



EE185524

# Quantization and Sampling Theory

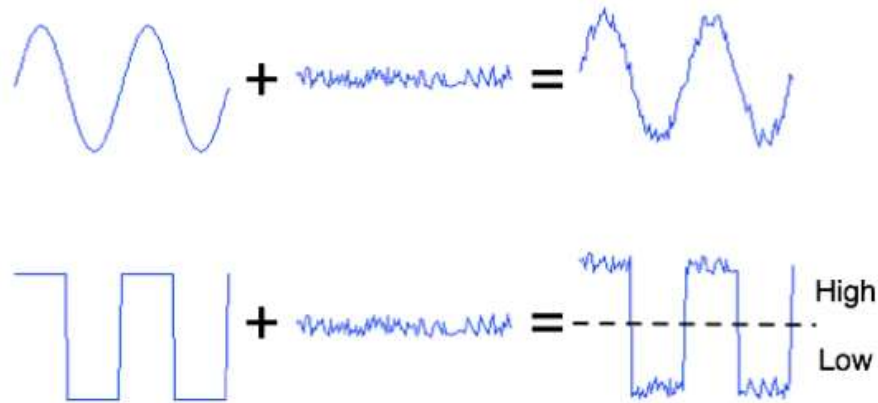
Yurid E. Nugraha  
TSP DTE FTEIC ITS

# Digital vs analog controller

- A digital control system uses digital electronics hardware, usually in the form of a programmed digital computer
- The evolution of microprocessors / embedded systems allowed their use as control elements:
  - met the stringent performance specifications needed in applications, and
  - have several advantages over their continuous-time counterparts:
- Sampling is thus inherent and may be necessary
- For some control system application, better system performance may be achieved by a digital control system design

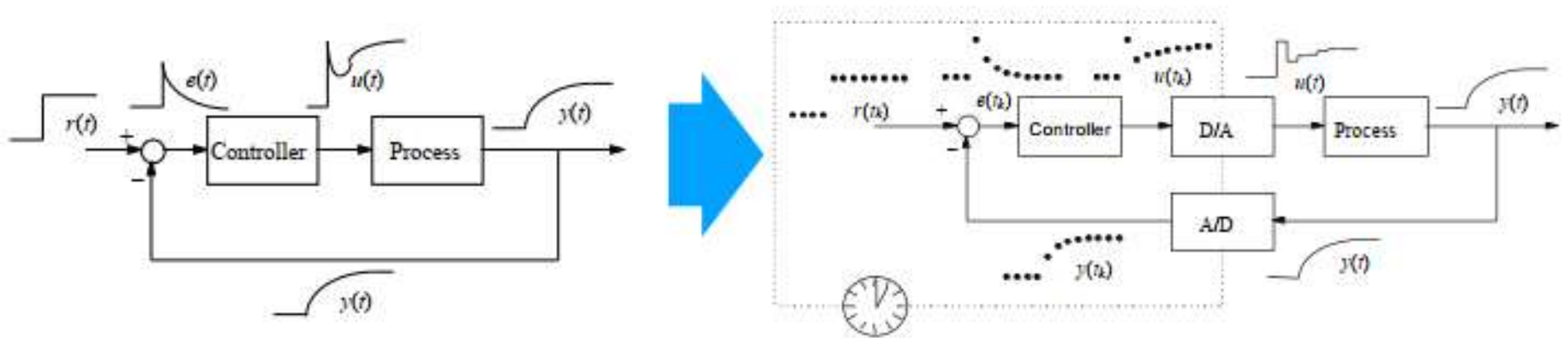
# Digital controller

- More reliable due to its improved noise immunity



- Need sampling to convert time-varying signals to discrete-time signals
- Quantization?

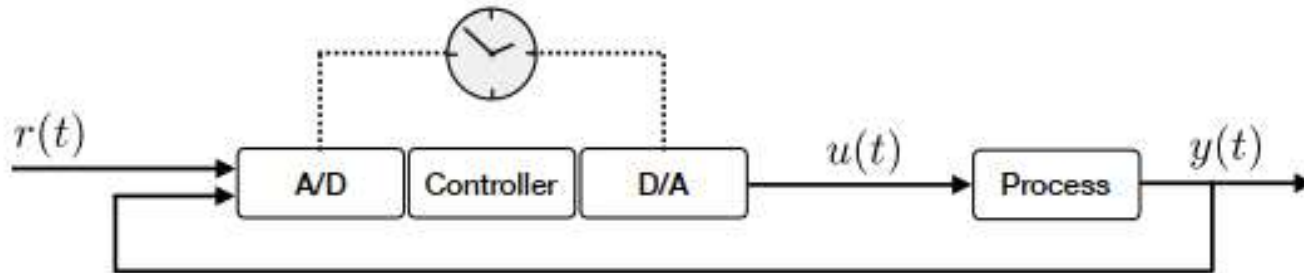
# Analog vs digital signals



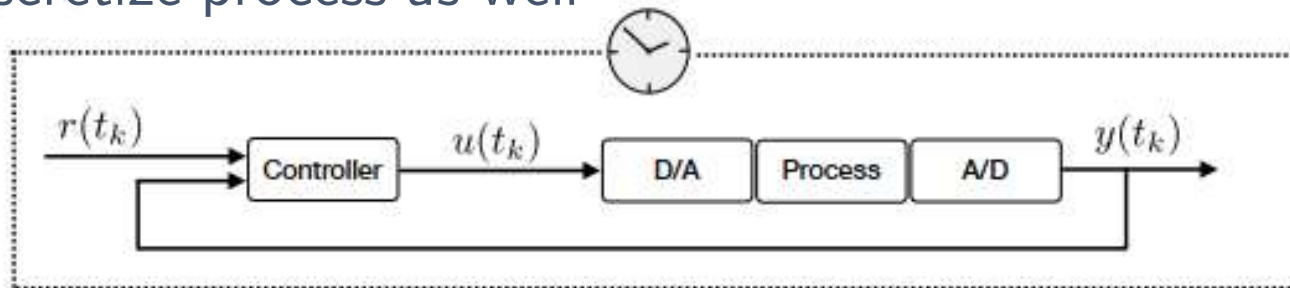
- Reconstruction of analog to digital signal (**and vice versa**) is only an approximation of the actual signal
- Some signal information might be lost or/and delayed in the process

# Digital controller design

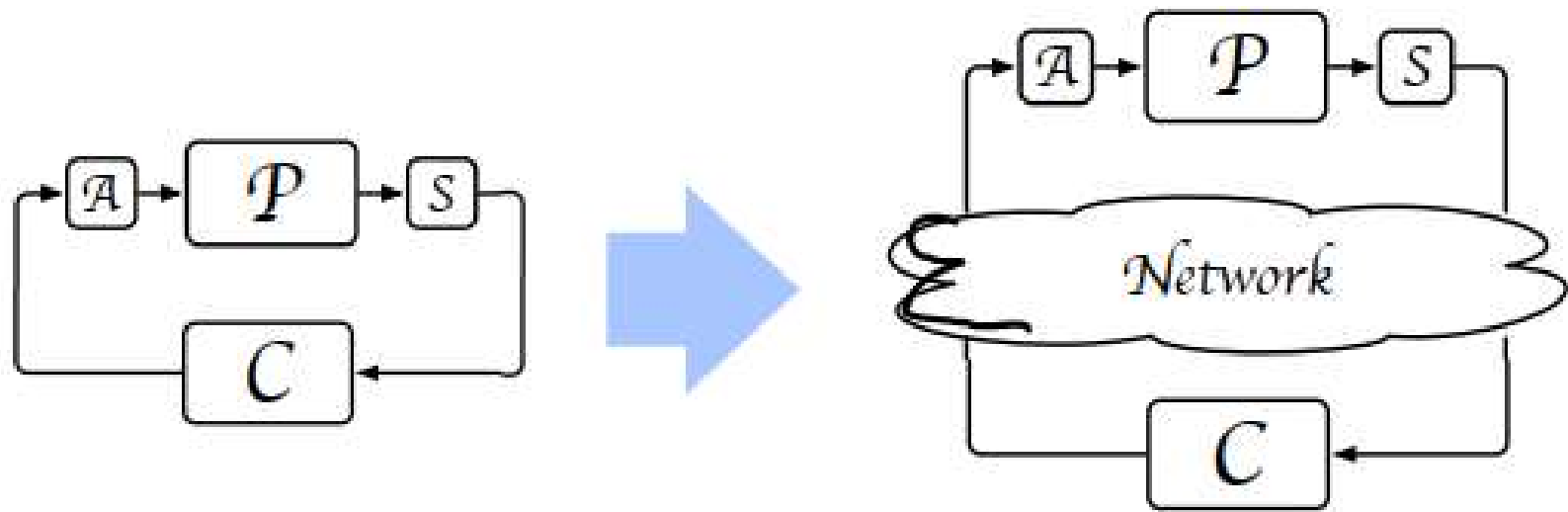
- Discretize analog controller (**more common**)



- Discretize process as well

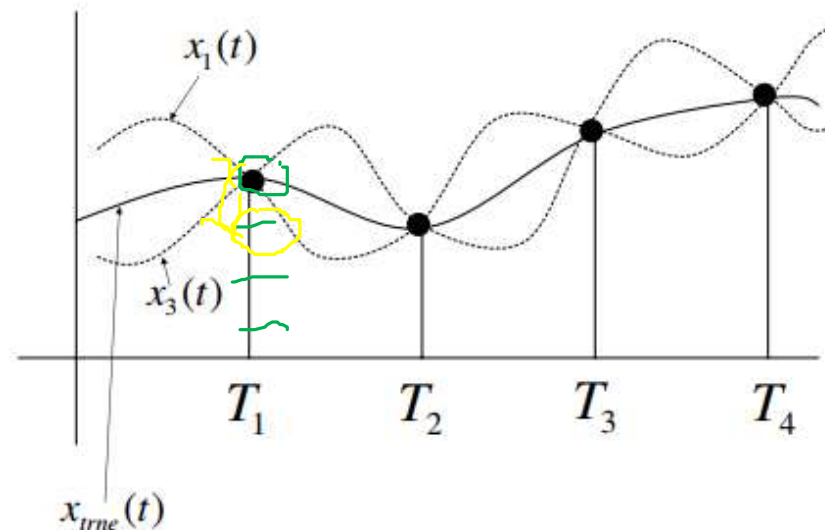


# Digital control in networked control



# Sampling

- (Usually) takes place in regular intervals, say  $T_s$
- Sampling frequency  $f_s = 1/T_s$
- Generally, cannot reconstruct signals fully from samples



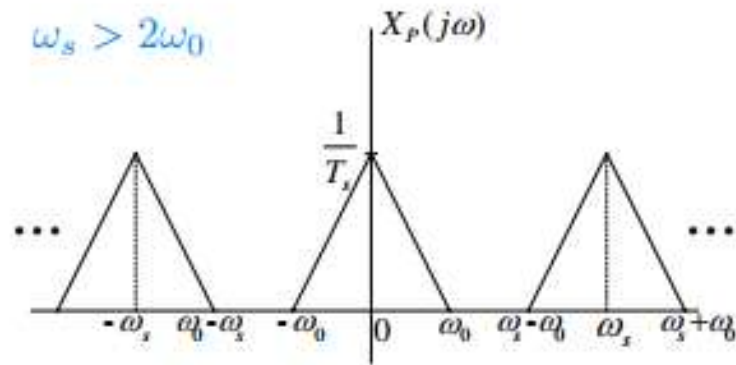


# Sampling criterion

- $x_c(t)$  can be uniquely determined by its samples  $x_c(nT_s)$  if the sampling angular frequency is at least twice as big as  $\omega_0$ :

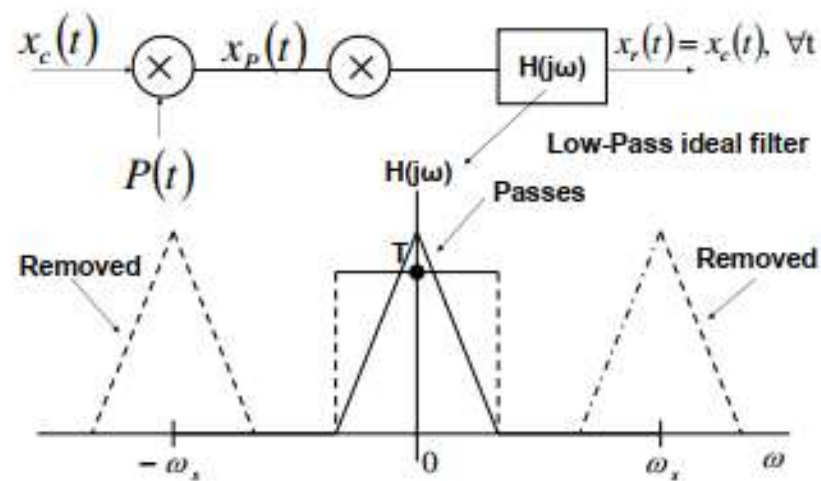
$$\omega_s = \frac{2\pi}{T_s} > 2\omega_0$$

- Nyquist angular frequency: minimum sampling angular frequency



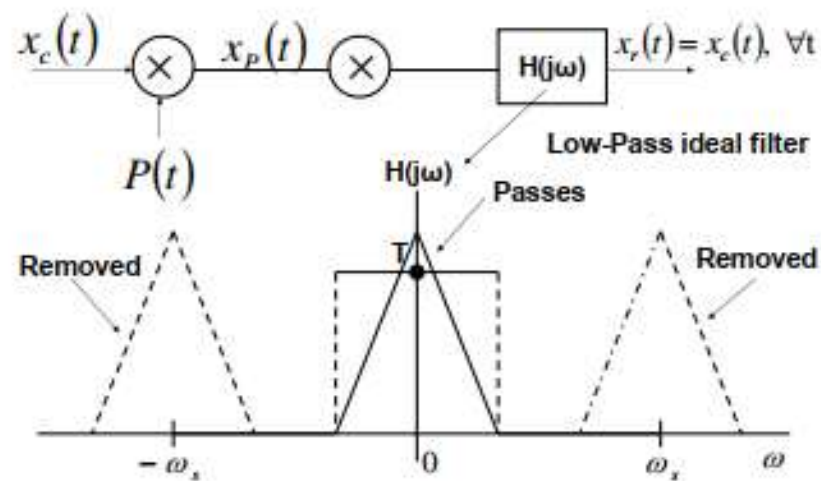
# Reconstruction

- D/A conversion
- Requires a low-pass filter w/ cut-off frequency:  
$$\omega_0 < \omega_c < \omega_s - \omega_0$$



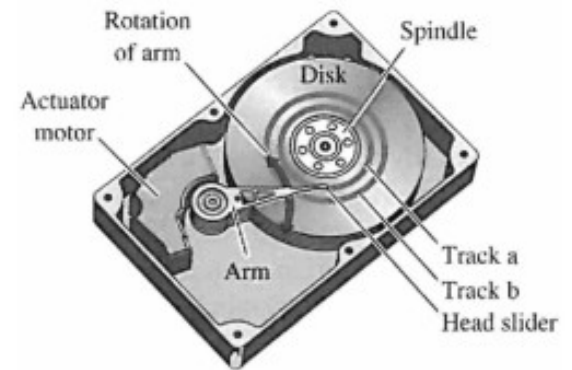
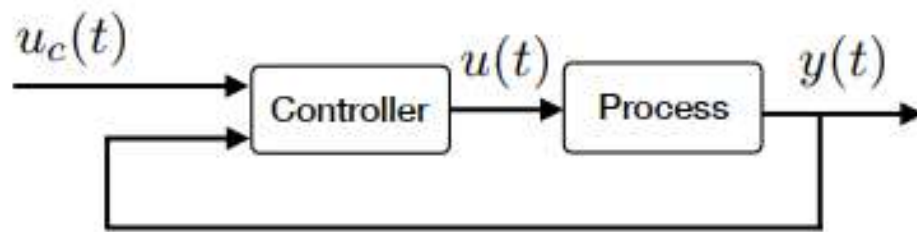
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## Example

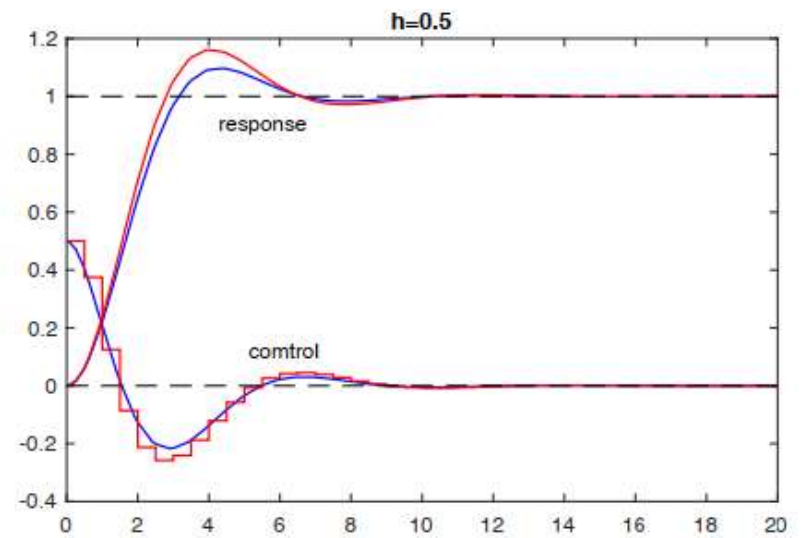
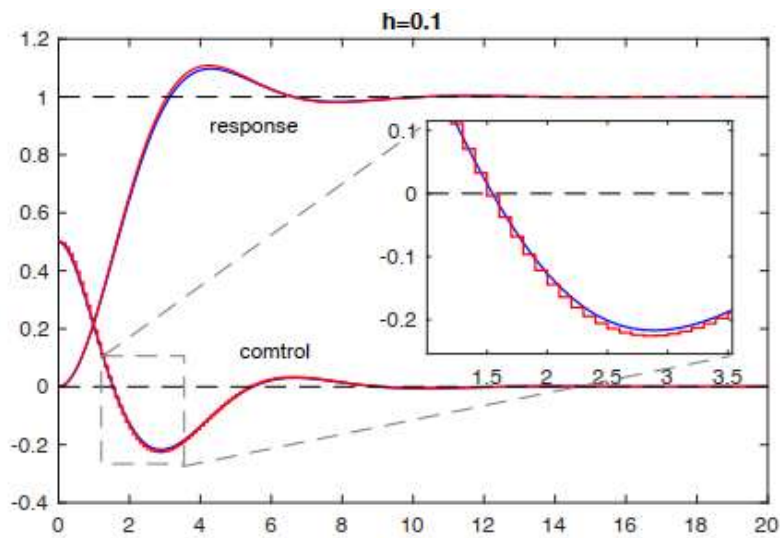
- Consider a plant with TF  $G(s) = \frac{k}{Js^2}$



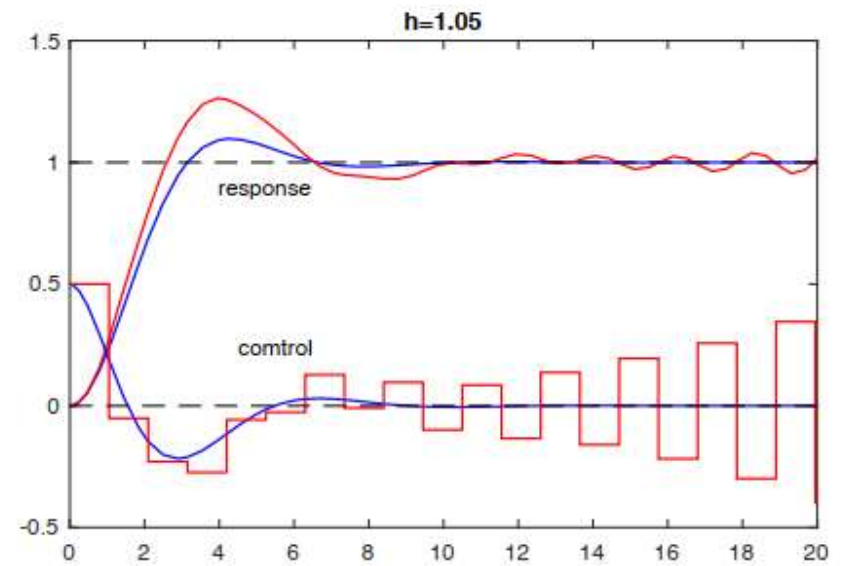
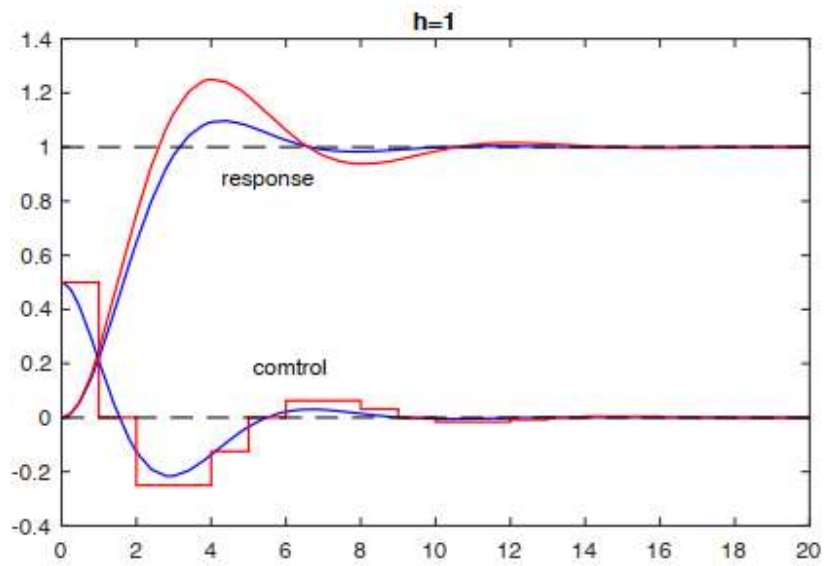
- Suppose controller  $U(s) = K \frac{b}{a} U_c(s) - K \frac{(s+p)}{s+a} Y(s)$
- Derivative is approximated w/ a difference  

$$\frac{x(t+h) - x(t)}{h} = -ax(t) + (a-b)y(t)$$

# Example



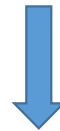
# Example



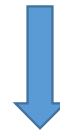
# Quantization

- The process of mapping input values from a large set (often continuous) to output values in a (countable) smaller set

- **Output:** fixed-point words (8-bit, 16-bit, and 24-bit)



256 levels



65536 levels



16.8 million

- An A/D converter produces these binary representation of the **sampled signals** at each sample time

# Fixed-point number representation

- A  $n$ -bit fixed-point binary number  $N$

$$N = \sum_{j=-m}^{n-1} b_j 2^j = \underbrace{b_{n-1} 2^{n-1} + \dots + b_0 2^0}_{\text{Integer portion}} + \underbrace{b_{-1} 2^{-1} + \dots + b_{-m} 2^{-m}}_{\text{Decimal portion}}$$

$$= (b_{n-1} b_{n-2} \dots b_0 \odot b_{-1} \dots b_{-m})_2, \quad b_j \in (0,1)$$

$\downarrow$                        $\downarrow$                        $\downarrow$   
MSB                      Binary point                      LSB



# Quantization error

- Depends on the type of arithmetic and type of quantization used

$$-\frac{q}{2} \leq e \leq \frac{q}{2}$$

where  $q := 2^{-C}$ ,  $C$  being the number of bits

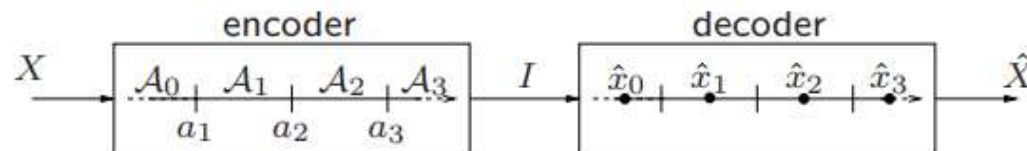
- More level  $\rightarrow$  lower noise
- Typically, original signal is much larger than LSB

# Quantizer



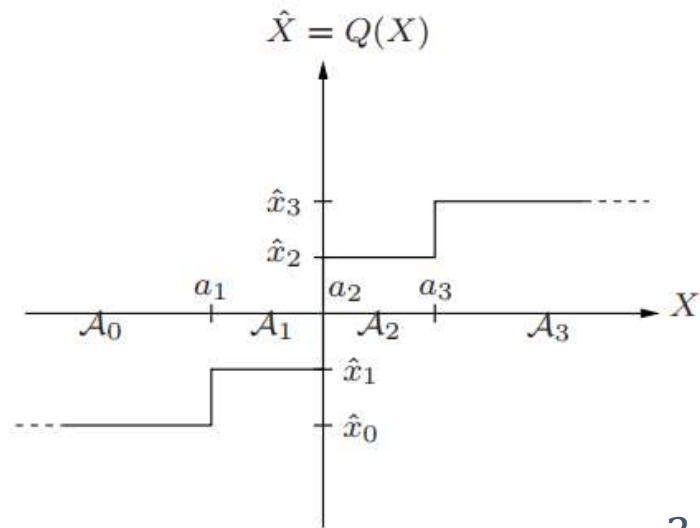
➤ Samples  $\mathbf{X} = X_N^1$  into a  $k$ -bit index, then produces approximation  $\hat{X}_1^N$

➤ **Example: Scalar quantizer**



➤ Encoder:  $X \in A_i \rightarrow I$ , Decoder:  $I \rightarrow \hat{X} = \hat{x}_i$

# Quantizer

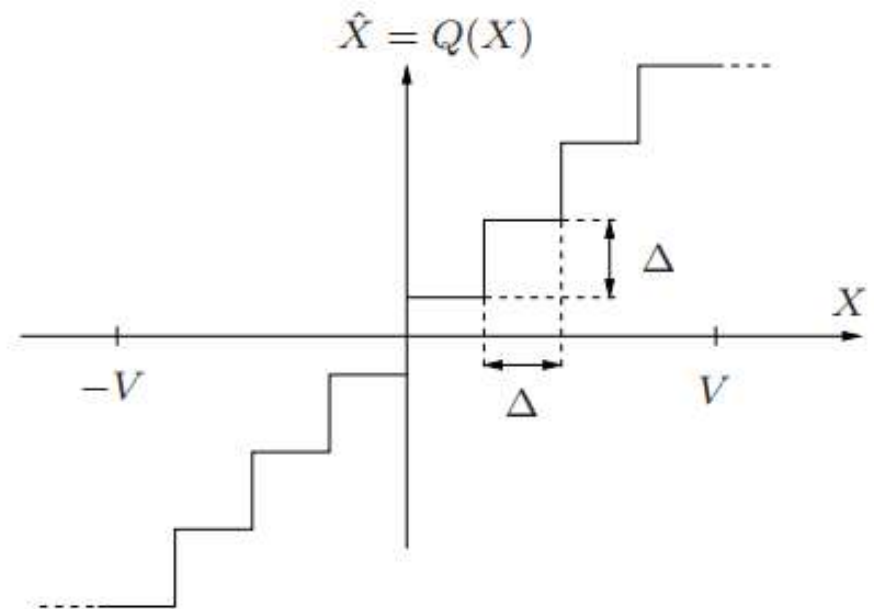


- MSE quantization distortion:  $D = \mathbb{E}[(X - \hat{X})^2]$
- Signal-to-quantization-noise ratio:  $SQNR = \frac{\mathbb{E}[X^2]}{\mathbb{E}[(X - \hat{X})^2]}$

# Uniform quantization

- Step size  $\Delta = \frac{2V}{2^k} = 2^{1-k}V$
- Quantization error  $\tilde{X} = X - Q(X)$ 
  - $X \in [-V, V] \rightarrow$  "granular region"
  - $|X| > V \rightarrow$  "overload"
- Quantization noise:

$$D = \underbrace{\int_{-V}^V (x - Q(x))^2 f(x) dx}_{\text{Granular distortion}} + \underbrace{\int_{|x|>V} (x - Q(x))^2 f(x) dx}_{\text{Overload distortion}}$$



# Quantization for nonuniform distribution

➤ Distortion is given by

$$D = \sum_{i=0}^{K-1} \int_{A_i} (x - x_i)^2 f(x) dx$$

$$= \sum_{i=0}^{K-1} \int_{a_i}^{a_{i+1}} (x - x_i)^2 f(x) dx$$

➤ Optimal encoding:

$$a_i = \frac{\hat{x}_{i-1} + \hat{x}_i}{2} \quad (\text{OE})$$

➤ Optimal decoding:

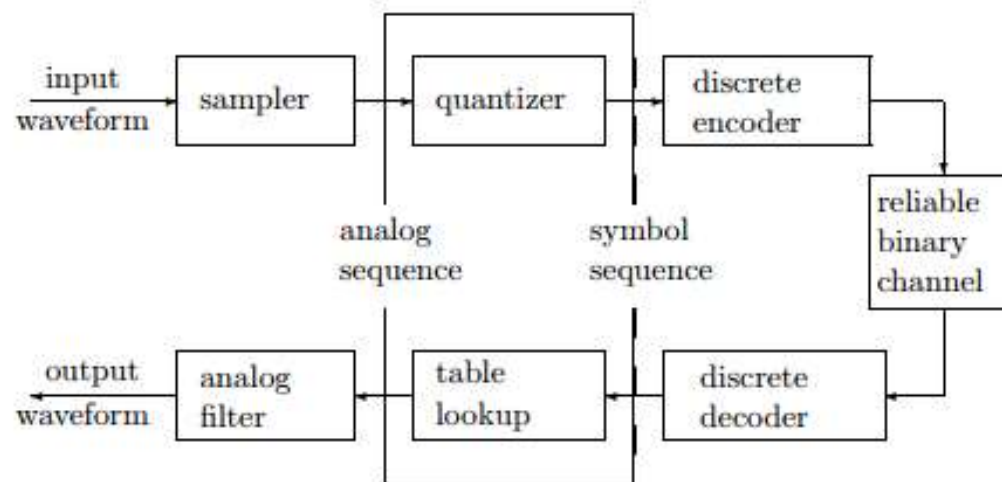
$$\hat{x}_i = \frac{\int_{a_i}^{a_{i+1}} x f(x) dx}{\int_{a_i}^{a_{i+1}} f(x) dx} \quad (\text{OD})$$

(assuming pdf known)

# Lloyd-Max algorithm

- Iteratively computes quantization variables  $\hat{x}_i$  and  $a_i$
- Assumption:  $f(x)$  known,  $a_0 = -V$ ,  $a_k = V$
- Steps:
  - Step 1: Assume a value for  $\hat{x}_0$
  - Step 2: Find  $a_1$  from (OE)
  - Step 3: Find  $\hat{x}_2$  from (OD)
  - ... etc.

# Quantization in communication systems



# Sustained oscillations and deadband effects

- When digital controllers are implemented with finite word length, **sustained oscillations** may appear at the controller output
- Consider the controller described by difference equation

$$y[k] = ay[k-1] + x[k]$$

where  $a = 0.5$ ,  $x[k] = 0.75\delta[k]$ ,  $y[-1] = 0$

$$\begin{aligned} k=0 &\rightarrow 0.75 \\ k=1 &\rightarrow 0.75 \times 0.5 \end{aligned}$$

- If the controller equation implemented with infinite word, then

$$y[k] = 0.75(0.5)^k$$

100



## Sustained oscillations and deadband effects

- If the controller equation implemented with *3-bit word*, then

$$y_q[k] = Q \left[ 0.5y_q[k-1] \right] + 0.75\delta[k]$$

- ..... stops at  $y_q[k] = \underline{0.125}$

# Interplay between sampling and quantization error

Quantization error may not be ignored

- For example, consider a controller w/ TF  $G(s) = \frac{10^4}{s+1}$
- Discretization w/ **impulse invariant approximation**.

$$G(z) = \frac{10^4}{1 - e^{-h}z^{-1}}$$

- **Unit impulse response:**  $g[m] = 10^4(e^{-h})^m$ ,  $m = 0, 1, 2, \dots$
- Thus,

$$\sum_{m=0}^{\infty} g^2[m] = 10^8 \sum_{m=0}^{\infty} e^{-2hm} = 10^8(1 + e^{-2h} + \dots) = \frac{10^8}{1 - e^{-2h}}$$

# Interplay between sampling and quantization error

Quantization error may not be ignored

➤ Variance of output:

$$\text{var}(y_e) = \text{var}(e) \sum_{m=0}^{\infty} g^2[m] = \left( \frac{2^{-2C}}{12} \right) \left( \frac{10^8}{1 - e^{-2}} \right)$$

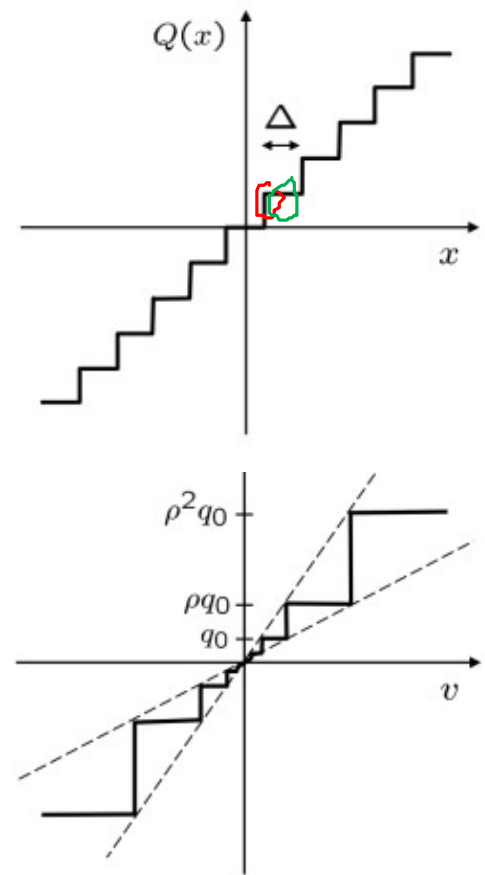
➤ If  $C$  fixed: decreasing  $h$  **increases** variance of output noise

➤ If  $h$  fixed: increasing  $C$  **decreases** variance of output noise

# Which quantizers?


Since a finite bits are transmitted, quantizers should be **designed**

- **Uniform** quantizers: divide space into equal sections
- **Logarithmic** quantizers: provide a finer quantization near the origin
- **Dynamic scaling**: a smaller scaling factor provides a fine quantization near origin; a larger one ensures large number fall within domain of quantization




Into control system...

$$\begin{aligned}x[k + 1] &= Ax[k] + Bu[k] + w[k] \\y[k] &= Cx[k] + v[k]\end{aligned}$$

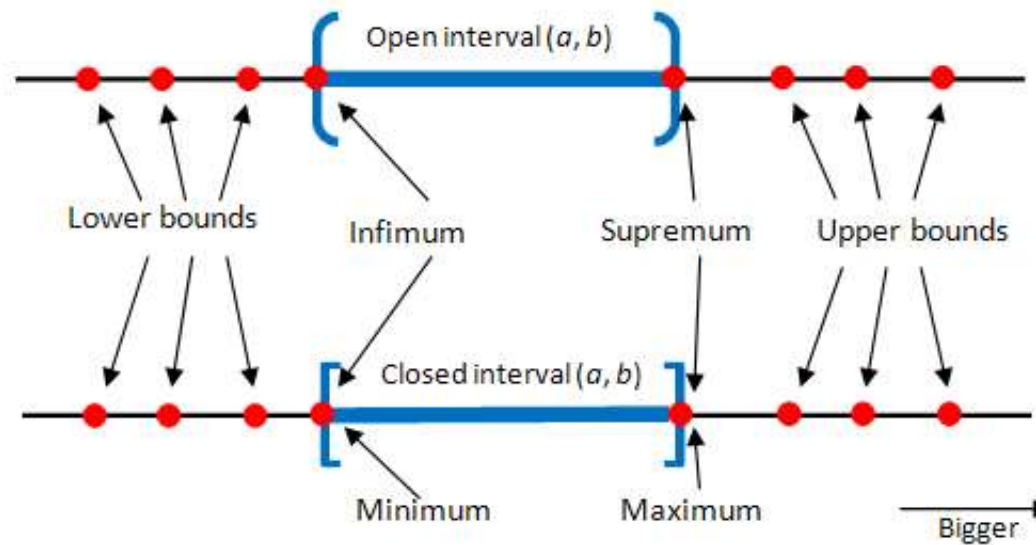


- **Objective:** Identify the trade-off between the unstable modes of the system and the channel's rate to guarantee stability
- **Solution:** Consider second moment stability:

$$\sup_{k \in \mathbb{N}} \mathbb{E}(\|X_k\|^2) < \infty$$



Into control system...



## Into control system...

Remote state estimation:

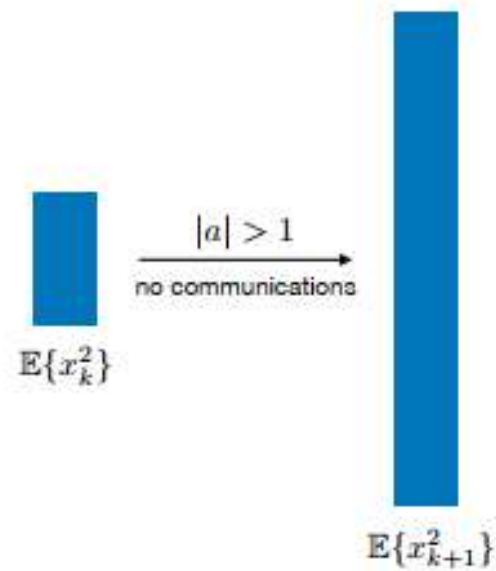


$$x[k + 1] = ax[k] + w[k]$$

$w[k]$  Gaussian w/ zero mean and variance  $\sigma_w^2$ :  
$$\mathbb{E}\{x_{k+1}^2\} = a^2 \mathbb{E}\{x_k^2\} + \sigma_w^2$$

Unstable if  $|a| > 1$

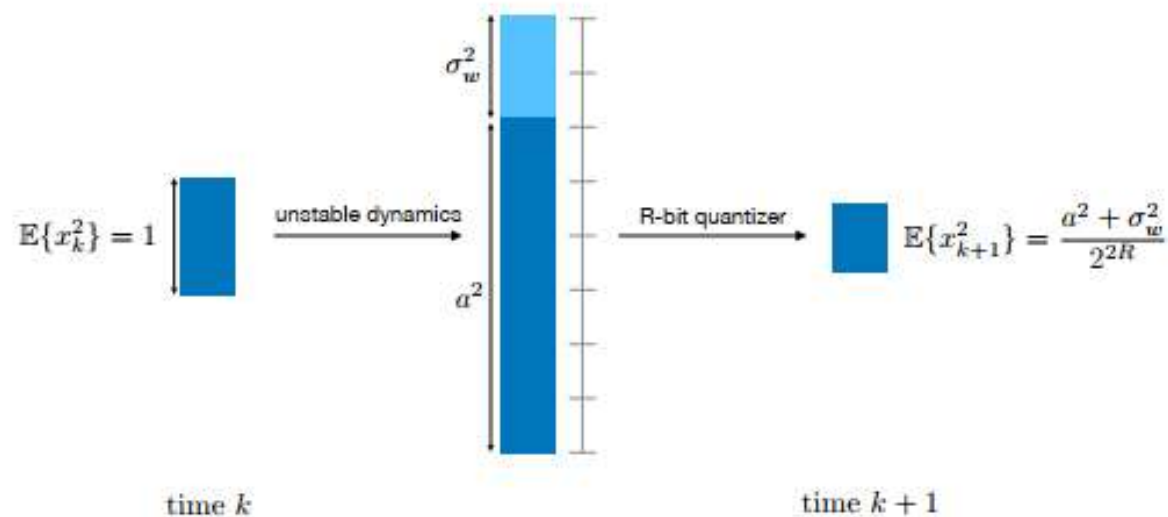
# Data rate theorem





# Data rate theorem

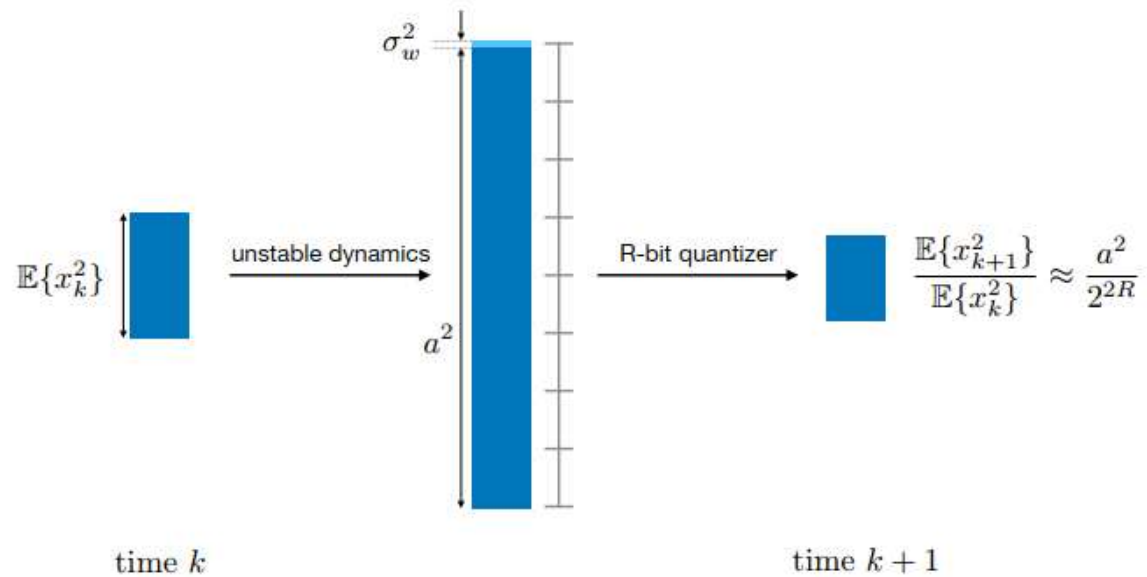
- The noise and bandwidth limitations of the channels are captured by modeling channels capable of transmitting only  $R$  bits in each time slot
- By transmitting enough bits at each time step, we can ensure the uncertainty decreases



# Data rate theorem

- $\mathbb{E}\{x_k^2\}$  grows larger each time
- Thus, (second moment) stability can be achieved if

$$\frac{a^2}{2^{2R}} < 1$$



# Conclusion

- Needs to carefully choose the sampling for stability and performance
- The finer the quantization, the better
- How much information is needed to be communicated by the quantizer in order to achieve a certain control objective?