

SLLN.

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A sequence of rnd variables:

$$\{X_k, k=1,2,\dots\} \text{ i.i.d.}$$

Assume $\mu_X := \mathbb{E}[X_1] < \infty$.

Then,

$$\frac{X_1 + \dots + X_n}{n} \xrightarrow{n \rightarrow \infty} \mu_X$$

Also: If a function g is such that

$g(X_1)$ is well defined
and $\mathbb{E}[g(X_1)] < \infty$,

then

$$\frac{g(X_1) + \dots + g(X_n)}{n} \xrightarrow{n \rightarrow \infty} \mathbb{E}[g(X_1)]$$

Risk-neutral Pricing.

$V(T)$ payoff of a European derivative security

P^* ... risk-neutral probab. measure (we get from the replicating portfolio)

$$e^{-rT} \mathbb{E}^*[V(T)] = \underline{V(0)}$$

Price of the option.

Monte Carlo Pricing.

Recipe: • Create simulated stock-price paths.

↑
from the risk-neutral dist'n.

- Apply the payoff function to the simulated stock-price paths.

Get: the possible payoffs realized for the stock-price draws

$$\Rightarrow \frac{v_1 + v_2 + \dots + v_n}{n} =: \bar{v} \quad n \dots \# \text{ of simulations}$$

is "close to" the expected risk-neutral payoff.

- Finally:

$e^{-rT} \bar{v}$... is the Monte Carlo price.

To increase accuracy by a factor of η , we need to increase the number of variates by η^2 .

$$\text{Var}[\bar{v}] = \frac{\text{Var}[v_i]}{n} \quad \uparrow$$

The limiting behavior

Define $R(0, T) := \ln \left(\frac{S(T)}{S(0)} \right)$ realized returns

They will be modeled as normally distributed.

Assume: • Realized returns over disjoint time periods are independent.

• Realized returns over time intervals of equal length are identically distributed.

ADDITIVE