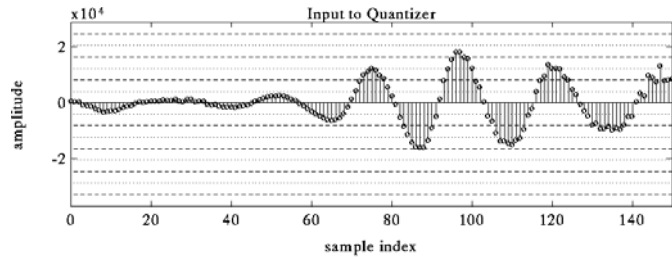
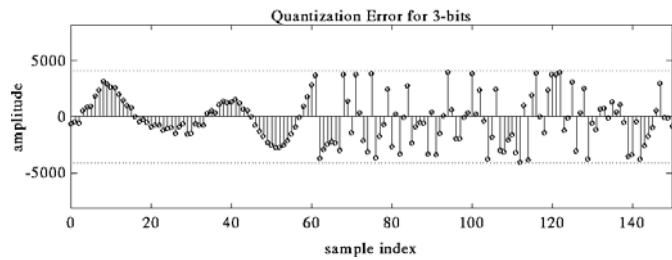




### 3-Bit Speech Quantization Error

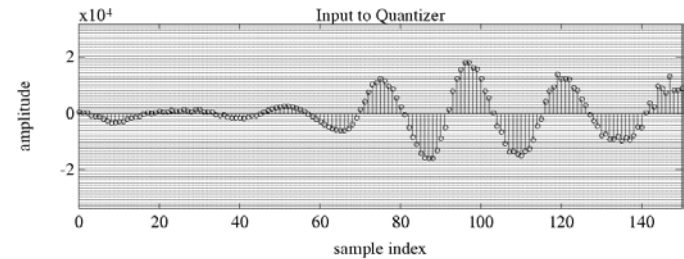


Input to quantizer

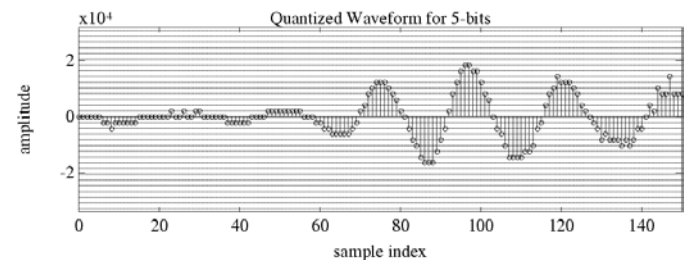


3-bit quantization error

### 5-Bit Speech Quantization

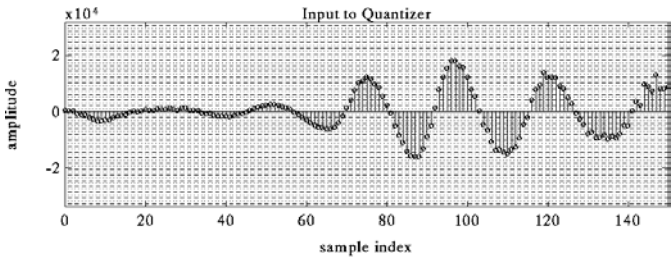


Input to quantizer

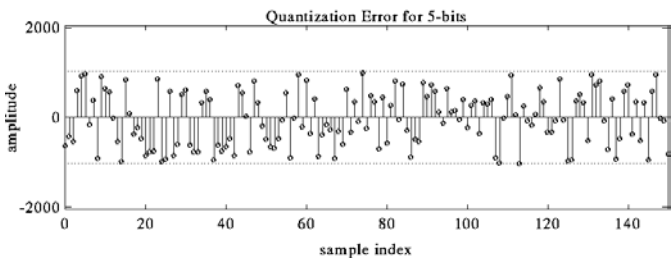


5-bit quantized waveform

### 5-Bit Speech Quantization Error



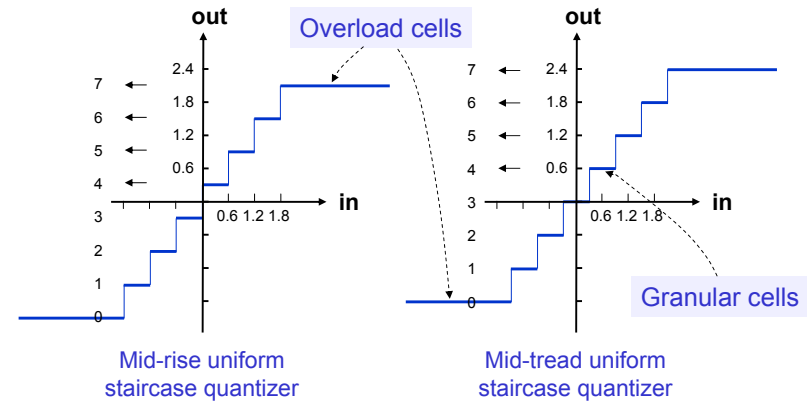
Input to quantizer



5-bit quantization error

### Scalar Quantization & Quantization Error

granular noise → error in granular cells  
 overload noise → error in overload cells



## Vector Quantization

- Distortion function:

$$\delta(x, y)$$

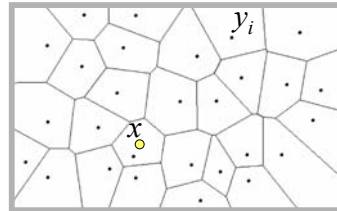
measures the dissimilarity between  $x$  and  $y$ .

- Centroid of a region  $Y$ :

$$\bar{y} = \min_y \int_Y \delta(x, y) dx$$

- Voronoi region:

$$Y_i = \{x; \delta(x, y_i) = \min_{j=1,2,\dots,N} \delta(x, y_j)\}$$



## Granular Quantization Error

- Each sample is quantized and each sample has a quantization error defined as

$$e[n] = \hat{x}[n] - x[n]$$

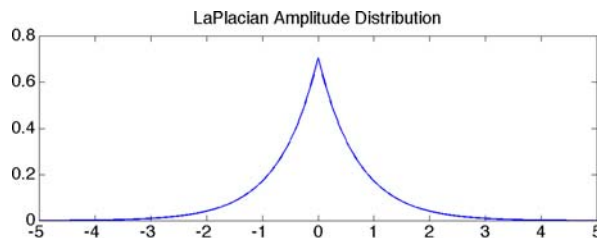
- Since each sample falls in an interval of length  $\Delta$ , and the quantized sample falls in the middle of that interval,

$$-(\Delta / 2) < e[n] \leq (\Delta / 2).$$

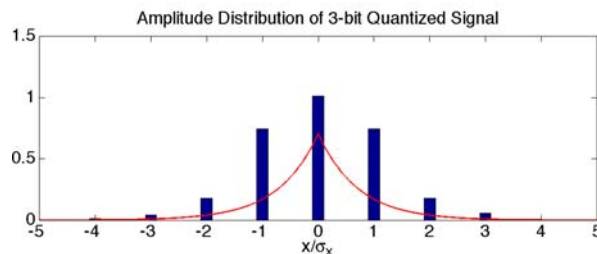
- We call this “quantization noise” because it seems to vary randomly. Clearly, the strength (power) of this noise is proportional to  $\Delta$ ; *i.e.*,

$$\sigma_e^2 = K\Delta^2$$

## Typical Amplitude Distributions



Laplacian distribution is often used as a model for speech signals

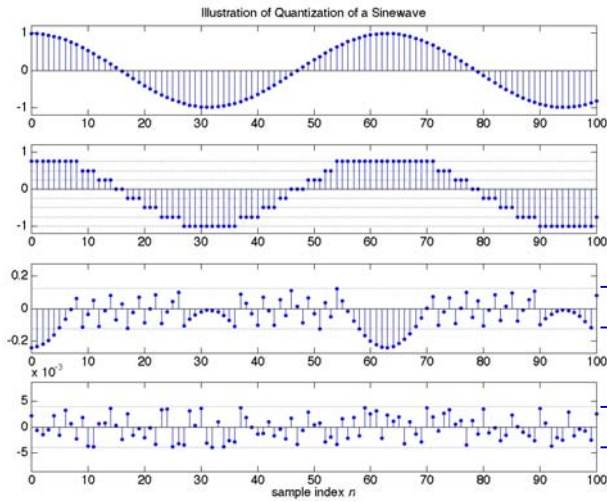


A 3-bit quantized signal has only 8 different values

## Probabilistic Model for Quantization

- We observed that the quantization error has very complicated variations that suggest a random or noise-like character.
- Random signals are represented by probability distributions and averages such as
  - Mean and mean-square (average power)
  - Histograms
  - Autocorrelation function
  - Power spectrum
- This is a good way to think about quantization noise.

## Quantization of a Sine Wave



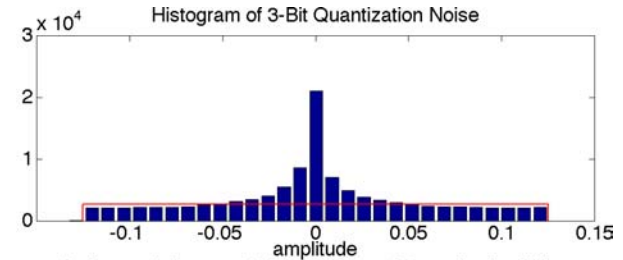
Unquantized  
sinewave

3-bit  
quantization  
waveform

3-bit  
quantization  
error

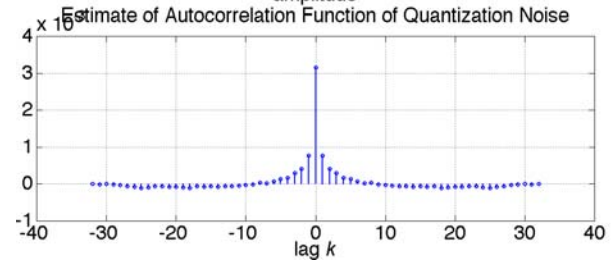
8-bit  
quantization  
error

## 3-Bit Quantization Noise

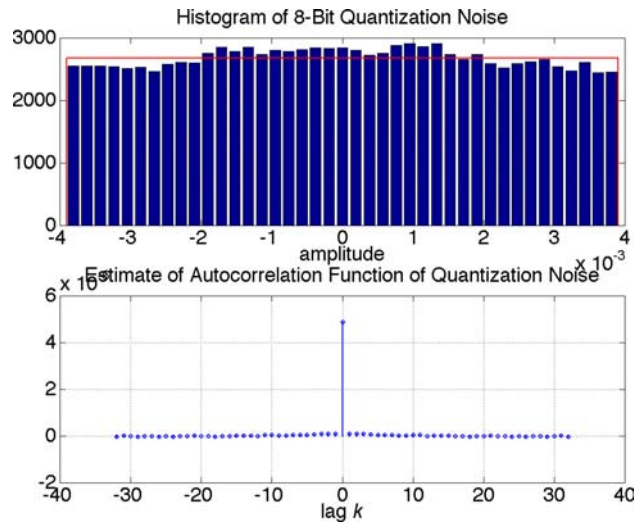


$$\Delta = 2 / 2^3$$

$$= 0.25$$



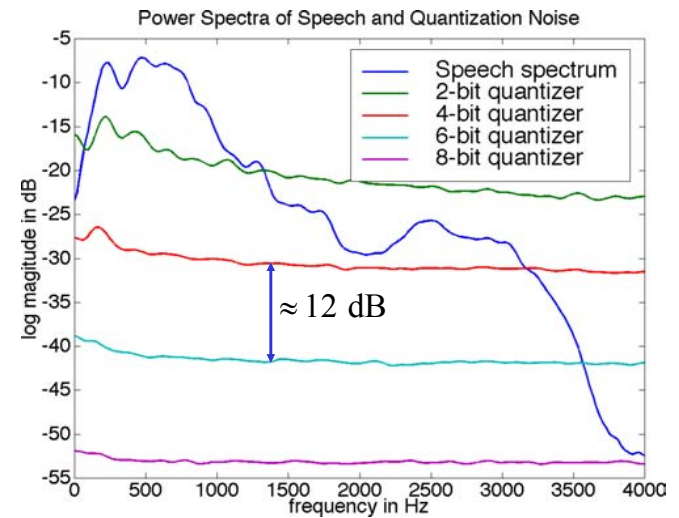
## 8-Bit Quantization Noise



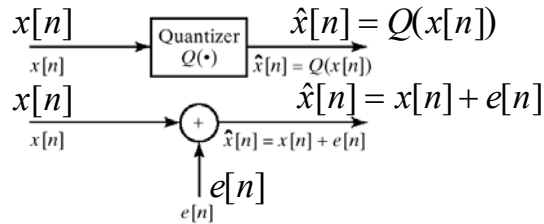
$$\Delta = 2 / 2^8$$

$$= 0.0078$$

## Spectra of Quantization Noise



## Linear Noise Model



- Error is uncorrelated with the input.
- Error is uniformly distributed over the interval  $-(\Delta/2) < e[n] \leq (\Delta/2)$ .
- Error is stationary white noise, (i.e. flat spectrum)

$$\Phi_{ee}(\omega) = \sigma_e^2 = \int_{-\Delta/2}^{\Delta/2} \frac{1}{\Delta} e^2 de = \frac{\Delta^2}{12}, \quad |\omega| \leq \pi$$

## Quantizer Signal-to-Noise Ratio

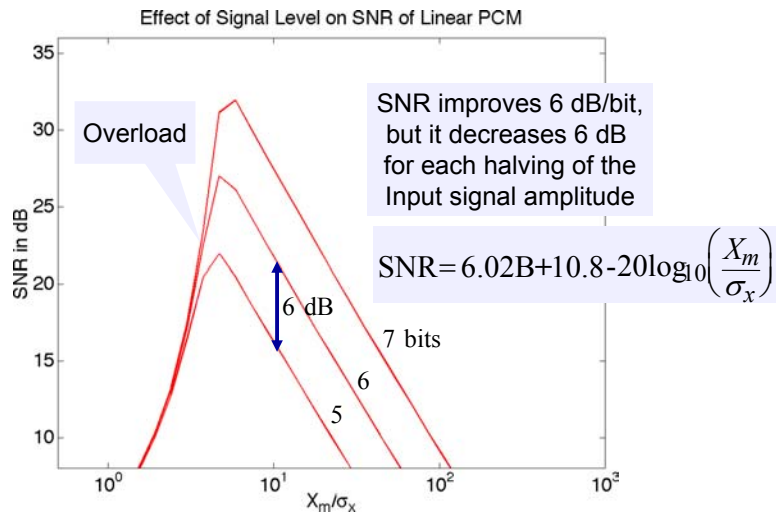
- Assume  $2^{(B+1)}$  levels and amplitude range  $2X_m$ . Then using a probabilistic analysis we obtain

$$\Rightarrow \underbrace{\Delta = \frac{2X_m}{2^{(B+1)}}}_{\text{step size}} = 2^{-B} X_m \Rightarrow \underbrace{\sigma_e^2 = \frac{2^{-2B} X_m^2}{12}}_{\text{noise power}}$$

- Therefore the quantizer SNR is:

$$\begin{aligned} \text{SNR} &= 10 \log_{10} \left( \frac{\sigma_x^2}{\sigma_e^2} \right) = 10 \log_{10} \left( \frac{12 \cdot 2^{2B} \sigma_x^2}{X_m^2} \right) \\ &= \mathbf{6.02B} + 10.8 - 20 \log_{10} \left( \frac{X_m}{\sigma_x} \right) \quad (\text{in dB}) \end{aligned}$$

## Variation of SNR with Signal Level



## Analysis of Fine Quantization

- Input is random variable  $x$  with pdf  $p_x(x)$
- Distortion is defined as MSE, i.e.

$$\delta(x, y) = \|x - y\|^2$$

- Quantizer  $Q$ ,  $Q(x) = y_i = \min_{j=1,2,\dots,N} \delta(x, y_j)$
- Reproduction codebook  $C$ ,  $C = \{y_j, j = 1, 2, \dots, N\}$
- Input space partition  $Y$ ,  $Y = \{Y_j, j = 1, 2, \dots, N\}$
- Average distortion

$$D(Q) = E[\|x - Q(x)\|^2] = \sum_{j=1}^N \int_{Y_j} \|x - y_j\|^2 p_x(x) dx$$

## Analysis of Fine Quantization (Large $N$ )

### Bennett assumptions

- $N$  is very large
- Partition is Voronoi – reproduction values are the regional centroids
- Partition is regular – covers the whole space such that overload cells have nearly zero probability measure
- Cell volumes  $V(Y_j) = \int_{Y_j} dx$  are all bounded and tiny
- $p_x(x)$  is smooth, meaning Riemann sums approach Riemann integral and mean value theorem applies

## Analysis of Fine Quantization (Large $N$ )

$$D(Q) = \sum_{j:V(Y_j)<\infty} \int_{Y_j} \|x - y_j\|^2 p_x(x) dx + \sum_{j:V(Y_j)=\infty} \int_{Y_j} \|x - y_j\|^2 p_x(x) dx$$

$$D(Q) \approx \sum_{j:V(Y_j)<\infty} \int_{Y_j} \|x - y_j\|^2 p_x(x) dx$$

$$\approx \sum_{j=1}^N p_x(y_j) \int_{Y_j} \|x - y_j\|^2 dx$$

since  $p_x(x)$  is smooth and mean value theorem applies

$$\text{But, } P_x(Y_j) = \int_{Y_j} p_x(x) dx \approx p_x(y_j) \int_{Y_j} dx = p_x(y_j) V(Y_j)$$

$$D(Q) \approx \sum_{j=1}^N P_x(Y_j) \int_{Y_j} \frac{\|x - y_j\|^2}{V(Y_j)} dx$$

← Relates to “normalized moment” – for further reading if interested

## Quantization Summary

- Quantization of signal values and results of computation is unavoidable in a digital system.
- We can analyze quantization error using a random noise model.
- The more bits in the number representation, the lower the noise. An oft-stated theorem is that “the signal-to-noise ratio increases 6 dB with each added bit”
  - however, remember that if the signal level decreases while keeping the quantizer step size the same, it is like throwing away bits!

## A/D & D/A Conversion

## Digital-to-Analog Conversion

- Reconstruction from sampled sequence using ideal lowpass filtering

$$X_r(j\Omega) = X(e^{j\Omega T})H_r(j\Omega)$$

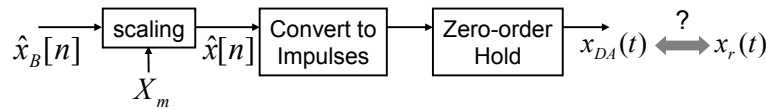
Discrete-time Fourier transform

$$\text{Ideal lowpass filter } H_r(j\Omega) = \begin{cases} T, & |\Omega| < \pi/T \\ 0, & |\Omega| > \pi/T \end{cases}$$

Ideal discrete to continuous conversion

$$x_r(t) = \sum_{n=-\infty}^{\infty} x[n] \frac{\sin[\pi(t-nT)/T]}{\pi(t-nT)/T}$$

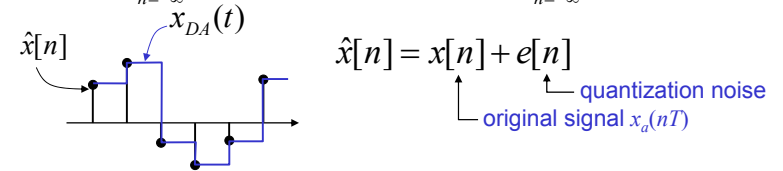
- Realizable implementation in circuitry – D/A converter



## D/A Conversion - II

$$x_{DA}(t) = \sum_{n=-\infty}^{\infty} \hat{x}[n]h_0(t-nT) = x_0(t) + e_0(t)$$

$$x_0(t) = \sum_{n=-\infty}^{\infty} x[n]h_0(t-nT) \quad e_0(t) = \sum_{n=-\infty}^{\infty} e[n]h_0(t-nT)$$

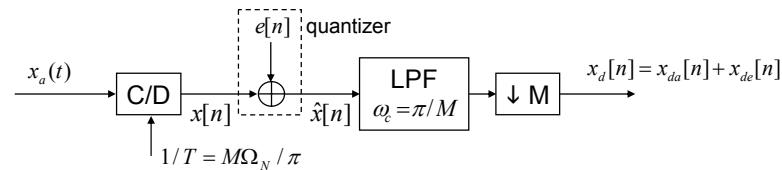


$$h_0(t) = \begin{cases} 1, & 0 < t < T \\ 0, & \text{otherwise} \end{cases} \iff H_0(j\Omega) = \frac{2 \sin(\Omega T/2)}{\Omega} e^{-j\Omega T/2}$$

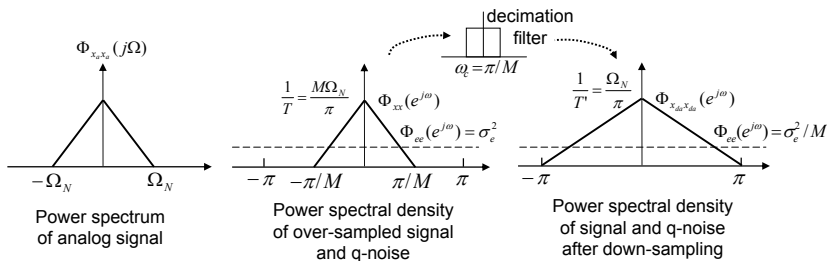
See Fig.4.54

Quantization noise aside, we need to compensate for the effect of zero order hold circuit to avoid spectral distortion  $\rightarrow \tilde{H}_r(j\Omega) = H_r(j\Omega)/H_0(j\Omega)$

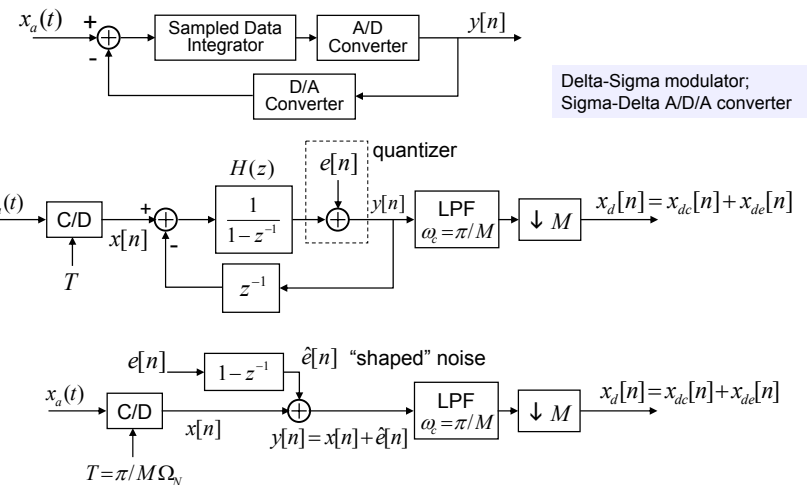
## Over-Sampling & Its Effect on Q-Noise



- $M$  is called oversampling ratio (above Nyquist rate)
- $M$  affects signal-to-quantization-noise ratio



## Oversampled A/D with Noise Shaping

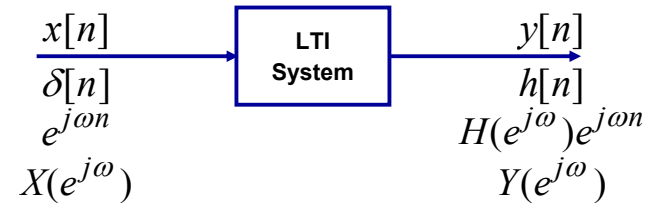


Delta-Sigma modulator;  
Sigma-Delta A/D/A converter

## Frequency Response of LTI Systems

## Convolution Theorem

$$y[n] = \sum_{k=-\infty}^{\infty} x[k]h[n-k] \Leftrightarrow Y(e^{j\omega}) = X(e^{j\omega})H(e^{j\omega})$$

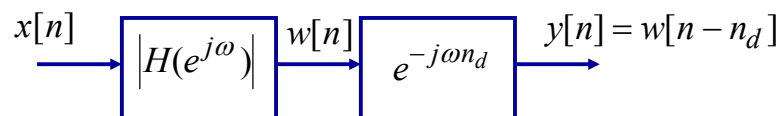


$$Y(e^{j\omega}) = H(e^{j\omega})X(e^{j\omega})$$

$$|Y(e^{j\omega})| = |H(e^{j\omega})||X(e^{j\omega})|$$

$$\angle Y(e^{j\omega}) = \angle H(e^{j\omega}) + \angle X(e^{j\omega})$$

## Linear Phase Means Delay



- Since we can consider the effects of magnitude and phase separately, it follows that linear phase of the form  $\angle H(e^{j\omega}) = -\omega n_d$  implies delay of  $n_d$  samples.
- Thus an ideal lowpass filter with delay has

$$h_{lp}[n] = \frac{\sin \omega_c (n - n_d)}{\pi (n - n_d)} \Leftrightarrow H(e^{j\omega}) = \begin{cases} e^{-j\omega n_d} & |\omega| < \omega_c \\ 0 & \omega_c < |\omega| < \pi \end{cases}$$

## Frequency Response Functions

$$H(e^{j\omega}) = |H(e^{j\omega})| e^{j\angle H(e^{j\omega})}$$

- Log-magnitude (in dB)

$$20 \log_{10} |H(e^{j\omega})|$$

- Phase (in radians)

$$\angle H(e^{j\omega}) = \arg [H(e^{j\omega})]$$

- Group delay (in samples)

$$\tau(\omega) = \text{grd} [H(e^{j\omega})] = -\frac{d}{d\omega} \{ \angle H(e^{j\omega}) \}$$



## Group Delay of a Linear Phase System

- Frequency response of a linear phase system is of the form

$$H(e^{j\omega}) = |H(e^{j\omega})| e^{-j\omega n_d}$$

- What is the group delay?

$$\begin{aligned} \tau(\omega) &= -\frac{d}{d\omega} \left\{ \angle H(e^{j\omega}) \right\} = -\frac{d}{d\omega} (-\omega n_d) \\ &= n_d \text{ samples} \end{aligned}$$

## Rational System Functions

- Consider a general difference equation of the form

$$\sum_{k=0}^N a_k y[n-k] = \sum_{k=0}^M b_k x[n-k]$$

- Rational system function of a causal and stable LTI system

$$H(z) = \frac{\sum_{k=0}^M b_k z^{-k}}{\sum_{k=0}^N a_k z^{-k}} = \frac{b_0 \prod_{k=1}^M (1 - c_k z^{-1})}{a_0 \prod_{k=1}^N (1 - d_k z^{-1})}$$

Causal  $\Rightarrow$

$$|z| > \max_k |d_k|$$

Stable  $\Rightarrow$

$$\max_k |d_k| < 1$$

## Impulse Response

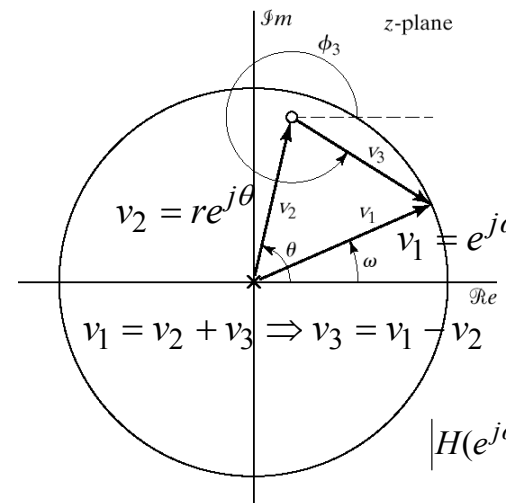
- We can make a partial fraction expansion of the rational system function:

$$H(z) = \underbrace{\left[ \sum_{r=0}^{(M-N)} B_r z^{-r} \right]}_{\text{if } M \geq N} + \sum_{k=1}^N \frac{A_k}{1 - d_k z^{-1}} \quad |z| > \max_k |d_k|$$

- The inverse  $z$ -transform gives the impulse response

$$h[n] = \underbrace{\left[ \sum_{r=0}^{(M-N)} B_r \delta[n-r] \right]}_{\text{if } M \geq N} + \sum_{k=1}^N A_k d_k^n u[n]$$

## Pole-Zero Plot



$$\begin{aligned} H(z) &= (1 - re^{j\theta} z^{-1}) \\ &= \frac{z - re^{j\theta}}{z} \end{aligned}$$

$$H(e^{j\omega}) = \frac{e^{j\omega} - re^{j\theta}}{e^{j\omega}}$$

$$= \frac{v_3}{v_1}$$

$$|H(e^{j\omega})| = \frac{|e^{j\omega} - re^{j\theta}|}{|e^{j\omega}|} = \frac{|v_3|}{|v_1|} = |v_3|$$

## Example

- System function:

$$H(z) = \frac{1}{(1 - re^{j\theta}z^{-1})(1 - re^{-j\theta}z^{-1})} = \frac{1}{1 - 2r \cos \theta z^{-1} + r^2 z^{-2}}$$

- Difference equation:

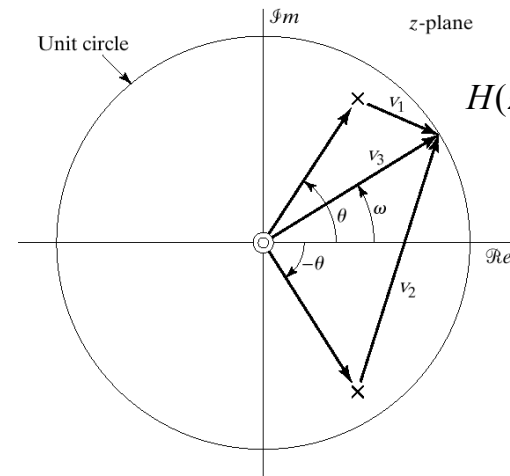
$$y[n] = 2r \cos \theta y[n-1] - r^2 y[n-2] + x[n]$$

- Impulse response:

$$H(z) = \frac{1}{(1 - re^{j\theta}z^{-1})} + \frac{1}{(1 - re^{-j\theta}z^{-1})}$$

$$h[n] = \frac{r^n \sin[\theta(n+1)]}{\sin \theta} u[n]$$

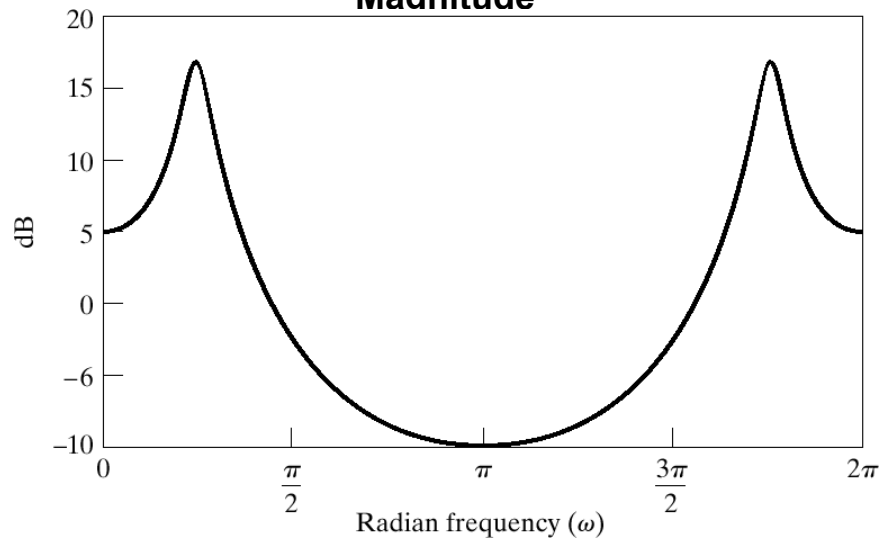
## Example



$$H(z) = \frac{z^2}{(z - re^{j\theta})(z - re^{-j\theta})}$$

$$|H(e^{j\omega})| = \frac{|v_3|^2}{|v_1| \cdot |v_2|}$$

## Magnitude



## Phase and Group Delay

