

PhD seminar @ PhD Away Days 2019  
Université du Bordeaux, Bordeaux, France

## What is martingale?

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A Digression



## A digression - conditional expectation

In the following, let  $(\Omega, \mathcal{F}, \mathbb{P})$  be a probability space.

Let  $A$  and  $B$  be two events. Then the conditional probability is given by

$$\mathbb{P}(A \mid B) = ?.$$

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The **conditional expectation** of  $X$  given  $Y = y$  is

$$\mathbb{E}(X \mid Y = y) = \sum_x x \mathbb{P}(X = x \mid Y = y).$$

## Example

We flip a coin twice.  $X \triangleq \#\{\text{heads in 1 flip}\}$

$Y \triangleq \#\{\text{heads in 2 flips}\}$

Get for all possible values of  $Y$ :

$$\mathbb{E}(X \mid Y = 0) = 0, \quad \mathbb{E}(X \mid Y = 1) = \frac{1}{2}, \quad \mathbb{E}(X \mid Y = 2) = 1.$$



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Martingales



## Setting

$(\Omega, \mathcal{F}, (\mathcal{F}_n)_{n \geq 0}, \mathbb{P})$  **filtered probability space**, i.e. probability space and

$\mathcal{F}_n \subset \mathcal{F}$  sub  $\sigma$ -algebras,  $\mathcal{F}_n \uparrow$  events observable up to time  $n$

We gamble with fixed bet and winning 1.

$$\xi_j \in L^1 \text{ iid}$$

winning in the  $n^{\text{th}}$  draw

$$S_n = \xi_1 + \dots + \xi_n$$

overall winning after  $n$  draws

$$S_0 = 0$$

$$\mathcal{F}_n = \sigma(\xi_1, \dots, \xi_n)$$

information about all draws  $1, 2, \dots, n$

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Let's vary the game and allow variable bets, i.e.

$$e_{n+1} = e_{n+1}(\xi_1, \dots, \xi_n) \geq 0 \quad \text{bet in the } n + 1^{\text{th}} \text{ draw.}$$

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**Question** What is a martingale?

**(Heuristic) Answer** A martingale is the generalisation of the concept of a (fair) game.

### Definition

Let  $I$  be an index set (ordered) and  $(\mathcal{F}_t)_{t \in I}$  a filtration. A stochastic process  $(X_t)_{t \geq 0}$  satisfying,  $\forall t \geq 0$ ,

- (i)  $X_t$  is  $\mathcal{F}_t$ -adapted, i.e.  $X_t \in \mathcal{F}_t$  measurable;
- (ii)  $X_t \in L^1$ , i.e.  $\mathbb{E}|X| < \infty$ ;
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Let  $X_1, \dots, X_n$  be iid random variables such that

$$\mathbb{P}(X_i = 1) = \mathbb{P}(X_i = -1) = \frac{1}{2}.$$

We think of  $X_i$  describing the result of a coin flipping game:

- win 1€ if coin comes up heads
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**Consider** **doubling strategy**, i.e. keep doubling the bet until we eventually win.  
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$W_n$  only takes two possible values,  $W_n \in \{1, -2^n + 1\}$  ( $n \in \mathbb{N}$ ). **Why?**

(1) Suppose we win for the first time on the  $n^{\text{th}}$  bet. Then

$$\begin{aligned}W_n &= -(1 + 2 + \dots + 2^{n-2}) + 2^{n-1} \\ &= -(2^{n-1} - 1) + 2^{n-1} \\ &= 1\end{aligned}$$

(2) If we have not yet won after  $n$  bets, then

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**Show**  $\mathbb{E}(W_{n+1} \mid W_n) = W_n \stackrel{\text{iteration}}{\implies} \mathbb{E}(W_{n+m} \mid W_n) = W_n$

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(1)  $W_n = 1$ : then  $\mathbb{P}(W_{n+1} = 1 \mid W_n = 1) = 1$ , so

$$\mathbb{E}(W_{n+1} \mid W_n = 1) = 1 = W_n.$$

(2)  $W_n = -2^n + 1$ : bet  $2^n$  on  $(n+1)^{\text{th}}$  toss, so  $W_{n+1} \in \{1, -2^{n+1} + 1\}$ . Clearly,

$$\mathbb{P}(W_{n+1} = 1 \mid W_n = -2^n + 1) = \frac{1}{2}$$

$$\mathbb{P}(W_{n+1} = -2^{n+1} + 1 \mid W_n = -2^n + 1) = \frac{1}{2}$$

so that

$$\begin{aligned}\mathbb{E}(W_{n+1} \mid W_n = -2^n + 1) &= 1 \cdot \frac{1}{2} + (-2^{n+1} + 1) \cdot \frac{1}{2} \\ &= -2^n + 1 = W_n.\end{aligned}$$

Applications



- Doob's optional stopping theorem  $\rightarrow$  on average, nothing can be won by stopping play based on the information obtainable w/o looking into the future
- most essential continuous time process, Brownian Motion, is a martingale (zero drift stochastic process)
- martingales are essential to stochastic integration:  $\mathbb{E}B_t^2 = t$ , Doob-Meyer decomposition.
- solve PDEs by stochastic/martingale methods
- stochastic differential geometry, stochastic calculus on manifolds (and beyond)
- finance: martingality of an asset is equivalent to not being able to conduct arbitrage through trades in that asset
- ...

## Most prominent example

### Example (Brownian motion)

Let  $(M, g)$  be a smooth Riemannian manifold,  $B = (B_t)$  Brownian motion on  $M$  and  $\mathbf{A} = \frac{1}{2}\Delta_M$  with  $\Delta_M$  the Laplace-Beltrami operator on  $M$ , i.e.  $\Delta = \sum_{i=1}^n \partial_i^2$  on  $\mathbb{R}^n$ . Then,

$$\forall f \in C_c^\infty(\mathbb{R}^n), t \geq 0 : \quad f(B_t) - f(B_0) - \int_0^t \frac{1}{2} \Delta f(B_s) ds$$

is a martingale.

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## Question

If the weather forecast was a martingale,  
what would be the best prediction for tomorrow?



Merci pour votre attention !

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