

Peter Lorenz

$$1.1) f(x) = \frac{1}{2} (\log(x))^2, \quad \text{domain } f = \{x | x > 0\}$$

$$f'(x) = 2 \cdot \frac{1}{2} (\log(x))^1 \cdot \frac{1}{x}$$
$$= \cancel{\log(x)} \cdot \frac{1}{x}$$

$\underbrace{\log(x)}_u \quad \underbrace{\frac{1}{x}}_v$

$$u' = \frac{1}{x}$$
$$v' = -x^{-2}$$

Produktregel:  $u \cdot v' + u' \cdot v$ :

$$= \log(x) \cdot (-x^{-2}) + \frac{1}{x} \cdot \frac{1}{x} = -\frac{\log(x)}{x^2} + \frac{1}{x^2} =$$
$$= \frac{1}{x^2} (1 - \log(x))$$

$$f'(x) \stackrel{!}{=} 0$$
$$\frac{1}{x^2} (1 - \log(x)) = 0 \quad | \cdot x^2$$
$$1 - \log(x) = 0 \quad | +1 \quad | \cdot (-1)$$
$$\log(x) = 1$$
$$e^{\log(x)} = e^1$$
$$\underline{\underline{x = e^1}}$$

A: Diese Funktion ist nicht konvex, da sie ab einer der Nullstelle in den Minusbereich geht.





$$1.2) \quad f(x) = x_1 \cdot x_2, \quad x \in \mathbb{R}^2$$

$$H_f(x) := \begin{pmatrix} \frac{\partial^2 f}{\partial x_1 \partial x_1}(x) & \frac{\partial^2 f}{\partial x_1 \partial x_2}(x) \\ \frac{\partial^2 f}{\partial x_2 \partial x_1} & \frac{\partial^2 f}{\partial x_2 \partial x_2} \end{pmatrix} = \begin{pmatrix} 0 & 1 \\ 1 & 0 \end{pmatrix}$$

Wenn die Hessematrix semi positiv definit ist, dann ist die Funktion konvex.

$\det(H_f(x)) < 0$  ... dann ist die Funktion  $f$  indefinit

$$\det(H_f(x)) = (0 \cdot 0) - (1 \cdot 1) = -1 \quad \text{indefinit und daher nicht konvex.}$$

$$1.3) \quad f(x) = \frac{x_1^2}{x_2}, \quad x \in \mathbb{R}^2, \quad x_2 > 0$$

$$\nabla f(x) = \begin{pmatrix} 2 \frac{x_1}{x_2} \\ -1 \cdot \frac{x_1^2}{x_2^2} \end{pmatrix}$$

$$\nabla^2 f(x) = \begin{pmatrix} \frac{2}{x_2} & -2 \cdot \frac{x_1}{x_2^2} \\ -2 \cdot \frac{x_1}{x_2^2} & 2 \cdot \frac{x_1^2}{x_2^3} \end{pmatrix} = \begin{pmatrix} \frac{1}{x_2} & -\frac{x_1}{x_2^2} \\ -\frac{x_1}{x_2^2} & \frac{x_1^2}{x_2^3} \end{pmatrix}$$

$$\begin{aligned} \det(\nabla^2 f(x)) &= \frac{1}{x_2} \cdot \left( \frac{x_1^2}{x_2^3} \right) - \left( \left( -\frac{x_1}{x_2^2} \right) \cdot \left( -\frac{x_1}{x_2^2} \right) \right) = \\ &= \frac{x_1^2}{x_2^4} - \frac{x_1^2}{x_2^4} = 0 \checkmark \end{aligned}$$

Dies heißt Die Funktion  $f$  ist konvex.





$$2.1) \text{ 2.1) } f(x) = \frac{1}{2} \|A \cdot x - b\|^2, \quad x \in \mathbb{R}^n, \quad b \in \mathbb{R}^m, \quad A \in \mathbb{R}^{m \times n}$$

$$a) \quad m = n = 2$$

$$\begin{aligned} f(x) &= \frac{1}{2} \left\| \begin{pmatrix} a_{11} & a_{12} \\ a_{21} & a_{22} \end{pmatrix} \cdot \begin{pmatrix} x_1 \\ x_2 \end{pmatrix} - \begin{pmatrix} b_1 \\ b_2 \end{pmatrix} \right\|^2 = \\ &= \frac{1}{2} \left\| \begin{pmatrix} a_{11}x_1 + a_{12}x_2 \\ a_{21}x_1 + a_{22}x_2 \end{pmatrix} - \begin{pmatrix} b_1 \\ b_2 \end{pmatrix} \right\|^2 = \\ &= \frac{1}{2} \left( \sqrt{(a_{11}x_1 + a_{12}x_2 - b_1)^2 + (a_{21}x_1 + a_{22}x_2 - b_2)^2} \right)^2 = \end{aligned}$$

$$\begin{aligned} \frac{\partial}{\partial x_1} f(x) &= \frac{1}{2} \left( 2(a_{11}x_1 + a_{12}x_2 - b_1) \cdot a_{11} + 2(a_{21}x_1 + a_{22}x_2 - b_2) a_{21} \right) \\ &= (a_{11}x_1 + a_{12}x_2 - b_1) a_{11} + (a_{21}x_1 + a_{22}x_2 - b_2) a_{21} \end{aligned}$$

$$\frac{\partial}{\partial x_2} f(x) = (a_{11}x_1 + a_{12}x_2 - b_1) a_{12} + (a_{21}x_1 + a_{22}x_2 - b_2) a_{22}$$

$$\begin{aligned} \nabla f(x) &= \begin{pmatrix} \frac{\partial}{\partial x_1} f(x) \\ \frac{\partial}{\partial x_2} f(x) \end{pmatrix} = \begin{pmatrix} a_{11} & a_{12} \\ a_{21} & a_{22} \end{pmatrix} \begin{pmatrix} a_{11}x_1 + a_{12}x_2 - b_1 \\ a_{21}x_1 + a_{22}x_2 - b_2 \end{pmatrix} \\ &= \begin{pmatrix} a_{11} & a_{12} \\ a_{21} & a_{22} \end{pmatrix} \begin{pmatrix} a_{11}x_1 + a_{12}x_2 - b_1 \\ a_{21}x_1 + a_{22}x_2 - b_2 \end{pmatrix} \\ &= A^T \cdot (A \cdot x - b) \end{aligned}$$

$$\begin{aligned}\nabla^2 f(x) &= A^T(A \cdot x - b) \\ &= A^T \cdot A \cdot x - A^T \cdot b\end{aligned}$$

$$\frac{\partial^2}{\partial x_1^2} f(x) = \begin{pmatrix} a_{11} & a_{12} \end{pmatrix} \cdot \begin{pmatrix} a_{11} \\ a_{12} \end{pmatrix}$$

$$\frac{\partial^2}{\partial x_2^2} f(x) = \begin{pmatrix} a_{12} & a_{22} \end{pmatrix} \cdot \begin{pmatrix} a_{12} \\ a_{22} \end{pmatrix}$$

$$\frac{\partial^2}{\partial x_1 \partial x_2} f(x) = \begin{pmatrix} a_{11} & a_{12} \end{pmatrix} \cdot \begin{pmatrix} a_{12} \\ a_{22} \end{pmatrix}$$

$$\frac{\partial^2}{\partial x_2 \partial x_1} f(x) = \begin{pmatrix} a_{12} & a_{22} \end{pmatrix} \cdot \begin{pmatrix} a_{11} \\ a_{12} \end{pmatrix}$$

$$\begin{aligned}\nabla^2 f(x) &= \begin{pmatrix} \frac{\partial^2}{\partial x_1^2} & \frac{\partial^2}{\partial x_1 \partial x_2} \\ \frac{\partial^2}{\partial x_2 \partial x_1} & \frac{\partial^2}{\partial x_2^2} \end{pmatrix} = \begin{pmatrix} \begin{pmatrix} a_{11} & a_{12} \end{pmatrix} \begin{pmatrix} a_{11} \\ a_{12} \end{pmatrix} & \begin{pmatrix} a_{11} & a_{12} \end{pmatrix} \begin{pmatrix} a_{12} \\ a_{22} \end{pmatrix} \\ \begin{pmatrix} a_{12} & a_{22} \end{pmatrix} \begin{pmatrix} a_{12} \\ a_{22} \end{pmatrix} & \begin{pmatrix} a_{12} & a_{22} \end{pmatrix} \begin{pmatrix} a_{11} \\ a_{12} \end{pmatrix} \end{pmatrix} = \\ &= \begin{pmatrix} \begin{pmatrix} a_{11} & a_{12} \end{pmatrix} \begin{pmatrix} a_{11} & a_{12} \\ a_{12} & a_{22} \end{pmatrix} \\ \begin{pmatrix} a_{12} & a_{22} \end{pmatrix} \begin{pmatrix} a_{11} & a_{12} \\ a_{12} & a_{22} \end{pmatrix} \end{pmatrix} = \\ &= \underline{\underline{A^T A}}\end{aligned}$$

Generalised:  $f(x) = \frac{1}{2} \|A \cdot x - b\|^2$

$$\nabla f(x) = \frac{1}{2} \cdot 2 \cdot A^T (A \cdot x - b)$$

$$\begin{aligned}\nabla^2 f(x) &= A^T (A \cdot x - b) \\ &= A^T \cdot A \cdot x - A^T \cdot b \\ &= ((A^T \cdot A))^T \cdot x - A^T \cdot b \\ &= (A^T \cdot A)^T = \underline{\underline{A^T \cdot A}}\end{aligned}$$



$$2.2) f(x) = \frac{1}{2} \| A(x + \lambda y) - b \|^2$$

$$f(\lambda) = \frac{1}{2} \left\| \begin{pmatrix} a_{11} & a_{12} \\ a_{21} & a_{22} \end{pmatrix} \left( \begin{pmatrix} x_1 \\ x_2 \end{pmatrix} + \lambda \cdot \begin{pmatrix} y_1 \\ y_2 \end{pmatrix} \right) - \begin{pmatrix} b_1 \\ b_2 \end{pmatrix} \right\|^2 =$$

$$= \frac{1}{2} \left\| \begin{pmatrix} a_{11} & a_{12} \\ a_{21} & a_{22} \end{pmatrix} \begin{pmatrix} x_1 + \lambda y_1 \\ x_2 + \lambda y_2 \end{pmatrix} - \begin{pmatrix} b_1 \\ b_2 \end{pmatrix} \right\|^2 =$$

$$= \frac{1}{2} \left\| \begin{pmatrix} a_{11}(x_1 + \lambda y_1) + a_{12}(x_2 + \lambda y_2) - b_1 \\ a_{21}(x_1 + \lambda y_1) + a_{22}(x_2 + \lambda y_2) - b_2 \end{pmatrix} \right\|^2 =$$

$$\frac{\partial}{\partial \lambda} f(\lambda) = \frac{1}{2} \left( \frac{(a_{11}(x_1 + \lambda y_1) + a_{12}(x_2 + \lambda y_2) - b_1)^2}{(a_{21}(x_1 + \lambda y_1) + a_{22}(x_2 + \lambda y_2) - b_2)^2} + \dots \right) =$$

$$= \frac{1}{2} ((a_{11}(x_1 + \lambda y_1) + a_{12}(x_2 + \lambda y_2) - b_1)^2 + (a_{21}(x_1 + \lambda y_1) + a_{22}(x_2 + \lambda y_2) - b_2)^2)$$

$$= \frac{1}{2} (2 \left( \frac{a_{11} y_1 + a_{12} y_2}{a_{21} y_1 + a_{22} y_2} \right) + 2 \left( \frac{a_{21} y_1 + a_{22} y_2}{a_{21} y_1 + a_{22} y_2} \right)) =$$

$$= \left( \frac{(a_{11} y_1 + a_{12} y_2) \cdot (a_{11}(x_1 + \lambda y_1) + a_{12}(x_2 + \lambda y_2) - b_1)}{(a_{21} y_1 + a_{22} y_2) \cdot (a_{21}(x_1 + \lambda y_1) + a_{22}(x_2 + \lambda y_2) - b_2)} \right) =$$

$$= \left( \begin{pmatrix} a_{11} & a_{12} \end{pmatrix} \begin{pmatrix} y_1 \\ y_2 \end{pmatrix} \right) \left( \begin{pmatrix} a_{21} & a_{22} \end{pmatrix} \begin{pmatrix} y_1 \\ y_2 \end{pmatrix} \right) =$$

$$= \left( \begin{pmatrix} y_1 & y_2 \end{pmatrix} \begin{pmatrix} a_{11} \\ a_{21} \end{pmatrix} \right) \left( \begin{pmatrix} y_1 & y_2 \end{pmatrix} \begin{pmatrix} a_{12} \\ a_{22} \end{pmatrix} \right) =$$

$$= \begin{pmatrix} y_1 & y_2 \end{pmatrix} \begin{pmatrix} a_{11} & a_{12} \\ a_{21} & a_{22} \end{pmatrix} = y^T A^T \quad \textcircled{1}$$

$$A(x + \lambda y) - b \quad \textcircled{2}$$

$$\Rightarrow \underline{\underline{y^T A^T (A(x + \lambda y) - b)}}$$

$$\frac{\partial}{\partial x} f(x) = y^T A^T (A(x + x_y) - b) =$$

$$= y^T A^T A \cdot (x + x_y - y^T A^T b) =$$

$$= y^T A^T A (x + x_y) - y^T A^T b =$$

$$= y^T A^T A x + y^T A^T A x_y - y^T A^T b =$$

$$= y^T A^T A x + y^T A^T A x_y - y^T A^T b =$$

$$= \quad + y^T A^T A y - \dots$$

$$\frac{\partial^2}{\partial x^2} f(x) =$$

$$\underline{\underline{y^T A^T A \cdot y}}$$



c) Generalize for arbitrary  $n$

$$f(x) = \left( \prod_{i=1}^n x_i \right)^{\frac{1}{n}}$$

$$\frac{\partial}{\partial x_j} f(x) = \frac{1}{n} f(x) \cdot \frac{1}{x_j}$$

$$\frac{\partial^2}{\partial x_j^2} f(x) = \frac{1}{n} \left( \frac{1}{n} \cdot \left( \prod_{i=1}^n x_i \right)^{\frac{1}{n}} \cdot \frac{\frac{n}{\prod_{i=1}^n x_i} x_i}{x_j} \cdot \frac{1}{x_j} + \left( \prod_{i=1}^n x_i \right)^{\frac{1}{n}} \cdot (-1) \cdot \frac{1}{x_j^2} \right) =$$

$$\frac{\partial^2}{\partial x_j^2} f(x) = \frac{1}{n} \left( \frac{1}{n} f(x) \cdot \frac{1}{x_j x_j} + (-1) f(x) \cdot \frac{1}{x_j^2 x_j} \right)$$

$$\begin{aligned} \frac{\partial^2}{\partial x_j^2} f(x) &= \frac{1}{n} f(x) \cdot \left( \left( \frac{1}{n} - 1 \right) \right) \cdot \frac{1}{x_j x_j} \\ &= \frac{1}{n} \left( \left( \frac{1}{n} - 1 \right) \right) \cdot f(x) \cdot \frac{1}{x_j x_j} \end{aligned}$$

$j \neq k$

$$\frac{\partial^2}{\partial x_j \partial x_k} f(x) = \frac{1}{n} \cdot \frac{1}{n} \left( \prod_{i=1}^n x_i \right)^{\frac{1}{n}-1} \frac{\frac{n}{\prod_{i=1}^n x_i} x_i}{x_k} \cdot \frac{1}{x_j}$$

~~$\frac{1}{n}$~~

$$\frac{\partial^2}{\partial x_j^2 \partial x_k^2} f(x) = \frac{1}{n} \left( \frac{1}{n} - 1 \right) f(x) \frac{1}{x_j x_j} = \frac{1}{n} \left( \frac{1}{n} - 1 \right) f(x) \frac{1}{x_j x_j}$$

$$\frac{\partial^2}{\partial x_j \partial x_k} f(x) = \frac{1}{n^2} f(x) \frac{1}{x_j x_k} = \frac{1}{n} \left( \frac{1}{n} \right) f(x) \frac{1}{x_j x_k}$$

$$\nabla^2 f(x) = \frac{1}{n} f(x) \cdot \begin{pmatrix} \left( \frac{1}{n} - 1 \right) \frac{1}{x_1^2} & \frac{1}{n} \cdot \frac{1}{x_1 x_2} & \dots & \frac{1}{n} \cdot \frac{1}{x_1 x_n} \\ \left( \frac{1}{n} - 1 \right) \frac{1}{x_2^2} & \dots & \dots & \dots \\ \vdots & \vdots & \ddots & \vdots \\ \left( \frac{1}{n} - 1 \right) \frac{1}{x_n^2} & \dots & \dots & \dots \end{pmatrix}$$