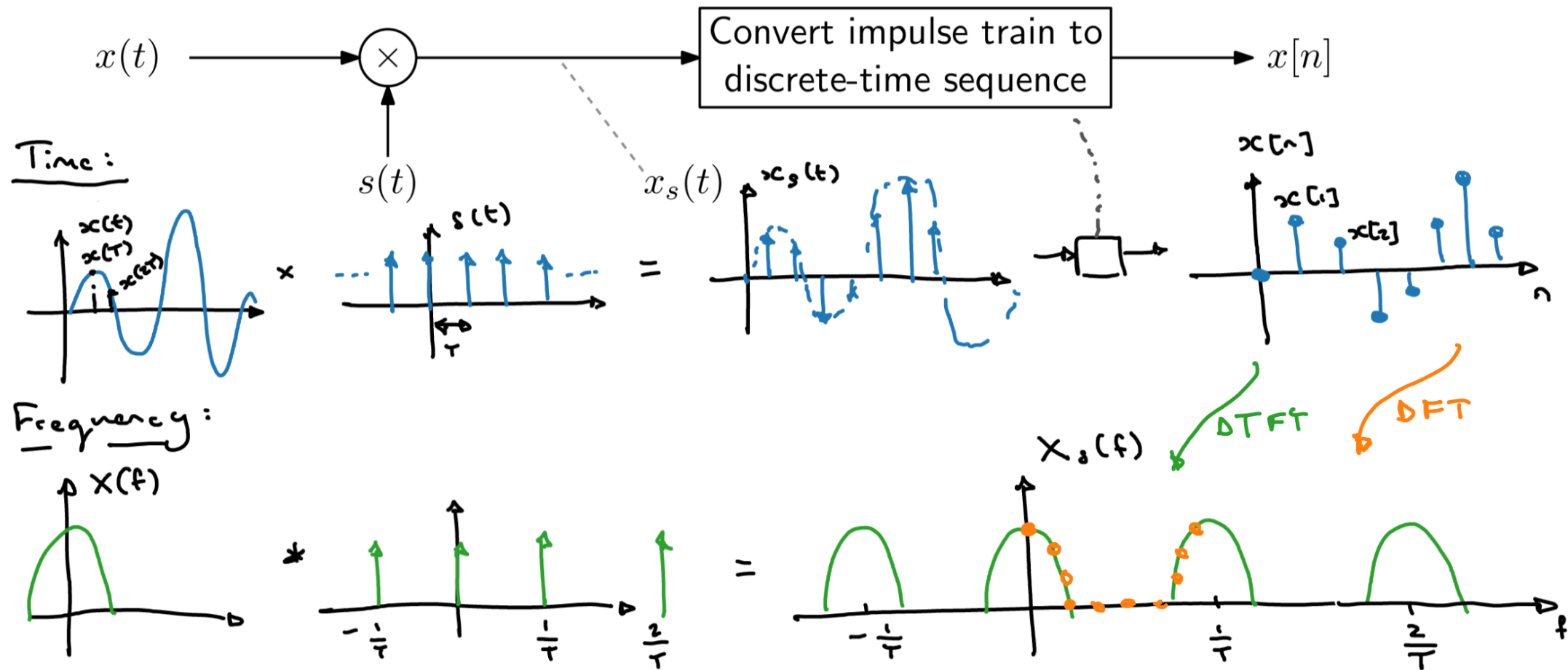


Discrete Fourier transform

Herman Kamper

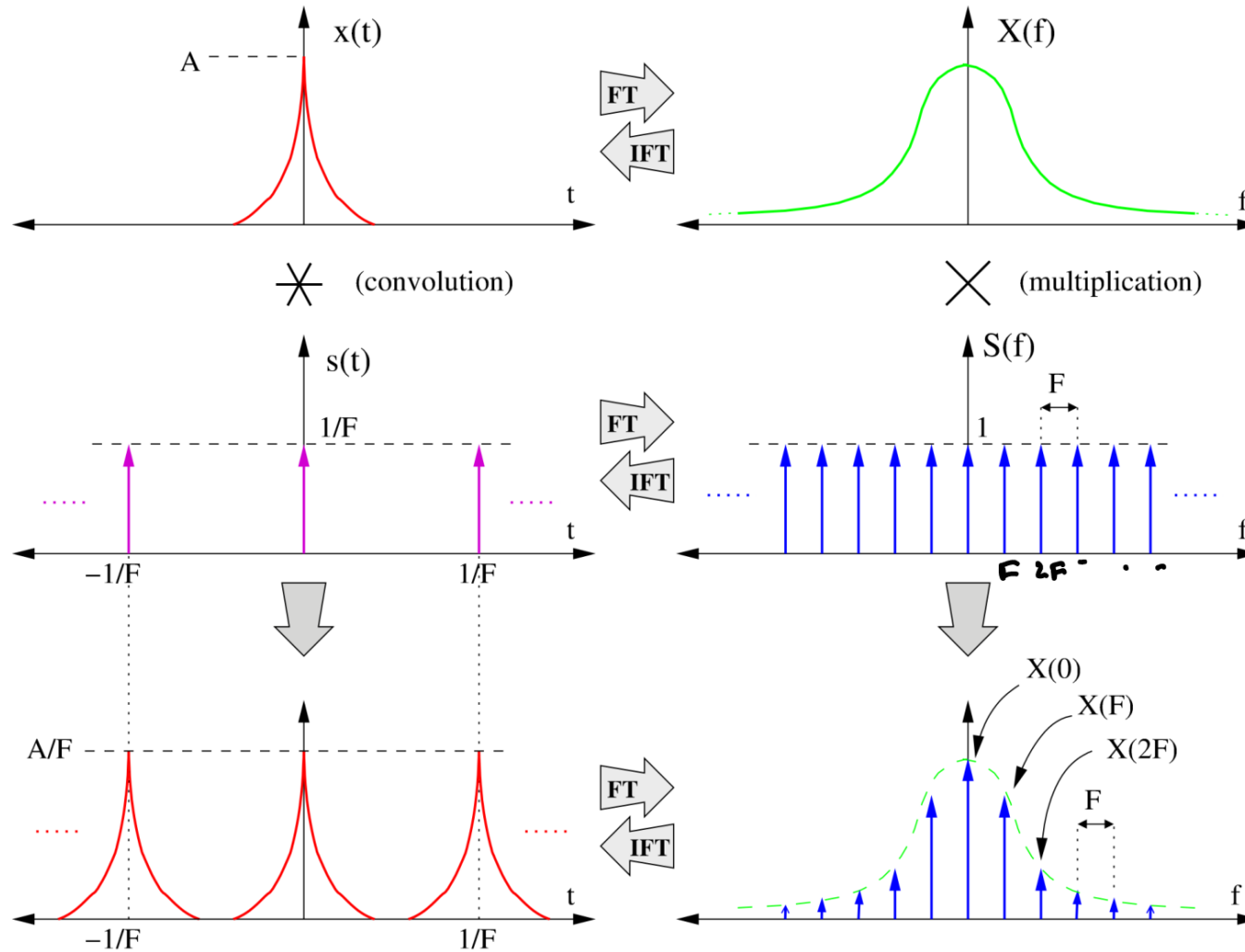
Sampling in time domain



Getting to the DFT

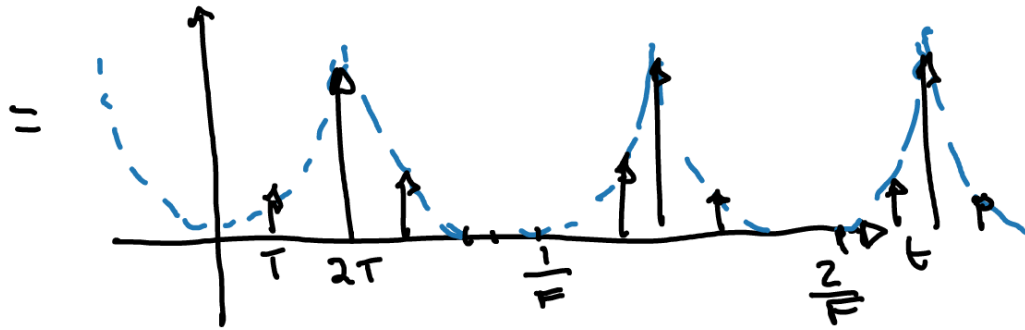
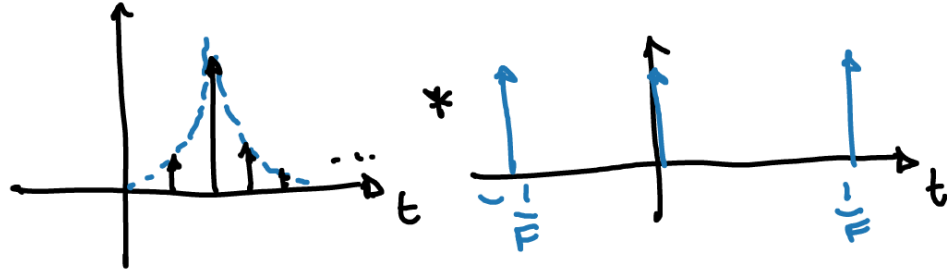
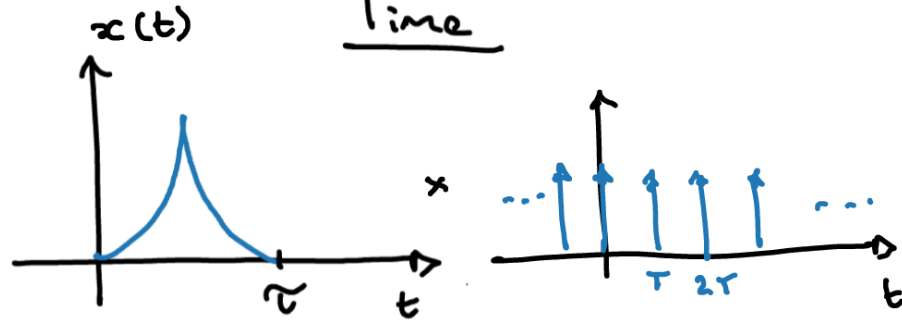
- ✓ • Discrete signal: Periodic continuous spectrum (DTFT)
- ✓ • Discrete spectrum: Periodic continuous signal
- ✓ • Discrete periodic signal: Both signal and spectrum discrete
- ✓ • But what if your signal is not periodic? Just hack it: window and repeat
 - Discrete Fourier transform (DFT): Discrete time and spectrum for arbitrary signals

Sampling in frequency domain

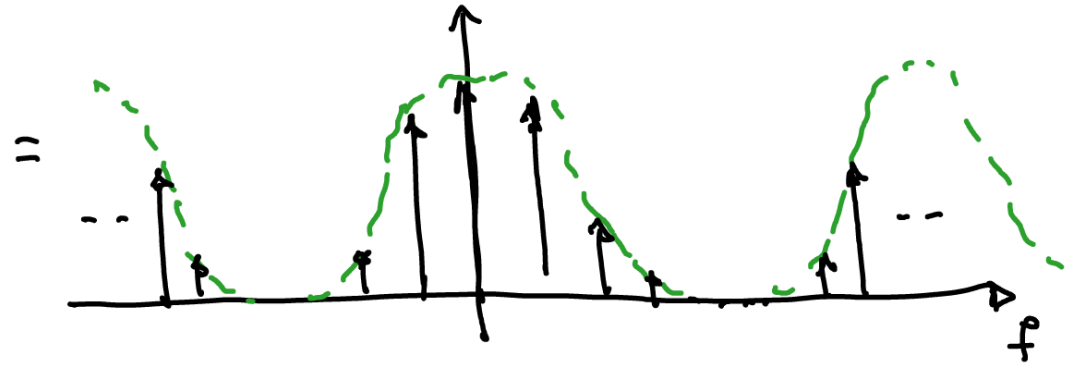
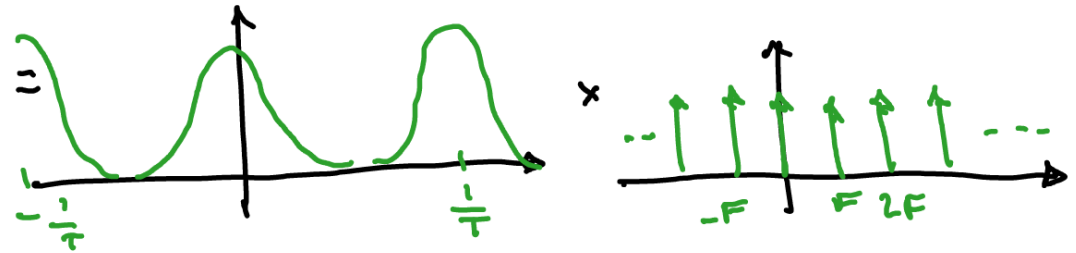
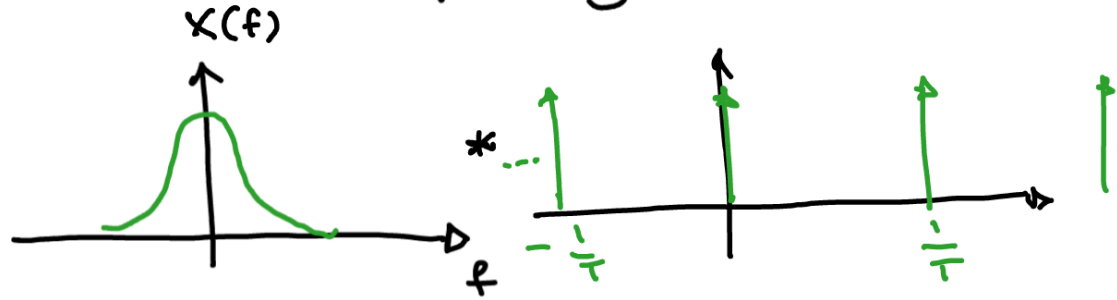


Sampling in time and frequency

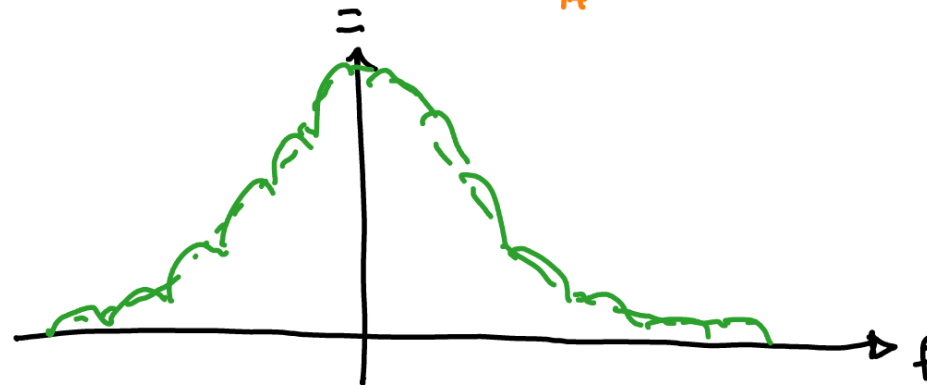
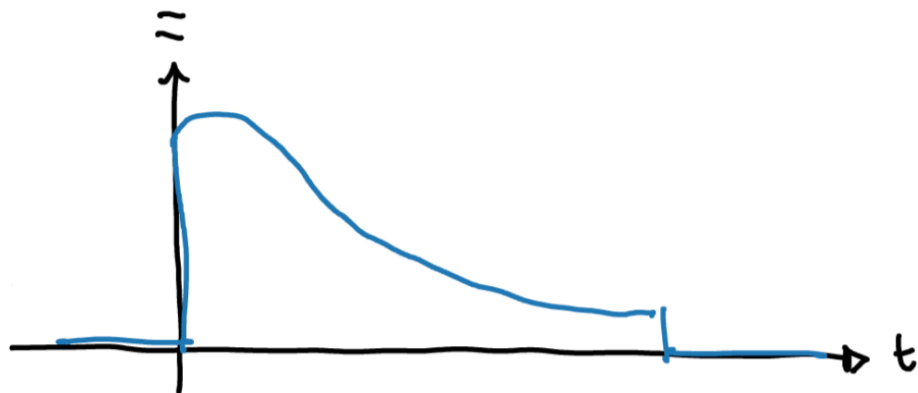
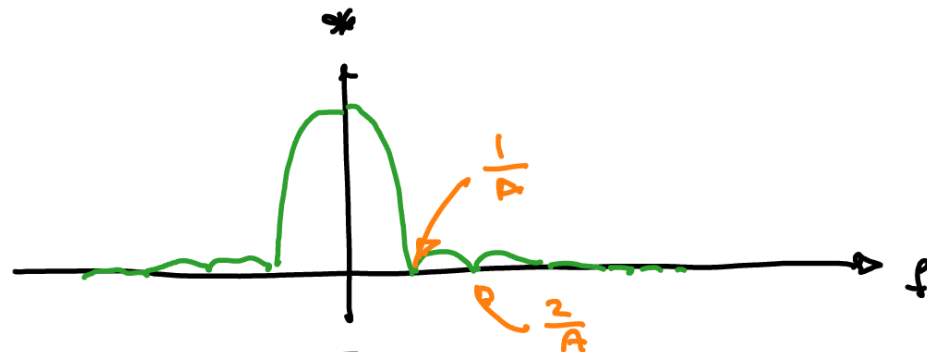
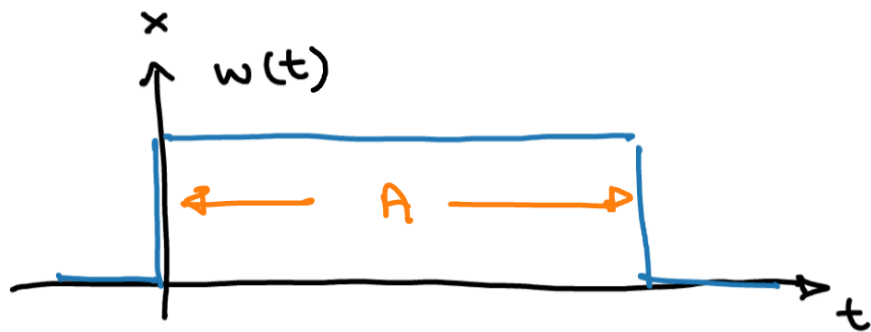
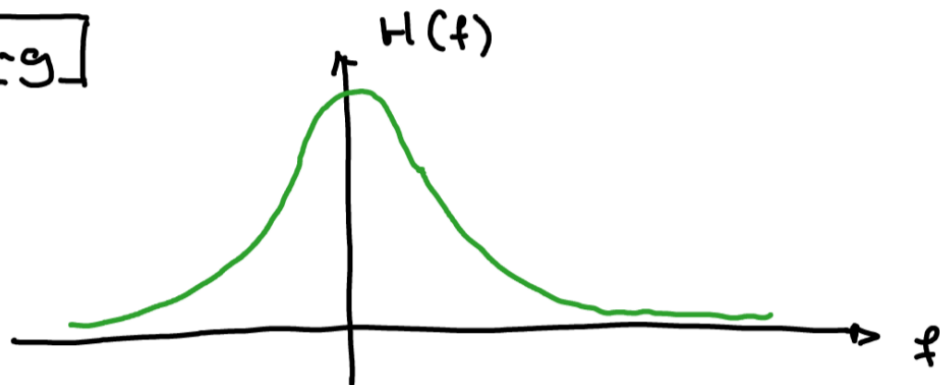
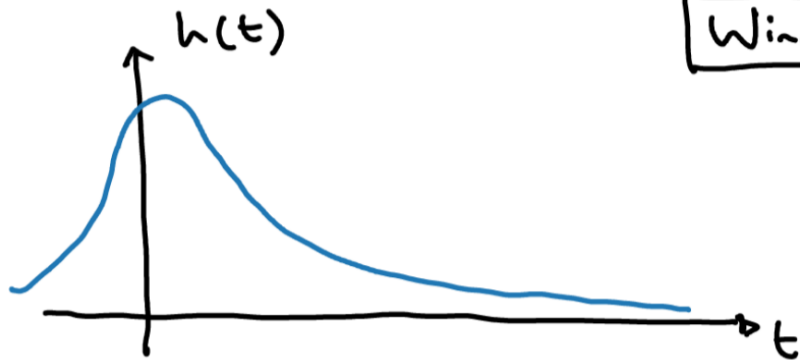
Time



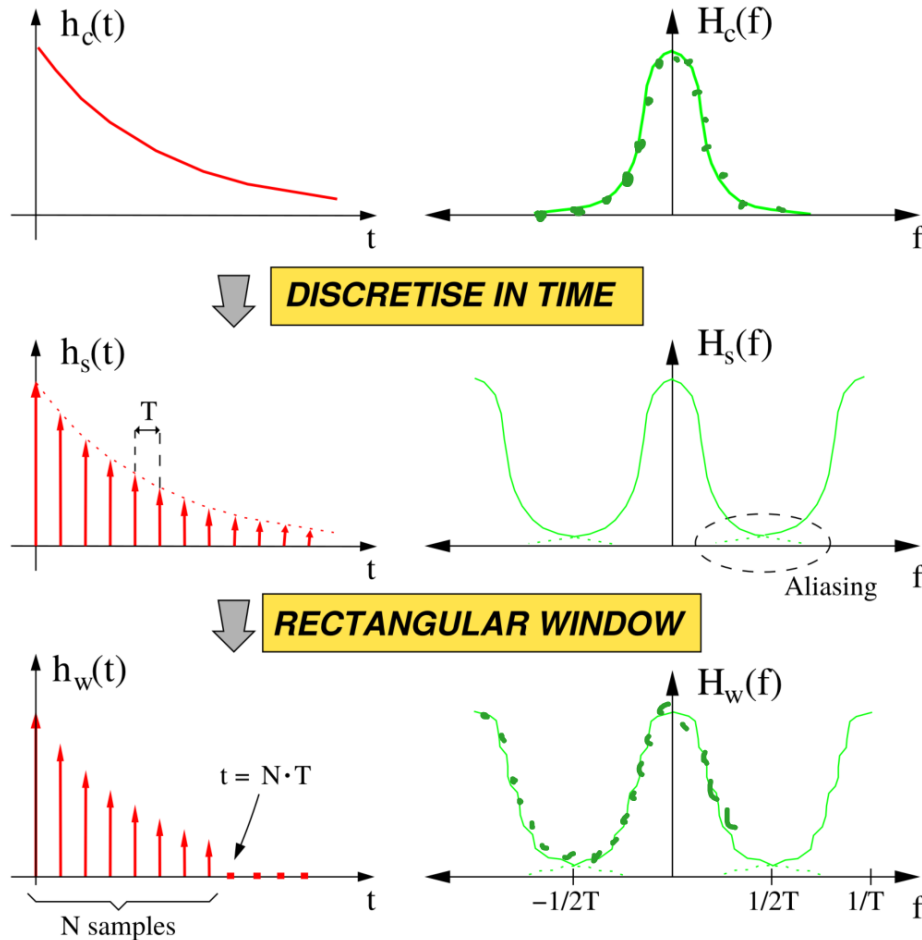
Frequency



Windowing



Discrete Fourier transform (DFT)



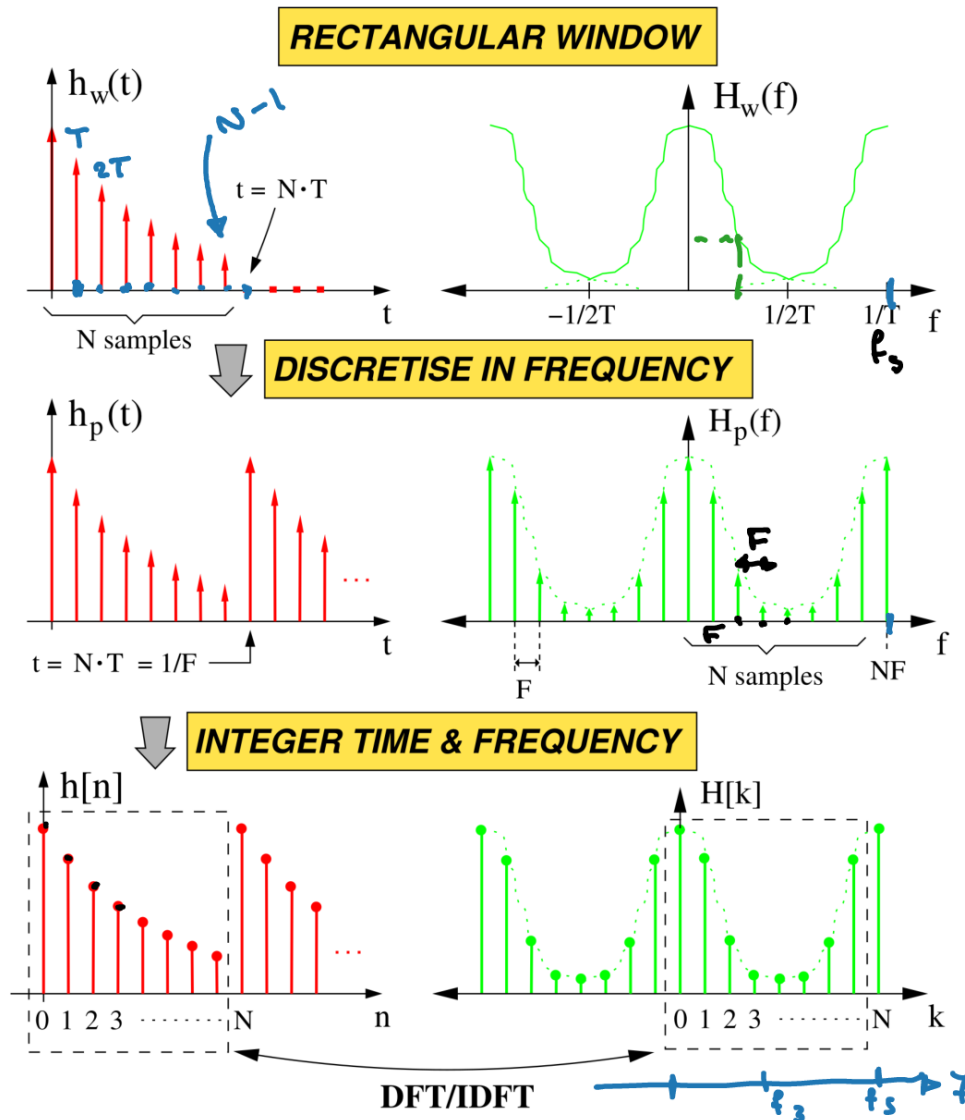
Sampling:

$$\text{DTFT: } H_s(f) = \sum_{n=-\infty}^{\infty} \overbrace{h(nT)}^{h[n]} \cdot e^{-j2\pi f nT}$$

Window:

$$\begin{aligned} H_w(f) &= H_s(f) * W(f) \\ &= \sum_{n=0}^{N-1} h(nT) \cdot e^{-j2\pi f nT} \end{aligned}$$

Discrete Fourier transform (DFT)



Window:

$$H_w(f) = H_s(f) * W(f)$$

$$= \sum_{n=0}^{N-1} h(nT) \cdot e^{-j2\pi f nT}$$

$$\frac{1}{F} = NT$$

Periodic extension:

$$H_p(f) = H_w(f) * \sum_{k=-\infty}^{\infty} \delta(f - kF)$$

Discrete Fourier transform:

$$h[n] = h(nT)$$

$$H[2] = H_p(2F)$$

$$H[k] = H_w(kF)$$

$$= \sum_{n=0}^{N-1} h[n] \cdot e^{-j2\pi kF nT}$$

$$kFT = k \frac{1}{NT}$$

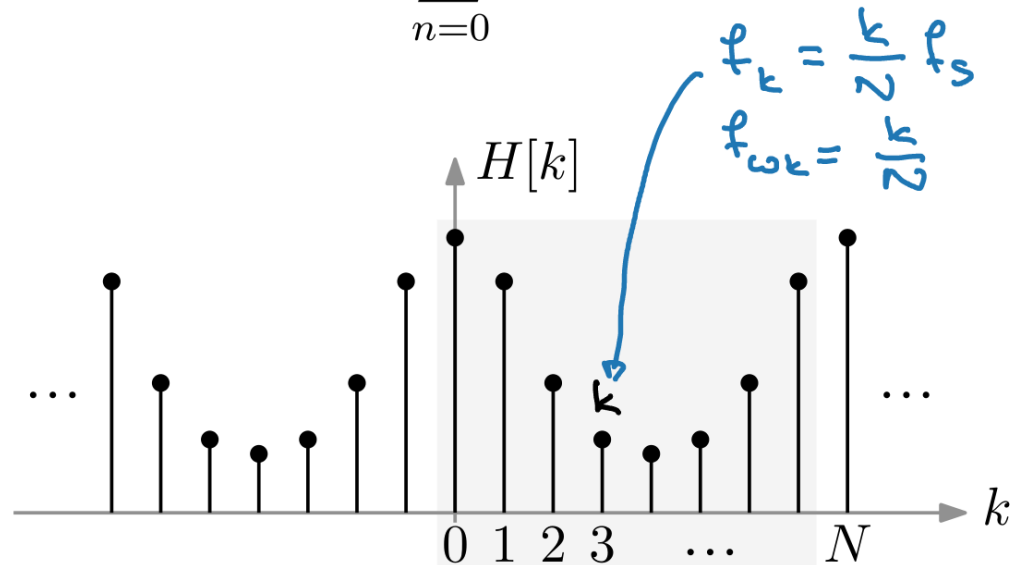
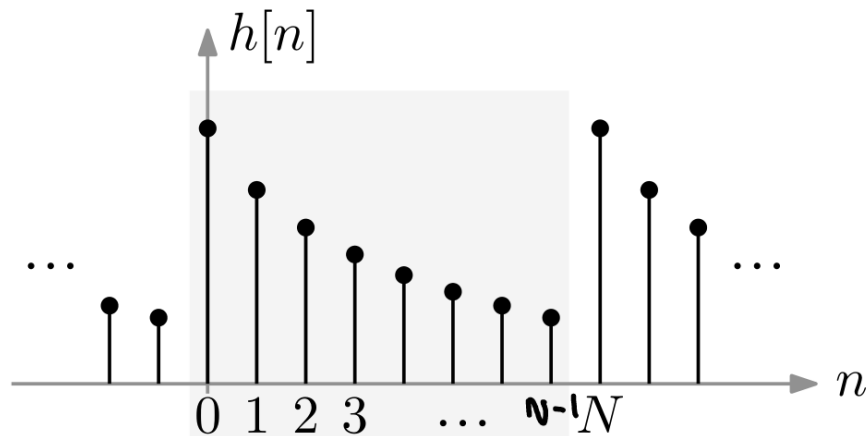
$$= \sum_{n=0}^{N-1} h[n] \cdot e^{-j2\pi \frac{k}{N} n}$$

IDFT:

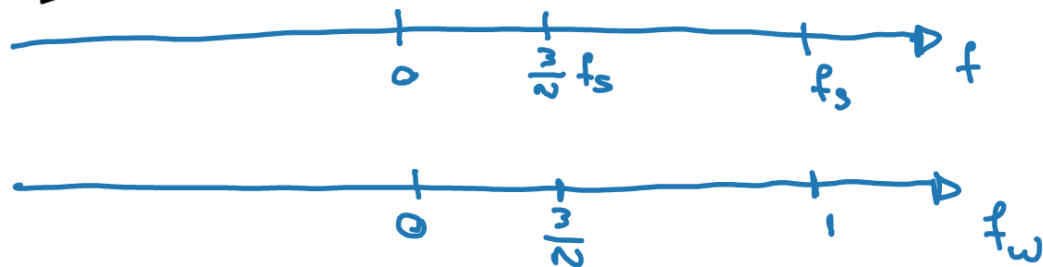
$$h[n] = \frac{1}{N} \sum_{k=0}^{N-1} H[k] e^{j2\pi kn/N}$$

DFT:

$$H[k] = \sum_{n=0}^{N-1} h[n] e^{-j2\pi kn/N}$$



DFT \rightarrow
 \leftarrow IDFT



Inverse discrete Fourier transform (IDFT)

An N -point discrete periodic signal can be expressed as the sum of N complex exponentials with discrete-time frequencies $\omega_k = \frac{2\pi k}{N}$.

If you don't believe me: you've proved it already in the DFT. The discrete $H[k]$ is given by a sum of N complex exponentials. We are now flipping things.

So we know that we should be able to write:

$$h[n] = \sum_{k=0}^{N-1} c_k e^{j2\pi nk/N}$$

with c_k unknown. The goal is to find all the c_k s.

Multiply the above equation by $e^{-j2\pi nk_1/N}$ on both sides, and take the sum over k_1 :

$$\begin{aligned} h[n]e^{-j2\pi nk_1/N} &= \left[\sum_{k=0}^{N-1} c_k e^{j2\pi nk/N} \right] e^{-j2\pi nk_1/N} \\ \sum_{k_1=0}^{N-1} h[n]e^{-j2\pi nk_1/N} &= \sum_{k_1=0}^{N-1} \sum_{k=0}^{N-1} c_k e^{j2\pi n(k-k_1)/N} \\ &= \sum_{k=0}^{N-1} c_k \sum_{k_1=0}^{N-1} e^{j2\pi n(k-k_1)/N} \\ &= c_{k_1} N \end{aligned}$$

That last step looks strange but it comes from

$$\sum_{n=0}^{N-1} e^{j2\pi kn/N} = \begin{cases} N & \text{if } k = 0, \pm N, \pm 2N, \dots \\ 0 & \text{otherwise} \end{cases}$$

which itself comes from the geometric series formula

$$\sum_{n=0}^{N-1} a^n = \begin{cases} N & \text{if } a = 1 \\ \frac{1-a^N}{1-a} & \text{if } a \neq 1 \end{cases}$$

So we have

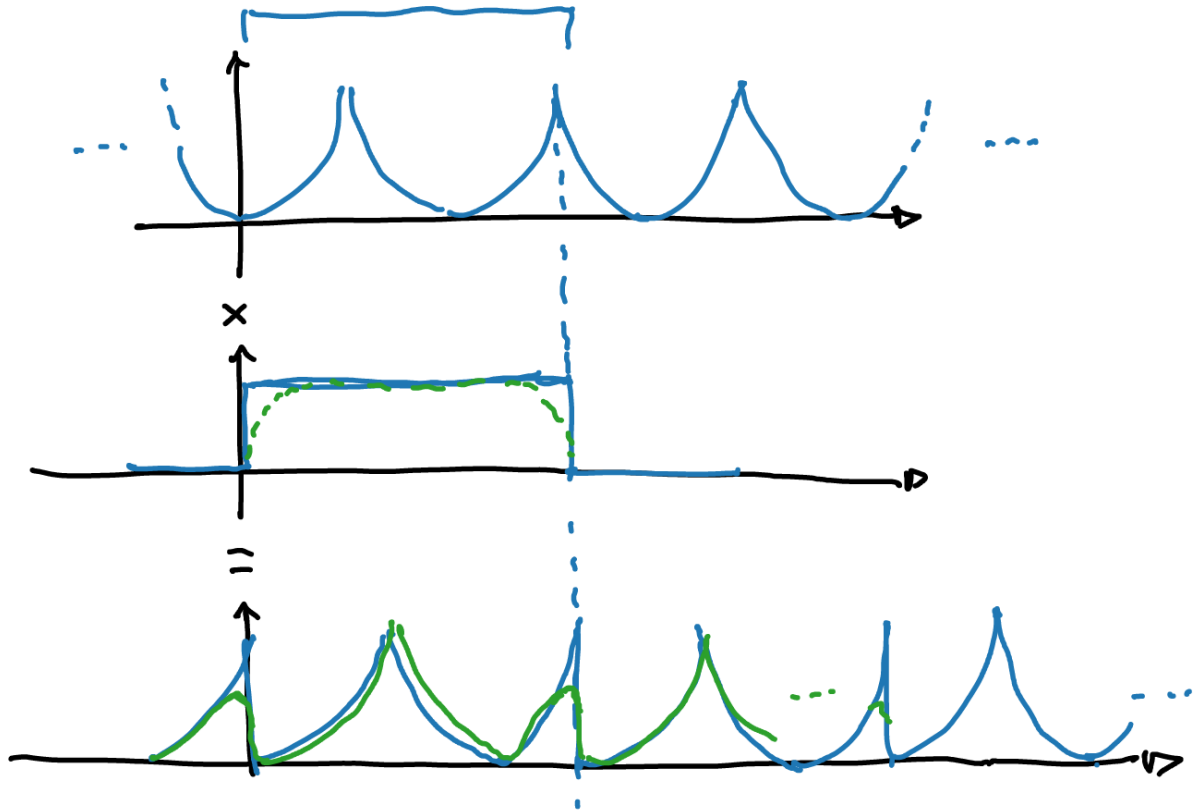
$$\begin{aligned} \sum_{k_1=0}^{N-1} h[n] e^{-j2\pi nk_1/N} &= c_{k_1} N \\ c_{k_1} &= \frac{1}{N} \sum_{k_1=0}^{N-1} h[n] e^{-j2\pi nk_1/N} \\ &= \frac{1}{N} H[k_1] \end{aligned}$$

If we plug this back into the equation where we started, we get the IDFT:

$$h[n] = \frac{1}{N} \sum_{k=0}^{N-1} H[k] e^{j2\pi nk/N}$$

How does the DFT mess up?

- Aliasing
- Windowing
- Periodicity



Windowing

