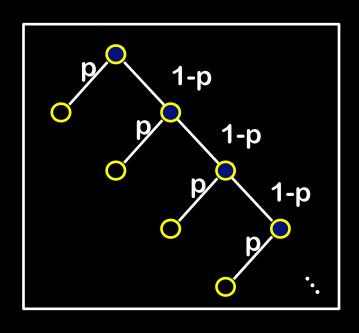
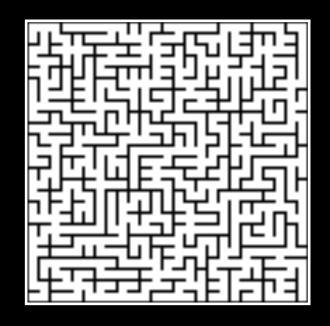
15-251

Great Theoretical Ideas in Computer Science

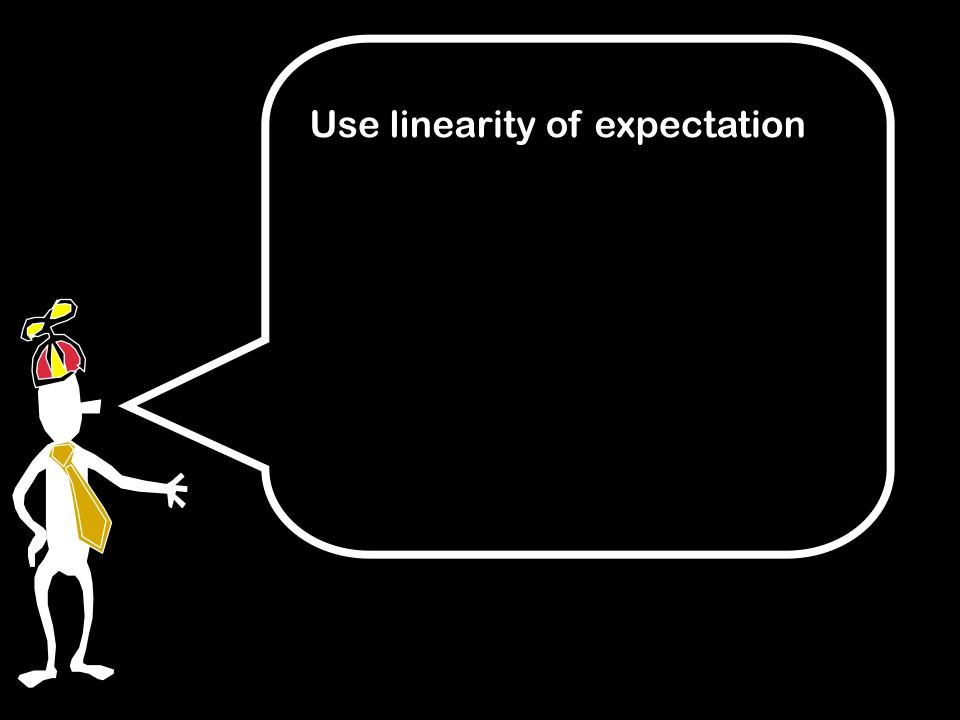
Infinite Sample Spaces and Random Walks

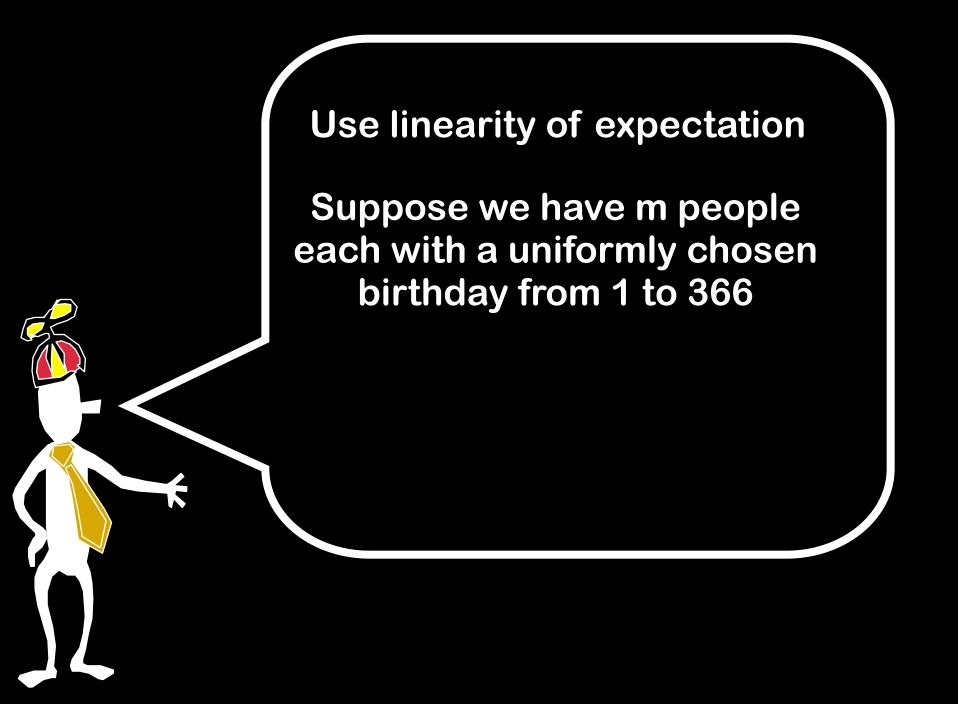
Lecture 11, September 30, 2008





See handout for probability review and some new stuff

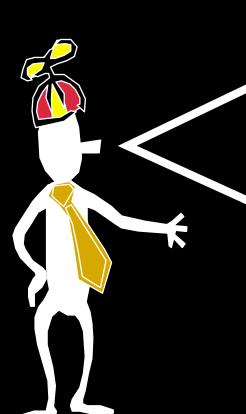




Use linearity of expectation

Suppose we have m people each with a uniformly chosen birthday from 1 to 366

X = number of pairs of people with the same birthday

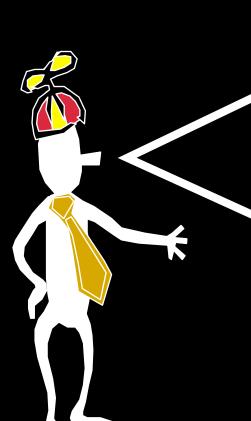


Use linearity of expectation

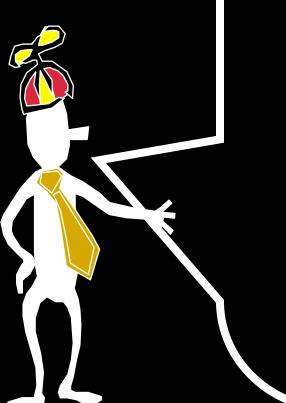
Suppose we have m people each with a uniformly chosen birthday from 1 to 366

X = number of pairs of people with the same birthday

E[X] = ?

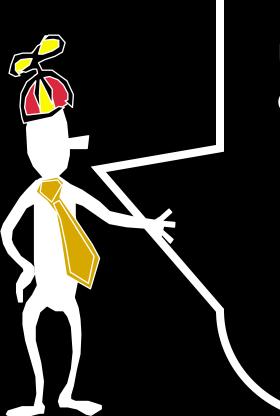


$$E[X] = ?$$



$$E[X] = ?$$

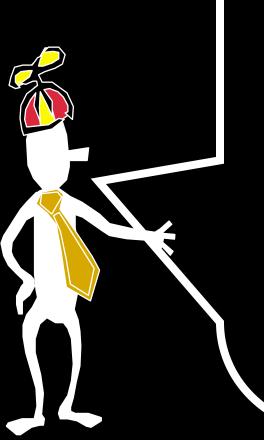
Use m(m-1)/2 indicator variables, one for each pair of people



$$\mathsf{E}[\mathsf{X}] = ?$$

Use m(m-1)/2 indicator variables, one for each pair of people

X_{jk} = 1 if person j and person k have the same birthday; else 0



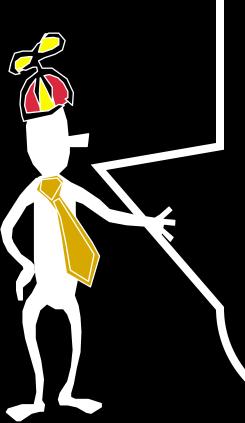
$$E[X] = ?$$

Use m(m-1)/2 indicator variables, one for each pair of people

X_{jk} = 1 if person j and person k have the same birthday; else 0

$$E[X_{jk}] = (1/366) 1 + (1 - 1/366) 0$$

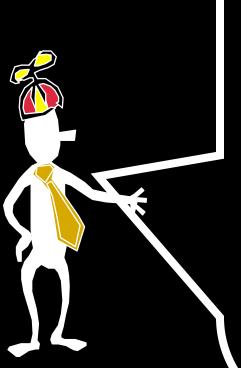
= 1/366



 X_{jk} = 1 if person j and person k have the same birthday; else 0

$$E[X_{jk}] = (1/366) 1 + (1 - 1/366) 0$$

= 1/366



 X_{jk} = 1 if person j and person k have the same birthday; else 0

$$E[X_{jk}] = (1/366) 1 + (1 - 1/366) 0$$

= 1/366

$$\mathsf{E}[\mathsf{X}] = \mathsf{E}[\; \Sigma_{\mathsf{j} \leq \mathsf{k} \leq \mathsf{m}} \; \mathsf{X}_{\mathsf{j}\mathsf{k}} \;]$$



X_{jk} = 1 if person j and person k have the same birthday; else 0

$$E[X_{jk}] = (1/366) 1 + (1 - 1/366) 0$$

= 1/366

$$E[X] = E[\sum_{j \le k \le m} X_{jk}]$$
$$= \sum_{j \le k \le m} E[X_{jk}]$$



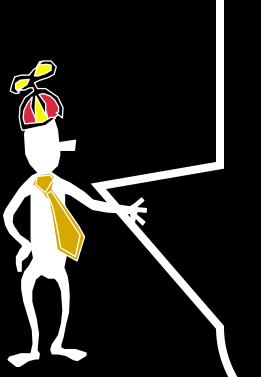
 X_{jk} = 1 if person j and person k have the same birthday; else 0

$$E[X_{jk}] = (1/366) 1 + (1 - 1/366) 0$$

= 1/366

$$E[X] = E[\sum_{j \le k \le m} X_{jk}]$$
$$= \sum_{j \le k \le m} E[X_{jk}]$$

 $= m(m-1)/2 \times 1/366$



Step Right Up...





Step Right Up...

You pick a number n ∈ [1..6]. You roll 3 dice. If any match n, you win \$1. Else you pay me \$1. Want to play?

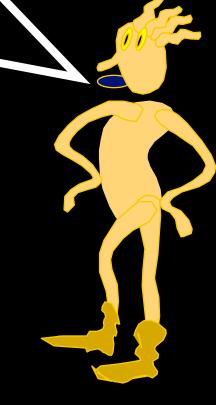


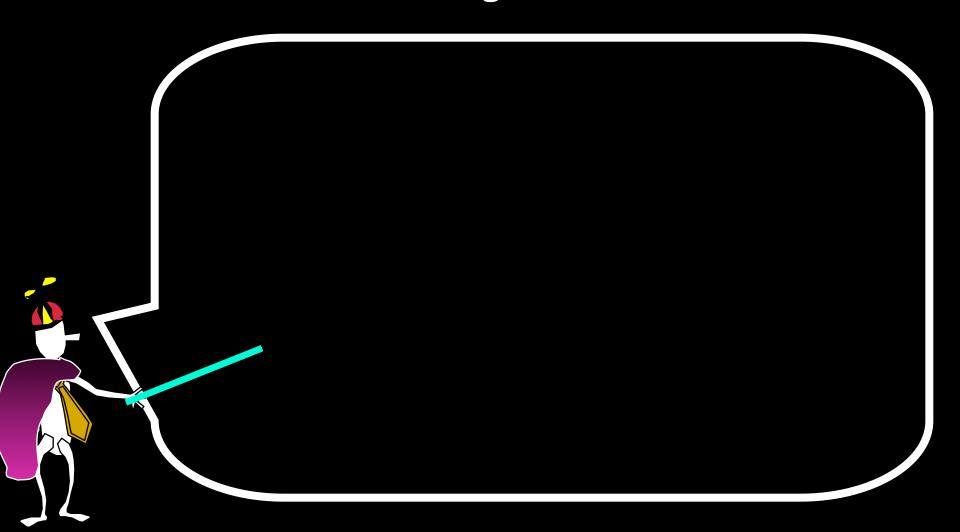


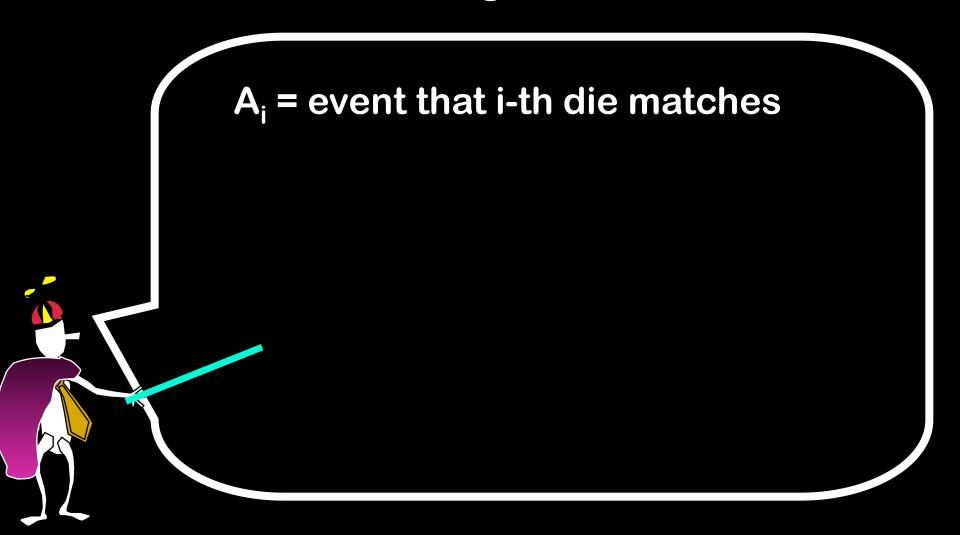
Step Right Up...

You pick a number n ∈ [1..6]. You roll 3 dice. If any match n, you win \$1. Else you pay me \$1. Want to play?



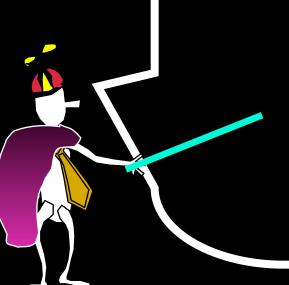








 $X_i = indicator RV for A_i$



A_i = event that i-th die matches

 $X_i = indicator RV for A_i$

Expected number of dice that match:

$$E[X_1+X_2+X_3] = 1/6+1/6+1/6 = \frac{1}{2}$$



A_i = event that i-th die matches

 $X_i = indicator RV for A_i$

Expected number of dice that match:

 $E[X_1+X_2+X_3] = 1/6+1/6+1/6 = \frac{1}{2}$

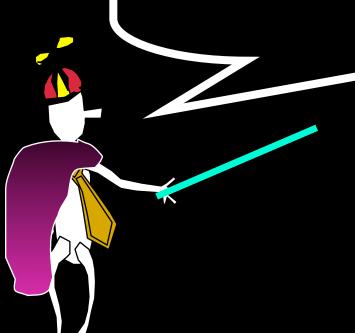
But this is not the same as Pr(at least one die matches)



Pr(at least one die matches)

= 1 - Pr(none match)

 $= 1 - (5/6)^3 = 0.416$



$$\sum_{i=0}^{\infty} \frac{1}{2^i} = ?$$



$$\sum_{i=0}^{\infty} \frac{1}{2^i} = ?$$



$$\sum_{i=0}^{\infty} \frac{1}{2^i} = ?$$
 Answer: 2



$$\sum_{i=0}^{\infty} \frac{1}{2^{i}} = ?$$

Answer: 2

0 1 1.5 2



$$\sum_{i=0}^{\infty} \frac{1}{2^i} = ?$$

Answer: 2



1

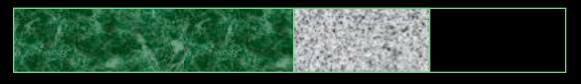
1.5

2



$$\sum_{i=0}^{\infty} \frac{1}{2^i} = ?$$

Answer: 2



1

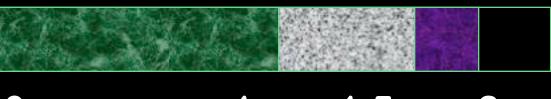
1.5

2



$$\sum_{i=0}^{\infty} \frac{1}{2^i} = ?$$

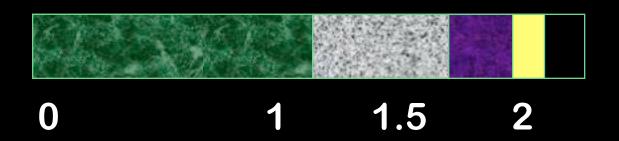
Answer: 2







$$\sum_{i=0}^{\infty} \frac{1}{2^i} = ?$$
 Answer: 2



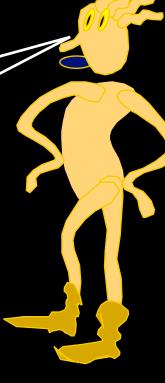


But it never actually gets to 2. Is that a problem?

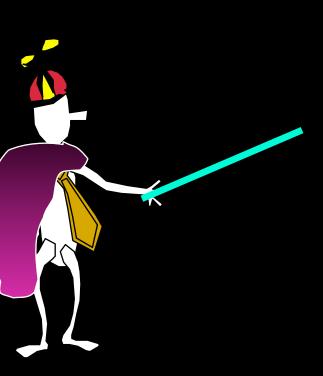
But it never actually gets to 2. Is that a problem?



No, we really mean the limit of the partial sums. In this case, the partial sum is 2-(½)ⁿ which goes to 2.



A related question



A related question

Suppose I flip a coin of bias p, stopping when I first get heads.

What's the chance that I:

•Flip exactly once?

Ans: p

•Flip exactly two times?

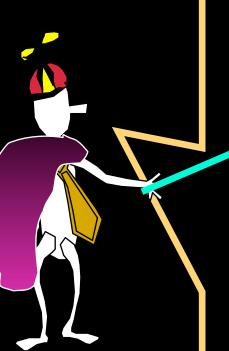
Ans: (1-p)p

•Flip exactly k times?

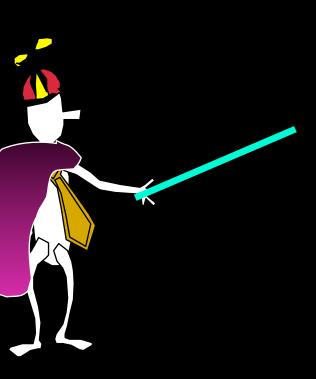
Ans: $(1-p)^{k-1}p$

•Eventually stop?

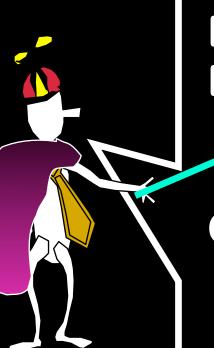
Ans: 1. (assuming p>0)



A related question



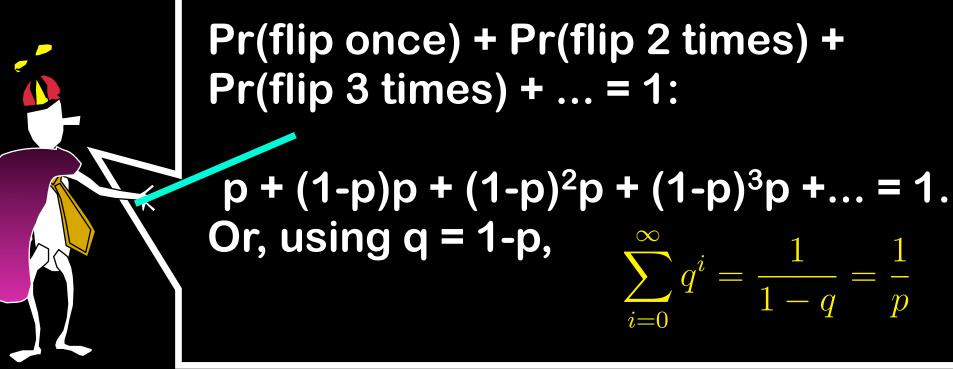
A related question



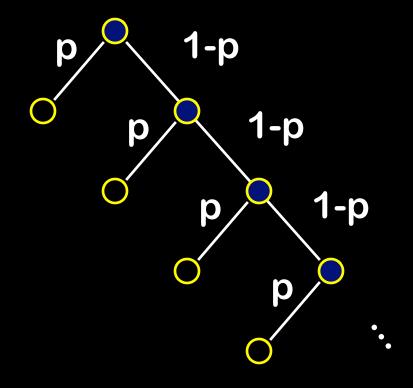
Pr(flip once) + Pr(flip 2 times) + Pr(flip 3 times) + ... = 1:

p +
$$(1-p)p + (1-p)^2p + (1-p)^3p + ... = 1$$
.
Or, using q = 1-p,

A related question



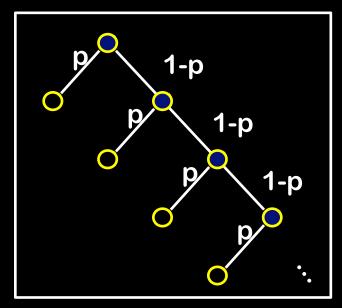
Pictorial view



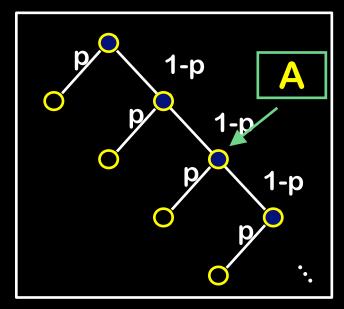
Sample space S = leaves in this tree.Pr(x) = product of edges on path to x.

If p>0, Pr(not halting by time n) goes to zero.

Suppose A is a node in this tree

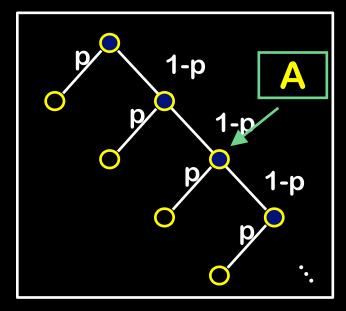


Suppose A is a node in this tree



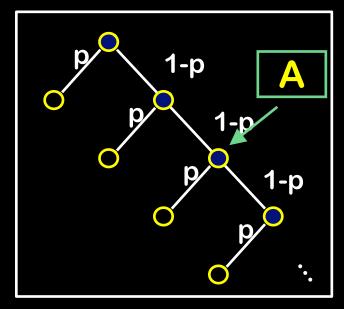
Suppose A is a node in this tree

Pr(x|A)=product of edges on path from A to x.



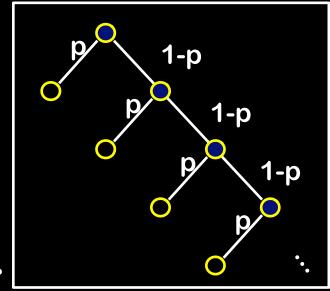
Suppose A is a node in this tree

Pr(x|A)=product of edges on path from A to x.



$$E[X] = \sum_{x} Pr(x)X(x).$$

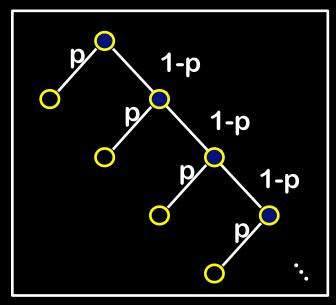
 $E[X|A] = \sum_{x \text{ in } A} Pr(x|A)X(x).$ I.e., it is as if we started the game at A.



Flip bias-p coin until heads.

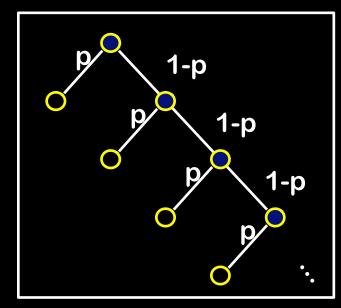
What is expected number of flips?

Let X = # flips.



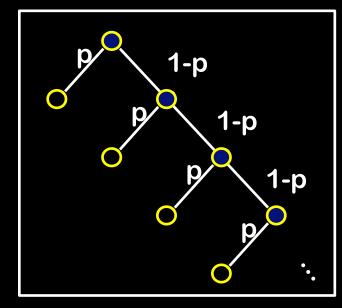
Let X = # flips.

B = event "1st flip is heads"



Let X = # flips.

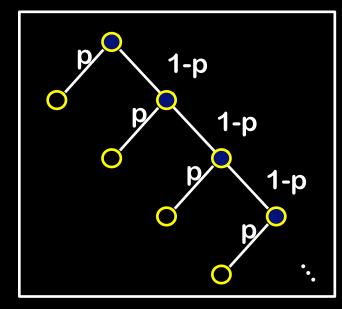
B = event "1st flip is heads"



 $E[X] = E[X|B] \times Pr(B) + E[X| not B] \times Pr(not B)$

Let X = # flips.

B = event "1st flip is heads"

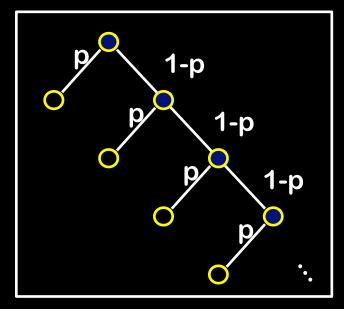


$$E[X] = E[X|B] \times Pr(B) + E[X| not B] \times Pr(not B)$$

= 1 \times p + (1 + E[X]) \times (1-p).

Let X = # flips.

B = event "1st flip is heads"



$$E[X] = E[X|B] \times Pr(B) + E[X| not B] \times Pr(not B)$$

= 1 \times p + (1 + E[X]) \times (1-p).

Solving:
$$p \times E[X] = p + (1-p)$$

 $\Rightarrow E[X] = 1/p$.

Infinite Probability spaces

Notice we are using infinite probability spaces here, but we really only defined things for <u>finite</u> spaces so far.

Infinite Probability spaces

Notice we are using infinite probability spaces here, but we really only defined things for <u>finite</u> spaces so far.

Infinite probability spaces can sometimes be weird.

Luckily, in CS we will almost always be looking at spaces that can be viewed as choice trees where

Pr(haven't halted by time t) goes to 0 as t gets large

General picture

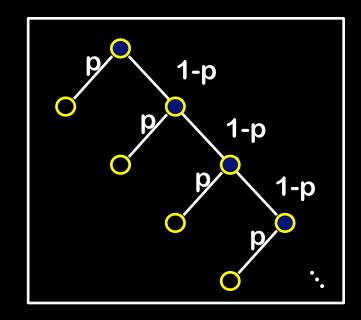
Let sample space S be leaves of a choice tree.

Let $\frac{S_n}{S_n} = \{ \text{leaves at depth } \leq n \}.$

For event A, let $A_n = A \cap S_n$.

If $\lim_{n} \Pr(S_n) = 1$, can define:

$$Pr(A)=\lim_{n} Pr(A_{n}).$$



Setting that doesn't fit our model

Event: "Flip coin until #heads > 2*#tails."

There's a reasonable chance this will <u>never</u> stop...

How to walk home drunk



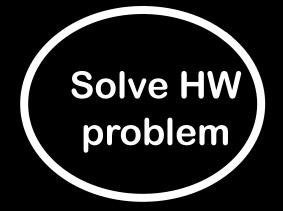
Drunk man will find way home, but drunk bird may get lost forever

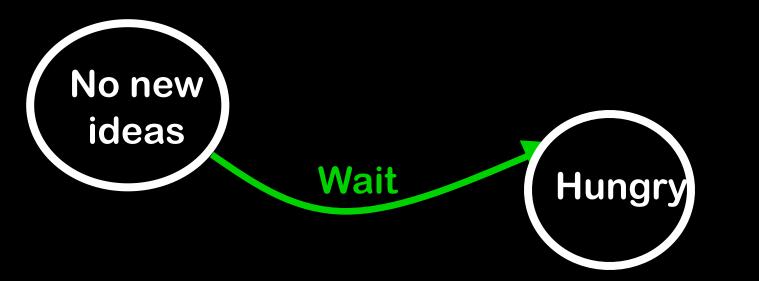
- Shizuo Kakutani

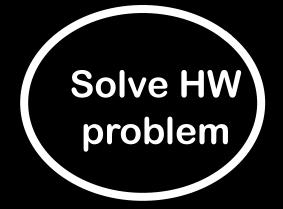


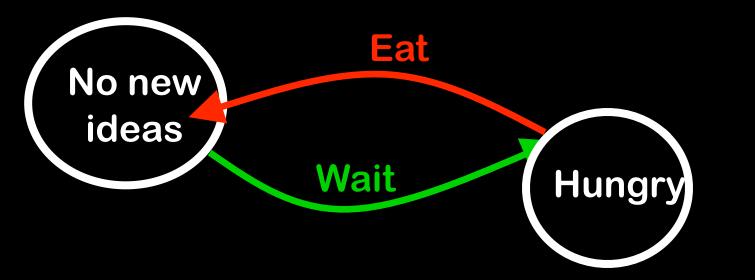


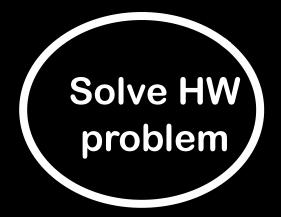


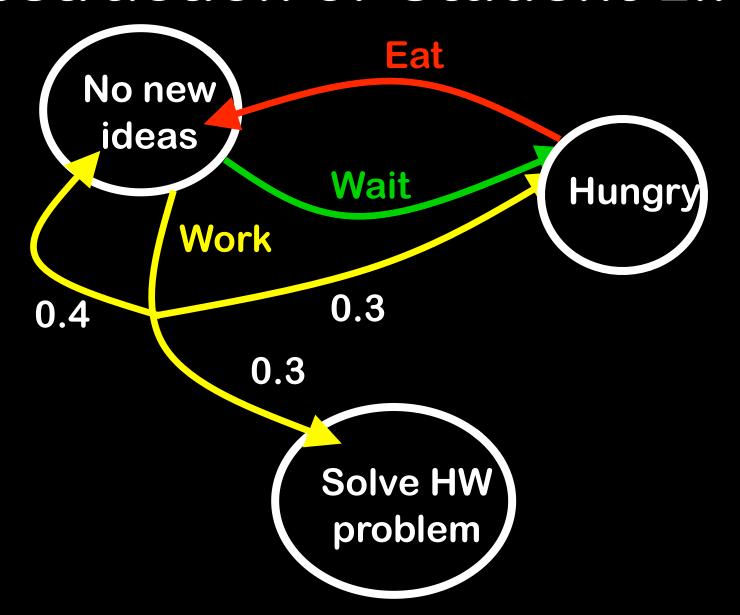


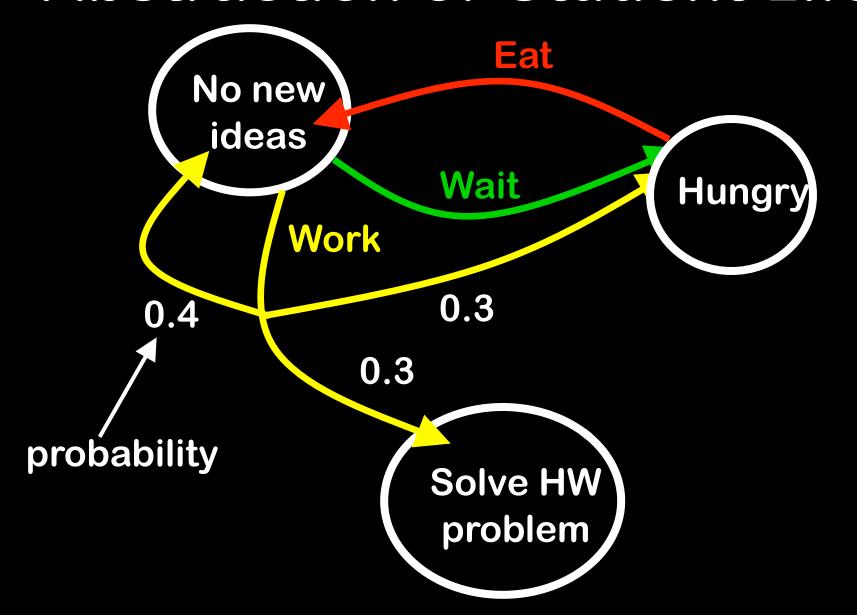


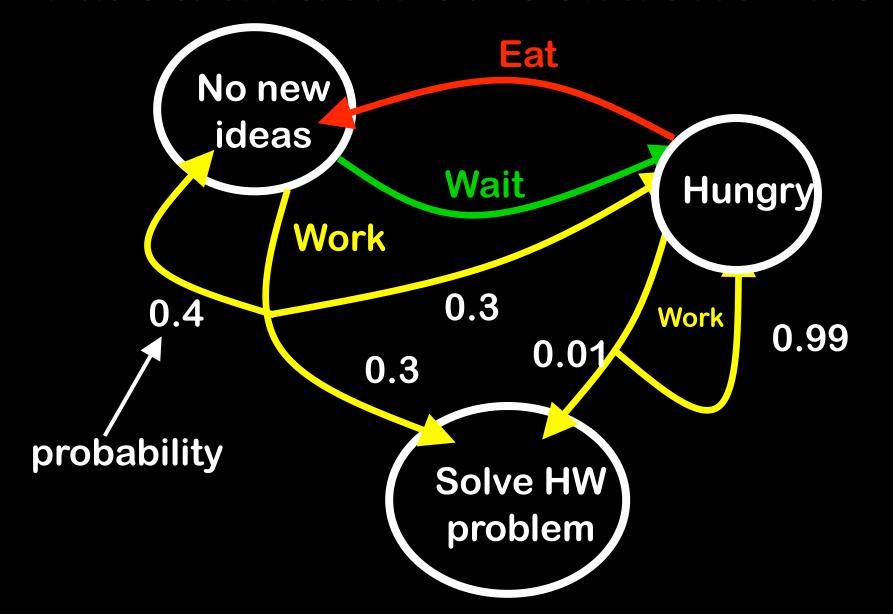


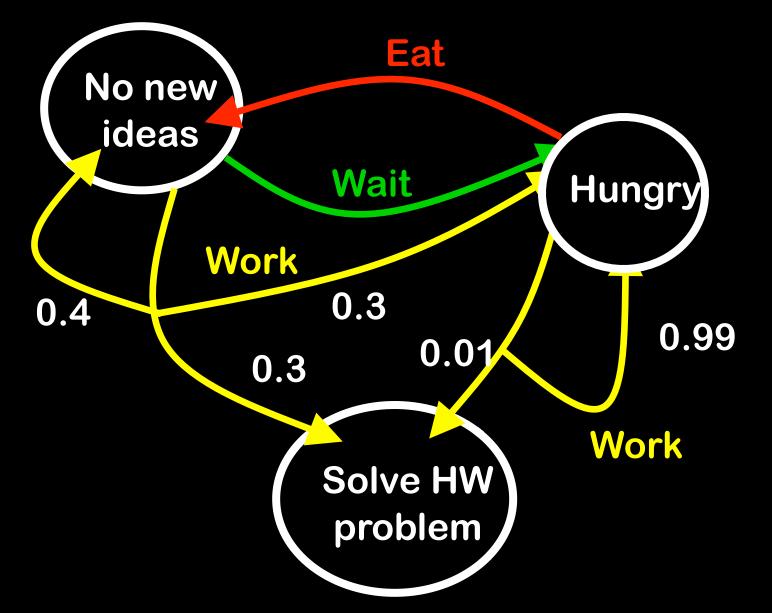












No new ideas

Wait

Hungry

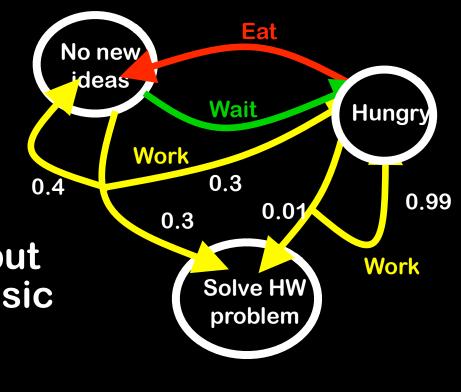
0.4

0.3

0.99

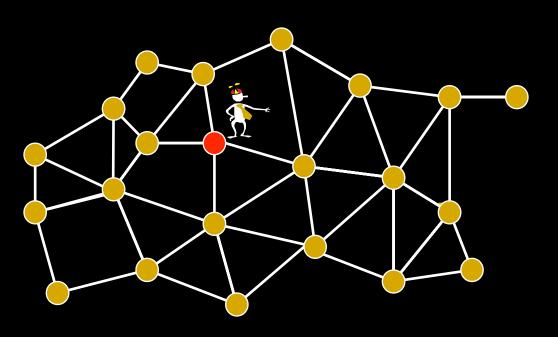
Solve HW problem

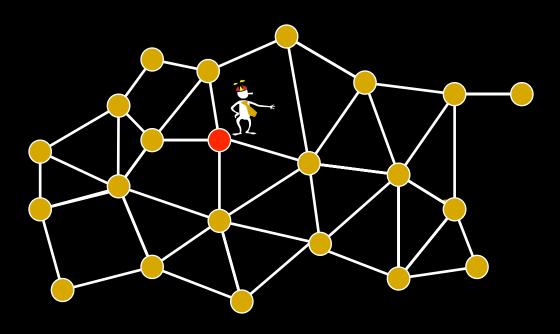
Like finite automata, but instead of a determinisic or non-deterministic action, we have a probabilistic action

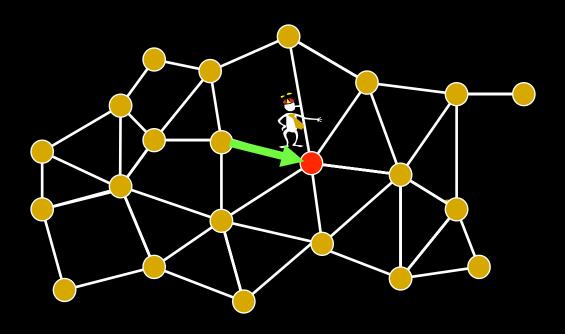


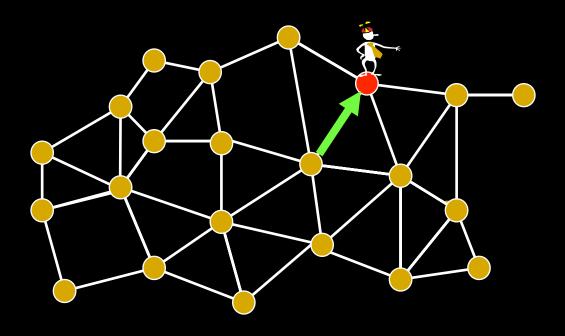
Like finite automata, but instead of a determinisic or non-deterministic action, we have a probabilistic action

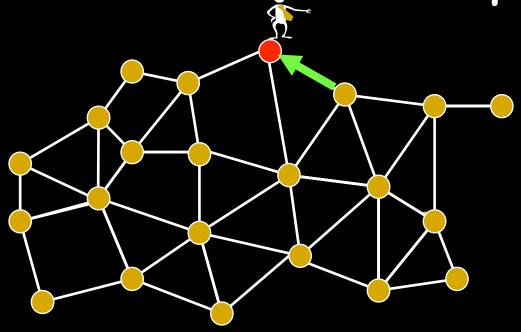
Example questions: "What is the probability of reaching goal on string Work, Eat, Work?"



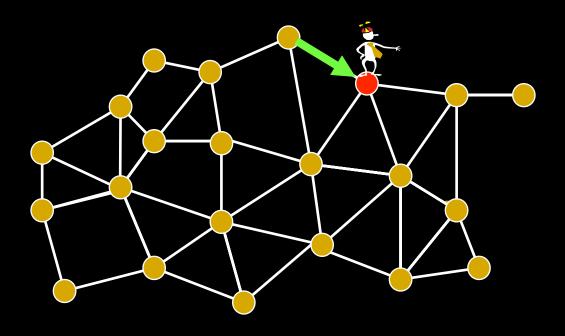








Simpler: Random Walks on Graphs



At any node, go to one of the neighbors of the node with equal probability

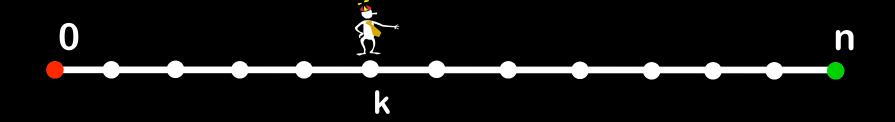
You go into a casino with \$k, and at each time step, you bet \$1 on a fair game

You go into a casino with \$k, and at each time step, you bet \$1 on a fair game

You leave when you are broke or have \$n

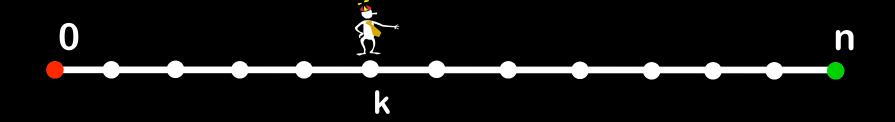
You go into a casino with \$k, and at each time step, you bet \$1 on a fair game

You leave when you are broke or have \$n



You go into a casino with \$k, and at each time step, you bet \$1 on a fair game

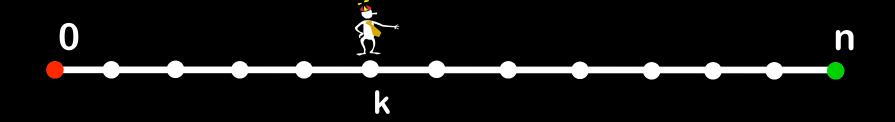
You leave when you are broke or have \$n



Question 1: what is your expected amount of money at time t?

You go into a casino with \$k, and at each time step, you bet \$1 on a fair game

You leave when you are broke or have \$n

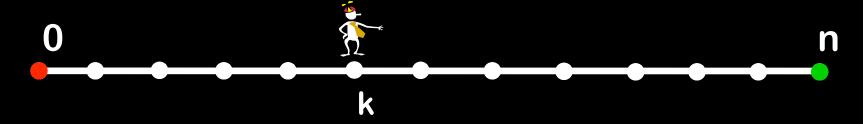


Question 1: what is your expected amount of money at time t?

Let X_t be a R.V. for the amount of \$\$\$ at time t

You go into a casino with \$k, and at each time step, you bet \$1 on a fair game

You leave when you are broke or have \$n

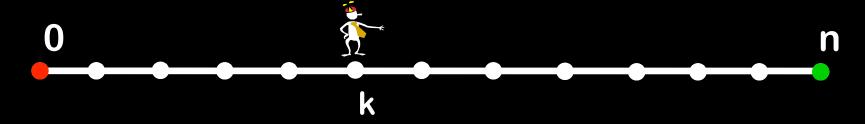


$$X_{t} = k + \delta_{1} + \delta_{2} + \dots + \delta_{t,}$$

(δ_i is RV for change in your money at time i)

You go into a casino with \$k, and at each time step, you bet \$1 on a fair game

You leave when you are broke or have \$n

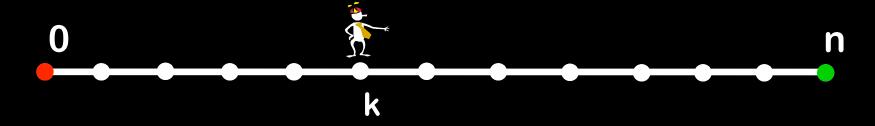


$$X_t = k + \delta_1 + \delta_2 + ... + \delta_{t,}$$

(δ_i is RV for change in your money at time i)
 $E[\delta_i] = 0$

You go into a casino with \$k, and at each time step, you bet \$1 on a fair game

You leave when you are broke or have \$n



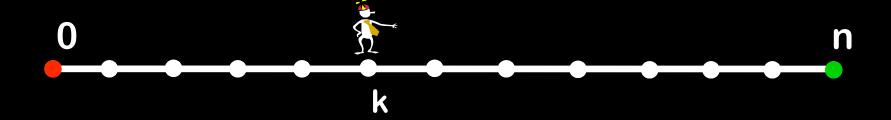
$$X_t = k + \delta_1 + \delta_2 + ... + \delta_{t,}$$

(δ_i is RV for change in your money at time i)
 $E[\delta_i] = 0$

So,
$$E[X_t] = k$$

You go into a casino with \$k, and at each time step, you bet \$1 on a fair game

You leave when you are broke or have \$n



$$E[X_t] = k$$

$$\begin{aligned} \mathsf{E}[\mathsf{X}_t] &= \mathsf{k} \\ \mathsf{E}[\mathsf{X}_t] &= \mathsf{E}[\mathsf{X}_t|\;\mathsf{X}_t = 0] \times \mathsf{Pr}(\mathsf{X}_t = 0) \\ &+ \mathsf{E}[\mathsf{X}_t|\;\mathsf{X}_t = n] \times \mathsf{Pr}(\mathsf{X}_t = n) \\ &+ \mathsf{E}[\;\mathsf{X}_t|\;\mathsf{neither}] \times \mathsf{Pr}(\mathsf{neither}) \end{aligned}$$

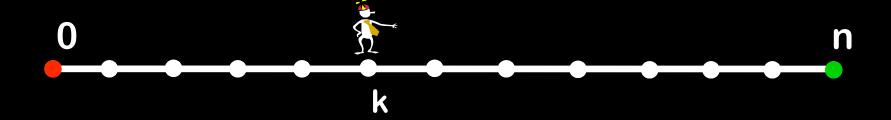
$$\begin{split} & \mathsf{E}[\mathsf{X}_t] = \mathsf{k} \\ & \mathsf{E}[\mathsf{X}_t] = \mathsf{E}[\mathsf{X}_t| \; \mathsf{X}_t = 0] \times \mathsf{Pr}(\mathsf{X}_t = 0) \\ & \quad + \mathsf{E}[\mathsf{X}_t| \; \mathsf{X}_t = n] \times \mathsf{Pr}(\mathsf{X}_t = n) \\ & \quad + \mathsf{E}[\; \mathsf{X}_t| \; \mathsf{neither}] \times \mathsf{Pr}(\mathsf{neither}) \\ & \quad \mathsf{k} = \mathsf{n} \times \mathsf{Pr}(\mathsf{X}_t = \mathsf{n}) \\ & \quad + (\mathsf{something}_t) \times \mathsf{Pr}(\mathsf{neither}) \end{split}$$

$$\begin{split} & \mathsf{E}[\mathsf{X}_t] = \mathsf{k} \\ & \mathsf{E}[\mathsf{X}_t] = \mathsf{E}[\mathsf{X}_t| \; \mathsf{X}_t = 0] \times \mathsf{Pr}(\mathsf{X}_t = 0) \\ & \quad + \mathsf{E}[\mathsf{X}_t| \; \mathsf{X}_t = n] \times \mathsf{Pr}(\mathsf{X}_t = n) \\ & \quad + \mathsf{E}[\; \mathsf{X}_t| \; \mathsf{neither}] \times \mathsf{Pr}(\mathsf{neither}) \\ & \mathsf{k} = \mathsf{n} \times \mathsf{Pr}(\mathsf{X}_t = n) \\ & \quad + (\mathsf{something}_t) \times \mathsf{Pr}(\mathsf{neither}) \\ & \mathsf{As} \; t \to \infty, \; \mathsf{Pr}(\mathsf{neither}) \to 0, \; \mathsf{also} \; \mathsf{something}_t < \mathsf{n} \\ & \quad \mathsf{Hence} \; \mathsf{Pr}(\mathsf{X}_t = n) \to \mathsf{k/n} \end{split}$$

Another Way To Look At It

You go into a casino with \$k, and at each time step, you bet \$1 on a fair game

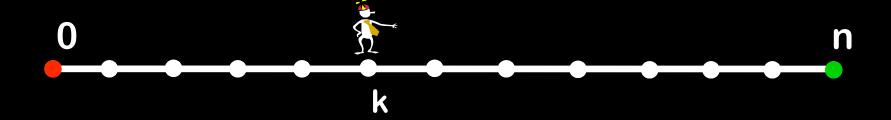
You leave when you are broke or have \$n



Another Way To Look At It

You go into a casino with \$k, and at each time step, you bet \$1 on a fair game

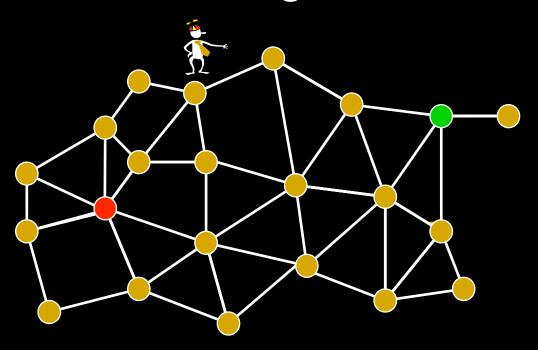
You leave when you are broke or have \$n



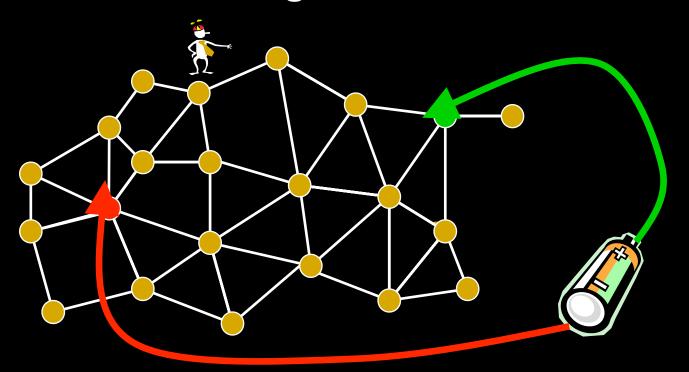
Question 2: what is the probability that you leave with \$n?

= probability that I hit green before I hit red

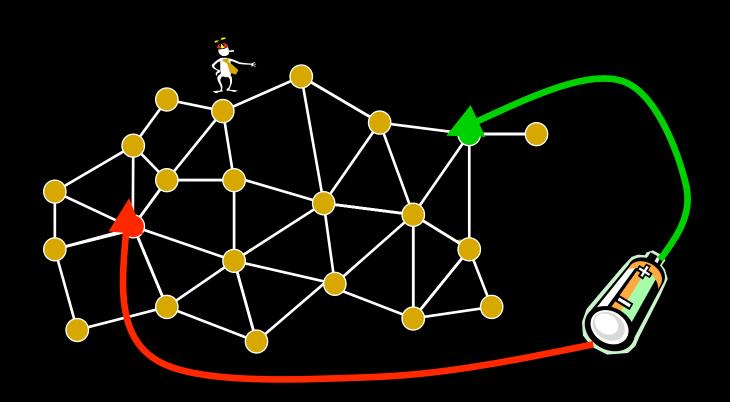
What is chance I reach green before red?

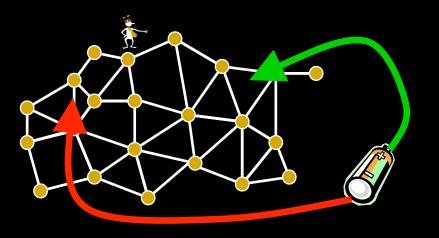


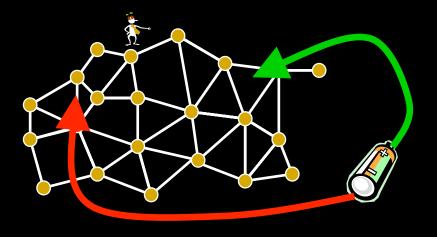
What is chance I reach green before red?



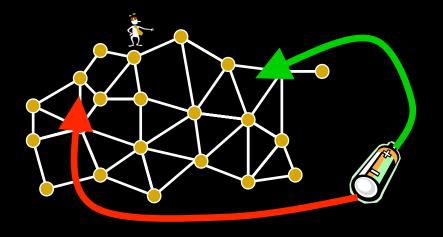
Same as voltage if edges are resistors and we put 1-volt battery between green and red





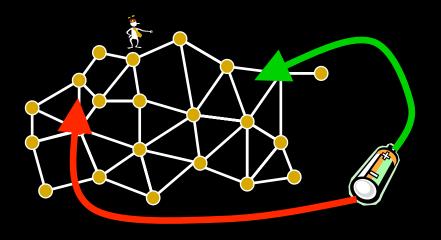


 $p_x = Pr(reach green first starting from x)$



 $p_x = Pr(reach green first starting from x)$

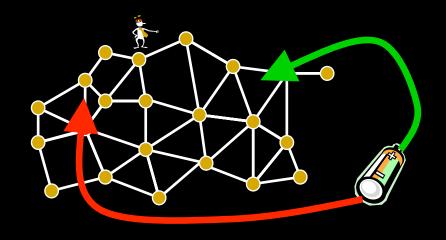
$$p_{green} = 1, p_{red} = 0$$



 $p_x = Pr(reach green first starting from x)$

$$p_{green} = 1, p_{red} = 0$$

And for the rest $p_x = Average_{y \in Nbr(x)}(p_y)$



 $p_x = Pr(reach green first starting from x)$

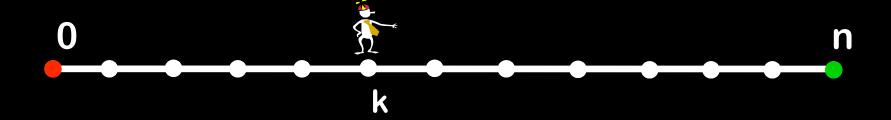
$$p_{green} = 1, p_{red} = 0$$

And for the rest p_x = Average $_{y \in Nbr(x)}(p_y)$ Same as equations for voltage if edges all have same resistance!

Another Way To Look At It

You go into a casino with \$k, and at each time step, you bet \$1 on a fair game

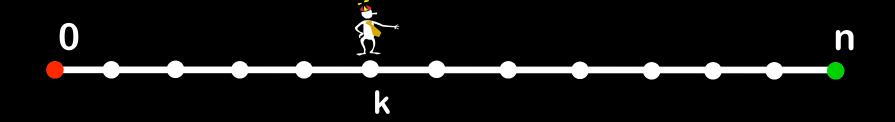
You leave when you are broke or have \$n

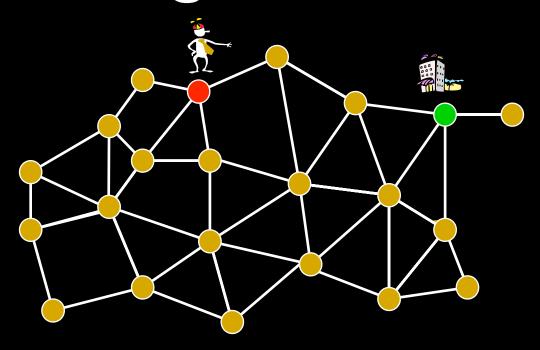


Another Way To Look At It

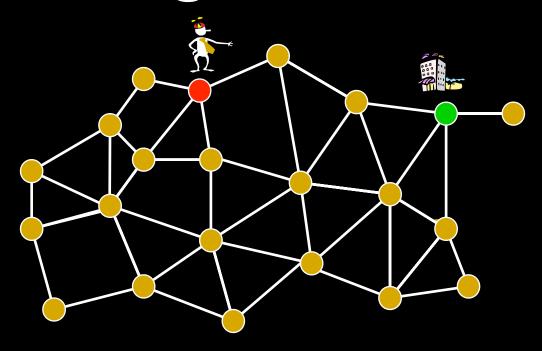
You go into a casino with \$k, and at each time step, you bet \$1 on a fair game

You leave when you are broke or have \$n



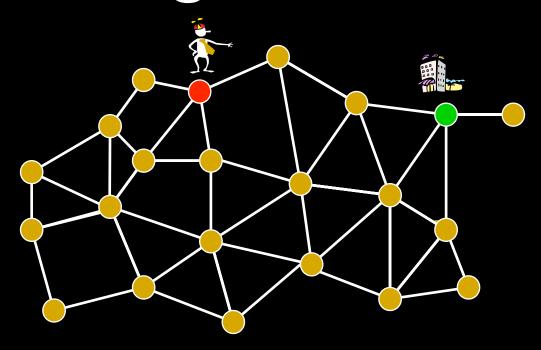


Lost in a city, you want to get back to your hotel How should you do this?



Lost in a city, you want to get back to your hotel How should you do this?

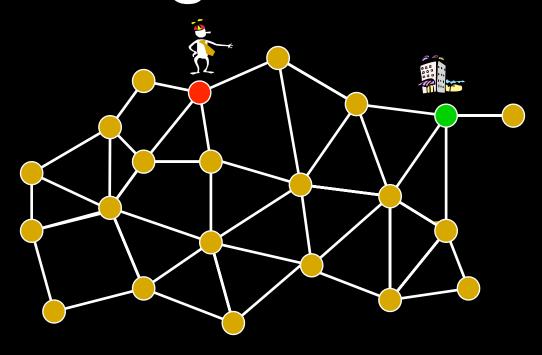
Depth First Search!



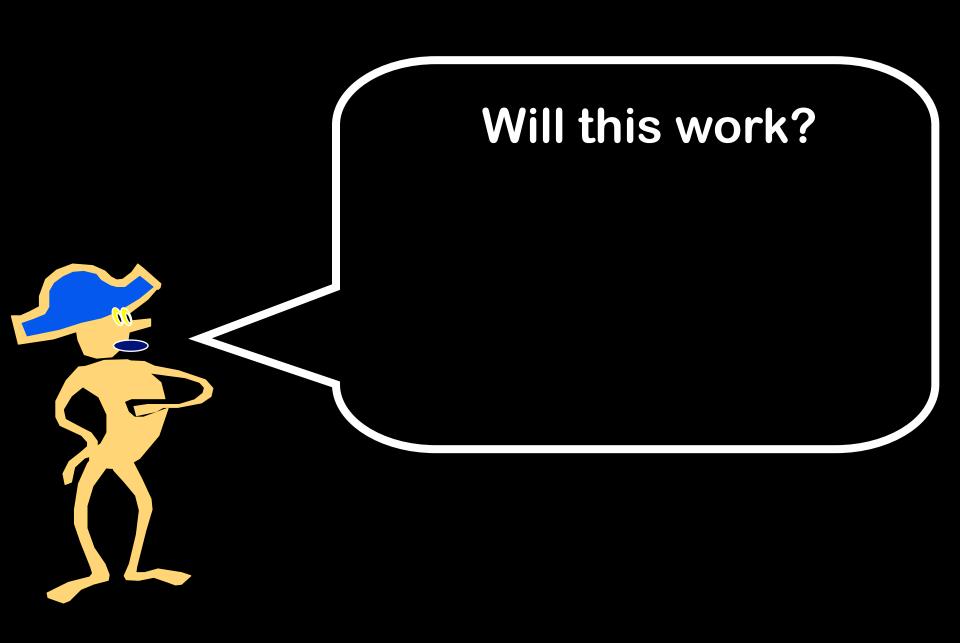
Lost in a city, you want to get back to your hotel How should you do this?

Depth First Search!

Requires a good memory and a piece of chalk

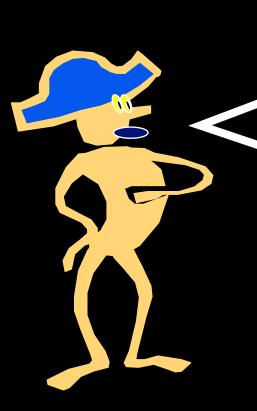


How about walking randomly?



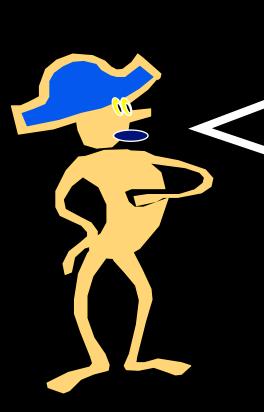








When will I get home?

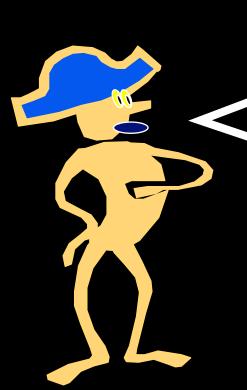




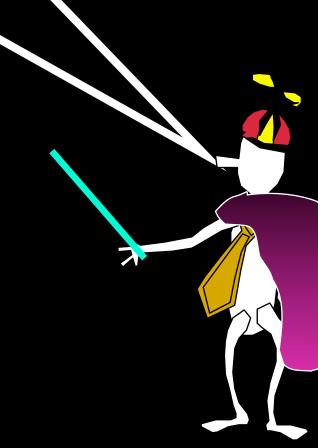
Is Pr[reach home] = 1?

When will I get home?

What is E[time to reach home]?



Pr[will reach home] = 1



Look at the first n steps

Look at the first n steps

There is a non-zero chance p₁ that we get home

Look at the first n steps

There is a non-zero chance p₁ that we get home

Also, $p_1 \ge (1/n)^n$

Look at the first n steps

There is a non-zero chance p₁ that we get home

Also, $p_1 \ge (1/n)^n$

Suppose we fail

Look at the first n steps

There is a non-zero chance p₁ that we get home

Also, $p_1 \ge (1/n)^n$

Suppose we fail

Then, wherever we are, there is a chance $p_2 \ge (1/n)^n$ that we hit home in the next n steps from there

Look at the first n steps

There is a non-zero chance p₁ that we get home

Also, $p_1 \ge (1/n)^n$

Suppose we fail

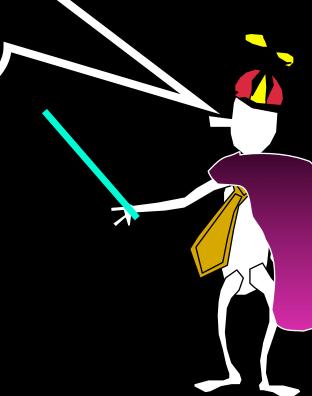
Then, wherever we are, there is a chance $p_2 \ge (1/n)^n$ that we hit home in the next n steps from there

Probability of failing to reach home by time kn = $(1 - p_1)(1 - p_2) \dots (1 - p_k) \rightarrow 0$ as $k \rightarrow \infty$

Furthermore:

Furthermore:

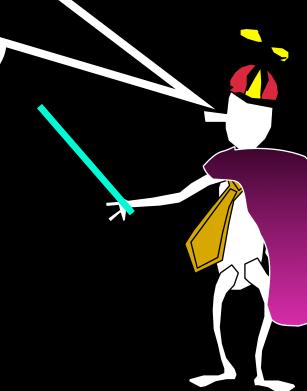
If the graph has nodes and m edges, then



Furthermore:

If the graph has n nodes and m edges, then

E[time to visit all nodes]
≤ 2m × (n-1)





If the graph has n nodes and m edges, then

E[time to visit all nodes]
≤ 2m × (n-1)

E[time to reach home] is at most this

```
Cover time (from u)

C<sub>u</sub> = E [ time to visit all vertices | start at u ]
```

Cover time (from u)

C_u = E [time to visit all vertices | start at u]

Cover time of the graph

```
Cover time (from u)

C_u = E [ time to visit all vertices | start at u ]
```

Cover time of the graph $C(G) = max_u \{ C_u \}$

```
Cover time (from u)

C<sub>u</sub> = E [ time to visit all vertices | start at u ]
```

Cover time of the graph

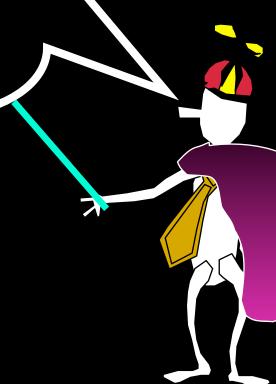
$$C(G) = max_u \{ C_u \}$$

(worst case expected time to see all vertices)

If the graph G has n nodes and m edges, then the cover time of G is

If the graph G has n nodes and m edges, then the cover time of G is

 $C(G) \leq 2m (n-1)$



If the graph G has n nodes and m edges, then the cover time of G is

 $C(G) \leq 2m (n-1)$

Any graph on n vertices has < n²/2 edges

If the graph G has n nodes and m edges, then the cover time of G is

 $C(G) \leq 2m (n-1)$

Any graph on n vertices has < n²/2 edges

Hence C(G) < n³ for all graphs G



Chance that we don't hit home by (2k)2m(n-1) steps is (½)k

True or False:

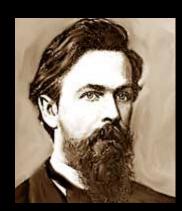
True or False:

If the average income of people is \$100 then more than 50% of the people can be earning more than \$200 each

True or False:

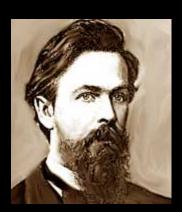
If the average income of people is \$100 then more than 50% of the people can be earning more than \$200 each

False! else the average would be higher!!!



Andrei A. Markov

If X is a non-negative r.v. with mean E[X], then



Andrei A. Markov

If X is a non-negative r.v. with mean E[X], then

 $Pr[X > 2 E[X]] \le \frac{1}{2}$

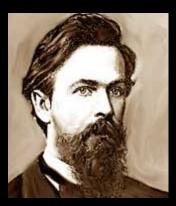


Andrei A. Markov

If X is a non-negative r.v. with mean E[X], then

 $Pr[X > 2 E[X]] \le \frac{1}{2}$

 $Pr[X > k E[X]] \leq 1/k$



Andrei A. Markov

Non-neg random variable X has expectation A = E[X]

Non-neg random variable X has expectation A = E[X]

$$A = E[X] = E[X | X > 2A] Pr[X > 2A] + E[X | X \le 2A] Pr[X \le 2A]$$

Non-neg random variable X has expectation A = E[X]

$$A = E[X] = E[X | X > 2A] Pr[X > 2A] + E[X | X \le 2A] Pr[X \le 2A]$$

 \geq E[X | X > 2A] Pr[X > 2A]

Non-neg random variable X has expectation A = E[X]

```
A = E[X] = E[X | X > 2A] Pr[X > 2A] + E[X | X \le 2A] Pr[X \le 2A]
```

 \geq E[X | X > 2A] Pr[X > 2A] (since X is non-neg)

Non-neg random variable X has expectation A = E[X]

```
A = E[X] = E[X | X > 2A] Pr[X > 2A] + E[X | X \le 2A] Pr[X \le 2A]
```

 $\geq E[X \mid X > 2A] Pr[X > 2A]$ (since X is non-neg)

Also, $E[X \mid X > 2A] > 2A$

Non-neg random variable X has expectation A = E[X]

$$A = E[X] = E[X | X > 2A] Pr[X > 2A] + E[X | X \le 2A] Pr[X \le 2A]$$

 \geq E[X | X > 2A] Pr[X > 2A] (since X is non-neg)

Also, $E[X \mid X > 2A] > 2A$

 \Rightarrow A \geq 2A \times Pr[X > 2A]

Non-neg random variable X has expectation A = E[X]

$$A = E[X] = E[X | X > 2A] Pr[X > 2A] + E[X | X \le 2A] Pr[X \le 2A]$$

 \geq E[X | X > 2A] Pr[X > 2A] (since X is non-neg)

Also, $E[X \mid X > 2A] > 2A$

- \Rightarrow A \geq 2A \times Pr[X > 2A]
 - $\Rightarrow \frac{1}{2} \ge \Pr[X > 2A]$

Non-neg random variable X has expectation A = E[X]

$$A = E[X] = E[X | X > 2A] Pr[X > 2A] + E[X | X \le 2A] Pr[X \le 2A]$$

 \geq E[X | X > 2A] Pr[X > 2A] (since X is non-neg)

Also, $E[X \mid X > 2A] > 2A$

- \Rightarrow A \geq 2A \times Pr[X > 2A]
 - $\Rightarrow \frac{1}{2} \ge \Pr[X > 2A]$

 $Pr[X > k \times expectation] \le 1/k$



Chance that we don't hit home by (2k)2m(n-1) steps is (½)k

Suppose I start at u

Suppose I start at u

E[time to hit all vertices | start at u] ≤ C(G)

Suppose I start at u

E[time to hit all vertices | start at u] ≤ C(G)

Hence, by Markov's Inequality:

Pr[time to hit all vertices > 2C(G) | start at u] $\leq \frac{1}{2}$

Pr [time to hit all vertices > 2C(G) | start at u] $\leq \frac{1}{2}$

Pr [time to hit all vertices > 2C(G) | start at u] $\leq \frac{1}{2}$

Suppose at time 2C(G), I'm at some node with more nodes still to visit

Pr [time to hit all vertices > 2C(G) | start at u] $\leq \frac{1}{2}$

Suppose at time 2C(G), I'm at some node with more nodes still to visit

Pr [haven't hit all vertices in 2C(G) more time | start at v] $\leq \frac{1}{2}$

Pr [time to hit all vertices > 2C(G) | start at u] $\leq \frac{1}{2}$

Suppose at time 2C(G), I'm at some node with more nodes still to visit

Pr [haven't hit all vertices in 2C(G) more time | start at v] $\leq \frac{1}{2}$

Chance that you failed both times $\leq \frac{1}{4} = (\frac{1}{2})^2$

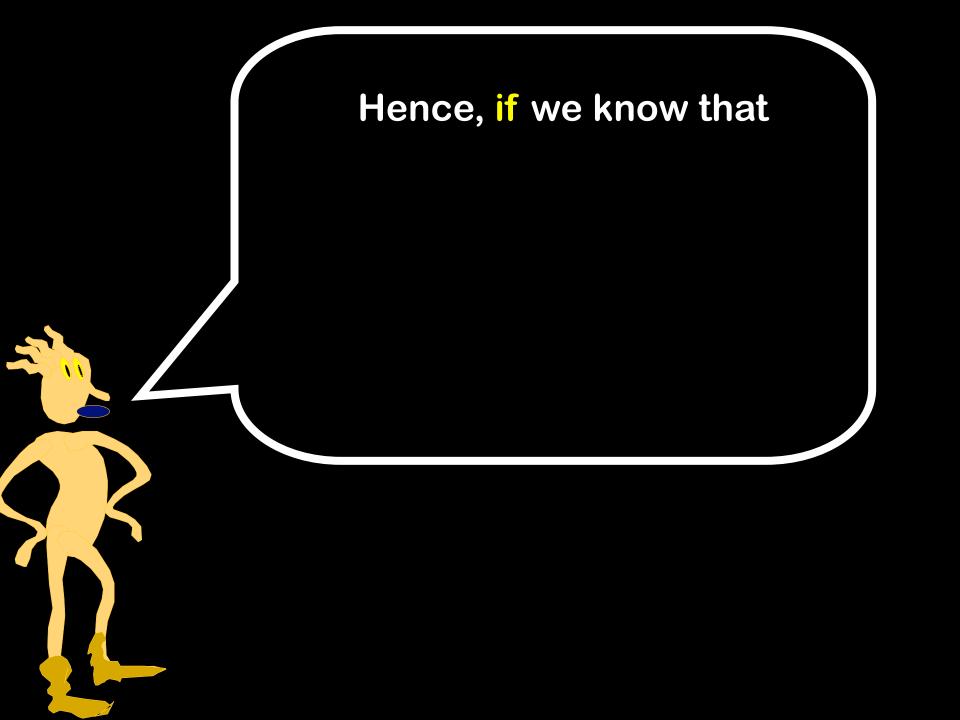
Pr [time to hit all vertices > 2C(G) | start at u] $\leq \frac{1}{2}$

Suppose at time 2C(G), I'm at some node with more nodes still to visit

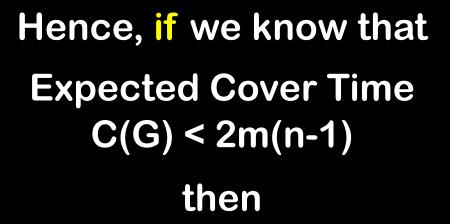
Pr [haven't hit all vertices in 2C(G) more time | start at $v \le \frac{1}{2}$

Chance that you failed both times $\leq \frac{1}{4} = (\frac{1}{2})^2$

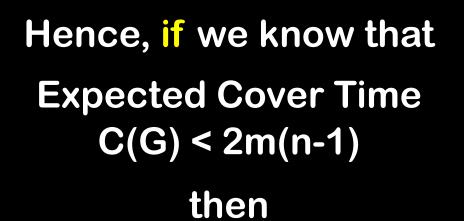
Hence, Pr[havent hit everyone in time $k \times 2C(G)$] $\leq (\frac{1}{2})^k$

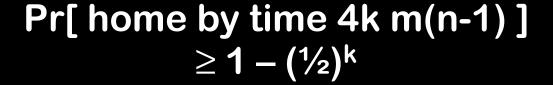


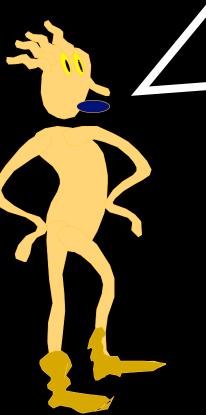












Random walks on infinite graphs

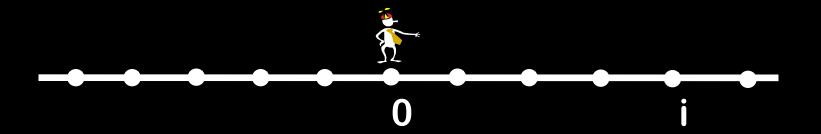


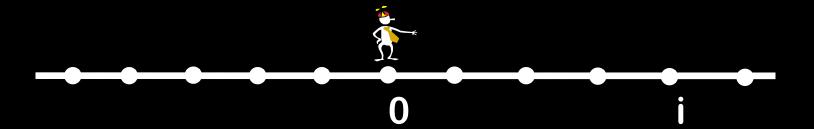


Drunk man will find way home, but drunk bird may get lost forever

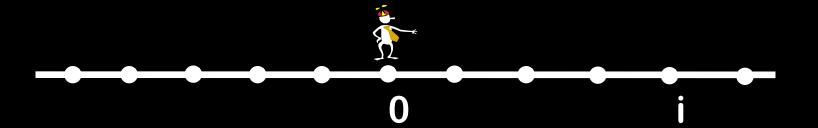
- Shizuo Kakutani



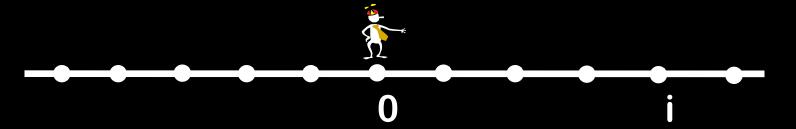




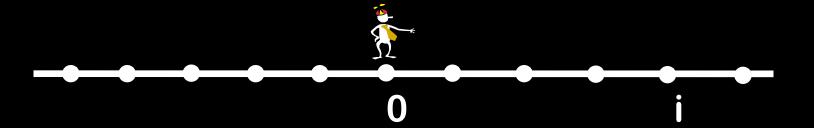
Flip an unbiased coin and go left/right



Flip an unbiased coin and go left/right Let X_t be the position at time t



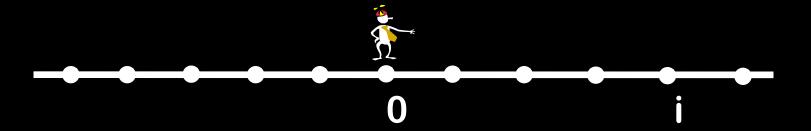
Flip an unbiased coin and go left/right Let X_t be the position at time t $Pr[X_t = i]$



Flip an unbiased coin and go left/right

Let X_t be the position at time t

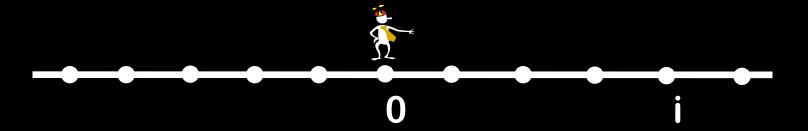
```
Pr[X_t = i] = Pr[\#heads - \#tails = i]
```



Flip an unbiased coin and go left/right

Let X_t be the position at time t

```
Pr[X_t = i] = Pr[\#heads - \#tails = i]
= Pr[\#heads - (t - \#heads) = i]
```



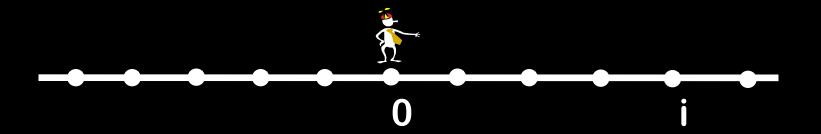
Flip an unbiased coin and go left/right

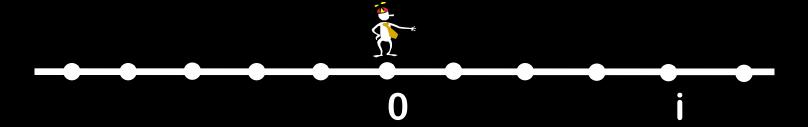
Let X_t be the position at time t

$$Pr[X_t = i] = Pr[\#heads - \#tails = i]$$

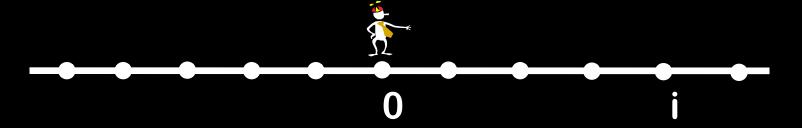
= $Pr[\#heads - (t - \#heads) = i]$

$$= \begin{bmatrix} t \\ (t+i)/2 \end{bmatrix} /2^t$$

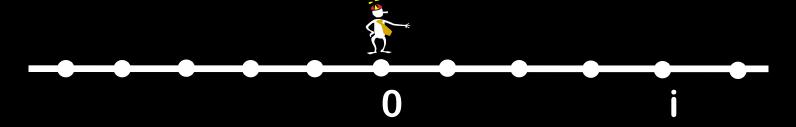




$$Pr[X_{2t} = 0] = \begin{bmatrix} 2t \\ t \end{bmatrix} /2^{2t}$$

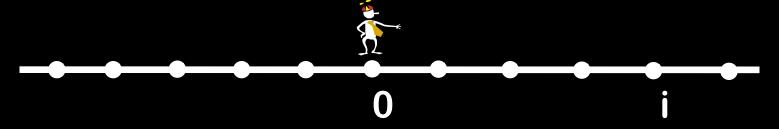


$$Pr[X_{2t} = 0] = \begin{bmatrix} 2t \\ t \end{bmatrix} /2^{2t} \le \Theta(1/\sqrt{t})$$



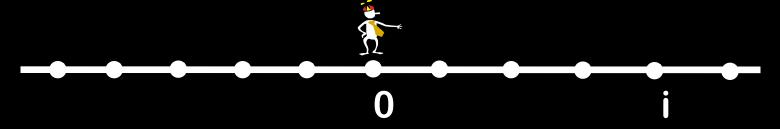
$$Pr[X_{2t} = 0] = \begin{bmatrix} 2t \\ t \end{bmatrix} /2^{2t} \le \Theta(1/\sqrt{t})$$

Stirling's approx



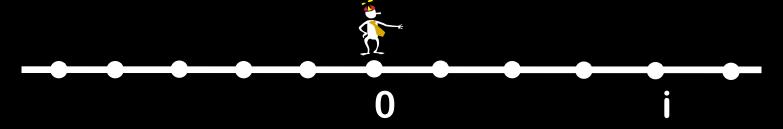
$$\Pr[X_{2t} = 0] = \begin{bmatrix} 2t \\ t \end{bmatrix} /2^{2t} \le \Theta(1/\sqrt{t})$$
 Stirling's approx

 Y_{2t} = indicator for $(X_{2t} = 0)$



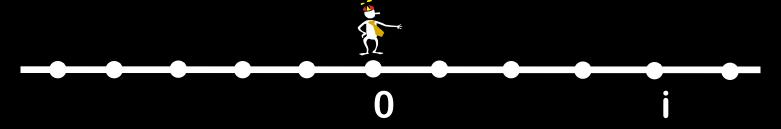
$$\Pr[X_{2t} = 0] = \begin{bmatrix} 2t \\ t \end{bmatrix} /2^{2t} \le \Theta(1/\sqrt{t})$$
 Stirling's approx

 $Y_{2t} = \text{indicator for } (X_{2t} = 0) \Rightarrow E[Y_{2t}] = \Theta(1/\sqrt{t})$



$$\Pr[X_{2t} = 0] = \begin{bmatrix} 2t \\ t \end{bmatrix} /2^{2t} \le \Theta(1/\sqrt{t})$$
 Stirling's approx

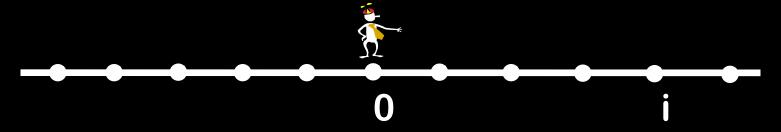
$$Y_{2t}$$
 = indicator for $(X_{2t} = 0) \Rightarrow E[Y_{2t}] = \Theta(1/\sqrt{t})$



$$\Pr[X_{2t} = 0] = \begin{bmatrix} 2t \\ t \end{bmatrix} /2^{2t} \le \Theta(1/\sqrt{t})$$
 Stirling's approx

$$Y_{2t}$$
 = indicator for $(X_{2t} = 0) \Rightarrow E[Y_{2t}] = \Theta(1/\sqrt{t})$

$$E[Z_{2n}] = E[\sum_{t=1...n} Y_{2t}]$$

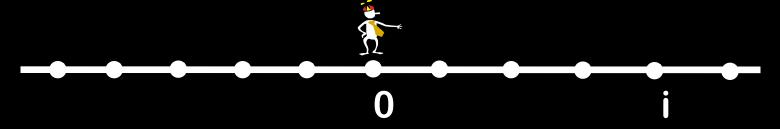


$$\Pr[X_{2t} = 0] = \begin{bmatrix} 2t \\ t \end{bmatrix} /2^{2t} \le \Theta(1/\sqrt{t})$$
 Stirling's approx

$$Y_{2t}$$
 = indicator for $(X_{2t} = 0) \Rightarrow E[Y_{2t}] = \Theta(1/\sqrt{t})$

$$E[Z_{2n}] = E[\sum_{t=1...n} Y_{2t}]$$

$$\leq \Theta(1/\sqrt{1} + 1/\sqrt{2} + ... + 1/\sqrt{n})$$



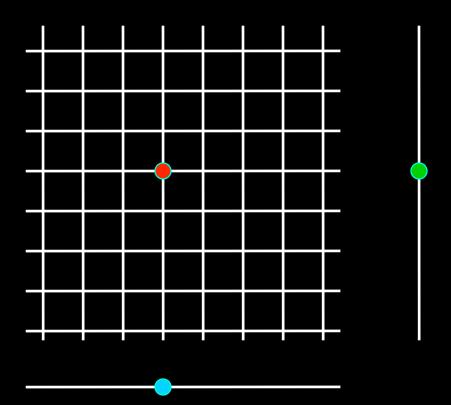
$$\Pr[X_{2t} = 0] = \begin{bmatrix} 2t \\ t \end{bmatrix} /2^{2t} \le \Theta(1/\sqrt{t})$$
 Stirling's approx

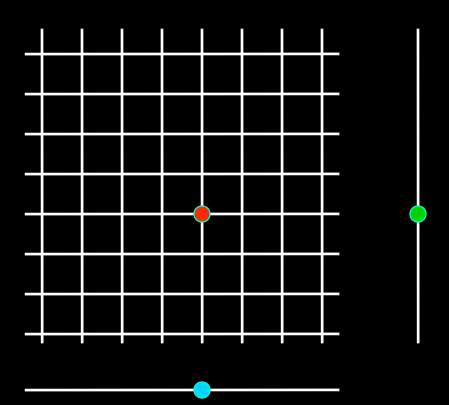
$$Y_{2t}$$
 = indicator for $(X_{2t} = 0) \Rightarrow E[Y_{2t}] = \Theta(1/\sqrt{t})$

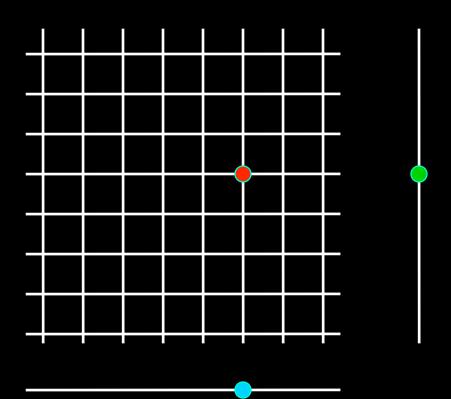
$$E[Z_{2n}] = E[\sum_{t=1...n} Y_{2t}]$$

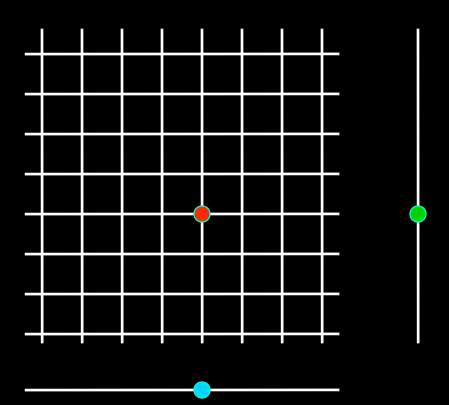
$$\leq \Theta(1/\sqrt{1} + 1/\sqrt{2} + ... + 1/\sqrt{n}) = \Theta(\sqrt{n})$$

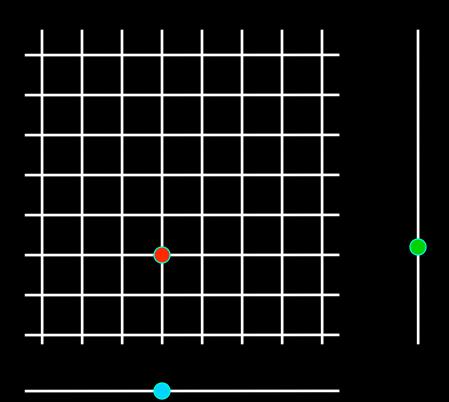
In n steps, you expect to return to the origin $\Theta(\sqrt{n})$ times!











Returning to the origin in the grid

both "line" random walks return
to their origins

Returning to the origin in the grid

both "line" random walks return
to their origins

```
Pr[visit origin at time t] = \Theta(1/\sqrt{t}) \times \Theta(1/\sqrt{t})
= \Theta(1/t)
```

Returning to the origin in the grid

both "line" random walks return
to their origins

Pr[visit origin at time t] =
$$\Theta(1/\sqrt{t}) \times \Theta(1/\sqrt{t})$$

= $\Theta(1/t)$

E[# of visits to origin by time n]
=
$$\Theta(1/1 + 1/2 + 1/3 + ... + 1/n) = \Theta(\log n)$$

Pr[visit origin at time t] = $\Theta(1/\sqrt{t})^3 = \Theta(1/t^{3/2})$

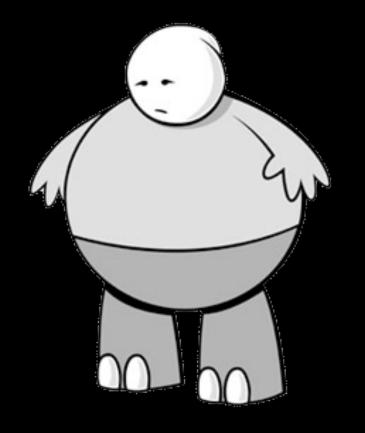
Pr[visit origin at time t] = $\Theta(1/\sqrt{t})^3 = \Theta(1/t^{3/2})$

 $\lim_{n\to\infty} E[\# of visits by time n] < K (constant)$

Pr[visit origin at time t] = $\Theta(1/\sqrt{t})^3 = \Theta(1/t^{3/2})$

lim_{n→∞} E[# of visits by time n] < K (constant)

Hence Pr[never return to origin] > 1/K



Here's What You Need to Know... **Conditional expectation**

Flipping coins with bias p
Expected number of flips
before a heads

Random Walk on a Line

Cover Time of a Graph

Markov's Inequality