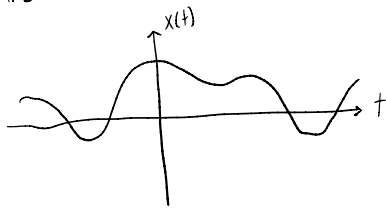


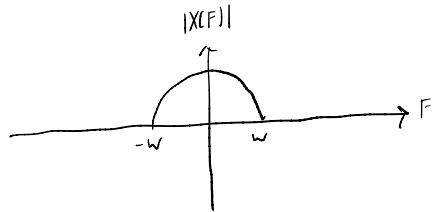
Sampling

The process by which we get from continuous waveforms to discrete ones

Time-Domain $x(t)$:

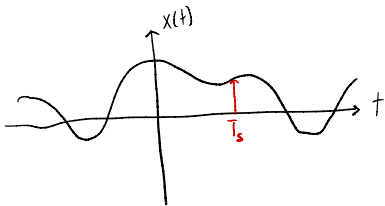


Frequency-Domain $X(F)$:

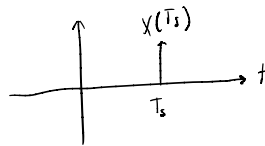


How to sample at a point $t = T_s$?

→ multiply by $\delta(t - T_s)$, then integrate (a dot product!)



→



sifting property

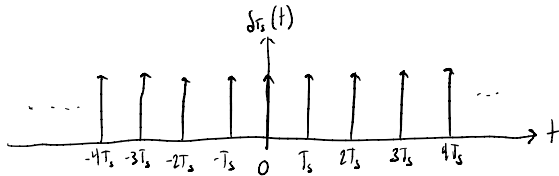
$$\rightarrow x(t = T_s) = \langle x(t), \delta(t - T_s) \rangle = \int_{-\infty}^{\infty} x(t) \delta(t - T_s) dt = x(T_s)$$

Okay, but one sample isn't much information...

Let's multiply our function by a "train" of delta functions

$$\text{Let } \delta_{T_s}(t) = \sum_{n=-\infty}^{\infty} \delta(t - nT_s)$$

$T_s = 1/F_s$

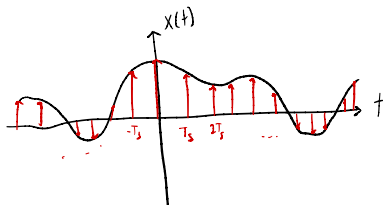


We get a "sampled sequence" by multiplying $x(t)$ by $\delta_{T_s}(t)$

$$\text{Let } \bar{x}(t) = x(t) \delta_{T_s}(t)$$

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

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



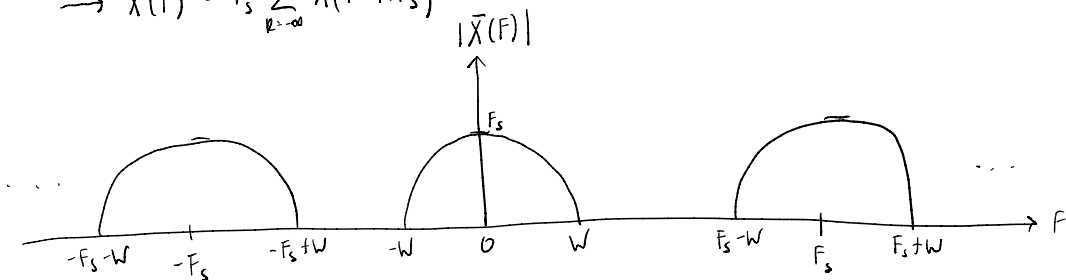
What does look like in the frequency domain?

• Multiplication in time is convolution in frequency

• So, $\bar{X}(F) = \mathcal{F}\{x(t)\} * \mathcal{F}\{\delta_T(t)\}$

$$= X(F) * \sum_{k=-\infty}^{\infty} \delta(F - kF_s)$$

$$\rightarrow \bar{X}(F) = F_s \sum_{k=-\infty}^{\infty} X(F - kF_s)$$



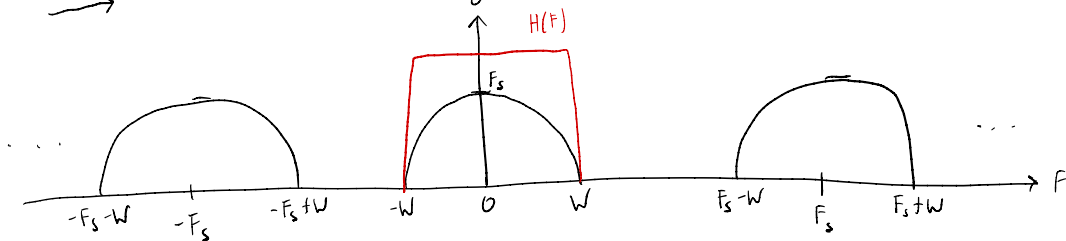
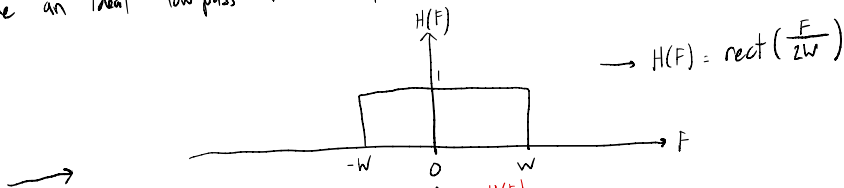
• Fourier transform becomes periodic! We have copies of original spectrum spaced out at integer multiples of sampling frequency and scaled in magnitude by F_s

$\bar{X}(t)$ is the "continuous" representation of the DT sequence $x[n]$

Q: Can we recover original $x(t)$ from $\bar{X}(t)/x[n]$?

A: Yes - we just need to filter out all the periodic copies of $X(F)$ from the spectrum of $\bar{X}(t)$

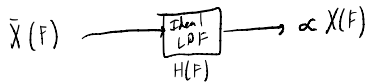
Define an ideal low-pass filter response $H(F)$:



$\rightarrow$ Then $\bar{X}(F)H(F) = F_s X(F)$

and we get

$$x(t) = \frac{1}{F_s} \mathcal{F}^{-1}\{\bar{X}(F)H(F)\}$$

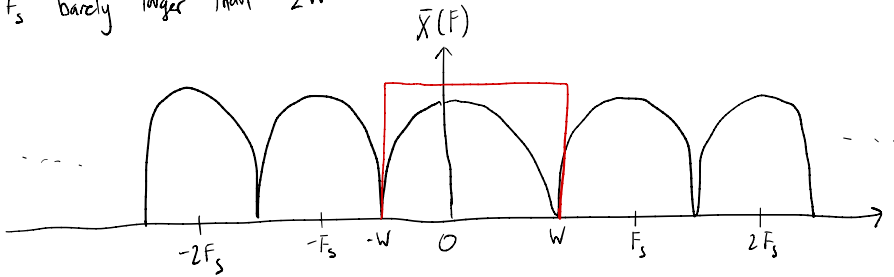


What happens if we change F_s ?

→ Copies of spectrum get closer or further apart

We can still recover original function if no overlap

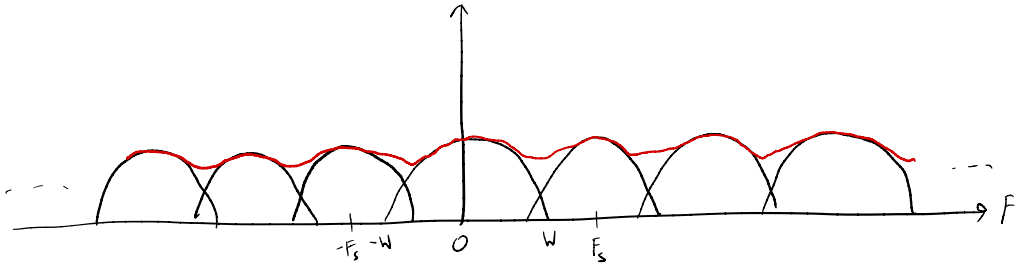
Ex. F_s barely larger than $2W$



Copies barely don't touch

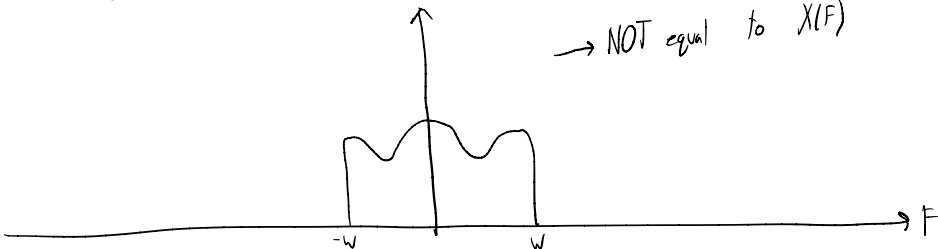
- Theoretically, w/ ideal LPF, can still recover $x(t)$

What if $F_s < 2W$?



The periodic copies of $X(F)$ overlap.

After applying ideal LPF:



→ NOT equal to $X(F)$

The original signal cannot be recovered. The sampling rate is too slow, leading to spectral overlap and permanent distortion of the original signal.

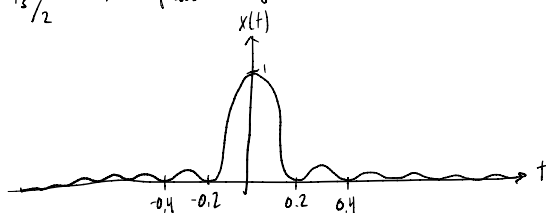
→ This overlap is called aliasing.

∴ A signal w/ max frequency W Hz ("band-limited" to W Hz) can be reconstructed exactly from its samples if the sampling frequency $F_s > 2W$

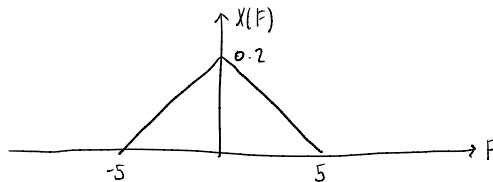
• The minimum sampling frequency $F_s = 2W$ is called the "Nyquist Rate" or the "Nyquist Frequency"

Another way: For a given set of samples taken at rate F_s then there is only one signal of bandwidth $B \leq F_s/2$ that passes through those samples

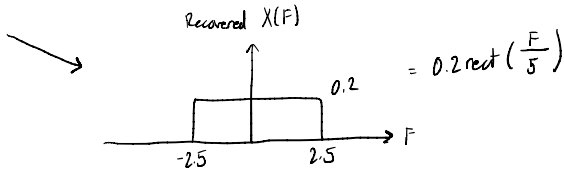
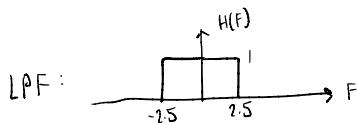
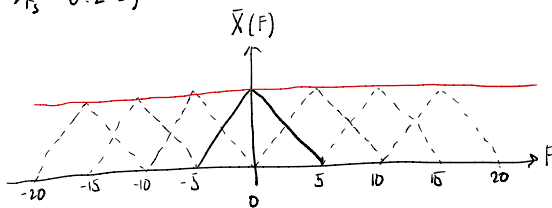
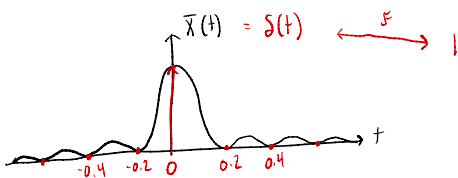
Example: $x(t) = \text{sinc}^2(5t)$



$$\rightarrow X(F) = 0.2 \Delta\left(\frac{F}{10}\right)$$

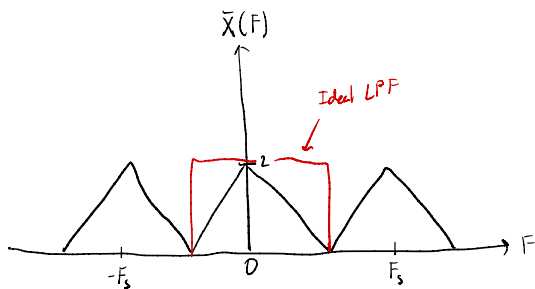
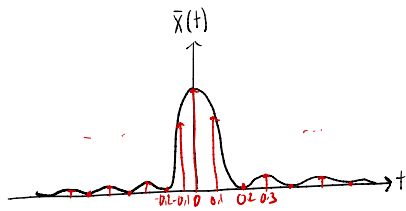


If we sample at $F_s = 5$ Hz ($T_s = 1/F_s = 0.2$ s)

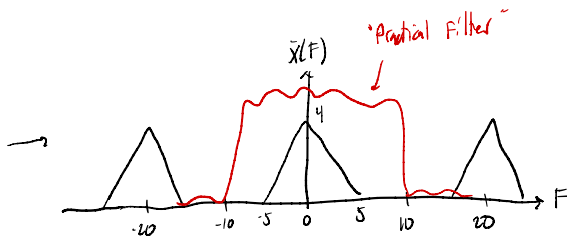


$$\rightarrow x(t) = \text{sinc}(5t) \neq \text{original } x(t)$$

Nyquist Rate $F_s = 10 \text{ Hz}$ ($T_s = 0.1 \text{ s}$)



Sample at $F_s = 20 \text{ Hz}$:



Another perspective on aliasing:

$$X_1(t) = \cos(2\pi F_1 t), \quad X_2(t) = \cos(2\pi F_2 t)$$

Suppose $F_2 = F_1 + kF_s$ for same F_s
(k integer)

int. multiple 2π

$$\rightarrow \bar{X}_1(nT_s) = \cos(2\pi F_1 nT_s)$$

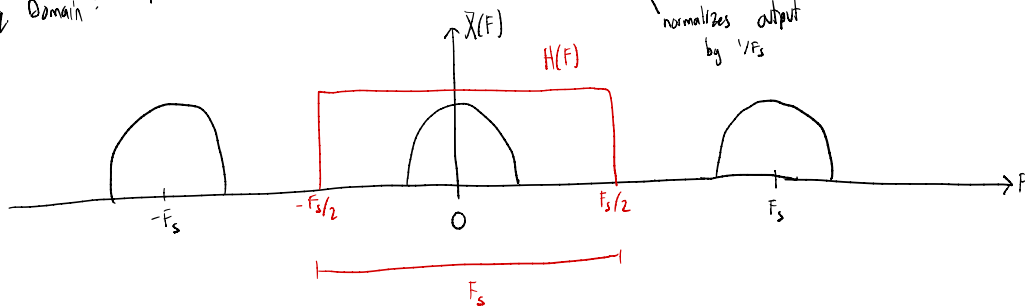
$$\bar{X}_2(nT_s) = \cos(2\pi(F_1 + kF_s)nT_s) = \cos(2\pi F_1 nT_s + 2\pi k n F_s T_s) = \cos(2\pi F_1 nT_s + 2\pi k n) = \cos(2\pi F_1 nT_s) = \bar{X}_1(nT_s)$$

$\therefore$ Sinusoids spaced in frequency by integer multiples of F_s are indistinguishable from one another

Matlab example

Reconstruction: Ok, we know how to sample - but how to reconstruct?

Freq. Domain: Filter out periodic replicas w/ $H(F) = \frac{1}{F_s} \text{rect}(F/F_s)$ (Assume signal band-limited below $F_s/2$)



Time - Domain:

$$H(F) = \frac{1}{F_s} \text{rect}(F/F_s) \xleftrightarrow{\mathcal{F}^{-1}} h(t) = \text{sinc}(F_s t)$$

$$X(F) = \bar{X}(F) H(F) \xleftrightarrow{\mathcal{F}^{-1}} x(t) = \bar{x}(t) * h(t)$$

$$= \left(\sum_{n=-\infty}^{\infty} x(nT_s) \delta(t - nT_s) \right) * \text{sinc}(F_s t)$$

$$= \sum_{n=-\infty}^{\infty} x(nT_s) \left(\delta(t - nT_s) * \text{sinc}(F_s t) \right)$$

$$= \sum_{n=-\infty}^{\infty} x(nT_s) \text{sinc}(F_s(t - nT_s))$$

Ideal Interpolation Formula

$\therefore$ Original signal can be reconstructed w/ a weighted sum of sinc functions

Matlab example w/ $-5 \leq t \leq 5$ and $-20 \leq t \leq 20$

Infinite non-causal sinc not possible in reality... normally truncate or use a window

$$\rightarrow h(t) = \text{sinc}(F_s t) w(t)$$

$$w(t) \propto$$

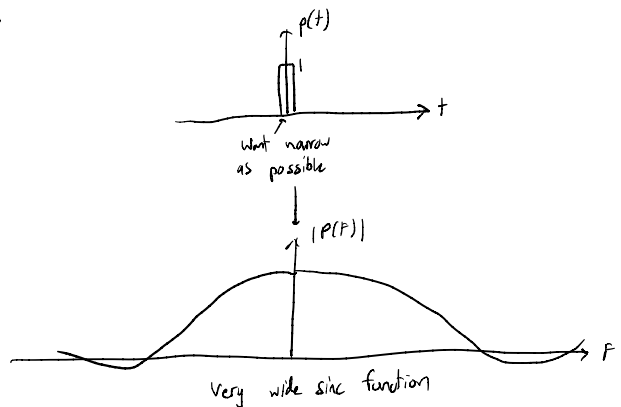
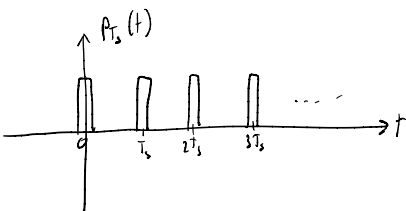

Practical Sampling

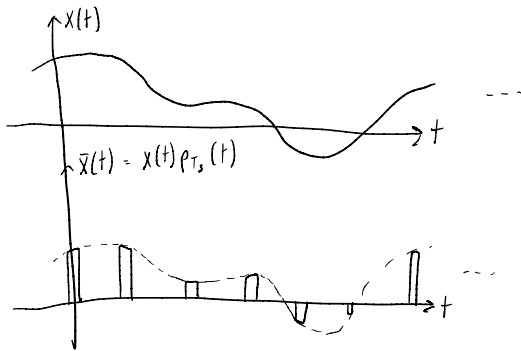
- In practice, we can't really make impulses $\delta(t)$ for sampling $(\delta_{T_s}(t))$

- Practical: A train of narrow pulses

$\rightarrow$ Suppose $p(t)$ is a single pulse

$$\text{Pulse Train: } P_{T_s}(t) = \sum_{n=-\infty}^{\infty} p(t - nT_s)$$





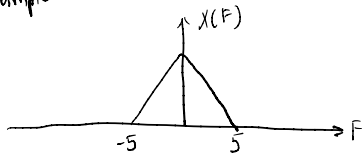
What does $\bar{X}(F)$ look like?

We get $p_{T_s}(t)$ by $p_{T_s}(t) = p(t) * \delta_{T_s}(t) = \sum_{n=-\infty}^{\infty} p(t) * \delta(t - nT_s)$

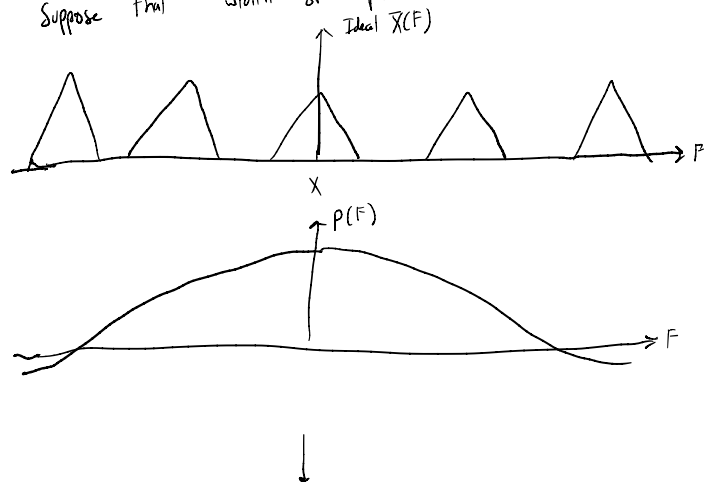
$$\rightarrow \bar{X}(F) = \mathcal{F}\{x(t)p_{T_s}(t)\} = \mathcal{F}\{x(t)\} * \mathcal{F}\{p_{T_s}(t)\} = X(F) * \underbrace{\mathcal{F}\{p_{T_s}(t)\}}_{p_{T_s}(t) = p(t) * \delta_{T_s}(t) \xrightarrow{\mathcal{F}} P(F)F_s \sum_{k=-\infty}^{\infty} \delta(F - kF_s)}$$

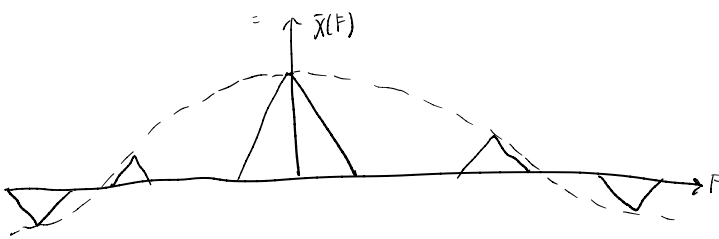
$$\text{So, } \bar{X}(F) = X(F) * \left(P(F)F_s \sum_{k=-\infty}^{\infty} \delta(F - kF_s) \right) = \underbrace{P(F)F_s}_{\text{Ideal } \bar{X}(F)} \sum_{k=-\infty}^{\infty} X(F - kF_s) = P(F) \bar{X}_{\text{ideal}}(F)$$

Previous Example: $x(t) = \text{sinc}^2(5t)$



Suppose that width of $p(t) \ll T_s \rightarrow \text{BW of } p(t) \gg F_s$





$\therefore$ As long as $P(F) \approx \text{constant}$ within the interval $-W \leq F \leq W$ we can approximately reconstruct $x(t)$

In other words, $p(t)$ must be narrower than the rate of change of $x(t)$

As $p(t)$ gets narrower, $\rightarrow \delta(t)$ and $P(F) \rightarrow 1 \therefore \text{ideal } \bar{X}(F)$ again