

# Stochastic Processes

Sam Qin

# Review: Random Variables

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## Expected Value:

$E[X] = \sum x \cdot p_x$  for discrete  $X$

$p_x$  denotes the probability  
of  $x$  occurring.

## Variance:

$$\text{Var}[X] = E[(X - E[X])^2]$$

$$\text{Var}[X] = E[X^2] - E[X]^2$$

# Stochastic Processes

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## Definition:

Sequence of random variables indexed by a set (usually time)

Often denoted  $\{X_N\}$  for a random variable  $X$  and indices  $N$  in a set  $T$ .

## Examples:

- repeated coin tosses
- brownian motion
- random walks

# Examples of Stochastic Process problems

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- Expected number of coin flips until you get two heads in a row?
- Repeatedly roll and add 2 standard die. Probability you roll a 12 before you roll two 7's in a row?
- Random walk on the number line. Start at 1, and stop if you hit 0 or 5. Probability you stop at 5?

# Markov Chain

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## **Markov Property:**

Future events occur  
independent of previous  
states

Aka, “Memorylessness”

## **Definition:**

A Markov chain is a  
stochastic process which  
has the Markov Property

# Modeling with Markov Chains

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## **Assumptions for simplicity:**

We have perfect information

There are a finite number  
of states

## **States in a Markov Chain:**

Each state is a “position”

Markov states transition  
between each other with  
some probabilities

# Modeling coin flips

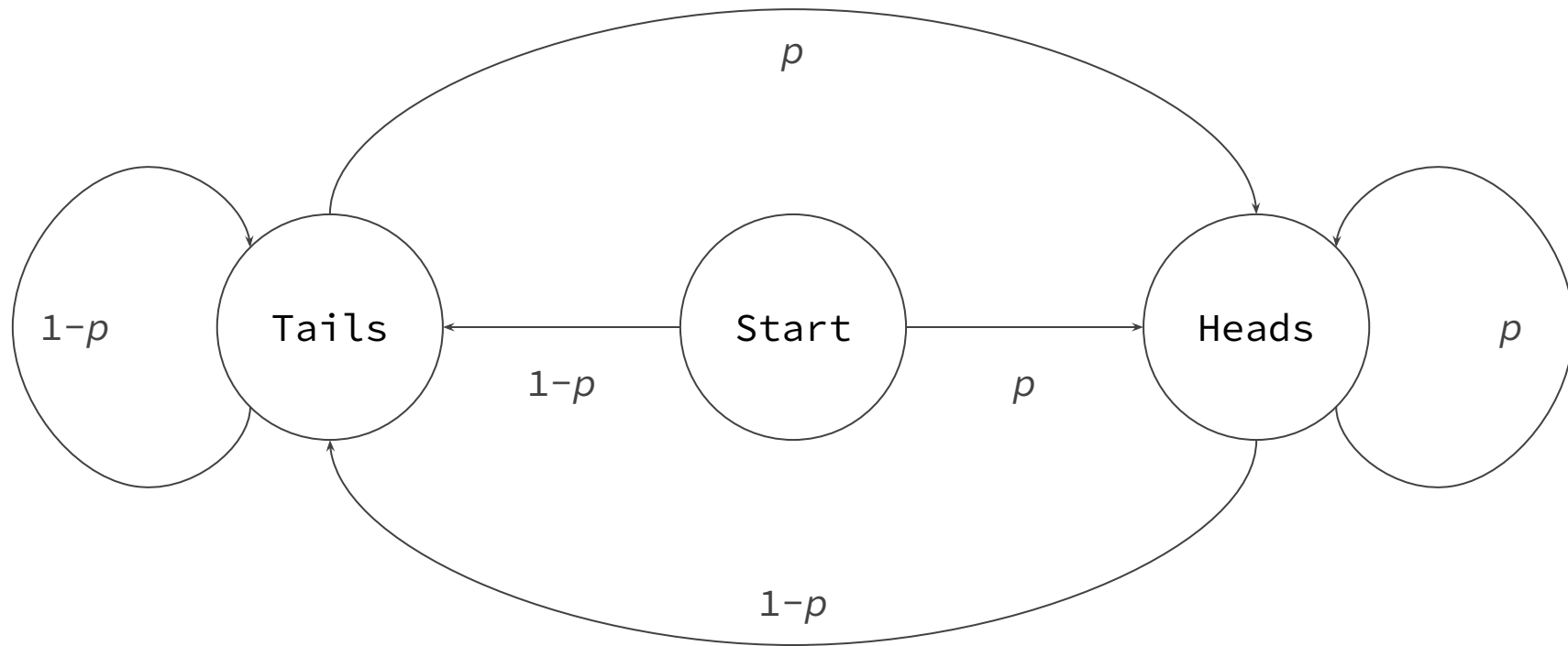
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How do we model a sequence of coin flips?

Assume the coin has probability  $p$  of landing heads and  $(1 - p)$  of landing tails.

## **States:**

- Starting
- Previous flip H
- Previous flip T



**Markov Chain representation of a coin flip sequence**



# Markov Chains as equations

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## Notation:

Let  $S_K$  denote the state  $K$ .

We have the following states:

$S_S \rightarrow \text{Start}$

$S_H \rightarrow \text{Heads}$

$S_T \rightarrow \text{Tails}$

## State Transitions:

$$S_S = 0$$

$$S_H = p \cdot S_H + p \cdot S_T$$

$$S_T = (1-p)S_H + (1-p)S_T$$

# Stopping Criteria

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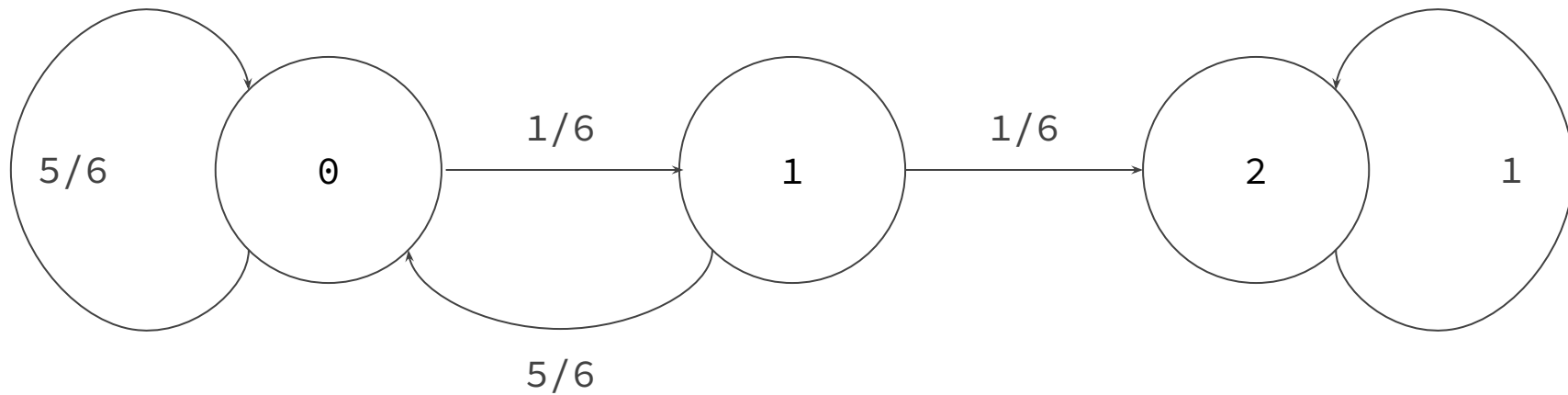
In the previous example, we have no stopping criteria, so the chain loops endlessly.

How do we represent stopping criteria?

## **Problem:**

Roll a die until you roll two 6's in a row.

What is the expected number of rolls?



**Markov Chain Representation**

# Absorption States

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- The state of two 6's in a row is an “absorption state”
- We stop the process once we reach this state.
- Usually represented by a loop edge with probability 1.

# Markov Chains for Expected Value

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## Notation:

Let  $E_T$  be the expected number of moves from state  $T$  to stopping.

We have the states  $E_0$ ,  $E_1$ ,  $E_2$ .

## Equations:

$$E_0 = 1 + \frac{5}{6}E_0 + \frac{1}{6}E_1$$

$$E_1 = 1 + \frac{5}{6}E_0 + \frac{1}{6}E_2$$

$$E_2 = 0$$

Recall  $E_2$  is our stopping criteria.

# Random Walks

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Particle on a grid which randomly moves one unit in a given direction.

A wide variety of problems arise.

## **Example:**

Consider the random walk with given conditions:

- Start at: 1
- Stop at 0, 10

What is the probability the particle stops at 10?

Expected number of steps?

# Markov Chain Approach

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## General Overview:

Create positional states:

0, 1, 2, ..., 10

Write out the system of equations for probability and for expected number of steps.

## Probabilistic Equations:

$$S_0 = 0, S_{10} = 1$$

$$S_i = \frac{1}{2}S_{i-1} + \frac{1}{2}S_{i+1}$$

## Expected Value Equations:

$$S_0 = 0, S_{10} = 0$$

$$S_i = \frac{1}{2}S_{i-1} + \frac{1}{2}S_{i+1} + 1$$

# Martingales

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## Definition:

A stochastic process which satisfies the conditional expected value

$$E[X_{t+1} | X_1, \dots, X_t] = X_t.$$

## Relation to Random Walks:

Random Walks are martingales.

$$X_{t+1} = X_t \pm 1.$$



# Martingale Trickery

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Why is this useful?

Let  $N$  be the last step, so

$X_N = 0$  or  $10$ .

What is  $E[X_N]$ ?

$X_N$  is a martingale process,  
so  $E[X_N] = X_0 = 1$ .

Let  $p_{10}$  be the probability  
 $X_N$  stops at  $10$ . Then,

$$E[X_N] = 10p_{10} + 0(1 - p_{10}) = 1$$

Solving,  $p_{10} = 1/10 = 0.1$

# Martingale Trickery, Expected Value

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## Variance of Random Walk:

Consider  $\text{Var}[X_t]$  now.

$$\text{Var}[X_t] = E[X_t^2] - E[X_t]^2.$$

Let  $X_t = \sum e_i$ , where  
 $e_i = \text{Bern}(-1, 1)$ .

$$E[X_t^2] = \sum e_i^2 = t$$

$$E[X_t^2] = t, \text{ so}$$

$$E[X_t^2 - t] = 0.$$

$$E[X_{t-1}^2 - (t-1)] = 0,$$

so the process

$V_t = X_t^2 - t$  is a  
Martingale.

# More trickery

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## End States:

Let  $N$  be the last state.

$$E[X_N^2 - N] = E[X_N^2] - N$$

By martingale properties,

$$E[X_N^2] - N = E[X_0^2] - 0, \text{ so}$$

$$N = E[X_N^2] - E[X_0^2]$$

## Solving:

Let  $p_{10}$  be the probability of stopping at 10. We know  $p_{10} = 10$

$$E[X_N^2] = 10^2 p_{10} + 0^2 (1 - p_{10})$$

$$N = 100 p_{10} - 1.$$

$$N = 9$$

# Summary

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- Random variables: expectation, variance
- Stochastic processes are an indexed sequence of random variables
- A Markov process is a “memoryless” process
- Martingales properties simplify random walks

Questions?

# Future continuations

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- Bertrand's Ballot Theorem
- Catalan Numbers