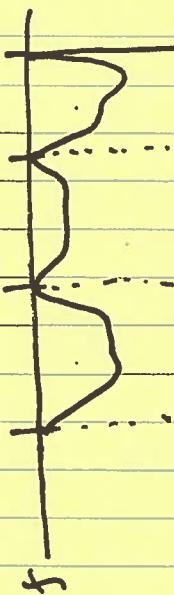


## Lecture 33

### Multirate signal processing

Taken a signal,  $x(n)$ , sampled at  $2f_3$ , and converted it to 3 signals sampled at 3 different sample rates.

→ multirate signal processing



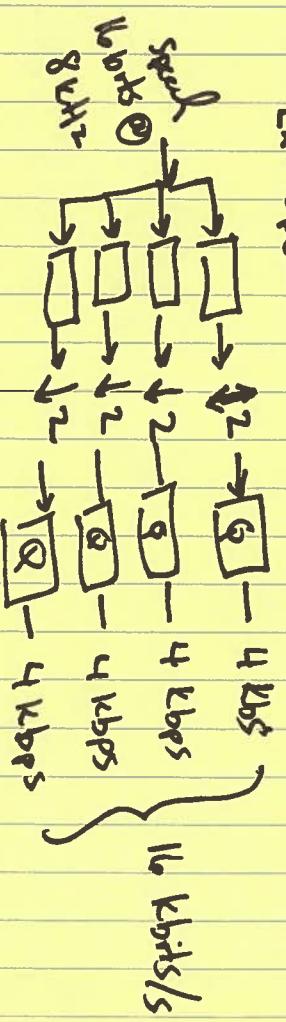
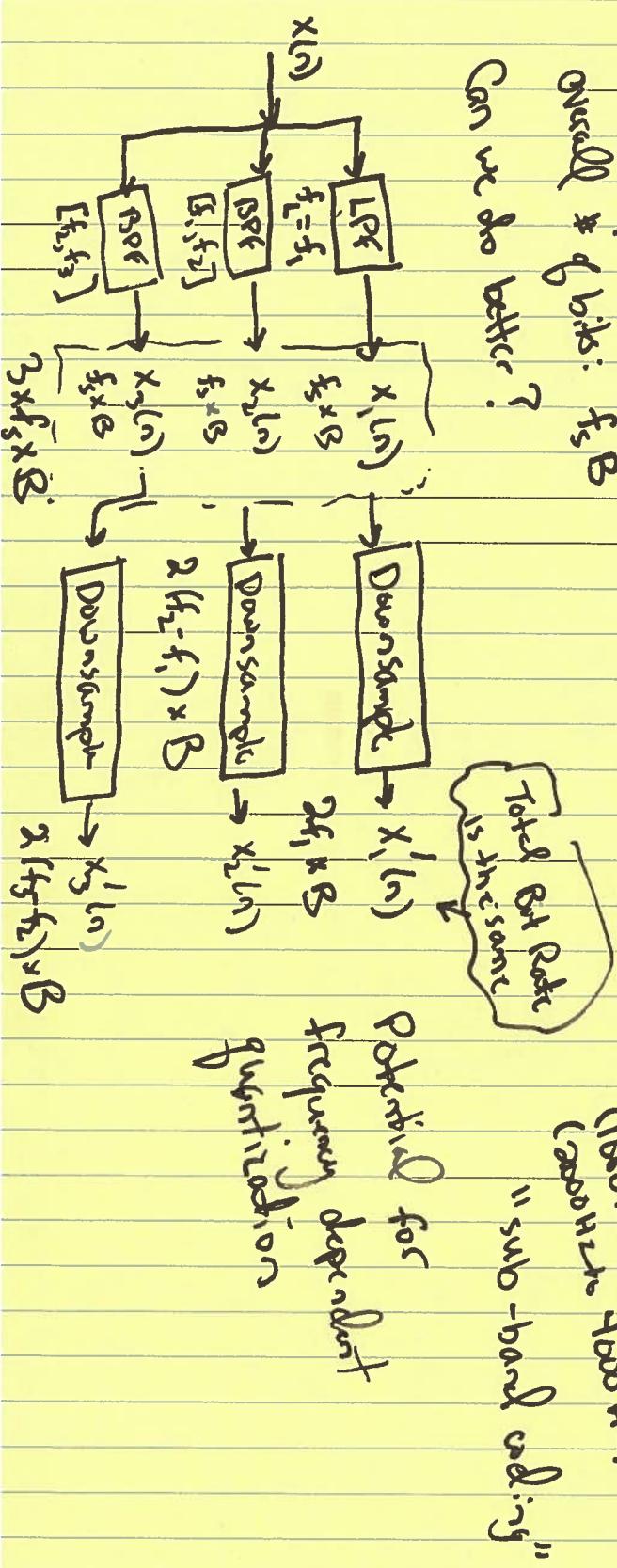
- $[0, f_1]$ : slowly-varying
- $[f_1, f_2]$ : not that much critical
- $[f_2, f_3]$ : slowly-varying

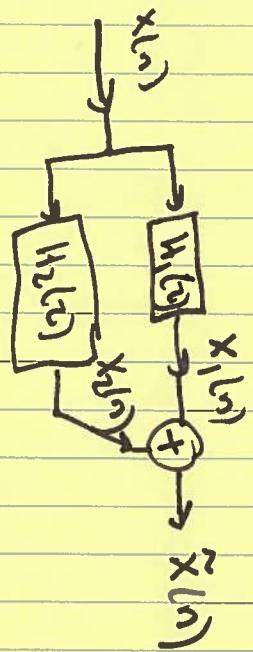
The sampling theorem states:  $f_s > 2f_3$

Analyses of quantization error: SNR  $\propto 6B$  (B bits)

Overall # of bits:  $f_s B$

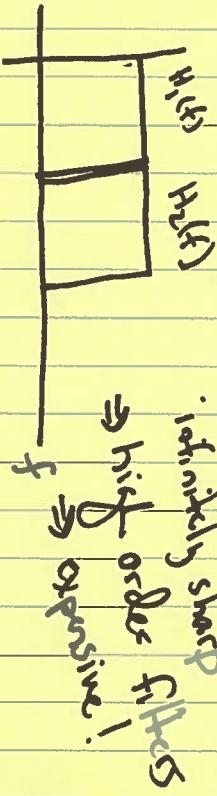
Can we do better?





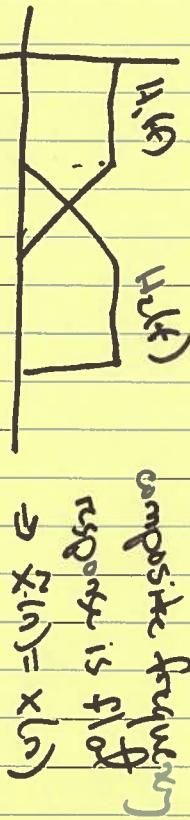
Requirement:  $\hat{x}(n) \approx x(n)$

How do we achieve this?



• infinitely sharp  
⇒ high order filters  
• ⇒ expensive!

More reasonable approach:



composite frequency  
response is flat  
 $\Rightarrow \hat{x}(n) = x(n)$

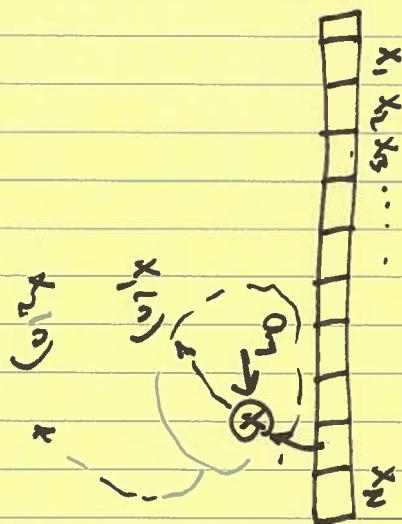
this problem has been extensively studied  
the result was a filter structure known  
as Quadrature mirror filters (QMF).  
Note: computation of QMF difficult.

Similarly have

$$x(n) = \sum_{m=0}^{M-1} x(m) h(n-m)$$

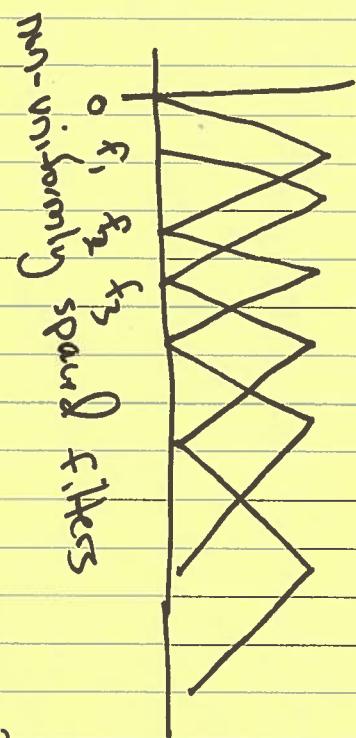
Recall filter transforms:

$$H(z) \rightarrow z \rightarrow \frac{1}{2} \rightarrow H_{\text{IRF}}$$



## Extensions to Pattern Recognition:

(3)



$$x^{(n)} \rightarrow \boxed{g(x^{(n)})} = \alpha y^{(n-1)} + x^{(n)}$$

$$\boxed{g(x^{(n)})} \rightarrow y^{(n)} = \alpha y^{(n-1)} + g(x^{(n)})$$

Replaced with an RBF.

