

Last time: Models for stock prices.

$$\hookrightarrow dS = \alpha S dt + \underbrace{\sigma S dW}_{\text{noisy fluctuations.}} \quad \text{B.M.}$$

① Probability Space (Ω, \mathcal{G}, P)

sample space
(non-empty set).

~~alg~~ σ -algebra

Represents events of which the probability is known.

$$\mathcal{G} = \{ \text{events } A \mid \uparrow \}. \quad \mathcal{G} \subseteq \mathcal{P}(\Omega).$$

$P \rightarrow$ Probability measure.

For every event A , $P(A) \in [0, 1]$.
represents the probability of A occurring.

(For every $A \in \mathcal{G}$, $P(A) \in [0, 1]$).

(1) $\mathcal{G} \rightarrow$ closed under complements. ($A \in \mathcal{G} \Rightarrow A^c \in \mathcal{G}$).

(2) \mathcal{G} is closed under countable unions.

(i.e. $\text{If } A_1, A_2, A_3, \dots \in \mathcal{G} \Rightarrow \bigcup_{i=1}^{\infty} A_i \in \mathcal{G}$).

(A σ -alg is a non empty collection of sets that satisfies (1) & (2).)

$$(3) \quad \emptyset \in \mathcal{G} \quad \& \quad \Omega \in \mathcal{G}$$

~~(4)~~

$$(4) \quad \text{If } A_1, \dots \in \mathcal{G} \text{ then } \bigcap_{i=1}^{\infty} A_i \in \mathcal{G}$$

$$(5) \quad \text{If } A, B \in \mathcal{G} \text{ then } A - B \in \mathcal{G}.$$

$$\left(\because A - B = A \cap B^c \right)$$

P is a Probability measure.

$$(1) \quad \text{For any } A \in \mathcal{G}, \quad P(A) \in [0, 1], \quad P(\emptyset) = 0 \quad \& \quad P(\Omega) = 1.$$

$$(2) \quad \text{If } A_1, A_2, \dots \in \mathcal{G} \text{ and are PAIRWISE DISJOINT} \\ (\text{i.e. } A_i \cap A_j = \emptyset \text{ when } i \neq j).$$

Then
$$P\left(\bigcup_{i=1}^{\infty} A_i\right) = \sum_{i=1}^{\infty} P(A_i).$$

① & ② is the definition of a probability measure.

③
$$P(A^c) = 1 - P(A)$$

④ If A_1, A_2, \dots is an increasing seq of evts.

$$A_1 \subseteq A_2 \subseteq A_3 \dots$$

Then
$$P\left(\bigcup_{i=1}^{\infty} A_i\right) = \lim_{i \rightarrow \infty} P(A_i).$$

④ ~~Proof~~ ⑤ If $A_1 \supseteq A_2 \supseteq A_3 \dots$ then

$$P\left(\bigcap_{i=1}^{\infty} A_i\right) = \lim_{i \rightarrow \infty} P(A_i)$$

Random Variables: A random variable is a fn $X: \Omega \rightarrow \mathbb{R}$.

Observe the event $\{X > 0\}$.

$$\hookrightarrow \{\omega \in \Omega \mid X(\omega) > 0\}$$

Can we compute $P(\{X > 0\})$? (= $P(X > 0)$).

Yes, provided $\{X > 0\} \in \mathcal{G}$

Require: for X to be a random variable.

then "all questions" about X will lead only to events in \mathcal{G} .

Eg: $\{X > \alpha\} \in \mathcal{G}$ for every $\alpha \in \mathbb{R}$.

OR $\{X \in [a, b)\} \in \mathcal{G}$ " " $a, b \in \mathbb{R}$.

OR $\{e^{X+10} < \sin(5X)\} \in \mathcal{G}$.

Def: We say X is a \mathcal{G} -measurable random variable.

if for every $\alpha \in \mathbb{R}$, $\{X \leq \alpha\} \in \mathcal{G}$.

Remark. If X & Y are 2 random variables.

Then $X+Y$, $X-Y$, $X+\alpha Y$, X^2 , $\frac{X}{Y}$
($\alpha \in \mathbb{R}$)
are all random variables.

Also, If f is any (non-constant) fn
(Eg $f(x) = x^2$,
 $f(x) = \ln(3x-5)$) & X is a R.V.

$f(X) = f \circ X$ is also a R.V.

Eg: Say $A \subseteq \Omega$.

Define $\mathbb{1}_A$ (\leftarrow fn with domain Ω & target \mathbb{R}).

$$\hookrightarrow \mathbb{1}_A \mapsto \mathbb{1}_A(\omega) = \begin{cases} 1 & \omega \in A \\ 0 & \omega \notin A. \end{cases}$$

(indicator fn of A).

Q: Is $\mathbb{1}_A$ a random variable?

$$\mathbb{1}_A \text{ is a RV} \iff A \in \mathcal{G}.$$

Eg 2: Say A_1, A_2, \dots, A_M are M events in \mathcal{G} .

Let $a_1, a_2, \dots, a_m \in \mathbb{R}$.

Define
$$X = \sum_{i=1}^M a_i \mathbb{1}_{A_i}$$

Such random variables are called SIMPLE RANDOM VARIABLES.

Expectations: The Expectation of a random variable is the "Lebesgue integral" (measures the "Expected reward").

If X is a RV,

Notation for Expectation of X is

$$EX = \int_{\Omega} X \, dP$$

Sup X is simple: $X = \sum_{i=1}^M a_i \mathbb{1}_{A_i}$

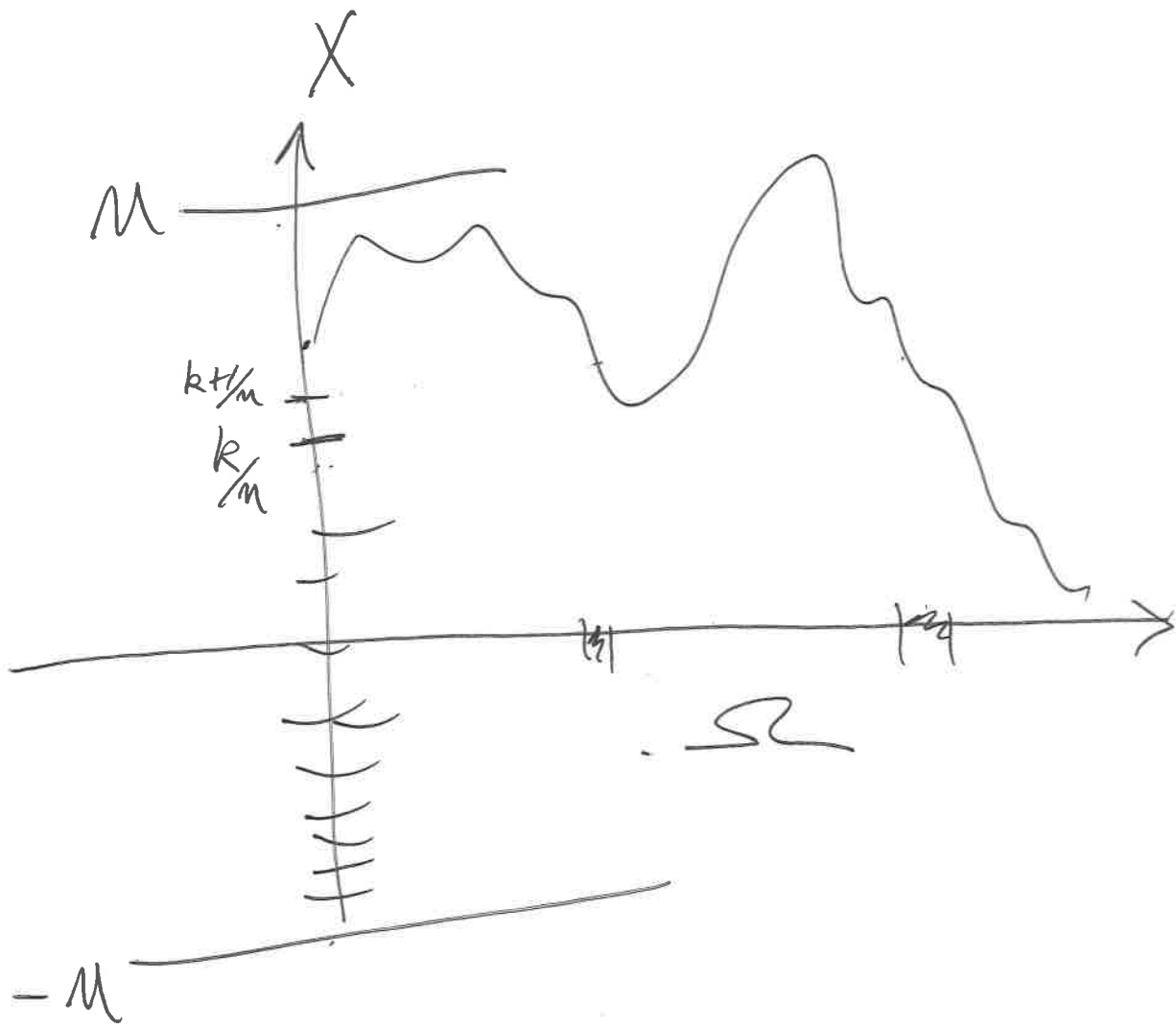
$$\text{Then } EX = \sum_{i=1}^M a_i P(A_i) = \sum_{i=1}^M a_i (P(X=a_i))$$

(if A_i 's are disjoint).

If X is not simple, approximate X by
simple RV's & take limits.

$$\text{Define } EX = \lim_{n \rightarrow \infty} E \left(\sum_{k=-n^2}^{n^2-1} \frac{k}{n} \mathbb{1}_{\left\{ \frac{k}{n} \leq X \leq \frac{k+1}{n} \right\}} \right)$$

↑
gives the Lebesgue integral.



To work with the expectations, going to use the following properties:

① linearity: If $\alpha \in \mathbb{R}$, X, Y are RV's.

$$\text{then } E(X + \alpha Y) = EX + \alpha EY.$$

② If $X \geq 0$ almost surely (i.e. $P(X \geq 0) = 1$).

$$\text{then } EX \geq 0.$$

Also if ^{additionally} $P(X > 0) > 0$, then $EX > 0$.

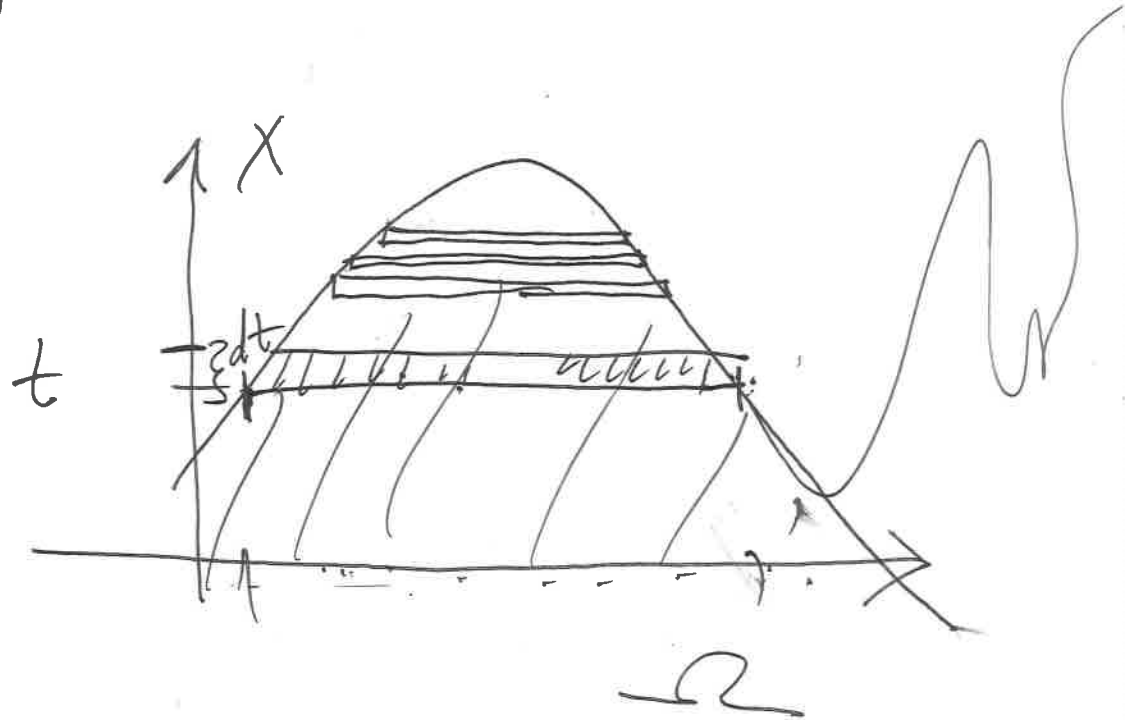
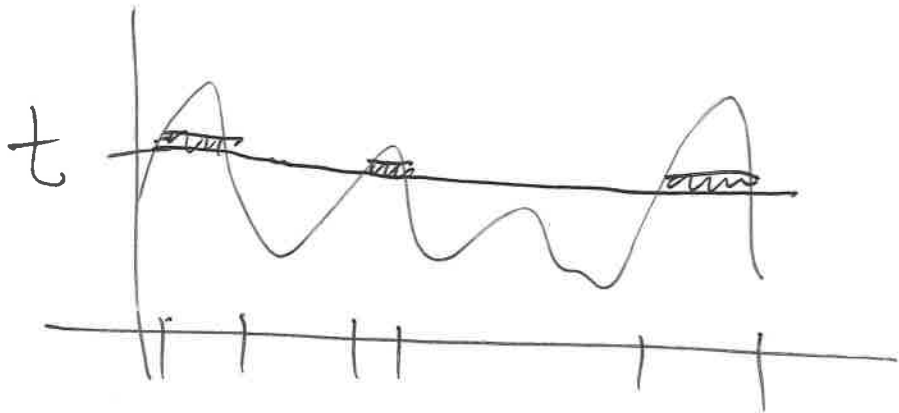
$$\Rightarrow \text{If } X \leq Y \text{ almost surely } \Rightarrow EX \leq EY$$

③ "layer cake formula"

If $X \geq 0$ almost surely.

then $EX = \int_0^\infty P(X \geq t) dt$

$\int_{\Omega} X dP$



More generally: φ is a non-negative increasing diff fun,

$$E \varphi(X) = \int_0^{\infty} \varphi'(t) P(X \geq t) dt$$

④ ~~Unconscious~~ ^{Lazy} Statistician Formula.

Say X has ~~PDF~~ PDF ϕ

& f is a non-random function.

$$EX = \int_{-\infty}^{\infty} x \phi(x) dx$$

$$E f(X) = \int_{-\infty}^{\infty} f(x) \phi(x) dx$$

① Independence

② Conditional Expectation.

Independence: ① A, B 2 events.

A is independent of B if $P(A \cap B) = P(A)P(B)$

$$(P(A|B) = P(A))$$

② X & Y 2 RV's.

(Every event obtained from a question about X)

is independent from every event obtained from a question about Y .

Eg: $\{X < \alpha\}$ is ind of $\{Y > \beta\}$
for every α, β .

$\sigma(X) \equiv$ σ -algebra generated by X

\equiv All events that can be observed from X .

$\sigma(X) =$ smallest σ -algebra that contains
the events $\{X \leq \alpha\}$ for every $\alpha \in \mathbb{R}$.

Can check $\sigma(X) \ni \{X \in (a, b)\}$

$\{e^X < \ln(5X)\}$.