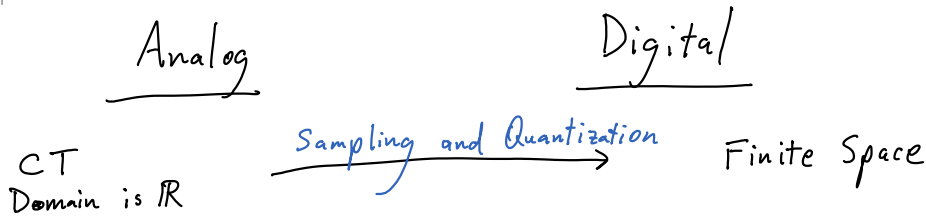
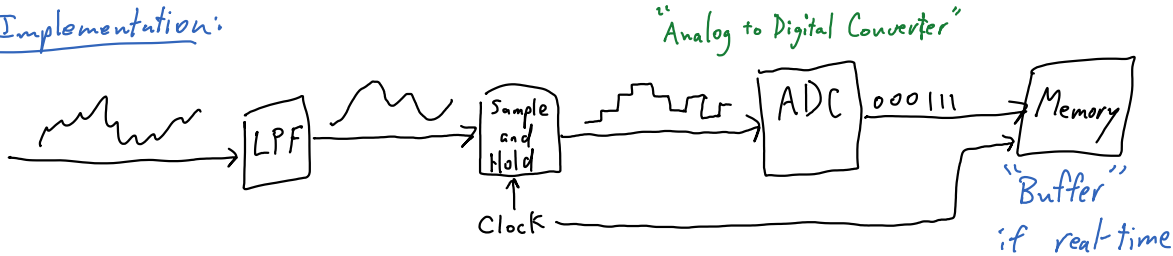


Lecture 12

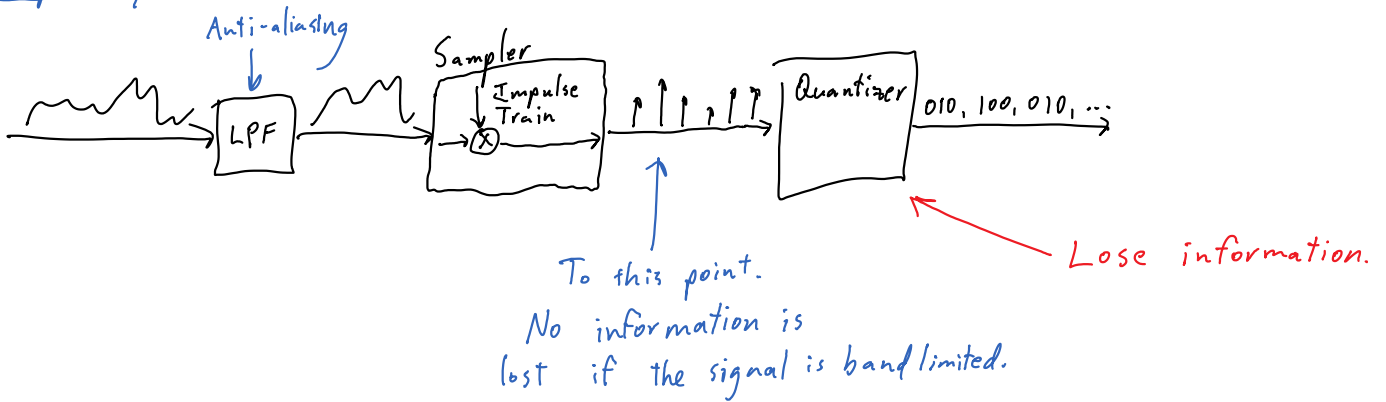
Monday, March 31, 2014  
3:39 PM



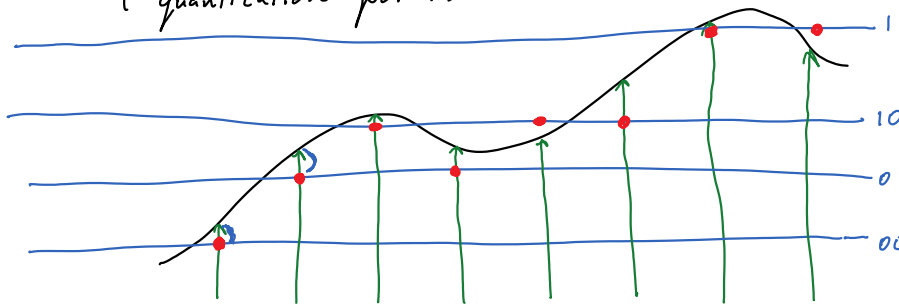
Implementation:



Conceptually:



Example: 2-bit quantization  
4 quantization points



00, 01, 10, 01, 10, 10, 11, 11

Let  $\hat{x}[n]$  be the reconstructed signal (red).

Then  $error[n] = x[n] - \hat{x}[n]$

error[n] "Noise"



CD Audio: 44.1 KHz sample rate  
 2 channels (stereo)  
 16 bit quantization (or 24)

$$2^{16} = 65,536 \text{ quant. points}$$

$$\Rightarrow 44,100 \times 2 \times 16 = 1,411,200 \text{ bits/sec.}$$

$$176,400 \text{ bytes/sec.}$$

$$10,584,000 \text{ bytes/min.}$$

10 MB/min

Doesn't count overhead.  
 Not even highest quality.

## Probability Primer:

$X$  is a random object.

Represents whether a team wins a game.

$$\Omega = \{ \text{win, loss, tie} \}$$

$$\begin{aligned} P(X = \text{win}) &= 70\% \\ P(X = \text{loss}) &= 20\% \\ P(X = \text{tie}) &= 10\% \end{aligned}$$

Non-negative  
 Sum to 1.

$$P(X = \text{win or } X = \text{loss}) = 70\% + 20\% = 90\%$$

Random Variable:  $\Omega = \mathbb{R}$

Discrete:  $X$  takes one of finitely many (or countably-many) values.

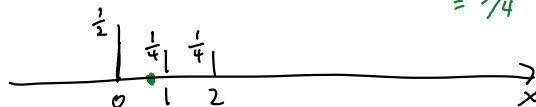
Probability mass function:

$$\text{Prob}(X=a) = p_x(a) \quad \forall a.$$

Example:

$$\text{Prob}(X \in S) = \sum_{a \in S} p_x(a)$$

$p_x$



$$EX = \left(\frac{1}{2}\right) \cdot 0 + \left(\frac{1}{4}\right) \cdot 1 + \left(\frac{1}{4}\right) \cdot 2 = \frac{3}{4}$$

$$\begin{aligned} \text{Var}(X) &= \left(\frac{1}{2}\right) \left(\frac{3}{4}\right)^2 + \left(\frac{1}{4}\right) \left(\frac{1}{4}\right)^2 \\ &\quad + \left(\frac{1}{4}\right) \left(\frac{5}{4}\right)^2 \\ &= \frac{18}{64} + \frac{1}{64} + \frac{25}{64} = \frac{44}{64} \end{aligned}$$

Rules:

$$p_x(x) \geq 0 \quad \forall x$$

$$\sum_x p_x(x) = 1$$

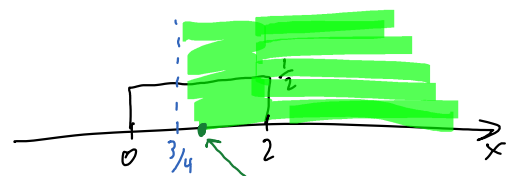
$$p_x(x) = \begin{cases} \frac{1}{2}, & x=0 \\ \frac{1}{4}, & x=1,2 \\ 0, & \text{else} \end{cases}$$

Continuous : Probability Density Function:

$$f_x(a) \quad \text{For } S \subset \mathbb{R}, \quad P(X \in S) = \int_S f_x(a) da$$

Example : Uniform Distribution:

$$\text{Unif}[0,2] = f_x(a) = \begin{cases} \frac{1}{2}, & 0 \leq a \leq 2 \\ 0, & \text{else} \end{cases}$$



Rules:

$$f_x(x) \geq 0 \quad \forall x$$

$$\int_{-\infty}^{\infty} f_x(x) dx = 1$$

Prob  $(X > \frac{3}{4})$

$$EX = \int_{-\infty}^{\infty} f_x(x) \cdot x dx$$

$$= \frac{1}{2} \int_0^2 x dx = \frac{1}{4} x^2 \Big|_0^2 = \frac{4}{4} = 1$$

Expected Value:

Mean of  $g(x)$ :  $EG(X) = \sum_x p_x(x) g(x) \quad \leftarrow \text{Discrete}$

$EG(X) = \int_{-\infty}^{\infty} f_x(x) g(x) dx \quad \leftarrow \text{Continuous}$

$$\text{Var}(X) = \int_{-\infty}^{\infty} f_x(x) (x-1)^2 dx$$

$$= \frac{1}{2} \int_0^2 (x-1)^2 dx$$

$$= \frac{1}{2} \int_{-1}^1 x^2 dx = \frac{1}{6}$$

Unif  $[a, b]$

$$\text{Var} = \frac{(b-a)^2}{12}$$

Mean of  $X$ .  $EX = \sum_x p_x(x) \cdot x$   
 First moment.

Law of Large Numbers: Let  $X_1, X_2, \dots, X_n$  be independent R.V.'s each have dist.  $p_x$

Sample Ave.  $\frac{1}{n} \sum_{i=1}^n g(X_i) \rightarrow EG(X)$

Second Moment:  $EX^2$

LLN:  $\frac{1}{n} \sum_{i=1}^n X_i^2 \rightarrow EX^2 \quad (\text{Power})$

Variance:  $E(X - EX)^2$

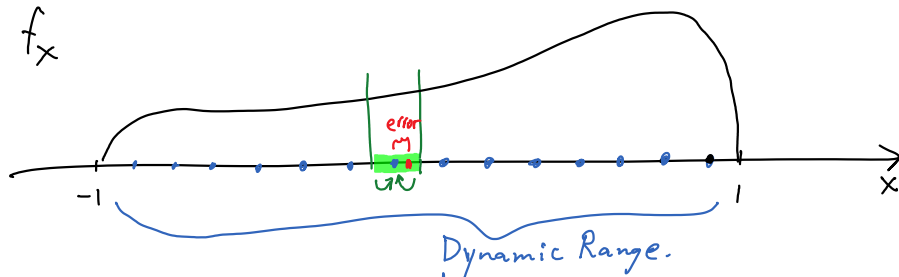
Standard Deviation:  $\sqrt{\text{Var}(X)}$

Variance:  $E (X - EX)$

Standard Deviation:  $\sqrt{\text{Var}(X)}$

Quantization:

Assume a distribution of signal  $x[n]$ .



Uniform Quantization: Equal spacing:

Error is uniformly distributed if quantization is fine enough.

$$|\text{error}| \leq \frac{DR}{2 (\# \text{ Quant. Points})}$$

$$= \frac{DR}{2 \cdot 2^b} \quad \text{where } b \text{ is number of bits for quant.}$$

Power of Quant. Noise:  $\frac{\left(\frac{DR}{2^b}\right)^2}{12}$   
(error)