

Geom - Av - 11/3/21

V sp. vet. su $\mathbb{R}$ con $\langle \cdot, \cdot \rangle$ prod scal, $\|\cdot\|$ norma.

$v_1, \dots, v_k$ è ortog. se $v_i \perp v_j \quad \forall i \neq j \quad (\langle v_i, v_j \rangle = 0)$
ortonormale se inoltre $\|v_i\| = 1$.

- sist. ortog. di vett $\neq 0$ è lin. indip.
- $v_1, \dots, v_m$ base ortog. $\Rightarrow v = \sum_{i=1}^m \frac{\langle v, v_i \rangle}{\|v_i\|^2} \cdot v_i$
ortonormale $\Rightarrow 1$

Q: come produrre basi ortogonali.

A: Gram-Schmidt "ortogonalizzazione"

Teo: data $v_1, \dots, v_m$ base di V il seguente procedimento fornisce una base ortonormale $w_1, \dots, w_m$:

- $w_1 = v_1 / \|v_1\|$

- $k \geq 1$ si pone $\tilde{w}_{k+1} = v_{k+1} - \sum_{i=1}^k \langle v_{k+1}, w_i \rangle \cdot w_i$

(sappiamo $w_1, \dots, w_k$ già fatti)

$$w_{k+1} = \tilde{w}_{k+1} / \|\tilde{w}_{k+1}\|$$

Es: in $\mathbb{R}^3$ $\begin{pmatrix} 1 \\ 2 \\ 1 \end{pmatrix}_{v_1} \begin{pmatrix} 3 \\ 1 \\ 4 \end{pmatrix}_{v_2} \begin{pmatrix} -1 \\ 0 \\ 2 \end{pmatrix}_{v_3}$

$$w_1 = \begin{pmatrix} 1 \\ 2 \\ -1 \end{pmatrix} / \sqrt{1+4+1} = \frac{1}{\sqrt{6}} \begin{pmatrix} 1 \\ 2 \\ -1 \end{pmatrix}$$

$$\tilde{w}_2 = \begin{pmatrix} 3 \\ 1 \\ 4 \end{pmatrix} - \left\langle \begin{pmatrix} 3 \\ 1 \\ 4 \end{pmatrix} \middle| \frac{1}{\sqrt{6}} \begin{pmatrix} 1 \\ 2 \\ -1 \end{pmatrix} \right\rangle \cdot \frac{1}{\sqrt{6}} \begin{pmatrix} 1 \\ 2 \\ -1 \end{pmatrix}$$

$$= \begin{pmatrix} 3 \\ 1 \\ 4 \end{pmatrix} - \frac{3+2-4}{6} \begin{pmatrix} 1 \\ 2 \\ -1 \end{pmatrix} = \begin{pmatrix} 3 \\ 1 \\ 4 \end{pmatrix} - \frac{1}{6} \begin{pmatrix} 1 \\ 2 \\ -1 \end{pmatrix} = \frac{1}{6} \begin{pmatrix} 17 \\ 4 \\ 25 \end{pmatrix}$$

verifica: vogliamo $\tilde{w}_2 \perp w_1$, cioè $\tilde{w}_2 \perp v_1$

$$\left\langle \begin{pmatrix} 1 \\ 2 \\ -1 \end{pmatrix} \middle| \begin{pmatrix} 17 \\ 4 \\ 25 \end{pmatrix} \right\rangle = 17+8-25=0 \quad \checkmark$$

Oss: $x = \alpha \cdot y \quad \alpha > 0$

$$\frac{\alpha}{\|x\|} = \frac{\alpha \cdot y}{\|\alpha \cdot y\|} = \frac{\alpha \cdot y}{\alpha \cdot \|y\|} = \frac{y}{\|y\|}$$

$$w_2 = \frac{1}{\sqrt{17^2+4^2+25^2}} \cdot \begin{pmatrix} 17 \\ 4 \\ 25 \end{pmatrix}$$

$$\tilde{w}_3 = \begin{pmatrix} -1 \\ 0 \\ 2 \end{pmatrix} - \frac{1}{6} \left\langle \begin{pmatrix} -1 \\ 0 \\ 2 \end{pmatrix} \middle| \begin{pmatrix} 1 \\ 2 \\ -1 \end{pmatrix} \right\rangle \cdot \begin{pmatrix} 1 \\ 2 \\ -1 \end{pmatrix} - \frac{1}{\sqrt{17^2+4^2+25^2}} \left\langle \begin{pmatrix} -1 \\ 0 \\ 2 \end{pmatrix} \middle| \begin{pmatrix} 17 \\ 4 \\ 25 \end{pmatrix} \right\rangle \cdot \begin{pmatrix} 17 \\ 4 \\ 25 \end{pmatrix}$$

= ...

$$w_3 = \tilde{w}_3 / \|\tilde{w}_3\|$$

Dimo: • $w_1 = v_1 / \|v_1\|$

• $k \geq 1$ si pone $\tilde{w}_{k+1} = v_{k+1} - \sum_{i=1}^k \langle v_{k+1} | w_i \rangle \cdot w_i$

(sappiamo $w_1, \dots, w_k$ già fatti)

$$w_{k+1} = \tilde{w}_{k+1} / \|\tilde{w}_{k+1}\|$$

Verifico per $k=1, \dots, n$ che w_k si può definire, è unitario, è ortogonale a $w_1, \dots, w_{k-1}$ e $\text{Span}(w_1, \dots, w_k) = \text{Span}(v_1, \dots, v_k)$.

$k=1$: tutto ok

Supponiamo vero per k e proviamo per $k+1$:

$$\bullet \tilde{w}_{k+1} = v_{k+1} - \underbrace{\sum_{i=1}^k \langle v_{k+1} | w_i \rangle \cdot w_i}_{\in \text{Span}(w_1, \dots, w_k) = \text{Span}(v_1, \dots, v_k)}$$

$\Rightarrow \neq 0$ poiché $v_1, \dots, v_{k+1}$ lin. indip.

$\Rightarrow$ si può definire; ✓

unitario ✓

ortog. ai precedenti : basta vederlo per $\tilde{w}_{k+1}$; $j \leq k$

$$\begin{aligned} \langle \tilde{w}_{k+1} | w_j \rangle &= \left\langle v_{k+1} - \sum_{i=1}^k \langle v_{k+1} | w_i \rangle \cdot w_i \mid w_j \right\rangle \\ &= \langle v_{k+1} | w_j \rangle - \sum_{i=1}^k \langle v_{k+1} | w_i \rangle \cdot \underbrace{\langle w_i | w_j \rangle}_{\delta_{ij} = \begin{cases} 1 & \text{se } i=j \\ 0 & \text{se } i \neq j \end{cases}} \\ &= \langle v_{k+1} | w_j \rangle - \langle v_{k+1} | w_j \rangle = 0, \quad \checkmark \end{aligned}$$

$$\bullet \text{Span}(w_1, \dots, w_{k+1}) = \text{Span}(v_1, \dots, v_{k+1})$$

$$\text{Span}(v_1, \dots, v_k, w_{k+1}) = \text{Span}(v_1, \dots, v_k, \tilde{w}_{k+1})$$

$$\overset{v_{k+1} + u}{\wedge}$$

$$\text{Span}(v_1, \dots, v_k)$$

✓

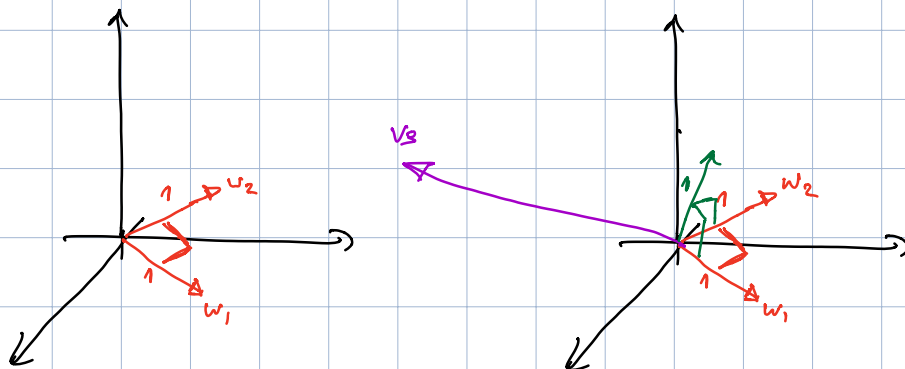


Oss: la matrice di cambio di base da $v_1, \dots, v_m$ a $w_1, \dots, w_m$ è

$$\begin{pmatrix} \langle w_1 | v_1 \rangle & * & * & & * \\ 0 & \langle w_2 | v_1 \rangle & * & & \vdots \\ \vdots & 0 & \langle w_3 | v_1 \rangle & & \vdots \\ 0 & 0 & 0 & \dots & 0 \end{pmatrix}$$

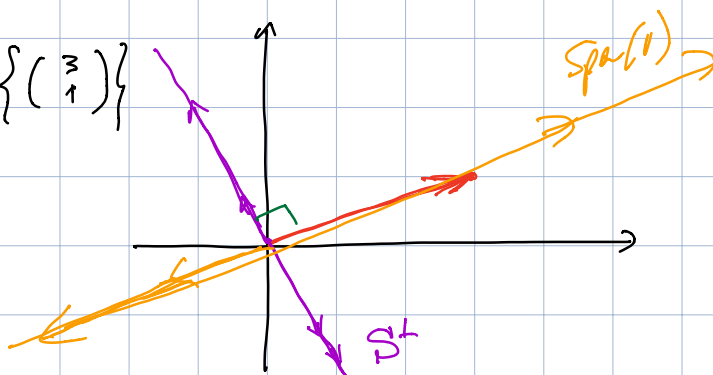
Con: dati $w_1, \dots, w_k$ base ortog. o ortonormale di $W \subset V$
 posso sempre completarli a base ortog. o ortonormale di V .

In fatti posso completare a caso e poi ortonormalizzare solo
 le parti dei vettori aggiunti.

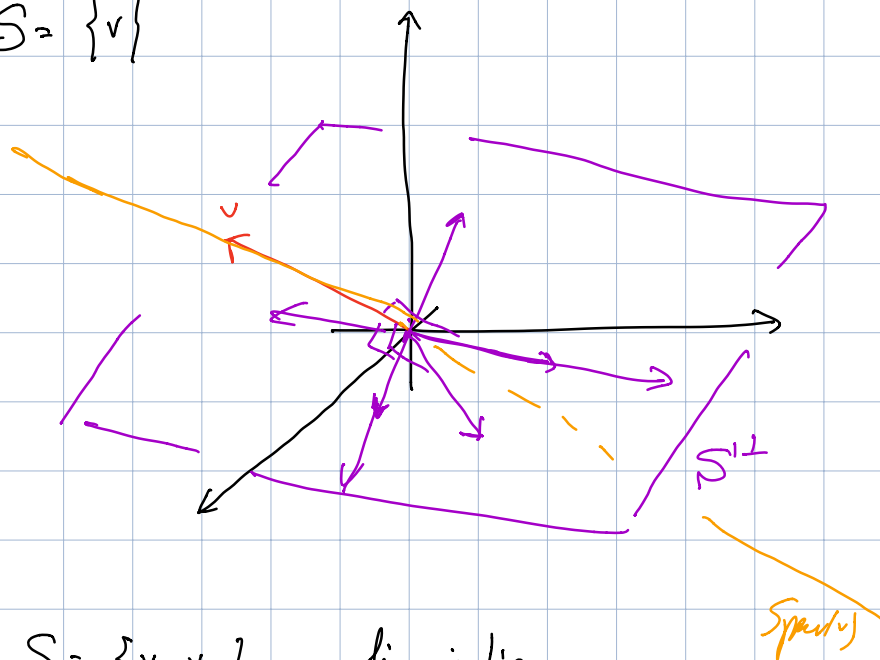


Def: dato $S \subset V$ chiamo ortogonale a S l'insieme:
 $S^\perp = \{v : \langle v | s \rangle = 0 \quad \forall s \in S\}.$

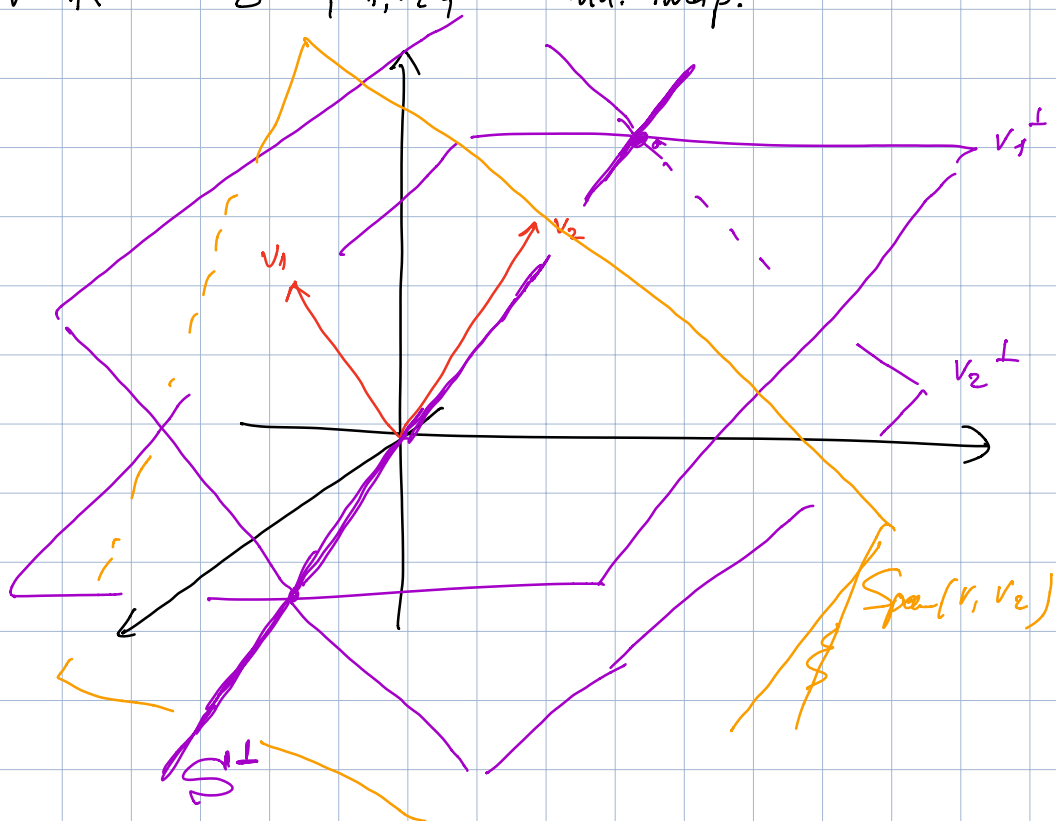
Es: $V = \mathbb{R}^2$, $S = \left\{ \begin{pmatrix} 3 \\ 1 \end{pmatrix} \right\}$



$$V = \mathbb{R}^3 \quad S = \{v\}$$



$$V = \mathbb{R}^2 \quad S = \{v_1, v_2\} \quad \text{lin. indep.}$$



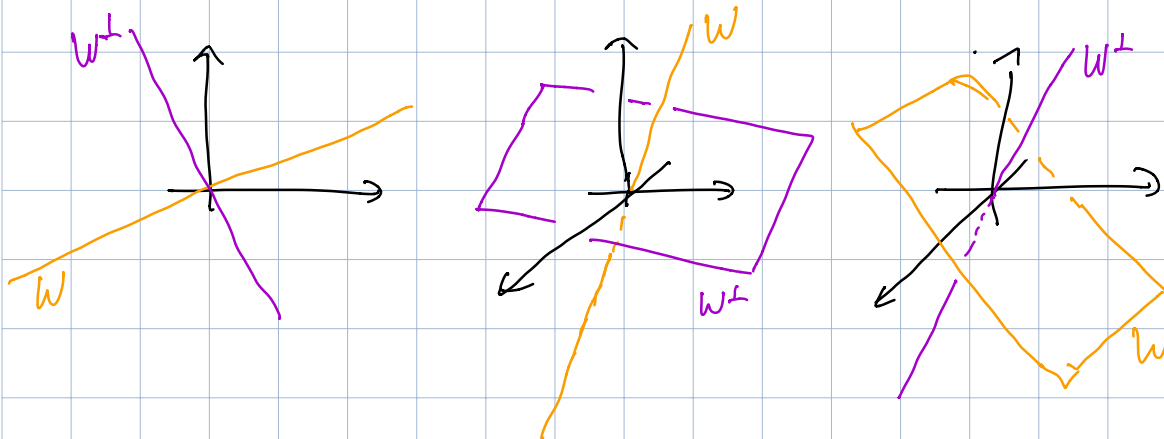
Oss: $S^\perp$ è un sottosp. vett. di V . Infatti:

$$v_1, v_2 \in S^\perp \quad \text{cioè} \quad \langle v_i, t \rangle = \langle v_i, t \rangle = 0 \quad \forall t \in S$$

$$\lambda_1, \lambda_2 \in \mathbb{R} \Rightarrow \langle \lambda_1 v_1 + \lambda_2 v_2, t \rangle = \lambda_1 \langle v_1, t \rangle + \lambda_2 \langle v_2, t \rangle = \lambda_1 \cdot 0 + \lambda_2 \cdot 0 = 0$$

Oss (esercizio): $S^\perp = (\text{Span}(S))^\perp$.

Prop: se $\dim_{\mathbb{R}}(V) < +\infty$ e W è sottosp. vett. di V
allora $V = W \oplus W^\perp$.



(Falso in $\dim = +\infty$).

Dimo: devo vedere che $W \cap W^\perp = \{0\}$ e $W + W^\perp = V$.

$$\begin{aligned} w \in W \cap W^\perp; \\ \langle w | w \rangle = 0 &\Rightarrow \|w\| = 0 \\ \uparrow \quad \cap \\ W &= W^\perp \Rightarrow w = 0 \\ &(\text{non vero } \dim < +\infty) \end{aligned}$$

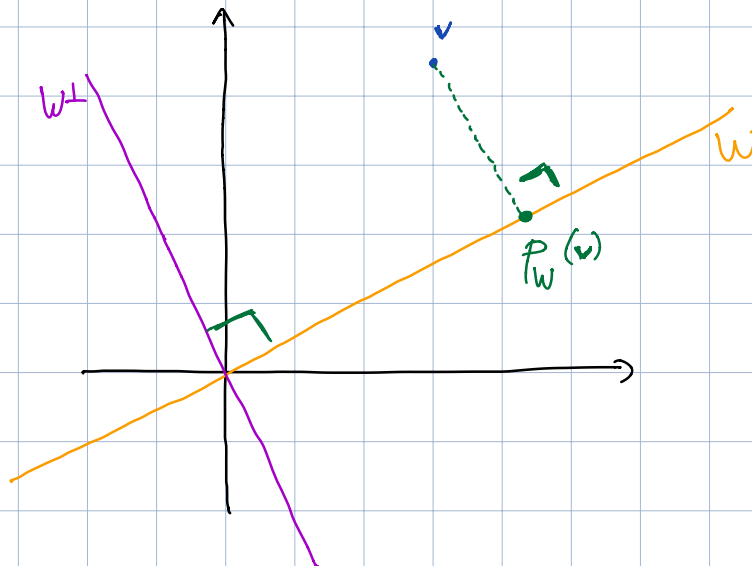
↓

Sia $W = \text{Span}(w_1, \dots, w_k)$
base ortonormale; completo a
base ortonormale $w_1, \dots, w_n$ di V .
Per $j > k$ ho $\langle w_j | w_i \rangle = 0$
 $i=1, \dots, k$

$$\Rightarrow w_j \in W^\perp; \text{ ora } u \in V$$

$$u = \underbrace{\alpha_1 w_1 + \dots + \alpha_k w_k}_{\substack{\cap \\ W}} + \underbrace{\alpha_{k+1} w_{k+1} + \dots + \alpha_n w_n}_{\substack{\cap \\ W^\perp}}$$

Def: la proiezione associata alla decomposizione $V = W \oplus W^\perp$ è detta proiezione ortogonale su W , detta P_W .



Oss (esercizio): Se $\dim(V) < +\infty$ e W è s.s.p. ho $(W^\perp)^\perp = W$.

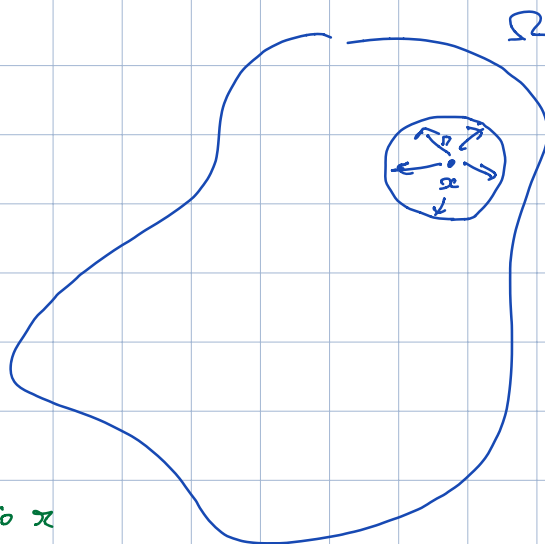
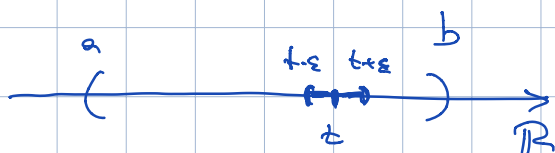
Calcolo differenziale in più variabili.

In una variabile si considerano $f: (a, b) \rightarrow \mathbb{R}$
 per calcolare $f'(t)$, $f''(t)$, ...

locus: intervallo aperto.

In n variabili analogamente $f: \Omega \rightarrow \mathbb{R}$ oppure $f: \Omega \rightarrow \mathbb{R}^k$
 $\Omega \subset \mathbb{R}^n$

con $\Omega \subset \mathbb{R}^n$ $\bar{\Omega}$ aperto:



$\forall x \in \Omega \exists r > 0$ t.c.
 $\{y \in \mathbb{R}^n: \|y-x\| < r\} \subset \Omega$

↑
 palla di centro x e raggio r

Il calcolo diff. in n variabili si fa in spazi $\subset \mathbb{R}^n$
 ma qui prendo solo $\Omega = \mathbb{R}^n$.

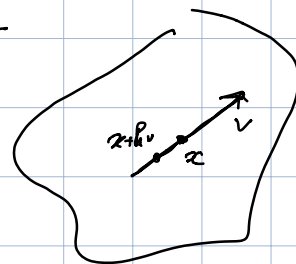
$$n=1 \quad f'(t) = \lim_{h \rightarrow 0} \frac{f(t+h) - f(t)}{h}$$

$n > 1$ • derivata in una direzione $v \in \mathbb{R}^n$, $v \neq 0$:

$$f: \mathbb{R}^n \rightarrow \mathbb{R}, \quad x \in \mathbb{R}^n$$

$$\frac{\partial f}{\partial v}(x) = \lim_{h \rightarrow 0} \frac{f(x+h \cdot v) - f(x)}{h}$$

In particolare ho derivata parziale di f in x
 nelle variabili x_i :



$$\frac{\partial f}{\partial x_i}(x) = \frac{\partial f}{\partial e_i}(x) = \lim_{h \rightarrow 0} \frac{f\left(\begin{pmatrix} x_1 \\ \vdots \\ x_i + h \\ \vdots \\ x_n \end{pmatrix}\right) - f(x)}{h}$$

oss: derivo rispetto a x_i considerando le altre variabili fissi.

Es: $f(x, y) = x^2 \cdot \sin(3x + 7y^2) + e^{5y}$

$$\frac{\partial f}{\partial x}(x, y) = 2x \cdot \sin(3x + 7y^2) + x^2 \cdot 3 \cdot \cos(3x + 7y^2)$$

$$\frac{\partial f}{\partial y}(x, y) = x^2 \cdot 14y \cdot \cos(3x + 7y^2) + 5 \cdot e^{5y}$$

Fatto: in ipotesi debole $\frac{\partial f}{\partial v}(x) = \sum_{i=1}^n \frac{\partial f}{\partial x_i}(x) \cdot v_i$.

Derivazione di composizione:

$$m=1 \quad (a, b) \xrightarrow{f} (c, d) \xrightarrow{g} \mathbb{R}$$

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

$$m > 1 \quad f: \mathbb{R}^m \rightarrow \mathbb{R}^k \quad f = \begin{pmatrix} f_1 \\ \vdots \\ f_k \end{pmatrix}$$

$$J_f(x) = \left(\frac{\partial f_i}{\partial x_j}(x) \right)_{\substack{i=1 \dots k \\ j=1 \dots m}} \in M_{k \times m}(\mathbb{R})$$

matrice jacobiana

$$\mathbb{R}^m \xrightarrow{f} \mathbb{R}^k \xrightarrow{g} \mathbb{R}^p$$

$$J(g \circ f)(x) = \underbrace{(Jg)(f(x))}_{p \times k} \cdot \underbrace{(Jf)(x)}_{k \times m}$$

$$\frac{\partial (g \circ f)}{\partial x_i}(x) = \sum_{j=1}^k \frac{\partial g}{\partial y_j}(f(x)) \cdot \frac{\partial f_j}{\partial x_i}(x)$$

————— 0 —————

Formula Taylor I ordine per $m=1$:

$$f(x+h) = f(x) + f'(x) \cdot h + \frac{1}{2} f''(x) \cdot h^2 + o(h^2)$$

trascurabile risp. ad h^2

x è di min. loc. se $f(x+h) \geq f(x) \quad \forall h \in (-\varepsilon, \varepsilon)$

- x min loc $\Rightarrow f'(x) = 0$, $f''(x) \geq 0$
- $f'(x) = 0$, $f''(x) > 0 \Rightarrow x$ min. loc.

Fissiamo $f: \mathbb{R}^m \rightarrow \mathbb{R}$. Chiamo gradiente di f in x

$$\text{grad}_x(f) = \left(\frac{\partial f}{\partial x_i}(x) \right)_{i=1, \dots, m} \in \mathbb{R}^m.$$

Poi si ossa $\frac{\partial^2 f}{\partial x_i \partial x_j}(x) = \frac{\partial}{\partial x_i} \left(\frac{\partial f}{\partial x_j} \right)(x)$

$$\left(\text{scrivo } \frac{\partial^2 f}{\partial x_i \partial x_i} = \frac{\partial^2 f}{\partial x_i^2} \right).$$

Chiamo matrice hessiana di f in x

$$H_x(f) = \left(\frac{\partial^2 f}{\partial x_i \partial x_j}(x) \right)_{i,j=1,\dots,n} \in M_{n \times n}(\mathbb{R}).$$

Teo: se f è decente $\frac{\partial^2 f}{\partial x_i \partial x_j} = \frac{\partial^2 f}{\partial x_j \partial x_i}$

cioè $H_x(f)$ è simmetrica.

Es: $f(x,y) = \cos(x^2 + 3y^3) + e^{x^5 \cdot y^3}$

$$\frac{\partial f}{\partial x} = -2x \cdot \sin(x^2 + 3y^3) + 5x^4 \cdot e^{x^5 \cdot y^3}$$

$$\frac{\partial f}{\partial y} = -24y^2 \cdot \sin(x^2 + 3y^3) + 3y^2 \cdot e^{x^5 \cdot y^3}$$

$$\frac{\partial^2 f}{\partial x^2} = -2 \cdot \sin(x) - 4x^2 \cdot \cos(x) + 20x^3 \cdot e^x + 25x^8 \cdot e^x$$

$$\frac{\partial^2 f}{\partial y \partial x} = -48xy^2 \cdot \cos(x) + 15x^4 \cdot y^2 \cdot e^x$$

$$\frac{\partial^2 f}{\partial x \partial y} = -48xy^2 \cdot \cos(x) + 15x^4 \cdot y^2 \cdot e^x$$

$$\frac{\partial^2 f}{\partial y^2} = \dots$$

Prop: se $w_1, \dots, w_k$ è base ortog. di $W \subset V$ stsp rett

$$p_W(v) = \sum_{i=1}^k \frac{\langle v | w_i \rangle}{\|w_i\|^2} \cdot w_i.$$

Dimo: possiamo completare $w_1, \dots, w_k$ a base stp. $w_1 \dots w_n$ di V ; inoltre $w_{k+1}, \dots, w_n \in W^\perp$. Ora:

$$v = \sum_{i=1}^n \frac{\langle v | w_i \rangle}{\|w_i\|^2} \cdot w_i = \underbrace{\sum_{i=1}^k \frac{\langle v | w_i \rangle}{\|w_i\|^2} \cdot w_i}_{\in W} + \underbrace{\sum_{i=k+1}^n \frac{\langle v | w_i \rangle}{\|w_i\|^2} \cdot w_i}_{\in W^\perp}$$

$$\Rightarrow p_W(v) = \sum_{i=1}^k \frac{\langle v | w_i \rangle}{\|w_i\|^2} \cdot w_i.$$

□

Prop: $p_W(v)$ è il punto di W più vicino a v .

