

PROOF:

[7b]

Sia  $\pi = \{t_0=0, t_1, \dots, t_{N-1}, t_N=T\}$  una partizione di  $[0, T]$ .

Definiamo

$$f_k := f(W_{t_k}) \quad , \quad \forall k = 0, \dots, N$$

$$\Delta_k := W_{t_k} - W_{t_{k-1}}$$

$$\begin{aligned} \Rightarrow f(W_T) - f(W_0) &= f_N - f_0 = (f_1 - f_0) + (f_2 - f_1) + \dots + \\ &+ \dots + (f_N - f_{N-1}) = \sum_{k=1}^N (f_k - f_{k-1}) \end{aligned}$$

$$\stackrel{\textcircled{=}}{=} \sum_{k=1}^N f'_{k-1} \cdot \Delta_k + \frac{1}{2} \sum_{k=1}^N f''_{k-1} \Delta_k^2 + \frac{1}{2} \sum_{k=1}^N [f''(W_{t_k}^*) - f''_{k-1}] \Delta_k^2,$$

Sviluppo in serie di Taylor dove  $t_k^* \in [t_{k-1}, t_k]$

$$=: I_1(\pi) + I_2(\pi) + I_3(\pi).$$

OSS:

$$\lim_{|\pi| \rightarrow 0^+} I_1(\pi) = \int_0^T f'(W_s) dW_s \quad \rightarrow \text{in quanto } f' \text{ limitato}$$

per i punti e sfruttando la def. di integrale di Ito

$$\lim_{|\pi| \rightarrow 0^+} I_2(\pi) = \int_0^T f''(W_s) ds \quad \rightarrow \text{come nella dim. della}$$

variabile quadratica di  $W$ ,  
mostrò il fatto che  $\langle W_s \rangle = s$   
(Analogamente "basta minimizzare")

Passo da provare che

$$\lim_{|\pi| \rightarrow 0^+} I_3(\pi) = 0 \quad \text{in } L^2 \quad \rightarrow \lim_{|\pi| \rightarrow 0} E[I_3(\pi) - 0]^2 = 0$$

(anche quadraticamente convergente in  $L^2$ )

Allora:

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$$\mathbb{E} \left[ \left( \mathbb{I}_3(\pi) \right)^2 \right] = \mathbb{E} \left[ \left( \sum_{k=1}^N (f''(W_{t_k}^*) - f''_{k-1}) \Delta_k^2 \right)^2 \right]$$

$$\stackrel{\textcircled{=}}{=} \mathbb{E} \left[ \sum_{k=1}^N (f''(W_{t_k}^*) - f''_{k-1})^2 \Delta_k^4 \right] +$$

NOTA: per il  
 (Proprietà di  
 indipendenza  
 tra)

$$+ 2 \mathbb{E} \left[ \sum_{k_1 < k_2} (f''(W_{t_{k_1}}^*) - f''_{k_1-1}) \Delta_{k_1}^2 \cdot (f''(W_{t_{k_2}}^*) - f''_{k_2-1}) \Delta_{k_2}^2 \right]$$

$$= \sum_{k=1}^N \mathbb{E} \left[ (f''(W_{t_k}^*) - f''_{k-1})^2 \Delta_k^4 \right] +$$

$$+ 2 \sum_{\substack{k_1, k_2=1 \\ k_1 < k_2}}^N \mathbb{E} \left[ (f''(W_{t_{k_1}}^*) - f''_{k_1-1}) (f''(W_{t_{k_2}}^*) - f''_{k_2-1}) \Delta_{k_1}^2 \Delta_{k_2}^2 \right]$$

$$=: L_1(\pi) + 2L_2(\pi)$$

Per quel che riguarda  $L_1(\pi)$ :  $f \in C_b^2$ ,  $\mathbb{E}[\Delta_k^4] = 3(t_k - t_{k-1})^2$   
 $\forall t_k \in (t_{k-1}, t_k]$ ,  $|f''(W_{t_k}^*) - f''_{k-1}|^2 \leq C \sup |f''|^2$ ,  $C_1$  costante

$$\Rightarrow L_1(\pi) \leq \sum_{k=1}^N \mathbb{E} \left[ C \sup |f''|^2 \Delta_k^4 \right] = C \sup |f''|^2 \sum_{k=1}^N \mathbb{E}[\Delta_k^4]$$

$$\leq 3 \cdot C_1 \cdot \overbrace{\sup |f''|^2}^{< \infty, \text{ poiché } f \in C_b^2 \text{ (derivata limitata)}} \sum_{k=1}^N (t_k - t_{k-1})^2 \leq \bar{C} \cdot \sum_{k=1}^N (t_k - t_{k-1})^2 \quad \bar{C} = C_1 \cdot C_2$$

$$\leq \bar{C} \sum_{k=1}^N \underbrace{(t_k - t_{k-1})}_{\leq |\pi|} \leq \bar{C} |\pi| \sum_{k=1}^N (t_k - t_{k-1})$$

$$= \bar{C} |\pi| \cdot t \xrightarrow{|\pi| \rightarrow 0} 0$$

$$\Rightarrow L_1(\pi) \xrightarrow{|\pi| \rightarrow 0} 0$$

Per quel che riguarda  $L_2(\pi)$ :

Ricordiamo la DISUGUAGLIANZA DI HÖLDER

$\forall p, q > 0$  t.c.  $\frac{1}{p} + \frac{1}{q} = 1$ , si ha

$$\mathbb{E}[|X \cdot Y|] \leq (\mathbb{E}[|X|^p])^{\frac{1}{p}} \cdot (\mathbb{E}[|Y|^q])^{\frac{1}{q}} \quad (DH)$$

$$\Rightarrow L_2(\pi) = \sum_{\substack{h, k=1 \\ h < k}}^N \underbrace{\mathbb{E}[(f''(W_{t_k}^x) - f''_{k-1}) (f''(W_{t_h}^x) - f''_{h-1})]^2}_{|X|} \underbrace{\Delta_k^2 \Delta_h^2}_{|Y|}$$

$$\stackrel{(DH)}{\leq} \underbrace{\sum_{\substack{h, k=1 \\ h < k}}^N \mathbb{E}[(f''(W_{t_k}^x) - f''_{k-1})^2 (f''(W_{t_h}^x) - f''_{h-1})^2]}_{f \in C^2 \Rightarrow \exists \text{ una costante}} \left( \mathbb{E}[\Delta_k^4 \Delta_h^4] \right)^{\frac{1}{2}}$$

una costante  $A = A(\pi)$

$$\leq A \cdot \sum_{\substack{k, h=1 \\ h < k}}^N (\mathbb{E}[\Delta_k^4 \Delta_h^4])^{1/2} \leq A \cdot \sum_{h < k}^N (\mathbb{E}[\Delta_k^4])^{1/2} (\mathbb{E}[\Delta_h^4])^{1/2}$$

$$= A \cdot \sum_{\substack{h, k=1 \\ h < k}}^N (3(t_k - t_{k-1}))^{1/2} (3(t_h - t_{h-1}))^{1/2}$$

$$= 3 \cdot A \cdot \sum_{h < k}^N \underbrace{(t_k - t_{k-1})}_{\leq \pi} \underbrace{(t_h - t_{h-1})}_{|\pi|}$$

~~Stima~~, OSS:

$$L^2 = \sum_{h, k=1}^N (t_k - t_{k-1})^2 = \sum_{k=1}^N (t_k - t_{k-1})^2 + 2 \sum_{\substack{h, k=1 \\ h < k}}^N (t_k - t_{k-1})(t_h - t_{h-1})$$

$$\Rightarrow \sum_{\substack{h, k=1 \\ h < k}}^N (t_k - t_{k-1})(t_h - t_{h-1}) \leq \frac{t}{2} - \frac{t}{2} = \frac{t}{2} \leq \frac{t}{|\pi|}$$

Quindi:

$$L_2(\pi) \leq 3 \cdot A \cdot \frac{1}{|\pi|} \xrightarrow{|\pi| \rightarrow \infty} 0$$

$$\Rightarrow I_3(\pi) \xrightarrow{|\pi| \rightarrow \infty} 0 \text{ in } L^2$$



DEF.  $X, Y$  processi continui, quadrato-integrabili. [7#]  
 (martingale)

$$\langle X, Y \rangle := \frac{1}{2} [\langle X+Y \rangle - \langle X-Y \rangle]$$

↳ COVARIANZA

PROPRIETÀ:

$\forall X, Y, Z$ , si ha

- 1)  $\langle X, Y \rangle = \langle Y, X \rangle$
- 2)  $\langle \alpha X + \mu Y, Z \rangle = \alpha \langle X, Z \rangle + \mu \langle Y, Z \rangle, \forall \alpha, \mu \in \mathbb{R}$
- 3)  $|\langle X, Y \rangle|^2 \leq \langle X \rangle \cdot \langle Y \rangle$

LEMMA TRIITO ~~VERSIONE MULTIVARIATA~~ (VERSIONE MULTIVARIATA)

Sia  $X$  un processo di Ito ~~generale~~  $N$ -dimensionale:

$$dX_t^{(j)} = \mu_t^{(j)} dt + \sum_{k=1}^d \sigma_t^{jk} dW_t^{(k)}, \quad \forall j=1, \dots, N$$

Sia  $F = F(t, x) \in C^{1,2}(\mathbb{R} \times \mathbb{R}^N)$ . Allora,

$$dF = \frac{\partial F}{\partial t} dt + \nabla F \cdot dX_t + \frac{1}{2} \sum_{j,k=1}^N \frac{\partial^2 F}{\partial x_t^{(j)} \partial x_t^{(k)}} d\langle X_t^{(j)}, X_t^{(k)} \rangle$$

$\left( \frac{\partial F}{\partial t}, \dots, \frac{\partial F}{\partial x_t^{(2)}} \right)$  prodotto scalare

dove  $\nabla F := \left( \frac{\partial F}{\partial x_t^{(1)}}, \dots, \frac{\partial F}{\partial x_t^{(N)}} \right)$

$$\text{e } \nabla F \cdot dX_t = \sum_{k=1}^N \frac{\partial F}{\partial x_t^{(k)}} dX_t^{(k)}$$

PROP (Lemma di Ito - versione generale)

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Sia  $\{X_t\}_{t \geq 0}$  un processo di Ito e sia  $f = f(t, x) \in C^{1,2}(\mathbb{R}^+ \times \mathbb{R}^n)$ . Allora,  $Y_t = f(t, X_t)$  è un processo di Ito e si ha

$$dY_t = \frac{\partial f(t, X_t)}{\partial t} dt + \frac{\partial f(t, X_t)}{\partial X_t} dX_t + \frac{1}{2} \frac{\partial^2 f(t, X_t)}{\partial X_t^2} \langle dX_t, dX_t \rangle.$$

PROOF: Analogo al caso dei Moti Browniani 

NB:  $dX_t = \mu(t, X_t) dt + \sigma(t, X_t) dW_t$

$$\begin{aligned} \Rightarrow dY_t &= \frac{\partial f}{\partial t} dt + \frac{\partial f}{\partial X} \left( \mu(t, X_t) dt + \sigma(t, X_t) dW_t \right) + \\ &\quad + \frac{1}{2} \frac{\partial^2 f}{\partial X^2} \sigma^2(t, X_t) dt \\ &= \left[ \frac{\partial f}{\partial t} + \mu(t, X_t) \frac{\partial f}{\partial X} + \frac{1}{2} \frac{\partial^2 f}{\partial X^2} \sigma^2(t, X_t) \right] dt + \sigma(t, X_t) \frac{\partial f}{\partial X} dW_t \end{aligned}$$

ESEMPIO: MOTO BROWNIANO GEOMETRICO.

$$\frac{dS_t}{S_t} = \mu dt + \sigma dW_t, \quad \mu \in \mathbb{R}, \sigma \in \mathbb{R}^+$$

$\Rightarrow$  È possibile determinare una espressione esplicita di  $S_t$ ?  $\rightarrow$  SÌ! (Usando Ito's lemma)

$$Y_t = f(t, S_t) = X_t(S_t)$$

[7-41]

$$\Rightarrow \frac{\partial f}{\partial t} = 0; \quad \frac{\partial f}{\partial S_t} = \frac{1}{S_t}; \quad \frac{\partial^2 f}{\partial S_t^2} = -\frac{1}{S_t^2}$$

$$\stackrel{(170)}{\Rightarrow} dY_t = \frac{\partial f}{\partial t} dt + \frac{\partial f}{\partial S} dS_t + \frac{1}{2} \frac{\partial^2 f}{\partial S_t^2} \langle dS, dS_t \rangle$$

$$= \frac{1}{S_t} \left[ \mu S_t dt + \sigma S_t dW_t \right] + \frac{1}{2} \left( -\frac{1}{S_t^2} \right) \langle \mu S_t dt + \sigma S_t dW_t, \mu S_t dt + \sigma S_t dW_t \rangle$$

$$= \mu dt + \sigma dW_t - \frac{1}{2} \frac{1}{S_t^2} \sigma^2 S_t^2 dt$$

$$\Rightarrow dX_t(S_t) = \left( \mu - \frac{\sigma^2}{2} \right) dt + \sigma dW_t$$

$$\Rightarrow X_t(S_t) - X_t(S_0) = \int_0^t \left( \mu - \frac{\sigma^2}{2} \right) ds + \int_0^t \sigma dW_s$$

$$\Rightarrow X_t(S_t) = X_t(S_0) + \left( \mu - \frac{\sigma^2}{2} \right) t + \sigma \int_0^t dW_s$$

$\underbrace{\int_0^t dW_s}_{W_t}$  " VERIFICA PER ESERCIZIO!"

$$\Rightarrow S_t = S_0 \exp \left\{ \left( \mu - \frac{\sigma^2}{2} \right) t + \sigma W_t \right\}$$