

MIT OpenCourseWare
<http://ocw.mit.edu>

16.36 Communication Systems Engineering
Spring 2009

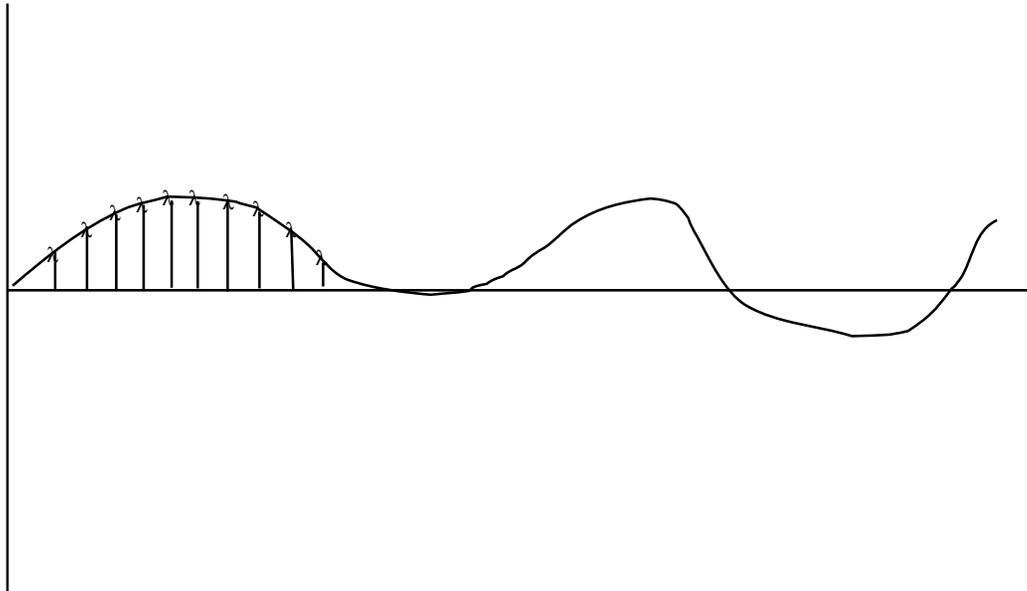
For information about citing these materials or our Terms of Use, visit: <http://ocw.mit.edu/terms>.

Lecture 2: The Sampling Theorem

Eytan Modiano

Sampling

- **Given a continuous time waveform, can we represent it using discrete samples?**
 - How often should we sample?
 - Can we reproduce the original waveform?



The Fourier Transform

- Frequency representation of signals

- **Definition:**
$$X(f) = \int_{-\infty}^{\infty} x(t) e^{-j2\pi ft} dt$$

$$x(t) = \int_{-\infty}^{\infty} X(f) e^{j2\pi ft} df$$

- **Notation:**

$$\mathbf{X(f) = F[x(t)]}$$

$$\mathbf{X(t) = F^{-1} [X(f)]}$$

$$\mathbf{x(t) \leftrightarrow X(f)}$$

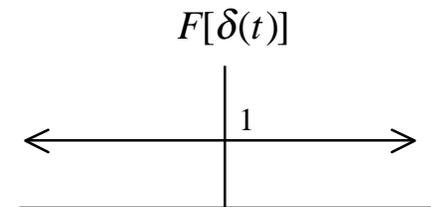
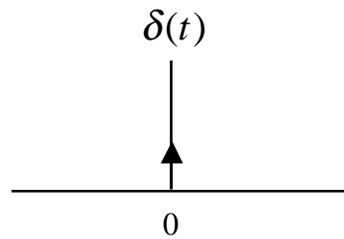
Unit impulse $\delta(t)$

$$\delta(t) = 0, \forall t \neq 0$$

$$\int_{-\infty}^{\infty} \delta(t) dt = 1$$

$$\int_{-\infty}^{\infty} \delta(t) x(t) dt = x(0)$$

$$\int_{-\infty}^{\infty} \delta(t - \tau) x(\tau) dt = x(\tau)$$



$$F[\delta(t)] = \int_{-\infty}^{\infty} \delta(t) e^{-j2\pi ft} dt = e^0 = 1$$

$$\delta(t) \leftrightarrow 1$$

Properties of the Fourier transform

- **Linearity**
 - $x_1(t) \Leftrightarrow X_1(f), x_2(t) \Leftrightarrow X_2(f) \Rightarrow \alpha x_1(t) + \beta x_2(t) \Leftrightarrow \alpha X_1(f) + \beta X_2(f)$
- **Duality**
 - $X(f) \Leftrightarrow x(t) \Rightarrow x(f) \Leftrightarrow X(-t)$ and $x(-f) \Leftrightarrow X(t)$
- **Time-shifting:** $x(t-\tau) \Leftrightarrow X(f)e^{-j2\pi f\tau}$
- **Scaling:** $F[(x(at))] = 1/|a| X(f/a)$
- **Convolution:** $x(t) \Leftrightarrow X(f), y(t) \Leftrightarrow Y(f)$ then,
 - $F[x(t)*y(t)] = X(f)Y(f)$
 - Convolution in time corresponds to multiplication in frequency and vice versa

$$x(t) * y(t) = \int_{-\infty}^{\infty} x(t - \tau) y(\tau) d\tau$$

Fourier transform properties (Modulation)

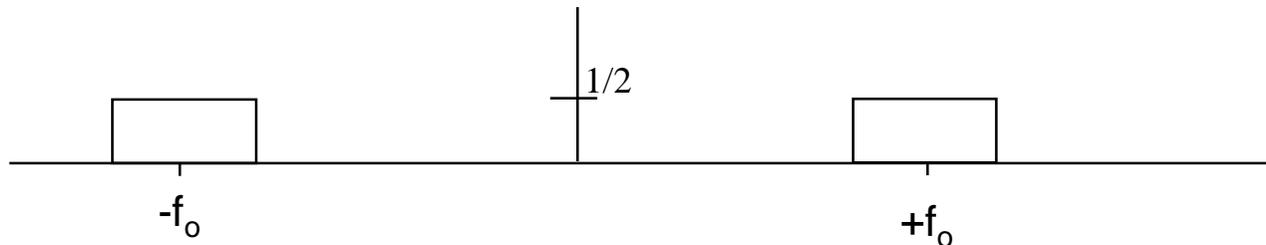
$$x(t)e^{j2\pi f_0 t} \Leftrightarrow X(f - f_0)$$

$$\text{Now, } \cos(x) = \frac{e^{jx} + e^{-jx}}{2}$$

$$x(t) \cos(2\pi f_0 t) = \frac{x(t)e^{j2\pi f_0 t} + x(t)e^{-j2\pi f_0 t}}{2}$$

$$\text{Hence, } x(t) \cos(2\pi f_0 t) \Leftrightarrow \frac{X(f - f_0) + X(f + f_0)}{2}$$

- **Example: $x(t) = \text{sinc}(t)$, $F[\text{sinc}(t)] = \Pi(f)$**
- **$Y(t) = \text{sinc}(t)\cos(2\pi f_0 t) \Leftrightarrow (\Pi(f - f_0) + \Pi(f + f_0))/2$**



More properties

- **Power content of signal** $\int_{-\infty}^{\infty} |x(t)|^2 dt = \int_{-\infty}^{\infty} |X(f)|^2 df$

- **Autocorrelation** $R_x(\tau) = \int_{-\infty}^{\infty} x(t)x^*(t-\tau)dt$

$$R_x(\tau) \Leftrightarrow |X(f)|^2$$

- **Sampling** $x(t_o) = x(t)\delta(t-t_o)$

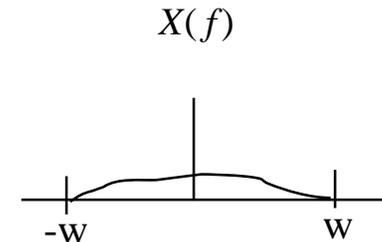
$$x(t) \sum_{n=-\infty}^{\infty} \delta(t-nt_o) = \text{sampled version of } x(t)$$

$$F\left[\sum_{n=-\infty}^{\infty} \delta(t-nt_o)\right] = \frac{1}{t_o} \sum_{n=-\infty}^{\infty} \delta\left(f - \frac{n}{t_o}\right)$$

The Sampling Theorem

- **Band-limited signal**
 - **Bandwidth < W**

$$X(f) = 0, \text{ for all } f, |f| \geq W$$



Sampling Theorem: If we sample the signal at intervals T_s where $T_s \leq 1/2W$ then signal can be completely reconstructed from its samples using the formula

$$x(t) = \sum_{n=-\infty}^{\infty} 2W' T_s x(nT_s) \text{sinc}[2W'(t - nT_s)]$$

Where, $W \leq W' \leq \frac{1}{T_s} - W$

With $T_s = \frac{1}{2W} \Rightarrow x(t) = \sum_{n=-\infty}^{\infty} x(nT_s) \text{sinc}[(\frac{t}{T_s} - n)]$

$$x(t) = \sum_{n=-\infty}^{\infty} x(\frac{n}{2W}) \text{sinc}[2W(t - \frac{n}{2W})]$$

Proof

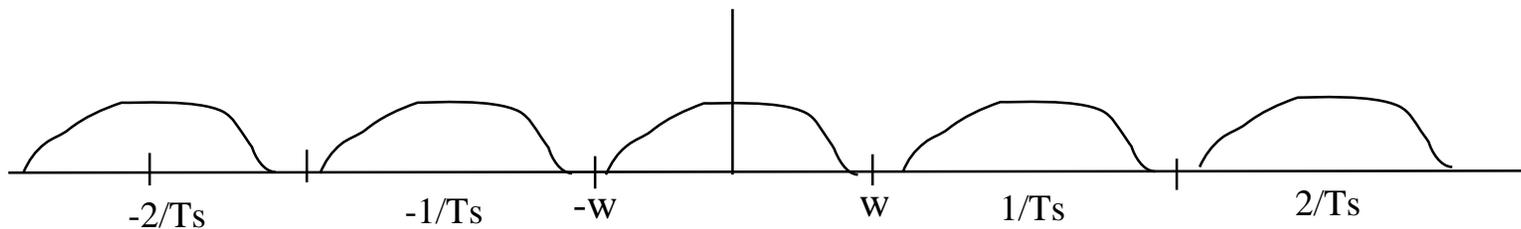
$$x_{\delta}(t) = x(t) \sum_{n=-\infty}^{\infty} \delta(t - nT_s)$$

$$X_{\delta}(f) = X(f) * F\left[\sum_{n=-\infty}^{\infty} \delta(t - nT_s)\right]$$

$$F\left[\sum_{n=-\infty}^{\infty} \delta(t - nT_s)\right] = \frac{1}{T_s} \sum_{n=-\infty}^{\infty} \delta\left(f - \frac{n}{T_s}\right)$$

$$X_{\delta}(f) = \frac{1}{T_s} \sum_{n=-\infty}^{\infty} X\left(f - \frac{n}{T_s}\right)$$

- **The Fourier transform of the sampled signal is a replication of the Fourier transform of the original separated by $1/T_s$ intervals**



Proof, continued

- If $1/T_s > 2W$ then the replicas of $X(f)$ will not overlap and can be recovered
- How can we reconstruct the original signal?
 - Low pass filter the sampled signal
- Ideal low pass filter is rectangular
 - Its impulse response is a sinc function
- Now the recovered signal after low pass filtering

$$H(f) = T_s \Pi\left(\frac{f}{2W}\right)$$

$$X(f) = X_\delta(f) T_s \Pi\left(\frac{f}{2W}\right)$$

$$x(t) = F^{-1}\left[X_\delta(f) T_s \Pi\left(\frac{f}{2W}\right)\right]$$

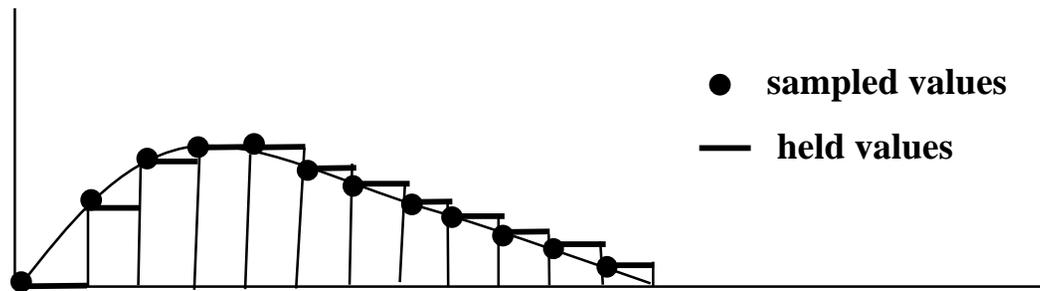
$$x(t) = \sum_{n=-\infty}^{\infty} x(nT_s) \text{Sinc}\left(\frac{t}{T_s} - n\right)$$

Notes about Sampling Theorem

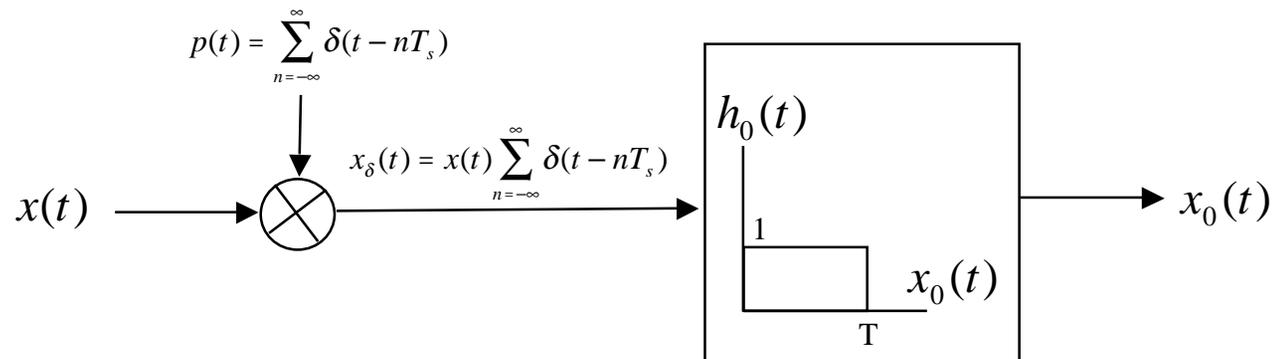
- **When sampling at rate $2W$ the reconstruction filter must be a rectangular pulse**
 - Such a filter is not realizable
 - For perfect reconstruction must look at samples in the infinite future and past
- **In practice we can sample at a rate somewhat greater than $2W$ which makes reconstruction filters that are easier to realize**
- **Given any set of arbitrary sample points that are $1/2W$ apart, can construct a continuous time signal band-limited to W**
- **Sampling using “impulses” is also not practical**
 - Narrow pulses are difficult to implement
 - In practice, sampling is done using small rectangular pulses or “zero-order-hold”

Zero-Order Hold

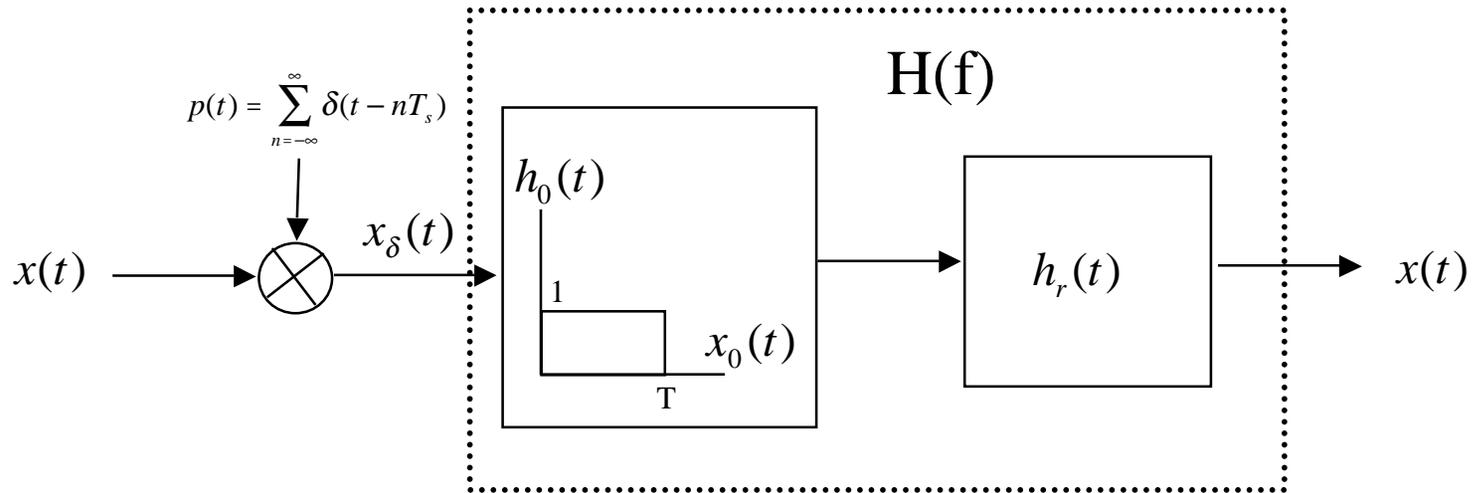
- A form of “interpolation”
- The sampled signal holds its value until the next sample time



- In principle, zero-order hold can be realized with a cascade of an impulse train sampling and an LTI system with rectangular impulse response



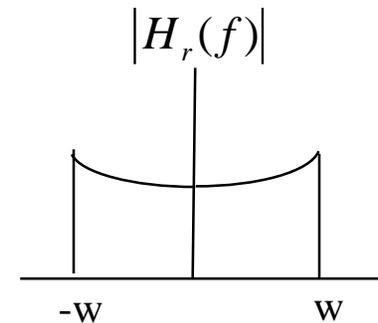
Reconstruction from zero-order hold



- We know from the sampling theorem that in order to reconstruct $x(t)$ from the impulse train samples on the left ($x_\delta(t)$) the filter on the right ($H(f)$) must be an ideal rectangular filter

$$H(f) = T_s \Pi\left(\frac{f}{2W}\right) = T_s \Pi(T_s f)$$

$$\left. \begin{aligned} H(f) &= T_s \Pi\left(\frac{f}{2W}\right) = H_0(f) H_r(f) \\ H_0(f) &= e^{-j\pi f T} \left(\frac{\sin(\pi f T)}{\pi f} \right) \end{aligned} \right\} \Rightarrow H_r(f) = \frac{T_s \Pi\left(\frac{f}{2W}\right) e^{j\pi f T}}{\sin(\pi f T)}$$



Aliasing

- **Sampling theorem requires that the signal be sampled at a frequency greater than twice its bandwidth**
- **When sampling at a frequency less than $2W$, the replicas of the frequency spectrum overlap and cannot be “separated” using a low pass filter**
- **This is referred to as aliasing**
 - Higher frequencies are “reflected” only lower frequencies
 - Signal cannot be recovered
- **The term aliasing refers to the fact that the higher frequency signals become indistinguishable from the lower frequency ones**