

# 5 Image Interpolation

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5.1 Interpolation of Image Signals

5.2 B-Splines

5.3 Image Up-Sampling

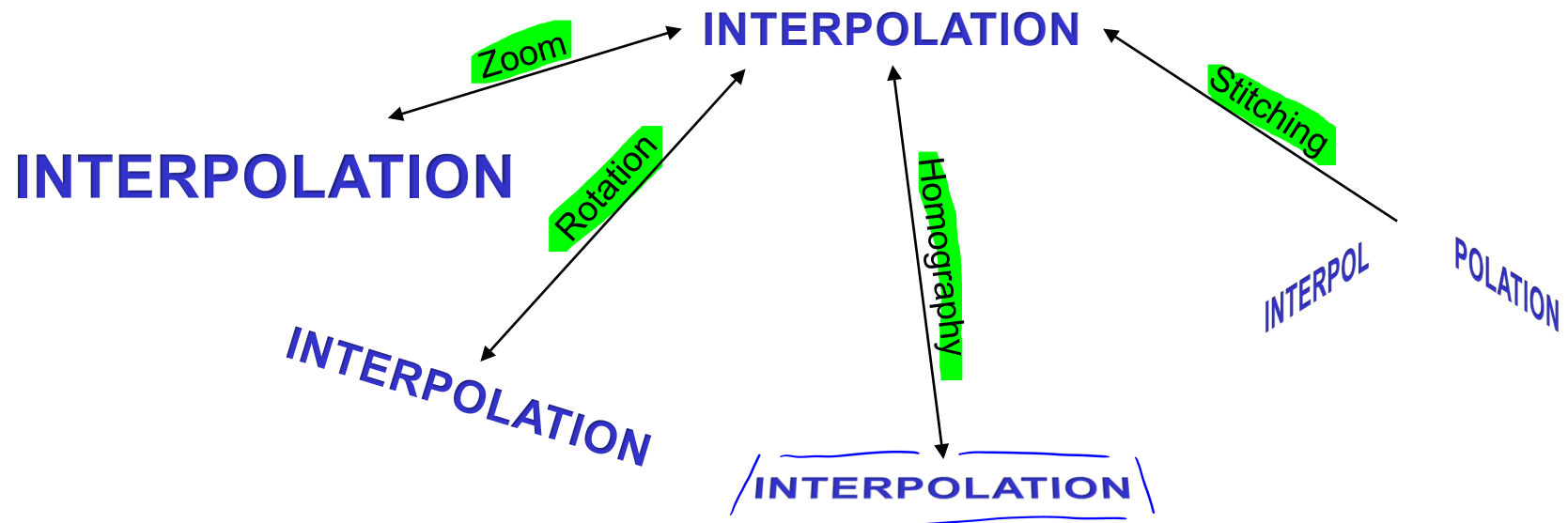
5.4 Triangulation-based Techniques

# 5.1 Interpolation of Image Signals

**Goal:** use available data to estimate sample values at new locations

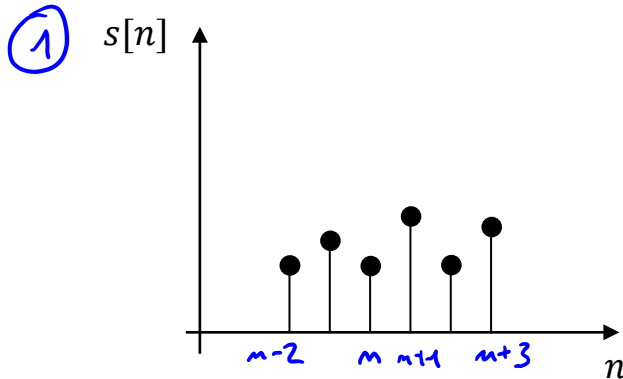
- Resize (zoom in/out)
- Distortions (affine transforms, homographies)
- Up-sampling (e.g. for sub-pixel accuracy)

*Not video interpolation  $\Rightarrow$  frame rate upconversion (FRUC) in this lecture*



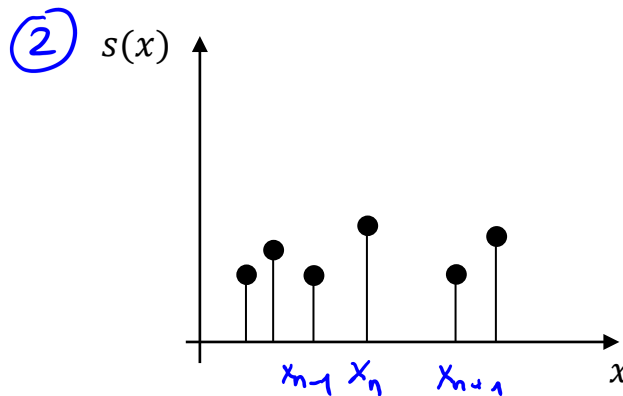
# Interpolation of Image Signals

**In general:** <sup>①</sup>equally vs <sup>②</sup>non-equally sampled points



**Advantage:** allows separability

- First interpolate rows then columns
- Nearest neighbor, Bi-linear (Bi-cubic), Lanczos, Splines



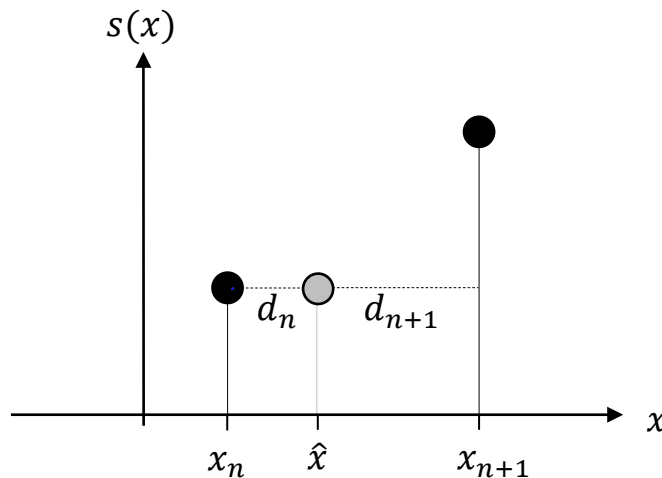
No assumption about sample locations

- Requires generic interpolation methods
  - Typically based on triangulation
- No easy generalization from 1D to 2D*

# Zero-Order Interpolation

Interpolated value

- **Copy** the value of the **nearest** available **neighbor** sample



Rule:

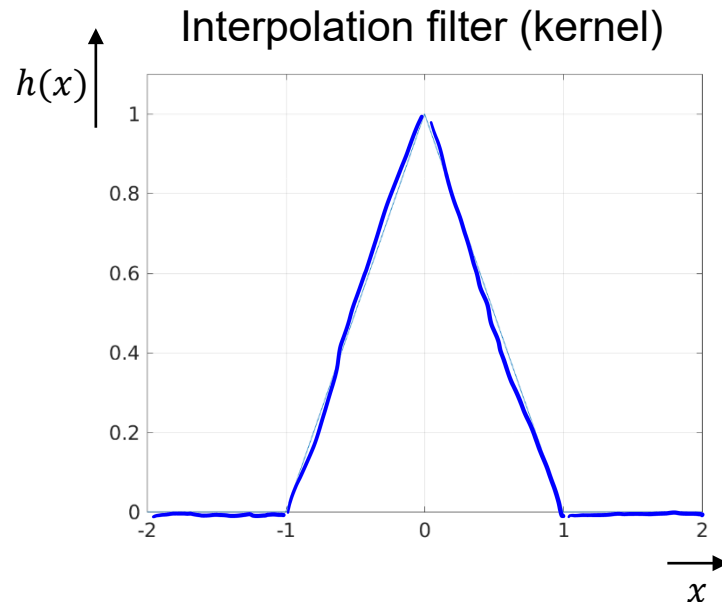
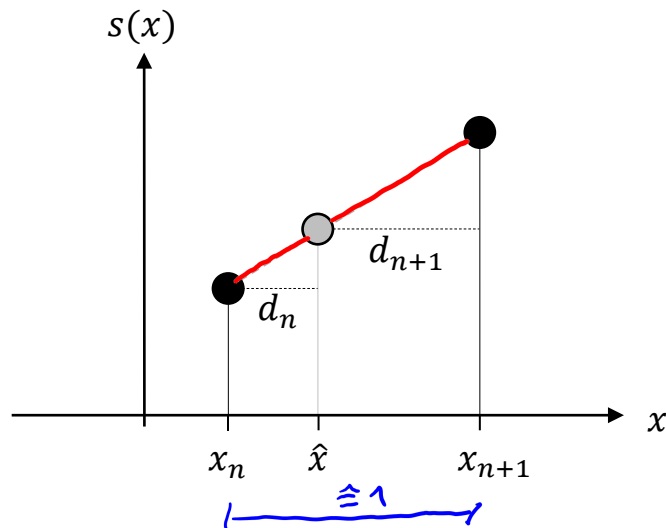
$$s(\hat{x}) = \begin{cases} s(x_n) & d_n \leq d_{n+1} \\ s(x_{n+1}) & d_{n+1} < d_n \end{cases}$$

Produces considerable artefacts, *but computationally easy*

- E.g. distortions of straight edges

# First-Order Interpolation

Fits straight line between two consecutive samples  $\Rightarrow$  **linear** interpolation



$$s(\hat{x}) = \frac{x_{n+1} - \hat{x}}{x_{n+1} - x_n} s(x_n) + \frac{\hat{x} - x_n}{x_{n+1} - x_n} s(x_{n+1})$$

$$= \frac{d_{n+1}}{d_n + d_{n+1}} s(x_n) + \frac{d_n}{d_n + d_{n+1}} s(x_{n+1})$$

For unitary pixel spacing:  $d_n + d_{n+1} = 1$

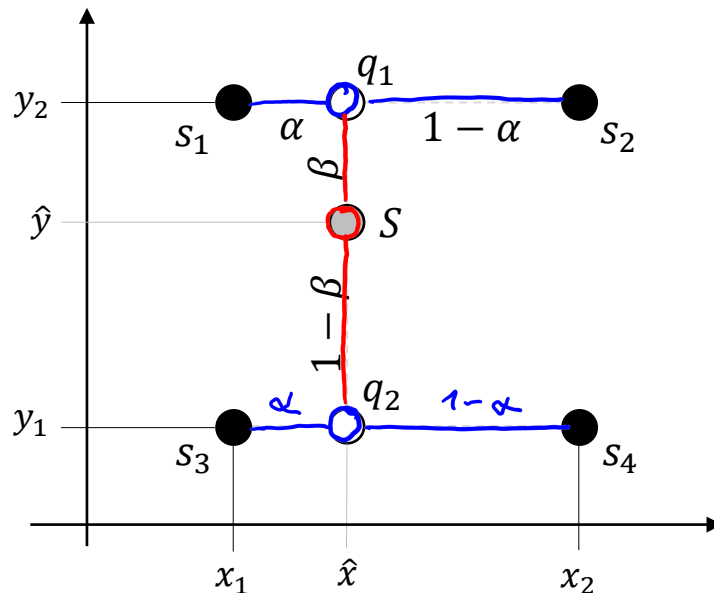
$$s(\hat{x}) = d_{n+1} s(x_n) + d_n s(x_{n+1})$$

$\text{or } s(\hat{x}) = \sum_{x_n} s(x_n) \cdot h(\hat{x} - x_n) \hat{=} \text{conv.}$

# Bilinear Interpolation $\Rightarrow 2D$

For **image** signals

- first applied to rows then to columns (separability) or vice versa



Normalized distances

$$\alpha = \frac{\hat{x} - x_1}{x_2 - x_1} \quad \beta = \frac{\hat{y} - y_1}{y_2 - y_1}$$

$\sim$  **Linear** interpolations

$$q_1 = (1 - \alpha)s_1 + \alpha s_2$$

$$q_2 = (1 - \alpha)s_3 + \alpha s_4$$

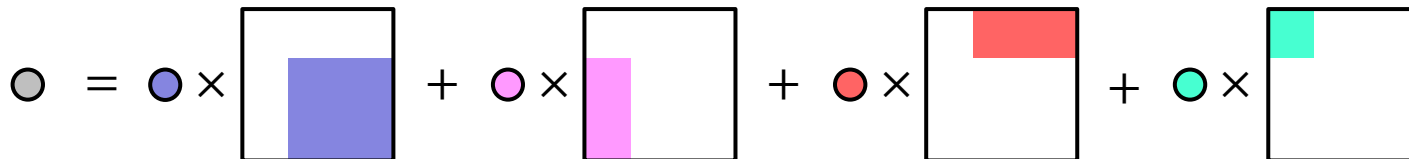
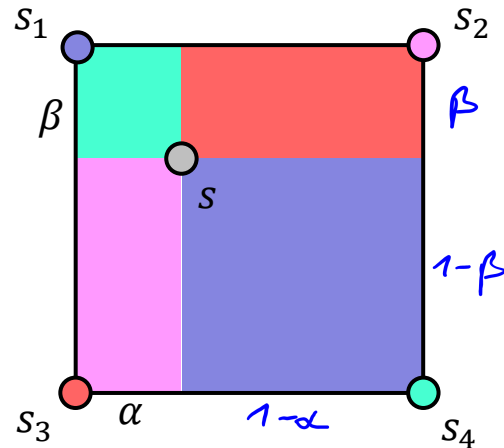
## Bilinear interpolation

$$s = (1 - \beta)q_1 + \beta q_2 \stackrel{(*)}{=} (1 - \beta)(1 - \alpha)s_1 + (1 - \beta)\alpha s_2 + \beta(1 - \alpha)s_3 + \beta\alpha s_4$$

# Bilinear Interpolation

**Interpretation:** area-based weighted linear combination

$$s = (1 - \beta)(1 - \alpha)s_1 + (1 - \beta)\alpha s_2 + \beta(1 - \alpha)s_3 + \beta\alpha s_4$$



# Ideal Interpolation

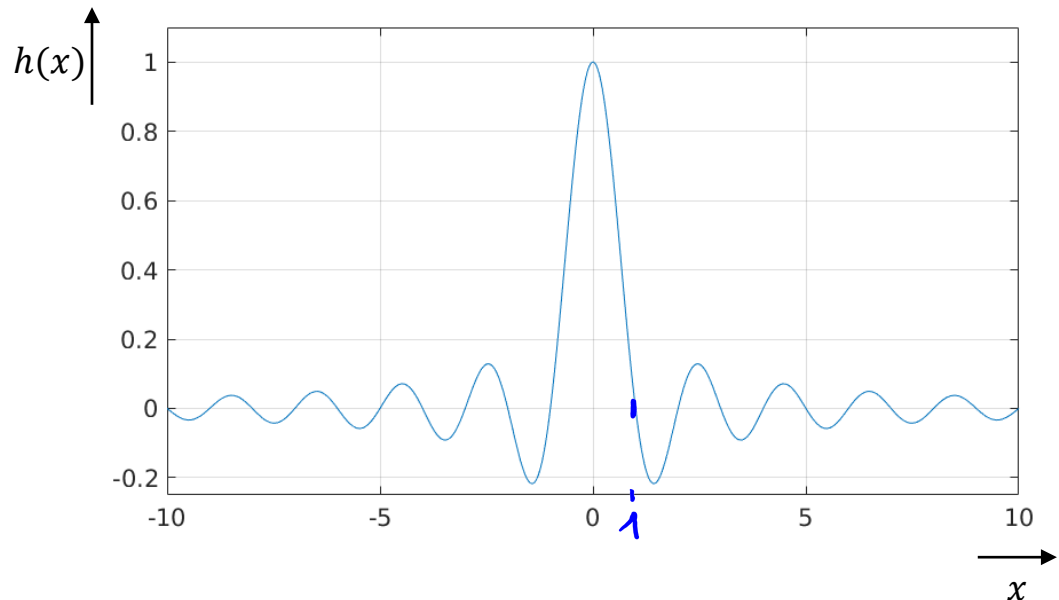
## Sinc filter

- Maintains frequency content (doesn't affect LF, doesn't add HF)
- Ideal low-pass filter (perfect cut-off HF)

$$h(x) = \frac{\sin(\pi x)}{\pi x} = \text{sinc}(x)$$

$$g(\hat{x}) = \sum_{x_n} g(x_n) \text{sinc}(\hat{x} - x_n)$$

(see 5-5)



## Problems

- Infinite summation required
- Blurring due to finite signal length and instationarity of image signals



# Lanczos Interpolator

Practical approximation of sinc filter

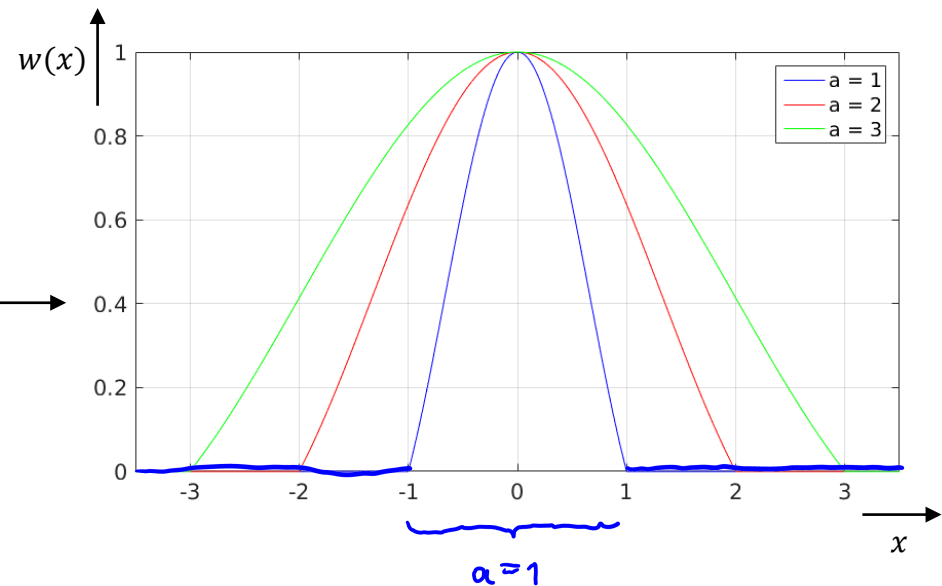
- Finite support



Kornél Lanczos  
(1893-1974)

Multiply sinc filter by **Lanczos window**

$$w(x) = \begin{cases} \text{sinc}\left(\frac{x}{a}\right) & -a < x < a \\ 0 & \text{otherwise} \end{cases}$$

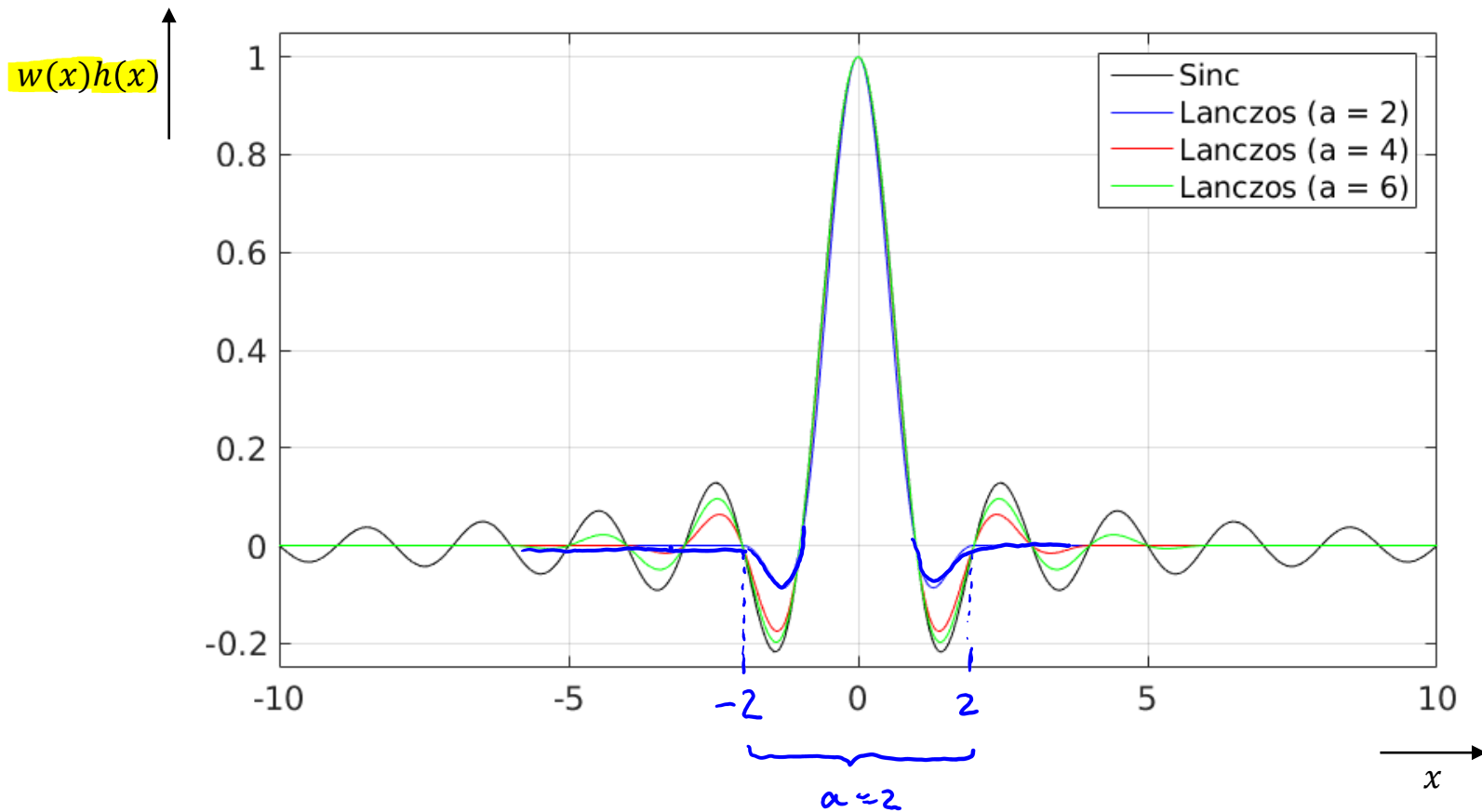


Parameter  $a$

- Determines filter width

# Lanczos Interpolator

## Comparison



# Examples (2× magnification)

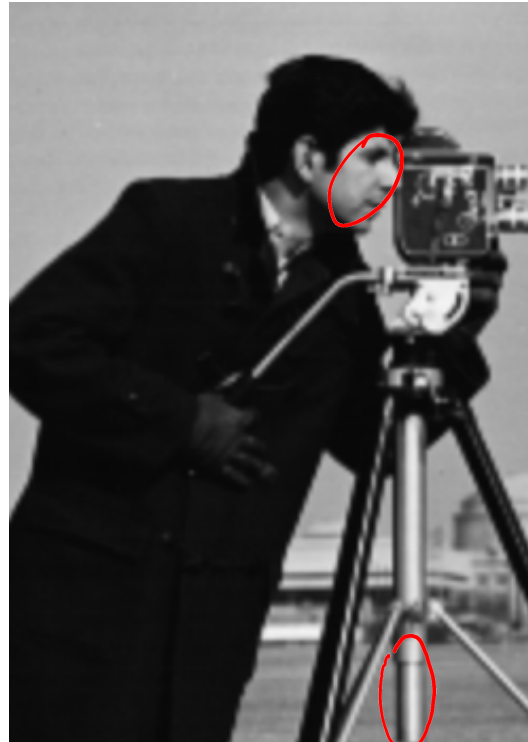
*"zero order"*

**Nearest** neighbor



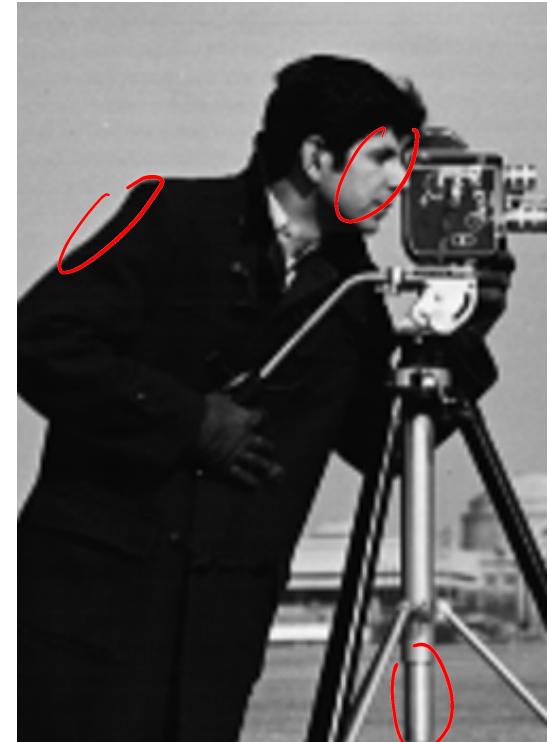
*"first order"*

**Linear** interpolation



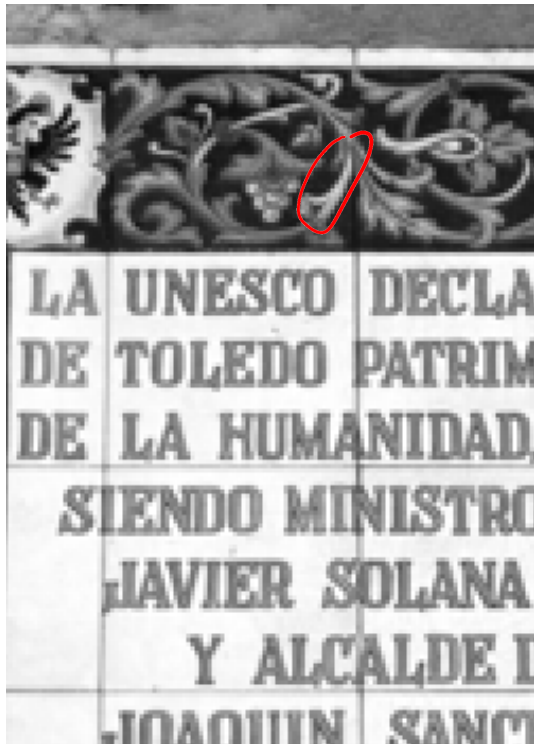
*"windowed ideal"*

**Lanczos** interpolation



# Examples (2× magnification)

Nearest neighbor



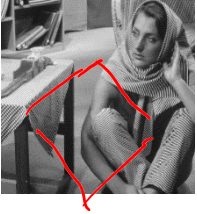
Linear interpolation



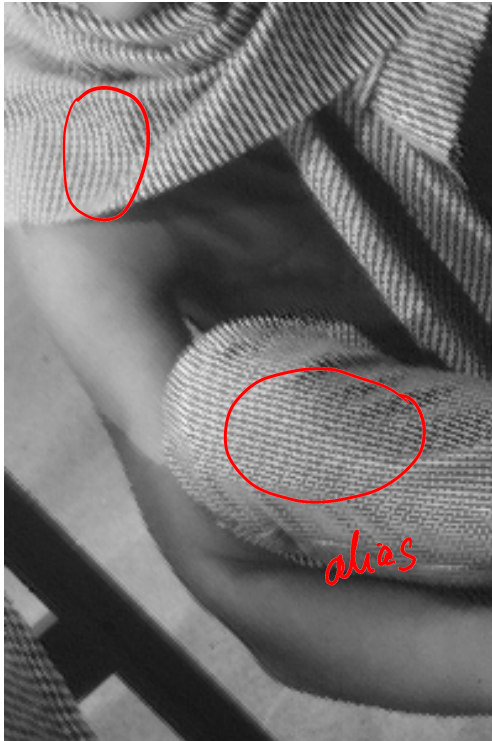
Lanczos interpolation



# Examples (45° rotation)



Nearest neighbor



Linear interpolation



Lanczos interpolation



# Examples (-45° rotation)



Nearest neighbor



Linear interpolation

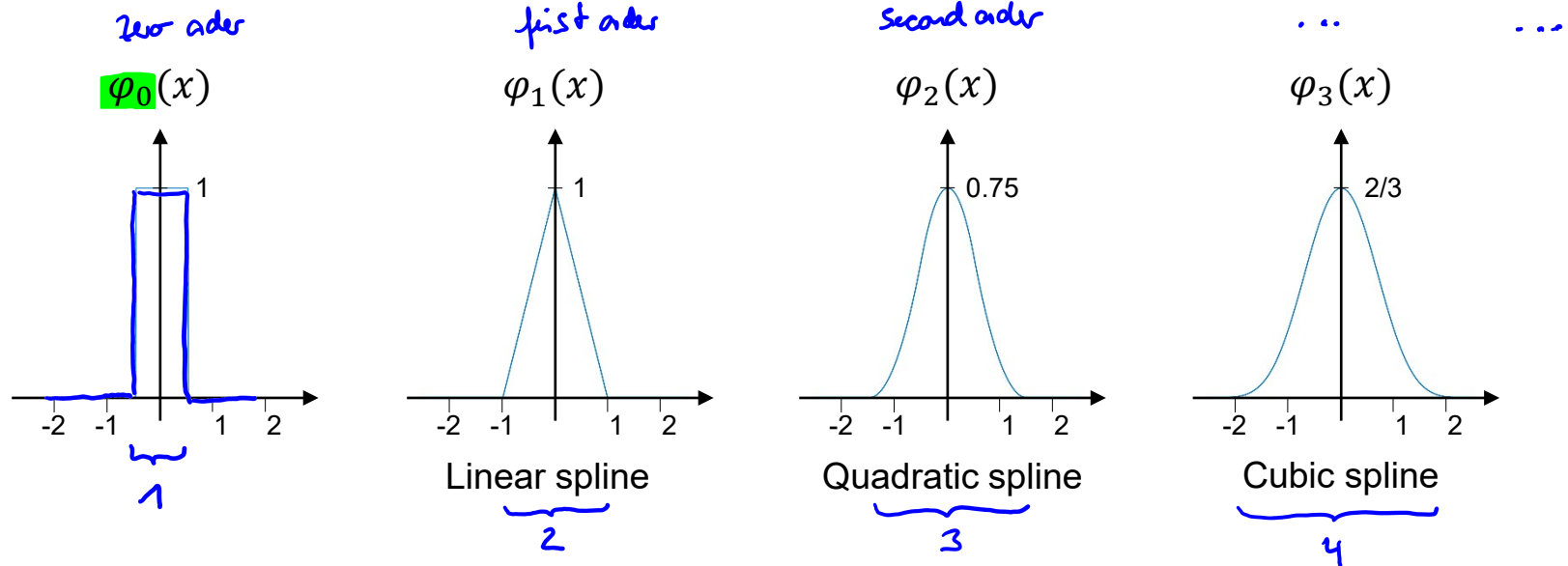


Lanczos interpolation



## 5.2 B-Splines Basis Functions

Construction rule for equidistant spacing of signal samples



Zero-order B-spline:

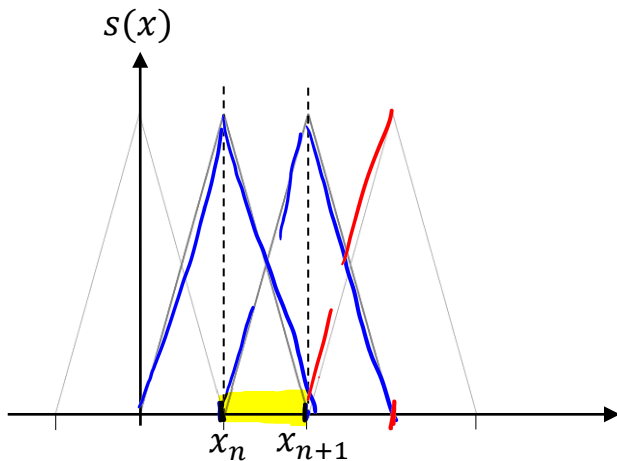
$$\varphi_0(x) = \begin{cases} 1 & -1/2 < x < 1/2 \\ 1/2 & |x| = 1/2 \\ 0 & \text{otherwise} \end{cases}$$

**Recursion:**

$$\varphi_n(x) = \varphi_0(x) * \varphi_{n-1}$$



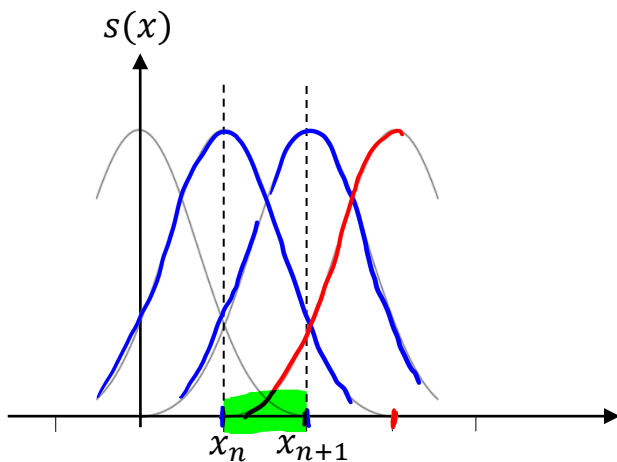
# Superposition of B-Splines



## Linear splines

- Two overlapping splines between two (consecutive) sampling points

$$s(x) = c_n \varphi_1(x - x_n) + c_{n+1} \varphi_1(x - x_{n+1})$$



## Cubic splines

- Four overlapping splines between four (consecutive) sampling points

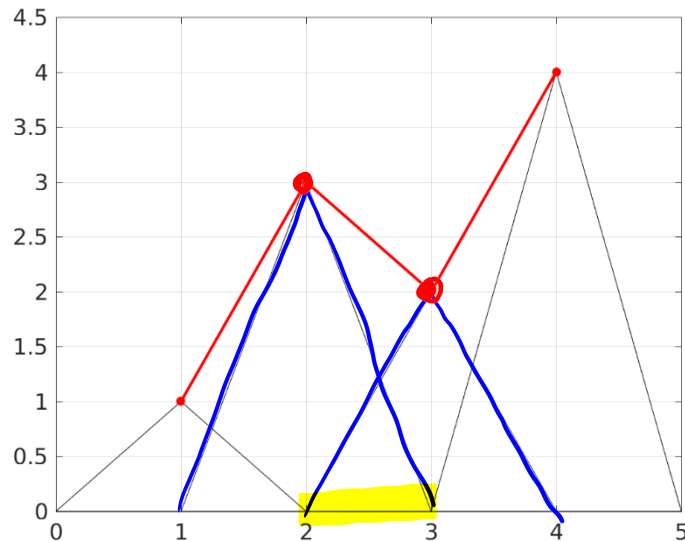
$$s(x) = c_{n-1} \varphi_3(x - x_{n-1}) + c_n \varphi_3(x - x_n) + c_{n+1} \varphi_3(x - x_{n+1}) + c_{n+2} \varphi_3(x - x_{n+2})$$



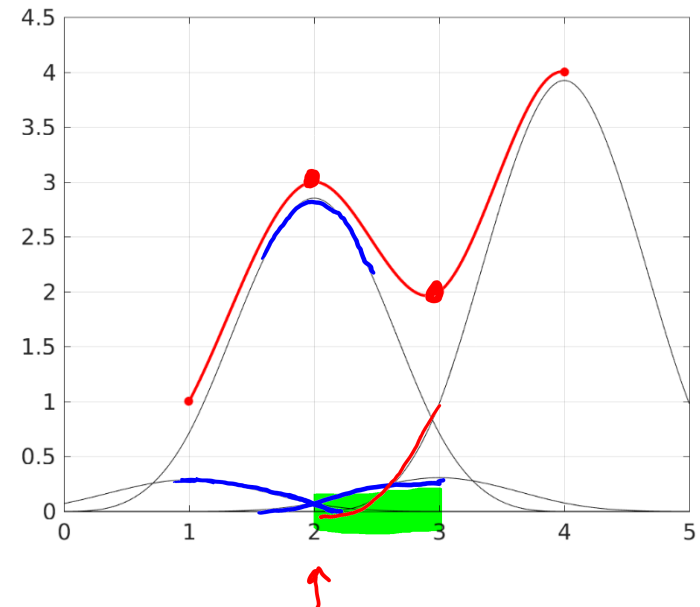
# Superposition of Weighted B-Splines

**Example:** Interpolate four sample values  $s(x) = [1, 3, 2, 4]$  at positions  $x_n = [1, 2, 3, 4]$

Linear splines



Cubic splines

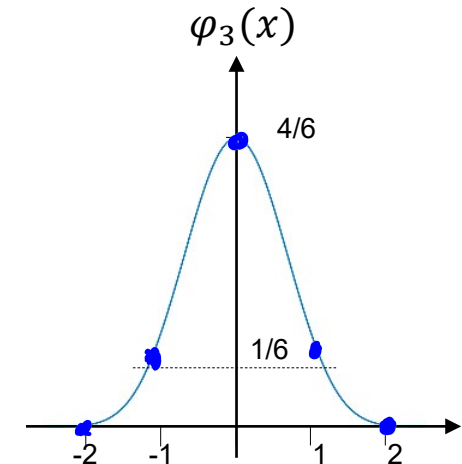


# Obtaining Spline Coefficients

Consider **cubic** spline interpolation

$$\varphi_3 = \begin{cases} \frac{2}{3} - x^2 + \frac{|x|^3}{2} & \text{for } 0 \leq |x| < 1 \\ \frac{(2 - |x|)^3}{6} & \text{for } 1 \leq |x| < 2 \\ 0 & \text{otherwise} \end{cases}$$

$$\varphi_3(-1) = \frac{1}{6} \quad \varphi_3(0) = \frac{4}{6} \quad \varphi_3(1) = \frac{1}{6}$$



## Constraint

- Spline interpolation reconstructs **exactly** available samples  $g(x_n)$

$$s(x_n) = \frac{1}{6}(c_{n-1} + 4c_n + c_{n+1})$$

$\uparrow$   
given point

# Obtaining Spline Coefficients

Matricial notation for signal  $s(x_n)$  for  $n = 0, 1, \dots, N - 1$

$$\begin{bmatrix} s(x_0) \\ s(x_1) \\ s(x_2) \\ \vdots \\ s(x_{N-1}) \end{bmatrix} = \frac{1}{6} \begin{bmatrix} 4 & 1 & 0 & 0 & \dots & 0 \\ 1 & 4 & 1 & 0 & \dots & 0 \\ 0 & 1 & 4 & 1 & \dots & 0 \\ \vdots & \ddots & \ddots & \ddots & \ddots & \vdots \\ 0 & \dots & 0 & 1 & 4 & 1 \\ 0 & \dots & 0 & 0 & 1 & 4 \end{bmatrix} \begin{bmatrix} c_0 \\ c_1 \\ c_2 \\ \vdots \\ c_{N-1} \end{bmatrix}$$

*Handwritten annotations:*

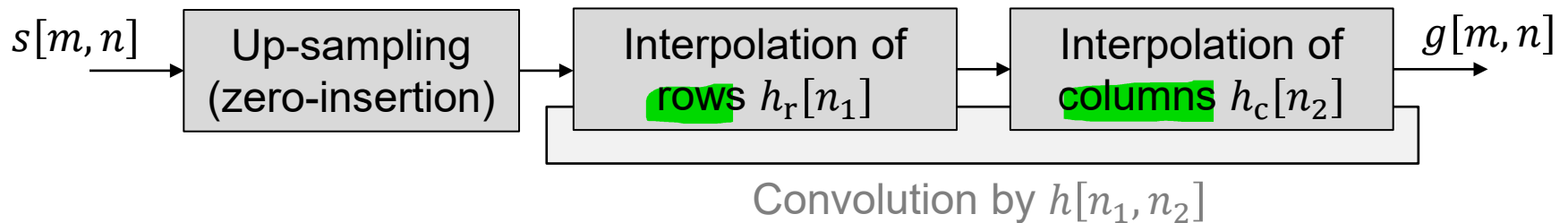
- An arrow points from the text "given sampling points" to the vector  $\begin{bmatrix} s(x_0) \\ \vdots \\ s(x_{N-1}) \end{bmatrix}$ .
- A blue bracket under the matrix is labeled "all weights of cubic spline at sampling points".
- A red bracket under the vector  $\begin{bmatrix} c_0 \\ \vdots \\ c_{N-1} \end{bmatrix}$  is labeled "unknown".

## 5.3 Image Up-sampling

Special case of image interpolation

- Both available and unknown samples are equally spaced

Up-sampling by factor  $L$



**Interpolation filter:**

$$h[m, n] = h_r[m] * h_c[n]$$

Row vector

Column vector

- Filter gain of  $L^2$  to compensate energy loss due to insertion of zeros

# Zero-Order Interpolation

## Zero-order hold filter (nearest neighbor)

- Repeat (hold) pixels along scan line, afterwards repeat each scan line

$$h_r[m] = h_c^T[n] = \mathbf{1}_{L \times 1} \longrightarrow h[m, n] = \mathbf{1}_{L \times L}$$

Example ( $L=2$ ):  $h[m, n] = \begin{bmatrix} 1 & 1 \\ 1 & 1 \end{bmatrix}$  2D impulse response function

$$\begin{bmatrix} 1 & 3 & 2 \\ 4 & 5 & 6 \end{bmatrix} \xrightarrow{\text{original image}} \begin{bmatrix} 1 & 0 & 3 & 0 & 2 & 0 \\ 0 & 0 & 0 & 0 & 0 & 0 \\ 4 & 0 & 5 & 0 & 6 & 0 \\ 0 & 0 & 0 & 0 & 0 & 0 \end{bmatrix} \xrightarrow{\text{zero-order interpolation}} \begin{bmatrix} 1 & 1 & 3 & 3 & 2 & 2 \\ 1 & 1 & 3 & 3 & 2 & 2 \\ 4 & 4 & 5 & 5 & 6 & 6 \\ 4 & 4 & 5 & 5 & 6 & 6 \end{bmatrix}$$

original image

# First-Order Interpolation

## Linear interpolation (first-order)

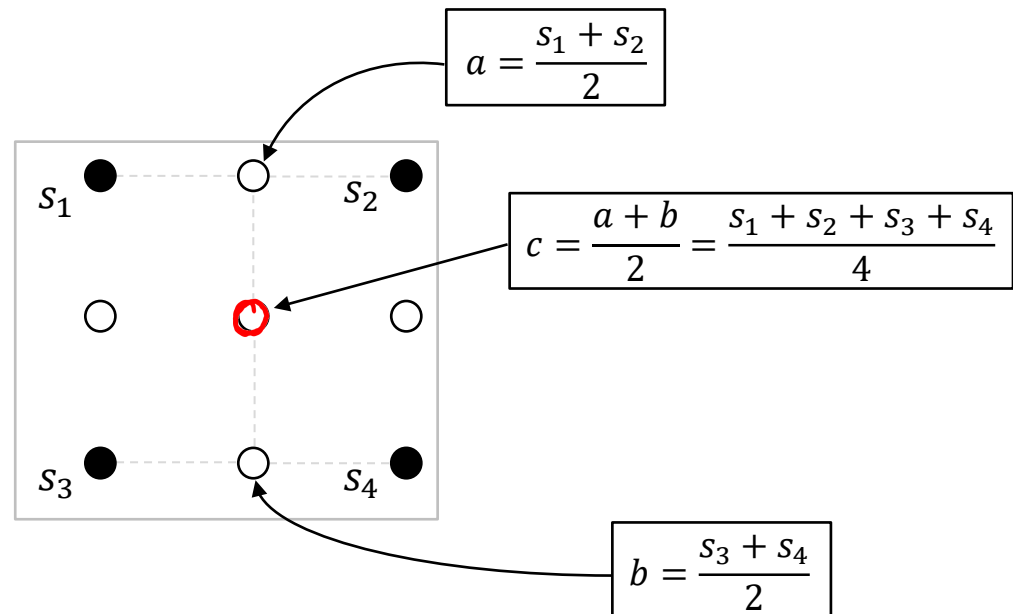
- Fits straight line between pixels along row / column

$$h_r[m] = h_c^T[n] = \frac{1}{L} [1 \ 2 \ \dots \ L-1 \ L \ L-1 \ \dots \ 2 \ 1]$$

Example ( $L=2$ )

$$h_r[m] = h_c^T[n] = \begin{bmatrix} \frac{1}{2} & 1 & \frac{1}{2} \end{bmatrix}$$

$$h[m,n] = \begin{bmatrix} 1/4 & 1/2 & 1/4 \\ 1/2 & \textcircled{1} & 1/2 \\ 1/4 & 1/2 & 1/4 \end{bmatrix}$$



## 5.4 Not-equally Spaced Samples

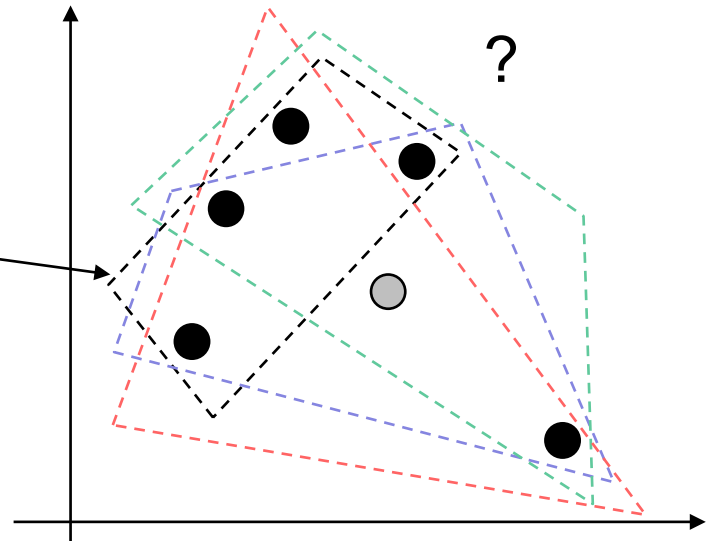
Sample location are arbitrary

- No regularity can be assumed
- No separability

Example: bilinear interpolation

- Which 4 samples to consider?

Four closest samples won't work (all on the same side)

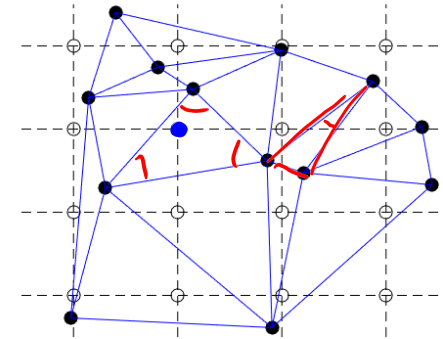


**Typical solution:** triangulation

# Triangulation

Triangulate the available mesh of samples

- Each unknown pixel is enclosed by a triangle
- Many possible triangulations



## Delaunay triangulation

- Most commonly used
- Maximizes the minimum angle in a triangle
  - Avoids extremely thin (degenerated) triangles
- Circumcircle of a triangle contains **only** its vertices (and no other samples)

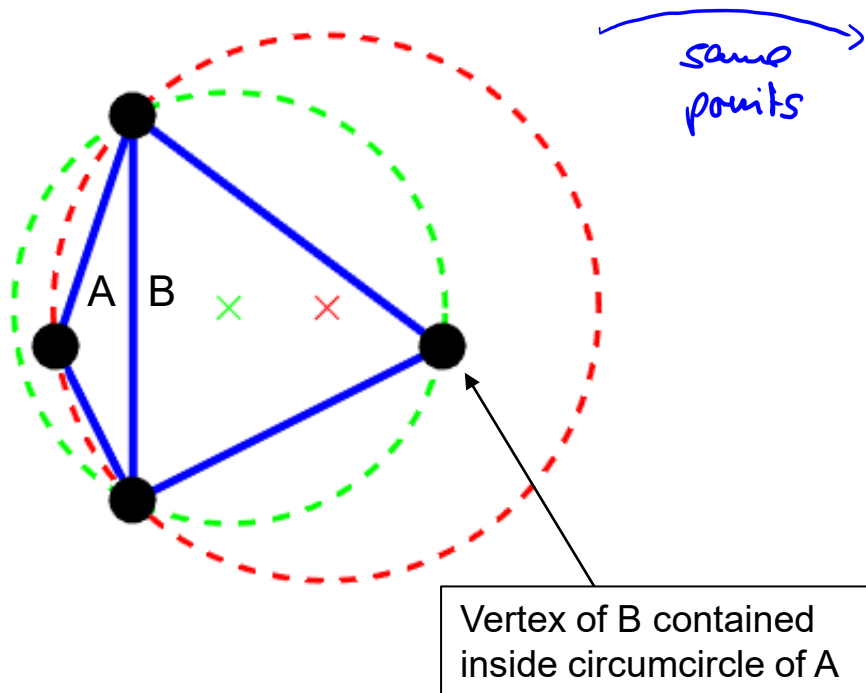


Boris Nikolaevich  
Delaunay (1890-1980)



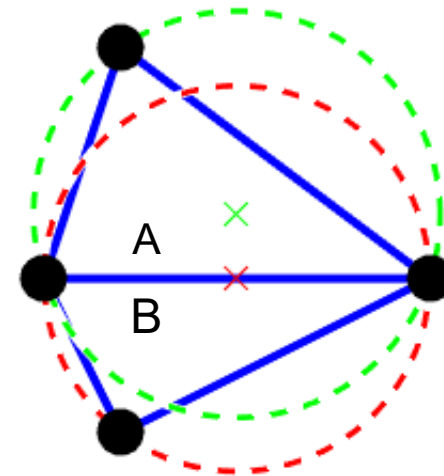
# Delaunay Triangulation

Possible triangulation



Angles in A:  $[135^\circ, \mathbf{27^\circ}, \mathbf{18^\circ}]$   
Angles in B:  $[63^\circ, 63^\circ, 54^\circ]$

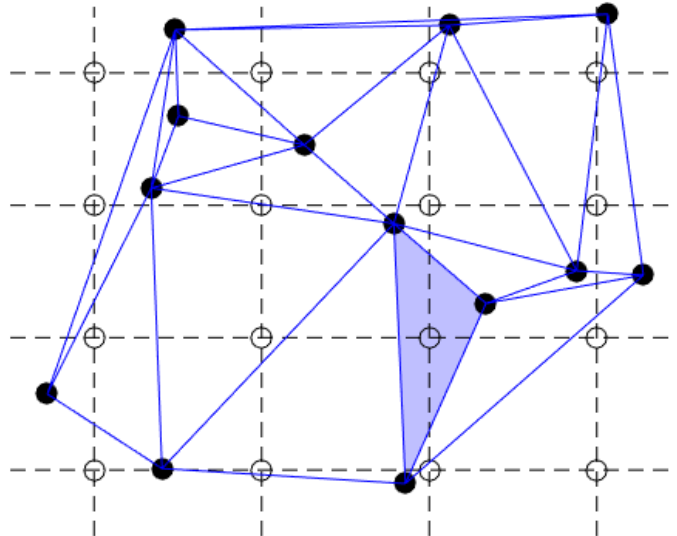
Delaunay triangulation



Angles in A:  $[63^\circ, 90^\circ, 27^\circ]$   
Angles in B:  $[71^\circ, 71^\circ, 38^\circ]$

# Linear Interpolation

Straightforward extension of 1D linear interpolation to 2D



Enclosing Delaunay triangle

- Three points define a plane (linear surface)

$$ax + by + cz + d = 0$$

$(x, y)$  sample coordinates  
 $z$  pixel value (color)

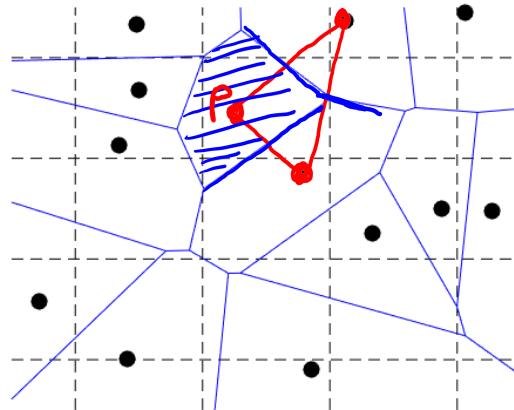
Interpolated value ( $z$ -coordinate) is obtained by evaluating plane equation at query position  $(x, y)$

# Natural Neighbor Interpolation

Based on **Voronoi tessellation**

- Dual graph of Delaunay triangulation
- Obtained by connecting centers of circumcircles (of each triangle)

$\hat{=}$  nearest  
neighbor  
classification  
(cf. Chap. 9)



**Voronoi cell** corresponding to sample point  $p$

- Region where all locations are closer to  $p$  than to any other sample point

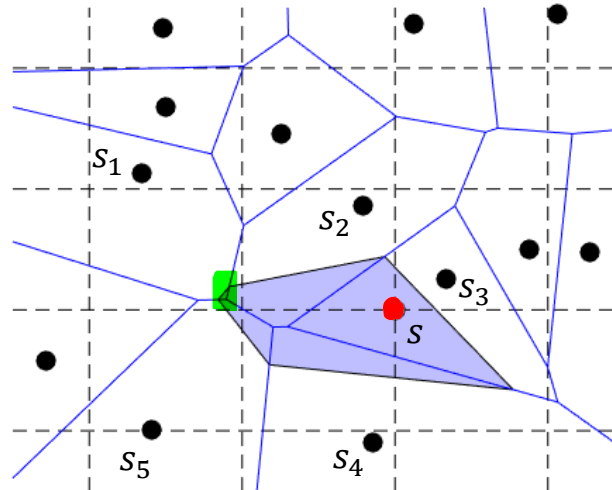


Georgy Feodosevich  
Voronoi (1868-1908)

# Natural Neighbor Interpolation

Place Voronoi cell at query point  $S \Rightarrow$  add another vertex, change tessellation accordingly

- Interpolated value  $s$  is based on (normalized) area overlap with amplitudes  $s_i$  of neighboring cells



$$s = s_1 \times \text{green square} + s_2 \times \text{triangle} + s_3 \times \text{triangle} + s_4 \times \text{triangle} + s_5 \times \text{triangle}$$