

# Interpolation

CS 205A:  
Mathematical Methods for Robotics, Vision, and Graphics

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# Announcement: Midterm

- ▶ Exams are graded (/50)
- ▶ Scores boosted by +5
- ▶ Hand back:
  - ▶ Available after class
  - ▶ Available from “assignment boxes” (Michela will email instructions)

## So Far

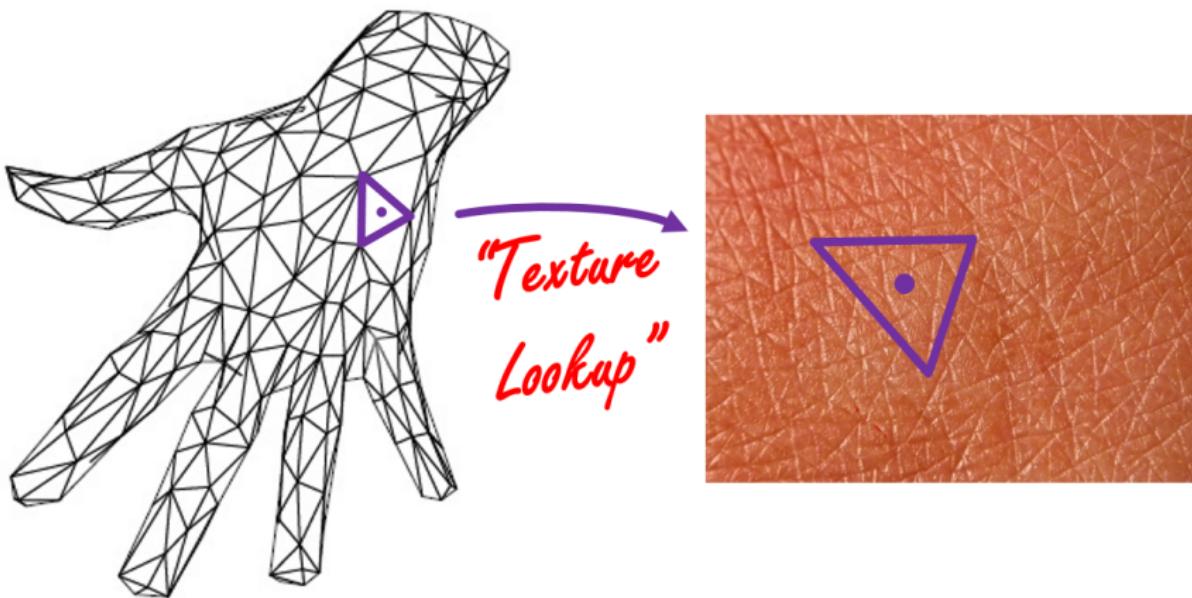
# Tools for *analyzing* functions:

# Roots, minima, ...

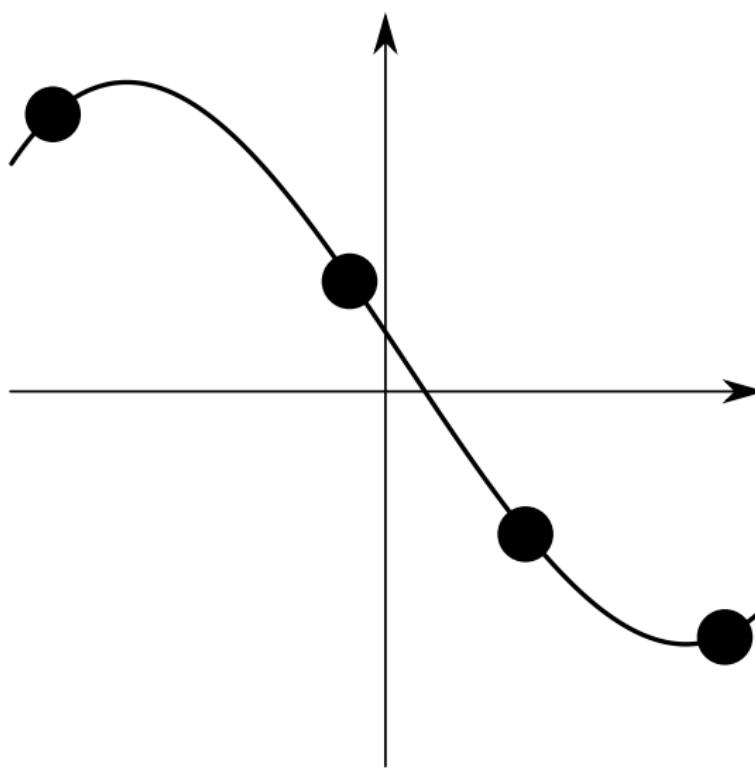
# Common Situation

The function **is**  
the unknown.

# Example: Rendering



# Example: Regression



# Input/Output

**Input:**  $\vec{x}_i \mapsto y_i$  (exactly)

**Output:**  $f(\vec{x})$  for  $\vec{x} \notin \{\vec{x}_i\}$

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Contrast with *regression*

## Initial Problem

$$f : \mathbb{R} \rightarrow \mathbb{R}$$

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**Given:**  $x_i \mapsto y_i$

## The Problem

$$\{f : \mathbb{R} \rightarrow \mathbb{R}\}$$

is a *huge* set.

# Common Strategy

Restrict search to a basis  $\phi_1, \phi_2, \dots$

$$f(x) = \sum_i a_i \phi_i(x)$$

**Need to compute:**  $\vec{a}$

# Monomial Basis

$$p_0(x) = 1$$

$$p_1(x) = x$$

$$p_2(x) = x^2$$

$$p_3(x) = x^3$$

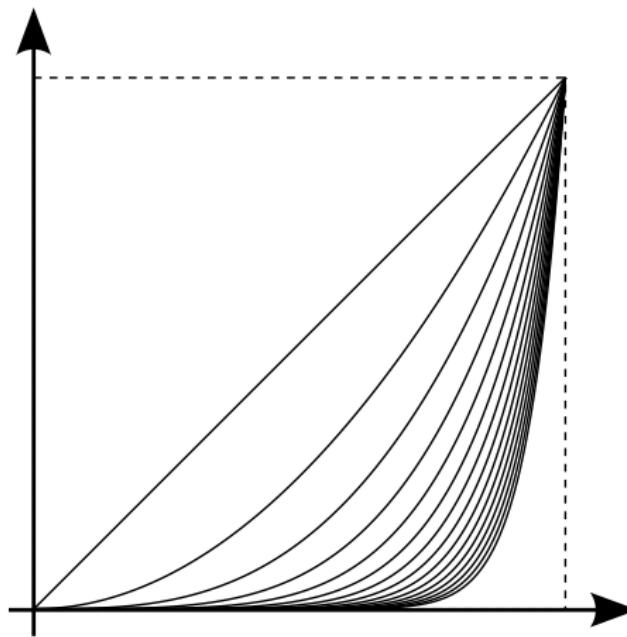
⋮ ⋮

$$f(x) = a_0 + a_1x + a_2x^2 + \cdots + a_{k-1}x^{k-1}$$

# Vandermonde System

$$\begin{pmatrix} 1 & x_1 & x_1^2 & \cdots & x_1^{k-1} \\ 1 & x_2 & x_2^2 & \cdots & x_2^{k-1} \\ \vdots & \vdots & \vdots & \cdots & \vdots \\ 1 & x_{k-1} & x_{k-1}^2 & \cdots & x_{k-1}^{k-1} \end{pmatrix} \begin{pmatrix} a_0 \\ a_1 \\ \vdots \\ a_{k-1} \end{pmatrix} = \begin{pmatrix} y_0 \\ y_1 \\ \vdots \\ y_{k-1} \end{pmatrix}$$

# Conditioning Issues



$p_k(x) = x^k$  look similar on  $[0, 1]$  for large  $k$

# Lagrange Basis

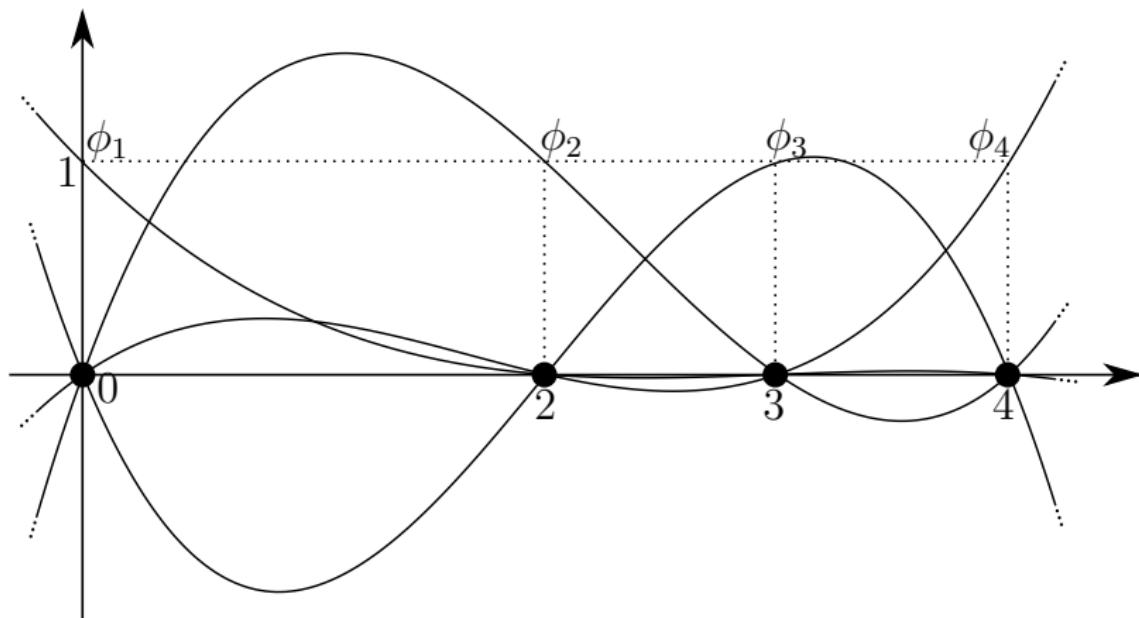
$$\phi_i(x) \equiv \frac{\prod_{j \neq i} (x - x_j)}{\prod_{j \neq i} (x_i - x_j)}$$

Still polynomial!

# Useful Property

$$\phi_i(x_\ell) = \begin{cases} 1 & \text{when } \ell = i \\ 0 & \text{otherwise} \end{cases}$$

# Illustration of Lagrange Basis



$$x_1 = 0, x_2 = 2, x_3 = 3, x_4 = 4$$

# Interpolation in Lagrange Basis

$$f(x) \equiv \sum_i y_i \phi_i(x)$$

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$O(n^2)$  time to evaluate.

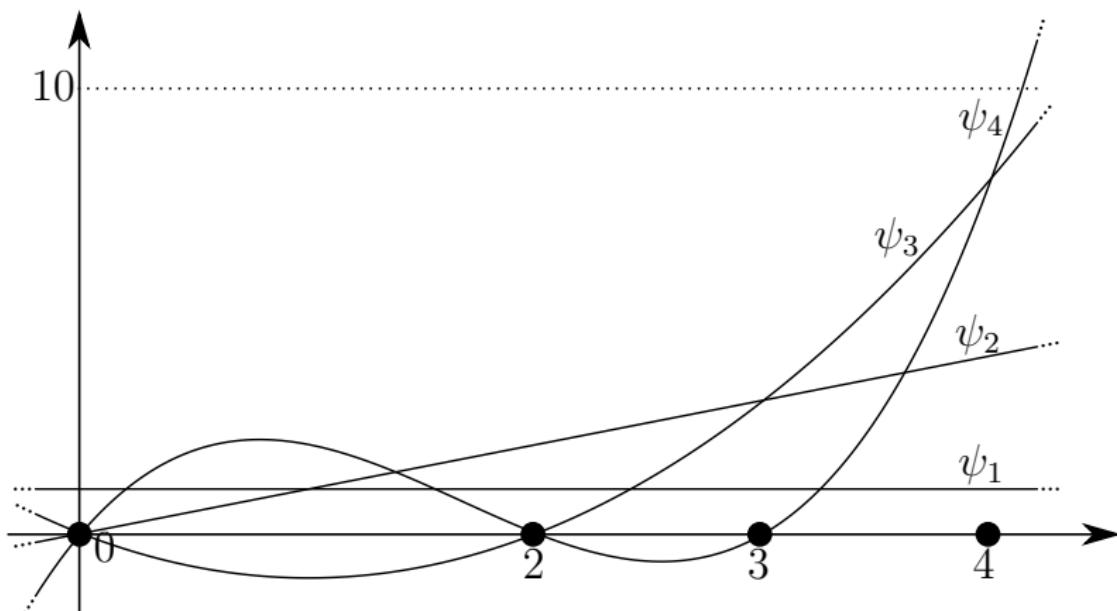
Numerical issues.

# Compromise: Newton Basis

$$\psi_i(x) = \prod_{j=1}^{i-1} (x - x_j)$$

$$\psi_1(x) \equiv 1$$

# Illustration of Newton Basis



$$x_1 = 0, x_2 = 2, x_3 = 3, x_4 = 4$$

# Evaluating in Newton

$$f(x_1) = c_1 \psi_1(x_1)$$

$$f(x_2) = c_1 \psi_1(x_2) + c_2 \psi_2(x_2)$$

$$f(x_3) = c_1 \psi_1(x_3) + c_2 \psi_2(x_3) + c_3 \psi_3(x_3)$$

⋮ ⋮

# Triangular System

$$\begin{pmatrix} \psi_1(x_1) & 0 & 0 & \cdots & 0 \\ \psi_1(x_2) & \psi_2(x_2) & 0 & \cdots & 0 \\ \psi_1(x_3) & \psi_2(x_3) & \psi_3(x_3) & \cdots & 0 \\ \vdots & \vdots & \vdots & \cdots & \vdots \\ \psi_1(x_k) & \psi_2(x_k) & \psi_3(x_k) & \cdots & \psi_k(x_k) \end{pmatrix} \begin{pmatrix} c_1 \\ c_2 \\ \vdots \\ c_k \end{pmatrix} = \begin{pmatrix} y_1 \\ y_2 \\ \vdots \\ y_k \end{pmatrix}$$

**Review:**  
Efficiency of solution?

# Triangular System

$$\begin{pmatrix} \psi_1(x_1) & 0 & 0 & \cdots & 0 \\ \psi_1(x_2) & \psi_2(x_2) & 0 & \cdots & 0 \\ \psi_1(x_3) & \psi_2(x_3) & \psi_3(x_3) & \cdots & 0 \\ \vdots & \vdots & \vdots & \cdots & \vdots \\ \psi_1(x_k) & \psi_2(x_k) & \psi_3(x_k) & \cdots & \psi_k(x_k) \end{pmatrix} \begin{pmatrix} c_1 \\ c_2 \\ \vdots \\ c_k \end{pmatrix} = \begin{pmatrix} y_1 \\ y_2 \\ \vdots \\ y_k \end{pmatrix}$$

**Review:**  
Efficiency of solution?  
 $O(n^2)$

# Important Point

All three methods yield the  
*same* polynomial.

# Rational Interpolation

$$f(x) = \frac{p_0 + p_1x + p_2x^2 + \cdots + p_mx^m}{q_0 + q_1x + q_2x^2 + \cdots + q_nx^n}$$

# Rational Interpolation

$$f(x) = \frac{p_0 + p_1x + p_2x^2 + \cdots + p_mx^m}{q_0 + q_1x + q_2x^2 + \cdots + q_nx^n}$$

$$y_i(q_0 + q_1x_i + \cdots + q_nx_i^n) = p_0 + p_1x_i + \cdots + p_mx_i^m$$

## Null space problem!

# Rational Interpolation

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Null space problem!

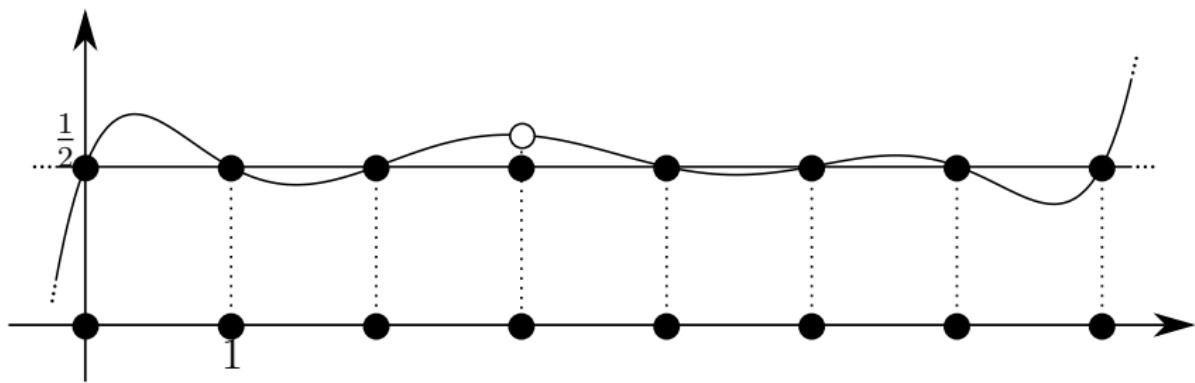
Scary example:  $n = m = 1; (0, 1), (1, 2), (2, 2)$

# Fourier Series

$$\cos(kx)$$

$$\sin(kx)$$

# Problem with Polynomials



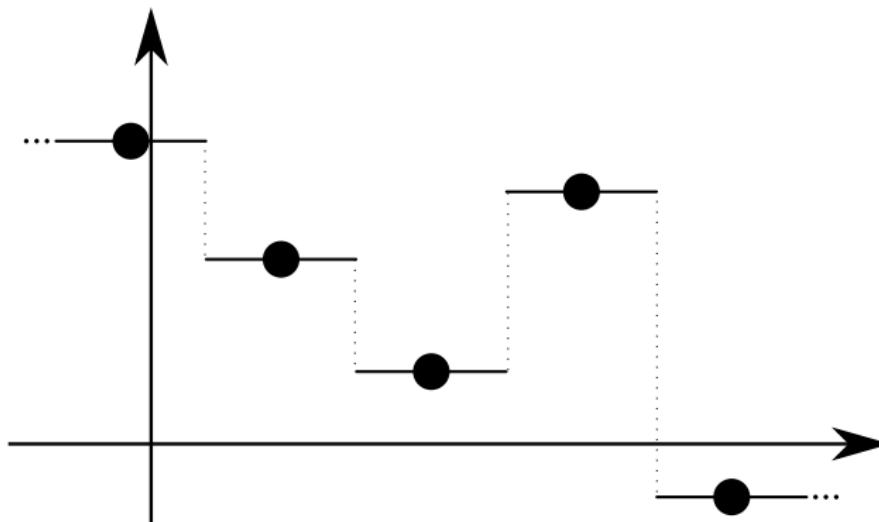
Local change can have global effect.

# Compact Support

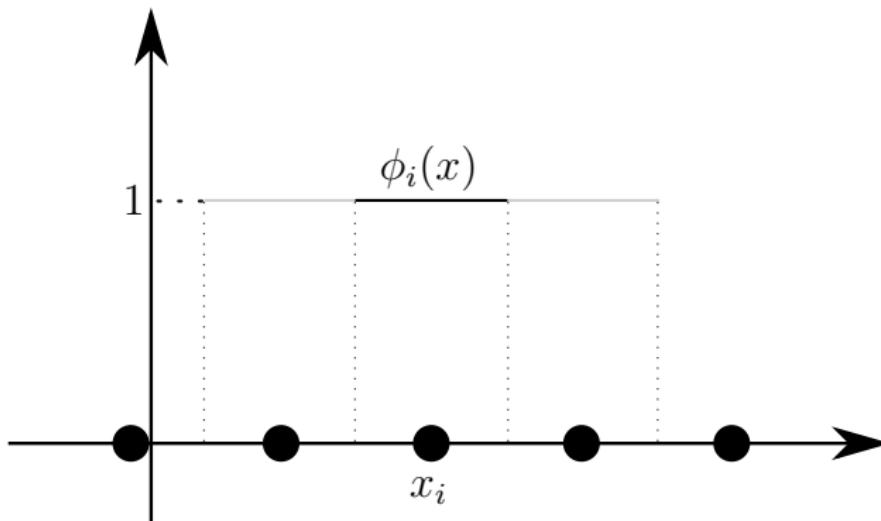
## Compact support

A function  $g(x)$  has *compact support* if there exists  $C \in \mathbb{R}$  such that  $g(x) = 0$  for any  $x$  with  $|x| > C$ .

# Piecewise Constant Interpolation

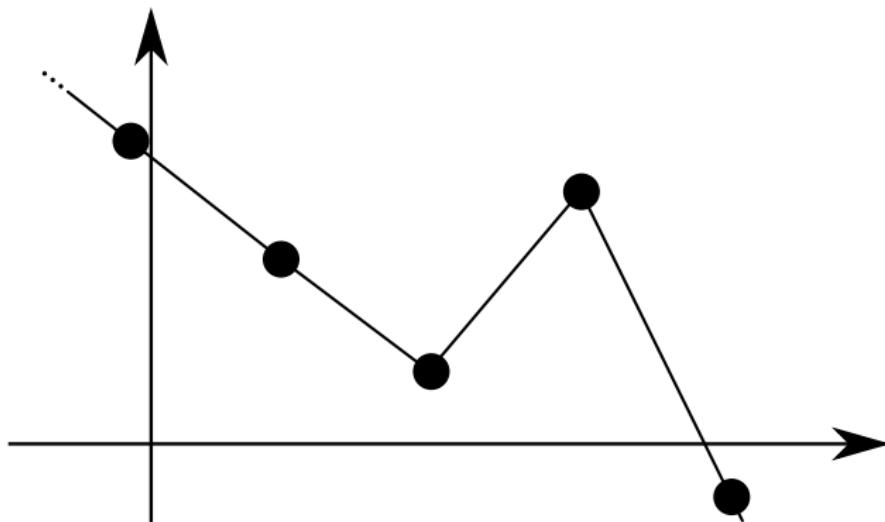


# Piecewise Constant Basis

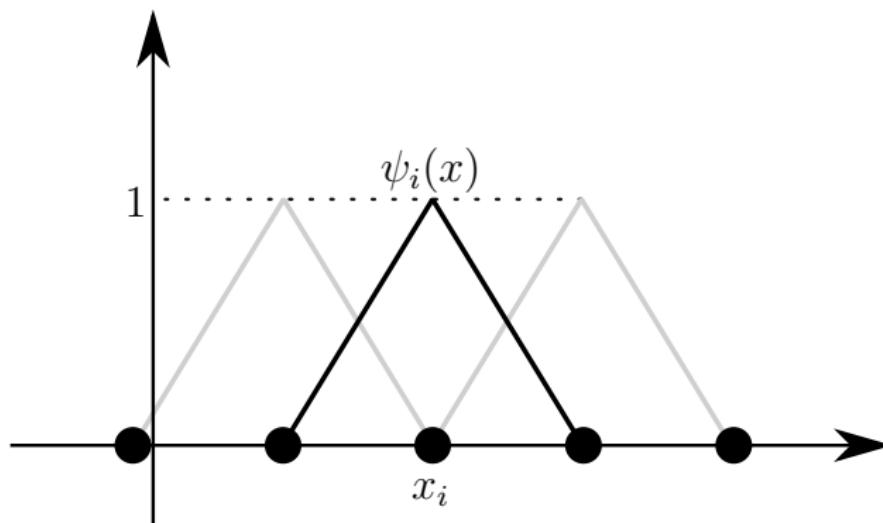


$$\phi_i(x) = \begin{cases} 1 & \text{when } \frac{x_{i-1}+x_i}{2} \leq x < \frac{x_i+x_{i+1}}{2} \\ 0 & \text{otherwise} \end{cases}$$

# Piecewise Linear Interpolation



# Piecewise Linear “Hat” Basis



$$\psi_i(x) = \begin{cases} \frac{x-x_{i-1}}{x_i-x_{i-1}} & \text{when } x_{i-1} < x \leq x_i \\ \frac{x_{i+1}-x}{x_{i+1}-x_i} & \text{when } x_i < x \leq x_{i+1} \\ 0 & \text{otherwise} \end{cases}$$

# Observation

Extra differentiability is possible and may look nicer but can be undesirable.

# Multidimensional Problem

$$f : \mathbb{R}^n \rightarrow \mathbb{R}$$

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**Given:**  $\vec{x}_i \mapsto y_i$

# Nearest-Neighbor Interpolation

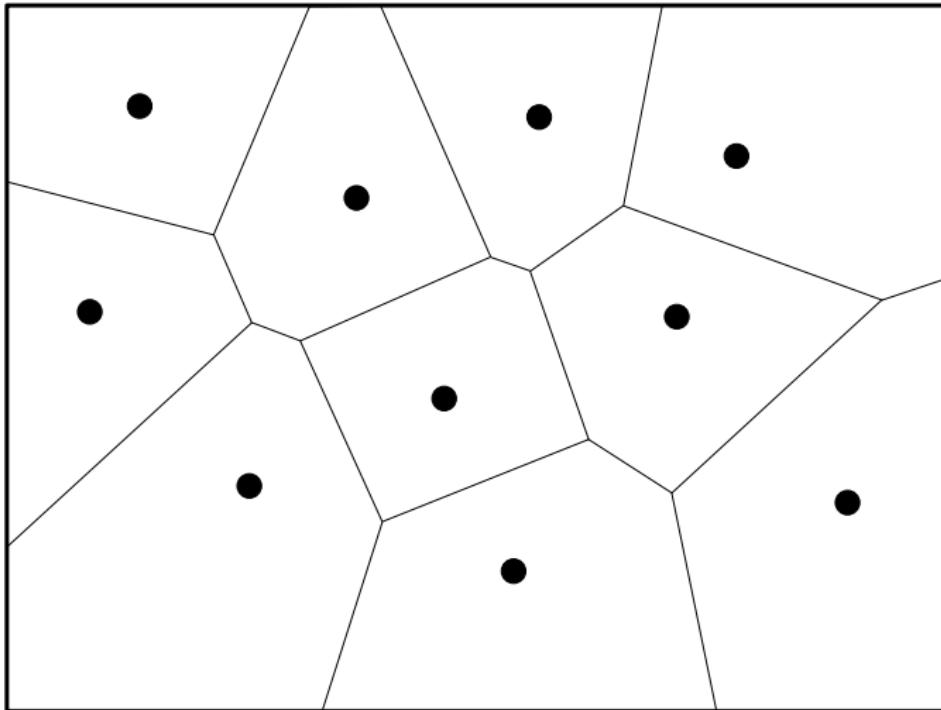
## Definition (Voronoi cell)

Given  $S = \{\vec{x}_1, \vec{x}_2, \dots, \vec{x}_k\} \subseteq \mathbb{R}^n$ , the *Voronoi cell* corresponding to  $\vec{x}_i$  is

$$V_i \equiv \{\vec{x} : \|\vec{x} - \vec{x}_i\|_2 < \|\vec{x} - \vec{x}_j\|_2 \text{ for all } j \neq i\}.$$

That is, it is the set of points closer to  $\vec{x}_i$  than to any other  $\vec{x}_j$  in  $S$ .

# Voronoi Cells



# Barycentric Interpolation

$n + 1$  points in  $\mathbb{R}^n$

$$\sum_i a_i \vec{x}_i = \vec{x}$$

$$\sum_i a_i = 1$$

# Barycentric Interpolation

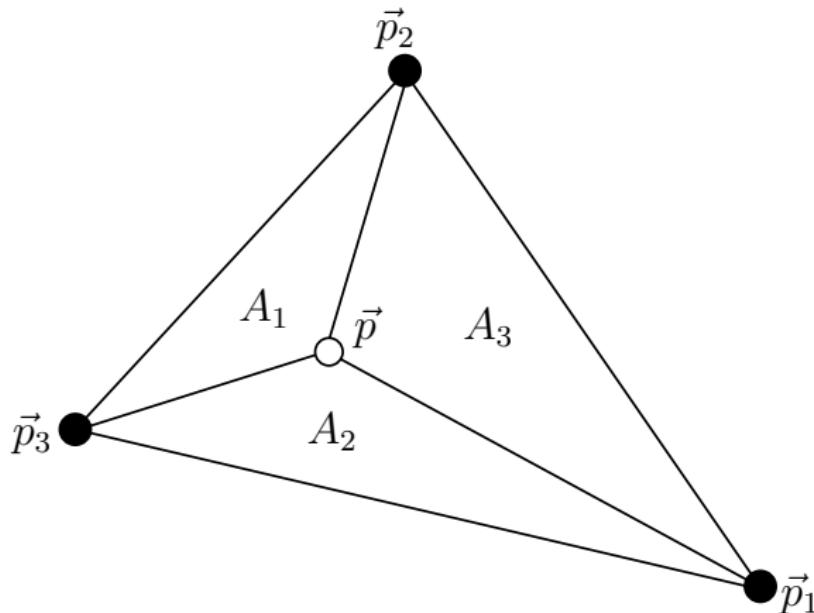
$n + 1$  points in  $\mathbb{R}^n$

$$\sum_i a_i \vec{x}_i = \vec{x}$$

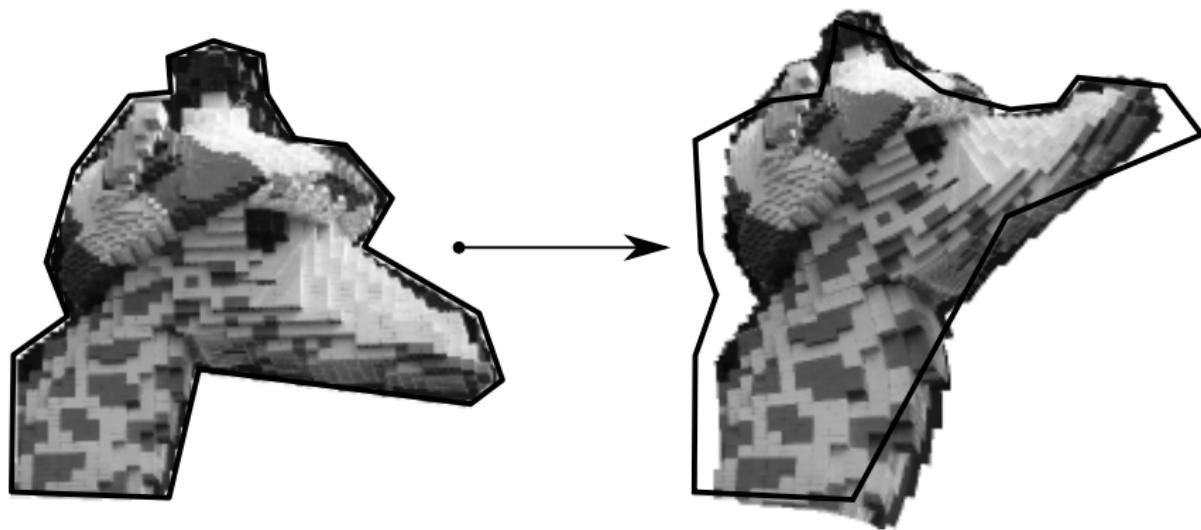
$$\sum_i a_i = 1$$

$$\longrightarrow f(\vec{x}) = \sum_i a_i(\vec{x}) y_i$$

# Interpretation in 2D

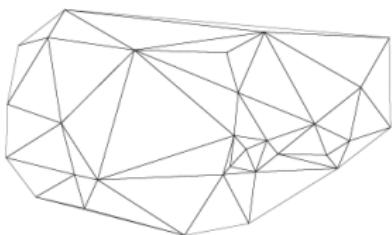


# Generalized Barycentric Coordinates

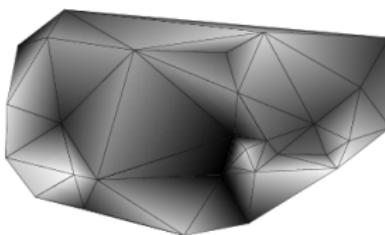


<http://www.eng.biu.ac.il/weberof/publications/complex-coordinates/>

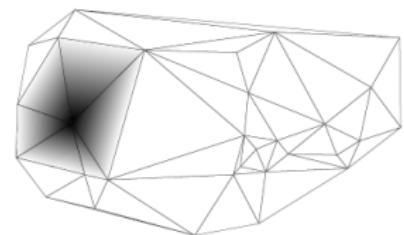
# Localized Barycentric Interpolation: Triangle Hat Functions



(a) Triangle mesh

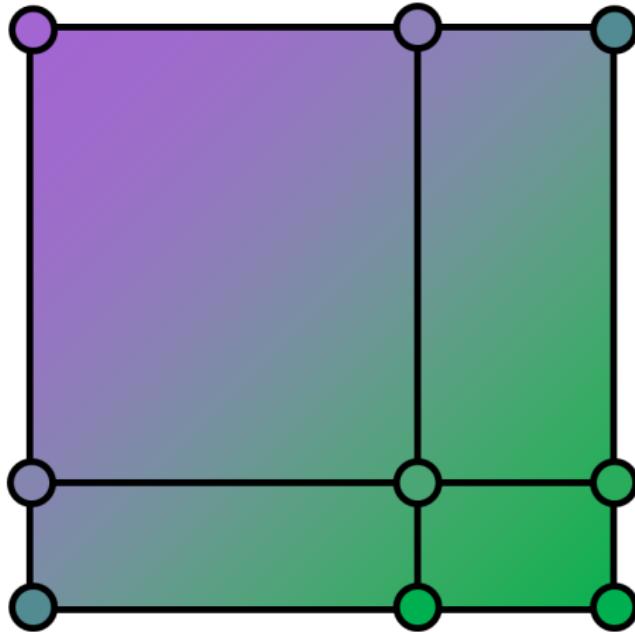


(b) Barycentric interpolation



(c) Hat function

# Interpolation on a Grid



# Linear Algebra of Functions

$$\langle f, g \rangle \equiv \int_a^b f(x)g(x) dx$$

# Measures “overlap” of functions!

# Orthogonal Polynomials

- ▶ **Legendre:** Apply Gram-Schmidt to  $1, x, x^2, x^3, \dots$
  - ▶ **Chebyshev:** Same, with weighted inner product

$$w(x) = \frac{1}{\sqrt{1-x^2}}$$

Nice oscillatory properties; minimizes ringing.

## Question

What is the *least-squares* approximation of  $f$  in a set of polynomials?

# Piecewise Polynomial Error

- ## ► Piecewise constant:

$$O(\Delta x)$$

- ### ► Piecewise linear:

$$O(\Delta x^2)$$