

BIEM/BIEF

A.Y. 2024/2025

BLAB

HANDOUTS

MATHEMATICS (MODULE 1)

-SECOND PARTIAL-

WRITTEN BY

ALESSANDRA VITIELLO



TEACHING DIVISION

“

This handout is written by students with no intention of replacing university materials.

It is a useful tool for studying the subject, but does not guarantee preparation as exhaustive and complete as the material recommended by the University.

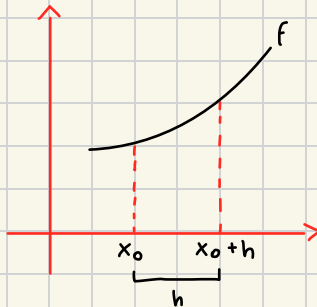


VARIABLE DIFFERENTIAL CALCULS

DIFFERENT QUOTIENT

given $f: A \subseteq \mathbb{R} \rightarrow \mathbb{R}$

x_0 interior point of A



INCREMENT / VARIATION OF x

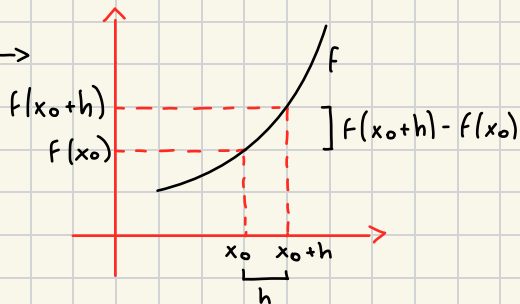
h = increment with sign \rightarrow

- $h > 0 \Rightarrow x_0 + h > x_0$
- $h < 0 \Rightarrow x_0 + h < x_0$

INCREMENT / VARIATION OF f

$f(x_0+h) - f(x_0)$ = increment with sign \rightarrow

- $> 0 \Rightarrow f(x_0+h) > f(x_0)$
- $< 0 \Rightarrow f(x_0+h) < f(x_0)$



DIFFERENCE QUOTIENT OF f AT x_0

$$\frac{f(x_0+h) - f(x_0)}{h} = \frac{\Delta f}{h} = \frac{\Delta f}{\Delta x}$$

DERIVATIVE

consider $f: A \subseteq \mathbb{R} \rightarrow \mathbb{R}$

x_0 interior point for A

increment of $x = h$

increment of $f = f(x_0+h) - f(x_0)$

different quotient of f at $x_0 = \frac{f(x_0+h) - f(x_0)}{h}$

if the limit: $\lim_{h \rightarrow 0} \frac{f(x_0+h) - f(x_0)}{h}$

\exists finite we say f has a derivative at x_0

we call derivative of f at x_0 the value $\rightarrow f'(x_0) = \lim_{h \rightarrow 0} \frac{f(x_0+h) - f(x_0)}{h}$

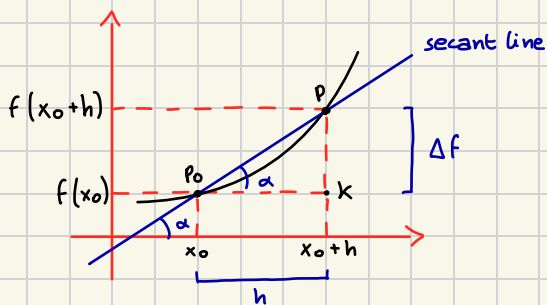
LEIBNIZ'S NOTION (alternative notion) \leftarrow

$$\frac{df}{dx}(x_0), \frac{dy}{dx}(x_0)$$

GEOMETRICAL MEANING OF $\frac{\Delta f}{h}$

$$\frac{\Delta f}{h} = \frac{Pk}{PoK} = \tan \alpha = m \text{ secant}$$

⇓
slope of the secant line



GEOMETRICAL MEANING OF $f'(x_0)$

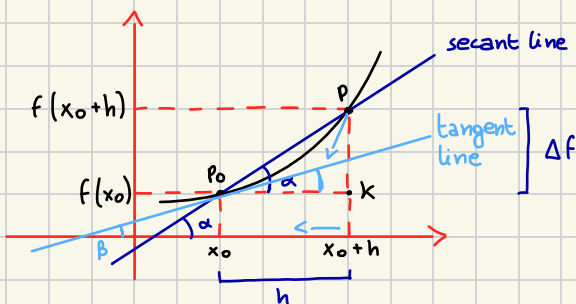
$$h \rightarrow 0 \Rightarrow x_0 + h \rightarrow x_0$$

$$P \rightarrow P_0$$

secant line \rightarrow tangent line at P_0

$$f'(x_0) = \lim_{h \rightarrow 0} \frac{\Delta f}{h} = \tan \beta = m \text{ tangent}$$

⇓
slope of the tangent line



↳ when the slope \exists (when the tangent line is not vertical)

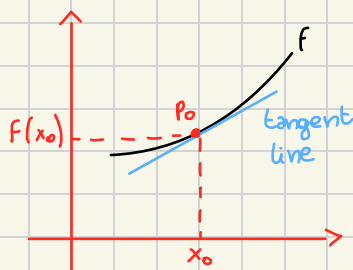
PROCEDURE TO WRITE THE TANGENT LINE TO f AT x_0

(when it's not vertical)

① tangent line to f at $x_0 =$ straight line passing through $P_0(x_0, f(x_0)) \rightarrow y - f(x_0) = m(x - x_0)$

② $m = f'(x_0) \rightarrow y - f(x_0) = f'(x_0)(x - x_0)$

⇓
equation of the tangent line

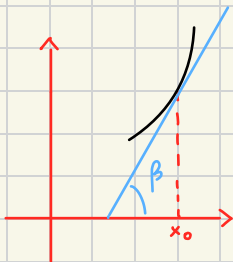


WHAT IF:

- f "quickly increasing" \rightarrow

$$f'(x_0) = \tan \beta$$

$$0 < \beta < \frac{\pi}{2} \Rightarrow \text{"close" to } \frac{\pi}{2}$$

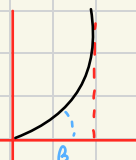


$$\rightarrow y = \tan x:$$

⇓

$f'(x_0) > 0$ and "large"

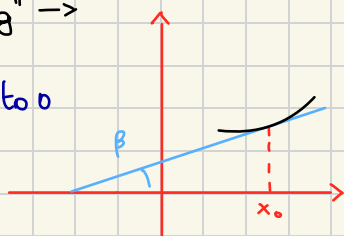
≈ 3



- f "slowly increasing" \rightarrow

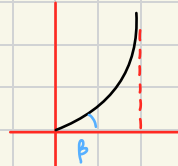
$$f'(x_0) = \tan \beta$$

$0 < \beta < \frac{\pi}{2} \Rightarrow$ "close" to 0



$$\rightarrow y = \tan x:$$

\Downarrow

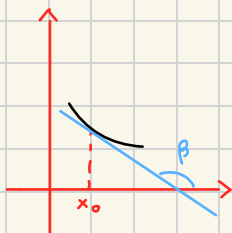


$f'(x_0) > 0$ and "small" $\approx \frac{1}{3}$

- f "slowly decreasing" \rightarrow

$$f'(x_0) = \tan \beta$$

$\frac{\pi}{2} < \beta < \pi \Rightarrow$ "close" to π



$$\rightarrow y = \tan x:$$

\Downarrow



$f'(x_0) < 0$ and "small"

$\approx -\frac{1}{3}$

- f "quickly decreasing" \rightarrow

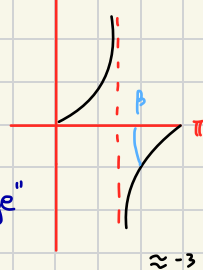
$$f'(x_0) = \tan \beta$$

$\frac{\pi}{2} < \beta < \pi \Rightarrow$ "close" to $\frac{\pi}{2}$



$$\rightarrow y = \tan x:$$

\Downarrow



$f'(x_0) < 0$ and "large"

≈ -3

BUT WHAT IF:

① $\lim_{h \rightarrow 0} \frac{\Delta f}{h} \exists$ but is not finite \rightarrow EXAMPLE $y = \sqrt[3]{x} \Rightarrow$

$$\text{at } x=0 \rightarrow \lim_{h \rightarrow 0} \frac{f(0+h) - f(0)}{h} =$$

$$\lim_{h \rightarrow 0} \frac{\sqrt[3]{h} - 0}{\frac{h}{\sqrt[3]{h}}} = \lim_{h \rightarrow 0} \frac{1}{\sqrt[3]{h^2}} = +\infty \Rightarrow \nexists f'(0)$$



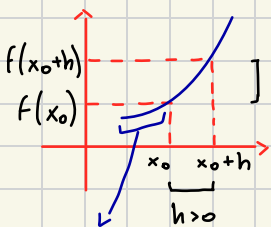
$x=0$ is inflection point with vertical tangent

$$\textcircled{2} \lim_{h \rightarrow 0} \frac{\Delta f}{h} \neq \Rightarrow$$

- RIGHT DERIVATIVE = $f'_+(x_0) = \lim_{h \rightarrow 0^+} \frac{f(x_0+h) - f(x_0)}{h}$ if the limit \exists finite

$f: A \subseteq \mathbb{R} \rightarrow \mathbb{R}$

x_0 such that \exists a right neighbourhood $B^+(x_0) \subseteq A$



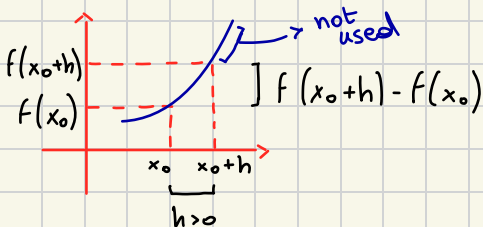
- right increment of $x = h$ (≥ 0)
- increment of $f = f(x_0+h) - f(x_0)$ (≥ 0)

• difference quotient = $\frac{f(x_0+h) - f(x_0)}{h}$

- LEFT DERIVATIVE = $f'_-(x_0) = \lim_{h \rightarrow 0^-} \frac{f(x_0+h) - f(x_0)}{h}$ if the limit \exists finite

$f: A \subseteq \mathbb{R} \rightarrow \mathbb{R}$

x_0 such that \exists a left neighbourhood $B^-(x_0) \subseteq A$



- left increment of $x = h$ (≤ 0)
- increment of $f = f(x_0+h) - f(x_0)$ (≤ 0)

• different quotient = $\frac{f(x_0+h) - f(x_0)}{h}$

THEOREM

$f: A \subseteq \mathbb{R} \rightarrow \mathbb{R}$, x_0 interior point

$f'(x_0) \exists \iff$ ① $f'_+(x_0) \exists$

② $f'_-(x_0) \exists$

③ they are equal

EXAMPLE of $\textcircled{2}$

$$y = |x| = \begin{cases} x & x \geq 0 \\ -x & x < 0 \end{cases} \Rightarrow \begin{aligned} f'_+(0) &= \lim_{h \rightarrow 0^+} \frac{f(0+h) - f(0)}{h} = \lim_{h \rightarrow 0^+} \frac{h-0}{h} = 1 \\ f'_-(0) &= \lim_{h \rightarrow 0^-} \frac{f(0+h) - f(0)}{h} = \lim_{h \rightarrow 0^-} \frac{-h-0}{h} = -1 \end{aligned}$$

$\Rightarrow f'(0) \nexists$

$\hookrightarrow \exists f'_+(x_0), \exists f'_-(x_0)$ but they are $\neq \Rightarrow x_0 = \text{corner}$

$$y = \sqrt{|x|} = \begin{cases} \sqrt{x} & x \geq 0 \\ \sqrt{-x} & x < 0 \end{cases} \Rightarrow$$

$$\lim_{h \rightarrow 0^+} \frac{f(0+h) - f(0)}{h} = \lim_{h \rightarrow 0^+} \frac{\sqrt{h} - 0}{h} = \lim_{h \rightarrow 0^+} \frac{1}{\sqrt{h}} = +\infty \Rightarrow f'_+(0) \neq$$

$\underbrace{\hspace{10em}}_{\substack{\text{ } \\ \rightarrow \sqrt{h} \cdot \sqrt{h} = (\sqrt{h})^2}}$

$$\lim_{h \rightarrow 0^-} \frac{f(0+h) - f(0)}{h} = \lim_{h \rightarrow 0^-} \frac{\sqrt{-h} - 0}{h} = \lim_{h \rightarrow 0^-} \frac{1}{-\sqrt{-h}} = -\infty \Rightarrow f'_-(0) \neq$$

$\underbrace{\hspace{10em}}_{\substack{\text{ } \\ \rightarrow -\sqrt{-h} \cdot -\sqrt{-h} = -(\sqrt{-h})^2 = -(-h) = h}}$

$$\hookrightarrow \exists \lim_{h \rightarrow 0^+} \frac{f(x_0+h) - f(x_0)}{h} = +\infty,$$

$$\exists \lim_{h \rightarrow 0^-} \frac{f(x_0+h) - f(x_0)}{h} = -\infty, \text{ or vice versa}$$

$\Rightarrow x_0 = \text{cusp point}$

COMPUTATION OF DERIVATES

= given $y = f(x)$, given the general interior point x , find $f'(x) \Rightarrow$ the derivative function, a function which at all points x will give the derivative of $f(x)$; if we want $f'(x_0)$, we just substitute $x = x_0$ in $f'(x_0)$

ELEMENTARY DERIVATES

① $y = k \Rightarrow y' = 0$ (\forall constant $k \in \mathbb{R}$)

proof: $f'(x) = \lim_{h \rightarrow 0} \frac{f(x_0+h) - f(x_0)}{h} = \lim_{h \rightarrow 0} \frac{k - k}{h} = \lim_{h \rightarrow 0} \frac{0}{k} = 0$

② $y = x^\alpha \Rightarrow y' = \alpha x^{\alpha-1}$

proof: $f'(x) = \lim_{h \rightarrow 0} \frac{f(x_0+h) - f(x_0)}{h} = \lim_{h \rightarrow 0} \frac{(x+h)^\alpha - x^\alpha}{h} =$

$$\lim_{h \rightarrow 0} \frac{[x(1+\frac{h}{x})]^\alpha - x^\alpha}{h} = \lim_{h \rightarrow 0} \frac{x^\alpha (1+\frac{h}{x})^\alpha - x^\alpha}{h} = \lim_{h \rightarrow 0} \frac{x^\alpha (1+\frac{h}{x})^\alpha - 1}{h}$$

$$\lim_{t \rightarrow 0} x^{\alpha-1} \frac{(1+t)^\alpha - 1}{t} = \alpha x^{\alpha-1}$$

\swarrow
 α

$$\frac{x^\alpha}{h} = \frac{x^{\alpha-1} \cdot h}{h} = x^{\alpha-1} \left[\frac{h}{x} \right] = t$$

$$③ y = e^x \Rightarrow y' = e^x / y = a^x \Rightarrow a^x \ln a$$

$$④ y = \ln x \Rightarrow y' = \frac{1}{x} / y = \log_a x$$

$$⑤ y = \sin x \Rightarrow y' = \cos x$$

$$⑥ y = \cos x \Rightarrow y' = -\sin x$$

$$⑦ y = \tan x \Rightarrow y' = \frac{1}{\cos^2 x} / \tan^2 x + 1$$

$$⑧ y = \tan^{-1} x \Rightarrow y' =$$

RULES ON DERIVATIVES

① DERIVATIVE OF A LINEAR COMBINATION

suppose f, g have a derivative $\forall x \in (a, b)$

suppose $\alpha, \beta \in \mathbb{R}$

$\alpha f + \beta g$ has a derivative $\forall x \in (a, b)$ and:

$$(\alpha f + \beta g)' = \alpha f' + \beta g'$$

PROOF:

$\forall x \in (a, b)$,

$$(\alpha f + \beta g)'(x) = \lim_{h \rightarrow 0} \frac{(\alpha f + \beta g)(x+h) - (\alpha f + \beta g)(x)}{h} =$$

$$\lim_{h \rightarrow 0} \frac{(\alpha f)(x+h) + (\beta g)(x+h) - (\alpha f)(x) - (\beta g)(x)}{h} =$$

$$\lim_{h \rightarrow 0} \left[\alpha \frac{f(x+h) - f(x)}{h} + \beta \frac{g(x+h) - g(x)}{h} \right] =$$

$$\alpha \lim_{h \rightarrow 0} \frac{f(x+h) - f(x)}{h} + \beta \lim_{h \rightarrow 0} \frac{g(x+h) - g(x)}{h} = \alpha f'(x) + \beta g'(x)$$

② DERIVATIVE OF A PRODUCT

suppose f, g have a derivative $\forall x \in (a, b)$

then fg has a derivative $\forall x \in (a, b)$

$$(fg)' = f'g + fg'$$

③ DERIVATIVE OF A QUOTIENT

suppose f, g have a derivative $\forall x \in (a, b)$

suppose $g(x) \neq 0 \forall x \in (a, b)$

then f/g has a derivative $\forall x \in (a, b)$ and:

$$\left(\frac{f}{g}\right)' = \frac{g f' + f g'}{g^2}$$

④ CHAIN RULE (DERIVATIVE OF A COMPOSITION)

suppose $f: A = (a, b) \rightarrow \mathbb{R}$, $g: B = (c, d) \rightarrow \mathbb{R}$, with $F(A) \subseteq g(B)$

suppose f has a derivative at $x \in A$, g has a derivative at $F(x) \in B$

then the composition $g \circ f: A = (a, b) \rightarrow \mathbb{R}$ has a derivative at x and:

$$(g \circ f)'(x) = g'(f(x)) \cdot f'(x)$$

⑤ DERIVATIVE OF THE INVERSE

suppose $f: (a, b) \rightarrow \mathbb{R}$, injective

suppose f has a derivative at $x_0 \in (a, b)$ with $f'(x_0) \neq 0$

then \exists the inverse function f^{-1} , it has a derivative at $y_0 = f(x_0)$

$$(f^{-1})'(y_0) = \frac{1}{f'(x_0)}$$

EXAMPLES

- f : exponential, f^{-1} : logarithm $\Rightarrow (\ln x)' = \frac{1}{(e^t)'} = \frac{1}{e^t} = \frac{1}{x}$
- f : \tan , f^{-1} : \tan^{-1} (arctan) $\Rightarrow (\tan^{-1} x)' = \frac{1}{(\tan y)'} = \frac{1}{1 + \tan^2 y} = \frac{1}{1 + x^2}$
- $y = x^{\frac{1}{x}} \Rightarrow e^{\ln(x)^{\frac{1}{x}}} = e^{\frac{1}{x} \ln x}$

$$y = e^t$$

$$t = \frac{1}{x} \ln x = x^{-1} \ln x$$

$$y' = e^t$$

$$t' = -x^{-2} \ln x \rightarrow x^{-1} \frac{1}{x}$$

$$= e^{\frac{1}{x} \ln x}$$

$$= -\frac{\ln x}{x^2} + \frac{1}{x^2}$$

$$= x^{\frac{1}{x}}$$

$$= \frac{1 - \ln x}{x^2}$$

$$y' = x^{\frac{1}{x}} \frac{1 - \ln x}{x^2}$$

PIECE-WISE DEFINED FUNCTIONS

$$\textcircled{1} y = |x| = \begin{cases} x & x \geq 0 \\ -x & x < 0 \end{cases} \Rightarrow y' = \begin{cases} 1 & x > 0 \\ -1 & x < 0 \end{cases}$$

$$y'_+(0) = \lim_{h \rightarrow 0^+} \frac{h-0}{h} = \textcircled{1} \Rightarrow y'(0) \neq$$

$$y'_-(0) = \lim_{h \rightarrow 0^-} \frac{-h-0}{h} = \textcircled{-1}$$

$$\textcircled{2} \quad y = \begin{cases} x^2 & x < 0 \\ x & x \geq 0 \end{cases} \Rightarrow y' = \begin{cases} 2x & x < 0 \\ 1 & x > 0 \end{cases}$$

$$y'_+(0) = \lim_{h \rightarrow 0^+} \frac{h-0}{h} = \textcircled{1} \Rightarrow y'(0) \nexists$$

$$y'_-(0) = \lim_{h \rightarrow 0^-} \frac{h^2-0}{h} = \textcircled{0}$$

$$\textcircled{3} \quad y = \begin{cases} x^2 & x < 0 \\ x^3 & x \geq 0 \end{cases} \Rightarrow y' = \begin{cases} 2x & x < 0 \\ 3x^2 & x > 0 \end{cases}$$

$$y'_+(0) = \lim_{h \rightarrow 0^+} \frac{h^3-0}{h} = \textcircled{0} \Rightarrow y'(0) \exists = 0$$

$$y'_-(0) = \lim_{h \rightarrow 0^-} \frac{h^2-0}{h} = \textcircled{0}$$

HIGHER ORDER DERIVATES

$$y = x^4 \Rightarrow y' = 4x^3 \Rightarrow y'' = 12x^2 \Rightarrow y''' = 24x \text{ (and so on)}$$

DERIVABILITY AND CONTINUITY

THEOREM: EXISTENCE OF DERIVATIVE \rightarrow CONTINUITY

suppose $f: (a,b) \rightarrow \mathbb{R}$ has a derivative $f'(x_0)$ at a point $x_0 \in (a,b)$
then f is continuous at the point x_0

proof

$$\text{suppose } \lim_{h \rightarrow 0} \frac{f(x_0+h) - f(x_0)}{h} = f'(x_0)$$

rewrite it using:

$$x = x_0 + h \Rightarrow h = x - x_0$$

$$h \rightarrow 0 = x \rightarrow x_0$$

therefore

$$\lim_{x \rightarrow x_0} \frac{f(x) - f(x_0)}{x - x_0} = f'(x_0)$$

then

$$\lim_{x \rightarrow x_0} [f(x) - f(x_0)] = \lim_{x \rightarrow x_0} \left[\frac{f(x) - f(x_0)}{x - x_0} (x - x_0) \right]$$

$$= \lim_{x \rightarrow x_0} \frac{f(x) - f(x_0)}{x - x_0} \cdot \lim_{x \rightarrow x_0} (x - x_0)$$

$$= f'(x_0) \cdot 0$$

$$= 0$$

but also

$$\lim_{x \rightarrow x_0} [f(x) - f(x_0)] = \lim_{x \rightarrow x_0} f(x) - \lim_{x \rightarrow x_0} f(x_0) = \lim_{x \rightarrow x_0} f(x) - f(x_0)$$

and therefore

$$\lim_{x \rightarrow x_0} f(x) - f(x_0) = 0 \Rightarrow \lim_{x \rightarrow x_0} f(x) = f(x_0)$$

meaning that $f(x)$ is continuous at x_0

but f continuous at $x_0 \not\Rightarrow f$ has a derivative at x_0

f DOES NOT HAVE A DERIVATIVE

- ① f is not continuous at x_0
- ② corners $\rightarrow f'_+(x_0), f'_-(x_0)$ finite and different
- ③ cusps $\rightarrow \lim_{h \rightarrow 0^+}$ and $\lim_{h \rightarrow 0^-}$ infinite and different
- ④ half-cusps \rightarrow one limit is finite and the other infinite
- ⑤ inflection points with vertical tangent

DERIVATIVE AT A BOUNDARY POINT

it is possible to define a derivative $f'(x_0)$ at a boundary point (ex. points a, b of a closed interval $[a, b]$) required that f has the one-sided derivative which is relevant

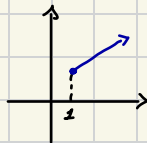
EXAMPLE

$y = x$ defined on $[1, 5]$, $x_0 = 1 \Rightarrow$

we say that f has a derivative at

$x_0 = 1$ because $f'_+(1) \exists$ finite

the same at $x_0 = 5$ because $f'_-(5) \exists$ finite



DIFFERENTIABILITY

$f: A \subseteq \mathbb{R} \rightarrow \mathbb{R}$, x_0 interior point for A

f is differentiable at $x_0 = f$ can be approximated with a straight line in a neighbourhood $\mathcal{U}(x_0)$ of x_0

FORMAL DEFINITION = $f: A \subseteq \mathbb{R} \rightarrow \mathbb{R}$, x_0 interior point for A

f is differentiable at x_0 if: \exists a neighbourhood $\mathcal{U}(x_0)$ of x_0 such that

$\forall x = x_0 + h \in \mathcal{U}(x_0)$

$$f(x) = \underbrace{f(x_0) + m(x-x_0)}_{\substack{\text{non vertical straight} \\ \text{line passing through } p_0}} + \underbrace{\theta(x-x_0)}_{\substack{\text{negligible} \\ \text{error}}} \quad \text{as } x \rightarrow x_0$$

non vertical straight line passing through p_0 negligible error

→ = tangent line

THEOREM: DIFFERENTIABILITY AND EXISTENCE OF A DERIVATIVE

$f: A \subseteq \mathbb{R} \rightarrow \mathbb{R}$, x_0 interior point for A

then

① f is differentiable at $x_0 \iff f$ has a derivative at x_0

② $m = f'(x_0)$

PROOF

since $f(x)$ is differentiable at x_0 , we have:

$$f(x) = f(x_0) + m(x-x_0) + \theta(x-x_0) \quad \text{as } x \rightarrow x_0$$

substitute $x = x_0 + h \rightarrow h = x - x_0$:

$$f(x_0+h) = f(x_0) + m \cdot h + \theta(h) \quad \text{as } h \rightarrow 0$$

$$f(x_0+h) - f(x_0) = m \cdot h + \theta(h) \quad \text{as } h \rightarrow 0$$

$$\frac{f(x_0+h) - f(x_0)}{h} = m + \frac{\theta(h)}{h}$$

$$= m + o(1) \quad \text{as } h \rightarrow 0$$

$$\left[\frac{\theta(h)}{h} = o(1) \text{ is true for the properties of } \theta \right]$$

we take the limit on both sides when $h \rightarrow 0$ and obtain:

$$\lim_{h \rightarrow 0} \frac{f(x_0+h) - f(x_0)}{h} = m$$

which means that $f(x)$ has a derivative $f'(x_0)$ at x_0 and:

$$f'(x_0) = m$$

<==

since $f(x)$ has a derivative at x_0 , we have:

$$\lim_{h \rightarrow 0} \frac{f(x_0+h) - f(x_0)}{h} = f'(x_0)$$

this means that in a neighbourhood $U(x_0)$ of x_0 , $\forall x = x_0+h \in U(x_0)$ we have:

$$\frac{f(x_0+h) - f(x_0)}{h} = f'(x_0) + \theta(1) \text{ as } h \rightarrow 0$$

therefore:

$$\begin{aligned} f(x_0+h) - f(x_0) &= f'(x_0)h + \theta(1) \cdot h \\ &= f'(x_0)h + \theta(h) \text{ as } h \rightarrow 0 \end{aligned}$$

[$\theta(1) \cdot h = \theta(h)$ is true for the propositions of θ]

and

$$f(x_0+h) = f(x_0) + f'(x_0)h + \theta(h)$$

substitute $x = x_0+h \rightarrow x - x_0 = h$:

$$f(x) = f(x_0) + f'(x_0)(x - x_0) + \theta(x - x_0)$$

therefore $f(x)$ is differentiable at x_0 calling m the number $m = f'(x_0)$

$$\text{DEF. OF DIFFERENTIABLE} = f(x) = f(x_0) + \underbrace{f'(x_0)(x - x_0)}_{\rightarrow \text{differential } df(x_0)} + \theta(x - x_0) \text{ as } x \rightarrow x_0$$

DIFFERENTIAL

$$\begin{aligned} df(x_0) &= f'(x_0)(x - x_0) \\ &= f'(x_0) h \\ &= f'(x_0) \Delta x \\ &= f'(x_0) dx \end{aligned} \quad \left. \vphantom{\begin{aligned} df(x_0) &= f'(x_0)(x - x_0) \\ &= f'(x_0) h \\ &= f'(x_0) \Delta x \\ &= f'(x_0) dx \end{aligned}} \right\} \text{equal writings}$$

$$\Rightarrow f'(x_0) = \frac{df(x_0)}{dx}$$

ROLE OF THE DIFFERENTIAL

$$f(x) - f(x_0) = f'(x_0)(x - x_0) + \underbrace{\theta(x - x_0)}_{\text{negligible error}} \Rightarrow \Delta f(x_0) = df(x_0)$$

the differential $df(x_0)$ is a good approximation of $\Delta f(x_0)$ at least when $x \rightarrow x_0$ (when the increment $x - x_0$ is "very small")

REMARK

f has a derivative at $x_0 \iff f$ is differentiable at x_0

$\Uparrow \Downarrow$

$\Uparrow \Downarrow$

f is continuous at x_0

DISCONTINUITIES OF A DERIVATIVE FUNCTION

consider a function $g(x)$ defined on an interval $I \subseteq \mathbb{R}$ and suppose $g(x)$ is the derivative of another function $f(x) \rightarrow$ not sure $g(x)$ is continuous

THEOREM

consider a function $g(x)$ defined on an interval $I \subseteq \mathbb{R}$

suppose $g(x) = f'(x) \Rightarrow g(x)$ is a derivative of the function $f(x)$

then:

- ① $g(x)$ has no eliminable discontinuity
- ② $g(x)$ has no jump
- ③ $g(x)$ may have essential discontinuities (at least one of the two limits $\lim_{x \rightarrow x_0^+} g(x)$ and $\lim_{x \rightarrow x_0^-} g(x) \neq$ or is $+\infty/-\infty$)

\implies

- $C^{(1)}(A) = \{f: A \subseteq \mathbb{R} \rightarrow \mathbb{R}, \text{ derivable with a continuous derivative on } A\}$
- $C^{(2)}(A) = \{f: A \subseteq \mathbb{R} \rightarrow \mathbb{R}, \text{ twice-derivable with a continuous second derivative on } A\}$
- $C^{(n)}(A) = \{f: A \subseteq \mathbb{R} \rightarrow \mathbb{R}, n\text{-times derivable with a continuous } n\text{-th derivative on } A\}$
- $C^{(\infty)}(A) = \{f: A \subseteq \mathbb{R} \rightarrow \mathbb{R}, \text{ infinitely many times derivable with all its derivatives being continuous on } A\}$

CONTINUITY

① $f: A \subseteq \mathbb{R} \rightarrow \mathbb{R}, x_0 \in A$

f is continuous at $x_0 \in A$ if:

$\forall \epsilon > 0 \exists \delta = \delta(\epsilon)$ such that

$\forall x \in A, |x - x_0| < \delta \Rightarrow |f(x) - f(x_0)| < \epsilon$

② $f: A \subseteq \mathbb{R} \rightarrow \mathbb{R}$

f is continuous on A if f is continuous at all points $x_0 \in A$

that is if:

$\forall x \in A, \forall \epsilon > 0 \exists \delta = \delta(x_0, \epsilon)$ such that

$\forall x \in A, |x - x_0| < \delta \Rightarrow |f(x) - f(x_0)| < \epsilon$

\Rightarrow

if we change $x_0 \in A$ also δ changes

sometimes given $\epsilon > 0$ we can choose the same $\delta = \delta(\epsilon)$ for every point $x_0 \in A \rightarrow$

f is UNIFORMLY CONTINUOUS on $A =$

$f: A \subseteq \mathbb{R} \rightarrow \mathbb{R}$

f is uniformly continuous on A if $\forall \epsilon > 0 \exists \delta = \delta(\epsilon)$ such that:

$\forall x, x_0 \in A, |x - x_0| < \delta \Rightarrow |f(x) - f(x_0)| < \epsilon$

f uniformly continuous on $A \xRightarrow{\iff} f$ continuous on A

\hookrightarrow if $f(x)$ is continuous on A but it increases/decreases too steeply then it is not uniformly continuous, however the two notions are equivalent when the domain $A \subseteq \mathbb{R}$ is compact

THEOREM

$f: A \subseteq \mathbb{R} \rightarrow \mathbb{R}, A$ compact

f is continuous on $A \iff f$ is uniformly continuous on A

LIPSCHITZ

a function $f: A \subseteq \mathbb{R} \rightarrow \mathbb{R}$ is Lipschitz on A if $\exists k > 0$ such that:

$\forall x_1, x_2 \in A, |f(x_1) - f(x_2)| \leq k |x_1 - x_2|$

\Rightarrow

$\exists k > 0$ such that:

$\forall x_1, x_2 \in A, \text{ with } x_1 \neq x_2$

$$\left| \frac{f(x_1) - f(x_2)}{x_1 - x_2} \right| \leq k$$

this condition is linked with the behaviour of the difference quotient of f

THEOREM

$f: A \subseteq \mathbb{R} \rightarrow \mathbb{R}$

if f is Lipschitz on $A \Rightarrow f$ is uniformly continuous on A

therefore, for a general set $A \subseteq \mathbb{R}$ and a function $f: A \subseteq \mathbb{R} \rightarrow \mathbb{R}$

Lipschitz $\xRightarrow{\neq}$ uniformly continuous $\xRightarrow{\neq}$ continuous

there are no general implications between derivability and the Lipschitz condition: Lipschitz \nleftrightarrow derivability

however

THEOREM

if $f: A = [a, b] \rightarrow \mathbb{R}$, then:

$C^1(A) \Rightarrow$ Lipschitz (\nleftrightarrow)

USES OF DIFFERENTIAL CALCULUS: THE SEARCH FOR LOCAL MAX/MIN

$f'(x_0) = 0$ [horizontal tangent] $\nleftrightarrow x_0$ local maximiser/minimiser

FERMAT'S THEOREM (necessary condition for local maximisers/minimisers)

$f: A \subseteq \mathbb{R} \rightarrow \mathbb{R}$, if:

- ① x_0 is an interior point for A [excludes boundary points]
- ② f has a derivative at x_0 [excludes points at which $\nexists f'(x_0)$]
- ③ x_0 is a local maximiser/minimiser]

then:

$f'(x_0) = 0 \Rightarrow x_0 =$ stationary point

APPLICATION EXAMPLE OF FERMAT'S THEOREM

$y = x^3 - 3x \rightarrow$ find local max/min

① domain $A = \mathbb{R} \rightarrow$ no boundary points

f has a derivative in $A \rightarrow$ no points at which $\nexists f'(x)$

$$② \quad y' = 3x^2 - 3 \rightarrow \text{set } y' = 0 \Rightarrow 3x^2 - 3 = 0$$

$$\Rightarrow 3x^2 = 3$$

$$x^2 = 1 \rightarrow x = 1, x = -1$$

we know nothing about
these points

we set $y' = 0$ because all interior points and such that $\exists f'(x)$ and for which we have $f'(x) \neq 0$ cannot be local max/min

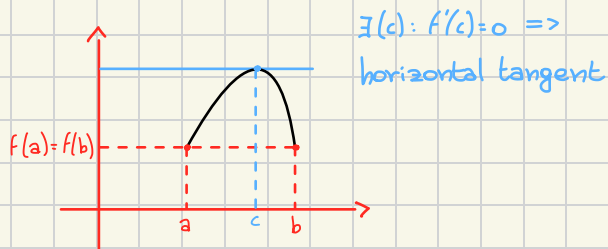
Fermat's theorem says that all such points are not useful because if they were local max/min they would have $y' = 0$ = necessary condition for local maximisers/minimisers

ROLLE'S THEOREM

$f: A \subseteq \mathbb{R} \rightarrow \mathbb{R}, A = [a, b]$, if:

- ① f is continuous in $[a, b]$
- ② f has a derivative in (a, b)
- ③ $f(a) = f(b)$

then, $\exists c \in (a, b): f'(c) = 0$

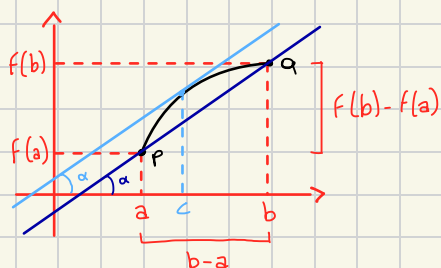


LAGRANGE'S THEOREM (mean value theorem) \rightarrow generalisation of Rolle's

$f: A \subseteq \mathbb{R} \rightarrow \mathbb{R}, A = [a, b]$, if:

- ① f is continuous in $[a, b]$
 - ② f has a derivative in (a, b)
- then, $\exists c \in (a, b): f'(c) = \frac{f(b) - f(a)}{b - a}$

\Rightarrow
 \exists at least one point $c \in (a, b)$: the tangent to the graph of $f(x)$ at c is parallel to the secant through $P(a, f(a))$, $Q(b, f(b))$



$\frac{f(b) - f(a)}{b - a}$ is the average rate (mean value) of variation of $f(x)$ in $[a, b]$;
 $\exists c: f'(c) = \frac{f(b) - f(a)}{b - a}$ means that $\exists c$ such that the instant rate of variation $f'(c)$ of $f(x)$ at c is equal to the average rate of variation $\frac{f(b) - f(a)}{b - a}$ of $f(x)$

INVERTIBILITY TEST (consequence of strict monotonicity test)

$f: A \subseteq \mathbb{R} \rightarrow \mathbb{R}$, if:

- ① A is an interval
 - ② f has a derivative in A
 - ③ $f'(x)$ always has the same sign in A [always > 0 or always < 0]
- $\Rightarrow f(x)$ is invertible in A (since $f(x)$ is strictly monotone on interval A)

SUFFICIENT CONDITION FOR LOCAL MAXIMISERS / MINIMISERS

$f: A \subseteq \mathbb{R} \rightarrow \mathbb{R}$, if:

- ① x_0 is an interior point of A
 - ② f has a derivative in a neighbourhood $B(x_0)$ of x_0
 - ③ $f'(x_0) = 0$ [$\Rightarrow x_0$ passes Fermat's theorem]
- * $\left[\begin{array}{l} \leq \\ \text{strict} \end{array} \right]$

then: *

- ① if $f'(x) \leq 0$ in $B^-(x_0)$ and $f'(x) \geq 0$ in $B^+(x_0) \Rightarrow x_0$ is a local minimiser
- ② if $f'(x) \geq 0$ in $B^-(x_0)$ and $f'(x) \leq 0$ in $B^+(x_0) \Rightarrow x_0$ is a local maximiser
- ③ if $f'(x) < 0$ in $B^-(x_0)$ and $f'(x) < 0$ in $B^+(x_0) \Rightarrow x_0$ is not a local max/min

EXAMPLE

Find local maximisers / minimisers of $y = x^3 - 3x$

\Rightarrow

① $y' = 3x^2 - 3 \rightarrow y' = 0$

$x = 1, x = -1$

② study the sign of y'

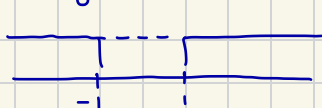
$y' > 0 \rightarrow 3x^2 - 3 > 0$

$3x^2 > 3$

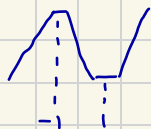
$x^2 > 1$

$x < -1$ or $x > 1$ [$y' < 0$ for $-1 < x < 1$]

③ we get



sign of y'

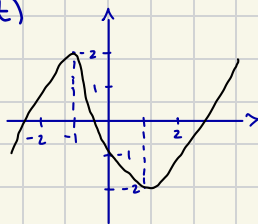


y increasing/decreasing

$\Rightarrow x = -1$ is a local maximiser (strict)

$x = 1$ is a local minimiser (strict)

$f(-1) = 2, f(1) = -2$



not global max/min

DE L'HÔPITAL'S THEOREM \Rightarrow computation of limits

INDETERMINATE FORM $\frac{0}{0}$

suppose f, g are differentiable on (a, b) , with $a, b \in \bar{\mathbb{R}} = \mathbb{R} \cup \{+\infty, -\infty\}$

suppose $x_0 \in [a, b]$

suppose $\lim_{x \rightarrow x_0} f(x) = \lim_{x \rightarrow x_0} g(x) = 0$

suppose $g'(x) \neq 0 \forall x \in (a, b)$

then:

if $\lim_{x \rightarrow x_0} \frac{f'(x)}{g'(x)} = L \in \bar{\mathbb{R}} \Rightarrow \lim_{x \rightarrow x_0} \frac{f(x)}{g(x)} = L$

INDETERMINATE FORM $\frac{\infty}{\infty}$

suppose f, g are differentiable on (a, b) , with $a, b \in \bar{\mathbb{R}} = \mathbb{R} \cup \{+\infty, -\infty\}$

suppose $x_0 \in [a, b]$

suppose $\lim_{x \rightarrow x_0} f(x) = +\infty / -\infty, \lim_{x \rightarrow x_0} g(x) = +\infty / -\infty$

suppose $g'(x) \neq 0 \forall x \in (a, b)$

then:

if $\lim_{x \rightarrow x_0} \frac{f'(x)}{g'(x)} = L \in \bar{\mathbb{R}} \Rightarrow \lim_{x \rightarrow x_0} \frac{f(x)}{g(x)} = L$

EXAMPLES

notable limits $\frac{0}{0}$

$$\textcircled{1} \lim_{x \rightarrow 0} \frac{e^x - 1}{x} \Rightarrow \lim_{x \rightarrow 0} \frac{e^x}{1} = 1 = \lim_{x \rightarrow 0} \frac{e^x - 1}{x} = 1$$

$$\textcircled{2} \lim_{x \rightarrow 0} \frac{\ln(1+x)}{x} \Rightarrow \lim_{x \rightarrow 0} \frac{\frac{1}{1+x}}{1} = 1 = \lim_{x \rightarrow 0} \frac{\ln(1+x)}{1} = 1$$

$$\textcircled{3} \lim_{x \rightarrow 0} \frac{(1+x)^\alpha - 1}{x} \Rightarrow \lim_{x \rightarrow 0} \frac{\alpha(1+x)^{\alpha-1}}{1} = \alpha = \lim_{x \rightarrow 0} \frac{(1+x)^\alpha - 1}{x} = \alpha$$

$$\textcircled{4} \lim_{x \rightarrow 0} \frac{\sin x}{x} \Rightarrow \lim_{x \rightarrow 0} \frac{\cos x}{1} = 1 = \lim_{x \rightarrow 0} \frac{\sin x}{x} = 1$$

scales of infinities $\frac{\infty}{\infty}$

$$\textcircled{1} \lim_{x \rightarrow +\infty} \frac{\ln x}{x} \Rightarrow \lim_{x \rightarrow +\infty} \frac{1/x}{1} = 0 = \lim_{x \rightarrow +\infty} \frac{\ln x}{x} = 0$$

$$\textcircled{2} \lim_{x \rightarrow +\infty} \frac{e^x}{x^2} \Rightarrow \lim_{x \rightarrow +\infty} \frac{e^x}{2x} = \lim_{x \rightarrow +\infty} \frac{e^x}{2} = +\infty = \lim_{x \rightarrow +\infty} \frac{e^x}{x^2} = +\infty$$

SUFFICIENT CONDITION FOR DERIVABILITY

$f: A \subseteq \mathbb{R} \rightarrow \mathbb{R}$, x_0 interior point for A , suppose:

① f is continuous in a neighbourhood $\mathcal{U}(x_0)$

② f has a derivative in $\mathcal{U}(x_0) - \{x_0\}$

then:

③ if $\lim_{x \rightarrow x_0^+} f'(x) \exists$ finite $\Rightarrow f'_+(x_0) = \lim_{x \rightarrow x_0^+} f'(x)$

④ if $\lim_{x \rightarrow x_0^-} f'(x) \exists$ finite $\Rightarrow f'_-(x_0) = \lim_{x \rightarrow x_0^-} f'(x)$

⑤ if $\lim_{x \rightarrow x_0} f'(x) \exists$ finite $\Rightarrow f'(x_0) = \lim_{x \rightarrow x_0} f'(x)$

TAYLOR'S EXPANSION / MACLAURIN'S EXPANSION

TAYLOR'S THEOREM (ORDER 2)

$f: A \subseteq \mathbb{R} \rightarrow \mathbb{R}$, then:

① x_0 interior point of A

② f has a derivative $f'(x)$ and second derivative $f''(x)$ in a neighbourhood $\mathcal{U}(x_0)$ [is twice differentiable]

then:

$\forall x = x_0 + h \in \mathcal{U}(x_0)$,

$$f(x) = \underbrace{f(x_0) + f'(x_0)(x-x_0) + \frac{f''(x_0)}{2}(x-x_0)^2}_{\substack{\text{tangent parabola at } x_0 \\ = \\ \text{Taylor's polynomial of degree 2}}} + \underbrace{\mathcal{O}(x-x_0)^2}_{\substack{\text{negligible error} \\ = \\ \text{Peano's remainder}}}$$

Taylor's expansion of order 2
as $x \rightarrow x_0$

- Taylor's formula / expansion \rightarrow polynomial + remainder \Rightarrow use order
- Taylor's polynomial \rightarrow no remainder \Rightarrow use degree

MACLAURIN'S FORMULA / EXPANSION (special case of Taylor when $x_0 = 0$)

$\forall x \in \mathcal{U}(0)$,

$$f(x) = \underbrace{f(0) + f'(0)x + \frac{f''(0)}{2}x^2}_{\substack{\text{tangent parabola} \\ = \\ \text{Maclaurin's polynomial}}} + \underbrace{\mathcal{O}(x^2)}_{\substack{\text{negligible error} \\ = \\ \text{Peano's remainder}}}$$

as $x \rightarrow x_0$

TAYLOR'S THEOREM (ORDER n)

$f: A \subseteq \mathbb{R} \rightarrow \mathbb{R}$, if:

① x_0 interior point of A

② f is n times differentiable in a neighbourhood $\mathcal{U}(x_0)$

then:

$$\forall x = x_0 + h \in \mathcal{U}(x_0)$$

$$f(x) = f(x_0) + f'(x_0)(x-x_0) + \frac{f''(x_0)}{2!}(x-x_0)^2 + \frac{f'''(x_0)}{3!}(x-x_0)^3 + \dots + \frac{f^{(n)}(x_0)}{n!}(x-x_0)^n$$

$$+ \mathcal{O}(x-x_0)^n$$

Taylor's polynomial of degree n

as $x \rightarrow x_0$

Peano's remainder

\hookrightarrow MACLAURIN'S FORMULA $\Rightarrow x_0 = 0$

MACLAURIN'S EXPANSION OF ELEMENTARY FUNCTIONS

① $y = e^x, x_0 = 0$

$$e^x = 1 + x + \frac{x^2}{2!} + \frac{x^3}{3!} + \dots + \frac{x^n}{n!} + \dots \quad x \rightarrow 0$$

② $y = \sin(x), x_0 = 0$

$$\sin(x) = x - \frac{x^3}{3!} + \frac{x^5}{5!} - \frac{x^7}{7!} + \dots + (-1)^n \frac{x^{2n+1}}{(2n+1)!} + \dots$$

③ $y = \cos(x), x_0 = 0$

$$\cos(x) = 1 - \frac{x^2}{2!} + \frac{x^4}{4!} - \frac{x^6}{6!} + \dots + (-1)^{n+1} \frac{x^n}{(2n)!} + \dots \quad x \rightarrow 0$$

④ $y = \ln(1+x), x_0 = 0$

$$\ln(1+x) = x - \frac{x^2}{2} + \frac{x^3}{3} - \frac{x^4}{4} + \dots + (-1)^{n+1} \frac{x^n}{n} + \dots \quad x \rightarrow 0$$

⑤ $y = (1+x)^\alpha, x_0 = 0$

$$(1+x)^\alpha = 1 + \alpha x + \frac{\alpha(\alpha-1)}{2!} x^2 + \frac{\alpha(\alpha-1)(\alpha-2)}{3!} x^3 + \dots + \frac{\alpha(\alpha-1)(\alpha-2)\dots(\alpha-(n-1))}{n!} x^n + \dots \quad x \rightarrow 0$$

II SUFFICIENT CONDITION FOR LOCAL MAXIMISERS/MINIMISERS

$f: A \subseteq \mathbb{R} \rightarrow \mathbb{R}$, suppose that:

① x_0 is an interior point of A

② f is n times differentiable in a neighbourhood $\mathcal{U}(x_0)$

③ $f'(x_0) = 0$

④ all the higher-order derivatives at x_0 are $= 0$ till $\exists n \geq 2$ such that $f^{(n)}(x_0) \neq 0$

then:

① if n is even and $f^{(n)}(x_0) > 0 \Rightarrow x_0$ is a local minimiser

② if n is even and $f^{(n)}(x_0) < 0 \Rightarrow x_0$ is a local maximiser

③ if n is odd $\Rightarrow x_0$ is not a local maximiser/minimiser

THEOREM (weaker because cannot be applied if $f''(x_0) = 0$)

$f: A \subseteq \mathbb{R} \rightarrow \mathbb{R}$, suppose that:

① x_0 is an interior point of A

② f is twice differentiable in a neighbourhood $\mathcal{U}(x_0)$

③ $f'(x_0) = 0$

then:

① if $f''(x_0) > 0 \Rightarrow x_0$ is a local minimiser

② if $f''(x_0) < 0 \Rightarrow x_0$ is a local maximiser

GENERAL PROCEDURE FOR LOCAL MAXIMUM / MINIMUM

given $f: A \subseteq \mathbb{R} \rightarrow \mathbb{R}$

- ① - points $x_0 \in A$ that are boundary points \rightarrow check separately
- points $x_0 \in A$ such that $\nexists f'(x_0) \rightarrow$ check separately
- ② all other points \rightarrow Fermat's necessary condition $\Rightarrow f'(x) = 0$
- ③ on those points which pass Fermat's condition \rightarrow
 - I sufficient condition \Rightarrow only $f'(x)$ but involves inequalities
or
 - II sufficient condition \Rightarrow also higher-order derivatives but no inequalities involved

CONCAVITY / CONVEXITY

THEOREM: CONVEXITY AND TANGENT

$f: A \subseteq \mathbb{R} \rightarrow \mathbb{R}$

if A is an open interval and f is differentiable on A , then:

f is convex in $A \Leftrightarrow \forall x_0, x \in A,$
 $f(x_0) + f'(x_0)(x - x_0) \leq f(x)$ [CONCAVE $\Rightarrow \geq f(x)$]

THEOREM: CONVEXITY TEST FOR DIFFERENTIABLE FUNCTIONS

$f: A \subseteq \mathbb{R} \rightarrow \mathbb{R}$

if A is an open interval and f is differentiable on A , then:

- ① f is convex in $A \Leftrightarrow f'$ is increasing in A [CONCAVE \Rightarrow decreasing]
- ② f is strictly convex in $A \Leftrightarrow f'$ is strictly increasing in A

THEOREM: CONVEXITY TEST FOR TWICE-DIFFERENTIABLE FUNCTIONS

$f: A \subseteq \mathbb{R} \rightarrow \mathbb{R}$

if A is an open interval and f is twice-differentiable on A , then:

- ① f is convex in $A \Leftrightarrow f'' \geq 0$ in A [CONCAVE $\Rightarrow f'' \leq 0$]
- ② $f'' > 0$ in $A \Rightarrow f$ is strictly convex in A [CONCAVE $\Rightarrow f'' < 0$]

GLOBAL MAXIMA / MINIMA

GENERAL METHODS TO SEE IF GLOBAL MAXIMA / MINIMA EXIST

- ① draw the graph
- ② use Weierstrass (or Tonelli) \rightarrow if the function $f(x)$ is continuous on a domain $A \subseteq \mathbb{R}$ which is a compact set $A \subseteq \mathbb{R}$, Weierstrass's theorem guarantees that both a global maximum and a global minimum \exists BUT it does not say how to find them

THEOREM: I SUFFICIENT CONDITION, GLOBAL VERSION

$f: A \subseteq \mathbb{R} \rightarrow \mathbb{R}$, if:

- ① A is an interval
- ② x_0 is an interior point of A
- ③ f has a finite derivative in A
- ④ $f'(x_0) = 0$

$\left[\begin{array}{l} \geq \\ \leq \end{array} \right]$ strict

then: \Leftarrow

- ③ if $f'(x) \leq 0 \quad \forall x \in A, x < x_0$ and $f'(x) \geq 0 \quad \forall x \in A, x > x_0$
then $x_0 =$ global minimiser
- ④ if $f'(x) \geq 0 \quad \forall x \in A, x < x_0$ and $f'(x) \leq 0 \quad \forall x \in A, x > x_0$
then $x_0 =$ global maximiser

THEOREM: II SUFFICIENT CONDITION, GLOBAL VERSION

$f: A \subseteq \mathbb{R} \rightarrow \mathbb{R}$, if:

- ① A is an interval
- ② x_0 is an interior point of A
- ③ f is twice differentiable in A
- ④ $f'(x_0) = 0$

$\left[\begin{array}{l} \geq \\ \leq \end{array} \right]$ strict

then: \Leftarrow

- ③ if $f''(x) \geq 0$ in A then $x_0 =$ global minimiser
- ④ if $f''(x) \leq 0$ in A then $x_0 =$ global maximiser

\rightarrow or MINIMUM

THEOREM: NECESSARY AND SUFFICIENT MAXIMUM CONDITION FOR A CONCAVE AND DIFFERENTIABLE FUNCTION

\hookrightarrow or CONVEX

$f: A \subseteq \mathbb{R} \rightarrow \mathbb{R}$, if:

- ① A is an interval
- ② f is concave (or convex) in A

③ f has a finite derivative in A

④ x_0 interior point of A

then:

$f'(x_0) = 0 \iff x_0$ is a local maximiser (or minimiser) $\iff x_0$ is a global maximiser (or minimiser)

HOW TO DRAW A GRAPH

① domain A

② limits at the boundary \rightarrow horizontal/vertical asymptotes

③ sign of $f(x)$ + intersections with coordinate axes

④ study of $f'(x) \rightarrow f$ increasing/decreasing + local max/min (I sufficient condition)

⑤ study of $f''(x) \rightarrow f$ convex/concave + local max/min (II sufficient condition)

⑥ graph + study of exceptional points + global max/min

EXAMPLE $\rightarrow y = \frac{x^2+2}{x^2-2}$

① domain

$$x^2 - 2 \neq 0 \Rightarrow x \neq \sqrt{2}, -\sqrt{2}$$

$$A = (-\infty, -\sqrt{2}) \cup (-\sqrt{2}, \sqrt{2}) \cup (\sqrt{2}, +\infty)$$

② limits / asymptotes

$$\lim_{x \rightarrow -\infty} \frac{x^2+2}{x^2-2} = 1^+$$

horizontal asymptote
 $y=1$ as $x \rightarrow -\infty$

$$\lim_{x \rightarrow +\infty} \frac{x^2+2}{x^2-2} = 1^+$$

horizontal asymptote
 $y=1$ as $x \rightarrow +\infty$

$$\lim_{x \rightarrow \sqrt{2}^-} \frac{x^2+2}{x^2-2} = -\infty$$

$$\lim_{x \rightarrow \sqrt{2}^+} \frac{x^2+2}{x^2-2} = +\infty$$

vertical asymptote
 $x = \sqrt{2}$

$$\lim_{x \rightarrow -\sqrt{2}^-} \frac{x^2+2}{x^2-2} = +\infty$$

$$\lim_{x \rightarrow -\sqrt{2}^+} \frac{x^2+2}{x^2-2} = -\infty$$

vertical asymptote
 $x = -\sqrt{2}$

③ sign / intersection with axes

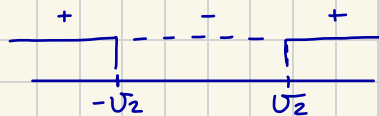
$$\frac{x^2+2}{x^2-2} > 0$$

$N > 0$ always

$D > 0$ if $x^2 - 2 > 0$

$$\Rightarrow x^2 > 2$$

$$\Rightarrow x < -\sqrt{2}, x > \sqrt{2}$$



intersection
with y-axes

$$\begin{cases} x=0 \\ y=-1 \end{cases}$$

intersection
with x-axes

$$\begin{cases} y=0 \\ \text{never} \end{cases}$$

④ $f'(x)$

$$\begin{aligned} f(x) = \frac{x^2+2}{x^2-2} &\rightarrow y' = \frac{(x^2-2)2x - (x^2+2)2x}{(x^2-2)^2} \\ &= \frac{2x^3 - 4x - 2x^3 - 4x}{(x^2-2)^2} \\ &= \frac{-8x}{(x^2-2)^2} \end{aligned}$$

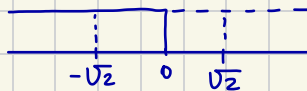
$$y' = 0 \rightarrow -8x = 0 \Rightarrow x = 0$$

$$y' > 0 \rightarrow \frac{-8x}{(x^2-2)^2} > 0$$

$D > 0$ (excluded $x = -\sqrt{2}, \sqrt{2}$ where $f(x) \nexists$)

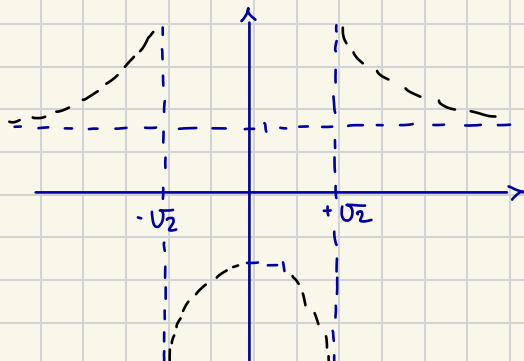
$$N > 0 \quad -8x > 0 \Rightarrow x < 0$$

$x = 0$ is local maximiser



⑤ $f''(x)$ not required

⑥ graph



**theorems and proofs for
differential calculus**

DERIVATIVE OF A CONSTANT

$$y = k \Rightarrow y' = 0 \quad (\forall \text{ constant } k \in \mathbb{R})$$

PROOF

$$f: \mathbb{R} \rightarrow \mathbb{R}, f(x) = k, \forall x \in \mathbb{R}, \forall k \neq 0:$$

$$f'(x) = \lim_{h \rightarrow 0} \frac{f(x_0+h) - f(x_0)}{h} = \lim_{h \rightarrow 0} \frac{k - k}{h} = \lim_{h \rightarrow 0} \frac{0}{k} = 0$$

DERIVATIVE OF $y = x^\alpha$

$$y = x^\alpha \Rightarrow y' = \alpha x^{\alpha-1}$$

PROOF

$$f'(x) = \lim_{h \rightarrow 0} \frac{f(x_0+h) - f(x_0)}{h} = \lim_{h \rightarrow 0} \frac{(x+h)^\alpha - x^\alpha}{h} =$$
$$\lim_{h \rightarrow 0} \frac{[x(1+\frac{h}{x})]^\alpha - x^\alpha}{h} = \lim_{h \rightarrow 0} \frac{x^\alpha (1+\frac{h}{x})^\alpha - x^\alpha}{h} = \lim_{h \rightarrow 0} \frac{x^\alpha (1+\frac{h}{x})^\alpha - 1}{h} =$$
$$\lim_{t \rightarrow 0} x^{\alpha-1} \frac{(1+t)^\alpha - 1}{t} = \alpha x^{\alpha-1}$$
$$\frac{x^\alpha}{h} = \frac{x^{\alpha-1} \cdot x}{h} =$$
$$x^{\alpha-1} \left[\frac{1}{\frac{h}{x}} \right] = t$$

DERIVATIVE OF $y = e^x$

$$f: \mathbb{R} \rightarrow \mathbb{R}, f(x) = a^x, a > 0$$

is derivable $\forall x \in \mathbb{R}$, $f': \mathbb{R} \rightarrow \mathbb{R}$ given by $f'(x) = a^x \log a$

in particular:

$$y = e^x \Rightarrow y' = e^x$$

PROOF

$$f'(x) = \lim_{h \rightarrow 0} \frac{f(x+h) - f(x)}{h} = \lim_{h \rightarrow 0} \frac{a^{x+h} - a^x}{h} = \lim_{h \rightarrow 0} \frac{a^x (a^h - 1)}{h} = a^x \lim_{h \rightarrow 0} \frac{a^h - 1}{h} = a^x \log a$$

DERIVATIVE OF A LINEAR COMBINATION

suppose f, g have a derivative $\forall x \in (a, b)$

suppose $\alpha, \beta \in \mathbb{R}$

$\alpha f + \beta g$ has a derivative $\forall x \in (a, b)$ and:

$$(\alpha f + \beta g)' = \alpha f' + \beta g'$$

f, g have a der. $\forall x \in (a, b)$

$\alpha, \beta \in \mathbb{R}$

$\alpha f + \beta g$ has a der. $\forall x \in (a, b)$

$$(\alpha f + \beta g)' = \alpha f' + \beta g'$$

PROOF:

$\forall x \in (a, b),$

$$(\alpha f + \beta g)'(x) = \lim_{h \rightarrow 0} \frac{(\alpha f + \beta g)(x+h) - (\alpha f + \beta g)(x)}{h} =$$

$$\lim_{h \rightarrow 0} \frac{(\alpha f)(x+h) + (\beta g)(x+h) - (\alpha f)(x) - (\beta g)(x)}{h} =$$

$$\lim_{h \rightarrow 0} \left[\alpha \frac{f(x+h) - f(x)}{h} + \beta \frac{g(x+h) - g(x)}{h} \right] =$$

$$\alpha \lim_{h \rightarrow 0} \frac{f(x+h) - f(x)}{h} + \beta \lim_{h \rightarrow 0} \frac{g(x+h) - g(x)}{h} = \alpha f'(x) + \beta g'(x)$$

DERIVATIVE OF $y = \ln x$

$$y = \ln x \Rightarrow y' = \frac{1}{x}$$

PROOF

$$f'(x_0) = \lim_{h \rightarrow 0} \frac{f(x_0+h) - f(x_0)}{h} = \lim_{h \rightarrow 0} \frac{\ln(x+h) - \ln x}{h} = \lim_{h \rightarrow 0} \frac{\ln\left(\frac{x+h}{x}\right)}{h}$$

$$= \lim_{h \rightarrow 0} \frac{1}{h} \ln\left(1 + \frac{h}{x}\right) = \lim_{h \rightarrow 0} \ln\left(1 + \frac{h}{x}\right)^{\frac{1}{h}}$$

$$\text{let } t = \frac{h}{x}; h = tx; \frac{1}{h} = \frac{1}{tx}; \frac{1}{t} \cdot \frac{1}{x}$$

$$= \lim_{t \rightarrow 0} \ln\left(1+t\right)^{\frac{1}{t} \cdot \frac{1}{x}} = \lim_{t \rightarrow 0} \ln\left[\left(1+t\right)^{\frac{1}{t}}\right]^{\frac{1}{x}} = \frac{1}{x} \lim_{t \rightarrow 0} \ln\left(1+t\right)^{\frac{1}{t}}$$

$$= \frac{1}{x} \ln\left[\underbrace{\lim_{t \rightarrow 0} \left(1+t\right)^{\frac{1}{t}}}_{=e}\right] = \frac{1}{x} \underbrace{\ln e}_{=1} = \frac{1}{x}$$

DERIVATIVE OF $y = \arctan x$

$f: [-\frac{\pi}{2}; \frac{\pi}{2}] \rightarrow \mathbb{R}$ given by $f(y) = \tan y$

then, $f^{-1}: \mathbb{R} \rightarrow [-\frac{\pi}{2}; \frac{\pi}{2}]$ is given by $f^{-1}(x) = \arctan x$

PROOF

from the derivative of the inverse function

$$\frac{d \arctan x}{dx} = \frac{1}{1 + \tan^2 y} = \frac{1}{1 + x^2}$$

RELATIONSHIP BETWEEN DERIVABILITY AND CONTINUITY

suppose $f: (a, b) \rightarrow \mathbb{R}$ has a derivative $f'(x_0)$ at a point $x_0 \in (a, b)$
then f is continuous at the point x_0

PROOF

$$\text{suppose } \lim_{h \rightarrow 0} \frac{f(x_0+h) - f(x_0)}{h} = f'(x_0)$$

rewrite it using:

$$x_0+h=x \Rightarrow h = x - x_0$$

$$h \rightarrow 0 = x \rightarrow x_0$$

therefore

$$\lim_{x \rightarrow x_0} \frac{f(x) - f(x_0)}{x - x_0} = f'(x_0)$$

then

$$\lim_{x \rightarrow x_0} [f(x) - f(x_0)] = \lim_{x \rightarrow x_0} \left[\frac{f(x) - f(x_0)}{x - x_0} (x - x_0) \right]$$

$$= \lim_{x \rightarrow x_0} \frac{f(x) - f(x_0)}{x - x_0} \cdot \lim_{x \rightarrow x_0} (x - x_0) = f'(x_0) \cdot 0 = 0$$

but also

$$\lim_{x \rightarrow x_0} [f(x) - f(x_0)] = \lim_{x \rightarrow x_0} f(x) - \lim_{x \rightarrow x_0} f(x_0) = \lim_{x \rightarrow x_0} f(x) - f(x_0)$$

and therefore

$$\lim_{x \rightarrow x_0} f(x) - f(x_0) = 0 \Rightarrow \lim_{x \rightarrow x_0} f(x) = f(x_0)$$

meaning that $f(x)$ is continuous at x_0

but f continuous at $x_0 \not\Rightarrow f$ has a derivative at x_0

RELATIONSHIP BETWEEN DERIVABILITY AND DIFFERENTIABILITY

$f: A \subseteq \mathbb{R} \rightarrow \mathbb{R}$, x_0 interior point for A

then

① f is differentiable at $x_0 \iff f$ has a derivative at x_0

② $m = f'(x_0)$

PROOF

since $f(x)$ is differentiable at x_0 , we have:

$$f(x) = f(x_0) + m(x - x_0) + \theta(x - x_0) \quad \text{as } x \rightarrow x_0$$

substitute $x = x_0 + h \rightarrow h = x - x_0$:

$$f(x_0 + h) = f(x_0) + m \cdot h + \theta(h) \quad \text{as } h \rightarrow 0$$

$$f(x_0 + h) - f(x_0) = m \cdot h + \theta(h) \quad \text{as } h \rightarrow 0$$

$$\frac{f(x_0 + h) - f(x_0)}{h} = m + \frac{\theta(h)}{h}$$

$$= m + \theta(1) \quad \text{as } h \rightarrow 0$$

$\left[\frac{\theta(h)}{h} = \theta(1) \right]$ is true for the properties of θ

we take the limit on both sides when $h \rightarrow 0$ and obtain:

$$\lim_{h \rightarrow 0} \frac{f(x_0 + h) - f(x_0)}{h} = m$$

which means that $f(x)$ has a derivative $f'(x_0)$ at x_0 and:

$$f'(x_0) = m$$

\Leftarrow

since $f(x)$ has a derivative at x_0 , we have:

$$\lim_{h \rightarrow 0} \frac{f(x_0 + h) - f(x_0)}{h} = f'(x_0)$$

this means that in a neighbourhood $U(x_0)$ of x_0 , $\forall x = x_0 + h \in U(x_0)$ we have:

$$\frac{f(x_0 + h) - f(x_0)}{h} = f'(x_0) + \theta(1) \quad \text{as } h \rightarrow 0$$

therefore:

$$\begin{aligned} f(x_0 + h) - f(x_0) &= f'(x_0)h + \theta(1) \cdot h \\ &= f'(x_0)h + \theta(h) \quad \text{as } h \rightarrow 0 \end{aligned}$$

$\left[\theta(1) \cdot h = \theta(h) \right]$ is true for the propositions of θ

and

$$f(x_0+h) = f(x_0) + f'(x_0)h + o(h)$$

substitute $x = x_0 + h \rightarrow x - x_0 = h$:

$$f(x) = f(x_0) + f'(x_0)(x - x_0) + o(x - x_0)$$

therefore $f(x)$ is differentiable at x_0 calling m the number
 $m = f'(x_0)$

FERMAT'S THEOREM

$f: A \subseteq \mathbb{R} \rightarrow \mathbb{R}$ defined on a set A in \mathbb{R}

C subset of A

f differentiable at an interior point $x_0 \in C$

x_0 maximiser/minimiser of f on C

then $f'(x_0) = 0 \Rightarrow x_0 = \text{stationary point}$

PROOF

let $x_0 \in C$ interior point and a local maximiser on C

\exists a neighbourhood $B_\epsilon(x_0)$ such that $f(x_0) \geq f(x) \forall x \in B_\epsilon(x_0) \cap C$

$h \in (0, \epsilon), x_0 + h \in B_\epsilon(x_0) \Rightarrow$

$$\frac{f(x_0+h) - f(x_0)}{h} \leq 0 \quad \forall h \in (0, \epsilon)$$

$$\lim_{h \rightarrow 0^+} \frac{f(x_0+h) - f(x_0)}{h} \leq 0$$

$h \in (-\epsilon, 0), x_0 + h \in B_\epsilon(x_0) \Rightarrow$

$$\frac{f(x_0+h) - f(x_0)}{h} \geq 0 \quad \forall h \in (-\epsilon, 0)$$

$$\lim_{h \rightarrow 0^-} \frac{f(x_0+h) - f(x_0)}{h} \geq 0$$

together:

$$0 \leq \lim_{h \rightarrow 0^-} \frac{f(x_0+h) - f(x_0)}{h} = \lim_{h \rightarrow 0} \frac{f(x_0+h) - f(x_0)}{h} = \lim_{h \rightarrow 0^+} \frac{f(x_0+h) - f(x_0)}{h} \leq 0$$

Since the hypothesis $f'(x_0)$ exists:

$$f'(x_0) = \lim_{h \rightarrow 0} \frac{f(x_0+h) - f(x_0)}{h} = 0$$

ROLLE'S THEOREM

$f: A \subseteq \mathbb{R} \rightarrow \mathbb{R}$, $A = [a, b]$, if:

- ① f is continuous in $[a, b]$
- ② f has a derivative in (a, b)
- ③ $f(a) = f(b)$

then, $\exists x_0 \in (a, b): f'(x_0) = 0$

PROOF

according to Weierstrass $\exists x_1, x_2 \in [a, b]$ such that:

$$f(x_1) = \min_{x \in [a, b]} f(x) = m$$

$$f(x_2) = \max_{x \in [a, b]} f(x) = \pi$$

if $m = \pi$ the f is constant, $f(x) = m = \pi$ so $f'(x) = 0 \forall x \in (a, b)$

if $m < \pi$ then at least one between x_1 and x_2 must be an interior point for (a, b) , in fact they can't be both boundary points since $f(a) = f(b)$

if $x_1 \in (a, b)$ for the Fermat's theorem $f'(x_1) = 0$, same in the case of x_2

LAGRANGE'S VALUE THEOREM

Let $f: [a, b] \rightarrow \mathbb{R}$ be continuous on $[a, b]$ and differentiable on (a, b)

then there exists $\hat{x} \in (a, b)$ such that:

$$f'(\hat{x}) = \frac{f(b) - f(a)}{b - a} *$$

PROOF

let $g: [a, b] \rightarrow \mathbb{R}$ be the auxiliary function defined by:

$$g(x) = f(x) - \left(f(a) + \frac{f(b) - f(a)}{b - a} (x - a) \right)$$

it is the difference between f and the straight line passing through the points $(a, f(a))$ and $(b, f(b))$

the function g is continuous on $[a, b]$ and differentiable on (a, b)

$$\text{moreover } g(a) = g(b) = 0$$

by Rolle's theorem there exists $\hat{x} \in (a, b)$ such that $g'(\hat{x}) = 0$, but

$$g'(x) = f'(x) - \frac{f(b) - f(a)}{b - a} \text{ and therefore } f'(\hat{x}) - \frac{f(b) - f(a)}{b - a} = 0$$

that is, \hat{x} satisfies condition *

CHARACTERIZATION OF FUNCTIONS WITH NULL DERIVATIVE

let $f: [a, b] \rightarrow \mathbb{R}$ be continuous on $[a, b]$ and differentiable on (a, b)

then $f'(x) = 0$ for every $x \in (a, b)$ if and only if f is constant, that is, if and only if there exists $k \in \mathbb{R}$ such that $f(x) = k \quad \forall x \in [a, b]$

PROOF

"only if" \Rightarrow

"if" part is the simple property of derivatives of derivability and continuity

let $x \in (a, b]$ and let us apply the mean value theorem on the interval $[a, x]$

it yields a point $\xi \in (a, x)$ such that:

$$0 = f'(\xi) = \frac{f(x) - f(a)}{x - a} \quad \text{that is } f(x) = f(a)$$

since x is any point in $(a, b]$ it follows that $f(x) = f(a)$ for any $x \in [a, b]$

CHARACTERIZATION OF FUNCTIONS WITH THE SAME DERIVATIVE

Let $f, g : [a, b] \rightarrow \mathbb{R}$ be continuous on $[a, b]$ and differentiable on (a, b)

then $f'(x) = g'(x) \quad \forall x \in (a, b)$ if and only if there exists $k \in \mathbb{R}$ such that $f(x) = g(x) + k$
 $\forall x \in [a, b]$

two functions that have the same first derivative are thus equal up to an additive constant k

PROOF

since "if" is obvious we prove "only if" \Rightarrow

let $h : [a, b] \rightarrow \mathbb{R}$ be the auxiliary function $h(x) = f(x) - g(x)$

we have $h'(x) = f'(x) - g'(x) = 0 \quad \forall x \in (a, b)$

therefore by the characterization of functions with null derivative h is constant on $[a, b]$

that is there exists $k \in \mathbb{R}$ such that $h(x) = k \quad \forall x \in [a, b]$ so $f(x) = g(x) + k \quad \forall x \in [a, b]$

MONOTONICITY TEST ON INTERVAL

let $f: (a, b) \rightarrow \mathbb{R}$ be a differentiable function with $a, b \in \bar{\mathbb{R}}$

then f is (globally) increasing on (a, b) if and only if $f'(x) \geq 0 \forall x \in (a, b)$

because of the clause $a, b \in \bar{\mathbb{R}}$ the interval (a, b) can be unbounded for example $(a, b) = \mathbb{R}$

a dual result with negativity of the derivative on (a, b) holds for the decreasing monotonicity

PROOF

"only if" \Rightarrow

suppose that f is increasing, let $x \in (a, b)$

$\forall h > 0$ we have $f(x+h) \geq f(x)$ hence $\frac{f(x+h) - f(x)}{h} \geq 0$

it follows that:

$$f'(x) = \lim_{h \rightarrow 0} \frac{f(x+h) - f(x)}{h} = \lim_{h \rightarrow 0^+} \frac{f(x+h) - f(x)}{h} \geq 0$$

"if" \Rightarrow

let $f'(x) \geq 0 \forall x \in (a, b)$

let $x_1, x_2 \in (a, b)$ with $x_1 < x_2$

by the mean value theorem there exists $\hat{x} \in [x_1, x_2]$ such that $f'(\hat{x}) = \frac{f(x_2) - f(x_1)}{x_2 - x_1}$

since $f'(\hat{x}) \geq 0$ and $x_2 - x_1 > 0$, this shows that $f(x_2) \geq f(x_1)$

STRICT MONOTONICITY TEST ON AN INTERVAL

let $f: (a, b) \rightarrow \mathbb{R}$ be a differentiable function, with $a, b \in \bar{\mathbb{R}}$

if $f'(x) > 0 \quad \forall x \in (a, b)$, then f is (globally) strictly increasing on (a, b)

PROOF

let $f'(x) > 0 \quad \forall x \in (a, b)$ and let $x_1, x_2 \in (a, b)$ with $x_1 < x_2$

by the mean value theorem there exists $c \in [x_1, x_2]$ such that $f'(c) = \frac{f(x_2) - f(x_1)}{x_2 - x_1}$

since $f'(c) > 0 \quad \forall c$ and $x_2 - x_1 > 0$, from $*$ it follows that $f(x_2) > f(x_1)$

MACLAURIN FORMULA FOR $\log(1+x)$

The function $f: (-1, \infty) \rightarrow \mathbb{R}$ given by $f(x) = \log(1+x)$ is n times differentiable at each point of its domain, with

$$f^{(n)}(x) = (-1)^{n+1} \frac{(n-1)!}{(1+x)^n} \quad \forall n \geq 1$$

Therefore, Taylor's expansion of order n of f at $x_0 > -1$ is:

$$\log(1+x) = \log(1+x_0) + \sum_{k=1}^n (-1)^{k+1} \frac{(x-x_0)^k}{k(1+x_0)^k} + o((x-x_0)^n)$$

A simple polynomial thus locally approximates the logarithmic function. In particular, the Maclaurin's expansion of order n of f is:

$$\log(1+x) = x - \frac{x^2}{2} + \frac{x^3}{3} - \dots + (-1)^{n+1} \frac{x^n}{n} + o(x^n) = \sum_{k=1}^n (-1)^{k+1} \frac{x^k}{k} + o(x^n)$$

N-VARIABLE DIFFERENTIAL CALCULUS

INTRODUCTION TO PARTIAL DERIVATIVES

consider an interior point x_0 of the domain $A \subseteq \mathbb{R}^2$

$x_0 = (x_{01}, x_{02})$ starting point

$h = (h_1, h_2)$ increment of x (vector in one of infinitely many directions)

$x_0 + h = (x_{01} + h_1, x_{02} + h_2)$ final point

$f(x_0) = f(x_{01}, x_{02})$

$f(x_0 + h) = f(x_{01} + h_1, x_{02} + h_2) - f(x_{01}, x_{02}) = \Delta f$ increment of f

\Rightarrow

we only accept to move in a direction which is \parallel to one of the axes

there are two such directions \rightarrow

① move \parallel to x_1 -axis

- $x_0 = (x_{01}, x_{02})$

- $h = (h, 0)$ increment of x when we move \parallel to x_1 -axis

$\hookrightarrow h$ is still a vector but it is now only determined by one number, its non-null 1 component h

- $x_0 + h = (x_{01} + h, x_{02})$

- $f(x_0) = f(x_{01}, x_{02})$

- $f(x_0 + h) = f(x_{01} + h, x_{02})$

- $f(x_0 + h) - f(x_0) = f(x_{01} + h, x_{02}) - f(x_{01}, x_{02}) = \Delta f$ increment of f when we move \parallel to x_1 -axis

- $\frac{\Delta f}{h} = \frac{f(x_0 + h) - f(x_0)}{h} = \frac{f(x_{01} + h, x_{02}) - f(x_{01}, x_{02})}{h}$

- partial derivative with respect to x_1 (if the increment h is \parallel to x_1 -axis)

$f'_{x_1}(x_0) = \lim_{h \rightarrow 0} \frac{f(x_{01} + h, x_{02}) - f(x_{01}, x_{02})}{h} \quad \exists \text{ finite}$

② move \parallel to x_2 -axis

- $x_0 = (x_{01}, x_{02})$

- $h = (0, h)$

- $x_0 + h = (x_{01}, x_{02} + h)$

- $f(x_0) = f(x_{01}, x_{02})$

- $f(x_0 + h) = f(x_{01}, x_{02} + h)$

- $f(x_0 + h) - f(x_0) = f(x_{01}, x_{02} + h) - f(x_{01}, x_{02}) = \Delta f$ increment of f when we move \parallel to x_2 -axis

- $\frac{\Delta f}{h} = \frac{f(x_0 + h) - f(x_0)}{h} = \frac{f(x_{01}, x_{02} + h) - f(x_{01}, x_{02})}{h}$

- partial derivative with respect to x_2
 $f'_{x_2}(\underline{x}_0) = \lim_{h \rightarrow 0} \frac{f(x_{01}, x_{02} + h) - f(x_{01}, x_{02})}{h} \exists \text{ finite}$

NOTIONS of partial derivatives of the function $f(x_1, x_2)$

$$f'_{x_1}(x_1, x_2), f'_{x_2}(x_1, x_2) \Rightarrow \frac{\partial f}{\partial x_1}(x_1, x_2), \frac{\partial f}{\partial x_2}(x_1, x_2)$$

f'_{x_1} is the derivative of $f(x_1, x_2)$ when only x_1 varies and x_2 remains constant; f'_{x_2} is the derivative of $f(x_1, x_2)$ when only x_2 varies and x_1 remains constant

EXAMPLE

$$f(x_1, x_2) = x_1^2 + x_2^3 \Rightarrow f'_{x_1}(x_1, x_2) = 2x_1, f'_{x_2}(x_1, x_2) = 3x_2^2$$

GRADIENT VECTOR

if $f(x_1, x_2)$ has finite partial derivatives f'_{x_1}, f'_{x_2} at \underline{x}_0 then the vector which contains the two partial derivatives is the gradient vector of f at $\underline{x}_0 \Rightarrow \text{grad } f(\underline{x}_0) = [f'_{x_1}(\underline{x}_0), f'_{x_2}(\underline{x}_0)]$ vector of numbers

with the general point $\underline{x} = (x_1, x_2) \Rightarrow \text{grad } f(x_1, x_2) = [f'_{x_1}(x_1, x_2), f'_{x_2}(x_1, x_2)]$
vector of functions

another notion $\Rightarrow \nabla f(\underline{x}_0) \rightarrow \nabla f(x_1, x_2)$ [$\nabla = \text{nabla}$]

FORMAL DEFINITION

$$f: A \subseteq \mathbb{R}^n \rightarrow \mathbb{R}$$

take an interior point \underline{x}_0 of A : $\underline{x}_0 = (x_{01}, x_{02}, \dots, x_{0n})$

take an increment h for which only one component is $\neq 0$ (the increment only involves one variable x_j): $h = (0, \dots, 0, h, 0, \dots, 0)$

$$\text{then: } \underline{x}_0 + h = (x_{01}, \dots, x_{0j-1}, x_{0j} + h, x_{0j+1}, \dots, x_{0n})$$

then consider:

$$f(\underline{x}_0) = f(x_{01}, \dots, x_{0n})$$

$$f(\underline{x}_0 + h) = f(x_{01}, \dots, x_{0j-1}, x_{0j} + h, x_{0j+1}, \dots, x_{0n})$$

and:

$$f(\underline{x}_0 + h) - f(\underline{x}_0) = f(x_{01}, \dots, x_{0j-1}, x_{0j} + h, x_{0j+1}, \dots, x_{0n}) - f(x_{01}, \dots, x_{0n}) = \Delta f$$

then take the difference quotient:

$$\frac{\Delta f}{h} = \frac{f(x_{01}, \dots, x_{0j-1}, x_{0j} + h, x_{0j+1}, \dots, x_{0n}) - f(x_{01}, \dots, x_{0n})}{h}$$

we can now define the partial derivative of f at \underline{x}_0 with respect to the variable x_j :

$$f'_{x_j}(\underline{x}_0) = \lim_{h \rightarrow 0} \frac{f(x_{01}, \dots, x_{0j-1}, x_{0j}+h, x_{0j+1}, \dots, x_{0n}) - f(x_{01}, \dots, x_{0n})}{h} \quad \exists \text{ finite}$$

$\forall j = 1, 2, \dots, n$

we define: $f'_{x_1}(\underline{x}_0), \dots, f'_{x_n}(\underline{x}_0) \left[\frac{\partial f}{\partial x_1}(\underline{x}_0), \dots, \frac{\partial f}{\partial x_n}(\underline{x}_0) \right]$

we apply it to a general point $\underline{x} = (x_1, \dots, x_n)$ from the function $f(x_1, \dots, x_n)$ we get n functions:

$$f'_{x_1}(x_1, \dots, x_n), \dots, f'_{x_n}(x_1, \dots, x_n) \left[\frac{\partial f}{\partial x_1}(x_1, \dots, x_n), \dots, \frac{\partial f}{\partial x_n}(x_1, \dots, x_n) \right]$$

called the n partial derivatives of the function $f(x_1, \dots, x_n)$

GRADIENT VECTOR

at \underline{x}_0 :

$\text{grad } f(\underline{x}_0) = (f'_{x_1}(\underline{x}_0), \dots, f'_{x_n}(\underline{x}_0))$ vector of numbers

at $\underline{x} = (x_1, \dots, x_n)$:

$\text{grad } f(x_1, \dots, x_n) = (f'_{x_1}(x_1, \dots, x_n), \dots, f'_{x_n}(x_1, \dots, x_n))$ vector of functions

INTRODUCTION TO DIFFERENTIABILITY

$f: A \subseteq \mathbb{R}^n \rightarrow \mathbb{R}$

consider an interior point \underline{x}_0 of A , then:

f is differentiable at \underline{x}_0 \rightarrow only strong notion

$\Downarrow \hat{\neq}$

$\Downarrow \hat{\neq}$

\exists partial derivatives of f at \underline{x}_0 $\not\equiv$ f is continuous at \underline{x}_0

CASE OF $f: A \subseteq \mathbb{R}^2 \rightarrow \mathbb{R}$

consider an interior point \underline{x}_0 of A

f is differentiable at \underline{x}_0 if:

$\forall \underline{x} = \underline{x}_0 + \underline{h} \in \mathcal{U}(\underline{x}_0)$,

$$f(\underline{x}) = \underbrace{f(\underline{x}_0) + \underline{m} \cdot (\underline{x} - \underline{x}_0)}_{\text{approximating plane}} + \underbrace{\theta(\|\underline{x} - \underline{x}_0\|)}_{\text{negligible error}}, \text{ as } \underline{x} \rightarrow \underline{x}_0$$

approximating plane negligible error

more explicitly:

$$f(x_1, x_2) = f(x_{01}, x_{02}) + (m_1, m_2) \begin{pmatrix} x_1 - x_{01} \\ x_2 - x_{02} \end{pmatrix} + \theta(\|\underline{x} - \underline{x}_0\|)$$

$$= \underbrace{f(x_{01}, x_{02}) + m_1(x_1 - x_{01}) + m_2(x_2 - x_{02})}_{\text{plane in 3D-space}} + \underbrace{\theta(\|\underline{x} - \underline{x}_0\|)}_{\text{negligible part}}, \text{ as } \underline{x} \rightarrow \underline{x}_0$$

GENERAL CASE $f: A \subseteq \mathbb{R}^n \rightarrow \mathbb{R}$

consider an interior point x_0 of A

f differentiable at x_0 if:

$$\forall x = x_0 + h \in \mathcal{U}(x_0),$$

$$f(x) = \underbrace{f(x_0) + m \cdot (x - x_0)}_{\text{approximating 1-degree part}} + \underbrace{o(\|x - x_0\|)}_{\text{negligible error}} \text{ as } x \rightarrow x_0$$

approximating 1-degree part negligible error

THEOREM: DIFFERENTIABILITY $\rightarrow \exists$ PARTIAL DERIVATIVES, CONTINUITY

$f: A \subseteq \mathbb{R}^n \rightarrow \mathbb{R}$

consider an interior point x_0 of A

if f is differentiable at x_0 , then:

① f has all partial derivatives

$$f'_{x_1}(x_0), \dots, f'_{x_n}(x_0), \text{ and } \underline{m} = \text{grad } f(x_0) = (f'_{x_1}(x_0), \dots, f'_{x_n}(x_0))$$

[$n=2 \Rightarrow$ the approximating plane is exactly the tangent plane at x_0]

② f is continuous at x_0

\Rightarrow

when f is differentiable at x_0 , then:

$$\forall x = x_0 + h \in \mathcal{U}(x_0),$$

$$f(x) = f(x_0) + \underbrace{\text{grad } f(x_0) \cdot (x - x_0)}_{\text{differentiable } df(x_0)} + \underbrace{o(\|x - x_0\|)}_{\text{negligible part}} \text{ as } x \rightarrow x_0$$

approximating 1-degree part

more explicitly:

$$\begin{aligned} df(x_0) &= \text{grad } f(x_0) \cdot (x - x_0) = [f'_{x_1}(x_0), \dots, f'_{x_n}(x_0)] \begin{bmatrix} x_1 - x_{01} \\ \vdots \\ x_n - x_{0n} \end{bmatrix} \\ &= f'_{x_1}(x_0)(x_1 - x_{01}) + \dots + f'_{x_n}(x_0)(x_n - x_{0n}) \\ &= f'_{x_1}(x_0) dx_1 + \dots + f'_{x_n}(x_0) dx_n \end{aligned}$$

EXAMPLE

given $f(x_1, x_2) = 3x_1^2 + 5x_1x_2^2$, write the differential df at $P(1,1) \Rightarrow$

$$f'_{x_1} = 6x_1 + x_2^2, f'_{x_2} = 10x_1x_2$$

therefore: $f'_{x_1}(1,1) = 11, f'_{x_2}(1,1) = 10$

$$df(1,1) = 11(x_1 - 1) + 10(x_2 - 1) = 11 dx_1 + 10 dx_2$$

THEOREM

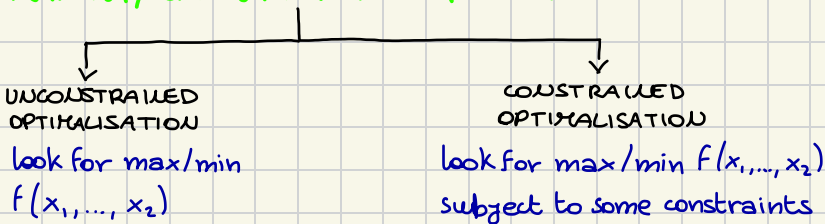
$$F: A \subseteq \mathbb{R}^n \rightarrow \mathbb{R}$$

if A is open and F has continuous partial derivatives on A
then F is differentiable on A
 \Rightarrow

given the set $A \subseteq \mathbb{R}^n$ we will use the notion:

$$C^1(A) = \left\{ \text{functions } F: A \subseteq \mathbb{R}^n \rightarrow \mathbb{R} \text{ that have continuous partial derivatives } \forall x \in A \right\}$$

MAXIMA/MINIMA FOR $F: A \subseteq \mathbb{R}^n \rightarrow \mathbb{R}$



UNCONSTRAINED OPTIMISATION

- ① necessary condition \Rightarrow Fermat's theorem
- ② sufficient condition

consider $F: A \subseteq \mathbb{R}^n \rightarrow \mathbb{R}$, if:

- ① x_0 is an interior point of A
- ② F has all partial derivatives at x_0
- ③ x_0 is a local maximiser/minimiser for F

then:

$$\text{grad } F(x_0) = \mathbf{0}$$

that is:

$$\begin{cases} F'_{x_1}(x_0) = 0 \\ \vdots \\ F'_{x_n}(x_0) = 0 \end{cases} \Rightarrow \begin{cases} \text{if } x_0 \text{ is such that } \text{grad } F(x_0) = \mathbf{0} \\ x_0 \text{ is called a stationary point of } F(x) \end{cases}$$

we CANNOT apply it when

- ① x_0 is a boundary point of A
- ② x_0 is a point at which the partial derivatives do not exist
- ③ Fermat's theorem is not invertible \Rightarrow if at a point x_0 we know that

$\text{grad } f(x_0) = 0$ (the tangent plane is horizontal)

GENERAL PROCEDURE TO FIND UNCONSTRAINED LOCAL MAXIMISERS / MINIMISERS

- ① find exceptional points \rightarrow
 - boundary points
 - points at which $f'_{x_1}, \dots, f'_{x_n}$ do not all \exists
- ② find all other points \rightarrow
 - necessary condition (Fermat's theorem)
 - sufficient condition

it may be that a point which passes the necessary condition gives neither a local maximum nor a local minimum

THEOREM: NECESSARY AND SUFFICIENT MAXIMUM (MINIMUM) CONDITION FOR A CONCAVE (CONVEX) AND DIFFERENTIABLE FUNCTION

$f: A \subseteq \mathbb{R}^n \rightarrow \mathbb{R}$, if:

- ① A is a convex set in \mathbb{R}^n
- ② f is concave (convex) in A
- ③ f is differentiable in A
- ④ x_0 interior point of A

then:

$\text{grad } f(x_0) = 0 \Leftrightarrow x_0$ is a local maximiser (minimiser) $\Leftrightarrow x_0$ is a global maximiser (minimiser)

**theorems and proofs for
n-variable differential
calculus**

FERMAT'S THEOREM IN \mathbb{R}^n

Let $f: A \subseteq \mathbb{R}^n \rightarrow \mathbb{R}$ be defined on a set A in \mathbb{R}^n and C a subset of A

suppose f is differentiable at an interior point \hat{x} of C

if \hat{x} is a local extremal point (maximizer/minimizer) of f on C then $\nabla f(\hat{x}) = 0$

PROOF

if \hat{x} is a local extremum of f on C there exists a neighborhood $U \subseteq C$ of \hat{x} such that $f(\hat{x}) \geq f(x)$ (maximizer) or $f(\hat{x}) \leq f(x)$ (minimizer) for all $x \in U$

by the differentiability of f at \hat{x} , for any direction v , we have:

$$f(\hat{x} + tv) = f(\hat{x}) + t \nabla f(\hat{x}) \cdot v + o(t) \quad \text{where } o(t) \text{ vanishes faster than } t \text{ as } t \rightarrow 0$$

at a local extremum $f(\hat{x} + tv) \leq f(\hat{x})$ (maximizer) or $f(\hat{x} + tv) \geq f(\hat{x})$ (minimizer) for small t

dividing and taking the limit as $t \rightarrow 0$, the directional derivative in any direction v is 0:

$$\nabla f(\hat{x}) \cdot v = 0 \quad \forall v$$

since the gradient is 0 in all directions, $\nabla f(\hat{x}) = 0$

LINEAR ALGEBRA

VECTORS

SUBSPACE = $V \subseteq \mathbb{R}^n, V \neq \emptyset$

V is a vector subspace of \mathbb{R}^n if:

① $\forall \underline{x}, \underline{y} \in V \Rightarrow \underline{x} + \underline{y} \in V$

② $\forall \underline{x} \in V, \forall \alpha \in \mathbb{R} \Rightarrow \alpha \underline{x} \in V$ (V is closed with respect to $\underline{x} + \underline{y}, \alpha \underline{x}$)

THEOREM

$V \subseteq \mathbb{R}^n, V \neq \emptyset$

V is a vector subspace of $\mathbb{R}^n \iff \forall \underline{x}, \underline{y} \in V, \forall \alpha, \beta \in \mathbb{R}, \alpha \underline{x} + \beta \underline{y} \in V$

(V is closed with respect to linear combinations $\alpha \underline{x} + \beta \underline{y}$)

THEOREM: SUBSPACES IN \mathbb{R}^2

$A \subseteq \mathbb{R}^2, A \neq \emptyset$

A is a subspace of $\mathbb{R}^2 \iff$ ① $A = \mathbb{R}^2$ or

② $A = \{ \underline{0} \}$ or

③ A is a straight line passing through $\underline{0}$, that is:

$$A = \left\{ \begin{pmatrix} x \\ y \end{pmatrix} \in \mathbb{R}^2 : ax + by = 0 \right\}$$

THEOREM: SUBSPACES IN \mathbb{R}^n

$A \subseteq \mathbb{R}^n, A \neq \emptyset$

A is a subspace of $\mathbb{R}^n \iff$ ① $A = \mathbb{R}^n$ or

② $A = \{ \underline{0} \}$ or

③ $A = \left\{ \begin{pmatrix} x_1 \\ \vdots \\ x_n \end{pmatrix} \in \mathbb{R}^n : \text{all the relations involving the components } x_1, \dots, x_n \text{ can be written as 1-degree equalities with no constants } \neq 0 \right\}$

necessary condition = if A is a subspace of $\mathbb{R}^n \rightarrow A$ must contain the origin $\underline{0}$

THEOREM: INTERSECTION OF SUBSPACES

the intersection of any collection of vector subspaces of \mathbb{R}^n is a vector subspace of \mathbb{R}^n

PROOF

suppose $\{V_i\}$ is a collection of vector subspaces of \mathbb{R}^n

since $\underline{0} \in V_i \forall i, \underline{0} \in \bigcap V_i$; therefore $\bigcap V_i \neq \emptyset$

consider $\underline{x}, \underline{y} \in \bigcap V_i$ and $\alpha, \beta \in \mathbb{R}$

since $\underline{x}, \underline{y} \in \bigcap V_i$, we have that $\underline{x}, \underline{y} \in V_i \forall i$

since every V_i is a vector subspace of \mathbb{R}^n

this implies that $\alpha \underline{x} + \beta \underline{y} \in V_i \forall i$

therefore $\alpha \underline{x} + \beta \underline{y} \in \bigcap V_i$, so $\bigcap V_i$ is a vector subspace of \mathbb{R}^n

LINEARLY DEPENDENT / INDEPENDENT VECTORS

= \downarrow

\downarrow =

$\underline{x}_1, \dots, \underline{x}_m \in \mathbb{R}^n$ are linearly dependent

if they are not linearly independent

that is if $\exists \alpha_1, \dots, \alpha_m \in \mathbb{R}$ such that:

$$\alpha_1 \underline{x}_1 + \dots + \alpha_m \underline{x}_m = \underline{0} \text{ and } \alpha_1, \dots, \alpha_m$$

are not all equal to zero

$\underline{x}_1, \dots, \underline{x}_m \in \mathbb{R}^n$ are linearly independent

if whenever $\alpha_1 \underline{x}_1 + \dots + \alpha_m \underline{x}_m = \underline{0} \Rightarrow$

$$\alpha_1 = \dots = \alpha_m = 0$$

- if $\underline{x}_1, \dots, \underline{x}_m$ contains $\underline{0}$, then $\underline{x}_1, \dots, \underline{x}_m$ are linearly dependent \Rightarrow

$$0 \underline{x}_1 + \dots + 0 \underline{x}_{m-1} + k \underline{0} = \underline{0} \text{ also when } k \neq 0$$

- one single vector $\underline{x} (\forall m \geq 1) \rightarrow$

• $\underline{x} \neq \underline{0}$ = linearly independent

• $\underline{x} = \underline{0}$ = linearly dependent

THEOREM ($m \geq 2$)

$\underline{x}_1, \dots, \underline{x}_m \in \mathbb{R}^n$, with $m \geq 2$, then:

① $\underline{x}_1, \dots, \underline{x}_m$ linearly independent = none of them can be written as a linear combination of the others

② $\underline{x}_1, \dots, \underline{x}_m$ linearly dependent = at least one of them can be written as a linear combination of the others

($m = 2$)

① linearly independent = if none of the vectors is a multiple of the other

② linearly dependent = at least one of them is a multiple of the other

FUNDAMENTAL VECTORS OF $\mathbb{R}^3 \Rightarrow$ linearly independent

$$\underline{e}_1 = \begin{pmatrix} 1 \\ 0 \\ 0 \end{pmatrix}, \underline{e}_2 = \begin{pmatrix} 0 \\ 1 \\ 0 \end{pmatrix}, \underline{e}_3 = \begin{pmatrix} 0 \\ 0 \\ 1 \end{pmatrix} \in \mathbb{R}^3$$

FUNDAMENTAL VECTORS OF $\mathbb{R}^n \Rightarrow$ linearly independent

$$e_1 = \begin{pmatrix} 1 \\ 0 \\ \vdots \\ 0 \end{pmatrix}, e_2 = \begin{pmatrix} 0 \\ 1 \\ \vdots \\ 0 \end{pmatrix}, \dots, e_n = \begin{pmatrix} 0 \\ \vdots \\ 1 \end{pmatrix} \in \mathbb{R}^n$$

remark

$x_1, \dots, x_m \in \mathbb{R}^n$, then:

- ① if $m > n$ they are linearly dependent
- ② if $m \leq n$ we don't know (depends on the vectors)

SPAN S

- consider a set S of vectors in \mathbb{R}^n
- $\text{span } S =$ intersection $\cap V_i$ of all vector subspaces V_i of \mathbb{R}^n which contains S
- $\text{span } S = \cap V_i$ is the smallest vector subspace of \mathbb{R}^n which contains S
- it is the vector subspace which is spanned generated by the set $S \rightarrow$ smallest enlargement of S with the property of being a vector space

remark

take $x_1, \dots, x_m \in \mathbb{R}^n$, then:

- ① if $m < n$ it is not possible that they span all \mathbb{R}^n
- ② if $m \geq n$ we don't know (depends on the vectors)

BASIS OF \mathbb{R}^n

take $x_1, \dots, x_m \in \mathbb{R}^n$

they are a basis of \mathbb{R}^n if:

- ① they are linearly independent (none of them is superfluous)
- ② they span all \mathbb{R}^n (enough to span \mathbb{R}^n)

if we call $S = \{x_1, \dots, x_m\}$, $\text{span } S = \mathbb{R}^n$

remark (i)

- ① $m > n \Rightarrow$ they can't be linearly independent so they are not a basis of \mathbb{R}^n
- ② $m < n \Rightarrow$ they can't span all \mathbb{R}^n , so they are not a basis of \mathbb{R}^n
- the number of vectors in a basis of \mathbb{R}^n must be $m = n$

remark (ii)

- suppose to take exactly n vectors in \mathbb{R}^n
- to check they are a basis of \mathbb{R}^n , two facts must be checked \rightarrow

① they are linearly dependent

② they span all \mathbb{R}^n

- this property [$x_1, \dots, x_n \in \mathbb{R}^n$ are linearly independent \Leftrightarrow they span all \mathbb{R}^n] holds \rightarrow if the vectors are the correct number only the linear independence needs to be checked

- ex. the fundamental basis of $\mathbb{R}^n \rightarrow$

$$e_1 = \begin{pmatrix} 1 \\ 0 \\ \vdots \\ 0 \end{pmatrix}, e_2 = \begin{pmatrix} 0 \\ 1 \\ \vdots \\ 0 \end{pmatrix}, \dots, e_n = \begin{pmatrix} 0 \\ \vdots \\ 0 \\ 1 \end{pmatrix} \in \mathbb{R}^n$$

the coefficients such that $x = \begin{pmatrix} x_1 \\ \vdots \\ x_n \end{pmatrix}$ can be written as a linear combination $x = \alpha_1 e_1 + \dots + \alpha_n e_n$ are exactly the components x_1, \dots, x_n

Dimension of \mathbb{R}^n

- the dimension of \mathbb{R}^n is n

- it gives the number of vectors in any basis of \mathbb{R}^n and it corresponds to the geometrical notion of dimension

remark

- any two vectors $e_j, e_k \in \mathbb{R}^n$ with $j \neq k$ are such that $e_j \perp e_k$ ($e_j \cdot e_k = 0 \Rightarrow$ they are orthogonal)

- all vectors $e_j \in \mathbb{R}^n$ are such that $\|e_j\| = 1$ ($e_j \cdot e_j = 1 \Rightarrow$ unitary norm)

- the two properties are collected together and conclude that any two vectors $e_j, e_k \in \mathbb{R}^n$ are orthonormal

- the fundamental basis e_1, \dots, e_n is an orthonormal basis

- in those situations in which we can't use exactly the fundamental basis we will try to use a basis constituted by orthogonal vectors or by orthonormal vectors (better)

THEOREM: ORTHOGONAL VECTORS ARE LINEARLY INDEPENDENT

consider k vectors $x_1, \dots, x_k \in \mathbb{R}^n$ with $2 \leq k \leq n$

if these vectors are z -by- z orthogonal and they do not include the null vector (if they are orthonormal), then they are linearly independent

THEOREM

consider an orthonormal basis x_1, \dots, x_n in \mathbb{R}^n

then $\forall v \in \mathbb{R}^n$ it is:

$$v = \alpha_1 x_1 + \dots + \alpha_n x_n \text{ with } \alpha_1 = v \cdot x_1, \dots, \alpha_n = v \cdot x_n$$

if $\underline{x}_1, \dots, \underline{x}_n$ is an orthonormal basis of \mathbb{R}^n there is a quick way to write the coefficients $\alpha_1, \dots, \alpha_n$ (exactly as it is for the basis $\underline{e}_1, \dots, \underline{e}_n$)

$$S = \{ \underline{x}_1, \dots, \underline{x}_m \}, \text{ SPAN } S \subset \mathbb{R}^n$$

\Rightarrow span S is a subspace of \mathbb{R}^n smaller than \mathbb{R}^n

THEOREM

suppose V is a subspace of \mathbb{R}^n , then \exists a set $S = \{ \underline{x}_1, \dots, \underline{x}_m \}$ such that:

$$V = \text{span } S$$

\Rightarrow every subspace of \mathbb{R}^n can be seen as the span S of some finite set S

BASIS OF A SUBSPACE

take a subspace $V \subseteq \mathbb{R}^n$

take $\underline{x}_1, \dots, \underline{x}_m \in V$

they are a basis of V if:

① they are linearly independent

② they span all V

\Rightarrow if we call $S = \{ \underline{x}_1, \dots, \underline{x}_m \}$, $\text{span } S = V$

THEOREM: PROPERTY OF UNIQUE WRITING

take a vector subspace $V \subseteq \mathbb{R}^n$

take $\underline{x}_1, \dots, \underline{x}_m \in V$

they are a basis of $V \Leftrightarrow \forall \underline{z} \in V$, \underline{z} can be written in a unique way as a linear combination of $\underline{x}_1, \dots, \underline{x}_m$

THEOREM

take a vector subspace $V \subseteq \mathbb{R}^n$

any two bases of V will contain the same number k of vectors

it will be:

$$k = n \Leftrightarrow V = \mathbb{R}^n$$

$$k < n \Leftrightarrow V \subset \mathbb{R}^n$$

DEFINITION OF A SUBSPACE

- take a vector subspace $V \subseteq \mathbb{R}^n$

- dimension of V = the number of vectors in any basis of V

- applied to $V = \mathbb{R}^n \Rightarrow \dim \mathbb{R}^n = n$

- applied to $V \subset \mathbb{R}^n \Rightarrow \dim V = k < n$

remark (i)

- consider a subspace $V \subseteq \mathbb{R}^n$ with dimension $k \leq n$
- take k vectors $x_1, \dots, x_k \in V$ (same number as $\dim V$)
- in order to guarantee they are a basis of V only one of the two properties needs to be checked
- x_1, \dots, x_k are linearly independent \Leftrightarrow they span all the space V

remark (ii)

if $V = \{0\}$, then:

- ① basis of V : empty set \emptyset
- ② dimension of V : 0

THEOREM: SUBSPACES IN \mathbb{R}^2

take $A \subseteq \mathbb{R}^2$, $A \neq \emptyset$

- A is a subspace of $\mathbb{R}^2 \Leftrightarrow$
- ① $A = \mathbb{R}^2$ $\dim A = 2$ or
 - ② $A = \{0\}$ $\dim A = 0$ or
 - ③ A is a straight line passing through 0 :
 $A = \left\{ \begin{pmatrix} x \\ y \end{pmatrix} \in \mathbb{R}^2 : ax + by = 0 \right\}$ $\dim A = 1$

THEOREM: SUBSPACES IN \mathbb{R}^3

take $A \subseteq \mathbb{R}^3$, $A \neq \emptyset$

- A is a subspace of $\mathbb{R}^3 \Leftrightarrow$
- ① $A = \mathbb{R}^3$ $\dim A = 3$ or
 - ② $A = \{0\}$ $\dim A = 0$ or
 - ③ A is a plane passing through 0 : \nearrow basis has 2 vectors
 $A = \left\{ \begin{pmatrix} x \\ y \\ z \end{pmatrix} : ax + by + cz = 0 \right\}$ $\dim A = 2$ or
 - ④ A is a straight line passing through 0 :
 $A = \left\{ \begin{pmatrix} x \\ y \\ z \end{pmatrix} \in \mathbb{R}^3 : a_1x + b_1y + c_1z = 0, a_2x + b_2y + c_2z = 0 \right\}$
 $\dim A = 1 \rightarrow$ basis has 1 vector

MATRICES

$$A = \begin{bmatrix} 2 & -3 & 4 \\ 1 & 0 & 1 \end{bmatrix} \rightarrow \text{row} \quad \Rightarrow 2 \text{ rows} + 3 \text{ columns} = 2 \times 3 \text{ matrix}$$

\downarrow
column

DEF. table containing numbers organised in rows and columns

more generally:

$$A = \begin{bmatrix} a_{11} & \dots & a_{1n} \\ \vdots & & \vdots \\ a_{m1} & \dots & a_{mn} \end{bmatrix}$$

$$A = [a_{ij}] \quad \begin{matrix} i = 1, \dots, m \\ j = 1, \dots, n \end{matrix} \Rightarrow \text{its elements have a double index which corresponds to double orders (by row / by column)}$$

also:

$$A = \begin{bmatrix} 2 & -3 & 1 \\ 4 & 5 & 0 \end{bmatrix} \Rightarrow \text{collection of column vectors}$$

$$A = \begin{bmatrix} 2 & -3 & 1 \\ 4 & 5 & 0 \end{bmatrix} \Rightarrow \text{collection of row vectors}$$

if $m = n$:

$$A = \begin{bmatrix} 3 & 1 & 2 \\ -1 & 2 & 0 \\ 1 & 3 & 5 \end{bmatrix} \Rightarrow \text{square matrix of order } n$$

\hookrightarrow main diagonal

SETS OF MATRICES

$$\mathcal{X}(m, n) = \{ \text{all } m \times n \text{ matrices} \}$$

$$\mathcal{X}(n) = \{ \text{all square matrices of order } n \}$$

OPERATIONS W $\mathcal{X}(m, n)$

① $A + B \Rightarrow$

$$A = \begin{bmatrix} 2 & -3 & 1 \\ 0 & 5 & 2 \end{bmatrix}, \quad B = \begin{bmatrix} 5 & -1 & 2 \\ 0 & 1 & 4 \end{bmatrix} \Rightarrow A + B = \begin{bmatrix} 7 & -4 & 3 \\ 0 & 6 & 6 \end{bmatrix}$$

DEF. $\forall A = [a_{ij}], B = [b_{ij}] \in \mathcal{R}(m, n)$, $C = A + B \in \mathcal{R}(m, n)$ and is given by:
 $c_{ij} = a_{ij} + b_{ij}$, $\forall i = 1, \dots, m; \forall j = 1, \dots, n$

PROPER.

- commutative = $\forall A, B \in \mathcal{R}(m, n)$, $A + B = B + A$
- associative = $\forall A, B, C \in \mathcal{R}(m, n)$, $(A + B) + C = A + (B + C) = A + B + C$
- \exists null matrix = $\exists 0 \in \mathcal{R}(m, n)$, $0 = \begin{pmatrix} 0 & \dots & 0 \\ \vdots & & \vdots \\ 0 & \dots & 0 \end{pmatrix}$, such that $\forall A \in \mathcal{R}(m, n)$,
 $A + 0 = A$
- \forall matrix \exists opposite = $\forall A = [a_{ij}] \in \mathcal{R}(m, n) \exists B = -A = [-a_{ij}] \in \mathcal{R}(m, n)$
such that $A + (-A) = 0$

② $\alpha A \Rightarrow$

$$A = \begin{bmatrix} 3 & -1 \\ 4 & 0 \\ 2 & 5 \end{bmatrix}, \alpha = 2 \Rightarrow \alpha A = \begin{bmatrix} 6 & -2 \\ 8 & 0 \\ 4 & 10 \end{bmatrix}$$

DEF. $\forall A = [a_{ij}] \in \mathcal{R}(m, n)$, $\forall \alpha \in \mathcal{R}$, $B = \alpha A \in \mathcal{R}(m, n)$ and is given by
 $b_{ij} = \alpha a_{ij}$, $\forall i = 1, \dots, m; \forall j = 1, \dots, n$

PROPER.

- distributive = $\forall \alpha, \beta \in \mathcal{R}$, $\forall A \in \mathcal{R}(m, n)$, $(\alpha + \beta)A = \alpha A + \beta A$
- distributive = $\forall \alpha \in \mathcal{R}$, $\forall A, B \in \mathcal{R}(m, n)$, $\alpha(A + B) = \alpha A + \alpha B$
- associative = $\forall \alpha, \beta \in \mathcal{R}$, $\forall A \in \mathcal{R}(m, n)$, $(\alpha \beta)A = \alpha(\beta A) = \alpha \beta A$
- number 1 is a unit for matrices = $\forall A \in \mathcal{R}(m, n)$, $1A = A$

③ $A \cdot B$ (row-column product) \Rightarrow

DEF. $\forall A \in \mathcal{R}(m, n)$, $\forall B \in \mathcal{R}(n, p)$, $C = A \cdot B \in \mathcal{R}(m, p)$ and is given by:

$$c_{ik} = (i\text{-th row of } A) \cdot (k\text{-th column of } B)$$

$$= (a_{i1} \ a_{i2} \ \dots \ a_{in}) \cdot \begin{pmatrix} b_{1k} \\ b_{2k} \\ \vdots \\ b_{nk} \end{pmatrix}$$

$$= a_{i1}b_{1k} + a_{i2}b_{2k} + \dots + a_{in}b_{nk}$$

$$= \sum_{j=1}^n a_{ij}b_{jk}, \forall i = 1, \dots, m; \forall k = 1, \dots, p$$

DEFECTS

- $A \cdot B$ is not an internal operation =
 $A \in \mathcal{K}(m, n), B \in \mathcal{K}(n, p) \Rightarrow C = A \cdot B \in \mathcal{K}(m, p)$ [but it becomes internal with square matrices $\rightarrow A \in \mathcal{K}(n), B \in \mathcal{K}(n) \Rightarrow C = A \cdot B \in \mathcal{K}(n)$]
- $A \cdot B$ is not commutative =
 - you can define $A \cdot B$ but not $B \cdot A$
 - you can define both but they are matrices of different kinds
 - you can define both and they are both of the same kind (only square matrices) but still usually $AB \neq BA$

PROPER.

- associative = $\forall A \in \mathcal{K}(m, n), \forall B \in \mathcal{K}(n, p), \forall C \in \mathcal{K}(p, q), (A \cdot B) \cdot C = A \cdot (B \cdot C)$
 $= A \cdot B \cdot C$
- associative = $\forall A \in \mathcal{K}(m, n), \forall B \in \mathcal{K}(n, p), \forall \alpha \in \mathcal{R}, (\alpha A) \cdot B = \alpha (A \cdot B) = \alpha AB$
- distributive = $\forall A \in \mathcal{K}(m, n), \forall B, C \in \mathcal{K}(n, p), A(B+C) = AB+AC$
- \exists 1-sided identity matrices = $\forall m, n \exists$ 1-sided identity matrices:
 $I_m = \begin{pmatrix} 1 & & 0 \\ & \ddots & \\ 0 & & 1 \end{pmatrix} \in \mathcal{K}(m),$
 $I_n = \begin{pmatrix} 1 & & 0 \\ & \ddots & \\ 0 & & 1 \end{pmatrix} \in \mathcal{K}(n),$
such that $\forall A \in \mathcal{K}(m, n)$:
 $I_m A = A \in \mathcal{K}(m, n)$
 $A I_n = A \in \mathcal{K}(m, n)$
- \exists identity matrix = $\forall n \exists$ identity matrix of order $n, I_n = \begin{bmatrix} 1 & & 0 \\ & \ddots & \\ 0 & & 1 \end{bmatrix} \in \mathcal{K}(n),$
such that $\forall A \in \mathcal{K}(n), A I_n = I_n A = A$

SQUARE MATRICES

$$A^2 = A \cdot A \rightarrow A^n = A \dots A \text{] } \rightarrow n \text{ times, } A^1 = A, A^0 = I_n \text{ [} a^0 = 1 \text{]}$$

④ A^T A transpose (defined \forall matrix) \Rightarrow

$$A = \begin{pmatrix} 3 & -1 & 2 \\ 4 & 7 & 0 \end{pmatrix} \rightarrow A^T = \begin{pmatrix} 3 & 4 \\ -1 & 7 \\ 2 & 0 \end{pmatrix}$$

PROPER.

- $(A^T)^T = A$
- $(A+B)^T = A^T + B^T$
- $(\alpha A)^T = \alpha A^T$
- $(AB)^T = B^T A^T$
- if A is a square matrix it is symmetric if: $a_{ij} = a_{ji} \forall i, j$, or if $A^T = A$

⑤ PARTICULAR SQUARE MATRICES \Rightarrow

- upper triangular = $A = \begin{bmatrix} 3 & -1 & 2 \\ 0 & 4 & -1 \\ 0 & 0 & 2 \end{bmatrix} \quad \forall i > j, a_{ij} = 0$

- lower triangular = $A = \begin{bmatrix} 3 & 0 & 0 \\ -1 & 2 & 0 \\ 4 & 1 & 0 \end{bmatrix} \quad \forall i < j, a_{ij} = 0$

- diagonal = $A = \begin{bmatrix} 3 & 0 & 0 \\ 0 & -1 & 0 \\ 0 & 0 & 0 \end{bmatrix} \quad \forall i \neq j, a_{ij} = 0$

- scalar = $A = \begin{bmatrix} 3 & 0 & 0 \\ 0 & 3 & 0 \\ 0 & 0 & 3 \end{bmatrix} \quad \text{diagonal} + a_{ii} = a$

VECTOR SPACES: EXAMPLES

- ① the set $P[x]$ of all polynomials in one variable x (operations: $p(x) + q(x), \alpha p(x)$)
- ② the set S of all convergent sequences $a_n: \mathbb{N} \rightarrow \mathbb{R}$ (operations: $a_n + b_n, \alpha a_n$)
- ③ the set $C(I) / C^1(I) / \dots / C^n(I)$ of all continuous/differentiable with continuous derivative / ... / n -times differentiable with continuous n^{th} derivative functions on an interval $I \subseteq \mathbb{R}$ (operations: $f + g, \alpha f$)
- ④ the set $C(A) / C^1(A)$ of all continuous/differentiable with continuous partial derivatives functions on a convex set $A \subseteq \mathbb{R}^n$ (operations: $f + g, \alpha f$)

LINEAR FUNCTIONS

a function $f: \mathbb{R}^n \rightarrow \mathbb{R}$ is said to be a linear function if:

- ① $\forall \underline{x}, \underline{y} \in \mathbb{R}^n, f(\underline{x} + \underline{y}) = f(\underline{x}) + f(\underline{y})$ [additivity]
- ② $\forall \underline{x} \in \mathbb{R}^n, \forall \alpha \in \mathbb{R}, f(\alpha \underline{x}) = \alpha f(\underline{x})$ [homogeneity]

THEOREM

a function $f: \mathbb{R}^n \rightarrow \mathbb{R}$ is a linear function $\Leftrightarrow \forall \underline{x}, \underline{y} \in \mathbb{R}^n, \forall \alpha, \beta \in \mathbb{R},$
 $f(\alpha \underline{x} + \beta \underline{y}) = \alpha f(\underline{x}) + \beta f(\underline{y})$

THEOREM

if f is a linear function $f: \mathbb{R}^n \rightarrow \mathbb{R}$, then:

- ① $f(\underline{0}) = 0$
- ② $\forall \underline{x}_1, \dots, \underline{x}_m \in \mathbb{R}^n, \forall \alpha_1, \dots, \alpha_m \in \mathbb{R}, f(\alpha_1 \underline{x}_1 + \dots + \alpha_m \underline{x}_m) = \alpha_1 f(\underline{x}_1) + \dots + \alpha_m f(\underline{x}_m)$

② f strictly increasing (strictly positive) $\Leftrightarrow a$ strongly positive

FUNCTIONS $f: \mathbb{R}^n \rightarrow \mathbb{R}^m$

ex. $f: \underline{x} = \begin{pmatrix} x_1 \\ x_2 \\ x_3 \end{pmatrix} \in \mathbb{R}^3 \rightarrow f(\underline{x}) = \begin{pmatrix} x_1 + 3x_3 \\ x_1 + x_2 + x_3 \end{pmatrix} \in \mathbb{R}^2$

a function $f: \mathbb{R}^n \rightarrow \mathbb{R}^m$ is said to be a linear function (linear operator) if:

- ① $\forall \underline{x}, \underline{y} \in \mathbb{R}^n, f(\underline{x} + \underline{y}) = f(\underline{x}) + f(\underline{y}) \Rightarrow$ additivity
- ② $\forall \underline{x} \in \mathbb{R}^n, \forall \alpha \in \mathbb{R}, f(\alpha \underline{x}) = \alpha f(\underline{x}) \Rightarrow$ homogeneity

THEOREM

a function $f: \mathbb{R}^n \rightarrow \mathbb{R}^m$ is a linear function $\Leftrightarrow \forall \underline{x}, \underline{y} \in \mathbb{R}^n, \forall \alpha, \beta \in \mathbb{R},$
 $f(\alpha \underline{x} + \beta \underline{y}) = \alpha f(\underline{x}) + \beta f(\underline{y})$

THEOREM

if f is a linear function $f: \mathbb{R}^n \rightarrow \mathbb{R}^m$, then:

- ① $f(\underline{0}) = \underline{0}$
- ② $\forall \underline{x}_1, \dots, \underline{x}_m \in \mathbb{R}^n, \forall \alpha_1, \dots, \alpha_m \in \mathbb{R}, f(\alpha_1 \underline{x}_1 + \dots + \alpha_m \underline{x}_m) = \alpha_1 f(\underline{x}_1) + \dots + \alpha_m f(\underline{x}_m)$

THEOREM: CONTINUITY OF LINEAR OPERATIONS

every linear operator $f: \mathbb{R}^n \rightarrow \mathbb{R}^m$ is continuous

that is: $\lim_{\underline{x} \rightarrow \underline{x}_0} f(\underline{x}) = f(\underline{x}_0), \forall \underline{x}_0 \in \mathbb{R}^n$

IMAGE, KERNEL

take a linear function $f: \mathbb{R}^n \rightarrow \mathbb{R}^m$

then we can always associate to f two different sets:

- ① image / image space $\Rightarrow f(\mathbb{R}^n) = \text{Im} f$
 $= \{ \underline{y} \in \mathbb{R}^m: \exists \underline{x} \in \mathbb{R}^n \text{ with } \underline{y} = f(\underline{x}) \} \subseteq \mathbb{R}^m$
- ② kernel $\Rightarrow \ker f = \{ \underline{x} \in \mathbb{R}^n: f(\underline{x}) = \underline{0} \} \subseteq \mathbb{R}^n$

DEFINITION \Rightarrow consider $f: \mathbb{R}^n \rightarrow \mathbb{R}^m$ linear, then: $\text{rank } f = \dim \text{Im} f$

THEOREM

$\ker f$ is always a vector subspace of \mathbb{R}^n

DEFINITION \Rightarrow consider $f: \mathbb{R}^n \rightarrow \mathbb{R}^m$ linear, then: nullity of $f = \dim \ker f$

THEOREM: RANK AND NULLITY

take $f: \mathbb{R}^n \rightarrow \mathbb{R}^m$ linear, then: $\dim \ker f = n - \dim \operatorname{Im} f$

THEOREM: SURJECTIVITY CONDITIONS

take $f: \mathbb{R}^n \rightarrow \mathbb{R}^m$ linear, then: f surjective $\Leftrightarrow \operatorname{Im} f = \mathbb{R}^m \Leftrightarrow \operatorname{rank} f = \dim \operatorname{Im} f = m$

THEOREM: INJECTIVITY CONDITIONS

take $f: \mathbb{R}^n \rightarrow \mathbb{R}^m$ linear, then: f injective $\Leftrightarrow \ker f = \{0\} \Leftrightarrow$ nullity of $f = \dim \ker f = 0 \Leftrightarrow \operatorname{rank} f = \dim \operatorname{Im} f = n$

THEOREM: BIJECTIVITY CONDITIONS

take $f: \mathbb{R}^n \rightarrow \mathbb{R}^m$ linear, then: f bijective \Leftrightarrow

- ① $m = n$
- ② $\operatorname{rank} f = m = n$

DETERMINANT OF A SQUARE MATRIX

particular cases \rightarrow

① A has order 1 $\Rightarrow A = (a) \Rightarrow \det A = a$

② A has order 2 $\Rightarrow A = \begin{pmatrix} a_{11} & a_{12} \\ a_{21} & a_{22} \end{pmatrix}$

ex. ① $A = \begin{pmatrix} 2 & 4 \\ 1 & 5 \end{pmatrix} \rightarrow \det A = 2 \cdot 5 - 4 \cdot 1 = 6 \neq 0 \Rightarrow$ linearly independent

② $A = \begin{pmatrix} 2 & 6 \\ 1 & 3 \end{pmatrix} \rightarrow \det A = 2 \cdot 3 - 6 \cdot 1 = 0 \Rightarrow$ linearly dependent

general case $\Rightarrow \det A = a_{11} \cdot a_{22} - a_{12} \cdot a_{21}$

③ A has order 3 $\Rightarrow A = \begin{pmatrix} a_{11} & a_{12} & a_{13} \\ a_{21} & a_{22} & a_{23} \\ a_{31} & a_{32} & a_{33} \end{pmatrix}$

ex. $A = \begin{pmatrix} 2 & 3 & 5 \\ -1 & 4 & -1 \\ 0 & 1 & -1 \end{pmatrix} \rightarrow \det A = 2 \cdot 4 \cdot (-1) + 3 \cdot (-1) \cdot 0 + 5 \cdot (-1) \cdot 1 = -14 \neq 0$

$L > \begin{pmatrix} 2 & 3 & 5 \\ -1 & 4 & -1 \\ 0 & 1 & -1 \end{pmatrix} \begin{matrix} 2 & 3 \\ -1 & 4 \\ 0 & 1 \end{matrix}$

SUBMATRICES

① submatrix of a matrix = given a matrix $A \in \mathbb{R}(m, n)$ we call submatrix of A any matrix which is obtained from A canceling out entire rows and/or columns

ex. $A = \begin{pmatrix} 2 & -1 & 5 \\ 3 & 2 & 4 \end{pmatrix} \checkmark \quad A = \begin{pmatrix} 2 & -1 & 5 \\ 3 & 2 & 4 \end{pmatrix} \times$

② complement submatrix = given a square matrix: $A = \begin{pmatrix} a_{11} & a_{12} & \dots & a_{1n} \\ \vdots & \vdots & & \vdots \\ a_{n1} & a_{n2} & \dots & a_{nn} \end{pmatrix}$ we call complement submatrix of the element a_{ij} the square submatrix of A which we obtain by canceling out the row i and the column j on which the element a_{ij} stands and we denote it as A_{ij}

LAPLACE'S DEFINITION

take a square matrix A of order n , then:

① if $n=1$, that is $A = (a)$, then $\det A = a$

② if $n > 1$ \rightarrow
ex. $A = \begin{pmatrix} 3 & 0 & 0 & -2 \\ 5 & 1 & 2 & 4 \\ 2 & -1 & 3 & 2 \\ 1 & 2 & 0 & 5 \end{pmatrix}$ we concentrate on the first row and define:

$$\det A = a_{11}(-1)^{1+1} \det A_{11} + a_{12}(-1)^{1+2} \det A_{12} + a_{13}(-1)^{1+3} \det A_{13} + a_{14}(-1)^{1+4} \det A_{14} + \dots + a_{1n}(-1)^{1+n} \det A_{1n}$$

DEF. MINOR = consider a matrix A ; consider a square submatrix B of A ; then $\det B$ is called a minor of the matrix A

DEF. COMPLEMENT MINOR = consider a square matrix A ; we call complement minor of the element a_{ij} the determinant $\det A_{ij}$ of the complement submatrix A_{ij} of the element a_{ij}

$$A = \begin{pmatrix} a_{11} & a_{12} & \dots & a_{1n} \\ \vdots & \vdots & & \vdots \\ a_{n1} & a_{n2} & \dots & a_{nn} \end{pmatrix}$$

DEF. COFACTOR (algebraic complement) = $(-1)^{i+j} \det A_{ij} \Rightarrow$ the complement minor $\det A_{ij}$ is multiplied by $+1/-1$ according to whether the sum of the indices $i+j$ is even/odd

\Rightarrow

① $n=1$: $A = (a)$, $\det A = a$

② $n > 1$: $\det A = \sum_{j=1}^n a_{1j} (-1)^{1+j} \det A_{1j} / \det A = a_{11} (\text{cofactor of } a_{11}) + a_{12} (\text{cofactor of } a_{12}) + \dots + a_{1n} (\text{cofactor of } a_{1n})$

$$\text{ex. } A = \begin{pmatrix} 3 & 0 & 0 & -2 \\ 5 & 1 & 2 & 4 \\ 2 & -1 & 3 & 2 \\ 1 & 2 & 0 & 5 \end{pmatrix} \rightarrow \det A = 3 \cdot \det \begin{pmatrix} 1 & 2 & 4 \\ -1 & 3 & 2 \\ 2 & 0 & 5 \end{pmatrix} - 2(-1) \det \begin{pmatrix} 5 & 1 & 2 \\ 2 & -1 & 3 \\ 1 & 2 & 0 \end{pmatrix} =$$

$$= 3 \cdot 9 + 2 \cdot (-17) = -7 \neq 0$$

FIRST LAPLACE'S THEOREM

$$\textcircled{1} \forall i, \sum_{j=1}^n a_{ij} \det A_{ij} (-1)^{i+j} = \det A \quad [\text{we can use any row } i]$$

$$\textcircled{2} \forall j, \sum_{i=1}^n a_{ij} \det A_{ij} (-1)^{i+j} = \det A \quad [\text{we can use any column } j]$$

\Rightarrow in exercises use the one with more zeros in it

THEOREM

consider $x_1, \dots, x_n \in \mathbb{R}^n$; build the square matrix with x_1, \dots, x_n as columns

$A = (x_1, \dots, x_n)$, then:

- x_1, \dots, x_n are linearly independent $\Leftrightarrow \det A \neq 0$
- x_1, \dots, x_n are linearly dependent $\Leftrightarrow \det A = 0$

BEHAVIOUR OF $\det A$ WITH RESPECT TO OPERATIONS

- ① $\det(A^T) = \det A$
- ② $\det(A \cdot B) = \det A \cdot \det B$
- ③ $\det(A+B) \neq \det A + \det B$
- ④ (if A has order n) $\det(\alpha A) = \alpha^n \det A$

INVERSE OF A SQUARE MATRIX

consider a square matrix of order n

we say that:

A is invertible or A has an inverse B

if \exists a square matrix B of order n such that

$$AB = BA = I_n$$

THEOREM: EXISTENCE AND UNIQUENESS

consider a square matrix A of order n

- ① A has an inverse $B \Leftrightarrow \det A \neq 0$
 - ② if A has an inverse B then B is unique
- $\Rightarrow B$ inverse of A ($B = A^{-1}$)

ADJOINT MATRIX

take a square matrix A

its adjoint matrix $\text{adj } A$ is the matrix we obtain with this procedure:

$$A \rightarrow A^* = \text{matrix of the cofactors of } A \rightarrow \text{adj } A = (A^*)^T$$

THEOREM

suppose A is a square of order n

suppose $\det A \neq 0$, then:

$$A^{-1} = \frac{1}{\det A} \text{adj } A$$

PROPERTIES

- ① $\forall A$ square, with $\det A \neq 0 \rightarrow (A^{-1})^{-1} = A$
- ② $\forall A$ square, with $\det A \neq 0 \rightarrow (A^T)^{-1} = (A^{-1})^T$
- ③ $\forall A, B$ square, with $\det A \neq 0, \det B \neq 0 \rightarrow (AB)^{-1} = B^{-1}A^{-1}$
- ④ $\forall A, B$ square, with $\det A \neq 0, \det B \neq 0 \rightarrow (A+B)^{-1} = A^{-1} + B^{-1}$
- ⑤ $\forall A$ square, with $\det A \neq 0, \forall \alpha \neq 0 \rightarrow (\alpha A)^{-1} = \frac{1}{\alpha} A^{-1}$
- ⑥ $\forall A$ square, with $\det A \neq 0 \rightarrow \det(A^{-1}) = \frac{1}{\det A}$

THEOREM

consider a linear function $f: \mathbb{R}^n \rightarrow \mathbb{R}^n$

call A its representation matrix which is a square matrix of order n , then:

- ① f invertible $\Leftrightarrow A$ invertible
- ② when f is invertible, the representation matrix of $f^{-1}: \mathbb{R}^n \rightarrow \mathbb{R}^n$ is the square matrix A^{-1}

RANK OF A MATRIX

RANK BY COLUMNS

given a $m \times n$ matrix A

the rank by columns of the matrix A is the maximum number of linearly independent columns in A

RANK BY ROWS

give a $m \times n$ matrix A ,

the rank by rows of the matrix A is the maximum number of linearly independent rows in A

we have:

THEOREM

given a $m \times n$ matrix A ,
the rank by columns of A is always coincident with the rank by rows of A
therefore, from now on, we will just call it the rank of A : it gives both the
maximum number of linearly independent columns and the maximum
number of linearly independent rows in A

THEOREM: COMPUTATION OF RANK

consider a $m \times n$ matrix A , then:

$\text{rank } A = k \iff k$ is the maximum order for which in A we have a non-null minor

KRONECKER'S THEOREM

consider a $m \times n$ matrix A

if \exists a non-null minor of A of order k , and all the minors of A of order $k+1$
which contains this non-null minor are null, then $\text{rank } A = k$

REMARK: THEOREM

consider a linear function $f: \mathbb{R}^n \rightarrow \mathbb{R}^m$

consider its representation matrix A , then:

- ① $\text{rank } f = k \iff \text{rank } A = k$
- ② f surjective $\iff \text{rank } A = m$ (n. of rows)
- f injective $\iff \text{rank } A = n$ (n. of columns)
- f bijective $\iff m = n, \text{rank } A = n$
- $[\det A \neq 0]$

LINEAR SYSTEMS

HOW TO WRITE IT

- explicit writing =

- one equation $\rightarrow a_1 x_1 + \dots + a_n x_n = b$
- m equations \rightarrow

$$\begin{cases} a_{11}x_1 + a_{12}x_2 + \dots + a_{1n}x_n = b_1 \\ a_{21}x_1 + a_{22}x_2 + \dots + a_{2n}x_n = b_2 \\ \vdots \\ a_{m1}x_1 + a_{m2}x_2 + \dots + a_{mn}x_n = b_m \end{cases}$$

$[a_{ij}: i = \text{equation index}, j = \text{variable index}]$

- matrix writing = $A \underline{x} = \underline{b} \Rightarrow$

$\underline{x} = \begin{pmatrix} x_1 \\ \vdots \\ x_n \end{pmatrix}$ vector of variables

$\underline{b} = \begin{pmatrix} b_1 \\ \vdots \\ b_n \end{pmatrix}$ vector of constant terms

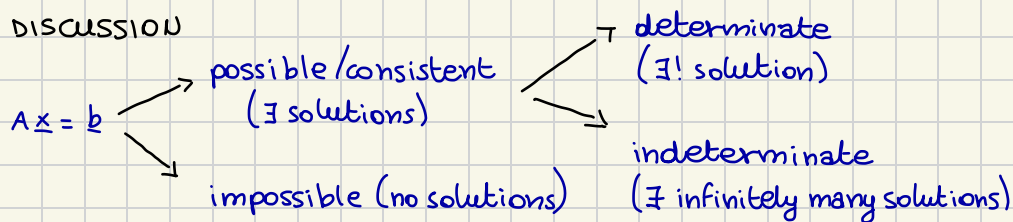
$$\Rightarrow \begin{pmatrix} a_{11} & \dots & a_{1n} \\ \vdots & & \vdots \\ a_{m1} & \dots & a_{mn} \end{pmatrix} \begin{pmatrix} x_1 \\ \vdots \\ x_n \end{pmatrix} = \begin{pmatrix} b_1 \\ \vdots \\ b_n \end{pmatrix}$$

$A = \begin{pmatrix} a_{11} & \dots & a_{1n} \\ \vdots & & \vdots \\ a_{m1} & \dots & a_{mn} \end{pmatrix}$ matrix of coefficients

- column writing = $a_1 x_1 + a_2 x_2 + \dots + a_n x_n = b \Rightarrow$

$$\begin{cases} a_{11}x_1 + a_{12}x_2 + \dots + a_{1n}x_n = b_1 \\ a_{21}x_1 + a_{22}x_2 + \dots + a_{2n}x_n = b_2 \\ \vdots \\ a_{m1}x_1 + a_{m2}x_2 + \dots + a_{mn}x_n = b_m \end{cases}$$

DISCUSSION



THEOREM: DISTINCTION BETWEEN DETERMINATE/INDETERMINATE CASES

given a linear system $A \underline{x} = \underline{b}$ with m equations and n variables, then:

- ① if $\text{rank } A < \text{rank } (A|\underline{b}) \Rightarrow$ impossible
- ② if $\text{rank } A = \text{rank } (A|\underline{b}) = n \Rightarrow$ determinate ($\exists!$ solution)
- ③ if $\text{rank } A = \text{rank } (A|\underline{b}) = k < n \Rightarrow$ indeterminate, with $n-k$ degrees of freedom (\exists infinitely many solutions)

TYPES OF SYSTEMS

- ① $A \underline{x} = \underline{0} \Rightarrow$ homogeneous linear systems always possible (the solution is always $\underline{x} = \underline{0}$)
- ② $A \underline{x} = \underline{b}$, $\underline{b} \neq \underline{0} \Rightarrow$ non-homogeneous linear systems (may be possible or not possible)

remark

looking for \underline{x} : $A \underline{x} = \underline{b} \approx$ looking for \underline{x} : $f(\underline{x}) = \underline{b}$

STRUCTURE OF SOLUTIONS

① homogeneous linear systems, $Ax = 0 \Rightarrow$

- always possible + one solution is always $x = 0$
- consider the set of all solutions: $V = \{x \in \mathbb{R}^n : Ax = 0\} = \text{Ker } f \Rightarrow$
 - the set of all solutions V is always a subspace of \mathbb{R}^n
 - dimension: $\dim V = n - \text{rank } A$ [by the rank nullity theorem]
 - therefore:
 - if $\text{rank } A = n$, one solution $x = 0$, $V = \{0\}$ and $\dim V = 0$
 - if $\text{rank } A = k < n$, infinitely many solutions which form a subspace V of \mathbb{R}^n with $\dim V = n - k$

② non-homogeneous linear systems, $Ax = b$, $b \neq 0 \Rightarrow$

- possible or impossible \rightarrow check if $\text{rank } A = \text{rank}(A|b)$ + $x = 0$ is never a solution
- consider the set of all solutions: $W = \{x \in \mathbb{R}^n : Ax = b, b \neq 0\}$, we have that $W = \{x \in \mathbb{R}^n : f(x) = b, b \neq 0\}$ [W is never a subspace of \mathbb{R}^n , $0 \notin W$]
- THEOREM: STRUCTURE OF SOLUTIONS IN THE NON-HOMOGENEOUS CASE
consider a linear system $Ax = b$, $b \neq 0$
suppose it is possible ($\text{rank } A = \text{rank}(A|b)$), then:
if you call x the general solution of $Ax = b$ and you call x^* one particular solution of $Ax = b$, you have $x = x^* + z$
where z is the general solution of the homogeneous linear system $Ax = 0$
the task of solving a non-homogeneous linear system $Ax = b$ can be reduced to finding one solution of $Ax = b$ and solving the homogeneous linear system $Ax = 0$ instead

SOLUTION

given $Ax = b$, suppose we checked that $Ax = b$ is possible, we can find the solutions using Cramer \rightarrow

$Ax = b$, with A square matrix of order n and $\det A \neq 0$

CRAMER'S RULE

consider $Ax = b$, with A square of order n

if $\det A \neq 0$, then: $x = \begin{pmatrix} x_1 \\ \vdots \\ x_n \end{pmatrix}$, and $x_1 = \frac{\det A_1}{\det A}, \dots, x_n = \frac{\det A_n}{\det A}$

where every matrix A_k is the matrix we obtain from matrix A , substituting the column k with the vector b

A not square or A square, $\det A = 0 \Rightarrow$

CRAMER'S METHOD (example of the general procedure)

consider a linear system $A \underline{x} = \underline{b}$ where A is any $m \times n$ matrix

$$A = \begin{pmatrix} 3 & 2 & 1 \\ -1 & 0 & -1 \\ 4 & 5 & -1 \\ 2 & 5 & -3 \end{pmatrix}, \quad \underline{b} = \begin{pmatrix} -1 \\ -1 \\ -6 \\ -8 \end{pmatrix}$$

① $\text{rank } A = \text{rank} \begin{pmatrix} 3 & 2 & 1 \\ -1 & 0 & -1 \\ 4 & 5 & -1 \\ 2 & 5 & -3 \end{pmatrix} \xrightarrow{B}$

$$\det B = 0 + 2 = 2 \neq 0$$

$$\det \begin{pmatrix} 3 & 2 & 1 \\ -1 & 0 & -1 \\ 4 & 5 & -1 \end{pmatrix} = 0 \quad \text{III}c = \text{I}c - \text{II}c$$

$$\det \begin{pmatrix} 3 & 2 & 1 \\ -1 & 0 & -1 \\ 2 & 5 & -3 \end{pmatrix} = 0 \quad \text{III}c = \text{I}c - \text{II}c$$

$$\Rightarrow \text{rank } A = 2$$

$$\text{rank}(A|\underline{b}) = \text{rank} \begin{pmatrix} 3 & 2 & 1 & -1 \\ -1 & 0 & -1 & -1 \\ 4 & 5 & -1 & -6 \\ 2 & 5 & -3 & -8 \end{pmatrix} \xrightarrow{B}$$

proceeding this way, since $\text{III}c = \text{I}c - 2\text{II}c$, we get that also $\text{rank}(A|\underline{b}) = 2$ therefore:

$A \underline{x} = \underline{b}$ is possible, indeterminate, with $3 - 2 = 1$ degree of freedom

② cancel out all the rows which have no intersection with the square submatrix B you find when computing the rank, that is:

$$A = \begin{pmatrix} 3 & 2 & 1 \\ -1 & 0 & -1 \\ 4 & 5 & -1 \\ 2 & 5 & -3 \end{pmatrix}, \quad \underline{b} = \begin{pmatrix} -1 \\ -1 \\ -6 \\ -8 \end{pmatrix} \Rightarrow \text{dependent} = \text{useless}$$

③ write the system explicitly:
$$\begin{cases} 3x_1 + 2x_2 + x_3 = -1 \\ -x_1 = -1 + x_3 \end{cases}$$

we move all the columns which have no intersection with the square submatrix B to the right-hand side of the equation

$$\begin{cases} 3x_1 + 2x_2 = -1 - x_3 \\ -x_1 = -1 + x_3 \end{cases} \quad \text{that is: } B = \begin{pmatrix} 3 & 2 \\ -1 & 0 \end{pmatrix}, \quad \underline{b} = \begin{pmatrix} -1 - x_3 \\ -1 + x_3 \end{pmatrix}$$

④ on the left we have a square matrix B with $\det B \neq 0$ so we use Cramer's rule

$\underline{x} = \begin{pmatrix} x_1 \\ x_2 \end{pmatrix}$, with:

$$x_1 = \frac{\det B_1}{\det B} = \frac{\det \begin{pmatrix} -1-x_3 & 2 \\ -1+x_3 & 0 \end{pmatrix}}{2} = \frac{2-2x_3}{2} = 1-x_3$$

$$x_2 = \frac{\det B_2}{\det B} = \frac{\det \begin{pmatrix} 3 & -1-x_3 \\ -1 & -1+x_3 \end{pmatrix}}{2} = \frac{-3+3x_3-1-x_3}{2} = -2+x_3$$

we find $k=2$ variables ($k=\text{rank } A$) in terms of $n-k=1$ free parameter (degree of freedom):

$$\begin{cases} x_1 = 1-x_3 \\ x_2 = -2+x_3 \\ x_3 = x_3 \end{cases}$$

PROBLEM OF LEAST SQUARES

consider a linear system: $A \underline{x} = \underline{b}$ with $A \in \mathcal{R}(m, n)$

suppose that it has no solutions (often the case when $m > n$) that is when you have more equations than variables

if the linear system has no solutions it means that:

$\nexists \underline{x} \in \mathbb{R}^n$ such that $A \underline{x} = \underline{b}$ [equivalent: $A \underline{x} - \underline{b} = \underline{0}$]

? $\exists \underline{x}^* \in \mathbb{R}^n$ such that \underline{x}^* minimises the error $\|A \underline{x} - \underline{b}\|$? = distance between the vector \underline{b} of the constants and the image $T(\underline{x}) = A \underline{x}$ of the linear operator $T: \mathbb{R}^n \rightarrow \mathbb{R}^m$ that is individuated by the matrix A

this error is null where \underline{x}^* solves the linear system $A \underline{x} = \underline{b}$, but in general we have: $\|A \underline{x} - \underline{b}\| \geq 0$

\underline{x}^* minimises the error $\|A \underline{x} - \underline{b}\|$ if and only if it minimises its square $(\|A \underline{x} - \underline{b}\|)^2$

DEF. a vector $\underline{x}^* \in \mathbb{R}^n$ is said to be a least squares solution of the linear system:

$A \underline{x} = \underline{b}$, if \underline{x}^* is a solution of the optimisation problem: $\min_{\underline{x} \in \mathbb{R}^n} (\|A \underline{x} - \underline{b}\|)^2$

that is if \underline{x}^* is the best vector \underline{x} that we can find in \mathbb{R}^n , in order to minimise the quantity $(\|A \underline{x} - \underline{b}\|)^2$

the name comes from the fact that $\|A \underline{x} - \underline{b}\|^2$ is a sum of squares

if the linear system $A \underline{x} = \underline{b}$ has a solution \underline{x} , it is also a least squares solution but a least squares solutions \underline{x}^* may exist also when the linear system $A \underline{x} = \underline{b}$ has no solution \underline{x}

THEOREM

consider the linear system $Ax = b$, with $A \in \mathbb{R}(m, n)$

suppose $m \geq n$

then if $\text{rank } A = n$ [if A has maximum rank], the optimisation problem:

$\min_{x \in \mathbb{R}^n} (\|Ax - b\|)^2$ has a solution and this solution is unique

**theorems and proofs for
linear algebra**

THE INTERSECTION OF VECTOR SUBSPACES

The intersection of any collection of vector subspaces of \mathbb{R}^n is a vector subspace

PROOF

Let $\{V_i\}$ be any collection of vector subspaces of \mathbb{R}^n

since $0 \in V_i \forall i$, we have $\bigcap_i V_i \neq \emptyset$

Let $x, y \in \bigcap_i V_i$ and $\alpha, \beta \in \mathbb{R}$

since $x, y \in V_i \forall i$ and therefore $\alpha x + \beta y \in V_i \forall i$ since each V_i is a vector subspace of \mathbb{R}^n

hence $\alpha x + \beta y \in \bigcap_i V_i$ and so $\bigcap_i V_i$ is a vector subspace of \mathbb{R}^n

CHARACTERIZATION WITH LINEAR COMBINATIONS

S subset of \mathbb{R}^n

$x \in \text{Span } S \iff x$ is a linear combination of vectors of S

$\implies \exists x_1, \dots, x_m \in S, \exists \alpha_1, \dots, \alpha_m \in \mathbb{R}$ such that $x = \alpha_1 x_1 + \dots + \alpha_m x_m$

PROOF

\implies only if

$V =$ set of all vectors $x \in \mathbb{R}^n$ that can be written as a linear combination of vectors of S

if $x, y \in V$ then x, y are linear combinations of vectors of S and therefore also $x+y$ is a linear combination of vectors of S which also implies $x+y \in V$

V is a vector subspace of \mathbb{R}^n containing S so it must be that $\text{span } S \subseteq V$, that is:

each $x \in \text{span } S$ can be written as a linear combination of vectors of S

\Leftarrow if

suppose $x \in \mathbb{R}^n$ (linear combination of vectors of S): $x = \alpha_1 x_1 + \dots + \alpha_m x_m$ for some

$\alpha_1, \dots, \alpha_m$ and $x_1, \dots, x_m \in S$

since $x_1, x_2 \in S$ they belong to $\text{span } S$

since $\text{span } S$ is a vector space also $\alpha_1 x_1 + \alpha_2 x_2$ must belong to $\text{span } S$

since $x_3 \in S$ it also belongs to $\text{span } S$

since $\text{span } S$ is a vector space also $\alpha_1 x_1 + \alpha_2 x_2 + \alpha_3 x_3$ belongs to $\text{span } S$

proceeding in this way in the end we get that also $x = \alpha_1 x_1 + \dots + \alpha_m x_m$ belongs to $\text{span } S$

as a consequence of the theorem if: $S = \{x_1, \dots, x_m \in \mathbb{R}^n\}$ then,
 $\text{span } S = \{x \in \mathbb{R}^n : \exists \alpha_1, \dots, \alpha_m \in \mathbb{R} \text{ such that } x = \alpha_1 x_1 + \dots + \alpha_m x_m\}$

PROPERTY OF UNIQUE WRITING FOR A BASIS

take $x_1, \dots, x_m \in \mathbb{R}^n$

they are a basis of $\mathbb{R}^n \iff \forall x \in \mathbb{R}^n, x$ can be written in a unique way as a linear combination of x_1, \dots, x_m

PROOF

\Rightarrow only if

suppose x_1, \dots, x_m are a basis of \mathbb{R}^n

by definition of basis every $x \in \mathbb{R}^n$ can be written as a linear combination of x_1, \dots, x_m

$$x = \alpha_1 x_1 + \dots + \alpha_m x_m$$

$$x = \beta_1 x_1 + \dots + \beta_m x_m$$

this means that: $(\alpha_1 - \beta_1) x_1 + \dots + (\alpha_m - \beta_m) x_m = 0$

but since x_1, \dots, x_m are linearly independent, this means that it must be:

$$\alpha_1 - \beta_1 = 0, \dots, \alpha_m - \beta_m = 0$$

$$\text{that is: } \alpha_1 = \beta_1, \dots, \alpha_m = \beta_m$$

which proves the unique possible writing

\Leftarrow if

suppose that $\forall x \in \mathbb{R}^n, x$ can be written in a unique way as a linear combination of x_1, \dots, x_m

this implies that $\text{span } S = \mathbb{R}^n$

we have to prove that x_1, \dots, x_m are linearly independent

but if we suppose that $\exists \alpha_1, \dots, \alpha_m$ such that: $\alpha_1 x_1 + \dots + \alpha_m x_m = 0$

it is enough to remark that also $0 x_1 + \dots + 0 x_m = 0$

since 0 can only be written in a unique way as a linear combination of

x_1, \dots, x_m this implies that: $\alpha_1 = 0, \dots, \alpha_m = 0$

that is: x_1, \dots, x_m are linearly independent

DETERMINATION OF COEFFICIENTS OF A VECTOR IN \mathbb{R}^n WITH RESPECT TO AN ORTHONORMAL BASIS

consider an orthonormal basis $\underline{x}_1, \dots, \underline{x}_n$ in \mathbb{R}^n

then for every $v = \alpha_1 \underline{x}_1 + \dots + \alpha_n \underline{x}_n$ with $\alpha_1 = v \cdot \underline{x}_1, \dots, \alpha_n = v \cdot \underline{x}_n$

meaning that if $\underline{x}_1, \dots, \underline{x}_n$ is an orthonormal basis of \mathbb{R}^n , there is a quick way to write the coefficients $\alpha_1, \dots, \alpha_n$

PROOF

let $\{\underline{x}_1, \underline{x}_2, \dots, \underline{x}_n\}$ be an orthonormal basis of \mathbb{R}^n

for any vector $v \in \mathbb{R}^n$ it can be expressed as a linear combination of the basis vectors:

$v = \alpha_1 \underline{x}_1 + \alpha_2 \underline{x}_2 + \dots + \alpha_n \underline{x}_n$ where $\alpha_1, \alpha_2, \dots, \alpha_n$ are the coefficients we want to determine

by the orthonormality of the basis:

$$\underline{x}_i \cdot \underline{x}_j = \begin{cases} 1 & \text{if } i=j \\ 0 & \text{if } i \neq j \end{cases}$$

take the dot product of both sides of $v = \sum_{i=1}^n \alpha_i \underline{x}_i$ with \underline{x}_k where $k \in \{1, 2, \dots, n\}$:

$$v \cdot \underline{x}_k = \left(\sum_{i=1}^n \alpha_i \underline{x}_i \right) \cdot \underline{x}_k$$

using the linearity of the dot product:

$$v \cdot \underline{x}_k = \sum_{i=1}^n \alpha_i (\underline{x}_i \cdot \underline{x}_k)$$

from the orthonormality property:

$$\underline{x}_i \cdot \underline{x}_k = \begin{cases} 1 & \text{if } i=k \\ 0 & \text{if } i \neq k \end{cases}$$

therefore all terms in the sum vanish except when $i=k$:

$$v \cdot \underline{x}_k = \alpha_k (\underline{x}_k \cdot \underline{x}_k) = \alpha_k \cdot 1 = \alpha_k$$

thus the coefficient α_k is given by:

$$\alpha_k = v \cdot \underline{x}_k$$

the coefficient $\alpha_1, \alpha_2, \dots, \alpha_n$ of the vector v with respect to the orthonormal basis $\{\underline{x}_1, \underline{x}_2, \dots, \underline{x}_n\}$

can be quickly determined as:

$$\alpha_k = v \cdot \underline{x}_k, \text{ for } k=1, 2, \dots, n$$

RIESZ REPRESENTATION THEOREM

a function $f: \mathbb{R}^n \rightarrow \mathbb{R}$ is linear if and only if there exists a unique vector $\chi \in \mathbb{R}^n$ such that:

$$f(x) = \chi \cdot x \quad \forall x \in \mathbb{R}^n$$

PROOF

\Leftarrow "if"

assume $f(x) = \chi \cdot x$ for some $\chi \in \mathbb{R}^n$

we show that f is linear

$$\text{for } x, y \in \mathbb{R}^n, f(x+y) = \chi \cdot (x+y) = \chi \cdot x + \chi \cdot y = f(x) + f(y)$$

$$\text{for } \alpha \in \mathbb{R} \text{ and } x \in \mathbb{R}^n, f(\alpha x) = \chi \cdot (\alpha x) = \alpha (\chi \cdot x) = \alpha f(x)$$

since f satisfies the additivity and homogeneity, f is linear

\Rightarrow "only if"

assume $f: \mathbb{R}^n \rightarrow \mathbb{R}$ is linear

we show that there exists a unique $\chi \in \mathbb{R}^n$ such that $f(x) = \chi \cdot x$

define χ as $\chi_i = f(e_i)$ for $i = 1, \dots, n$ where e_i is the i -th standard basis vector of \mathbb{R}^n

for any $x = (x_1, x_2, \dots, x_n) \in \mathbb{R}^n$, $x = \sum_{i=1}^n x_i e_i$

and by linearity of f , $f(x) = f(\sum_{i=1}^n x_i e_i) = \sum_{i=1}^n x_i f(e_i) = \sum_{i=1}^n x_i \chi_i = \chi \cdot x$

if $f(x) = y^* \cdot x$ for another vector y^* then $\chi \cdot x = y^* \cdot x \quad \forall x \in \mathbb{R}^n$

choosing $x = e_i$, we get $\chi_i = y_i^*$, thus $\chi = y^*$

RIESE REPRESENTATION THEOREM FOR OPERATORS $f: \mathbb{R}^n \rightarrow \mathbb{R}^m$

- ① a function $f: \mathbb{R}^n \rightarrow \mathbb{R}^m$ is linear \Leftrightarrow there exists a matrix $A \in \mathcal{K}(m, n)$ called the representation matrix of f such that $\forall \underline{x} = \begin{pmatrix} x_1 \\ \vdots \\ x_n \end{pmatrix} \in \mathbb{R}^n$ we have: $f(\underline{x}) = A\underline{x}$
- ② when it exists this matrix A is unique

PROOF

\Leftarrow "if"

suppose $\exists A \in \mathcal{K}(m, n)$ such that:

$f(\underline{x}) = A\underline{x} \quad \forall \underline{x} \in \mathbb{R}^n$, then:

- ① $\forall \underline{x}, \underline{y} \in \mathbb{R}^n, f(\underline{x} + \underline{y}) = A(\underline{x} + \underline{y}) = A\underline{x} + A\underline{y} = f(\underline{x}) + f(\underline{y})$
- ② $\forall \underline{x} \in \mathbb{R}^n, \forall \alpha \in \mathbb{R}, f(\alpha \underline{x}) = A(\alpha \underline{x}) = \alpha(A\underline{x}) = \alpha f(\underline{x})$

\Rightarrow "only if"

we use the fact that:

$\forall \underline{x} = \begin{pmatrix} x_1 \\ \vdots \\ x_n \end{pmatrix} \in \mathbb{R}^n$ we can write $\underline{x} = x_1 \underline{e}_1 + \dots + x_n \underline{e}_n$

suppose f is linear, $\forall \underline{x} \in \mathbb{R}^n$ we have:

$$\begin{aligned} f(\underline{x}) &= f(x_1 \underline{e}_1 + \dots + x_n \underline{e}_n) \\ &= f(x_1 \underline{e}_1) + \dots + f(x_n \underline{e}_n) \\ &= x_1 f(\underline{e}_1) + \dots + x_n f(\underline{e}_n) \\ &= \underline{\alpha}_1 x_1 + \dots + \underline{\alpha}_n x_n = A\underline{x} \end{aligned}$$

if we consider the matrix $A = (\underline{a}_1, \dots, \underline{a}_n) = (f(\underline{e}_1), \dots, f(\underline{e}_n))$ whose columns are given by the vectors: $\underline{\alpha}_1 = f(\underline{e}_1), \dots, \underline{\alpha}_n = f(\underline{e}_n)$

② uniqueness \Rightarrow

suppose there exist $A, A' \in \mathcal{K}(m, n)$ such that $\forall \underline{x} \in \mathbb{R}^n, f(\underline{x}) = A\underline{x} = A'\underline{x}$

if we take $\underline{x} = \underline{e}_1$ we find:

$$f(\underline{e}_1) = A\underline{e}_1 = \underline{\alpha}_1$$

$$f(\underline{e}_1) = A'\underline{e}_1 = \underline{\alpha}'_1$$

where $\underline{\alpha}_1$ is the first column of A and $\underline{\alpha}'_1$ is the first column of A'

therefore $\underline{\alpha}_1 = \underline{\alpha}'_1$

taking $\underline{x} = \underline{e}_2$ we prove that it is $\underline{\alpha}_2 = \underline{\alpha}'_2$ and so on, therefore $A = A'$

THE IMAGE SPACE IS A SUBSPACE SPANLED BY THE IMAGE OF STANDARD BASIS

- ① $f(\mathbb{R}^n) = \mathcal{I}mf$ is always a vector subspace of \mathbb{R}^m
- ② a spanning set for $\mathcal{I}mf$ is always: $f(\underline{e}_1), \dots, f(\underline{e}_n)$

PROOF

① consider $\underline{y}_1, \underline{y}_2 \in \mathcal{I}mf, \alpha_1, \alpha_2 \in \mathbb{R}$

then $\exists \underline{x}_1 \in \mathbb{R}^n : \underline{y}_1 = f(\underline{x}_1), \exists \underline{x}_2 \in \mathbb{R}^n : \underline{y}_2 = f(\underline{x}_2)$

since f is linear, $f(\alpha_1 \underline{x}_1 + \alpha_2 \underline{x}_2) = \alpha_1 f(\underline{x}_1) + \alpha_2 f(\underline{x}_2) = \alpha_1 \underline{y}_1 + \alpha_2 \underline{y}_2$

this means that $\alpha_1 \underline{y}_1 + \alpha_2 \underline{y}_2 \in \mathcal{I}mf$

therefore $\mathcal{I}mf$ is a vector subspace of \mathbb{R}^m

② consider $\underline{y} \in \mathcal{I}mf$

then $\exists \underline{x} \in \mathbb{R}^n : \underline{y} = f(\underline{x})$

considering the fundamental basis $\underline{e}_1, \dots, \underline{e}_n$ of \mathbb{R}^n , we have that $\underline{x} = x_1 \underline{e}_1 + \dots + x_n \underline{e}_n$

since f is linear we have $\underline{y} = f(\underline{x}) = f(x_1 \underline{e}_1 + \dots + x_n \underline{e}_n) = x_1 f(\underline{e}_1) + \dots + x_n f(\underline{e}_n)$

this means that \underline{y} is a linear combination of $f(\underline{e}_1), \dots, f(\underline{e}_n)$

therefore $f(\underline{e}_1), \dots, f(\underline{e}_n)$ is a spanning set for $\mathcal{I}mf$

THE KERNEL IS A SUBSPACE

if $T: \mathbb{R}^n \rightarrow \mathbb{R}^m$ is linear then $\ker T$ is a vector subspace of \mathbb{R}^n

PROOF

let $x, x' \in \ker T$, i.e. $T(x) = 0$ and $T(x') = 0$

for every $\alpha, \beta \in \mathbb{R}$ we have: $T(\alpha x + \beta x') = \alpha T(x) + \beta T(x') = \alpha 0 + \beta 0 = 0$

thus, $\alpha x + \beta x' \in \ker T$ and this proves that $\ker T$ is a vector subspace of \mathbb{R}^n

DETERMINANT OF THE INVERSE MATRIX

for any invertible square matrix A , the determinant of its inverse satisfies:

$$\det(A^{-1}) = \frac{1}{\det(A)} \text{ where } \det(A) \neq 0$$

PROOF

by definition of the inverse matrix, $A \cdot A^{-1} = I$, where I is the identity matrix
taking the determinant of both sides, we have $\det(A \cdot A^{-1}) = \det(I)$

using the determinant property $\det(AB) = \det(A) \cdot \det(B)$, this becomes:

$$\det(A) \cdot \det(A^{-1}) = \det(I)$$

since the determinant of the identity matrix is 1 ($\det(I) = 1$), it follows that:

$$\det(A) \cdot \det(A^{-1}) = 1$$

rearranging we find: $\det(A^{-1}) = \frac{1}{\det(A)}$

this result holds as long as A is invertible, which implies $\det(A) \neq 0$

KRONECKER - CAPELLI'S THEOREM

given a linear system $A\underline{x} = \underline{b}$ with m equations and n variables,

$A\underline{x} = \underline{b}$ is possible $\Leftrightarrow \text{rank } A = \text{rank } (A | \underline{b})$

PROOF

$A\underline{x} = \underline{b}$ is possible \Leftrightarrow

$\exists \underline{x} = \begin{pmatrix} x_1 \\ \vdots \\ x_n \end{pmatrix} \in \mathbb{R}^n: A\underline{x} = \underline{b} \Leftrightarrow$

$\exists x_1, \dots, x_n \in \mathbb{R}: \underline{b}$ can be written as a linear combination $\underline{b} = x_1 \underline{a}_1 + \dots + x_n \underline{a}_n$ of the columns of $A \Leftrightarrow$

\underline{b} is linearly dependent on the columns of $A \Leftrightarrow$

$\text{rank } A = \text{rank } (A | \underline{b})$

CRAUER'S THEOREM

consider the system $Ax = b$ where A is a square matrix of order n :

- ① the system $Ax = b$ is determinable (it has a unique solution) if and only if $\det(A) \neq 0$
- ② if $\det(A) \neq 0$ then the solution is given by $x = A^{-1}b$

PROOF

① if $\det(A) \neq 0$ the matrix A is invertible by definition of the determinant
the invertibility of A ensures the existence of A^{-1} , the inverse of A

for the system $Ax = b$, multiplying both sides by A^{-1} , we get: $A^{-1}(Ax) = A^{-1}b$

using the property $A^{-1}A = I$, this simplifies to $x = A^{-1}b$

since A^{-1} exists this is the unique solution

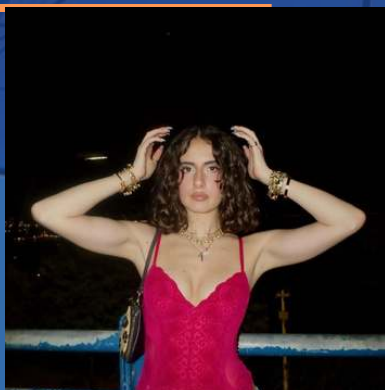
② if $\det(A) = 0$, the matrix A is not invertible, which implies A is singular

for a singular matrix, the system $Ax = b$ may have no solution (if b lies outside the column space of A) or infinitely many solutions if b lies in the column space of A but A does not have full rank

in either cases the system is not uniquely solvable (it is not determinable)

the system $Ax = b$ is uniquely solvable if and only if $\det(A) \neq 0$ in which case the solution is $x = A^{-1}b$

FOR DOUBTS OR SUGGESTIONS ON THE HANDOUTS



ALESSANDRA VITIELLO

alessandra.vitiello@studbocconi.it

[@alexvitiello](https://www.instagram.com/alexvitiello)

+39 344 491 6912

FOR INFO ON THE TEACHING DIVISION



VITTORIA NASONTE

vittoria.nasonte@studbocconi.it

[@_vittorian_](https://www.instagram.com/_vittorian_)

+39 3274441476



ELENA CACIOLI

elena.cacioli@studbocconi.it

[@elenacaciolii_](https://www.instagram.com/elenacaciolii_)

+39 3928931605



TEACHING DIVISION



OUR PARTNERS

700+
CLUB



ETHAN
SUSTAINABILITY

DELIVERY VALLEY

NO GENDER KITCHEN

LA PIADINERIA

