

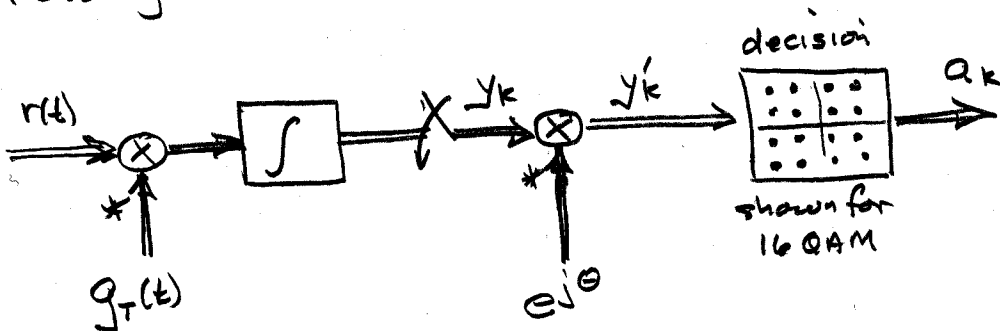
## 7.4 Coherent and Quasi-Coherent Detection

7.4.1

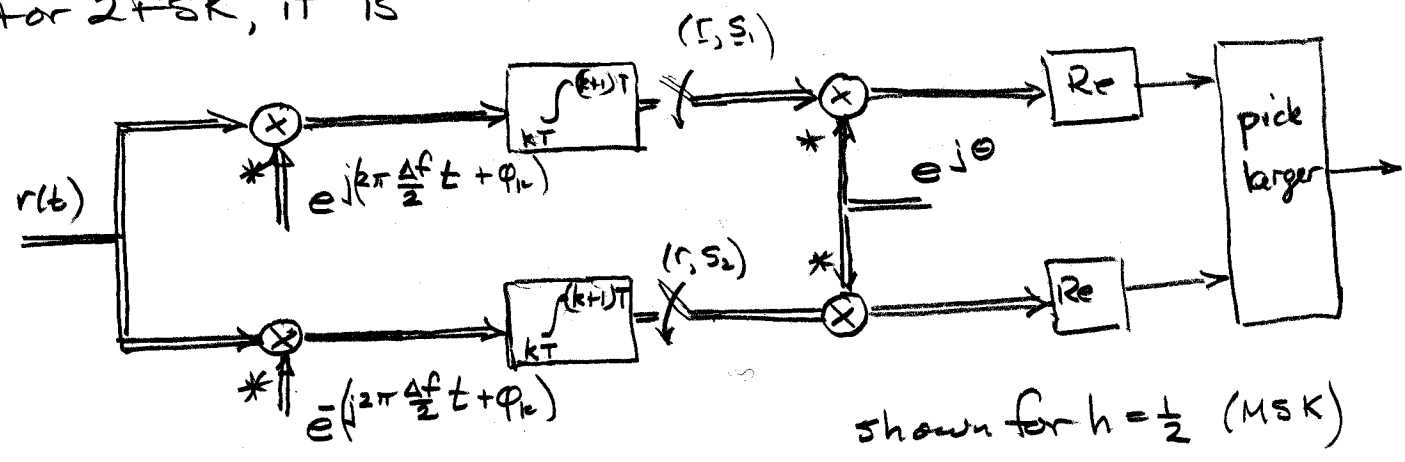
- "coherent detection": you have a perfect phase ref  $\hat{\theta} = \theta$  and you use it to de-rotate the received signal
- "quasi-coherent detection": you have a reasonably good phase reference, and you decide to use it as if it were perfect and hope for the best.
- Not too much to say here. Just:
  - structure of coherent receivers
  - performance of coherent receivers
  - justification of the quasi-coherent approach

### Coherent Receiver Structures

- For linear modulation it is just what we have been doing



• For 2FSK, it is



- receive  $r(t) = e^{j\theta} s_i(t) + n(t)$

lump  $e^{j\theta}$  with  $s_i(t)$  to make modified signals  $s'_i(t) = e^{j\theta} s_i(t)$

- with some orthonormal basis, we have

$$r = s'_i + n$$

- equiprob signals, MAP detector has

$$\hat{i} = \arg \min_i |r - s'_i|^2 = \arg \min_i (r - s'_i)^+ (r - s'_i)$$

$$= \arg \min_i (|r|^2 - 2 \operatorname{Re}[s'_i{}^+ r] + |s'_i|^2)$$

$$= \arg \max_i \operatorname{Re}[r, s'_i]$$

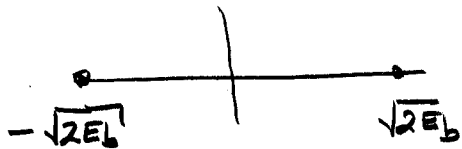
$$= \arg \max_i \operatorname{Re} \left[ e^{-j\theta} \int r(t) s_i^*(t) dt \right]$$

a reminder that things are complex

## Performance of Coherent Detection

- Coherent detectors eliminate phase offset, so performance is just what we derived in Sections 5 and 6

- Example: BPSK, a binary antipodal set



noise variance, each dimension

$$\sigma_w^2 = N_0$$

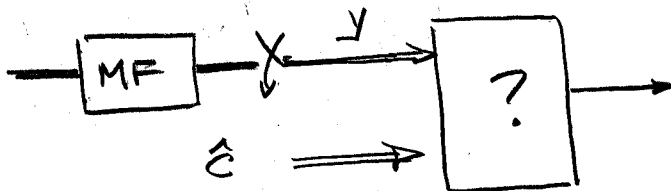
$$P_b = Q\left(\frac{\sqrt{2E_b}}{\sigma_w}\right) = Q\left(\sqrt{\frac{2E_b}{N_0}}\right) = Q(\sqrt{2}\sigma_b) \text{ as before.}$$

## Justification of Quasi-Coherent Structure

- Suppose we follow a feedforward or pilot based approach and create an estimate of the gain  $c = e^{j\theta}$

$$\hat{c} = c + w \quad \begin{array}{l} w \text{ is Gaussian noise} \\ \text{var } \sigma_w^2 = \frac{1}{2} E[|w|^2] \end{array}$$

- We have for linear modulation:



- Received sample is  $y = c s_i + v$  complex scalar

but  $c = \hat{c} - w$ , so

$$y = \hat{c} s_i - \underbrace{w s_i}_{u, \text{ Gaussian}} + v = \hat{c} s_i + u$$

$$\sigma_{u_i}^2 = \sigma_w^2 E_i + \nu$$

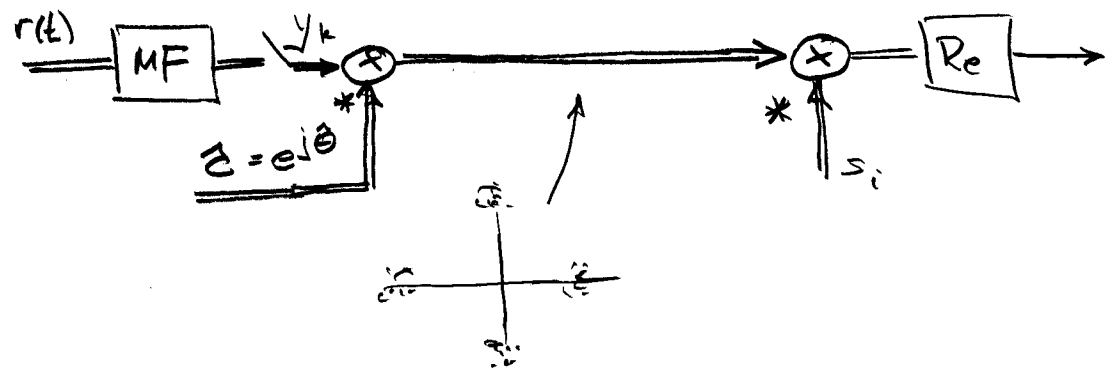
Since we have  $\hat{c}$  at hand, we should do

$$\hat{i} = \underset{i}{\operatorname{argmax}} P(y | s_i, \hat{c}) = \frac{1}{2\pi\sigma_{u_i}^2} \exp\left(-\frac{|y - \hat{c} s_i|^2}{2\sigma_{u_i}^2}\right)$$

$$= \underset{i}{\operatorname{argmin}} \left( \frac{|y - \hat{c} s_i|^2}{2\sigma_{u_i}^2} + \ln(\sigma_{u_i}^2) \right) = \underset{i}{\operatorname{argmin}} \left( |y - \hat{c} s_i|^2 + 2\sigma_{u_i}^2 \ln(\sigma_{u_i}^2) \right)$$

$$= \underset{i}{\operatorname{argmin}} \left( |y|^2 - 2\operatorname{Re}[y s_i^* \hat{c}^*] + E|s_i|^2 + 2\sigma_{u_i}^2 \ln(\sigma_{u_i}^2) \right)$$

If equal energy  $\hat{i} = \underset{i}{\operatorname{argmax}} \operatorname{Re}[y s_i^* \hat{c}^*]$



- Performance? Difficult integrals of Bessel functions.  
Ugh. Closed form in terms of Marcum Q-function.