

Lecture 21

NONNEGATIVE MATRICES

(and applications to PageRank)

* PSET 3 is out! Due 10/30 9PM.

* Checkups (revamped!)

* Midterm Re-dos (signup sheet)

* OFFICE HOURS (P.)

Wed 10/28
11AM

Eigenvalues / Eigenvectors

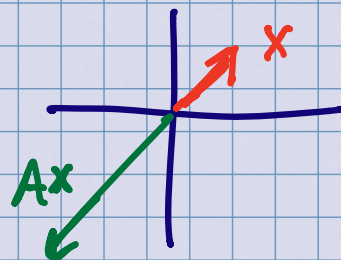
$A \in \mathbb{R}^{n \times n}$
(square)

$n \times n$ $n \times 1$ 1×1 $n \times 1$

$$A X = \lambda X$$

↑
matrix

↑
scalar
(eigenvalue)



eigenvalues
↓

DIAGONALIZATION:
(sometimes)

$$A = T D T^{-1}$$

↑
eigenvectors

Symmetric Matrices

$$A = A^T$$

- All eigenvalues are real
- Full basis of orthogonal eigenvectors

eigenvectors
↓

$$\Rightarrow A = T D T^T$$

↑
eigenvalues

(always diagonalizable)

T is orthogonal
($T T^T = T^T T = I$)

NON NEGATIVE MATRICES

A square matrix where all entries are non-negative numbers

$$A_{ij} \geq 0 \quad \forall ij$$

EX: $\begin{bmatrix} 1 & 3 & 2 \\ 0 & 4 & 3 \\ 5 & 7 & 1 \end{bmatrix}$ ✓ $\begin{bmatrix} 0 & 1 \\ 10 & 2 \end{bmatrix}$ ✓ ~~$\begin{bmatrix} 2 & -3 & 4 \\ 5 & 1 & 2 \\ 0 & 3 & -1 \end{bmatrix}$~~

IF A, B ARE NONNEGATIVE MATRICES,
ARE THESE ALSO NONNEGATIVE?

• $A + B$ ✓ $a_{ij} + b_{ij}$

• $A \cdot B$ ✓ $\sum_{k=1}^n a_{ik} b_{kj}$

• Ax ✓
(x is a non ng. vector)

• A^{-1} ✗

$$A = \begin{bmatrix} 1 & 1 \\ 0 & 1 \end{bmatrix}$$

$$A^{-1} = \begin{bmatrix} 1 & -1 \\ 0 & 1 \end{bmatrix}$$

WHAT ABOUT EIGENVALUES/EIGENVECTORS?

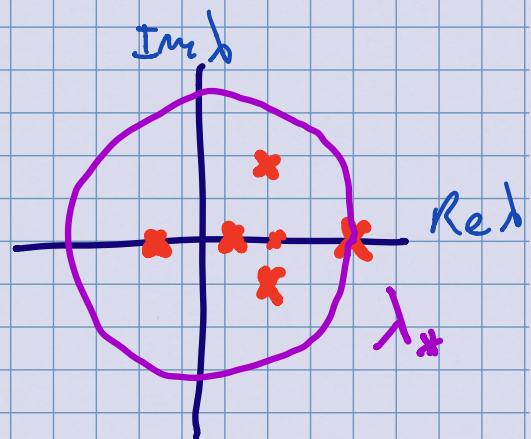
SOMETHING VERY INTERESTING HAPPENS!

ASSUME $A_{ij} > 0$ (strictly positive).

THEN, THERE EXISTS A REAL, POSITIVE

EIGENVALUE $\lambda_* > 0$ SUCH THAT:

- λ_* HAS MULTIPLICITY 1 (SINGLE EIGENVALUE).



- ALL OTHER EIGENVALUES HAVE SMALLER ABSOLUTE VALUE $|\lambda_i| < \lambda_*$
- THE EIGENVECTOR v_* ASSOCIATED w/ λ_* IS ALSO POSITIVE $(v_*)_i > 0$

EXAMPLE:

(Lecture 17)

$$A = \begin{bmatrix} 1 & 2 \\ 3 & 2 \end{bmatrix}$$

all
positive
entries

$$\begin{cases} \lambda_1 = 4 \\ \lambda_2 = -1 \end{cases}$$

SANITY CHECK!

TRACE?

DETERMINANT?

Eigenvalue $\lambda_1 = 4$ is positive ✓

Eigenvector $\begin{pmatrix} 2 \\ 3 \end{pmatrix}$ has positive entries

(What if I use Julia,
and it gives me the
eigenvector $\begin{pmatrix} -4 \\ -6 \end{pmatrix}$?)

Why is this true?

• Recall the power method

$$X_{k+1} = \text{normalize}(AX_k) = \frac{AX_k}{\|AX_k\|}$$

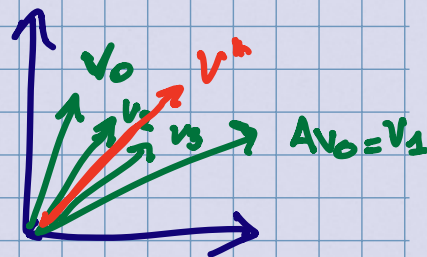
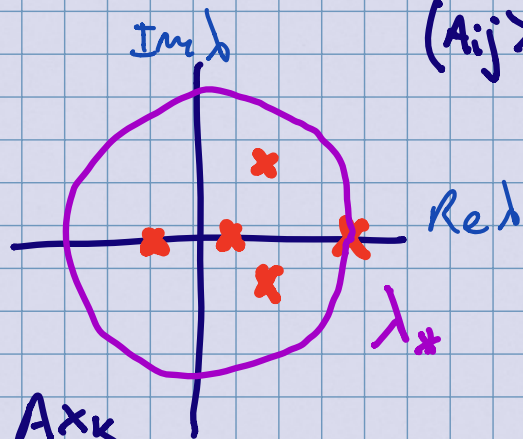
for a "random" initial vector X_0

⇒ Notice that if $X_0 > 0$ (componentwise),
then $X_k > 0$ for all k . $A_{ij} > 0$

⇒ X_k converges to eigenvector $V_* > 0$

⇒ The sequence $X_k^T A X_k \rightarrow \lambda_* > 0$

Eigenvalues of A
($A_{ij} > 0$)



MARKOV MATRICES

(also "STOCHASTIC" MATRICES).

A nonnegative matrix (as before),
with the additional property that
the sum of each row (or each column)
is equal to 1.

"row stochastic" or "column stochastic".

• Why do we care?

Because then (in either case)

the matrix has $\lambda = 1$ as an eigenvalue,
with the corresponding eigenvector nonnegative.

why?

$$\text{ROW STOCHASTIC} \Leftrightarrow A \begin{bmatrix} 1 \\ \vdots \\ 1 \end{bmatrix} = 1 \cdot \begin{bmatrix} 1 \\ \vdots \\ 1 \end{bmatrix} \Rightarrow \left(\begin{array}{l} \lambda = 1 \text{ is} \\ \text{an eigenvalue} \end{array} \right)$$

$$\text{and } \lambda(A) = \lambda(A^T)$$

EXAMPLE:

$$A = \frac{1}{4} \begin{bmatrix} 1 & 2 \\ 3 & 2 \end{bmatrix} \quad \left(\begin{array}{c} \text{column} \\ \text{stochastic} \end{array} \right)$$

$$\Rightarrow \lambda_* = 1 \quad V_* = \begin{pmatrix} 2/5 \\ 3/5 \end{pmatrix}$$

normalized so
all entries add up to 1.

Transpose

$$B = \frac{1}{4} \begin{bmatrix} 1 & 3 \\ 2 & 2 \end{bmatrix} \begin{array}{l} = 1 \\ = 1 \end{array} \quad \left(\begin{array}{c} \text{row - stochastic} \end{array} \right)$$

$$\Rightarrow \lambda_* = 1 \quad V_* = \begin{pmatrix} 1 \\ 1 \end{pmatrix}$$

Surprising?

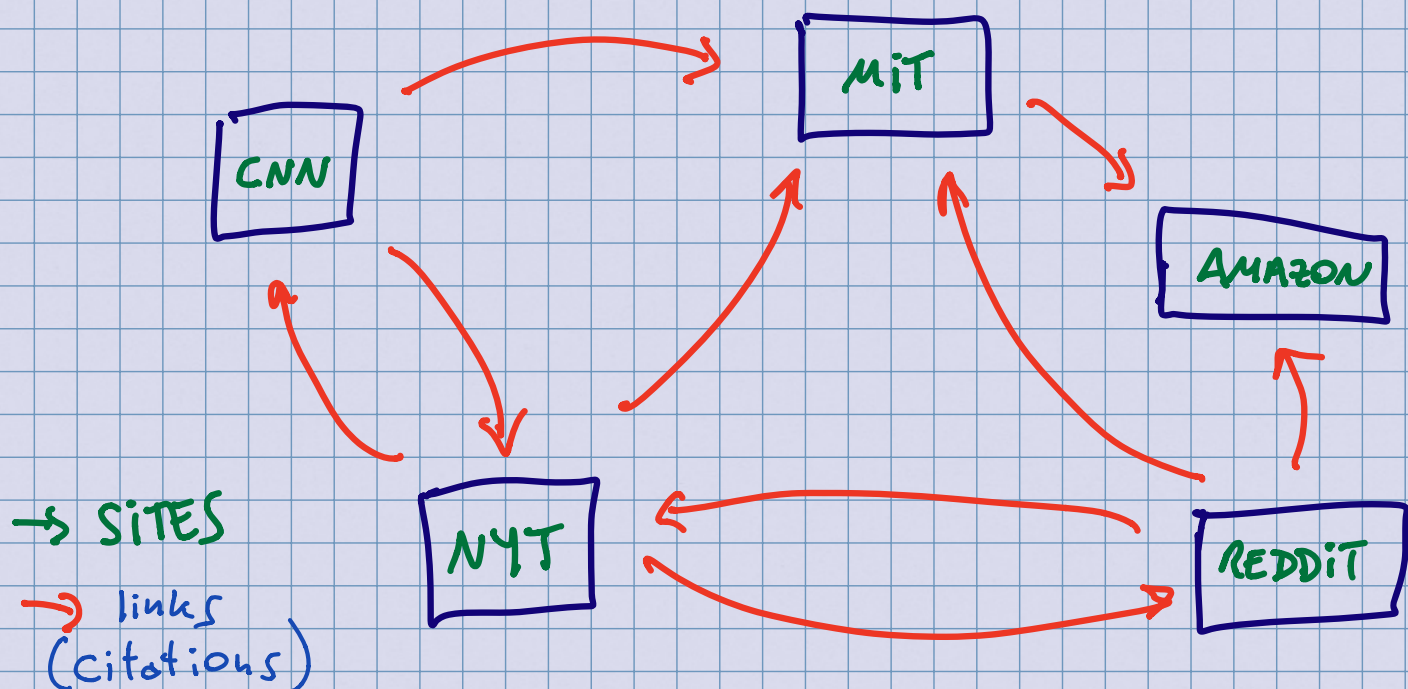
PAGE RANK :

How does a search engine (e.g. Google) rank websites?

What makes a site more "important" than others?

many models (PageRank, HITS, etc...)

Here's a simple one:



(Def)

THE IMPORTANCE (RANK) OF A SITE
DEPENDS ON HOW MANY (AND WHICH)
IMPORTANT SITES POINT TO IT.

- RECURSIVE (OR "FIXED POINT" DEFINITION)

typo!

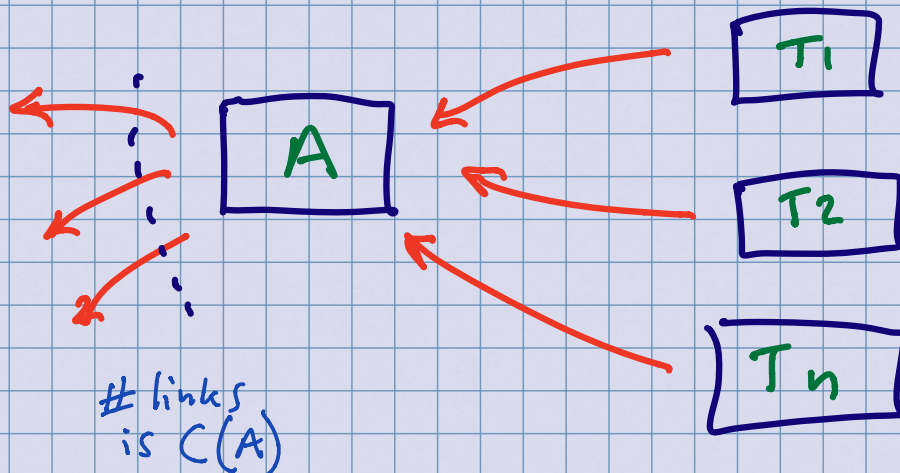
(from the Brin/Pager paper)

We assume page A has pages $T_1 \dots T_n$ which point to it (i.e., are citations). The parameter d is a damping factor which can be set between 0 and 1. We usually set d to 0.85. There are more details about d in the next section. Also $C(A)$ is defined as the number of links going out of page A . The PageRank of a page A is given as follows:

$$PR(A) = \frac{(1-d)}{N} + d (PR(T_1)/C(T_1) + \dots + PR(T_n)/C(T_n))$$

linear equations in PR

Note that the PageRanks form a probability distribution over web pages, so the sum of all web pages' PageRanks will be one.

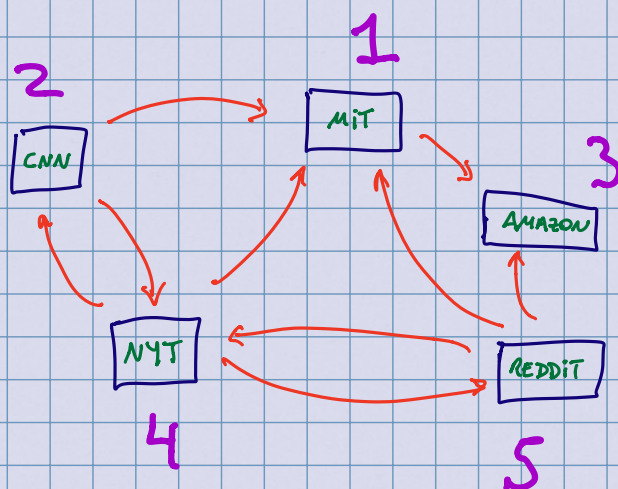


Writing it as a matrix... (linear system)

$$\begin{bmatrix} PR(T_1) \\ \vdots \\ PR(T_n) \end{bmatrix} = \frac{1}{N} \begin{bmatrix} 1-d \\ 1-d \\ \vdots \\ 1-d \end{bmatrix} + d \cdot \begin{bmatrix} \text{WEIGHTED} \\ \text{ADJACENCY} \\ \text{MATRIX OF} \\ \text{GRAPH} \end{bmatrix} \begin{bmatrix} PR(T_1) \\ \vdots \\ PR(T_n) \end{bmatrix}$$

$$\begin{bmatrix} 0 & 1/2 & 0 & 1/3 & 1/3 \\ 0 & 0 & 0 & 1/3 & 0 \\ 1 & 0 & 0 & 0 & 1/3 \\ 0 & 1/2 & 0 & 0 & 1/3 \\ 0 & 0 & 1 & 1/3 & 0 \end{bmatrix}$$

(columns add up to 1)



- Parameter d chosen appropriately (mostly, to ensure matrix is strictly positive)

for $d = 0.85 \Rightarrow PR \approx (0.19, 0.07, 0.28, 0.15, 0.31)$

- key ^{practical} difficulty: this is a HUGE problem! (~ 2 Billion websites)

As an eigenvalue problem:

$$\left(\begin{array}{l} p \text{ is normalized} \\ \text{so } e^T p = 1, \\ \text{where } e = \begin{pmatrix} 1 \\ \vdots \\ 1 \end{pmatrix} \end{array} \right)$$

From previous page:

$$\begin{bmatrix} PR(T_1) \\ \vdots \\ PR(T_n) \end{bmatrix} = \frac{1}{N} \begin{bmatrix} 1-d \\ 1-d \\ \vdots \\ 1-d \end{bmatrix} + d \cdot \begin{bmatrix} \text{WEIGHTED} \\ \text{ADJACENCY} \\ \text{MATRIX OF} \\ \text{GRAPH} \end{bmatrix} \begin{bmatrix} PR(T_1) \\ \vdots \\ PR(T_n) \end{bmatrix}$$

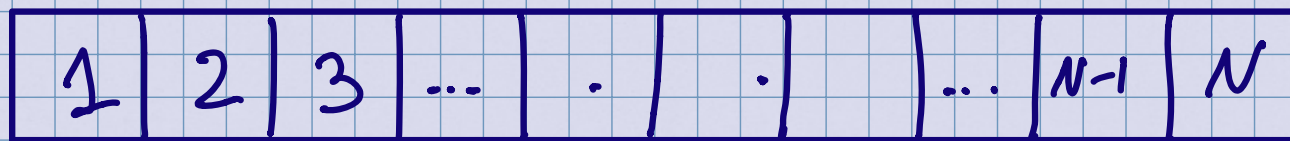
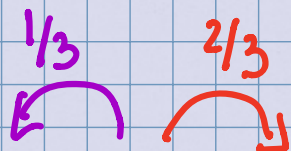
$$p = \left(\frac{1-d}{N} \right) e + d G p$$

$$p = \left[\left(\frac{1-d}{N} \right) e e^T + d G \right] p$$

(since $e^T p = 1$)

Makes every entry
strictly positive

HOPPING RABBIT



- AT EACH TIME STEP, THE BUNNY FLIPS A COIN ^(BIASED).
 - WITH PROB $\frac{1}{3}$ JUMPS TO THE **LEFT**
 - WITH PROB $\frac{2}{3}$ JUMPS TO THE **RIGHT**
- IF HE REACHES THE ENDS, BOUNCES BACK
 - IF AT BOX 1, AND JUMPS LEFT, REMAINS AT 1
 - IF AT BOX N, AND JUMPS RIGHT, REMAINS AT N

THE RABBIT HAS BEEN JUMPING ALL DAY LONG,
AND YOU SHOW UP LATE ONE AFTERNOON
TO VISIT HIM.

(ALTERNATIVE INTERPRETATION)

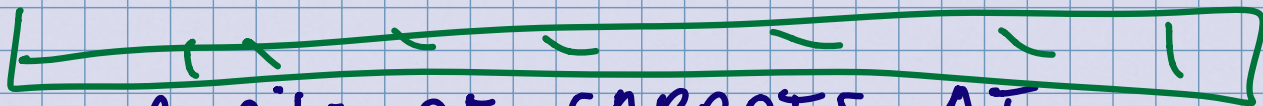
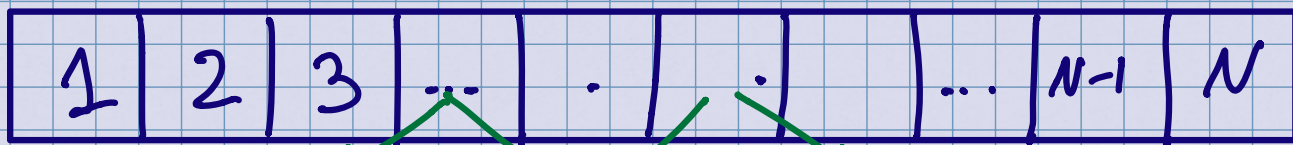
$\frac{1}{3}$ $\frac{2}{3}$



$\frac{1}{3}$ $\frac{2}{3}$



$\frac{1}{3}$ $\frac{2}{3}$



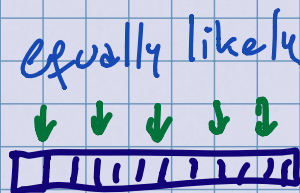
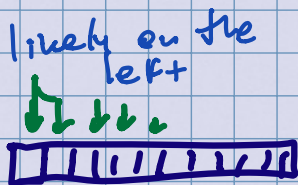
- A PILE OF CARROTS AT
EVERY BOX

- EVERY TIMESTEP, WE SPLIT
THEM.

IF YOU DON'T LIKE RABBITS, YOU CAN THINK, E.G.,
ABOUT PARTICLES (ELECTRONS ON A LATTICE)
UNDER AN EXTERNAL FIELD.

(OR MANY OTHER THINGS)

Q: WHERE WOULD YOU EXPECT TO FIND THE RABBIT?
(e.g., left-side vs. right side?)



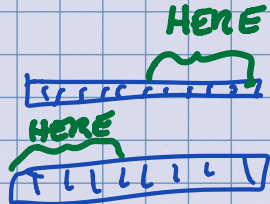
SAY $N=10$.

Q: IF I BET YOU WITH ODDS 1 TO 1
THAT THE RABBIT IS ON THE RIGHT HALF,
WOULD YOU ACCEPT THE BET?

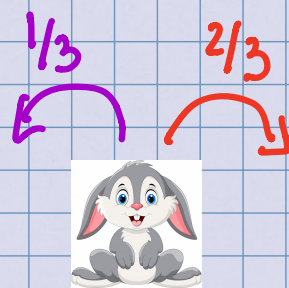
2 to 1?

10 to 1?

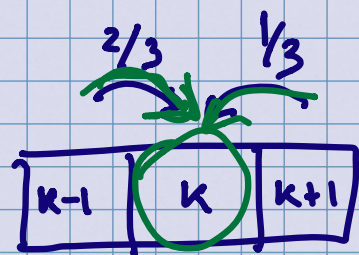
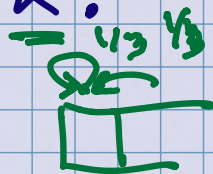
100 to 1?

(x to 1 :  , YOU PAY ME 1
I PAY YOU x)

HOW TO SOLVE THIS?



Let p_k be the probability that the rabbit is in box k .



Then:

$$p_k = \frac{2}{3} p_{k-1} + \frac{1}{3} p_{k+1}$$

(for $2 \leq k \leq N-1$)

$$\begin{bmatrix} p_1 \\ p_2 \\ \vdots \\ p_k \\ \vdots \\ p_N \end{bmatrix} = \begin{bmatrix} \cancel{1/3} & \cancel{2/3} & & & \\ 2/3 & 0 & 1/3 & & \\ & 2/3 & 0 & 1/3 & \\ & & 2/3 & 0 & 1/3 \\ & & & \ddots & \ddots \\ & & & & 2/3 & 0 & 1/3 \\ & & & & & 2/3 & 2/3 \end{bmatrix} \begin{bmatrix} p_1 \\ p_2 \\ \vdots \\ p_k \\ \vdots \\ p_N \end{bmatrix}$$

$p = A p$

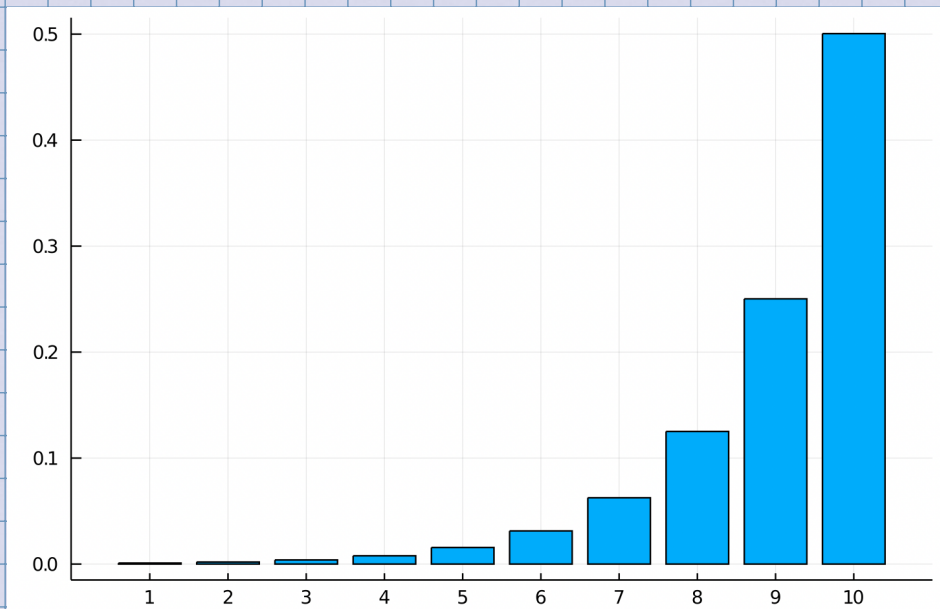
(and $\sum p_i = 1$)

- A is a non negative matrix
- Every column adds up to 1.

The vector $\mathbf{p}$ of probabilities
is the solution of a linear system.

Alternatively, it is the eigenvector of $\mathbf{A}$
associated with the eigenvalue 1.

In our case ($N=10$), we have



3%

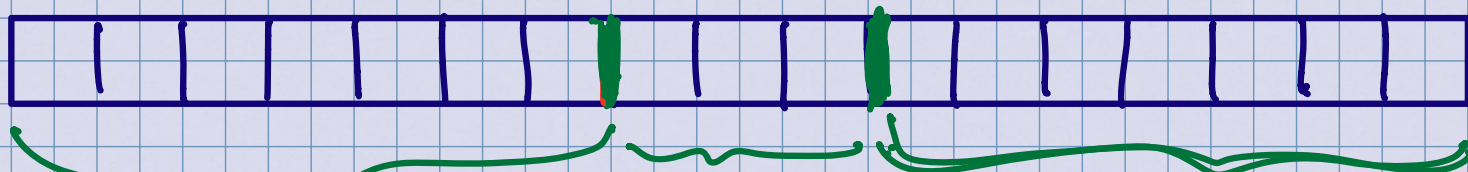
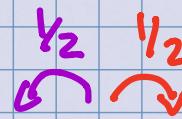
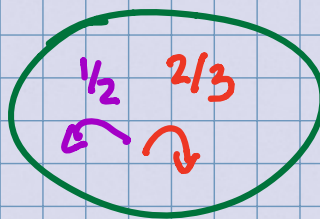
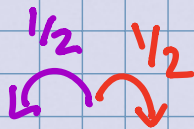
97%

probabilities
 p_i
(after solving
the linear
system)

"Fair"
odds
 $\Rightarrow$ are
 $\sim 1:32$

CHALLENGE!

"TUNNEL EFFECT"



REGION 1

REGION 2

REGION 3

(EQUALLY
LIKELY)

(BIASED
TO RIGHT)

(EQUALLY
LIKELY)

- WHAT DO YOU EXPECT WILL HAPPEN?
- HOW DOES THE SIZE OF THE MIDDLE REGION AFFECT THE RESULTS?

"MARKOV CHAINS"