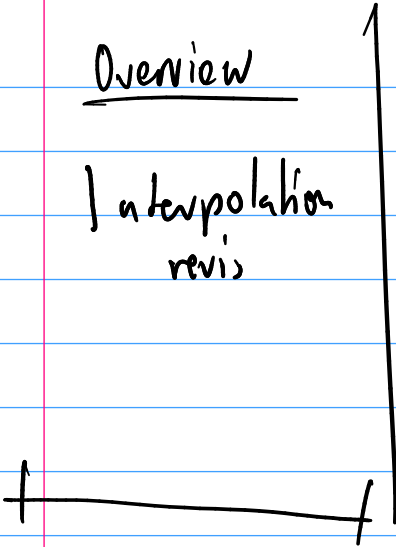


Overview

Interpolation  
revis



## Recap: Interpolation

Starting point: Looking for a linear combination of functions  $\varphi_i$  to hit given data points  $(x_i, y_i)$ .

Interpolation becomes solving the linear system:

$$y_i = f(x_i) = \sum_{j=0}^{N_{\text{func}}} \alpha_j \underbrace{\varphi_j(x_i)}_{V_{ij}} \quad \Leftrightarrow \quad V\alpha = \mathbf{y}.$$

Want unique answer: Pick  $N_{\text{func}} = N \rightarrow V$  square.

$V$  is called the (generalized) Vandermonde matrix.

Main lesson:

$$\underbrace{V}_{\text{coefficients}} = \text{(values at nodes)}.$$

$\ll \epsilon$

# Rethinking Interpolation

We have so far always used monomials  $(1, x, x^2, x^3, \dots)$  and equispaced points for interpolation. It turns out that this has *significant problems*.

**Demo:** Monomial interpolation

## **Demo:** Choice of Nodes for Polynomial Interpolation

# Interpolation: Choosing Basis Function and Nodes

Both function basis and point set are under our control. What do we pick?

Ideas for basis functions:

- ▶ Monomials  $1, x, x^2, x^3, x^4, \dots$
- ▶ Functions that make  $V = I \rightarrow$  'Lagrange basis'
- ▶ Functions that make  $V$  triangular  $\rightarrow$  'Newton basis'
- ▶ Splines (piecewise polynomials)
- ▶ Orthogonal polynomials
- ▶ Sines and cosines
- ▶ 'Bumps' ('Radial Basis Functions')

Ideas for nodes:

- ▶ Equispaced
- ▶ 'Edge-Clustered' (so-called Chebyshev/Gauss/... nodes)

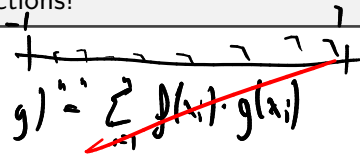
## Better Conditioning: Orthogonal Polynomials

What caused monomials to have a terribly conditioned Vandermonde?

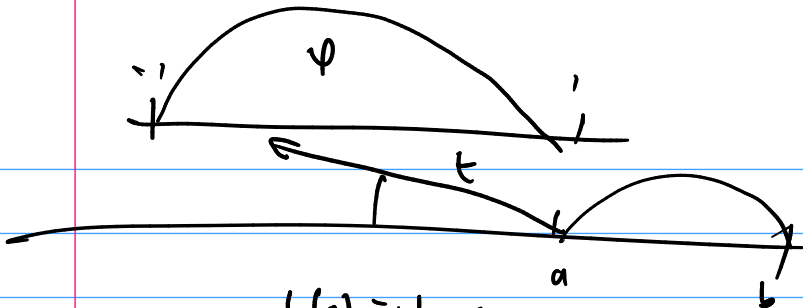
What's a way to make sure two vectors are *not* like that?

orthogonal

But polynomials are functions!


$$\cancel{(f, g) = \sum_{i=1}^n f(x_i) \cdot g(x_i)}$$
$$\rightarrow (f, g) = \int_{-1}^1 f(x) g(x) dx$$

$$(f, g) = \int_{-1}^1 f(x) g(x) \frac{1}{\sqrt{1-x^2}} dx$$



$$t(a) = -1 \quad \leftarrow$$

$$t(b) = 1 \quad \leftarrow$$

$$t(x) = mx + n$$

$$m = \frac{2}{b-a}$$



## Better Conditioning: Orthogonal Polynomials (II)

But how can I practically compute the Legendre polynomials?

## Another Family of Orthogonal Polynomials **Chebyshev**

Three equivalent definitions:

- ▶ Result of Gram-Schmidt with weight  $1/\sqrt{1-x^2}$

What is that weight?

$$T_0, T_1, T_2, T_3$$

$$\rightarrow \triangleright T_k(x) = \cos(k \cos^{-1}(x)) \leftarrow$$

- ▶  $T_{k+1}(x) = 2xT_k(x) - T_{k-1}(x)$  ← recurrence relation

**Demo:** Chebyshev interpolation part I

What are good nodes to use with Chebyshev polynomials?

$$T_0, T_1$$

$$T_k = \cos(k \cdot \cos^{-1}(x_i))$$

$$x_i = \cos(\dots)$$

## Chebyshev Nodes

Might also consider zeros (instead of roots) of  $T_k$ :

$$x_i = \cos \left( \frac{2i+1}{2k} \pi \right) \quad (i = \underline{1}, \dots, \underline{k}).$$

The Vandermonde for these (with  $T_k$ ) can be applied in  $O(N \log N)$  time, too.

It turns out that we were still looking for a good set of interpolation nodes.

We came up with the criterion that the nodes should bunch towards the ends. Do these do that?

**Demo:** Chebyshev interpolation part II

## Calculus on Interpolants

Suppose we have an interpolant  $\tilde{f}(x)$  with  $f(x_i) = \tilde{f}(x_i)$  for  $i = 1, \dots, n$ :

$$\tilde{f}(x) = \alpha_1\varphi_1(x) + \dots + \alpha_n\varphi_n(x)$$

How do we compute the derivative of  $\tilde{f}$ ?

Suppose we have function values at nodes  $(x_i, f(x_i))$  for  $i = 1, \dots, n$  for a function  $f$ . If we want  $f'(x_i)$ , what can we do?

## About Differentiation Matrices

How could you find coefficients of the derivative?

Give a matrix that finds the second derivative.