



# DIGITAL COMMUNICATION

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# BASEBAND SHAPING

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## Background on Random Processes Power Spectra of Discrete PAM Signals

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# RANDOM PROCESSES

## Definition and Basics

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- Recall that a random variable  $X: \Omega \rightarrow \mathbb{R}$  such that some constraints are satisfied
- A random process  $X(t, \omega)$  is an indexed set of random variables
- At a high level, it is a collection of continuous-time or discrete-time signals, for every  $\omega$
- For a given  $t$ ,  $X(t, \omega)$  is a random variable
  
- Examples: Stock market index, rainfall, weather forecasting
- Examples: Message signal in communications, receiver noise, received signal etc.
  
- A random process is said to be “deterministic”, if future values can be predicted from past samples. Example:  $X(t) = A \cos(\omega t + \phi)$ , where  $\phi \sim U(0, 2\pi)$
- A “non-deterministic” random process is not deterministic

# RANDOM PROCESSES

## Statistical Parameters



- Recall that for a given  $t$ ,  $X(t, \omega)$  is a random variable
- Therefore, moments on  $X(t)$  can be defined for a given value of  $t$
- Let  $\mu_X(t) = \mathbb{E}(X(t))$  -- This is the ensemble average
- Let  $\sigma_X^2(t) = \mathbb{E}([X(t) - \mu_X(t)]^2)$  -- These two constitute the first order statistics
- Let  $m_X^2(t) = \mathbb{E}(X^2(t))$  -- This is the mean squared value
  
- Consider two time instances  $t_1$  and  $t_2$ . Then for all  $t_1, t_2 \in \mathbb{R}$
- $R_{XX}(t_1, t_2) = \mathbb{E}(X(t_1) X(t_2))$  -- This is called as the autocorrelation function
- $C_{XX}(t_1, t_2) = \mathbb{E}([X(t_1) - \mu_X(t_1)][X(t_2) - \mu_X(t_2)])$  -- Autocovariance function
- Note that  $R_{XX}(t_1, t_1) = m_X^2(t_1)$
- Note that  $C_{XX}(t_1, t_1) = \sigma_X^2(t_1)$

- Recall that for a given  $t$ ,  $X(t, \omega)$  is a random variable
- A random process  $X(t, \omega)$  is said to be first order stationary, if its PDF at any given time instances  $t$  and  $t + \Delta t$  are equal, i.e.,  $f_X(x; t) = f_X(x; t + \Delta t)$
- A random process  $X(t, \omega)$  is said to be stationary in strict sense if
$$f(x_1, \dots, x_n; t_1, \dots, t_n) = f(x_1, \dots, x_n; t_1 + \Delta t, \dots, t_n + \Delta t), \text{ for all } t_1, \dots, t_n, \Delta t \in \mathbb{R}, n \in \mathbb{N}$$
- A random process  $X(t, \omega)$  is said to be stationary in wide sense (WSS) if
  - $\mu_X(t) = \mu$ , for all  $t \in \mathbb{R}$
  - $R_{XX}(t_1, t_2) = R_{XX}(t_2 - t_1) = R_{XX}(\tau)$ , for all  $t_1, t_2 \in \mathbb{R}$
- For continuous-time process,  $R_{XX}(\tau) = R_{XX}(t, t + \tau)$
- For discrete-time process,  $R_{XX}(k) = R_{XX}(n, n + k)$
- The PSD of a WSS process is the FT of its ACF:  $S_X(f) = FT\{R_{XX}(\tau)\}$

# POWER SPECTRAL DENSITY (PSD)

## Power Spectrum of Discrete PAM Signals

- Let  $v(t)$  denote a basic pulse of duration  $T_b$  seconds
- The discrete PAM signal is a random process, defined as

$$X(t) = \sum_{k=-\infty}^{\infty} A_k v(t - kT_b)$$

- Here,  $A_k$  is a discrete random variable, whose value depends on the  $k^{th}$  bit and the chosen format (such as unipolar, polar, bipolar or Manchester)

$$\text{Unipolar : } A_k = \begin{cases} a & \text{symbol 1} \\ 0 & \text{symbol 0} \end{cases}$$

$$\text{Polar/Manchester : } A_k = \begin{cases} a & \text{symbol 1} \\ -a & \text{symbol 0} \end{cases}$$

$$\text{Bipolar : } A_k = \begin{cases} a, -a & \text{alternating 1s} \\ 0 & \text{symbol 0} \end{cases}$$

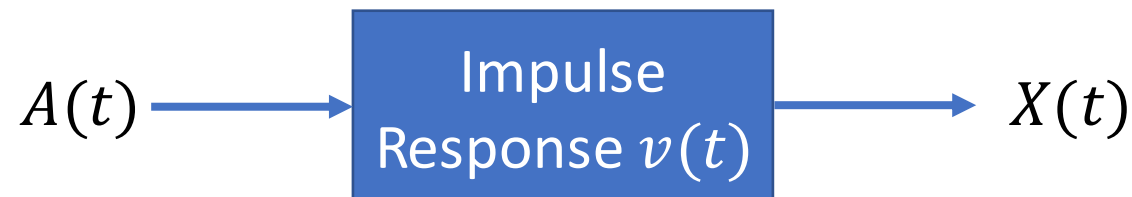
# POWER SPECTRAL DENSITY (PSD)

## Power Spectrum of Discrete PAM Signals

- We can rewrite  $X(t)$  as follows

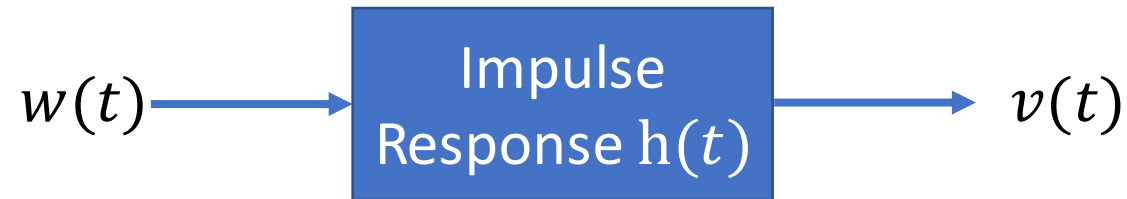
$$\begin{aligned} X(t) &= \sum_{k=-\infty}^{\infty} A_k \{ \delta(t - kT_b) * v(t) \} \\ &= \left\{ \sum_{k=-\infty}^{\infty} A_k \delta(t - kT_b) \right\} * v(t) \\ &= A(t) * v(t) \end{aligned}$$

- Therefore, the generation of  $X(t)$  can be viewed as an output of an LTI system, with input  $A(t)$  and impulse response  $v(t)$



# POWER SPECTRAL DENSITY (PSD)

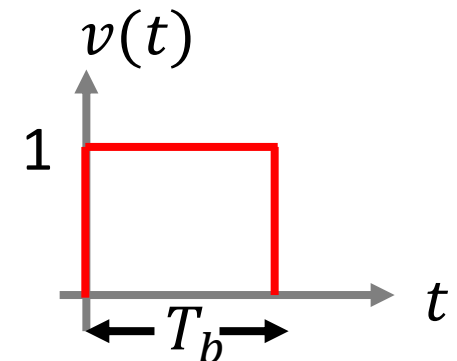
## Power Spectrum of Discrete PAM Signals



- Assume that  $v(t)$  is an output of an LTI system with response  $h(t)$  and input  $w(t)$
- When  $w(t)$  is a WSS process,  $v(t)$  is also WSS and  $S_v(f) = |H(f)|^2 S_w(f)$
- A discrete PAM signal is a cyclo-stationary process (periodic and stationary). Hence,

$$S_X(f) = \frac{|V(f)|^2}{T_b} S_A(f)$$

- Therefore, to evaluate  $S_X(f)$ , we need  $|V(f)|^2$
- Assume that  $v(t)$  is a rectangular function with unit amplitude





# POWER SPECTRAL DENSITY (PSD)

## Power Spectrum of Discrete PAM Signals



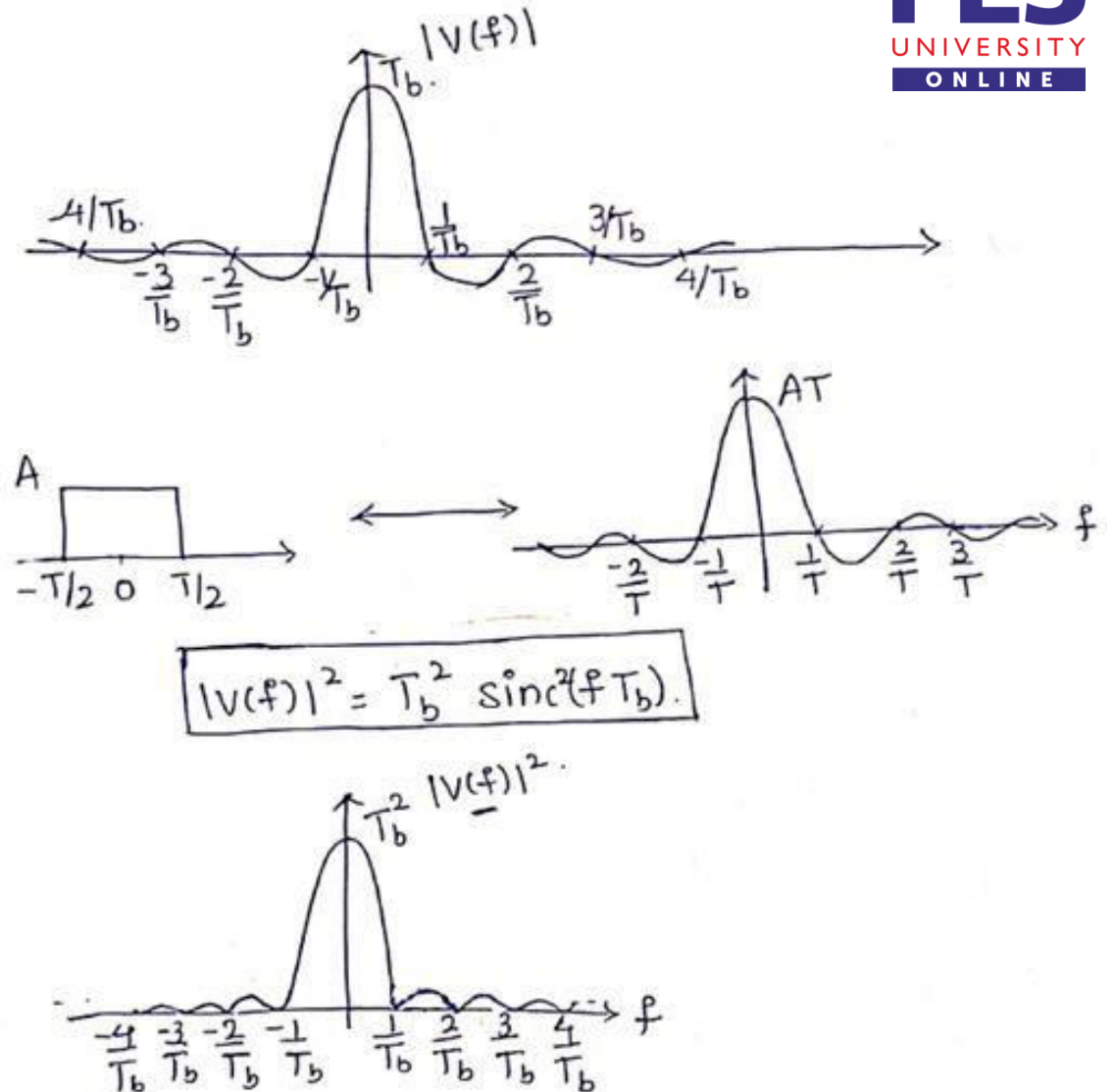
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$$\begin{aligned} V(f) &= \int_0^{T_b} v(t) e^{-j2\pi ft} dt \\ &= \frac{-1}{j2\pi f} e^{-j2\pi ft} \Big|_0^{T_b} = \frac{1}{j2\pi f} [1 - e^{-j2\pi f T_b}] \\ &= \frac{e^{-j\pi f T_b}}{\pi f} \left[ \frac{e^{j\pi f T_b} - e^{-j\pi f T_b}}{2j} \right] \\ &= e^{-j\pi f T_b} \frac{\sin(\pi f T_b)}{\pi f} \end{aligned}$$

$$\therefore |V(f)| = \frac{\sin(\pi f T_b)}{\pi f} = \frac{T_b \sin(\pi f T_b)}{\pi f T_b}$$

$$\therefore |V(f)| = T_b \text{sinc}(f T_b)$$

$$\boxed{\therefore |V(f)|^2 = T_b^2 \text{sinc}^2(f T_b)}$$



# POWER SPECTRAL DENSITY (PSD)

## Power Spectrum of Discrete PAM Signals

- Now that we have evaluated  $V(f)$ , we next need to evaluate  $S_A(f)$
- Recall that the power spectrum is the Fourier transform of the autocorrelation
- A sequence of samples  $x_k$  can be represented either as a discrete-time sequence  $x(n)$  or a continuous-time signal  $x(t)$

$$x(n) = \sum_{k=-\infty}^{\infty} x_k \delta(n - k)$$

$$x(t) = \sum_{k=-\infty}^{\infty} x_k \delta(t - kT)$$

- Now, in the Fourier representation

$$x(n) \xleftrightarrow{\text{F.T.}} X(e^{j\omega}) = \sum_{k=-\infty}^{\infty} x_k e^{-j\omega k}$$

$$x(t) \xleftrightarrow{\text{F.T.}} X(f) = \int_{t=-\infty}^{\infty} x(t) e^{-j2\pi ft} dt$$

$$\begin{aligned} X(f) &= \int_{t=-\infty}^{\infty} \sum_k x_k \delta(t - kT) e^{-j2\pi ft} dt \\ &= \sum_k x_k \left\{ \int_{t=-\infty}^{\infty} \delta(t - kT) e^{-j2\pi ft} dt \right\} \end{aligned}$$

# POWER SPECTRAL DENSITY (PSD)

## Power Spectrum of Discrete PAM Signals

- Recall that the power spectrum is the Fourier transform of the autocorrelation
- Therefore for us, the sequence of interest  $x_k = R_A(k)$
- From the previously shown development, we can write

$$\begin{aligned} X(f) &= \int_{t=-\infty}^{\infty} \sum_k x_k \delta(t - KT) e^{-j2\pi ft} dt \\ &= \sum_k x_k \left\{ \int_{t=-\infty}^{\infty} \delta(t - KT) e^{-j2\pi ft} dt \right\} \end{aligned}$$

$$X(f) = \sum_k x_k e^{-j2\pi fKT}$$

$$S_A(f) = \sum_{n=-\infty}^{\infty} R_A(n) e^{-j2\pi fnT_b}$$

where  $R_A(n) = E[A_k A_{k-n}]$

# POWER SPECTRUM OF PAM SIGNALS

## Power Spectrum of Discrete PAM Signals

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- Next, we intend to calculate the power spectra for the following cases
  - Unipolar NRZ
  - Bipolar NRZ
  - Polar NRZ
  - Manchester coding
- In each case, we first evaluate  $S_A(f)$  from  $R_A(n)$  and then evaluate  $S_X(f)$  using

$$S_X(f) = \frac{|V(f)|^2}{T_b} S_A(f)$$

- Note that we have shown

$$|V(f)|^2 = T_b^2 \text{sinc}^2(fT_b)$$



# THANK YOU

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