

• Modello Trinomiale:

$$B_n = B_{n-1} (1+r)$$

$$S_n = S_{n-1} (1+\mu_n) \quad , \quad n = 1, \dots, N$$

(VERSIONE INCOMPLETA)

$$1 + \mu_n \sim p_1 \delta_u + p_2 \delta_m + (1-p_1-p_2) \delta_d$$

$$\left(P(\mu_n = u-1) = p_1, \quad P(\mu_n = m-1) = p_2 \right.$$

$$\left. P(\mu_n = d-1) = 1-p_1-p_2 \right)$$

$(\mu_n)_n$ indipendenti

• Dato $X \in \mathbb{m}_{\mathcal{F}_N}$: come lo prezziamo?

$$H_n^Q = \mathbb{E}^Q \left[\frac{B_n}{B_N} X \mid \mathcal{F}_n \right]$$

↓

prezzo
risk-neutral

non introduce arbitraggi

per alcuna scelta di Q .

Copertura? short-fall risk minimization

$$\min_{(\alpha, \beta) \in A} \mathbb{E}^P \left[\underbrace{(X - V_n^{(\alpha, \beta)})^+}_{L_\alpha} \right]$$

$\left. \begin{cases} + & \text{se sottorepli chiama} \\ 0 & \text{se super-repli chiama} \end{cases} \right\}$

es. | $N = 2$, $S_0 = 1$

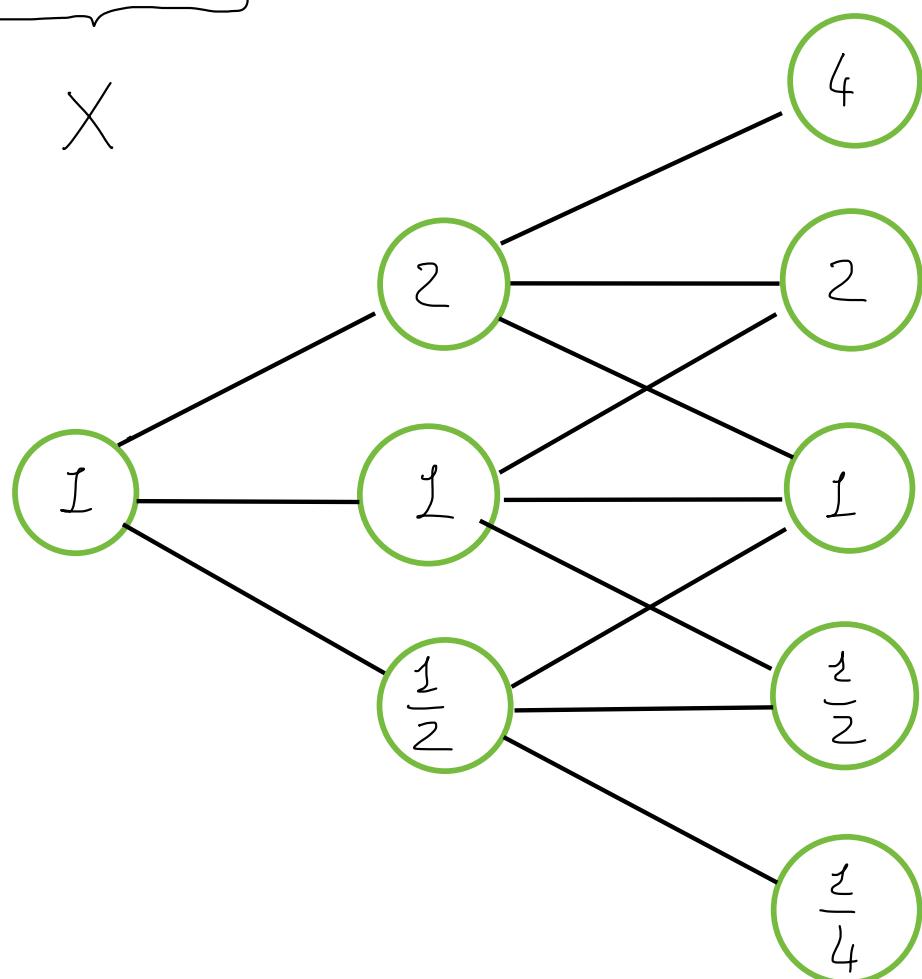
$$I + \mu_n = \begin{cases} 1/2 & \text{con } p_1 = 1/3 \\ 1 & \text{con } p_2 = 1/3 \\ 2 & \text{con } p_3 = 2/3 \end{cases} \quad (\text{indipend.})$$

$$\mathbb{P}(\mu_n = -\frac{1}{2}) = \mathbb{P}(\mu_n = 0)$$

$$= \mathbb{P}(\mu_n = 1) = \frac{1}{3}$$

$$r = 0$$

$$\underbrace{F(S_2)}_{X} = (S_2 - 1)^+$$



Risolvere: $\min_{(\alpha, \beta) \in A_{\geq 0}} \mathbb{E}^P \left[\underbrace{\left((S_{2-1})^+ - V_2^{(\alpha, \beta)} \right)^+}_{U(V_2^{(\alpha, \beta)}, S_2)} \right]$

$$\mathcal{U}(v, s) = ((s - I)^+ - v)^+$$

$$\mathcal{A}_{\geq 0} := \{(a, b) \in \mathcal{A} : V_n^{(a, b)} \geq 0, n = 0, 1, 2\}$$

- condizione sufficiente per $(a, b) \in \mathcal{A}_{\geq 0}$:

$$\begin{cases} V_0 \geq 0 \\ -\frac{V_{n-1}}{S_{n-1}} \leq a_n \leq \frac{2V_{n-1}}{S_{n-1}} \end{cases} \quad (*)$$

$$V_n = V_{n-1} + a_n S_{n-1} \mu_n$$

↑
autofinanziamento.

$$= V_{n-1} + \begin{cases} -\frac{a_n S_{n-1}}{2} & \text{se } \mu_n = -\frac{1}{2} \\ 0 & \text{se } \mu_n = 0 \\ a_n \cdot S_{n-1} & \text{se } \mu_n = 1 \end{cases}$$

$$\geq 0$$

- algoritmo di programm. dinamica:

$$W_2(v, s) = \mathcal{U}(v, s) = ((s - I)^+ - v)^+$$

$$W_{n-1}(v, s) = \min_{\alpha \in [-\frac{v}{s}, \frac{2v}{s}]} \mathbb{E}^P \left[\underbrace{W_n(v + s\alpha\mu_n, s(1 + \mu_n))}_{G_n(v, s, \mu_n; \alpha)} \right]$$

per $n = 1, 2$

- $n=2$:

$$W_1(v, 2) = \min_{\alpha \in [-\frac{v}{2}, v]} \mathbb{E}^P \left[U(v + 2\alpha\mu_2, 2(1+\mu_2)) \right]$$

$$\stackrel{?}{=} \min_{\alpha \in [-\frac{v}{2}, v]} \mathbb{E}^P \left[\left((2(1+\mu_2)-1)^+ - (v + 2\alpha\mu_2) \right)^+ \right]$$

$$\frac{1}{3} \left[(0 - (v - \underbrace{\alpha}_{0}))^+ + (1 - v)^+ \right.$$

$$\left. + (3 - v - 2\alpha)^+ \right]$$

$$\min_{\alpha \in [-\frac{v}{2}, v]} (-) = (3 - 3v)^+$$

per $\alpha = v$

$$\stackrel{?}{=} \min_{\alpha \in [-\frac{v}{2}, v]} \frac{1}{3}(1-v)^+ + \frac{1}{3}(3-v-2\alpha)^+$$

$$= \frac{1}{3}(1-v)^+ + (1-v)^+ = \frac{4}{3}(1-v)^+$$

$$\arg \min_{\alpha \in [-\frac{v}{2}, v]} = v \Rightarrow \alpha_2(v, z) = v$$

$$W_1(v, z) = \min_{\alpha \in [-v, 2v]} \mathbb{E}^P \left[W_2(v + \alpha \mu_2, \mu_2^+) \right]$$

" " " "

$$= \min_{\alpha \in [-v, 2v]} \mathbb{E}^P \left[(\mu_2^+ - (v + \alpha \mu_2))^+ \right]$$

" " " "

$$= \frac{1}{3} (2 - 3v)^+ \quad \text{con } \alpha = 2v$$

$$\Rightarrow \alpha_2(v, z) = 2v$$

$$W_1(v, \frac{z}{2}) = \min_{\alpha \in [-2v, 4v]} \mathbb{E}^P [\dots]$$

" " " "

$$= 0 \quad \forall \alpha \in [-2v, 4v]$$

$$\Rightarrow \alpha_2(v, \frac{z}{2}) = ? \text{ qualsiasi}$$

$$W_0(v, \lambda) = \min_{\alpha \in [-v, 2v]} \mathbb{E}^P \left[W_1 \left(v + \alpha \mu_1, \lambda + \mu_1 \right) \right]$$

//

$$\frac{1}{3} \left[W_1 \left(v - \frac{\alpha}{2}, \frac{\lambda}{2} \right) + W_1 \left(v, \lambda \right) + W_1 \left(v + \alpha, 2 \right) \right]$$

$$= \frac{5}{9} (1 - 3v)^+ \quad \text{con min in } \alpha = 2v$$

$$\Rightarrow \alpha_1(v, \lambda) = 2v$$

$V_2^\alpha \geq X$ IP-q.c.

oss: $v \geq \frac{2}{3} \Rightarrow W_0(v, \lambda) = 0$

$$= \min_{\alpha} \mathbb{E}^P \left[(X - V_2^\alpha)^+ \right]$$

↳ candidato per il prezzo di X in 0

• Procediamo in avanti:

- V_0 fissato (≥ 0)
- $\alpha_1 = \alpha_1(v, \lambda) = 2v$

$$- V_1 = V_0 + \alpha_1 S_0 \cdot \mu_1$$

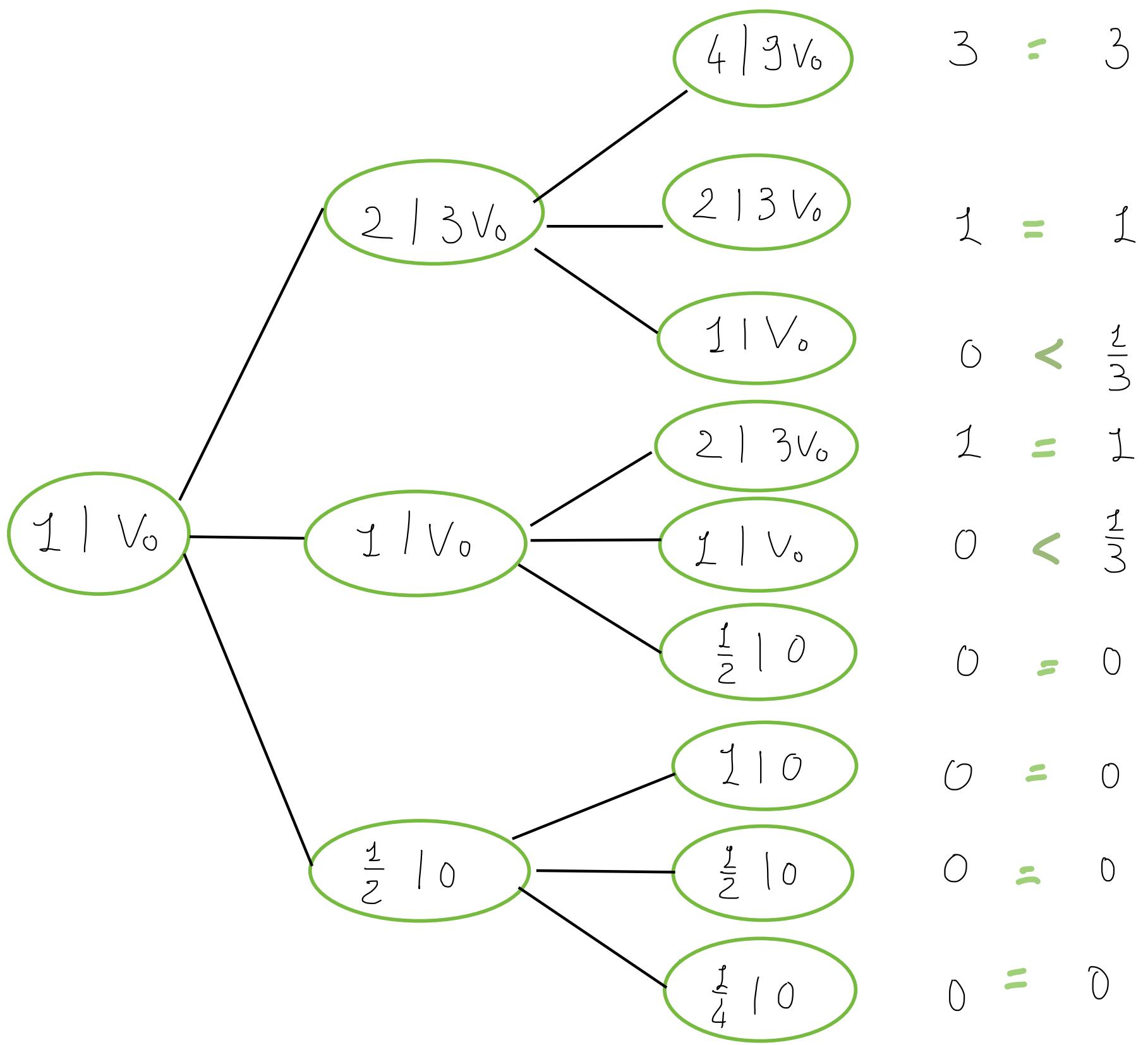
$$= V_0 + \begin{cases} -V_0 & \text{se } \mu_1 = -\frac{1}{2} \\ 0 & \text{se } \mu_1 = 0 \\ 2V_0 & \text{se } \mu_1 = 1 \end{cases}$$

$$- \alpha_2 = \alpha_2(V, S)$$

$$= \begin{cases} 0 & \text{se } S_1 = \frac{1}{2} \quad (\text{se } \mu_2 = -\frac{1}{2}) \\ 2V_0 & \text{se } S_1 = 1 \quad (\text{se } \mu_2 = 0) \\ V_0 & \text{se } S_1 = 2 \quad (\text{se } \mu_2 = 1) \end{cases}$$

$$- V_2 = V_1 + \dots$$

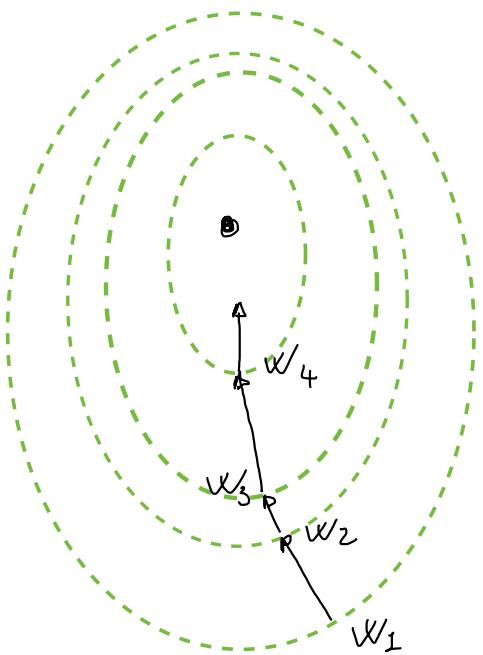
$\sqrt{2}$
 Payoff $(V_0 = \frac{1}{3})$



• Discesa stocastica del gradiente :

$$\begin{cases} w_1 = 0 \\ \\ w_{n+1} = w_n - \gamma \cdot \nabla f(w_n) , \quad n = 1, \dots, N-1 \end{cases}$$

fissato convessa



• 2 possibilità :

- $w_{\text{out}} := w_N$

- $w_{\text{out}} := \frac{1}{N} \sum_{n=1}^N w_n$

↳ nostra scelta

lemma

Sia $A \subset \mathbb{R}^d$ aperto e convesso, e

sia $f: A \rightarrow \mathbb{R}$ funzione. Allora

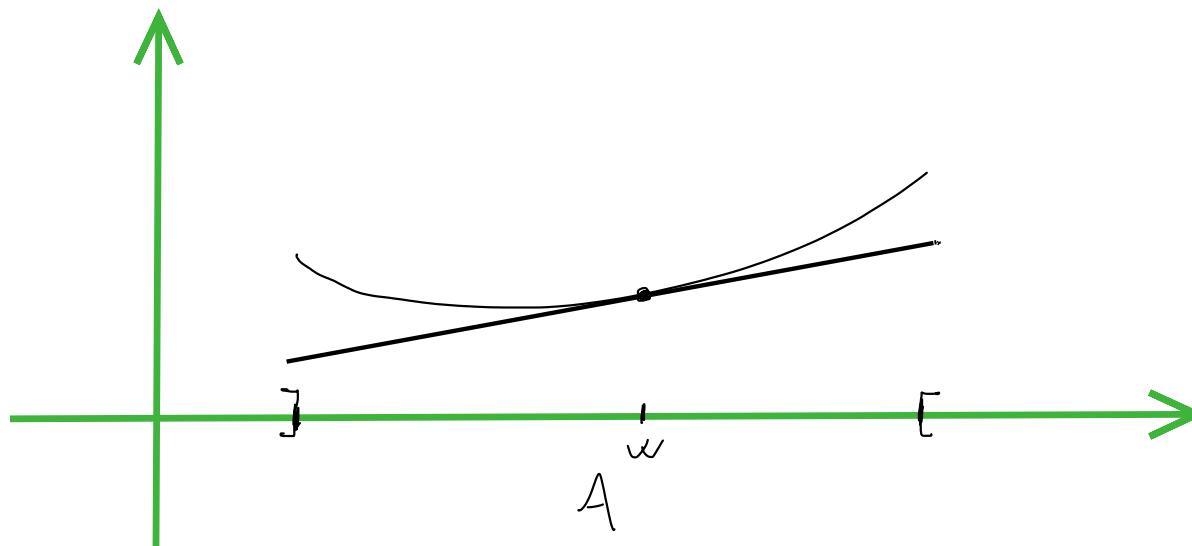
f è convessa \Leftrightarrow

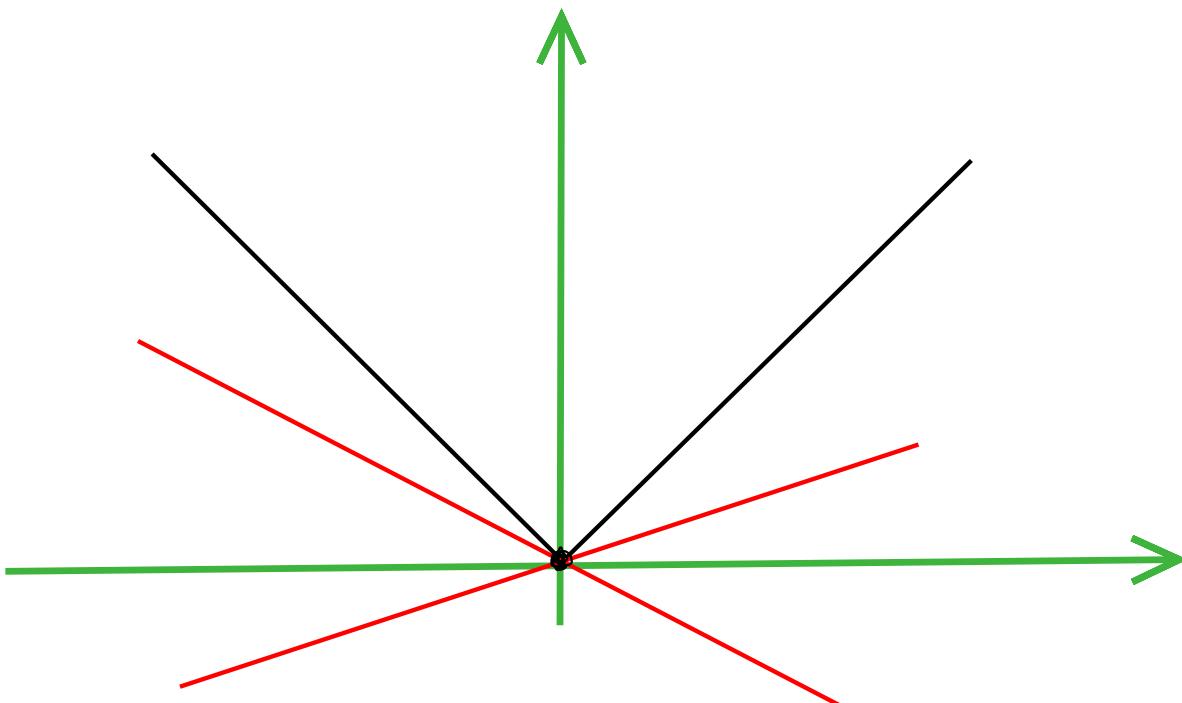
$\forall w \in A \quad \exists v = v(w) \in \mathbb{R}^d :$

$$f(u) \geq f(w) + \langle u - w, v \rangle \quad \forall u \in A$$

Inoltre, se f è differenziale, $v = v(w)$

è unico e $v(w) = \nabla f(w)$





dim] classica.

$$\begin{aligned}
 & f \text{ è differenziabile} \\
 & \frac{f(u) - f(w)}{u - w} \geq v > f'(w) \\
 \Rightarrow & \liminf_{u \rightarrow w} \frac{f(u) - f(w)}{u - w} > f'(w)
 \end{aligned}$$

ASSURDO

notazione] Dato $A \subset \mathbb{R}^d$, $\rho > 0$. Diciamo
che f è ρ -Lipschitziana se

$$\|f(x) - f(y)\| \leq \rho \|x - y\|, \quad x, y \in A$$

teorema] Sia A aperto e convesso. Sia

$f: A \rightarrow \mathbb{R}$ convessa, differenz e p -Lipschitziana.

Sia poi $w^* \in A$: $\|w^*\| \leq R$.

Ponendo $\gamma := \frac{R}{\rho \sqrt{n}}$, e assumendo che

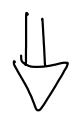
$w_n \in A \quad \forall n=1, \dots, N$, vale :

$$f(\bar{w}) - f(w^*) \leq \frac{R \rho}{\sqrt{N}}$$

$$\frac{1}{N} \sum_{n=1}^N w_n$$

corollario | Sotto le ipotesi del teorema
precedente,

$$w^* \in \arg \min_{\|w\| \leq R} f(w)$$



$$|f(\bar{w}) - f(w^*)| \leq \frac{R \rho}{\sqrt{N}}$$

dipende
da N

In particolare:

$$N \geq \frac{R^2 p^2}{\varepsilon^2} \Rightarrow |f(\bar{w}) - f(w^*)| \leq \varepsilon$$

Oss.

$$f(\bar{w}) \xrightarrow[N \rightarrow +\infty]{} f(w^*)$$

$$\text{non } \bar{w} \xrightarrow[N \rightarrow +\infty]{} w^*$$

es.

$$f(x, y) = x^2, \quad w_1 = (1, 1)$$

$$w^* = (0, 0)$$

$$\bar{w}_n \rightarrow (0, 1) \neq w^*$$

lemma

Sia $v_1, \dots, v_n \in \mathbb{R}^d$ sequenza arbitraria.

Sia $w^* \in \mathbb{R}^d$ vettore arbitrario.

Ponendo:

$$\begin{cases} w_1 = 0 \\ \vdots \\ w_{n+1} = w_n - \gamma v_n \end{cases}$$

vale:

(*)

$$\sum_{n=1}^N \langle v_n, w_n - w^* \rangle \leq \frac{\|w^*\|^2}{2\gamma} + \frac{\gamma}{2} \sum_{n=1}^N \|v_n\|^2$$

In particolare:

Se $\gamma = \frac{R}{\rho\sqrt{N}}$ con $R, \rho > 0$ tali che

$$\|v_n\| \leq \rho, \quad \|w^*\| \leq R$$

Allora:

$$\frac{1}{N} \sum_{n=1}^N \langle v_n, w_n - w^* \rangle \leq \frac{R\rho}{\sqrt{N}}$$

||

$$\frac{R^2}{2\gamma N} + \frac{N}{2N} \rho^2$$

dim | (lemma)

$$\langle w_n - w^*, v_n \rangle = \frac{1}{\gamma} \langle w_n - w^*, \gamma v_n \rangle$$

$$= \frac{1}{2\gamma} \left(-\underbrace{\|w_n - w^* - \gamma v_n\|^2}_{+ \|w_n - w^*\|^2 + \gamma^2 \|v_n\|^2} \right)$$

$$= \frac{1}{2\gamma} \left(-\|w_{n+1} - w^*\|^2 + \|w_n - w^*\|^2 \right)$$

$$+ \frac{\gamma}{2} \| v_n \|^2$$

$$\begin{aligned}
& \sum_{n=1}^N \langle w_n - w^*, v_n \rangle = \frac{1}{2\gamma} \left(\|w_1 - w^*\|^2 - \|w_{N+1} - w^*\|^2 \right) \\
& + \frac{\gamma}{2} \sum_{n=1}^N \|v_n\|^2 \\
& \leq \frac{1}{2\gamma} \|w_1 - w^*\|^2 + \frac{\gamma}{2} \sum_{n=1}^N \|v_n\|^2 \quad \# \\
& \quad \text{O}
\end{aligned}$$