

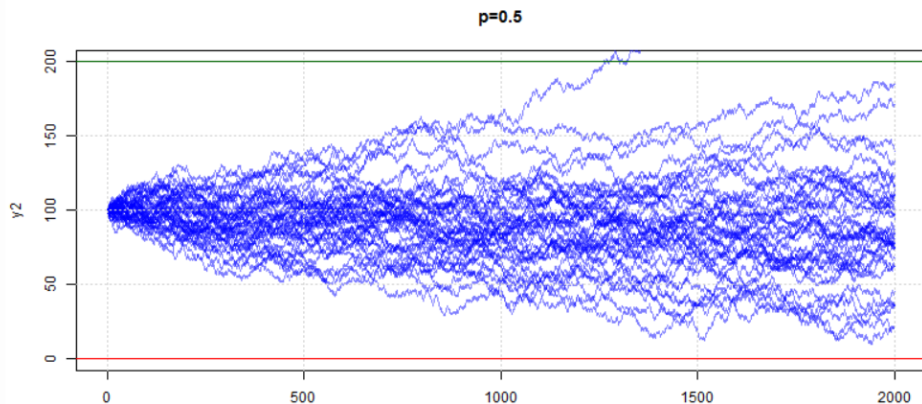
# Martingales and Applications

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# Gambler's Ruin

Lets play a game: You have a 50% chance of gaining \$1 and a 50% chance of losing \$1. Starting with  $K$  dollars, what are the odds you hit  $B$  before hitting 0?



# Martingales

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Let  $X = (X_1, X_2, \dots, X_t, \dots)$  be a sequence of real-valued random variables.

This history of the sequence  $X$  is captured by an ascending family of sigma algebra's  $\mathcal{F} = (\mathcal{F}_t)_{t \in I}$  called a filtration. We say that  $X$  is adapted to the filtration if  $\mathcal{F}_t$  contains the entire history of the process  $X$  up to time  $t$ .

## Definition (Martingale)

Let  $X$  be a sequence of random variables and  $\mathcal{F}$  a filtration. We say  $X$  is a martingale if

- $X$  is adapted to  $\mathcal{F}$
- $E(|X_i|) < \infty$  for  $(i = 1, 2, \dots)$
- $E[X_{t+s} | \mathcal{F}_t] = X_t$

# Gambler's Ruin is a Martingale

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For our coinflip game, let  $Y_n$  denote the outcome of the  $n$ th coinflip and  $X_n$  denote the total amount of money we have at time  $n$ . A heads is recorded as a 1 and a tails as a  $-1$ .

We can show that this is a martingale.

- $X$  is adapted to  $\mathcal{F}$
- $E[|X_t|] < \infty$  because  $E[X_t - X_0] \leq t$
- $E[X_{t+1}|\mathcal{F}_t] = X_t$

$$= E[X_t + Y_{t+1}|\mathcal{F}_t] = X_t + E[Y_{t+1}|\mathcal{F}_t] = X_t \text{ because } E[Y_{t+1}] = 1 * 1/2 + (-1) * 1/2 = 0$$

Note: This is because our flips are i.i.d

# Stopping times

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## Definition (Stopping time)

A stopping time  $\tau$  is an almost-surely finite time-valued random variable defined by some condition on the process.

## Lemma

*If  $\tau$  is a stopping time, then  $X_{t \wedge \tau}$  is a martingale. ( $t \wedge \tau$  means  $\min(t, \tau)$ )*

## Proof.

Intuitively, given that  $X_t$  is a martingale, then if  $t < \tau$ , we then our MTG  $X_t$  is still walking. However, if  $t \geq \tau$  then that indicates that our stopping time,  $\tau$  has been reached. If the stopping time has been reached, we know in the next step we are at our present position. □

# Optional Stopping Theorem

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## Theorem (Optional Stopping Theorem)

If  $(X, \mathcal{F})$  is a martingale,  $X$  is pointwise uniformly bounded, and  $\tau$  is a stopping time, then

$$\mathbb{E}[X_\tau] = \mathbb{E}[X_0] \quad (1)$$

## Proof.

$X_{t \wedge \tau}$  is a martingale. Therefore for any time  $t$

$$\mathbb{E}[X_{t \wedge \tau}] = \mathbb{E}[X_{0 \wedge \tau}] = \mathbb{E}[X_0]. \quad (2)$$

As  $t$  goes to infinity  $X_{t \wedge \tau}$  converges to  $X_\tau$ . Therefore by boundedness the result holds as  $t$  goes to infinity.  $\square$

# Returning to Gambler's Ruin



Returning to our game: You have a 50% chance of gaining \$1 and a 50% chance of losing \$1. Starting with  $K$  dollars, what are the odds you hit  $B$  before hitting 0?

Let  $\tau$  be the time it takes to hit 0 or hit  $B$ . Applying the optional stopping theorem, we have

$$\mathbb{E}[X_\tau] = \mathbb{E}[X_0] = K \quad (3)$$

$X_\tau$  is equal to  $B$  or 0 so

$$K = \mathbb{E}[X_\tau] = B \cdot \mathbb{P}[X_\tau = B] + 0\mathbb{P}[X_\tau = 0]. \quad (4)$$

Therefore

$$\mathbb{P}[X_\tau = B] = \frac{K}{B}. \quad (5)$$

# Brownian Motion

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Brownian motion is a continuous time analogue of a simple random walk.

It was discovered when R. Brown observed irregular motion of tiny particles.

## Definition (Brownian Motion)

A real-valued stochastic process  $W(\cdot)$  is called a Brownian motion if

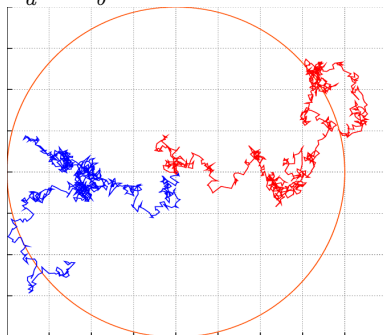
- $W(0) = 0$  almost surely
- $W(t) - W(s)$  is  $\mathcal{N}(0, t - s)$  for all  $t \geq s \geq 0$
- For all times  $0 < t_1 < \dots < t_n$ , the random variables  $W(t_1), W(t_2) - W(t_1), \dots, W(t_n) - W(t_{n-1})$  are independent
- $W(t)$  is almost surely continuous in time  $t$



# Introduction to the Problem

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Let  $X_t, Y_t$  be two independent Brownian motions starting at 0. Let  $T$  be the first time that  $(X_t, Y_t)$  lies on the ellipse  $\frac{X^2}{a^2} + \frac{Y^2}{b^2} = 1$ .



# Solution

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## Lemma

If  $W$  is a Brownian motion, then  $W(t)^2 - t$  is a martingale.

## Proof.

Let  $M(t) = W(t)^2 - t$ . Then

$$\mathbb{E}[M(t+s)|\mathcal{F}_t] = \mathbb{E}[(W(t+s) - W_t + W(t))^2 - (t+s)|\mathcal{F}_t] \quad (6)$$

Expanding the square gives

$$\mathbb{E}[(W(t+s) - W(t))^2 + 2(W(t+s) - W(t))W(t) + W(t)^2 - t - s|\mathcal{F}_t] \quad (7)$$

As  $W(t+s) - W(t)$  is independent to the history of the process up to time  $t$  we have

$$\mathbb{E}[(W(t+s) - W(t))^2] + W(t)^2 - t - s = W(t)^2 - t. \quad (8)$$

□

## Solution (2)

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Let  $M(t) = \frac{X_t^2 - t}{a^2} + \frac{Y_t^2 - t}{b^2}$ . Let  $\tau$  be defined as

$$\inf\{t : \frac{X_t^2}{a^2} + \frac{Y_t^2}{b^2} = 1\}. \quad (9)$$

$M(t)$  is a martingale and  $\tau$  is a stopping time so  $M(t \wedge \tau)$  is a martingale. Applying the optional stopping theorem we get that

$$0 = \mathbb{E}[M(0)] = \mathbb{E}[M(\tau)] = \mathbb{E}\left[\frac{X_\tau^2 - \tau}{a^2} + \frac{Y_\tau^2 - \tau}{b^2}\right]. \quad (10)$$

Breaking up the right side:

$$\mathbb{E}\left[\frac{X_\tau^2}{a^2} + \frac{Y_\tau^2}{b^2} - \frac{\tau}{a^2} - \frac{\tau}{b^2}\right] = 1 - \frac{\mathbb{E}[\tau]}{a^2} - \frac{\mathbb{E}[\tau]}{b^2} \quad (11)$$

So we're left with

$$E[\tau] = \frac{1}{\frac{1}{a^2} + \frac{1}{b^2}} = \frac{a^2 b^2}{a^2 + b^2} \quad (12)$$

# Conclusion

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We looked at Martingales, Brownian Motion, and the Optional Stopping Theorem

## Definitions (Recap)

- For Martingales:  $E[X_{t+s}|\mathcal{F}_t] = X_t$
- For Brownian Motion:  $W(t) - W(s)$  is  $\mathcal{N}(0, t - s)$  for all  $t \geq s \geq 0$
- For Optional Stopping Theorem:  $E[X_0] = E[X_\tau]$

These ideas help us compute and understand recursive events directly, and can be extended to games that are not fair.

Fin

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*Thank You for listening, Any Questions?*