

DTMC - SIMULATION!

WE START FROM A DTMC $D: (S, \Pi, P_0)$

PROBLEM: GENERATE / SAMPLE A TRAJECTORY

$$S_0 S_1 S_2 S_3 \dots$$

$$P(S_0 S_1 S_2 S_3 \dots) = p_0 \pi_{0,1} \pi_{1,2} \pi_{2,3} \dots$$

SIMULATION (D, N_{\max}) // output: $S_0 S_1 \dots S_{N_{\max}}$

$N \leftarrow 0$ // CURRENT STEP

$S \leftarrow \text{SAMPLE}(P)$ // CURRENT STATE

WHILE $N < N_{\max}$

$S \leftarrow \text{SAMPLE}(\Pi_{S,p})$

$N \leftarrow N+1$

Oss: Se S è finito,
SAMPLE è facile
da implementare!
(più o meno)

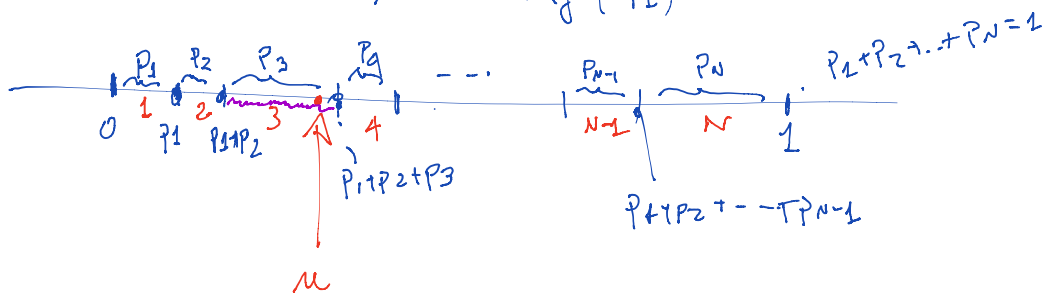
Sia $S = \{s_1, \dots, s_N\}$ e p una distrib. di prob. su S

$$p = (p_1 \dots p_N), \quad p_i \geq 0, \quad \sum_i p_i = 1$$

HO A DISPOSIZIONE UN PSEUDO RANDOM

NUMBER GENERATOR $\text{rand}()$

CHE RITORNA $u \sim \text{Unif}(0,1)$



SAMPLE (P) $P = (P_1, \dots, P_n)$

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u ← RAND() // Unif(0,1)
σ ← 0
i ← 0
WHILE σ < u
  i ← i + 1
  σ ← σ + P_i
RETURN i

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Expected # of comparisons C

$$E[C] = \sum_i i \cdot p_i = E[\bar{P}]$$

$$\vec{p} \quad 0.1 \quad 0.2 \quad 0.3 \quad 0.4 \quad E[\bar{P}] = 0.1 \cdot 1 + 0.2 \cdot 2 + 0.3 \cdot 3 + 0.4 \cdot 4 = 3$$

$$\vec{q} \quad 0.4 \quad 0.3 \quad 0.2 \quad 0.1 \quad E[\bar{Q}] = 0.4 \cdot 1 + 0.3 \cdot 2 + 0.2 \cdot 3 + 0.1 \cdot 4 = 2$$

ANALYSIS OF SIMULATION DATA

PROBLEMS TO SOLVE.

LET $X = (S, \Pi, P)$ A DTMC. LET $g: S \rightarrow \mathbb{R}$

LET $m > 0$ GIVEN

$$a) E[g(X_m)]$$

X_m : r.v. state of X at step m

state property

$$\gamma: S^m \rightarrow \mathbb{R}, \text{ for } m \text{ given}$$

trajectory property

$$b) E[\gamma(X_0, \dots, X_m)]$$

trajectory from time 0 to m , X_0, \dots, X_m

IF X HAS A STEADY STATE DISTRIBUTION π , THEN

$$c) E[g(X_\infty)] := E_\pi[g(S_1)]$$

SOLUTION: PRODUCE A SAMPLE OF a), b), OR c)
 BY SIMULATING X , AND ANALYZE THIS SAMPLE
 STATISTICALLY.

Focus on a), b)

a) b) SIM (X, g, m)

So... S_m ← SIMULATE (X, m)

RETURN $g(S_m)$
 $f(S_0 \dots S_m)$

STD CAMPIONANDO $Y \sim g(X_m) / f(X_0 \dots X_m)$

$y_1 \dots y_N$ OBSERVATIONS OF Y

PROBLEM: ESTIMATE $E[Y]$ from $y_1 \dots y_N$

SAMPLE MEAN ESTIMATOR

$$\bar{Y}_N = \frac{1}{N} \sum_{i=1}^N y_i$$

$y_i \sim Y_i \equiv Y, Y_1, \dots, Y_N$ iid

$\bar{Y}_N = \frac{1}{N} \sum_{i=1}^N Y_i$ IS A R.V.!
 INDEP + IDENTICALLY
 DISTRIBUTED

\bar{Y}_N IS CONSISTENT

if $\bar{Y}_N \rightarrow E[Y]$

UNBIASED

$$E[\bar{Y}_N] = E[Y]$$

$$E[\bar{Y}_N] = \frac{1}{N} \sum_{i=1}^N E[Y_i] = \frac{1}{N} \cdot N \cdot E[Y] = E[Y] \quad \underline{\text{UNBIASED}}$$

STRONG LAW OF LARGE NUMBER

$$\lim_{N \rightarrow \infty} \bar{Y}_N \rightarrow E[Y] \quad \text{a.s.} \quad \left[\begin{array}{l} Y_i, \bar{Y}_N : \Omega \rightarrow \mathbb{R} \\ \uparrow \\ \text{no prob } \tau \end{array} \right]$$

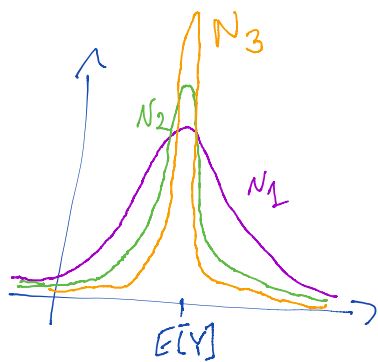
$$P\left(\left\{ \omega \mid \lim_{N \rightarrow \infty} \bar{Y}_N(\omega) = E[Y] \right\}\right) = 1$$

Fisso $\omega \in \Omega$

CENTRAL LIMIT THEOREM

$$\sqrt{N} \frac{\bar{Y}_N - E[Y]}{\sqrt{\text{VAR}[Y]}} \xrightarrow{\text{weakly}} \mathcal{N}(0, 1) \quad \left(\begin{array}{l} \text{gaussian} \\ \text{with zero} \\ \text{mean and} \\ \text{variance } 1 \end{array} \right)$$

$$\left[\bar{Y}_N \sim \mathcal{N}\left(E[Y], \frac{\text{VAR}[Y]}{N}\right) \right]$$



$$N_1 < N_2 < N_3 !$$

$$\text{VAR}[\bar{Y}_N] = \text{VAR}\left[\frac{1}{N} \sum_{i=1}^N Y_i\right] \stackrel{\text{indip.}}{=} \frac{1}{N^2} \sum \text{VAR}[Y_i] = \frac{\text{VAR}[Y]}{N}$$

SAMPLE VARIANCE

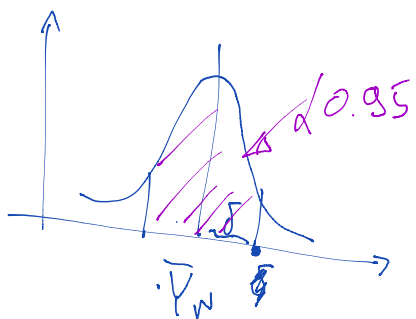
$$\bar{S}_N^2 = \frac{1}{N-1} \sum_{i=1}^N (Y_i - \bar{Y}_N)^2$$

fatti il conticino

$$= \frac{1}{N-1} \sum Y_i^2 - \frac{N}{N-1} \bar{Y}_N^2$$

$$E[\bar{S}_N^2] = \frac{N}{N-1} \left(E[Y_i^2] - E[\bar{Y}_N^2] + E[Y]^2 - E[Y]^2 \right)$$

$$= \frac{N}{N-1} \left(\text{VAR}[Y] - \underbrace{\text{VAR}[\bar{Y}_N]}_{\frac{\text{VAR}[Y]}{N}} \right) = \text{VAR}[Y]$$



$$W\left(\frac{\bar{Y}_N}{\sqrt{N}}, \frac{\bar{S}_N^2}{N}\right) = W_N$$

TROVA δ t.c.

$$P(W_N \in [\bar{Y}_N - \delta, \bar{Y}_N + \delta]) = \alpha \quad (\alpha = 0.95)$$

$$Z \sim W(0,1) \quad \phi(z) = P(Z \leq z) \quad \text{cdf}$$

$$P(|z| < z) = P(z < z) - P(z < -z) = \alpha$$

$$\Phi(z) - (1 - \Phi(z)) = \alpha$$

$$\Rightarrow z_{\frac{\alpha}{2}} = \Phi^{-1}\left(\frac{\alpha+1}{2}\right)$$

$$\delta = z_{\frac{\alpha}{2}} \cdot \text{STD}[\bar{Y}_N] = z_{\frac{\alpha}{2}} \cdot \frac{\sqrt{S_N^2}}{\sqrt{N}}$$

CONFIDENCE INTERVAL

$$I_c = \left[\bar{Y}_N - \frac{1}{\sqrt{N}} \cdot z_{\frac{\alpha}{2}} \cdot \sqrt{S_N^2}, \bar{Y}_N + \frac{1}{\sqrt{N}} \cdot z_{\frac{\alpha}{2}} \cdot \sqrt{S_N^2} \right]$$

It's a random quantity

$$P_2(\bar{Y}_N \in I_c) \approx \alpha \quad \left(\begin{array}{l} \text{per } N \text{ grande} \\ \text{cibastanza} \\ N \geq 30 \end{array} \right)$$

Nel caso b), tipically $f(s_0, \dots, s_m) \in \{0, 1\}$

In questo framework posso formalizzare REACHABILITY ABSORP.

