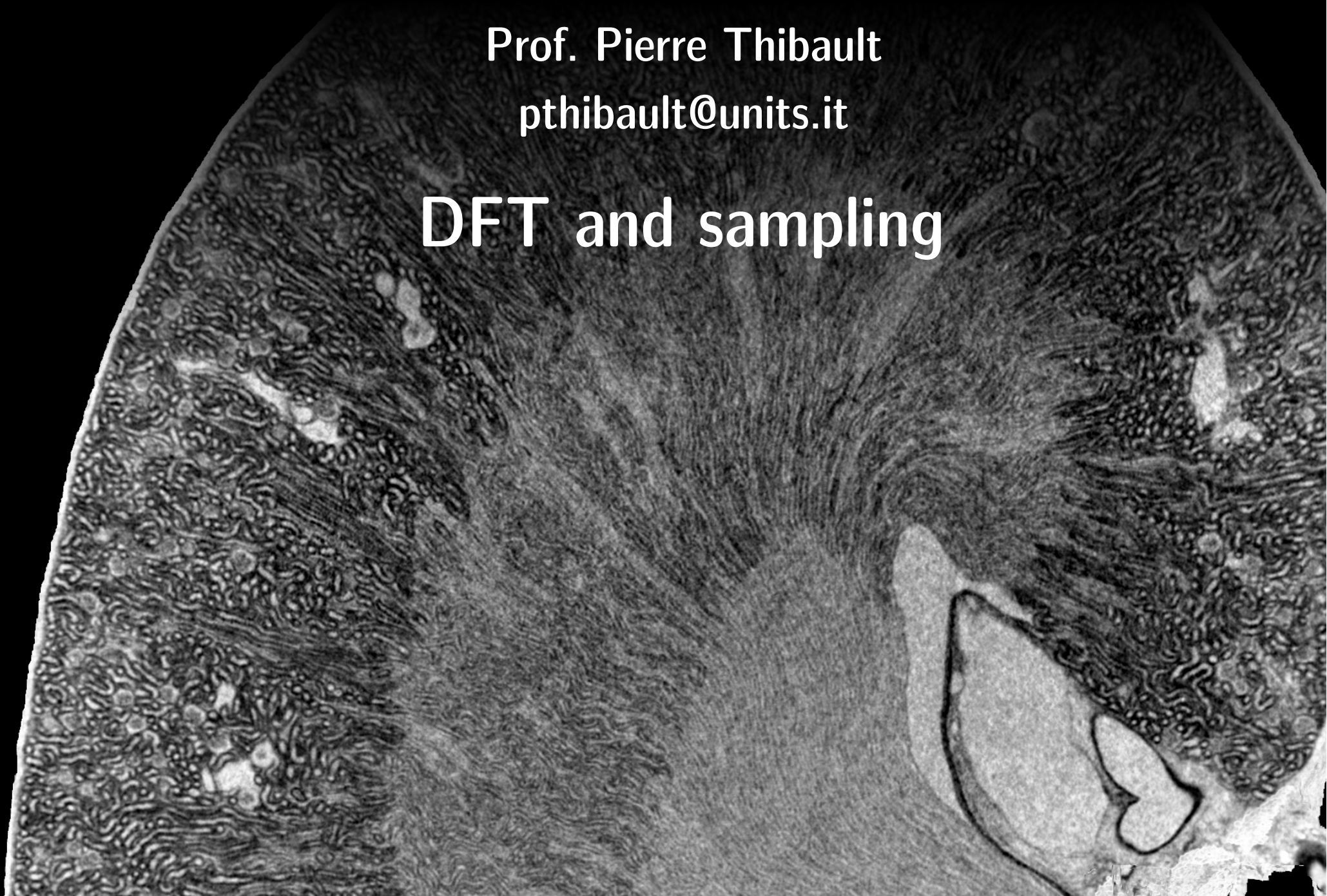


Image Processing for Physicists

Prof. Pierre Thibault

pthibault@units.it

DFT and sampling



Overview

- Discrete Fourier transform
 - Nyquist theorem
 - Undersampling and Aliasing
- Interpolation (resampling)

Reminder: Dirac distribution

- “sifting” property

$$\int_{-\infty}^{\infty} f(x) \delta(x-x_0) dx = f(x_0)$$

“selects” value of f
at position x_0 .

- normalization

- relation to Fourier transforms

$$\mathcal{F}\{1\} = \int_{-\infty}^{\infty} e^{-2\pi i u x} dx = \delta(u)$$

Periodic signals

$f(x)$: Periodic function with period p

Fourier synthesis equation

$$f(x) = \mathcal{F}^{-1}\{F\} = \int_{-\infty}^{\infty} F(u) e^{2\pi i u x} du$$

continuous variable u

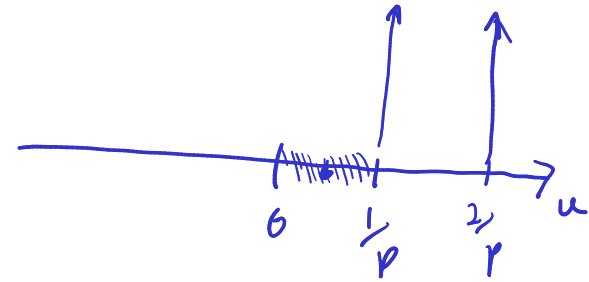
but also:

$$f(x) = \sum_{k=-\infty}^{\infty} c_k e^{2\pi i k x / p}$$

Fourier series discrete variable

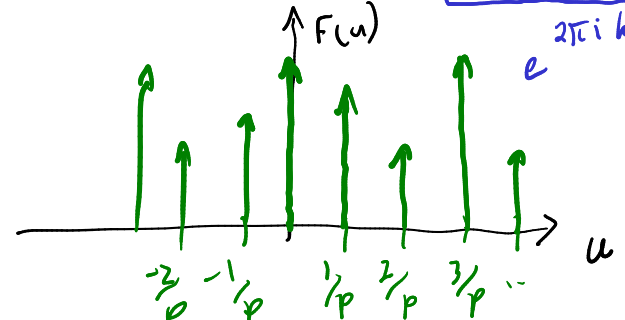
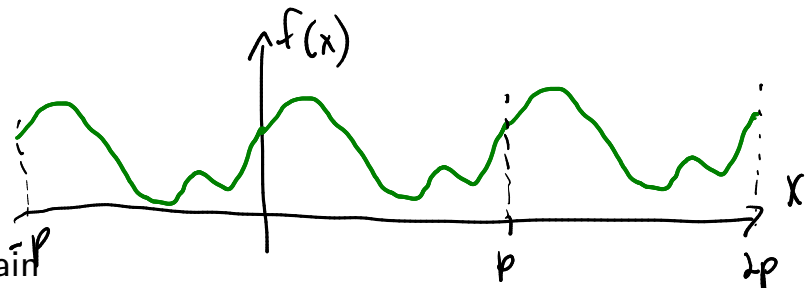
Therefore

$$F(u) = \sum_{k=-\infty}^{\infty} c_k \delta(u - k/p)$$



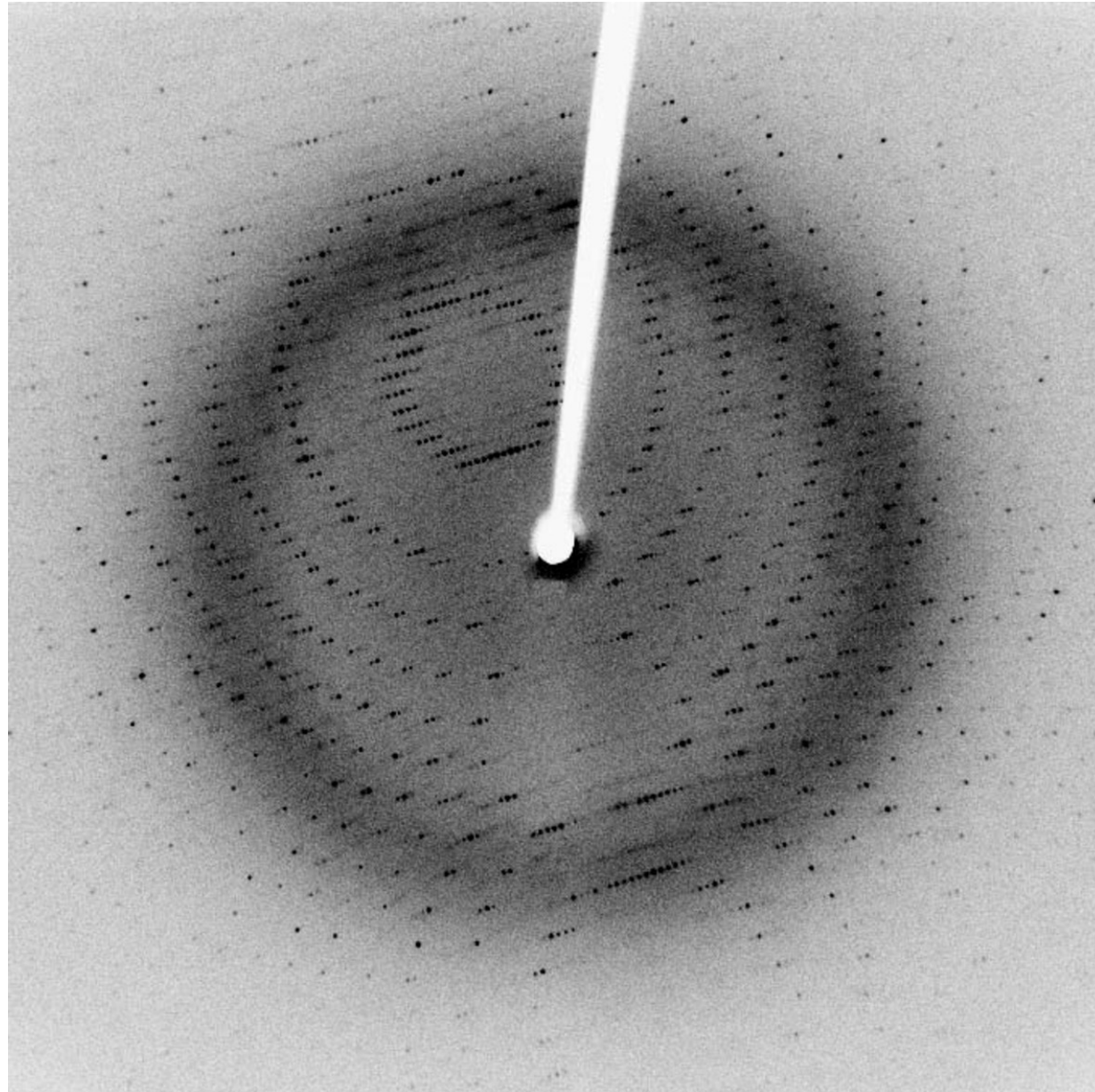
[because. $\int_{-\infty}^{\infty} F(u) e^{2\pi i u x} du = \int_{-\infty}^{\infty} \sum_k c_k \delta(u - k/p) e^{2\pi i u x} du = \sum_k c_k \int_{-\infty}^{\infty} \delta(u - k/p) e^{2\pi i u x} du$

$e^{2\pi i k x / p}$



Periodic signals

X-ray diffraction by a crystal



"Bragg peaks"
⇕
Dirac deltas
caused by
periodicity

The Dirac comb

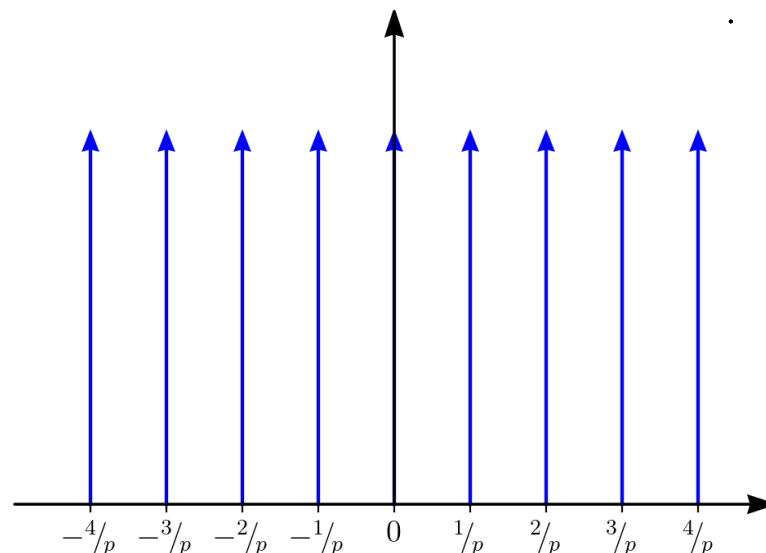
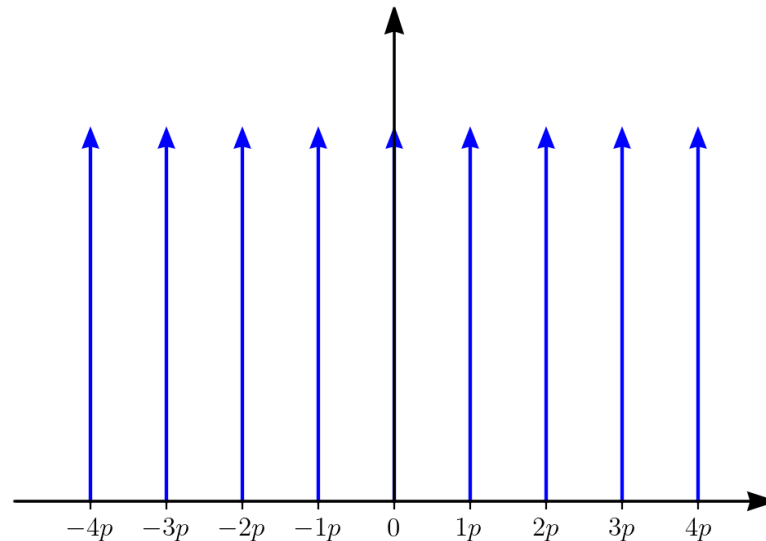
A periodic "function" made of Dirac functions

$$\Delta_p(x) = \sum_{n=-\infty}^{\infty} \delta(x - np)$$

$$\Delta_{\frac{1}{p}}(u) = \sum_{k=-\infty}^{\infty} \delta(u - \frac{k}{p})$$

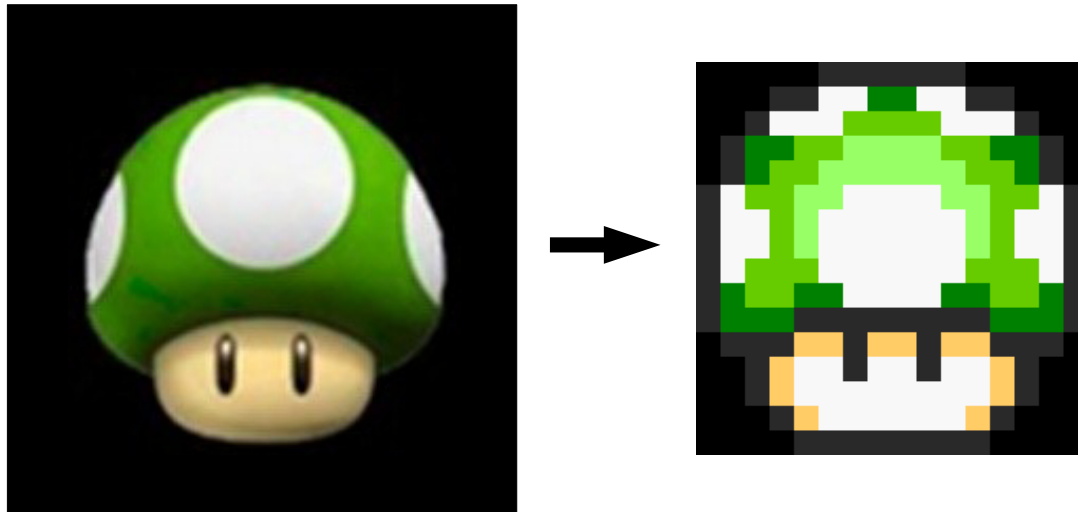
$$\mathcal{F}\{\Delta_p(x)\} = \frac{1}{p} \Delta_{\frac{1}{p}}(u)$$

Fourier transform of a Dirac comb is a Dirac comb

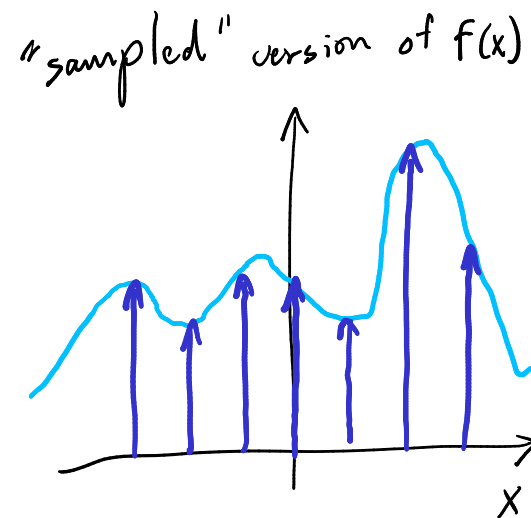
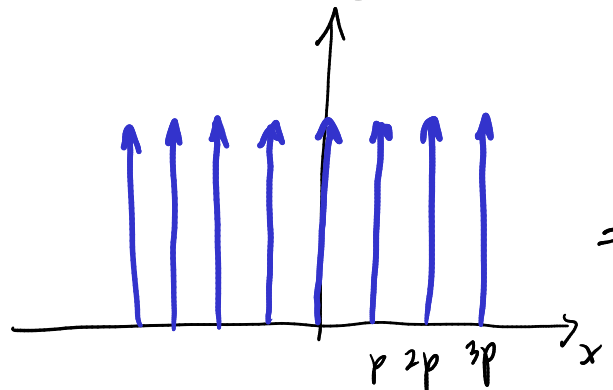
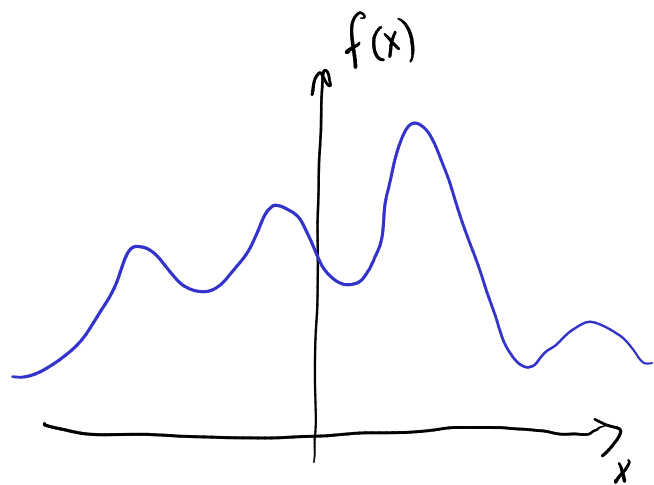


The discrete Fourier transform

- additional ingredients needed:
 - sampling in space
 - finite field of view in space
 - sampling in frequency domain
 - finite frequency band
- discrete approximation of some continuous function



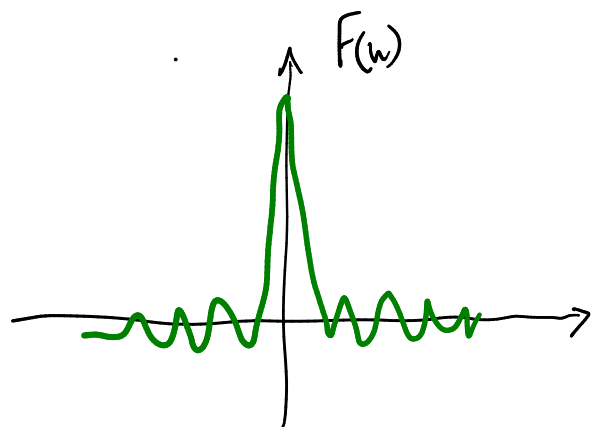
Sampling



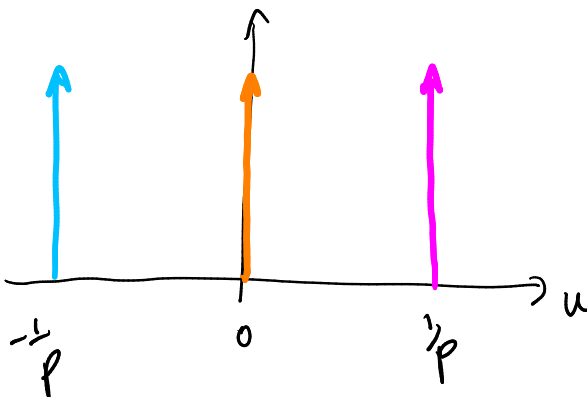
↓

↓ F

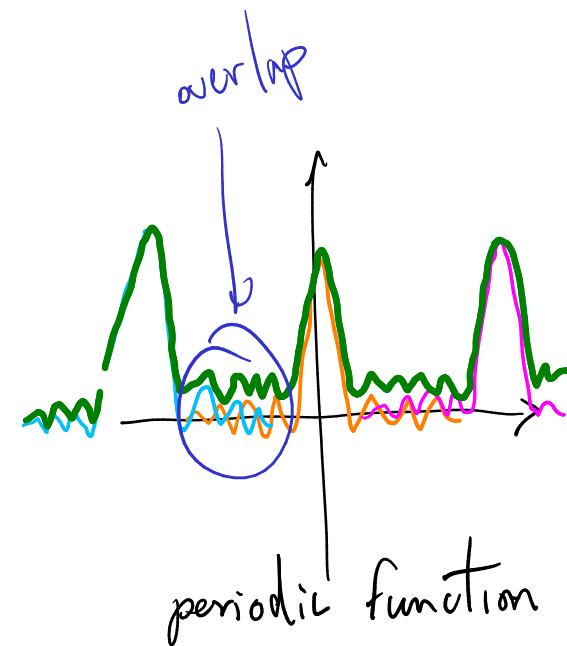
↓



*

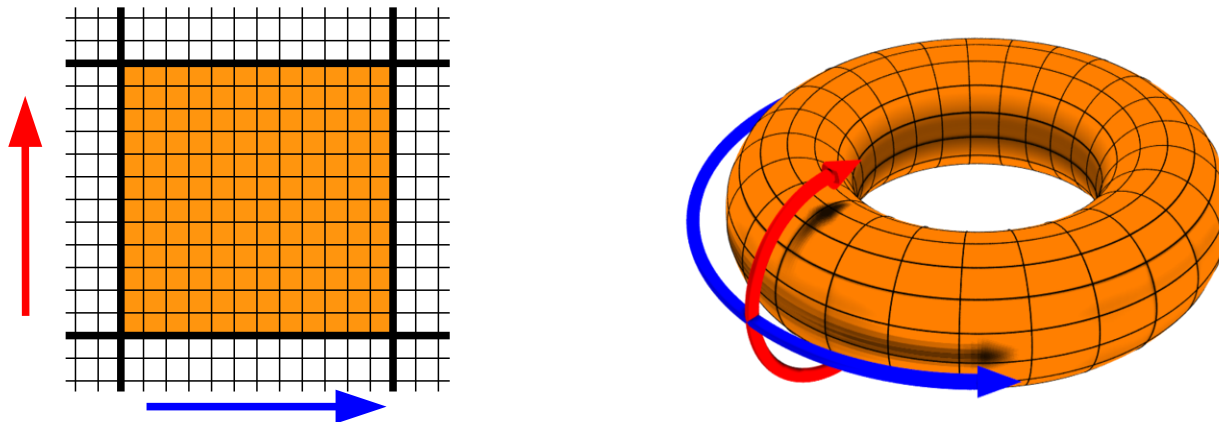


=



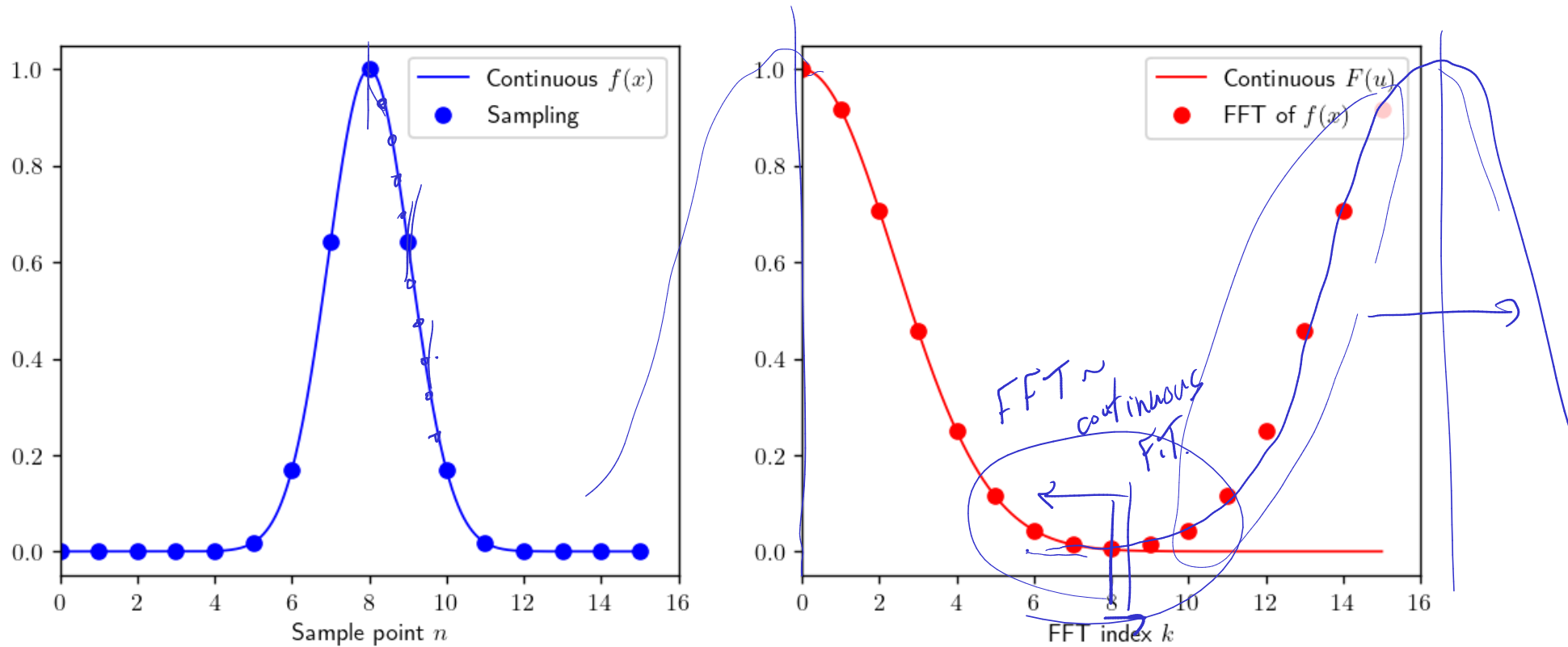
Discrete Fourier Transform

- A **periodic** function has a **discrete** spectrum in the Fourier domain;
 - A function with **discrete** values in the spatial domain is **periodic** in the Fourier domain;
- ⇒ A periodic and discrete function has a periodic and discrete Fourier transform.



DFT example

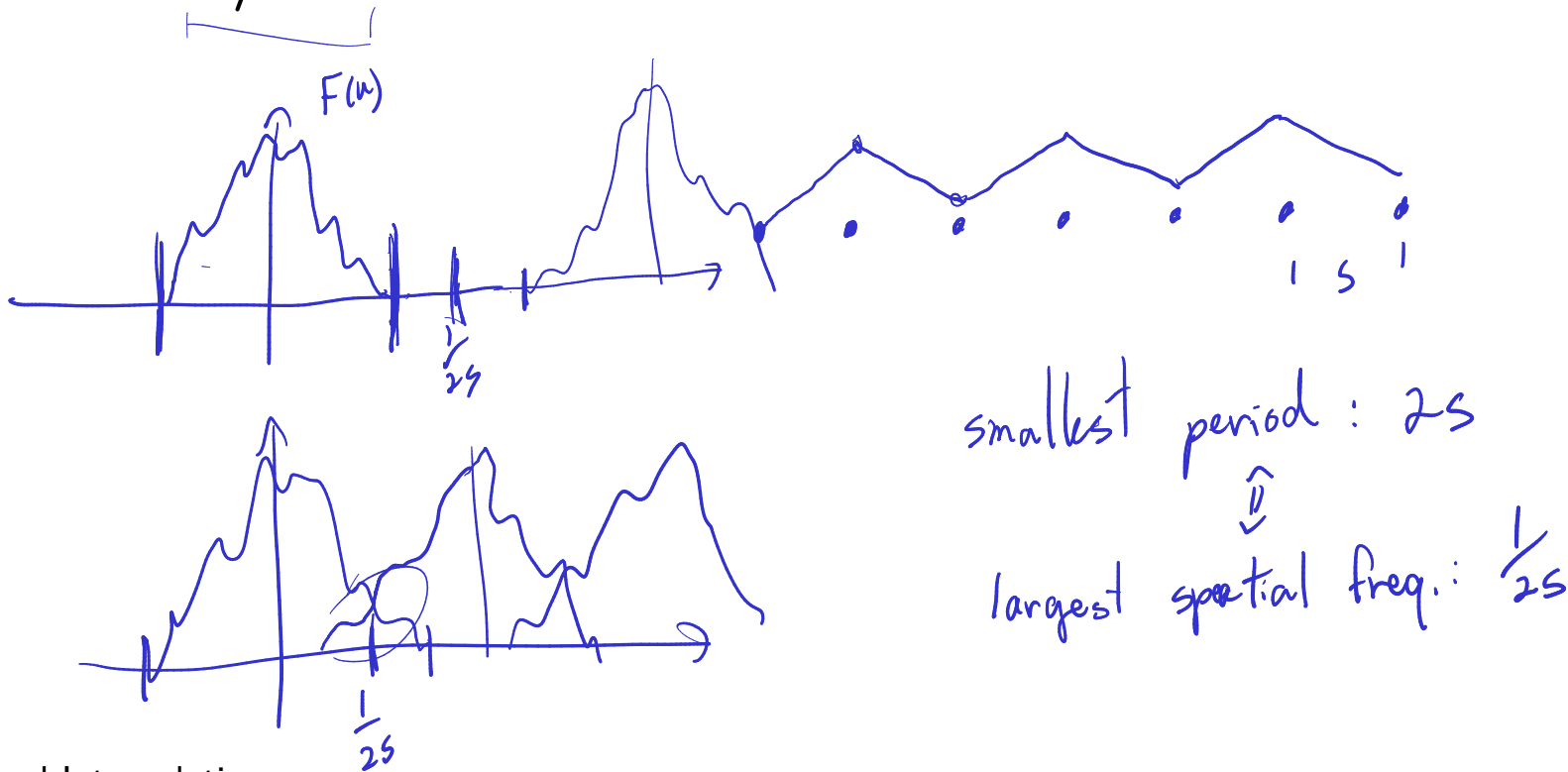
- Example: relation between space, sampling and frequency



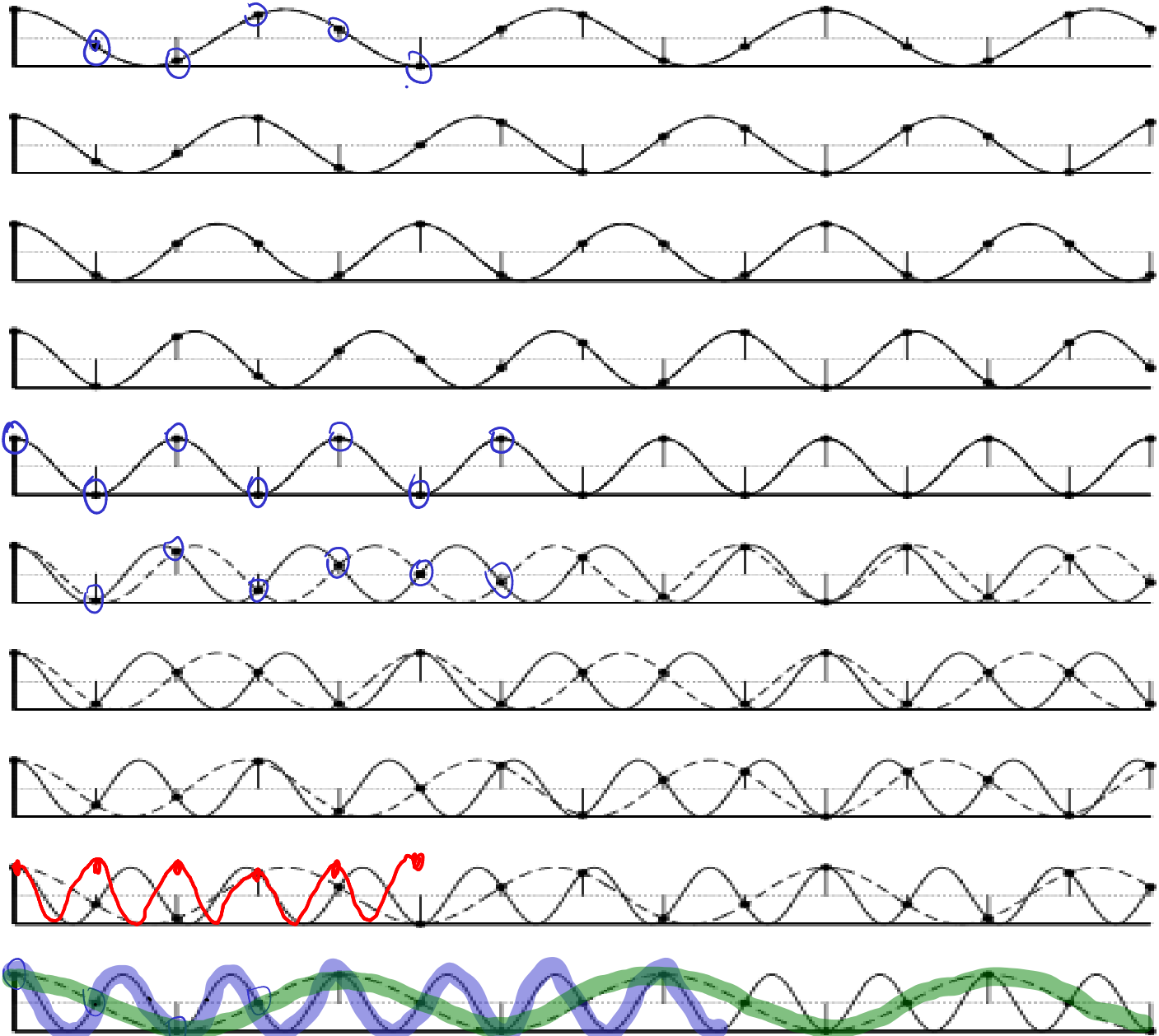
zero frequency component is in the top left corner output array.

The Nyquist-Shannon sampling theorem

“The largest frequency that can be represented in a signal sampled at intervals s is $1/2s$ ”

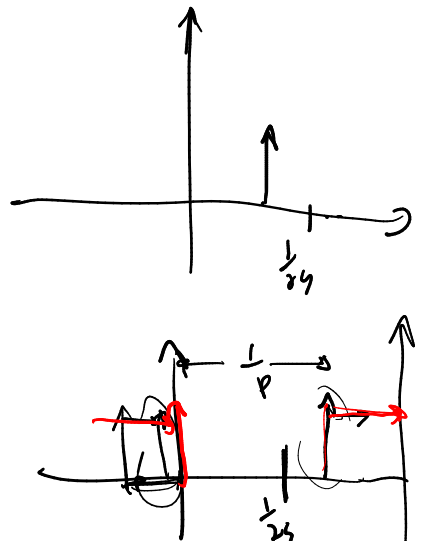


Undersampling and aliasing



Nyquist →

p=2s

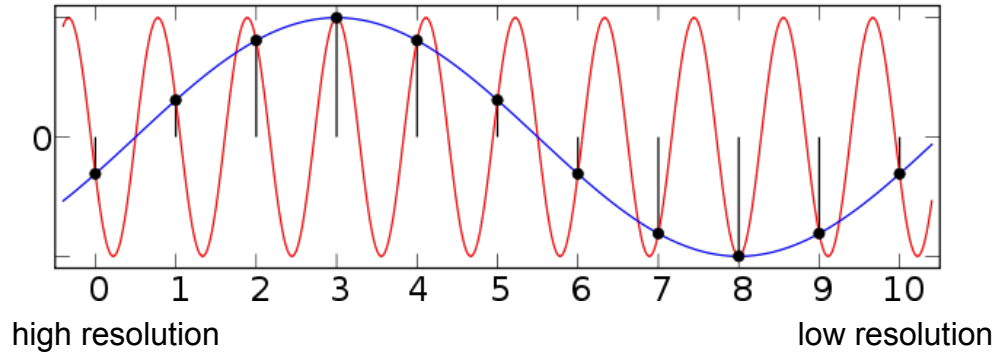


Sampling and Interpolation

"aliasing": higher frequency is mapped onto a lower frequency

Aliasing

Moiré: after resampling, high spatial frequencies appear as low spatial frequencies

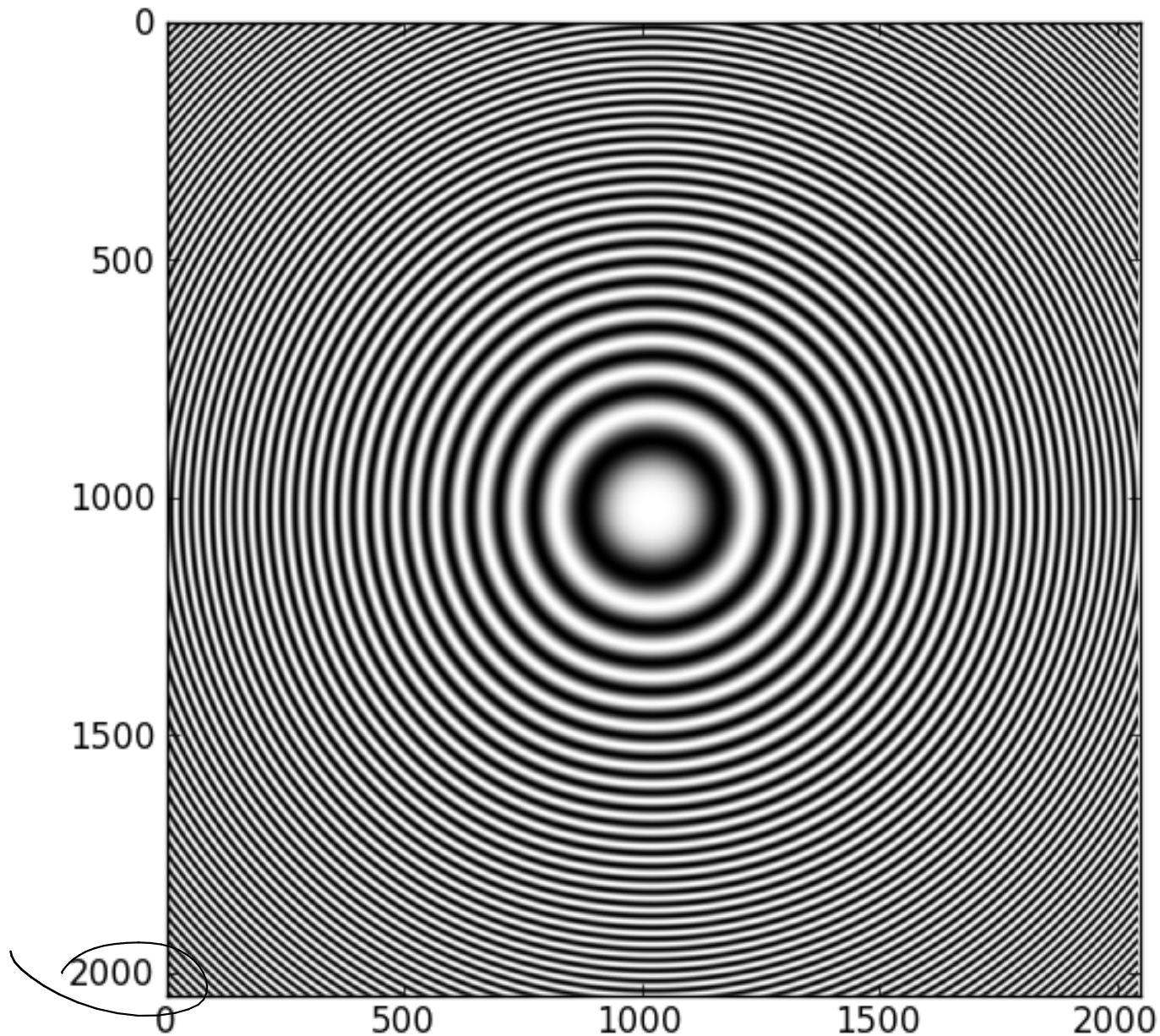


aliasing

source: <http://wikipedia.org>

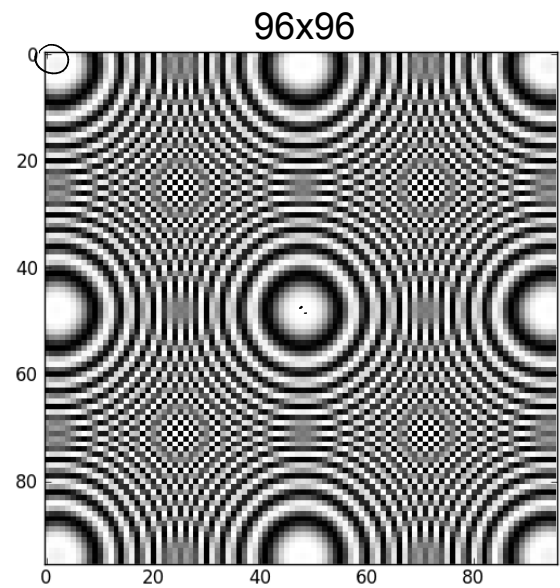
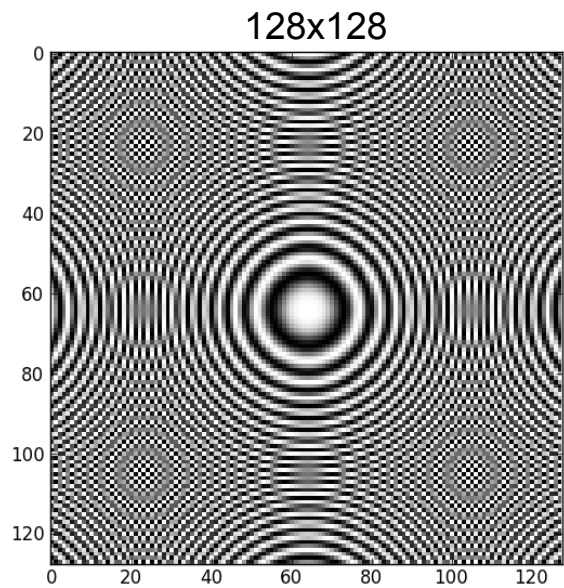
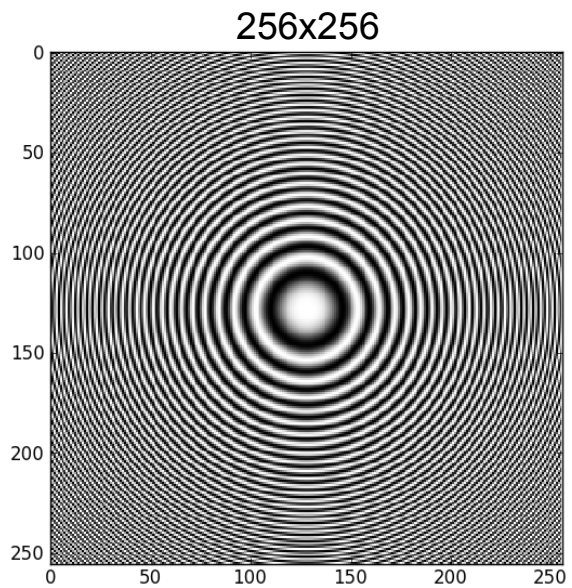
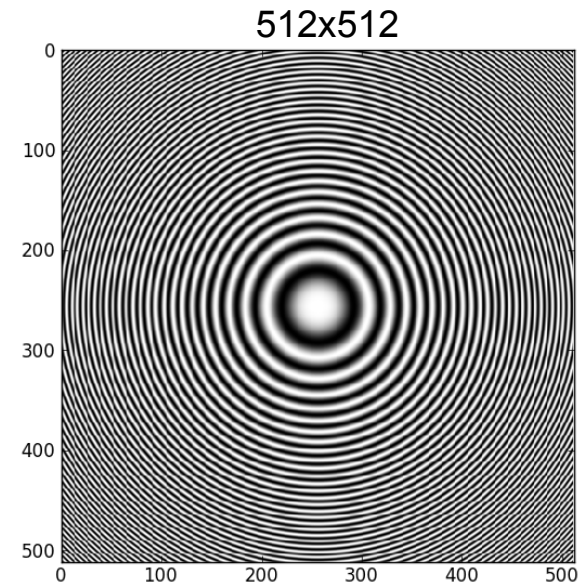
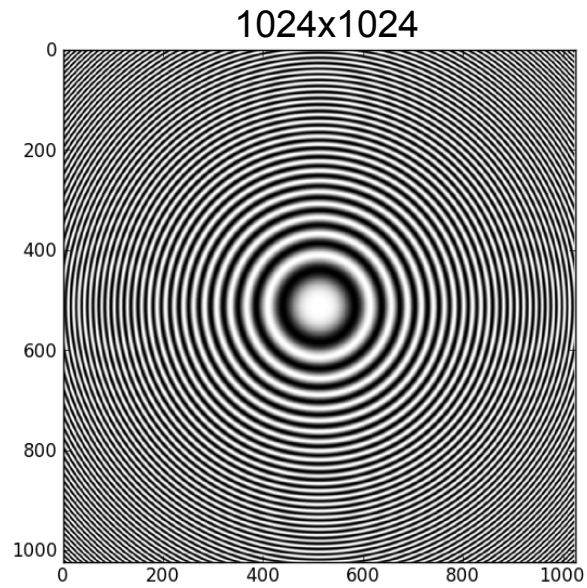
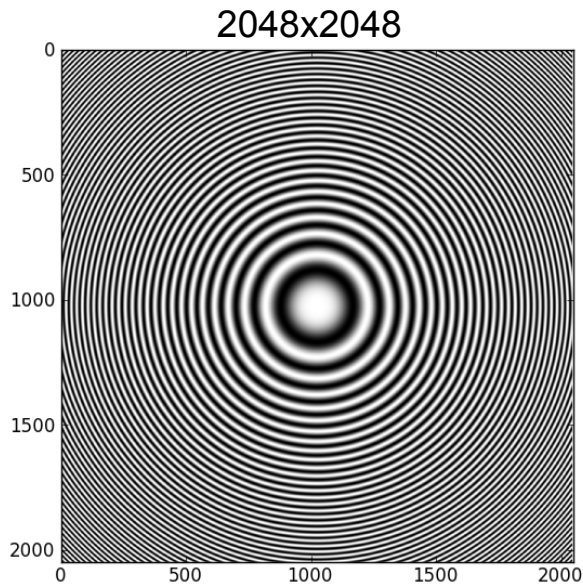
Undersampling

“Fresnel zone” test pattern: radial linear increase in spatial frequency



$\cos(r^2, a)$
 $\cos(r \cdot (ra))$
↑
frequency increases
radially
linearly with r

Undersampling & aliasing

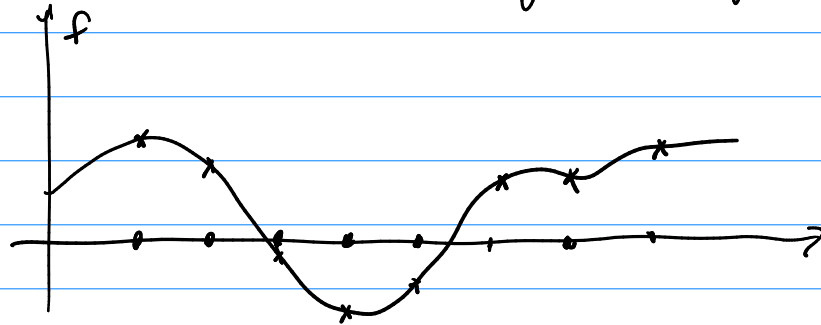


Sampling - summary

* what is sampling? The approximation of a signal (function) with a discrete set of values.

$$f(x) \longrightarrow f_n = f(x_n) \quad \begin{array}{l} \{f_n\} \text{ samples of } f \\ \{x_n\} \text{ sampling positions} \end{array}$$

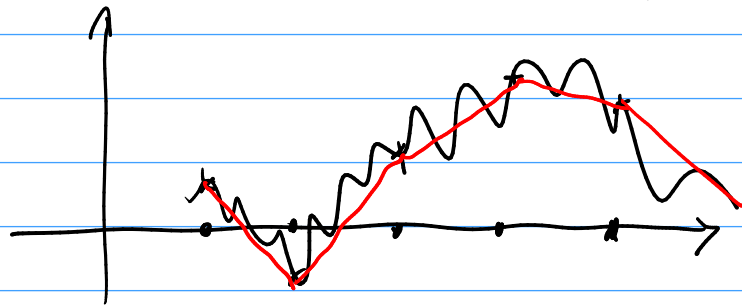
Most often $\{x_n\}$ is on a regular grid.



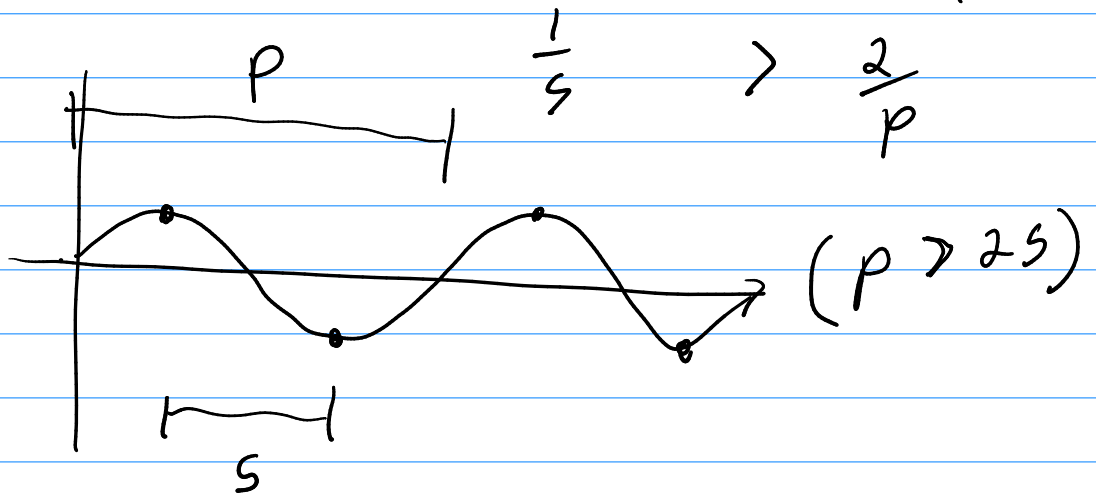
sampling is an unavoidable step in all data acquisition almost

Can any signal be sampled - given a sampling rate?

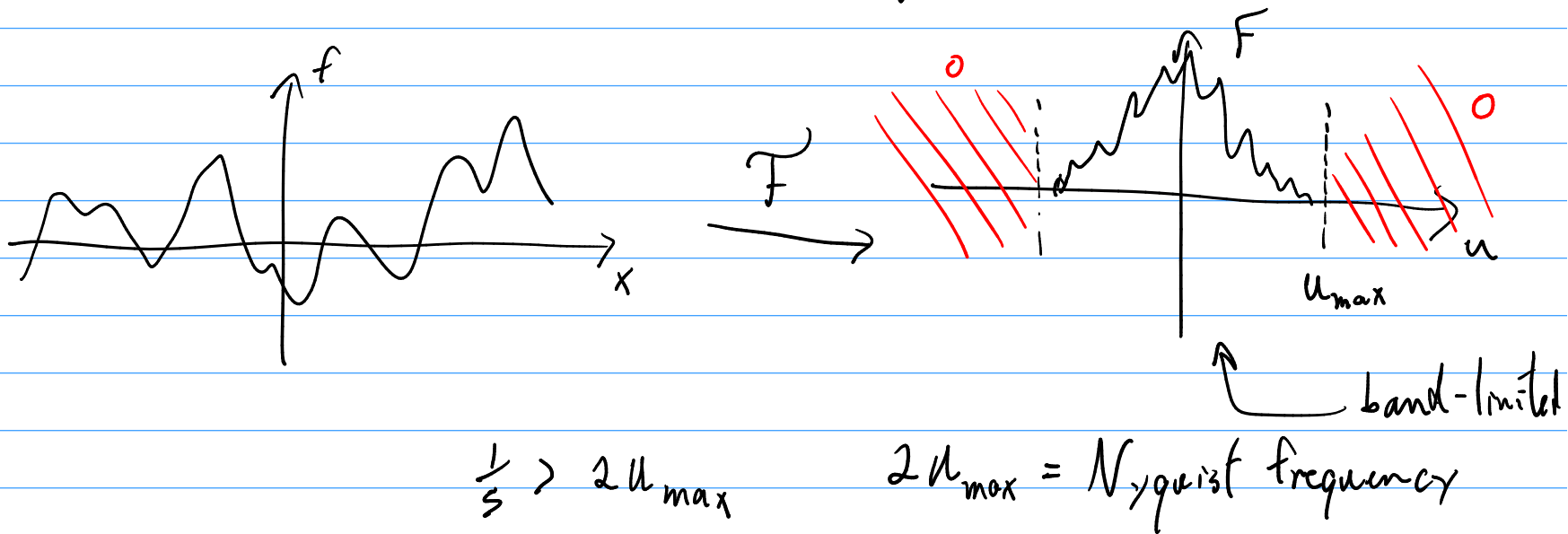
No. Too high frequencies cannot be faithfully reproduced with a sampled signal.



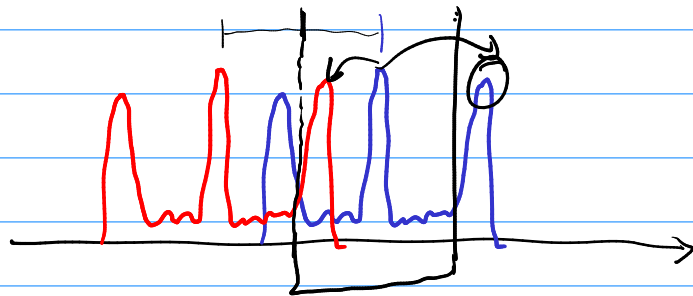
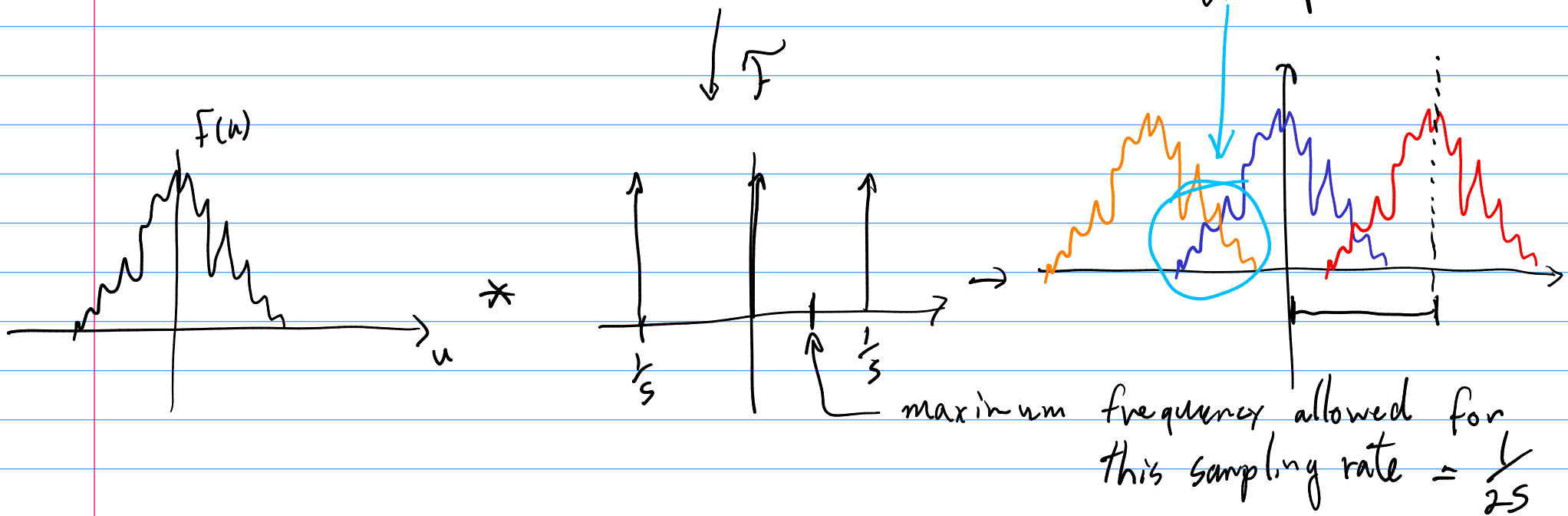
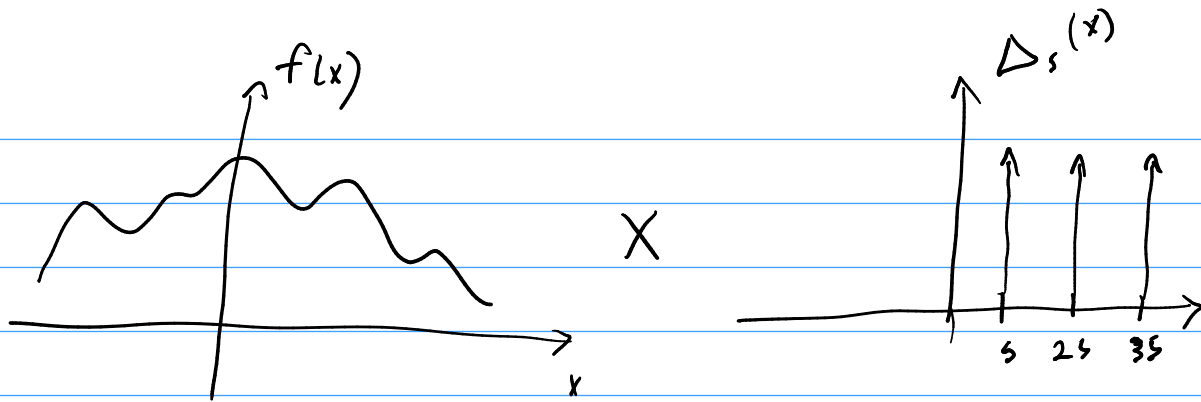
Condition: Nyquist theorem: sampling rate $>$ 2 highest frequency



⇒ The only signals that can be faithfully reproduced by sampling are "band limited": they have a maximum frequency above which the signal is 0.

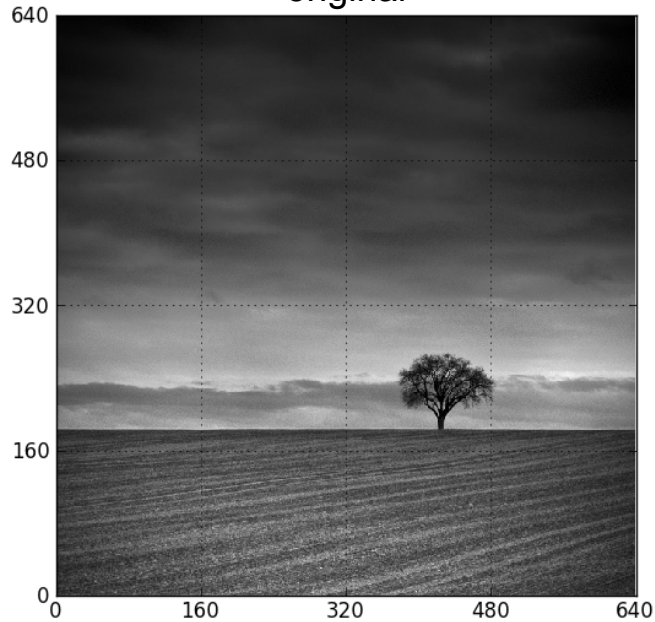


* What happens if $\frac{1}{s} < 2u_{max}$? Undersampling



Fourier space translation

original



amplitude of Fourier spectrum

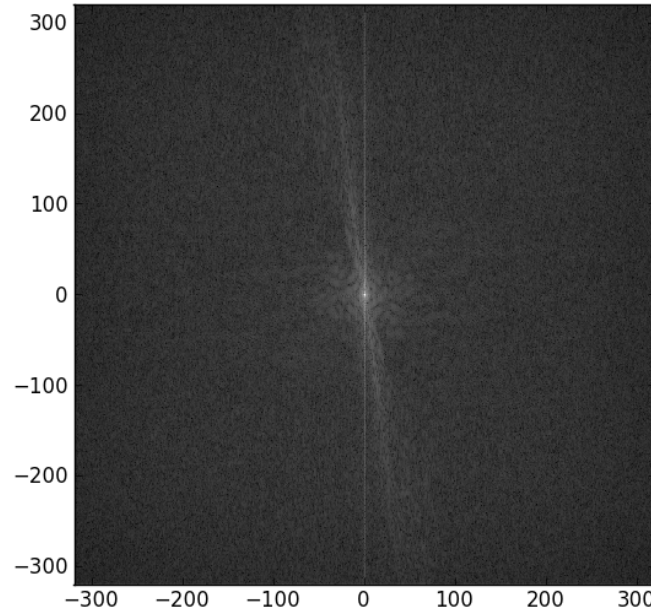


Image shifting using shifting property of FT

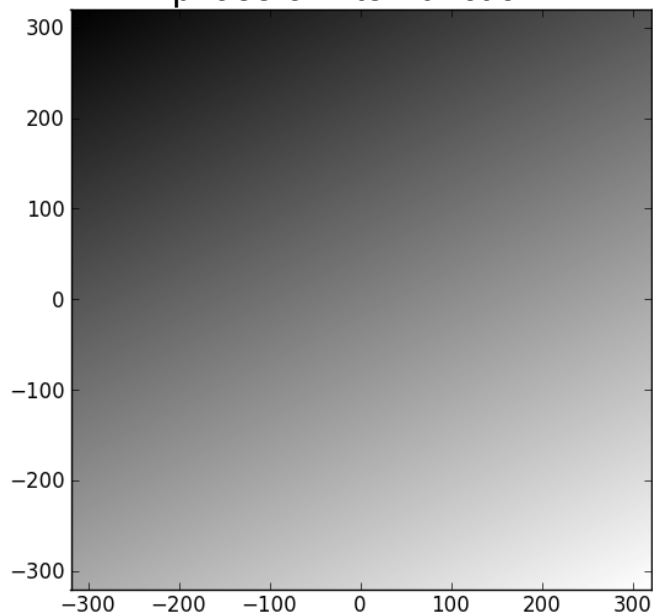
$$f(x+x_0)$$

$$\downarrow \mathcal{F}$$

$$F(u) e^{-2\pi i x_0 u}$$

$$\mathcal{F}\{f(\vec{r}+\vec{r}_0)\} = F(\vec{u}) e^{-2\pi i \vec{r}_0 \cdot \vec{u}}$$

phase of filter function



shifted image

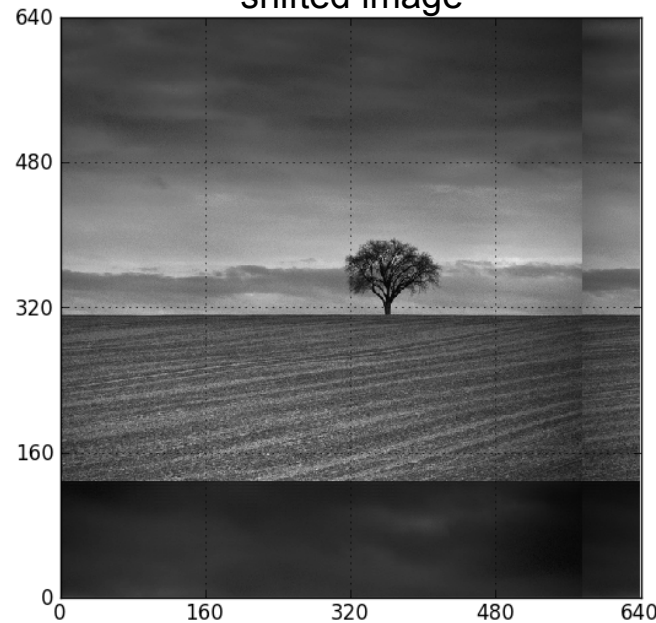
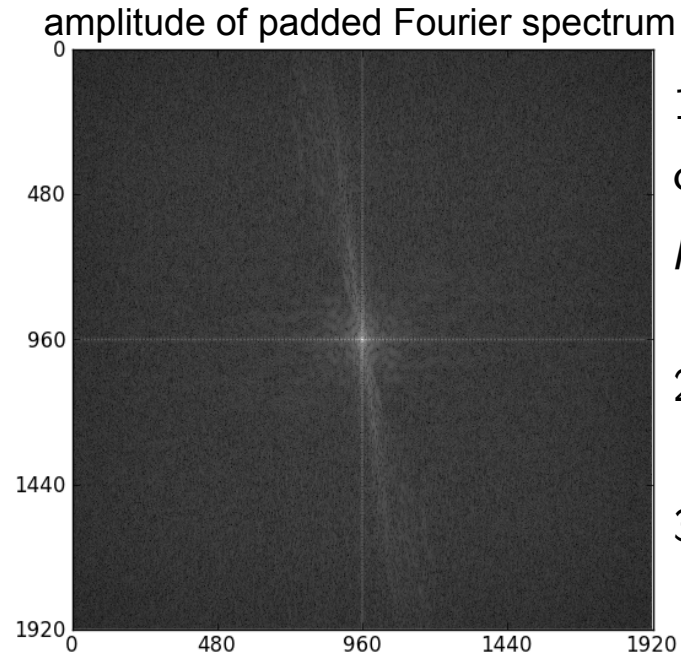
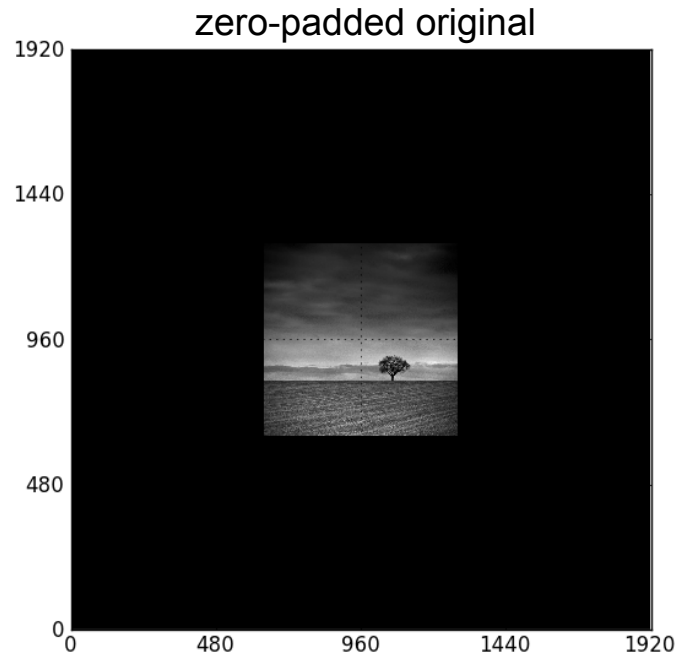


Image gets wrapped around

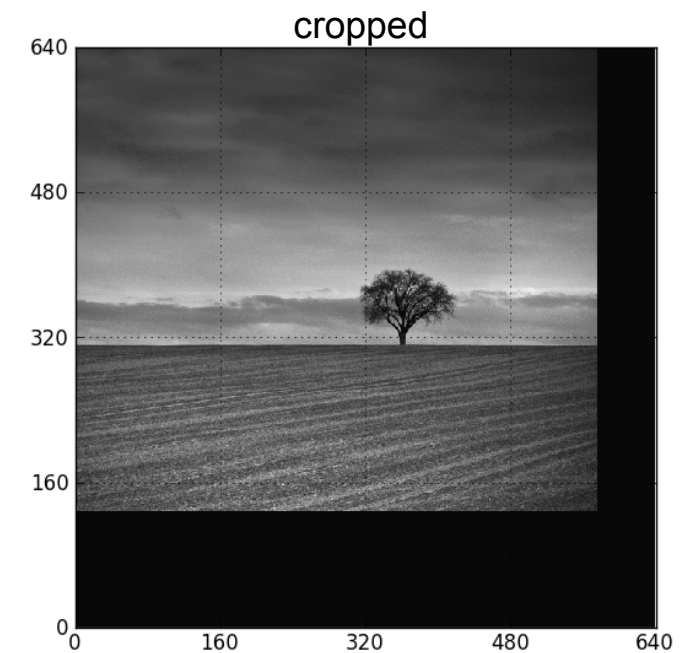
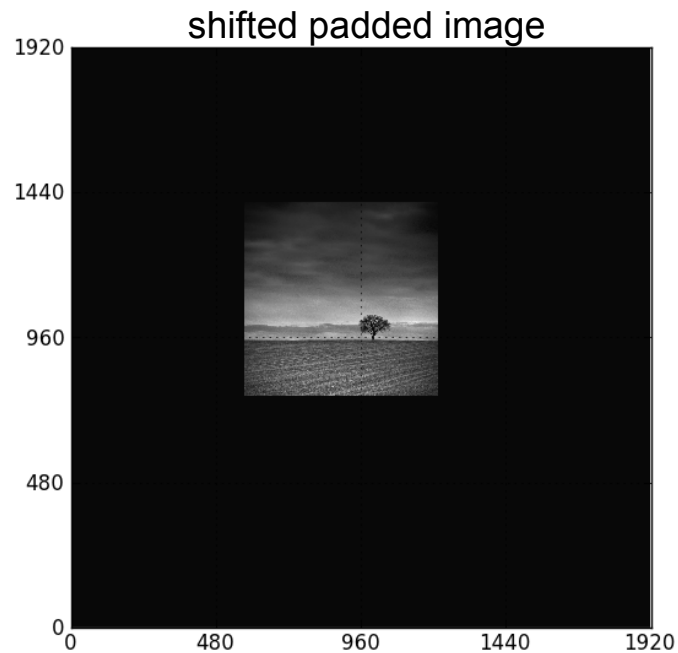
Zero-padding



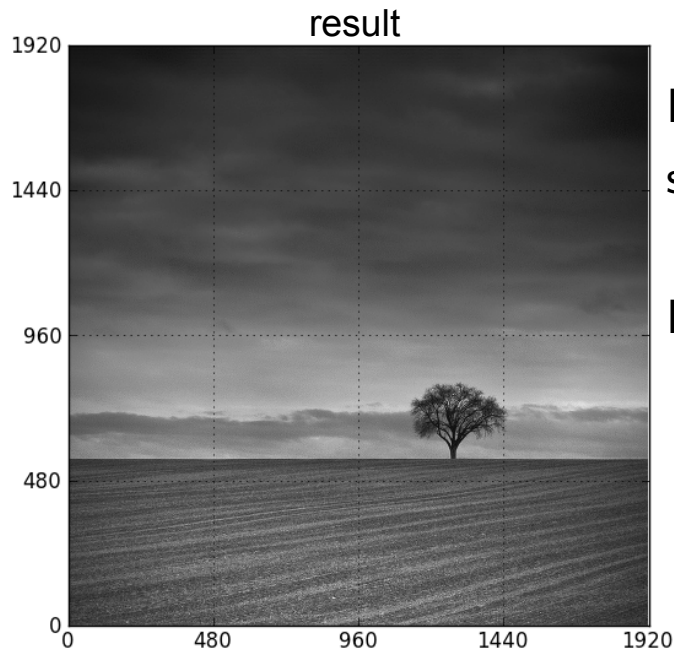
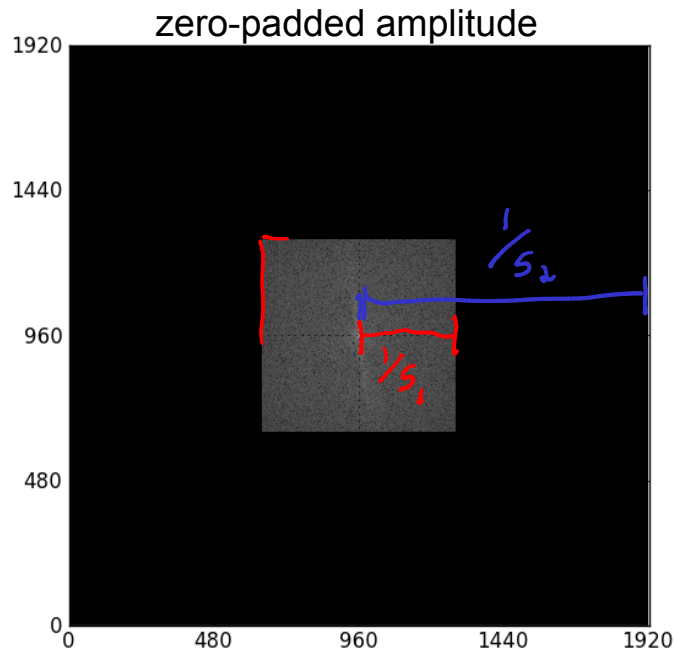
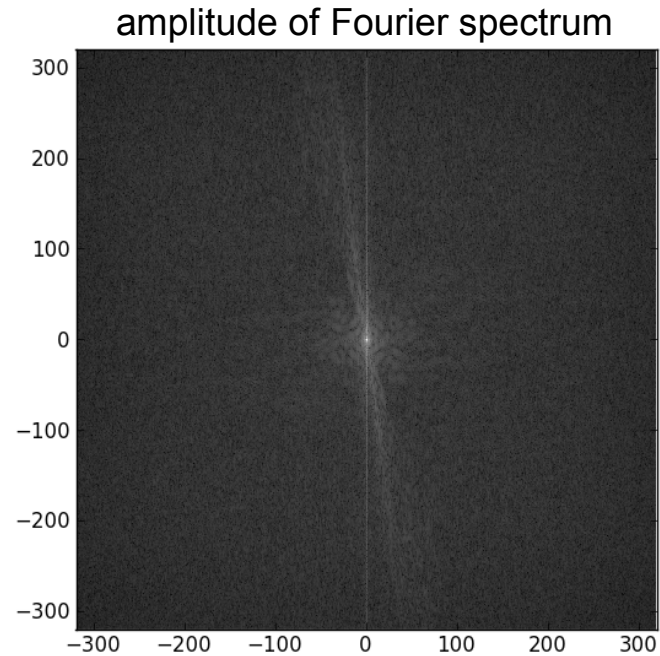
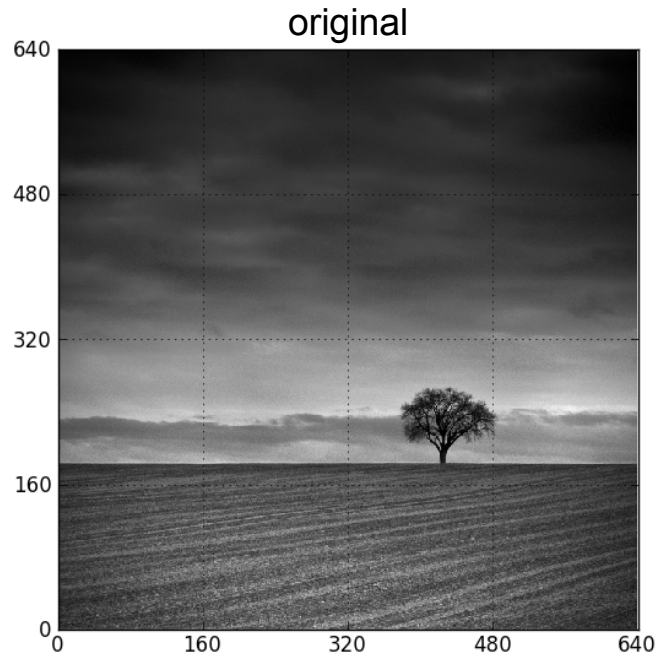
1. Add zeros around original image (*zero-padding*)

2. Shift using FT

3. Crop result



Zero-padding in Fourier space



Result: increased sampling!

Is it magic? ← NO.
this is interpolation

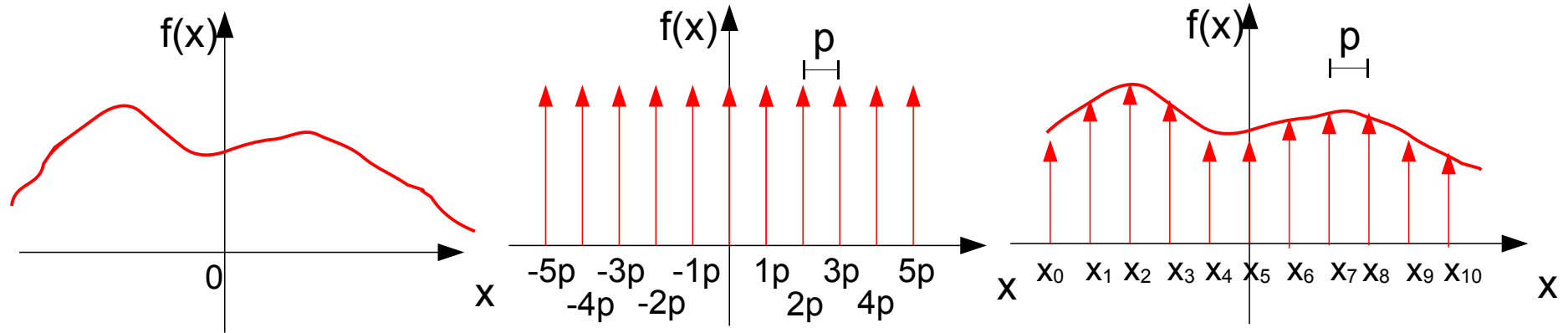
$$\frac{1}{s_2} = 3 \frac{1}{s_1}$$

$$s_2 = \frac{s_1}{3}$$

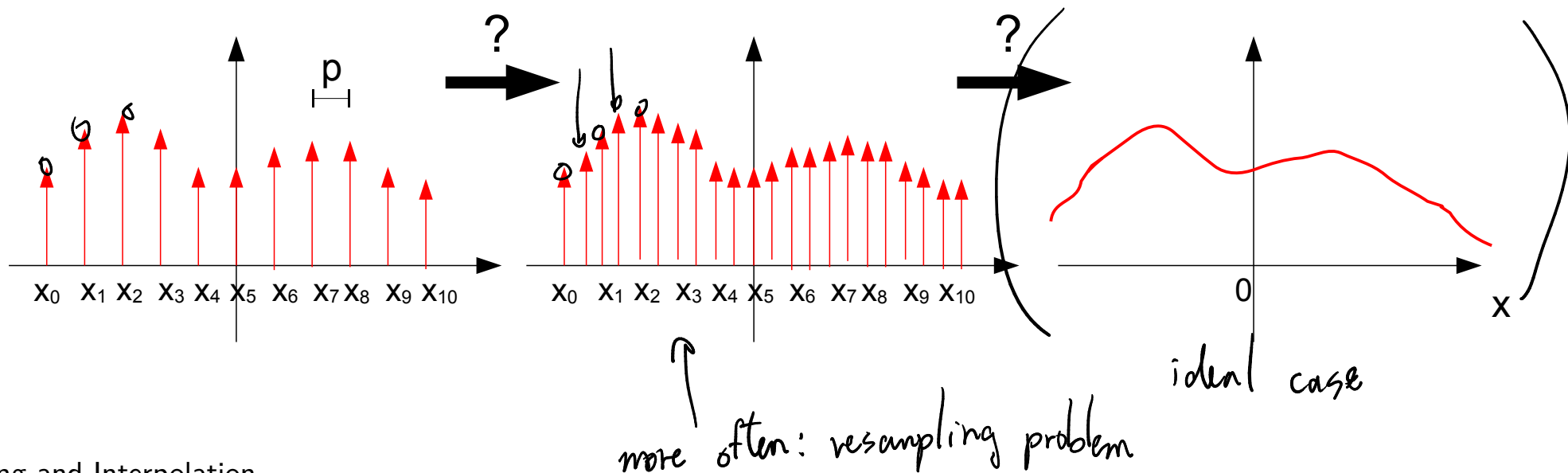
← pixels are $\frac{1}{3}$ of original

Interpolation

- Discrete sampling of a continuous function



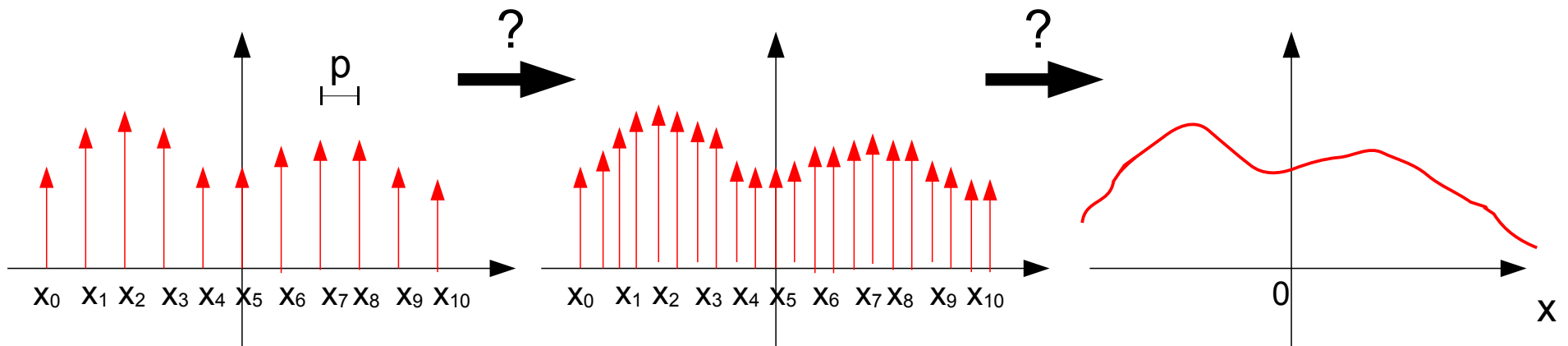
- Reconstruct original function from sampled data?



Interpolation

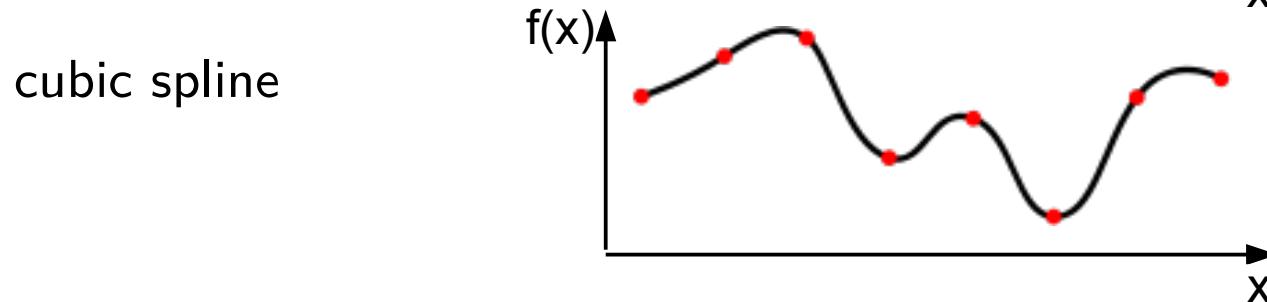
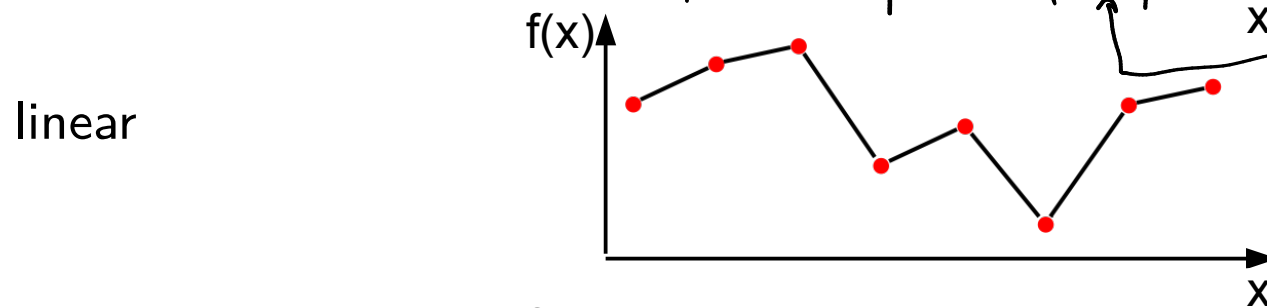
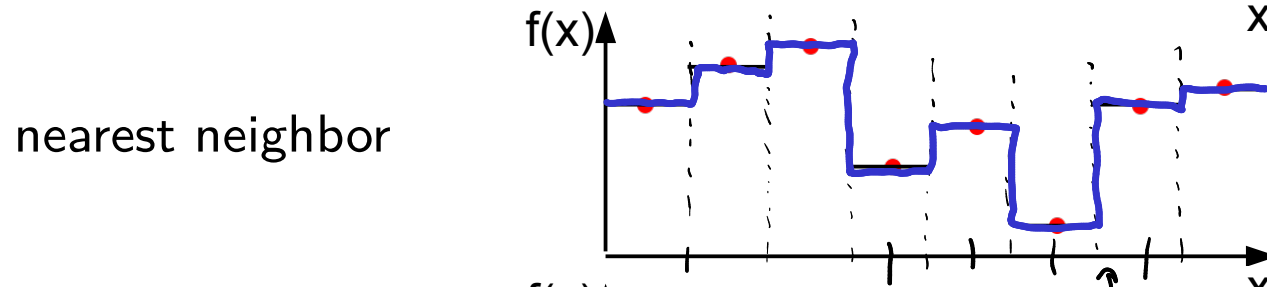
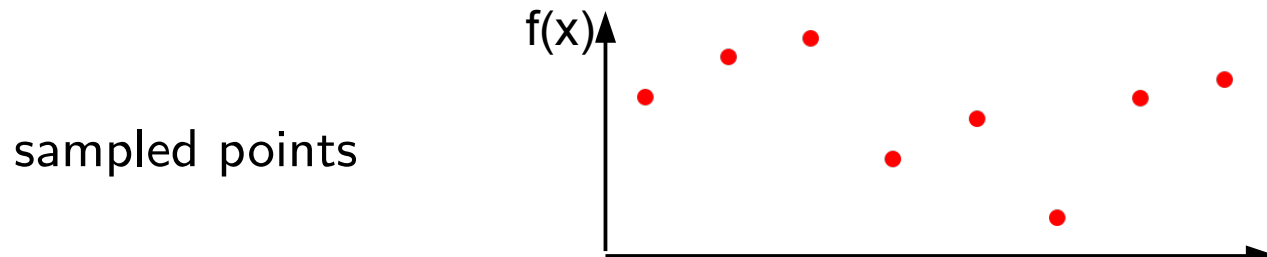
Finding unknown points between known ones

- wide field, many different approaches
- closely related to approximation theory and curve fitting



Interpolation

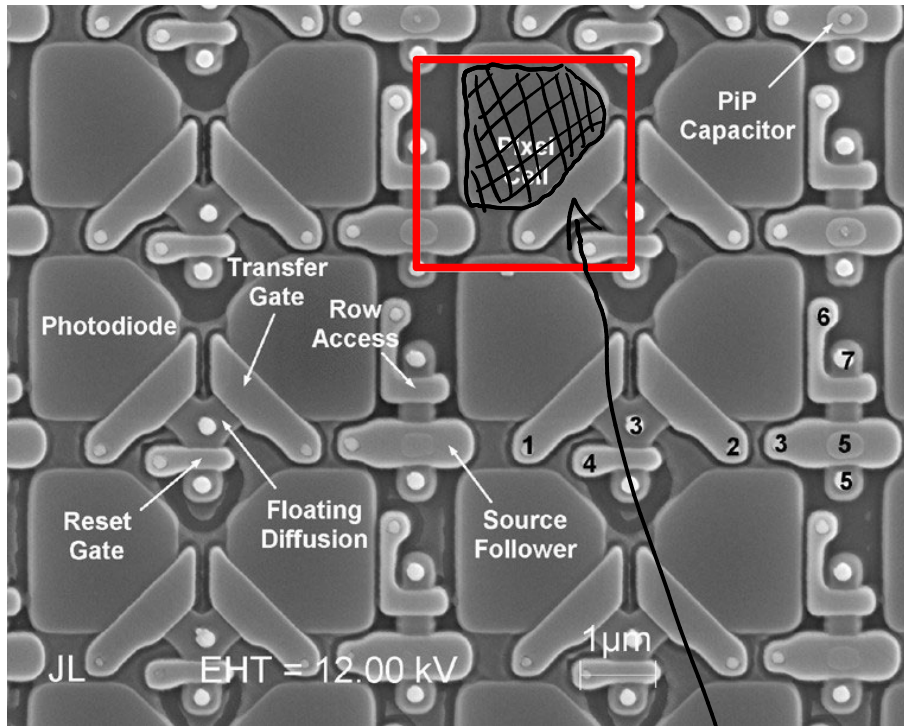
Various “classical” interpolation methods available



*sharp transitions midway
between samples*

Pixels

- distinguish between detector pixels, image pixels and screen pixels
 - detector pixels are rarely square
 - image pixels are commonly, but not necessarily square
 - screen pixels are rarely square



Detector pixel

region
sensitive
to light

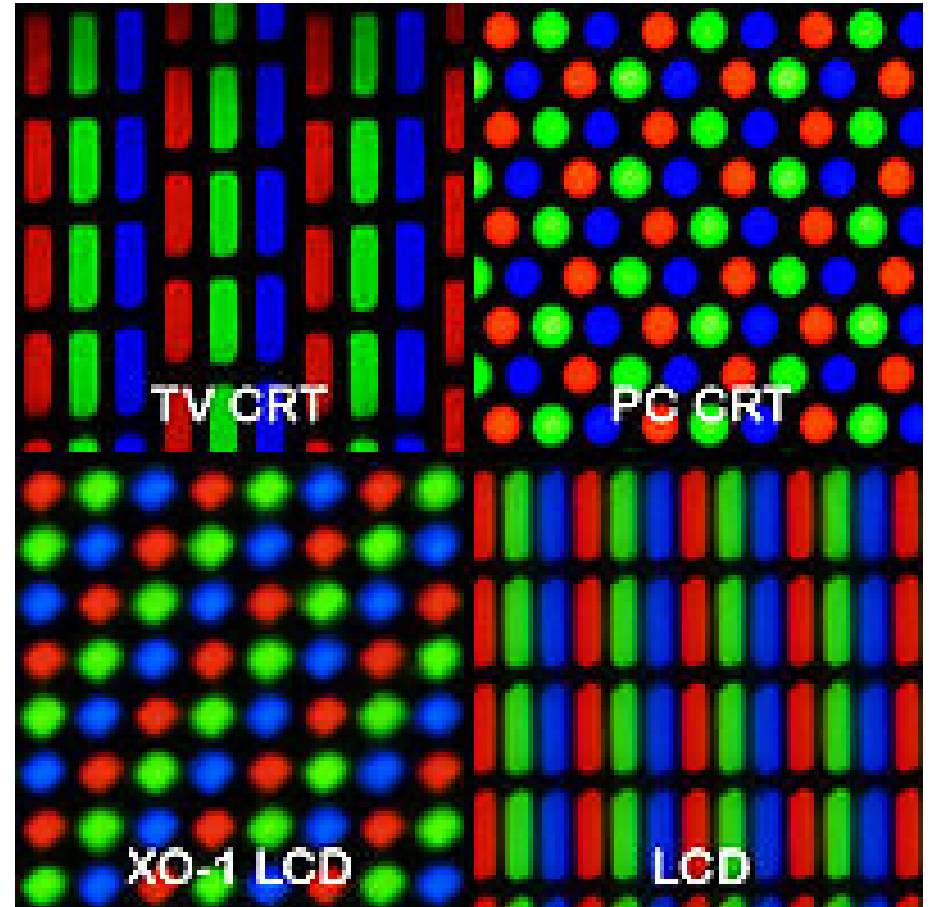
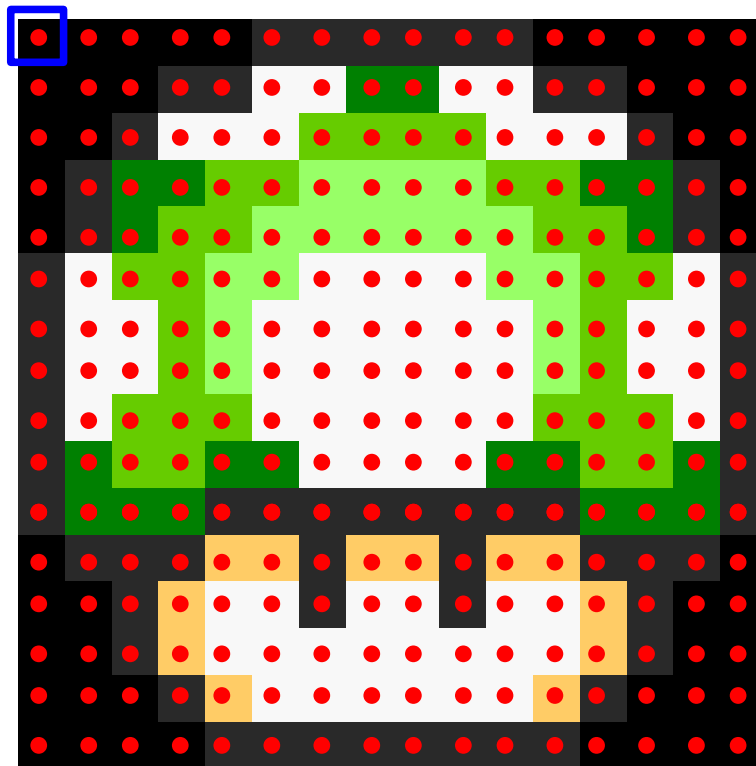


Image pixels

Images are discrete samples of a continuous function

- ...with coordinates
- ...and values (voltage at coordinate, integral over pixel area, ...)
- ...represented by pixel basis functions on a sampling grid



Linear interpolation

- Interpolation as an operator

$$f(x) = \mathcal{L}\{f_n\}$$

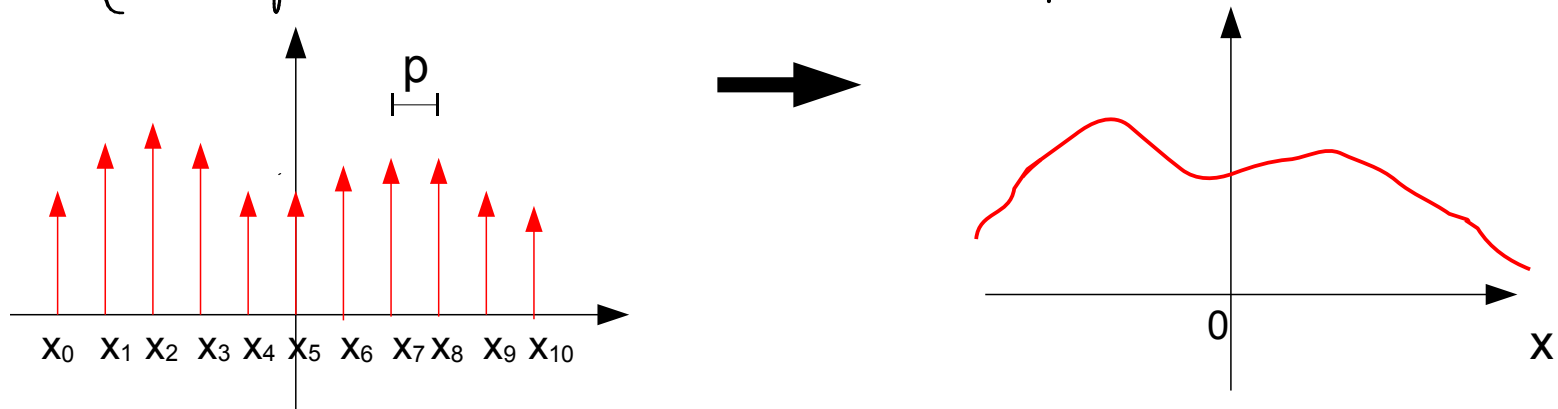
- Linear ^{ity of} interpolation operator

$$\mathcal{L}\{f_n + g_n\} = \mathcal{L}\{f_n\} + \mathcal{L}\{g_n\}$$

- Shift invariance

$$\mathcal{L}\{f_{n+n_0}\} = f(x + n_0 s)$$

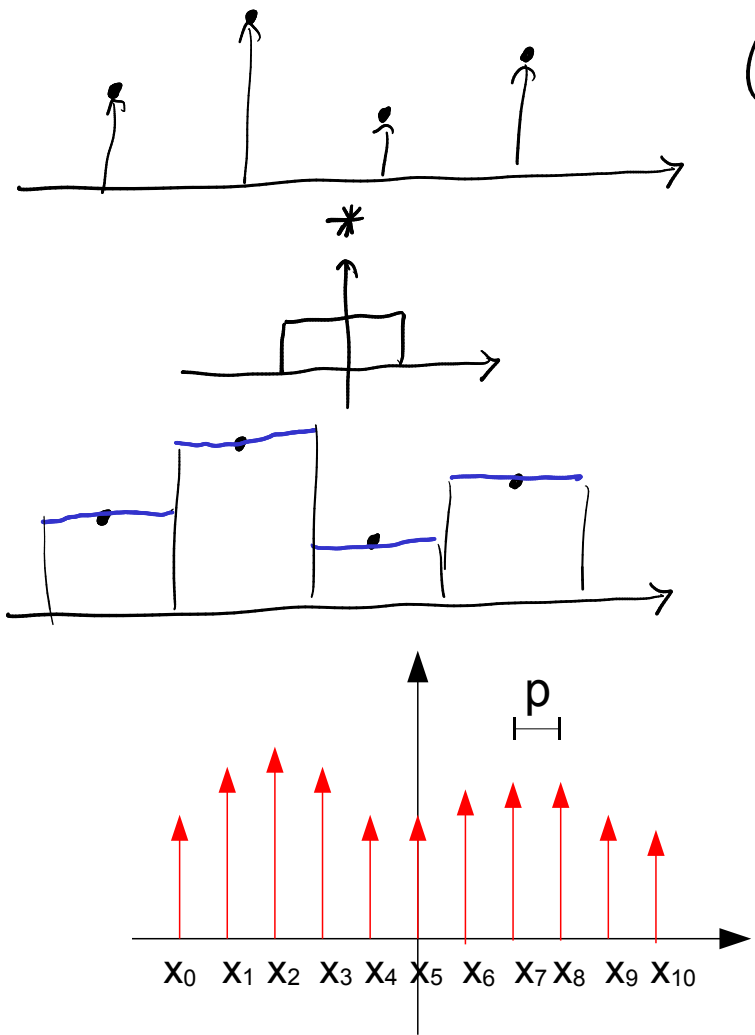
- Kernel (interpolation seen as a convolution)



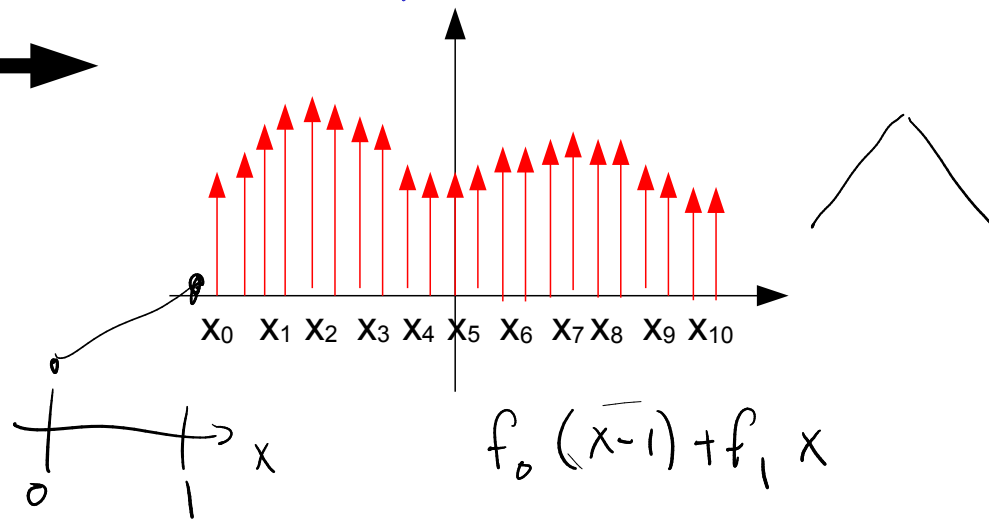
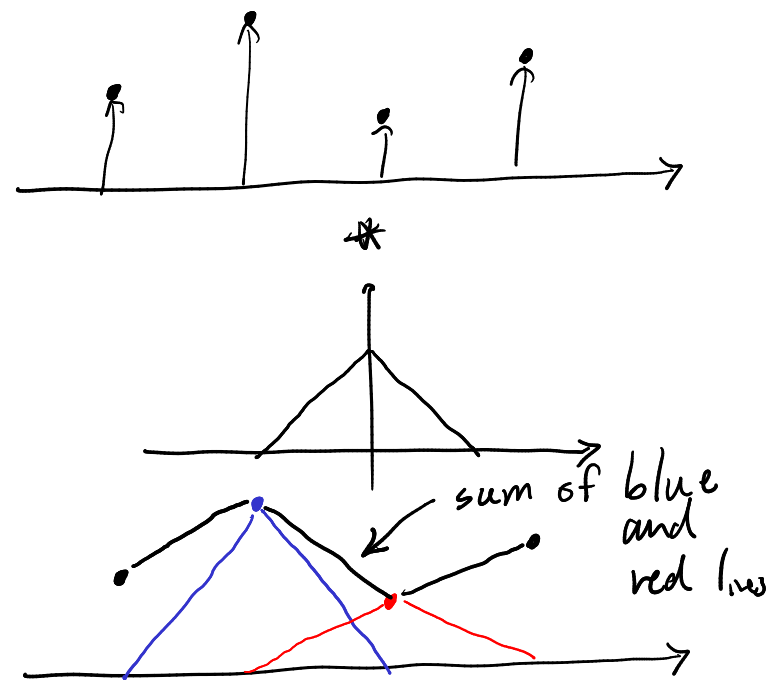
Linear interpolation

- Linear interpolation can be written as a convolution with a kernel (e.g. a basis function)

nearest neighbor = convolution with box function ("rect" function in 1D)

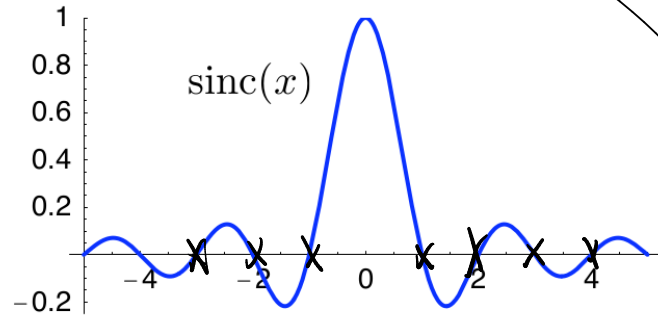


linear interpolation



Linear interpolation

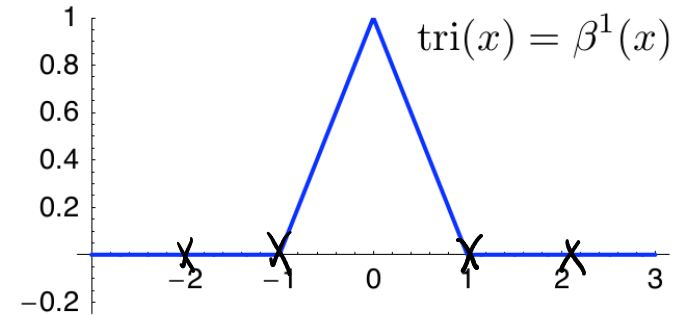
■ Bandlimited



Interpolation condition:

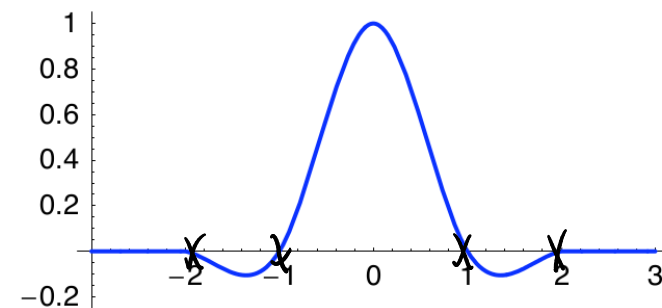
$$\varphi_{\text{int}}(k) = \delta_k = \begin{cases} 1, & k = 0 \\ 0, & \text{otherwise} \end{cases}$$

■ Piecewise linear



tri x tri

■ Cubic convolution



[Keys, 1981; Karup-King 1899]

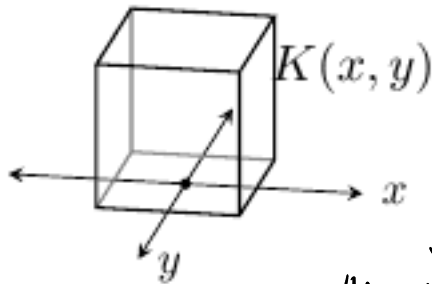
source: http://bigwww.epfl.ch/tutorials/unser_isbi_06_part1

Interpolation via convolution

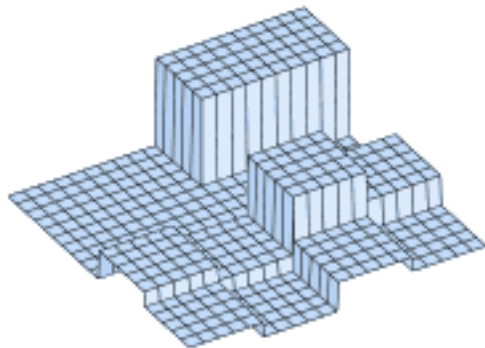
2D interpolation

- Make 2D interpolation linear in each variable

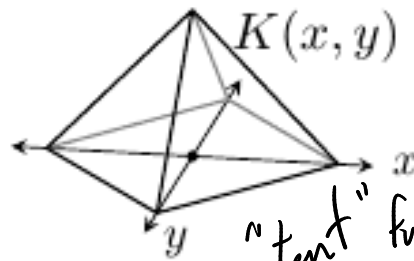
nearest neighbor



"box" function

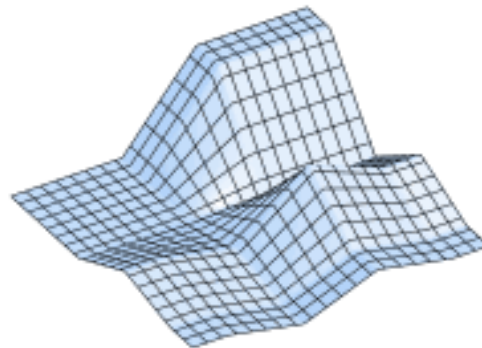


bilinear

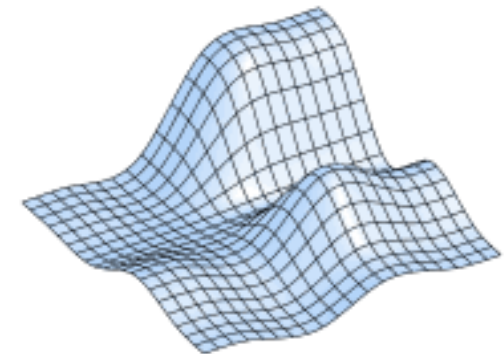
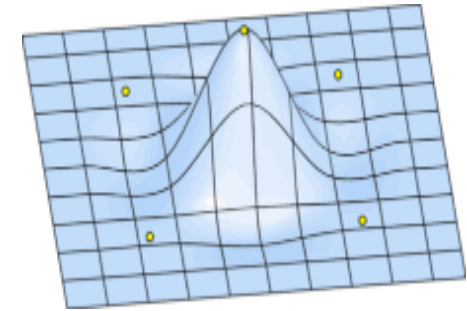


"tent" function

$$= K(x) K(y)$$



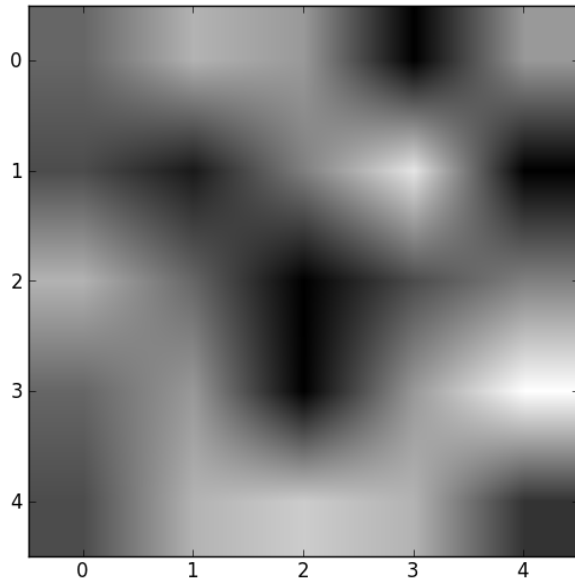
bicubic



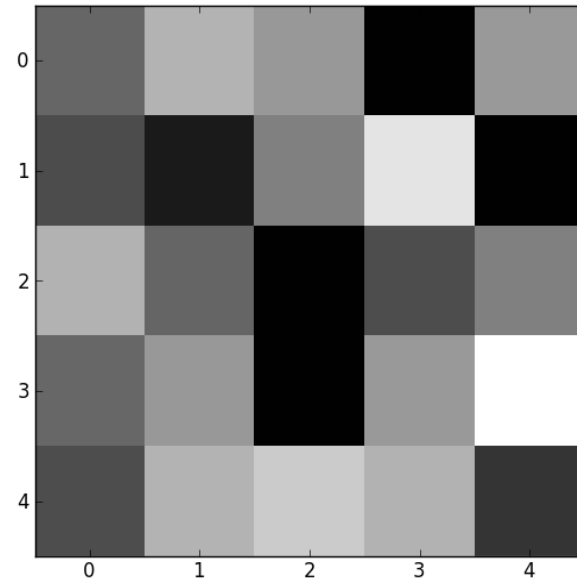
source: http://www.ipol.im/pub/art/2011/g_lmii/

Python plotting

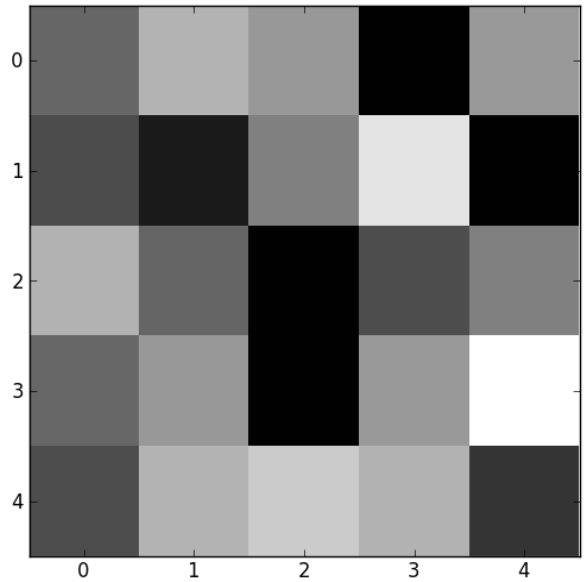
`plt.imshow(im)`



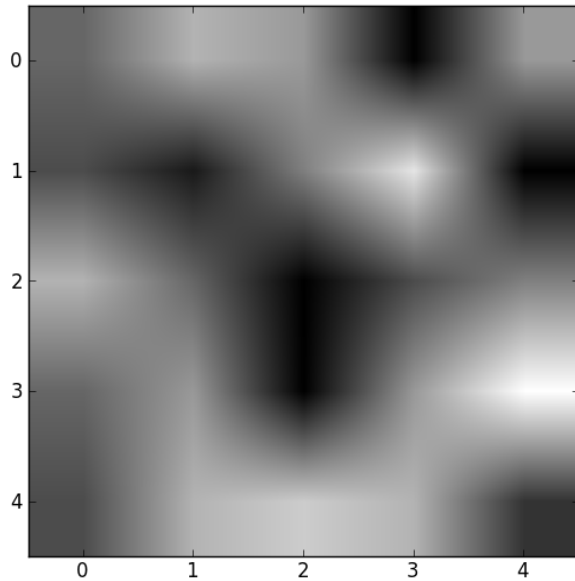
`plt.imshow(im, interpolation='none')`



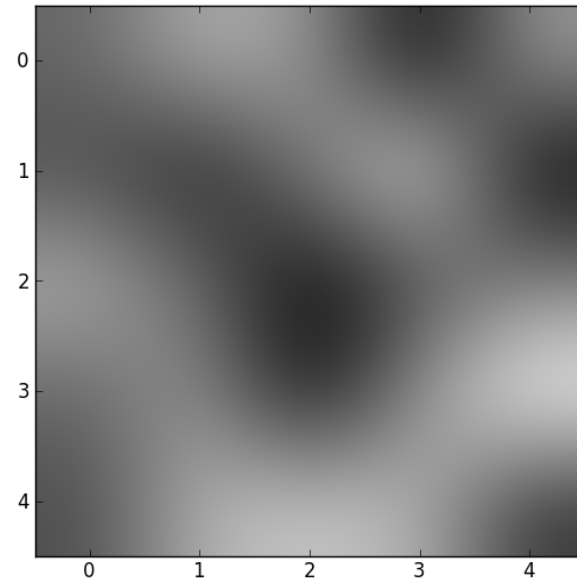
`plt.imshow(im, interpolation='nearest')`



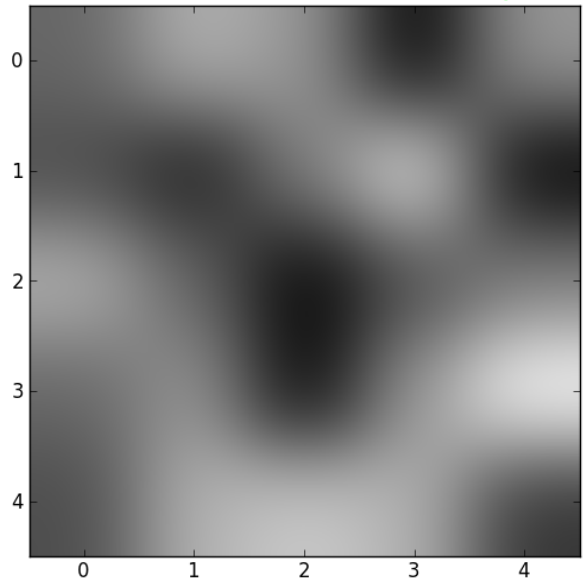
`plt.imshow(im, interpolation='bilinear')`



`plt.imshow(im, interpolation='bicubic')`

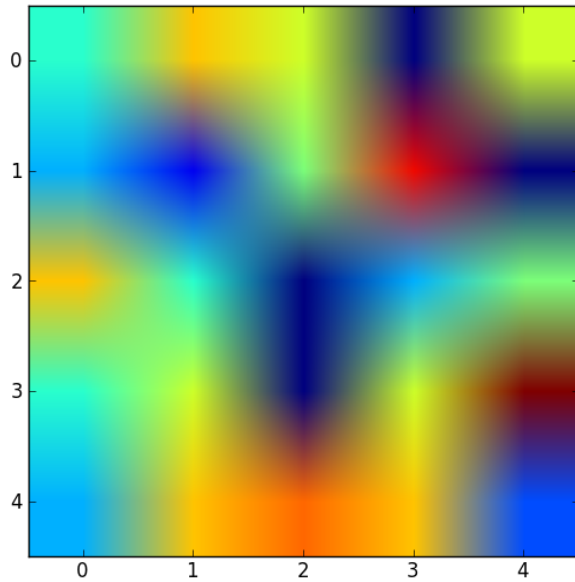


`plt.imshow(im, interpolation='gaussian')`

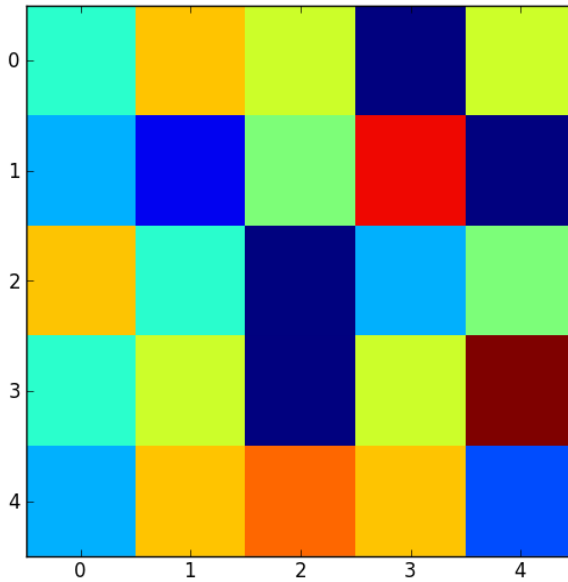


Python plotting

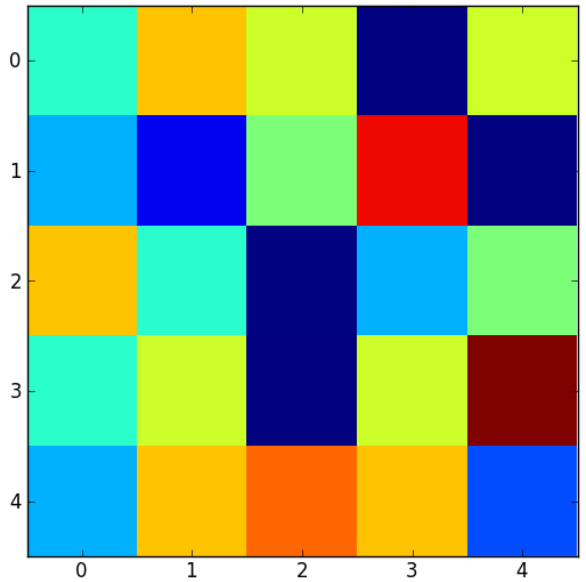
`plt.imshow(im)`



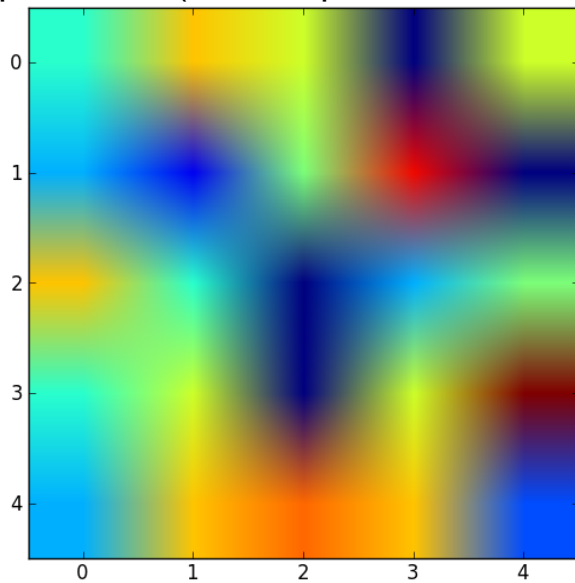
`plt.imshow(im, interpolation='none')`



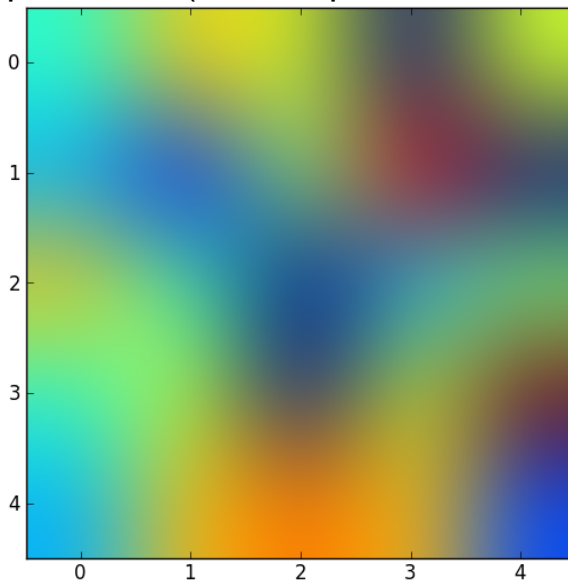
`plt.imshow(im, interpolation='nearest')`



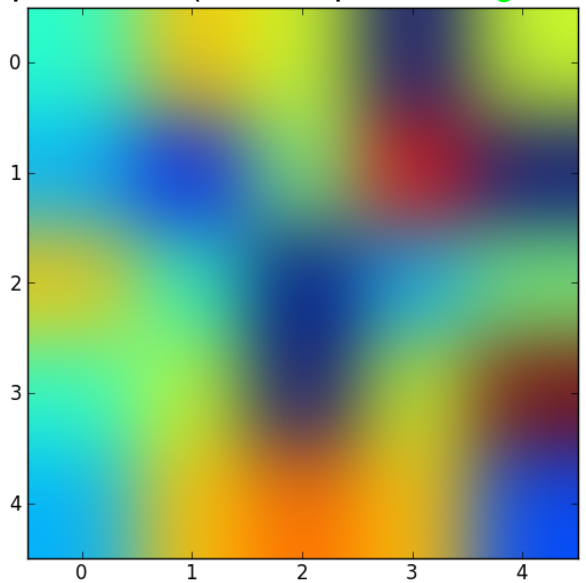
`plt.imshow(im, interpolation='bilinear')`



`plt.imshow(im, interpolation='bicubic')`

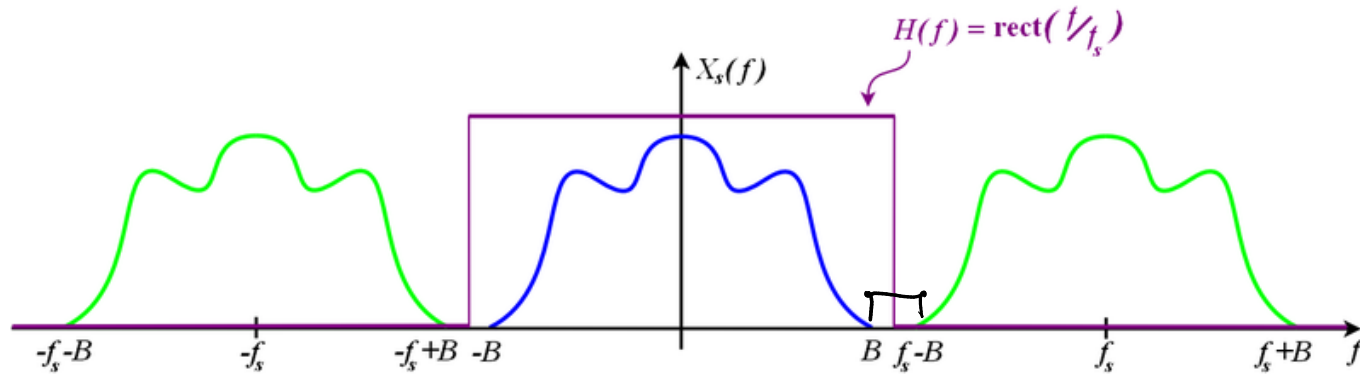


`plt.imshow(im, interpolation='gaussian')`

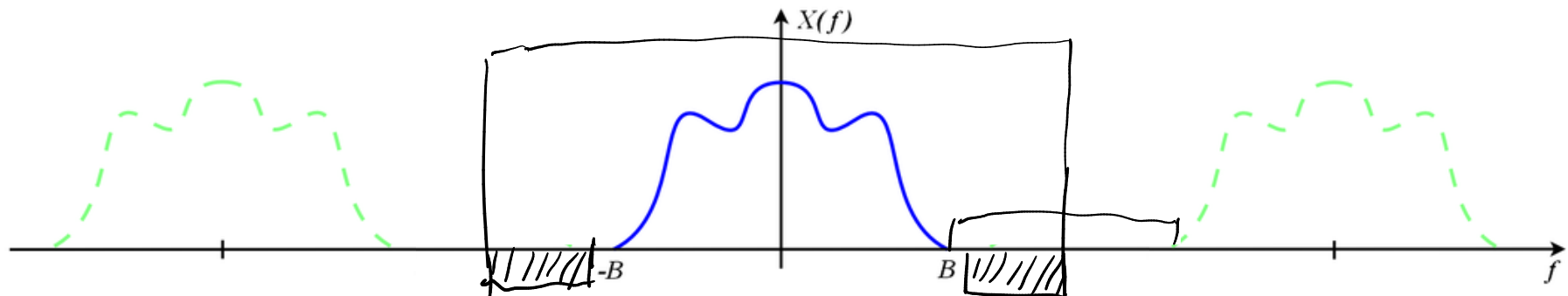
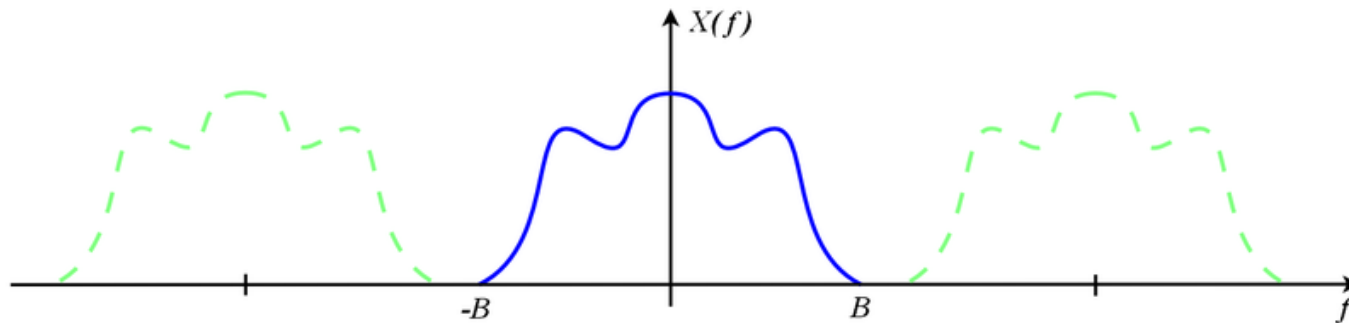


Sinc interpolation and zero-padding

Also known as “Whittaker–Shannon interpolation”



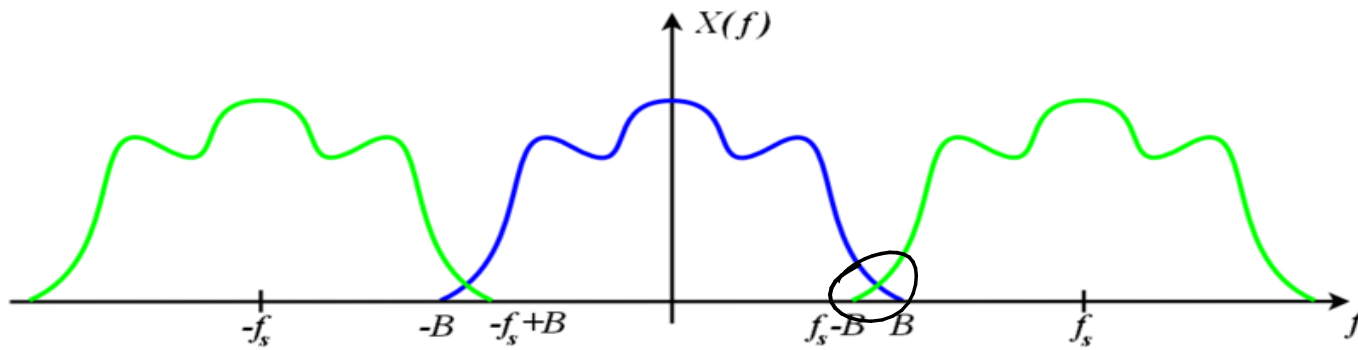
(Fourier space)



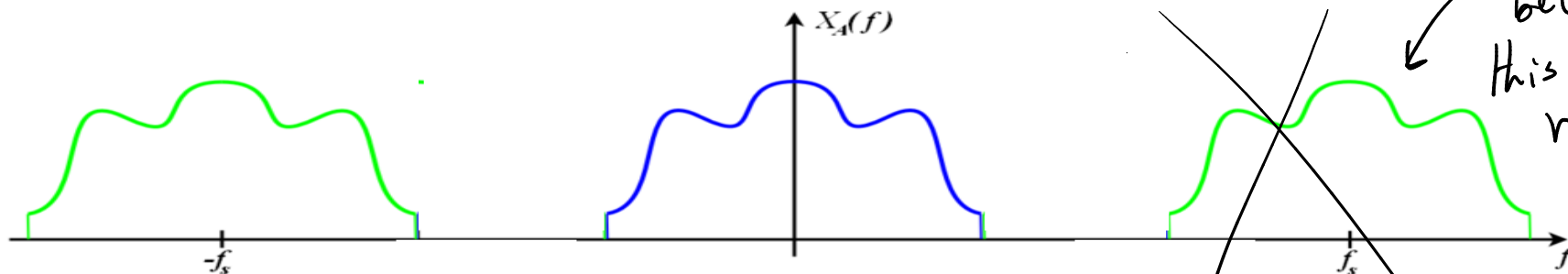
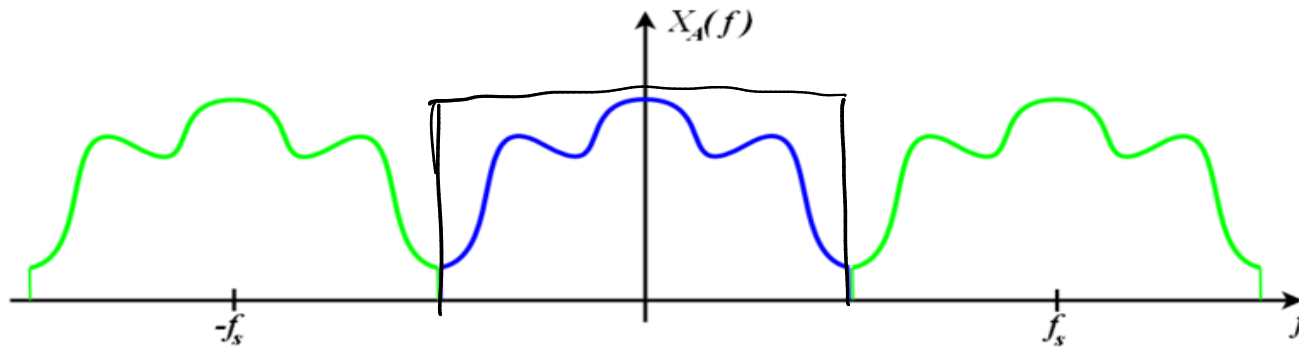
↑ zero-padding

Sinc interpolation and zero-padding

Also known as “Whittaker–Shannon interpolation”



Under sampled signal



non-zero because this is a resampling operation

if interpolation

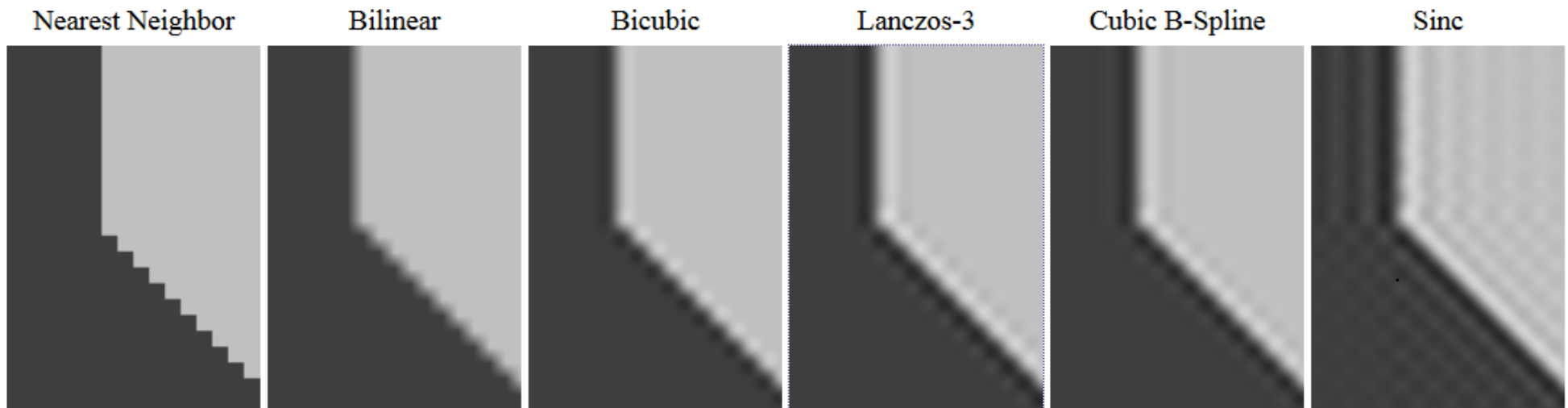
Reconstruction from samples

- Sinc interpolation can perfectly reconstruct a function from its samples if

- equivalent
- sampled at a rate higher than Nyquist rate
 - bandlimited up to Nyquist frequency
 - no aliasing

related to "ringing artifact" "Gibbs phenomenon"

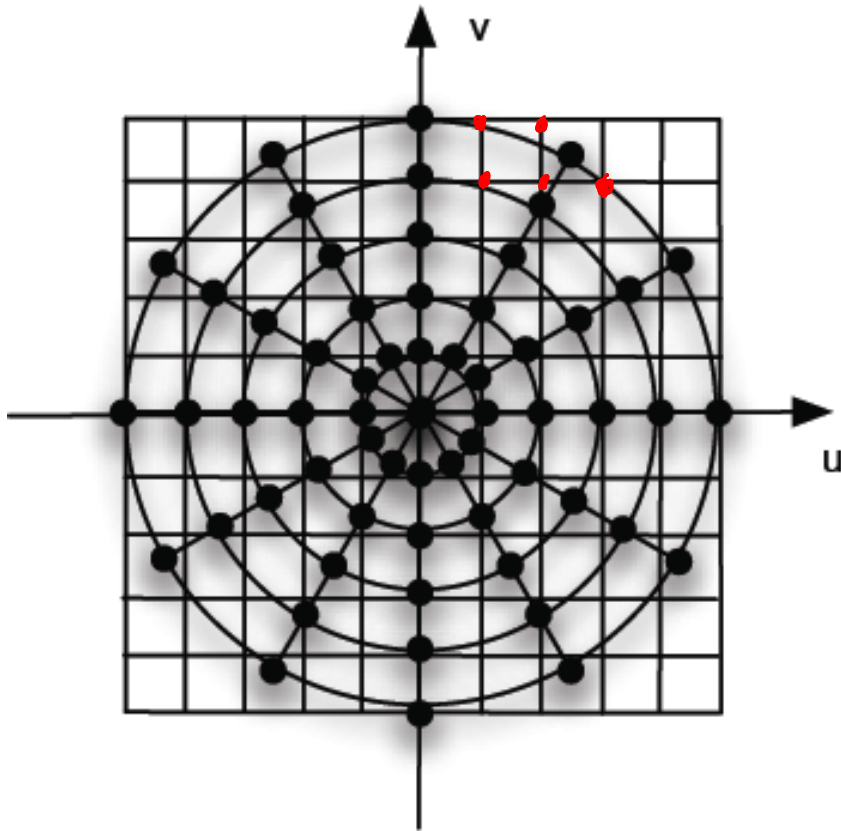
- Sinc interpolation introduces ringing otherwise, due to leakage of aliased frequencies



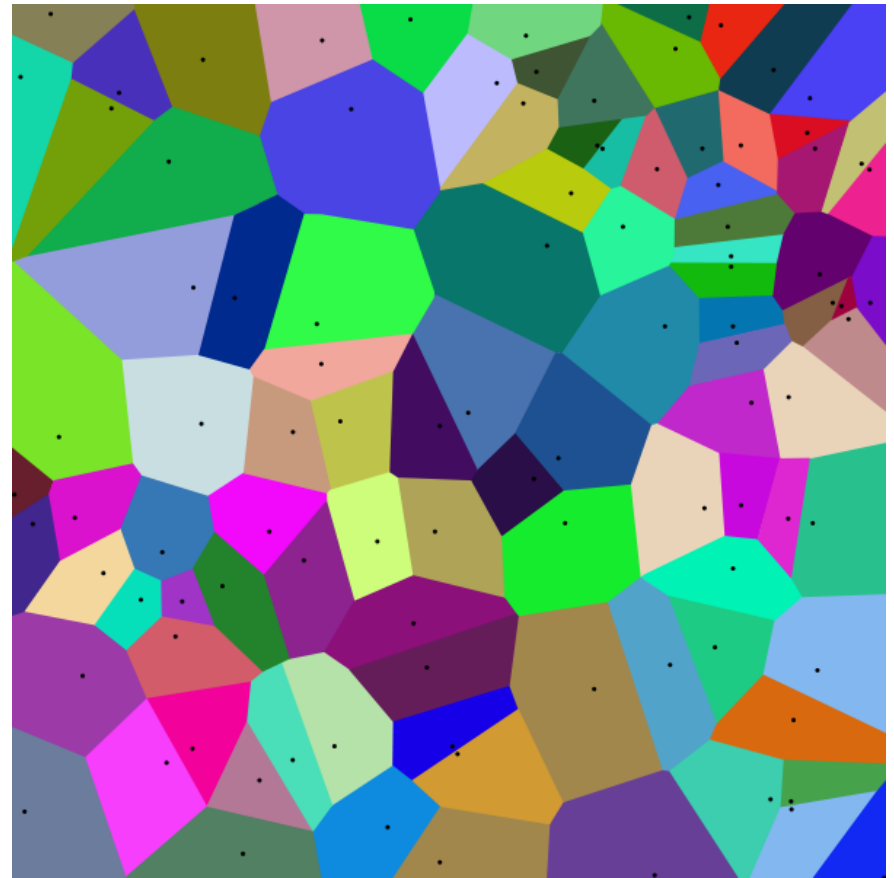
Linear interpolation of a step edge: a balance between staircase artifacts and ripples.

Other Interpolation

- Change from polar to cartesian grid
- Linear, but not translation invariant



polar vs. cartesian sampling



irregular sampling

Summary

- Images can be represented as a sampling grid and pixel basis functions
- Need for interpolation arises when changing the grid
- Linear and translation invariant interpolation can be written as a convolution with an interpolation kernel function
- Typical interpolation kernels include nearest neighbor, linear, cubic and higher B-spline interpolation
- Zero-padding in one domain equals sinc interpolation in the other
- “ideal” sinc interpolation may lead to ringing artifacts