# Differentiation

Derivative	Notation	
$f'(a) = \lim_{h \to 0} \frac{f(a+h) - f(a)}{h}$	$f\in \mathcal{C}$	$f$ is continuous in $\mathbb{R}^n$
$h \to 0$ h	$f \in \mathcal{C}(S)$	f is continuous in $S$
Total derivative	$f \in \mathfrak{C}$	f is differentiable on $\mathbb{R}^n$
$f \in U \subset \mathbb{R}^n \to \mathbb{R}^m, \ L \in \mathbb{R}^{m \times n}$	$f \in \mathfrak{C}(S)$	f is differentiable on $S$
•	$f \in \mathfrak{C}_T$	$f$ is total differentiable on $\mathbb{R}^n$
$f'(a) := L \iff \lim_{a \to x} \frac{  f(x) - f(a) - L(x - a)  }{  x - a  } = 0$	$f \in \mathfrak{C}_T(S)$	f is total differentiable on $S$
$\int (a) \cdot \Delta (a) \cdot \Delta (a) = \int (a) \cdot \Delta (a) \cdot \Delta (a) = \int (a) \cdot \Delta (a) \cdot \Delta (a) = \int (a) \cdot \Delta (a) \cdot \Delta (a) = \int (a) \cdot \Delta (a) \cdot \Delta (a) = \int (a) \cdot \Delta (a) \cdot \Delta (a) = \int (a) \cdot \Delta (a) \cdot \Delta (a) = \int (a) \cap (a$	$f \in \mathfrak{C}_P$	$f$ is partial differentiable on $\mathbb{R}^n$
We then call $L$ the <b>Jacobian matrix</b> and usually de-	$f \in \mathfrak{C}_P(S)$	f is partial differentiable on $S$
note it by $f'(a)$ .	$f \in \mathcal{C}^1$	$f$ is continuously differentiable on $\mathbb{R}^n$
In the one dimensional case no distinction has to be	$f \in \mathcal{C}^1(S)$	f is continuously differentiable on $S$

In the one dimensional case no distinction has to be made between the total derivative and the derivative.

We use  $\mathfrak{C}^n$  to denote an n-times differentiable function.

Note that  $f \in \mathfrak{C}$  and  $f \in \mathfrak{C}(S)$  is only used in the one

dimensional	case.

	dimensional case.	
Elementary functions	Hyperbolic functions	Trigonometric functions
$f(x)   o  rac{d}{dx}f(x)$	$f(x) \rightarrow \frac{d}{dx}f(x)$	$f(x) \rightarrow \frac{d}{dx}f(x)$
$\begin{array}{cccccccccccccccccccccccccccccccccccc$	$\begin{aligned} \sinh x &\to \cosh x \\ \cosh x &\to \sinh x \\ \tanh x &\to -\tanh^2 x + 1 = \frac{1}{\cosh^2 x} \\ \coth x &\to -\coth^2 x + 1 = \frac{1}{\cosh^2 x} \\ \operatorname{arsinh} x &\to \frac{1}{\sqrt{x^2 + 1}} \\ \operatorname{arcosh} x &\to \frac{1}{\sqrt{x^2 - 1}} \\ \operatorname{artanh} x &\to \frac{1}{1 - x^2} \\ \operatorname{arcoth} x &\to \frac{1}{1 - x^2} \\ \operatorname{arcoth} x &\to \frac{1}{2 - x} \\ \operatorname{arcoth} x &\to \frac{1}{2 - x} \\ \operatorname{arcosh} x &\to \frac{1}{$	$ \begin{array}{rcl} \sin x & \to & \cos x \\ \cos x & \to & -\sin x \\ \tan x & \to & \tan^2 x + 1 = \frac{1}{\cos^2 x} \\ \cot x & \to & -\cot^2 x - 1 = \frac{-1}{\sin^2 x} \\ \arctan x & \to & \frac{1}{\sqrt{1-x^2}} \\ \arctan x & \to & \frac{1}{\sqrt{1-x^2}} \\ \arctan x & \to & \frac{1}{1+x^2} \\ \arctan x & \to & \frac{1}{1+x^2} \\ \operatorname{arccot} x & \to & \frac{-1}{1+x^2} \\ \operatorname{arccot} x & \to & \frac{-\cos x}{\sin^2 x} \\ \sec x &= \frac{1}{\cos x} & \to & \frac{\sin x}{\cos^2 x} \end{array} $
Single variable rules $\begin{array}{rcl} (cf)' &=& cf' & \text{Linearity 1} \\ (f \pm g)' &=& f' \pm g' & \text{Linearity 2} \\ (fg)' &=& f'g + fg' & \text{Product rule} \\ \left(\frac{f}{g}\right)' &=& \frac{f'g - fg'}{g^2} & \text{Quotient rule} \\ \left(\frac{1}{f}\right)' &=& -\frac{f'}{f^2} & \text{Reciprocal rule} \\ (f \circ g)' &=& (f' \circ g) \cdot g' & \text{Chain rule} \\ (f^{-1})' &=& \frac{1}{f' \circ f^{-1}} & \text{Inverse function rule} \end{array}$	Mean value theorem $f \in C([a,b]) \land f \in \mathfrak{C}((a,b)) \Longrightarrow$ $\exists c \in (a,b) : f'(c) = \frac{f(b) - f(a)}{b - a}$ Logarithmic derivative $(\ln f)' = \frac{f'}{f}$	
Real-valued functions $f: D \subset \mathbb{R}^n \to \mathbb{R}$ where D is open. $\boldsymbol{x}: I \subset \mathbb{R} \to \mathbb{R}^n$	$\begin{array}{c} \textbf{Vector-valued functions} \\ \boldsymbol{f}: B \subset \mathbb{R}^l \to \mathbb{R}^m \\ \boldsymbol{g}: D \subset \mathbb{R}^n \to \mathbb{R}^l \end{array}$	Gradient, Curl, Divergence
Partial derivative	Jacobian matrix	$\nabla f$ $O_{x_1} f$ $(f)^T$
$\frac{\partial f}{\partial x} = \partial_x f$ all variables except $x$	If $\mathbf{f} \in \mathcal{C}^1(B)$	$\nabla f = \begin{bmatrix} \partial_{x_1} f \\ \vdots \\ \partial_{x_n} f \end{bmatrix} = (f')^T$
$\partial_x$ are considered to be constants	$\begin{bmatrix} \partial_{x_1} f_1 & \dots & \partial_{x_n} f_1 \end{bmatrix}$	$\begin{bmatrix} & & & \\ & & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & & \\ & & & & \\ & & & & \\ & & & & \\ & & & & \\ & & & & \\ & & & & \\ & & & & \\ & & & & \\ & & & & \\ & & & & \\ & & & & \\ $
If all partial derivates exist we call the function <b>partial differentiable</b> .	$\boldsymbol{f}' = \begin{bmatrix} \partial_{x_1} f_1 & \dots & \partial_{x_n} f_1 \\ \vdots & \ddots & \vdots \\ \partial_{x_1} f_m & \dots & \partial_{x_n} f_m \end{bmatrix}$	$\operatorname{div} \boldsymbol{f} = \langle \nabla, \boldsymbol{f} \rangle$
$ \begin{array}{ll} f \in \mathfrak{C}_P(D) & \not\Longrightarrow & f \in \mathcal{C}^1(D) \\ f \in \mathfrak{C}_T(D) & \longleftarrow & f \in \mathcal{C}^1(D) \end{array} $	Chain rule	Hessian matrix $\begin{bmatrix} \partial^2 f & -\partial^2 f \end{bmatrix}$
$f \in \mathfrak{C}_T(D) \implies f' = [\partial_{x1}f \; \partial_{x2}f \; \dots]$	$ \begin{array}{c} \boldsymbol{f} \in \mathcal{C}^1(B), \boldsymbol{g} \in \mathfrak{C}_P(D) \implies \\ (\boldsymbol{f} \circ \boldsymbol{g})' = (\boldsymbol{f}' \circ \boldsymbol{g}) \cdot \boldsymbol{g}' \end{array} $	$\begin{bmatrix} \frac{\partial^2 f}{\partial x_1^2} & \cdots & \frac{\partial^2 f}{\partial x_n x_1} \end{bmatrix}$
Schwarz's theorem	$(\mathbf{j} \circ \mathbf{g}) = (\mathbf{j} \circ \mathbf{g}) \cdot \mathbf{g}$ from that	$\mathcal{H}_f = \begin{bmatrix} & & & \\ & & & \\ & & & \\ & & & & & \\ & & & & \\ & & & & \\ & & & & & \\ & & & & & \\ & & & & \\ & & & & \\ & & & & \\ & & & & & \\ & & & & & \\ & & & & \\$
$f \in \mathcal{C}^2(D) \implies \frac{\partial^2 f}{\partial xy} = \frac{\partial^2 f}{\partial yx}$	$\frac{\partial_t f(x(t), y(t))}{\partial_t f(x(t), y(t))} = \partial_x f \cdot \partial_t x + \partial_y f \cdot \partial_t y$	$\left[\begin{array}{ccc} \frac{\partial^2 \boldsymbol{f}}{\partial x_1 x_n} & \cdots & \frac{\partial^2 \boldsymbol{f}}{\partial x_n^2} \end{array}\right]$
$\int \mathcal{C} (D) \longrightarrow \partial_{xy} - \partial_{yx}$ Chain rule	$O_{tJ}(x(t), y(t)) = O_{xJ} \circ O_{t}x + O_{yJ} \circ O_{t}y$ Mean value theorem	in the two dimensional case
$f \in \mathcal{C}^1(D), \ \boldsymbol{x} \in \mathfrak{C}(I) \implies$	$f \in \mathcal{C}^1(B) \implies$	$egin{aligned} \mathcal{H}_f = \left[ egin{aligned} \partial_{xx} oldsymbol{f} & \partial_{yx} oldsymbol{f} \ \partial_{xy} oldsymbol{f} & \partial_{yy} oldsymbol{f} \end{aligned}  ight] \end{aligned}$
$\frac{d}{dt}f(\boldsymbol{x}(t)) = f'(\boldsymbol{x}(t)) \cdot \dot{\boldsymbol{x}}(t) = \sum_{i=1}^{n} \frac{\partial f}{\partial x_i}(\boldsymbol{x}(t)) \cdot \dot{x}_i(t)$	$\exists \theta_i \in (0,1) : f_i(\boldsymbol{x}) - f_i(\boldsymbol{a}) = f'_i(\boldsymbol{a} + \theta_i(\boldsymbol{x} - \boldsymbol{a}))(\boldsymbol{x} - \boldsymbol{a})$	$\int \left[ \partial_{xy} \boldsymbol{f} \ \partial_{yy} \boldsymbol{f} \right]$
Mean value theorem	Inverse function rule	
$f \in \mathcal{C}^1(D) \implies \\ \exists \theta \in (0,1) : f(\boldsymbol{x}) - f(\boldsymbol{a}) = f'(\boldsymbol{a} + \theta(\boldsymbol{x} - \boldsymbol{a}))(\boldsymbol{x} - \boldsymbol{a})$	$f: D \subset \mathbb{R}^n \to \mathbb{R}^n$	
	$D \text{ open } \wedge \mathbf{f} \in \mathcal{C}^1 \wedge \exists \mathbf{f}^{-1} \\ \Longrightarrow (\mathbf{f}^{-1})'(\mathbf{y}) = [\mathbf{f}'(\mathbf{f}^{-1}(\mathbf{y}))]^{-1}$	
Directional derivative	$\Rightarrow (I - ) (\mathbf{y}) =  I (I - (\mathbf{y})) ^{-1}$	

 $\partial_{\boldsymbol{v}} f = \langle \nabla f, \boldsymbol{v} \rangle$ 

# **Applications of Differentiation**

#### L'Hôpital's rule

 $\lim f(x) = \lim g(x) = 0$ A zero (root) of f can be computed by (if  $x_0$ is choosen near a zero)  $\lim_{x \to \infty} f(x) = \lim_{x \to \infty} g(x) = \pm \infty \quad \Longrightarrow \quad$  $f'(x_n) \neq 0 \Longrightarrow$  $x_{n+1} = x_n - \frac{f(x_n)}{f'(x_n)}$  $\lim_{x \to a} \frac{f(x)}{g(x)} = \lim_{x \to a} \frac{f'(x)}{g'(x)}$ Τf Monotonicity Newton's method (multivariate)  $f \in \mathcal{C}^1$  $f' \geq 0$  $f'(x_n)(x_{n+1}-x_n) = -f(x_n)$ monotonically increasing  $\Leftrightarrow$  $f' \stackrel{-}{\leq} 0$ monotonically decreasing  $\Leftrightarrow$ f' > 0strictly increasing  $\implies$ f' < 0 $\implies$ strictly decreasing Linear approximation (tangent) at a Taylor series (of f at a)  $f \in \mathcal{C}^{\infty} \implies$  $f \in \mathcal{C}^1 \implies$  $T(x, a, f) = \sum_{k=0}^{\infty} \frac{f^{(k)}(a)}{k!} (x - a)^k$  $f(x) \approx f(a) + f'(a) \cdot (x - a)$ Approximation by a polynomial of degree n at aTaylor's theorem  $f \in \mathcal{C}^{n+1}(I) \land x, a \in I \implies$  $f(x) = \sum_{k=0}^{n} \frac{f^{(k)}(a)}{k!} + R_{n+1}(x, a)$  $R_{n+1}(x, a) = \frac{1}{n!} \int_{a}^{x} (x - t)^{n} f^{(n+1)}(t) dt$  $\lim_{x \to \infty} R_n(x) = 0 \iff$ f(x) = T(x, a, f)Remainder In this case the function at x is (Lagrange form)  $R_{n+1}(x,a) = \frac{f^{(n+1)}(\xi)}{(n+1)!}(x-a)^{n-1}$ equal to its taylor series. Remainder (Cauchy form)  $\xi \in (x,a) \lor \xi \in (a,x)$ We often don't know the remainder explicitely; However we can bound it by  $\mathcal{O}(x^{n+1})$ .

Single variable Extrema

### $f:[a,b]\to\mathbb{R}$ $\exists r \forall x : |x - c| < r \implies f(x) < f(c)$ local minima

We call a, b and all points  $c \in [a, b]$  where f isn't differentiable or where f'(x) = 0 a critical point.

#### Fermat's theorem

 $c \text{ extrema} \Longrightarrow c \text{ critical point}$ 

#### First derivative test

If f' changes sign at c, f has an extremum at c. If the sign changes from + to - the function has a maximum at c.

#### Second derivative test

 $f^{\prime\prime}(c) < 0 \quad \Longrightarrow \quad f \text{ has a maxima at } c$  $f''(c) > 0 \implies f$  has a minima at c $f''(c) = 0 \implies$  the test is inconclusive Existence theorem

 $f \in \mathcal{C}^1([a,b]) \land f \in \mathfrak{C}((a,b))$  $\implies f$  has a maxima and minima on [a, b]

#### Zeroes of the gradient

	$\nabla(uv)$	=	$u \cdot  abla \cdot  abla$	$v + v \cdot  abla u$	
u, v linearily indepe	endent	$\implies$	u = 0	$0 \wedge v = 0$	
u, v linearily depe	endent	$\implies$	equat	ion with	
			fewer	variables	
Linear dependence	in the	2D	case is	determined	by
$\det[\nabla v \ \nabla u] = 0.$					

Elementary	' tay	lor series	
$\frac{1}{1-x}$	=	$\sum_{n\geq 0} x^n$	x  < 1
$\frac{x^m}{1-x}$	=	$\sum x^n$	$ x <1,\ m\in\mathbb{N}_{\geq0}$
$\frac{x}{(1-x)^2}$	=	$\sum_{n\geq 1}^{\infty} nx^n$	x  < 1
$(1+x)^{z}$	=	$\sum_{\substack{n \ge n \\ n \ge 1}}^{n \ge m} nx^n$ $\sum_{\substack{n \ge 0 \\ n \ge 0}}^{n} (\frac{z}{n}) \cdot x^n$	$ x <1,\ z\in\mathbb{C}$
$\sqrt{1+x}$	=	$\sum_{n\geq 0} \frac{(-1)(2n)!}{(1-2n)n!^2 4^n} x^n$	
$e^x$	=	$\sum_{n\geq 0}^{-} \frac{x^n}{n!}$	
$\log(1-x)$			x  < 1

## Newton's method

Multivariable Extrema

Second derivative test

 $\exists r \forall \boldsymbol{x} : || \boldsymbol{x} - \boldsymbol{c} || < r \Longrightarrow f(\boldsymbol{x}) < f(\boldsymbol{c})$ 

f has a critical point at c iff  $\nabla f(c) = 0$ .

 $\mathcal{H}_f(c)$  positive definite  $\implies$  minima  $\mathcal{H}_f(c)$  negative definite  $\implies$  maxima

 $oldsymbol{H}_1 = a_{11} \quad oldsymbol{H}_2 = \left[ egin{array}{cc} a_{11} & a_{12} \ a_{21} & a_{22} \end{array} 
ight] \quad \dots \quad oldsymbol{H}_n = oldsymbol{A}$ 

 $\begin{array}{rcl} \boldsymbol{A} \text{ positive definite} & \Leftarrow & \forall i : \det \boldsymbol{H}_i > 0 \\ \boldsymbol{A} \text{ negative definite} & \Leftarrow & \forall i : \operatorname{sgn}(\det \boldsymbol{H}_i) \end{array}$ 

 $f: D \subset \mathbb{R}^n \to \mathbb{R}$ 

 $\mathcal{H}_f(c)$  indefinit

Definiteness

 $f \in \mathfrak{C}_P \land D$  open

Implicit function theorem

 $f: D \subset \mathbb{R}^{n+m} \to \mathbb{R}^m, \ (x, y) \mapsto f(x, y)$  $oldsymbol{f} \in \mathcal{C}^1 \ \land \ D$  open  $f' = \begin{bmatrix} \frac{\partial f}{\partial x} & \frac{\partial f}{\partial y} \end{bmatrix}$  $\boldsymbol{f}(\boldsymbol{a}, \boldsymbol{b}) = 0 \quad \wedge \quad \det \frac{\partial \boldsymbol{f}}{\partial \boldsymbol{u}}(\boldsymbol{a}, \boldsymbol{b}) \neq 0$ 

Then there exists an open neighbourhood  $U \subset \mathbb{R}^{n+m}$  of  $(\boldsymbol{a}, \boldsymbol{b})$  such that there exits a function  $\boldsymbol{g}:\mathbb{R}\to\mathbb{R}^m$  which describes the solutions of f(a, b) = 0 in U.

Multivariate approximation Multivariate Taylor series  $(Df(a))(\cdot) := \langle \nabla f, (\cdot) \rangle$  $T(\boldsymbol{x}, \boldsymbol{a}, f) = \sum_{n=0}^{\infty} \frac{1}{n!} D^n f(\boldsymbol{a}) \left(\boldsymbol{x} - \boldsymbol{a}, \dots, \boldsymbol{x} - \boldsymbol{a}\right)$ The Taylor series is generalized componentwise to vector-valued functions.

local

maxima

⇒ no extremum

minima

Extrema with constraints

 $f: \mathbb{R}^n \to \mathbb{R}$  $g_k: \mathbb{R}^n \to \mathbb{R}$ 

 $1 < k < m ~~\wedge~~ f \in \mathfrak{C}_P ~~\wedge~ D$ open The maxima of f under the constraints of  $g_1, g_2, \dots = 0.$ 

First Method

 $[g(x,y)=0 \Longleftrightarrow y=h(x)]$  $\implies f(x, h(x))' = 0$ 

Second method

## $\{(x,g(x,y))|g(x,y)=0\}=\{(x(t),y(t)),t\in\mathbb{R})$ $\implies f(x(t), y(t))' = 0$

#### Lagrange multipliers

 $L(x_1,\ldots,x_n,\lambda_1\ldots\lambda_m) = f + \sum_{i=1}^m \lambda_i g_i$ Then the extrema are given by  $\nabla L = \mathbf{0}$ 

### Hyperbolic/Trigonometric taylor series

 $=(-1)^{i}$ 

 $\sum_{n \ge 0} \frac{(-1)^n}{(2n+1)!} x^{2n+1}$  $\sin x =$  $\sin x = \sum_{n \ge 0} \frac{(-1)^n}{(2n+1)!} x^{2n}$   $\cos x = \sum_{n\ge 0} \frac{(-1)^n}{(2n)!} x^{2n}$   $\tan x = \sum_{n\ge 1} \frac{B_{2n}(-4)^n (1-4)^n}{(2n)!} x^{2n-1} \quad |x| < \frac{\pi}{2}$   $\arcsin x = \sum_{n\ge 0} \frac{(-2n)!}{4^n n!^2 (2n+1)} x^{2n+1} \quad |x| \le 1$   $\arccos x = \frac{\pi}{2} - \arcsin x$   $\arctan x = \sum_{n\ge 0} \frac{(-1)^n}{(2n+1)} x^{2n+1} \quad |x| \le 1$   $\sec x = \sum_{n\ge 0} \frac{(-1)^n E_{2n}}{(2n)!} x^{2n}$   $W_0(x) = \sum_{n\ge 1} \frac{(-n)^{n-1}}{n!} x^n \quad |x| < \frac{1}{e}$ 

The taylor series of sinh/cosh/tanh can be formed by removing  $(-1)^n$  from the formulas of sin/cos/tan; The taylor series holds for  $\mathbb{R}$ ,  $\mathbb{R}$ , and  $|x| < \pi/2$  respectively. The formulas of arcsinh/arccosh/arctanh can be formed by adding  $(-1)^n$  to the formulas for arcsin/arccos/arctan; They hold for |x| < 1.

 $B_n$  are the Bernoulli numbers.  $E_n$  are the Euler numbers.