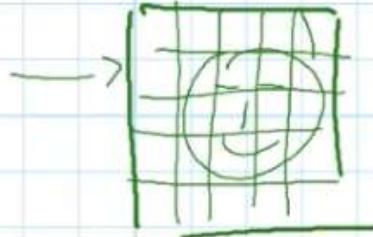


Lineare Ausgleichsrechnung



$$Ax = s$$

Zeile $\hat{=}$ Strahl

$\dim(x)$ $\hat{=}$ Anzahl der Kästchen

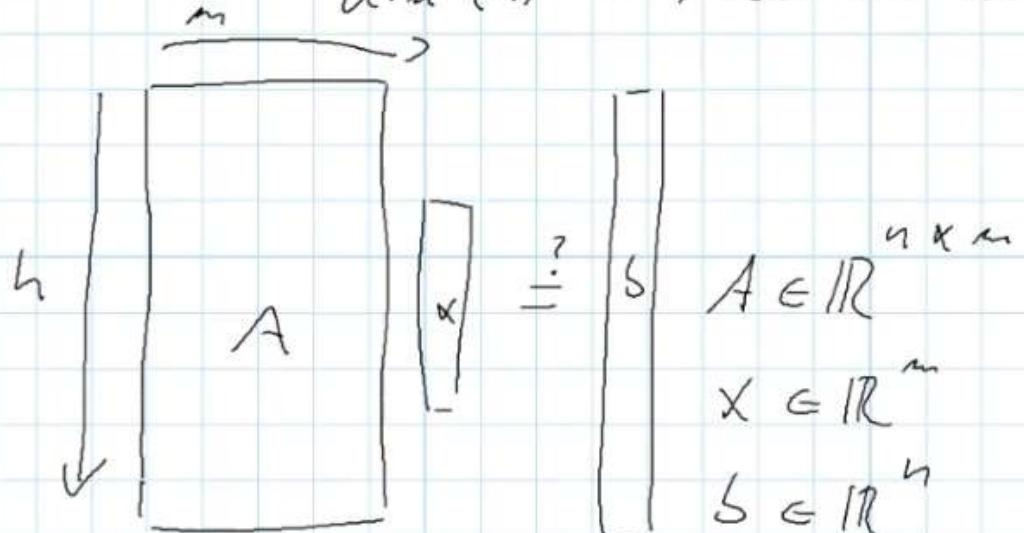
$$r = Ax - b +$$

$$\|r\|_2 = \sqrt{r_0^2 + r_1^2 + \dots + r_{n-1}^2}$$

$$\arg \min_{x \in \mathbb{R}^m} \|r\|_2$$

(=)

$$\arg \min_{x \in \mathbb{R}^m} \|r\|_2^2$$



$\hat{A}x \neq s$ $A \in \mathbb{R}^{n \times m}$, $s \in \mathbb{R}^n$, $x \in \mathbb{R}^m$

$$r = Ax - s$$

$$\underset{x \in \mathbb{R}^m}{\text{arg min}} \|Ax - s\|_2^2 = \|r\|_2^2$$

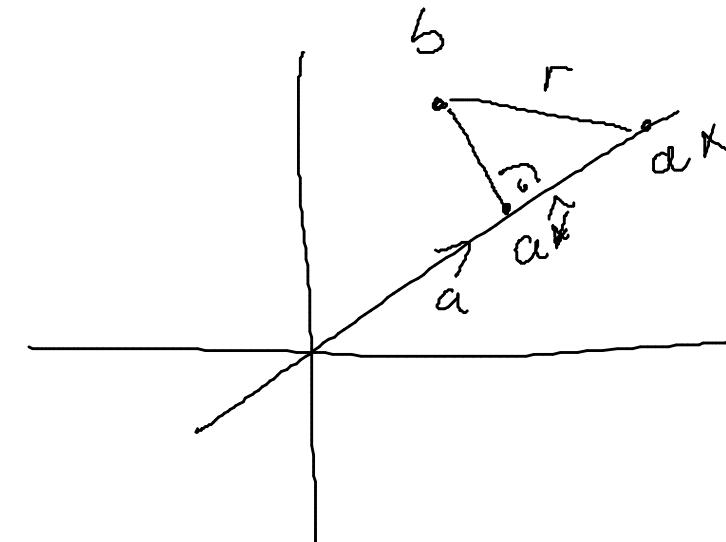
Gerade $a \in \mathbb{R}^n$, $a \in \mathbb{R}^2$, $x \in \mathbb{R}$, $s \in \mathbb{R}^2$

$$\begin{aligned}\|r\|_2^2 &= \|a\hat{x} - s\|_2^2 + \|a\hat{x} - ax\|_2^2 \\ &= \|a\hat{x} - s\|_2^2 + \|a(\hat{x} - x)\|_2^2 \quad \Rightarrow \quad x = \hat{x}\end{aligned}$$

$$0 = a^T r = a^T (Ax - s) = a^T a x - a^T s \quad (\Rightarrow)$$

$$a^T a x = a^T s \quad \rightarrow \quad x = \frac{a^T s}{a^T a}$$

Gilt für alle n !



$$A^T A \alpha = A^T s \quad \text{Normalengleichung}$$

$$\alpha = (A^T A)^{-1} A^T s$$

$$A\alpha = \underbrace{A(A^T A)^{-1} A^T}_{\text{Projektor}} s$$

Projektor

$$P = A(A^T A)^{-1} A^T$$

$$P^2 = P$$

$$A^T (A\alpha - s) = 0$$

$$\begin{pmatrix} a_0 & \vdots \\ a_1 & \vdots \\ \vdots & \vdots \end{pmatrix} (A\alpha - s) = \begin{pmatrix} 0 \\ 0 \\ \vdots \end{pmatrix}$$

Eigenschaften von $A^T A$

Symmetrische $(A^T A)^T = A^T A^T = A^T A$

Positiv semidefinit
(PSD) $x^T A^T A x = (Ax)^T Ax = y^T y = \|y\|^2 \geq 0$

A hat vollen Spaltenrang $\Rightarrow A^T A$ regulär
= Spalten linear unabhängig $\hat{\Leftrightarrow} \det(A^T A) \neq 0$

$$A^T A x \neq 0 \quad \Leftrightarrow \quad x \neq 0 \\ \Rightarrow x^T A^T A x > 0 \quad \text{Positiv definiert (PD)}$$

Eigenschaften von PD-Matrizen $\beta = A^T \tilde{A}$

$$(0, \dots, 0, 1, 0, \dots, 0) \beta \begin{pmatrix} 0 \\ \vdots \\ 0 \\ 1 \\ \vdots \\ 0 \end{pmatrix} > 0 \Rightarrow s_{ii} > 0$$

$$(x_0, \dots, x_i, 0, x_{i+1}, \dots) \beta \begin{pmatrix} x_0 \\ \vdots \\ x_i \\ 0 \\ \vdots \\ x_{i+1} \end{pmatrix} > 0 \Rightarrow$$

Löschen von Spalte/Zeile i
erhält PD

$$\left(\begin{array}{cccc|c} 1 & * & * & * & \\ * & 1 & * & * & \\ * & * & 1 & * & \\ * & * & * & 1 & \\ \hline 0 & * & * & * & 0 \end{array} \right)$$

β ist PD dann auch $s \cdot \beta$ für $s > 0$

A, B sind PD

" $(A+B)$

$$x^T(A+B)x = x^T A x + x^T B x > 0 > 0$$

$\Rightarrow sA + tB$ PD für $s, t > 0$

$$\det(\beta) > 0$$

$$\det(I(1-\epsilon) + \beta \cdot \epsilon) \neq 0 \text{ für } \epsilon \in [0, 1]$$

PD

keine Nullstellen in $[0, 1]$

Cholesky

$$\beta = A^T A$$

$$\beta = L L^T$$

$$\begin{matrix} b_{00} & b_{10} & b_{20} & \dots \\ b_{10} & b_{11} & b_{21} & \\ b_{20} & b_{21} & b_{22} & \\ \vdots & \vdots & \vdots & \end{matrix} = \left(\begin{array}{cccc} l_{00} & 0 & 0 & \dots \\ l_{10} & l_{11} & 0 & \\ l_{20} & l_{21} & l_{22} & \\ \vdots & \vdots & \vdots & \end{array} \right) \left(\begin{array}{cccc} l_{00} & l_{10} & l_{20} & \dots \\ 0 & l_{11} & l_{21} & \dots \\ 0 & 0 & l_{22} & \dots \\ \vdots & \vdots & \vdots & \end{array} \right) = \begin{matrix} l_{00} & l_{00} \cdot l_{10} & l_{00} l_{20} & l_{00} l_{30} \\ (l_{10} \cdot l_{00}) l_{11}^2 & l_{10}^2 + l_{11}^2 & l_{10} l_{20} + l_{11} l_{21} & \dots \\ \dots & \dots & \dots & \dots \end{matrix}$$

$$s_{00} = l_{00}^2 \rightarrow l_{00} = \sqrt{s_{00}}$$

$$s_{10} = l_{00} \cdot l_{10} \rightarrow l_{10} = \frac{s_{10}}{l_{00}}$$

$$s_{11} = l_{10}^2 + l_{11}^2 \rightarrow l_{11} = \sqrt{s_{11} - l_{10}^2} \quad 0 < s_{11} - l_{10}^2 = s_{11} - \frac{s_{10}^2}{l_{00}^2} = s_{11} - \frac{s_{10}^2}{s_{00}}$$

$$b_{20} = l_{10} l_{00} + l_{20} l_{11} \rightarrow \quad \Rightarrow \quad s_{00} s_{11} - s_{10}^2 > 0$$

$$l_{20} = \frac{s_{20} - l_{10} l_{11}}{l_{11}}$$

$$l_{ii} = \sqrt{s_{ii} - \sum_{k=0}^{i-1} l_{ik}^2}$$

$$l_{ij} = \frac{1}{l_{ii}} \left(s_{ij} - \sum_{k=0}^{i-1} l_{ik} l_{jk} \right)$$

$$A x \neq s \rightarrow A^T A x = A^T s \quad \xrightarrow{\text{Cholesky}} \quad L L^T x = c \quad (= A^T s)$$

$$y = L^T x \rightarrow Ly = c \quad \text{Vorwärtsch.} \rightarrow y$$

$$y \rightarrow L^T x = y \quad \text{Rückwärtseins.} \rightarrow x$$

$$\begin{aligned} & (a_0 x_0 - s)^2 + \\ & + (a_1 x_1 - s) \end{aligned}$$

$$(a_0 x_0 + c_0 - s)^2 + (a_1 x_1 + c_1 - s)^2$$

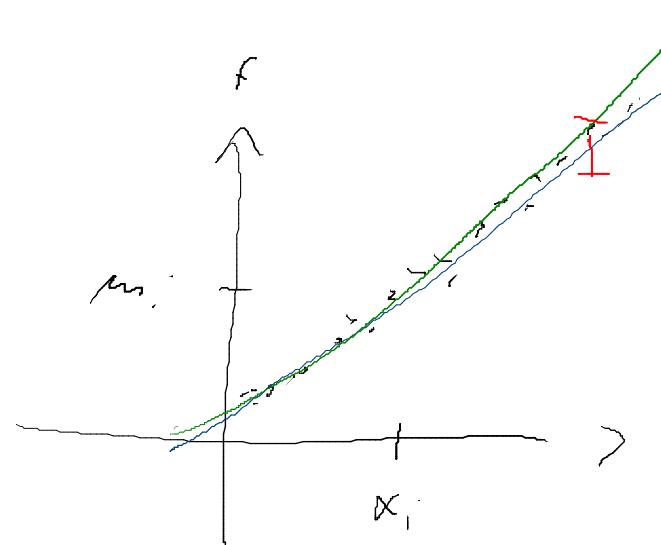
$$\beta = U^T D V \quad U^T U = I$$

$$\beta^{-1} = U^T D^{-1} V$$

$$\beta \beta^{-1} = U^T D V U^T D^{-1} V$$

\tilde{U}

Lineare Regression



Gesucht: $f(x_i) = m_i$

$$\underset{f \in \Theta}{\arg \min} \sum_i (f(x_i) - m_i)^2$$

$$f(x) = c_0 + c_1 \cdot x + c_2 \cdot x^2$$

$$m_0 = f(x_0) = c_0 + c_1 \cdot x_0 + c_2 \cdot x_0^2$$

$$m_1 = f(x_1) = c_0 + c_1 \cdot x_1 + c_2 \cdot x_1^2$$

}

$$\left(\begin{array}{ccc|c} 1 & x_0 & x_0^2 \\ 1 & x_1 & x_1^2 \\ 1 & x_2 & x_2^2 \end{array} \right) \left(\begin{array}{c} c_0 \\ c_1 \\ c_2 \end{array} \right) = \left(\begin{array}{c} m_0 \\ m_1 \\ m_2 \end{array} \right)$$

$$X^T X c = X^T m$$

$$f(x) = \sum c_i \beta_i(x)$$

$$\beta_i(x) = 1, x, x^2, \sin(x), e^{-x}$$