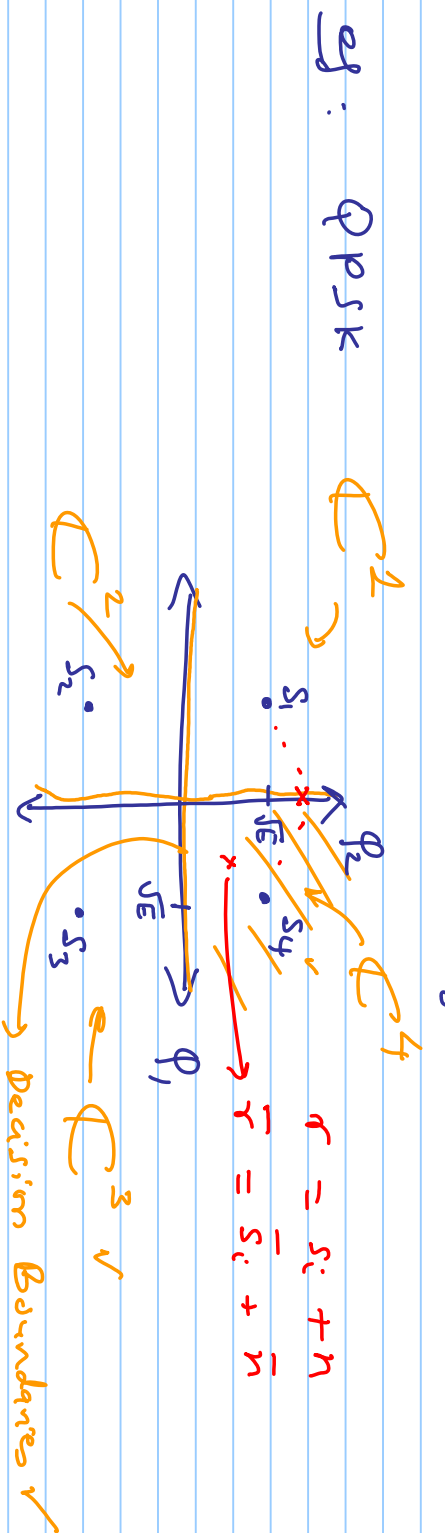
$\forall j \neq i$ &
 $j = 1, \dots, M$

exp(.) \rightarrow Monotonic fn.

Nearest Neighbors Taking log (-) on both sides and recognizing the (-)ve sign

Pick $\hat{s} = s_i$ iff $(r - s_i)^2 \leq (r - s_j)^2 \quad \forall j \neq i \& j=1 \dots M$

Decision Boundaries & Decision Regions



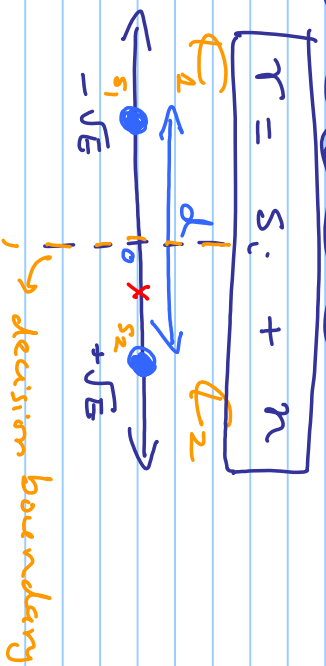
$\mathcal{R} \rightarrow$ Real
 $\mathcal{C} \rightarrow$ Complex

→ Probability of Symbol Error (SER)

$$s_i \in \{ +\sqrt{E_s}, -\sqrt{E_s} \} \quad p(s_1) = p(s_2) = \frac{1}{2}$$

$$\text{noise } n \in \mathcal{N}(0, \frac{N_0}{2}) \quad \checkmark$$

$$E[n] = 0 ; E[n^2] = \frac{N_0}{2} = \sigma_n^2$$



Let s_1 be sent
 decision error $r = -\sqrt{E_s} + n$

$$p(e) = \underbrace{p(e|s_1)}_{p_1} p(s_1) + p(e|s_2) p(s_2)$$

$$P_1 = P(r \in C_2 \mid s_1 \text{ sent}) = \int_{C_2} p(r|s_1) dr$$

$$= P(r \geq 0 \mid s_1 \text{ sent})$$

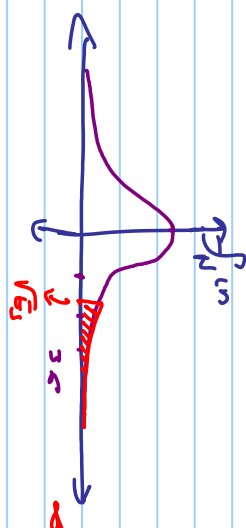
$$= P(n \geq \sqrt{E_s} \mid s_1 \text{ sent})$$



$$= P(n \geq \sqrt{E_s})$$

← n & s : are statistically independent

$$= \int_{\sqrt{E_s}}^{\infty} f_n(n) dn = \int_{\sqrt{E_s}}^{\infty} \frac{1}{\sqrt{\pi} N_0} e^{-n^2/N_0} dn$$



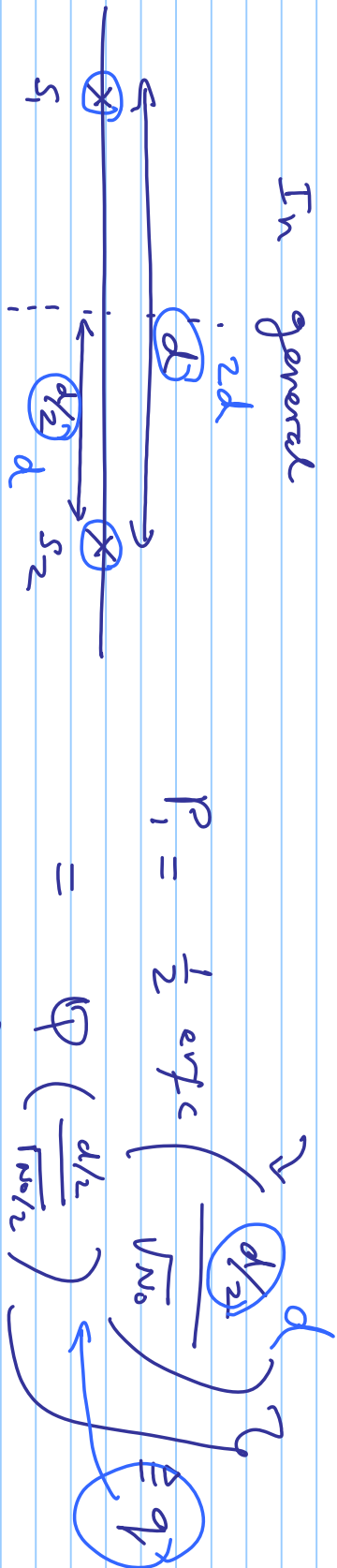
use $\beta = \frac{n}{\sqrt{N_0}} \Rightarrow d\beta = \frac{dn}{\sqrt{N_0}}$ when $n = \sqrt{E_s}$, $\beta = \sqrt{\frac{E_s}{N_0}}$

$$P_1 = \frac{1}{\sqrt{\pi}} \int_{\sqrt{\frac{E_s}{N_0}}}^{\infty} e^{-\beta^2} d\beta = \frac{1}{2} \operatorname{erfc}\left(\sqrt{\frac{E_s}{N_0}}\right)$$

$$= Q\left(\sqrt{\frac{E_s}{N_0/2}}\right)$$

$$= Q$$

In general



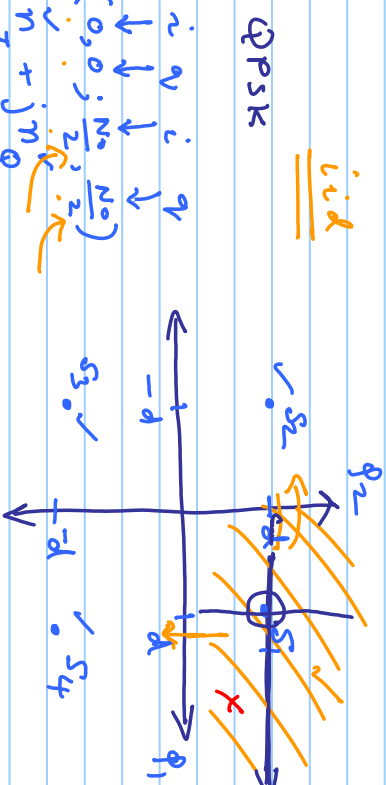
decision boundary (\perp bisectors connecting the 2 points)

Ex #1 :

$$r = s_i + n$$

QPSK

iid

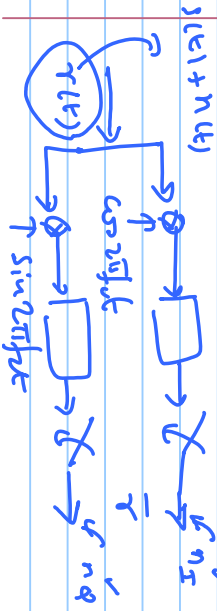


complex $\rightarrow r = s_i + n$

$$s_i = s_{iI} + j s_{iQ}$$

uncorrelated $\rightarrow n_I, n_Q$

$$E[n_I n_Q] = 0$$



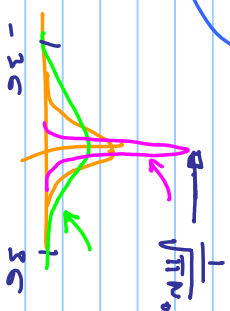
QPSK signal sent thru an AWGN channel

$P_e \rightarrow$ Avg. Probability of Symbol Error

$\hookrightarrow P_e \rightarrow$ Avg. \hookrightarrow Great Symbol Detection

$$P_e = 1 - P_c$$

$$\frac{1}{\sqrt{\pi N_0}} e^{-\frac{r^2}{2N_0}}$$

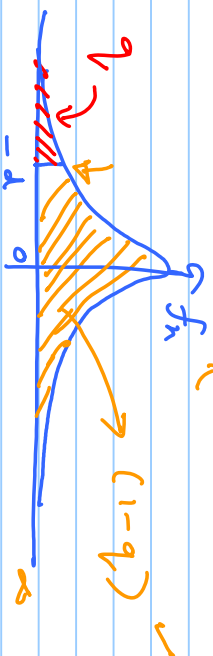
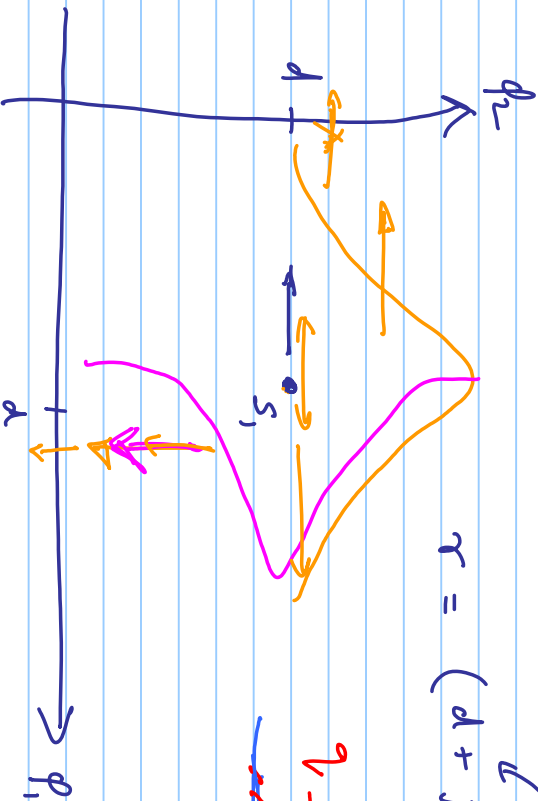


$$r = s_i + n$$



$$r = e^{j\theta} s_i + n$$

$$r = (d + jd) + (n_I + jn_Q)$$



$$P_c = \sum_{i=1}^4 P(c|s_i) P(s_i) = \frac{1}{4}$$

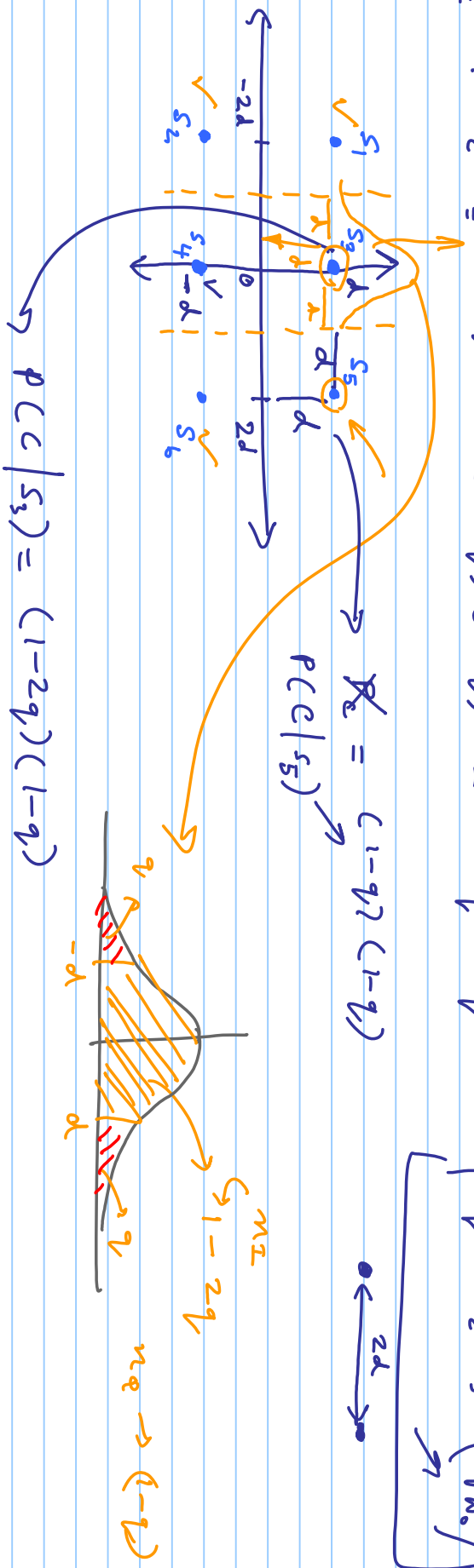
$$\therefore P(c|s_1) = (1-q) \cdot (1-q)$$

$$= P_c$$

$$P_e = 1 - P_c = 1 - (1 - 2q)(1 - q) = 2q - q^2$$

$$q = \frac{1}{2} \operatorname{erfc} \left(\frac{d}{\sqrt{N_0}} \right)$$

Ex #2

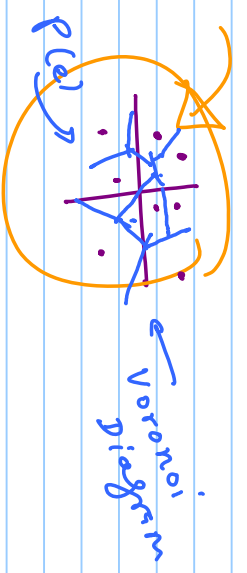


$$P_c = (1 - 2q)(1 - q)$$

$$P(c|s_3) = (1 - 2q)(1 - q)$$

$$P(c) = \frac{1}{6} (1 - q)^2 + \frac{2}{6} (1 - 2q)(1 - q)$$

$$P(c) = 1 - P(e)$$



$$P(s_i) = \frac{1}{M} \quad \forall i: s_1, s_2, \dots, s_M$$

$$r = s_i + n \rightarrow \mathcal{N}(0, \frac{N_0}{2})$$