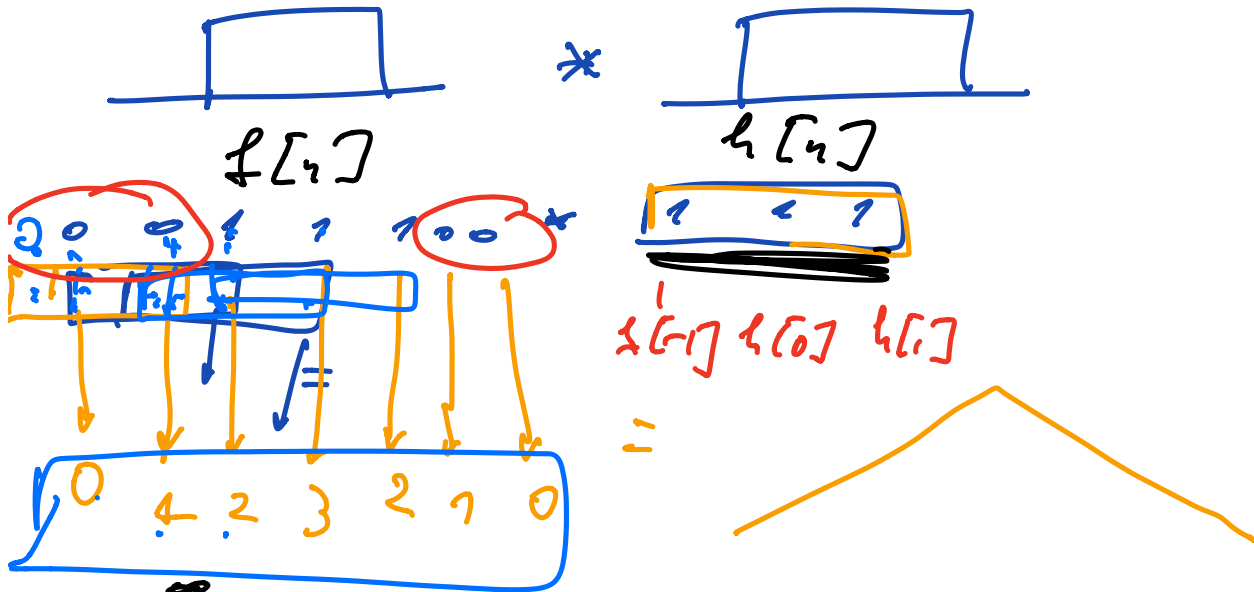


Few examples :



$$\sum_{k=-\infty}^{\infty} f[k] h[n-k]$$

COVID 7-day average $\frac{1}{2} | \frac{1}{2} | \frac{1}{2} | \dots | \frac{1}{2}$

$h[-k]$ $\begin{matrix} \frac{1}{2} \dots \frac{1}{2} \\ h[-6] \dots h[0] \end{matrix}$ $h[0] h[1] \dots h[6]$

$\dots \dots \dots 3.5 \dots 4.0 \dots$

value of inner product

Convolution in probabilities

$$x: 1 \ 2 \ 3 \ 4 \ 5 \ 6 \quad p(x) = \frac{1}{6}$$

$$y: 1 \ 2 \ 3 \ 4 \ 5 \ 6 \quad p(y) = \frac{1}{6}$$

$$z = x + y = 2, 3, \dots, 12$$

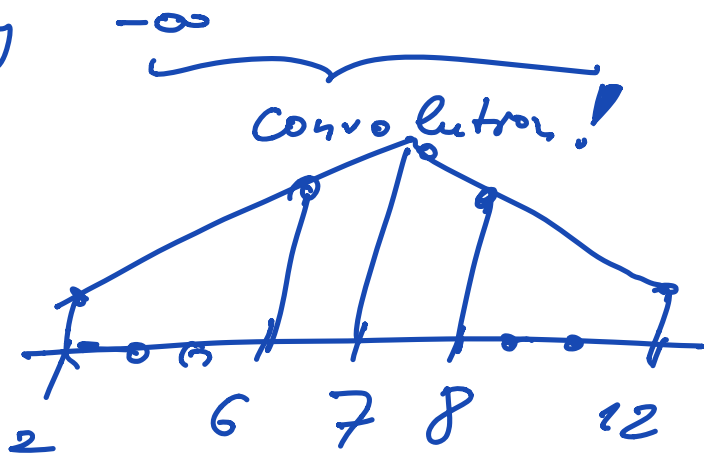
$$p(z) = p(x)p(y) = \int_{-\infty}^{\infty} p(x)p(z-x)dx$$

where $z = x + y$

$$p(8) = \frac{5}{36}$$

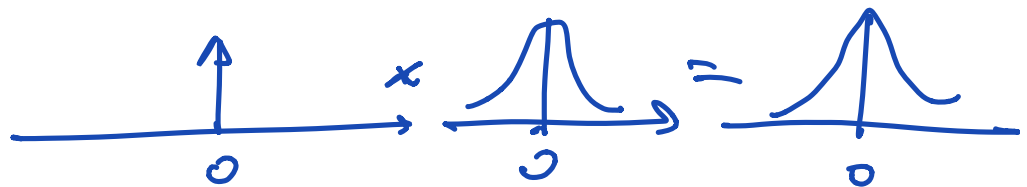
$$p(7) = \frac{6}{36}$$

$$p(6) = \frac{5}{36}$$

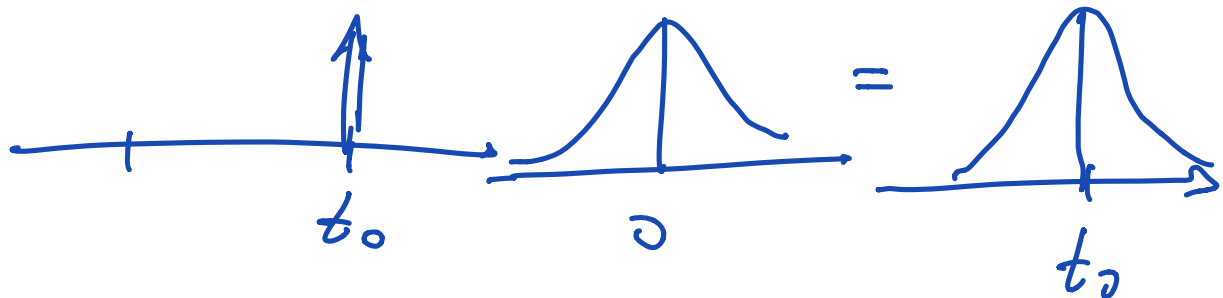


$$p(2) = p(12) = \frac{1}{36}$$

Dirac: $\delta(t) * h(t) = h(t)$



$$\delta(t-t_0) * h(t) = h(t-t_0)$$



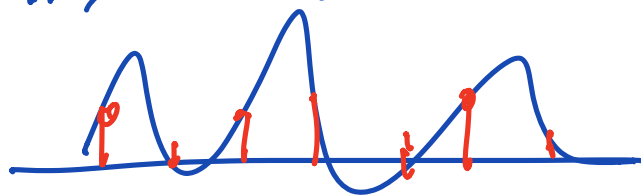
Dirac series
"comb"
impulse train

"
xeroxing"
←

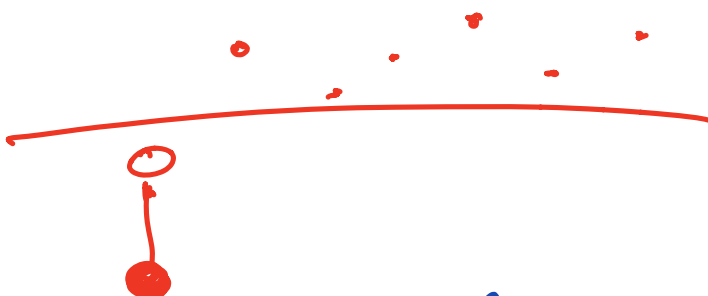
Sampling

multiply

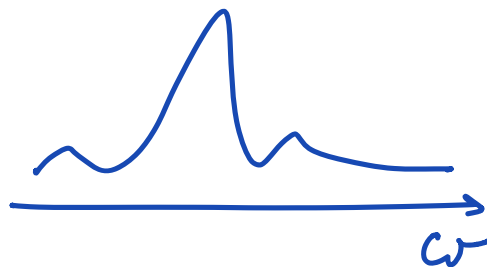
Continuous



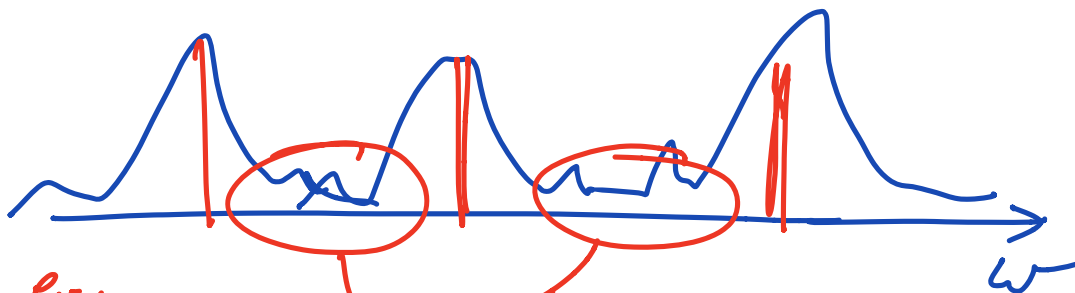
↓



*



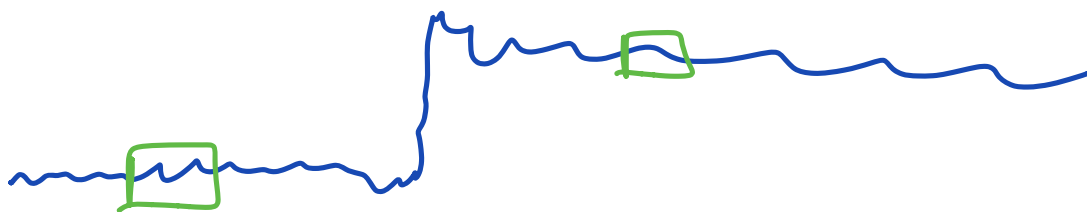
↓



Sampling
freq $\geq 30\text{ kHz}$
 ≥ 2 max freq.
 $\hookrightarrow 20\text{ kHz}$

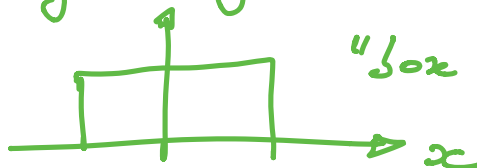
aliasing

edge detection



Smooth first !

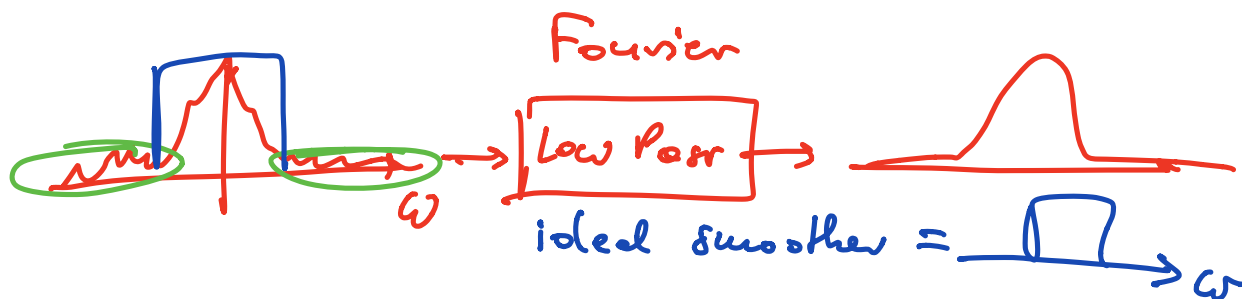
is averaging - good smoothing filter ?



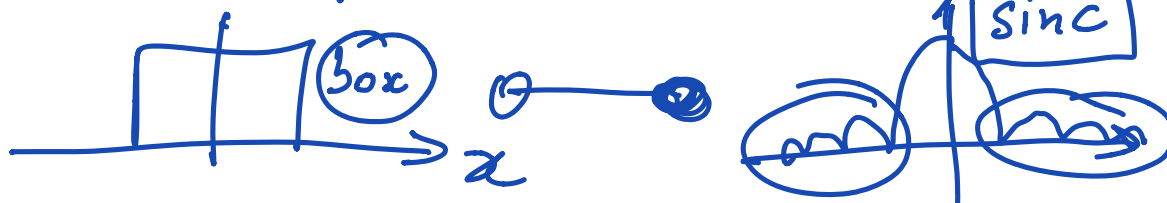
"box"

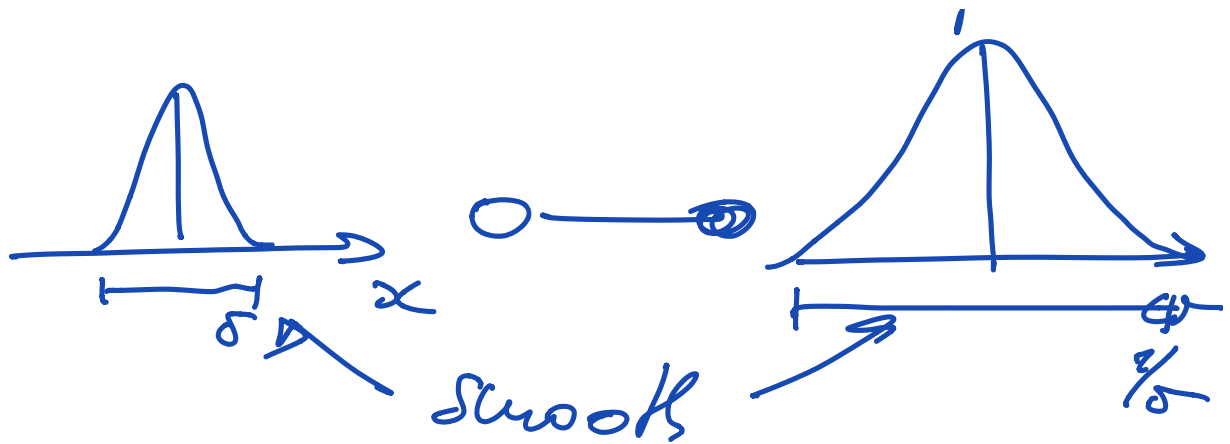
creates additional edges

smoothing is low-pass filtering



A box in space





GAUSSIAN

edge detection

1. smooth with g
2. first-derivative

$$f \rightarrow \boxed{} \rightarrow e$$

$$\frac{d(f * g)}{dx}$$

demo: Convolution commutes with differentiation

$$\frac{d(f * g)}{dx} = f * \frac{dg}{dx}$$

$$\frac{d}{dx} \int_{-\infty}^{\infty} f(t) g(x-t) dt \quad \xrightarrow{\text{proof 1}} \quad \frac{d}{dx} \int_{-\infty}^x f(t) g(x-t) dt$$

→ proof 2 via Fourier

$$f \mapsto F$$

$$f * g \mapsto FG$$

$$g \mapsto G$$

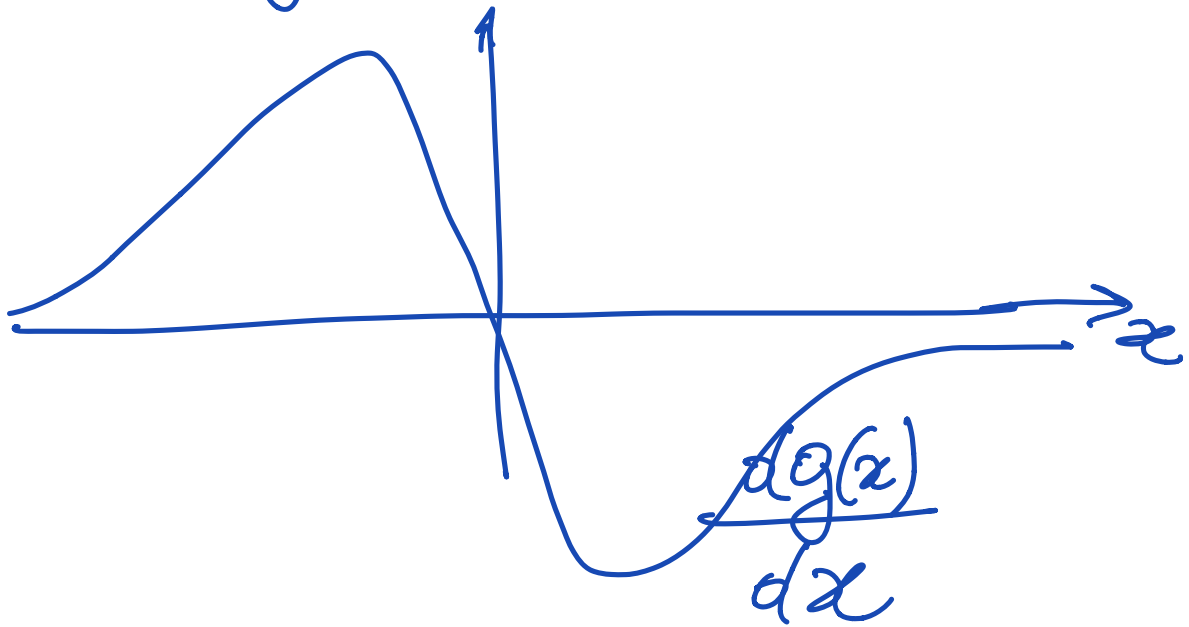
$$\frac{d}{dx} \mapsto j\omega$$

$$\frac{d}{dx} (f * g) \mapsto j\omega FG$$

$$\text{INV}(F) \times \text{inv}(j\omega G) \xrightarrow{\text{inverse Fourier}}$$

$$f * \frac{dg}{dx}$$

To detect an edge
we convolve with
first derivative of
a Gaussian ! out



and we find the
maximum !