

Interpolation

DATA POINTS

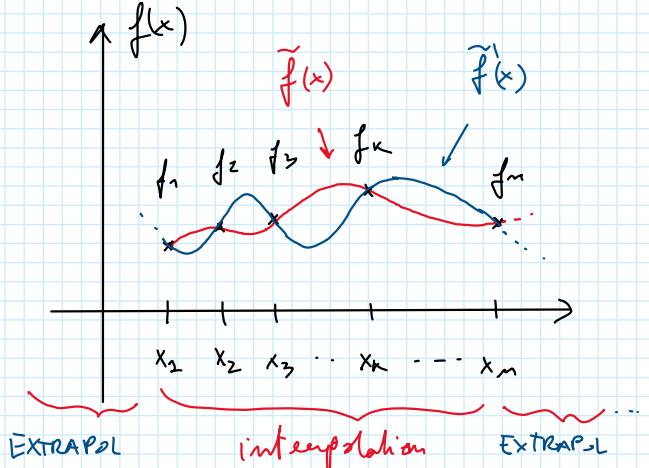
Def given $f(x)$, Known only at certain points x_1, x_2, \dots, x_m

$$f(x_1) = f_1, f(x_2) = f_2, \dots, f(x_m) = f_m$$

interpolation is the process of constructing $\tilde{f}(x)$ such that.

$$\tilde{f}(x_k) = f_k, \quad \forall k = 1, \dots, m \quad (1)$$

↑
DATA POINT
↓
 $f(x_k)$



→ $\tilde{f}(x)$ can be used to ESTIMATE

$$f(x) \quad \forall x \in [x_1, \dots, x_m]$$

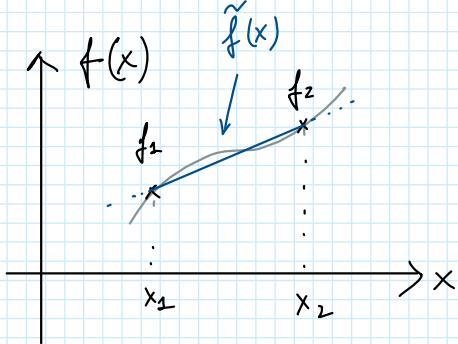
NOTE: $\tilde{f}(x)$ is not unique $\rightarrow \infty$ of $\tilde{f}(x)$ that satisfy the condition (1)

for uniqueness \Rightarrow RESTRICTION of $\tilde{f}(x)$

Polynomial interpolation

• restrict $\tilde{f}(x)$ polynomial of degree $m-1$ (m - DATA POINTS)

2 DATA POINTS



coefficients

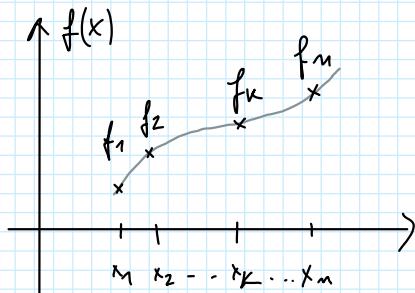
$$\tilde{f}(x) = a_0 + a_1 x \quad \leftarrow \text{Polynomial of order 1}$$

"find a_0 and a_1 such that"

$$\begin{cases} \tilde{f}(x_1) = a_0 + a_1 x_1 = f_1 \\ \tilde{f}(x_2) = a_0 + a_1 x_2 = f_2 \end{cases}$$

$$\begin{cases} f(x_1) = a_0 + a_1 x_1 &= f_1 \\ \tilde{f}(x_2) = a_0 + a_1 x_2 &= f_2 \end{cases}$$

m DATA POINTS



$$\tilde{f}(x) = a_0 + a_1 x + a_2 x^2 + a_3 x^3 + \dots + a_{m-1} x^{m-1}$$

"find a_0, a_1, \dots, a_{m-1} such that"

$$\begin{cases} \tilde{f}(x_1) = a_0 + a_1 x_1 + a_2 x_1^2 + \dots + a_{m-1} x_1^{m-1} = f_1 \\ \tilde{f}(x_2) = a_0 + a_1 x_2 + a_2 x_2^2 + \dots + a_{m-1} x_2^{m-1} = f_2 \\ \vdots \\ \tilde{f}(x_m) = a_0 + a_1 x_m + a_2 x_m^2 + \dots + a_{m-1} x_m^{m-1} = f_m \end{cases}$$

matrix form:

$$\left[\begin{array}{cccc} 1 & x_1 & x_1^2 & \dots \\ 1 & x_2 & x_2^2 & \dots \\ \vdots & \vdots & \vdots & \vdots \\ 1 & x_m & x_m^2 & \dots \end{array} \right] \left[\begin{array}{c} a_0 \\ a_1 \\ \vdots \\ a_{m-1} \end{array} \right] = \left[\begin{array}{c} f_1 \\ f_2 \\ \vdots \\ f_m \end{array} \right]$$

$\underbrace{\quad}_{[\mathbf{V}]}$ VANDERMONDE MATRIX

\Rightarrow solve $[\mathbf{V}] \{a\} = \{f\} \Rightarrow$ find $[\mathbf{V}^{-1}]$ inverse of $[\mathbf{V}]$

$$[\mathbf{V}^{-1}] [\mathbf{V}] \{a\} = [\mathbf{V}^{-1}] \{f\}$$

$$\underbrace{\quad}_{\begin{bmatrix} I_d \\ \mathbf{0} \end{bmatrix}} \rightarrow \underbrace{\begin{bmatrix} I_d \\ \mathbf{0} \end{bmatrix}}_{\{a\}} \quad \downarrow \quad \begin{bmatrix} f_1 \\ f_2 \\ \vdots \\ f_m \end{bmatrix}$$

$$[V^{-1}] \begin{Bmatrix} \{a\} \end{Bmatrix} \rightarrow \begin{Bmatrix} \{f_m\} \end{Bmatrix}$$

$$\{a\} = [V^{-1}] \{f\}$$

$[V]$ is invertible if $\text{Det}[V] \neq 0 \Rightarrow [V]$ is linearly independent. ✓

numerically.

$[V]$ often ILL-CONDITIONED matrix

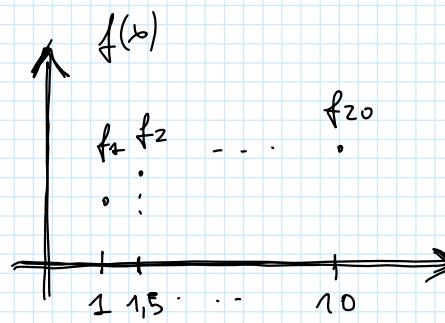
EXAMPLE: $f(x) \rightarrow \tilde{f}(x)$

$$m = 20$$

$$x_1 = 1$$

$$x_{20} = 10$$

$$[V] = \begin{bmatrix} 1 & 1 & 1^2 & \dots & 1^{19} \\ 1 & 1.5 & 1.5^2 & \dots & 1.5^{19} \\ \vdots & & \vdots & & \vdots \\ 1 & 10 & 10^2 & \dots & 10^{19} \end{bmatrix}$$



v_k - k-th entry of $[v]$

$$\frac{\min(v_k)}{\max(v_k)} = \frac{1}{10^{19}} = 10^{-19}$$

$$\epsilon \sim 2,2 \cdot 10^{-16}$$

\Rightarrow first rows are "AS IF" they were zero)



MATRIX is "numerically" SINGULAR

K

CONDITION number
(of a matrix)

unified measure of the sensitivity of the solution of $[A]\{x\} = \{b\}$ to small perturbations in $[A]$ or $\{b\}$

↑
internal perturbations
(roundoff errors)

↗
external perturbations

ILL-conditioned linear system
huge variations in $\{x\}$ for small variations in $\{b\}$

Def: $K([A]) = \|A\|_2 \|A^{-1}\|_2$

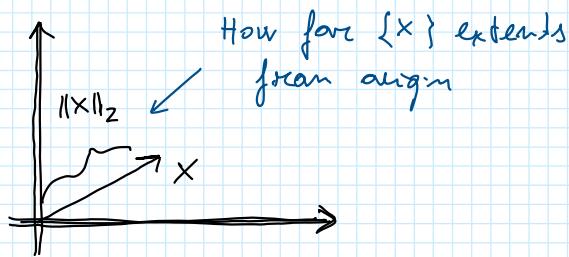
A

How does $\{x\}$ extend

$$\text{Def: } \kappa([A]) = \|A\|_2 \|A^{-1}\|_2$$

2-norm (EUCLIDEAN NORM)
of a VECTOR

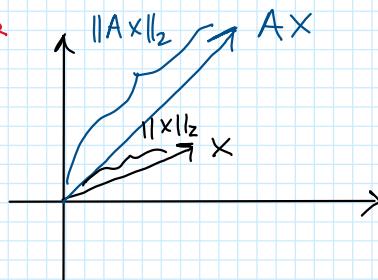
$$\|x\|_2 = \sqrt{\sum_{k=1}^m x_k^2}$$



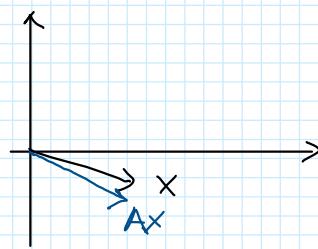
2-norm of a matrix Ax = vector

$$\|A\|_2 = \max \left(\frac{\|Ax\|_2}{\|x\|_2} \right) = M$$

STRETCHING of
x induced by A



$$\frac{1}{\|A^{-1}\|_2} = \min \left(\frac{\|Ax\|_2}{\|x\|_2} \right) = m$$



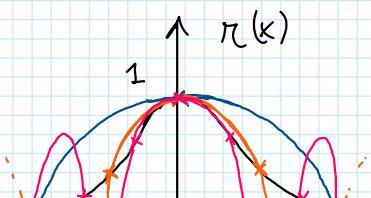
$$\kappa = \underbrace{\|A\|_2}_{M} \cdot \underbrace{\|A^{-1}\|_2}_{m} = \frac{M}{m} \leftarrow \text{maximum stretching of } x \text{ induced by } [A]$$

if $\kappa \sim 1 \Rightarrow$ system is STABLE with respect to perturbations \Rightarrow well-cond.

$\kappa \gg 1 \Rightarrow$ unstable \Rightarrow ill-cond.

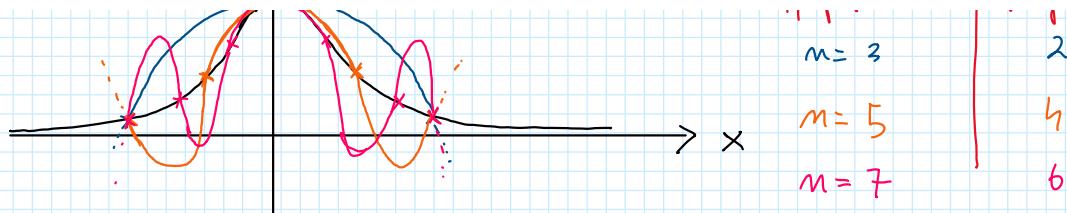
Polynomial interpolation is oscillatory

$$\text{Ex - RUNGE function } r(x) = \frac{1}{1 + \alpha x^2}, \quad \alpha = 25$$



Polynomial interp

# points	Degree
$m = 3$	2



if we increase the polynomial order \Rightarrow larger oscillations
near the boundaries of
the sampling interval

PIECEWISE LINEAR INTERPOLATION (1D)

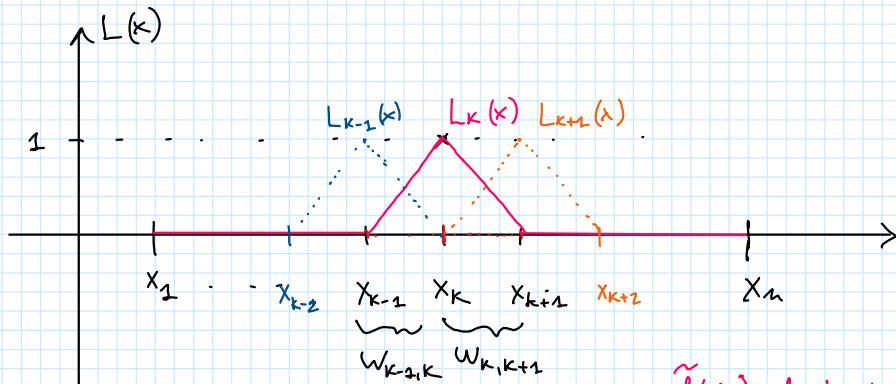
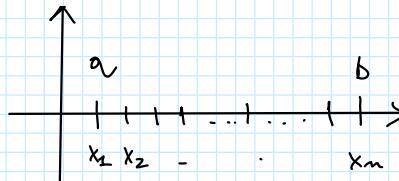
IDEA: construct an interpolating function $\tilde{f}(x)$ in $[a, b]$
from a SET of PIECEWISE-LINEAR POLYNOMIALS
"HAT" FUNCTIONS (BASIS FUNCTIONS)

$$\tilde{f}(x) = L_1(x)f_1 + L_2(x)f_2 + \dots + L_K(x)f_K + \dots + L_n(x)f_n$$

↑ ↑

each polynomial is $\neq 0$ only on a
SUBSET of $[a, b]$

STRATEGY: introduce set of
 n -points (nodes) to
subdivide $[a, b]$ $n-1$ ELEMENTS
(INTERVALS)



$$W_{k,k+1} = [x_k, x_{k+1}]$$

$$W_{k-1,k} = [x_{k-1}, x_k]$$

$$\Omega_k = W_{k-1,k} \cup W_{k,k+1}$$

REQUIREMENTS:

$$\tilde{f}(x_k) = f_k \quad L_k(x_k) + \underbrace{f_{k+1} L_{k+1}(x_k)}_{=1} + \dots + f_{n-1} L_{n-1}(x_k) = f_k$$

must be $= 0$
to have $\tilde{f}(x_k) = f_k$

$$\begin{cases} L_k(x_k) = 1 \\ L_k(x) = 0, \forall x \notin \Omega_k \\ L_k(x) \text{ linear} \end{cases}$$