Review

Now...

 $P(0) \land ((\forall n)(P(n) \implies P(n+1) \equiv (\forall n \in N) P(n).$

 $P(0) \land ((\forall n)(P(n) \Longrightarrow P(n+1) \equiv (\forall n \in N) P(n).$ Thm: For all $n \ge 1$, $8|3^{2n} - 1$.

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Induction on n.

 $P(0) \wedge ((\forall n)(P(n) \Longrightarrow P(n+1) \equiv (\forall n \in N) P(n).$ **Thm:** For all $n \ge 1$, $8|3^{2n} - 1$. Induction on n. Base: $8|3^2 - 1$.

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Induction Step: Prove P(n+1)
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3^{2n+2}-1 =
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3^{2n+2} - 1 = 9(3^{2n}) - 1
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 $P(0) \land ((\forall n)(P(n) \Longrightarrow P(n+1) \equiv (\forall n \in N) P(n).$ **Thm:** For all $n \ge 1$, $8|3^{2n} - 1$.

Induction on n.

Base: 8|3² - 1.

Induction Hypothesis: Assume P(n): True for some n. ($3^{2n} - 1 = 8d$)

Induction Step: Prove P(n+1)

 $3^{2n+2} - 1 = 9(3^{2n}) - 1$ (by induction hypothesis)

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 $3^{2n+2} - 1 = 9(3^{2n}) - 1$ (by induction hypothesis) = 9(8d + 1) - 1

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$$3^{2n+2} - 1 = 9(3^{2n}) - 1$$
 (by induction hypothesis)
= $9(8d+1) - 1$
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 $P(0) \land ((\forall n)(P(n) \Longrightarrow P(n+1) \equiv (\forall n \in N) P(n).$ Thm: For all $n \ge 1$, $8|3^{2n} - 1$.

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$$3^{2n+2} - 1 = 9(3^{2n}) - 1$$
 (by induction hypothesis)
= $9(8d + 1) - 1$
= $72d + 8$
= $8(9d + 1)$

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Divisible by 8.

n-men, n-women.

n-men, n-women.

Each person has completely ordered preference list

n-men, n-women.

Each person has completely ordered preference list contains every person of opposite gender.

n-men, n-women.

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Pairing.

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Pairing.

Set of pairs (m_i, w_j) containing all people *exactly* once.

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Pairing.

Set of pairs (m_i, w_j) containing all people *exactly* once. How many pairs?

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Pairing.

Set of pairs (m_i, w_j) containing all people *exactly* once. How many pairs? *n*.

People in pair are **partners** in pairing.

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Rogue Couple in a pairing.

A m_i and w_k who like each other more than their partners

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Stable Pairing.

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Pairing with no rogue couples.

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Pairing with no rogue couples.

Does stable pairing exist?

No, for roommates problem.

Traditional Marriage Algorithm:

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Each Day:

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All men propose to favorite woman who has not yet rejected him.

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Useful Algorithmic Definitions:

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Man crosses off woman who rejected him.

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Useful Algorithmic Definitions: Man **crosses off** woman who rejected him. Woman's current proposer is "**on string.**"

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"Propose and Reject."

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- \implies M proposed to W
- \implies W ended up with someone she liked better than *M*.

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rogue couple (M,W)

 \implies M proposed to W

 \implies W ended up with someone she liked better than *M*. Not rogue couple!

G = (V, E)

G = (V, E)V - set of vertices.

 $\begin{array}{l} G = (V, E) \\ V \text{ - set of vertices.} \\ E \subseteq V \times V \text{ - set of edges.} \end{array}$

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Adjacent, Incident, Degree.

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Thm: Sum of degrees is 2|E|.

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Thm: Sum of degrees is 2|E|. Edge is incident to 2 vertices.

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Pair of Vertices are Connected:

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Adjacent, Incident, Degree. In-degree, Out-degree.

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Pair of Vertices are Connected: If there is a path between them.

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Connected Component: maximal set of connected vertices.

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Pair of Vertices are Connected: If there is a path between them.

Connected Component: maximal set of connected vertices.

Connected Graph: one connected component.

Graph Algorithm: Eulerian Tour

Thm: Every connected graph where every vertex has even degree has an Eulerian Tour; a tour which visits every edge exactly once.

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Recurse on connected components.

Thm: Every connected graph where every vertex has even degree has an Eulerian Tour; a tour which visits every edge exactly once.

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Recurse on connected components. Put together.

Thm: Every connected graph where every vertex has even degree has an Eulerian Tour; a tour which visits every edge exactly once.

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Recurse on connected components. Put together.

Property: walk visits every component.

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Recurse on connected components. Put together.

Property: walk visits every component. Proof Idea: Original graph connected.

Thm: Every connected graph where every vertex has even degree has an Eulerian Tour; a tour which visits every edge exactly once.

Algorithm:

Take a walk using each edge at most once.

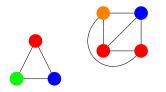
Property: return to starting point. Proof Idea: Even degree.

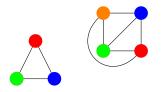
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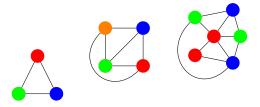
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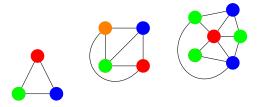


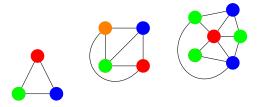




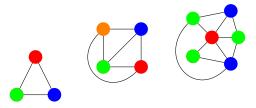






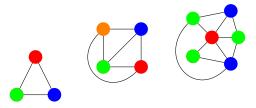


Given G = (V, E), a coloring of a *G* assigns colors to vertices *V* where for each edge the endpoints have different colors.



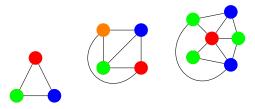
Notice that the last one, has one three colors.

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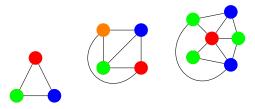
Notice that the last one, has one three colors. Fewer colors than number of vertices.

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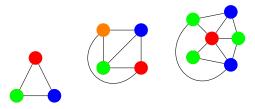
Notice that the last one, has one three colors. Fewer colors than number of vertices. Fewer colors than max degree node.

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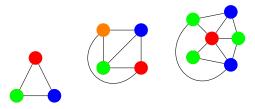
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Interesting things to do.

Given G = (V, E), a coloring of a *G* assigns colors to vertices *V* where for each edge the endpoints have different colors.

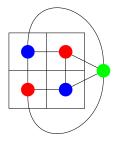


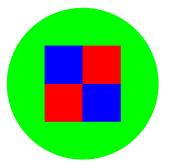
Notice that the last one, has one three colors. Fewer colors than number of vertices. Fewer colors than max degree node.

Interesting things to do. Algorithm!

Planar graphs and maps.

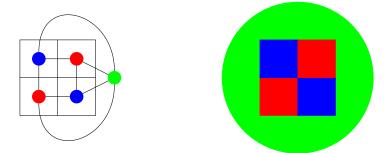
Planar graph coloring \equiv map coloring.





Planar graphs and maps.

Planar graph coloring \equiv map coloring.



Four color theorem is about planar graphs!

Theorem: Every planar graph can be colored with six colors.

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Recall: $e \le 3v - 6$ for any planar graph where v > 2.

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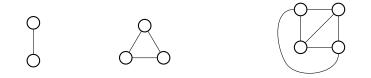
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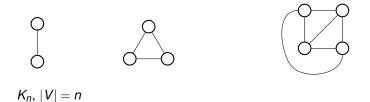
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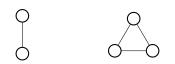
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 $K_n, |V| = n$

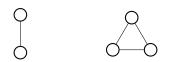
every edge present.





 $K_n, |V| = n$

every edge present. degree of vertex?





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Very connected.





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Very connected. Lots of edges:

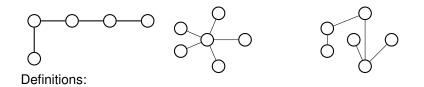




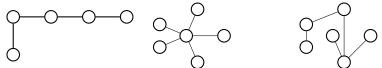
 $K_n, |V| = n$

every edge present. degree of vertex? |V| - 1.

Very connected. Lots of edges: n(n-1)/2.



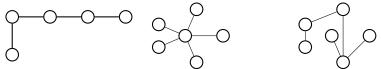




Definitions:

A connected graph without a cycle.



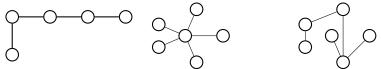


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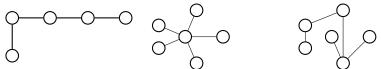


Definitions:

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A connected graph where any edge removal disconnects it.



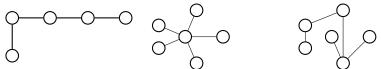
Definitions:

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An acyclic graph where any edge addition creates a cycle.



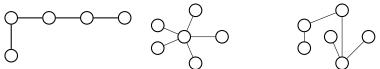
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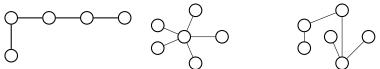
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Property:

Can remove a single node and break into components of size at most |V|/2.

Hypercubes.

Hypercubes. Really connected.

Hypercubes. Really connected. $|V|\log|V|$ edges!

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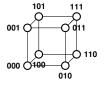
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0 1 O----C





Arithmetic modulo *m*. Elements of equivalence classes of integers.

Arithmetic modulo m. Elements of equivalence classes of integers. $\{0, \ldots, m-1\}$

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Arithmetic modulo m.
Elements of equivalence classes of integers.
\{0, \ldots, m-1\}
and integer i \equiv a \pmod{m}
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Additive inverses are intuitively negative numbers.

 $3^{-1} \pmod{7}?$

3⁻¹ (mod 7)? 5

 $3^{-1} \pmod{7}? 5$ $5^{-1} \pmod{7}?$

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Proof: a and b inverses of x (mod n)

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See,

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See,... no inverse!

x has inverse modulo m if and only if gcd(x,m) = 1.

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Group structures more generally.

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Group structures more generally.

Proof Idea:

 $\{0x, \ldots, (m-1)x\}$ are distinct modulo *m* if and only if gcd(x, m) = 1.

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Extended-gcd(x, y)

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Extended-gcd(x, y) returns (d, a, b)

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Extended-gcd(x, y) returns (d, a, b)
d = gcd(x, y) and d = ax + by
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Extended-gcd(x, y) returns (d, a, b)
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Multiplicative inverse of (x, m).

x has inverse modulo m if and only if gcd(x,m) = 1.

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Confirm: -119 + 120 = 1 $d = e^{-1} = -17 = 43 = \pmod{60}$

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S(C) = D(C).

RSA:

$$N = p, q$$

 e with gcd $(e, (p-1)(q-1))$.
 $d = e^{-1} \pmod{(p-1)(q-1)}$.

Public Key Cryptography:

 $D(E(m,K),k) = (m^e)^d \mod N = m.$

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3⁶ (mod 7)?



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3⁶ (mod 7)? 1. Fermat: p = 7, p - 1 = 6

```
3^6 \pmod{7}? 1. Fermat: p = 7, p - 1 = 6
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21 = (3)(7)
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2<sup>1</sup>4 (mod 21)?
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```

Sum Rule: For disjoint sets *S* and *T*, $|S \cup T| = |S| + |T|$

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Answer: $|S| + |T| - |S \cap T| = 10^9 + 10^9 - 10^8$.



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Why?

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Halt is undecidable.

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Reductions from Halt give more undecidable problems.

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Reductions use program for problem A to solve HALT.

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Reductions from Halt give more undecidable problems.

Reductions use program for problem A to solve HALT.

Concept 1: can call program A

Concept 2:One can modify text of input program (to HALT).

Probability Review

1. True or False

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CS70: Review of Probability.

Probability Review

- 1. True or False
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- 6. Common Mistakes

CS70: Review of Probability.

Probability Review

- 1. True or False
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- 3. Quiz 1: G
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- 6. Common Mistakes

• Ω and *A* are independent.

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$$X_1, \ldots, X_n$$
 i.i.d. $\implies var(\frac{X_1 + \cdots + X_n}{n}) = var(X_1)$.

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$$X_1, ..., X_n$$
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 i.i.d. $\implies \frac{X_1 + \dots + X_n - nE[X_1]}{n\sigma(X_1)} \rightarrow \mathcal{N}(0, 1).$

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 i.i.d. $\implies \frac{X_1 + \dots + X_n - nE[X_1]}{n\sigma(X_1)} \rightarrow \mathcal{N}(0, 1)$. False: \sqrt{n}

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$$Pr[A \setminus B] \ge Pr[A] - Pr[B]$$
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$$X_1, ..., X_n$$
 i.i.d. $\implies var(\frac{X_1 + \dots + X_n}{n}) = var(X_1)$. False: $\times \frac{1}{n}$

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$$\frac{\exp\{-\lambda 5\}}{\exp\{-\lambda 3\}} = \exp\{-\lambda 2\}.$$

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WLLN

▶ WLLN (7)

- WLLN (7)
- MMSE

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- MMSE (6)

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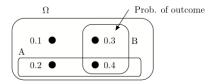
- WLLN (7)
- MMSE (6)
- Projection property (8)
- Chebyshev (2)

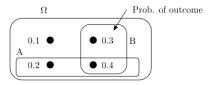
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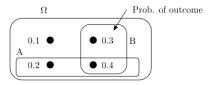
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- Markov's inequality

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- Markov's inequality (1)



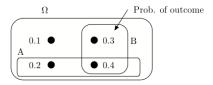


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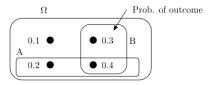


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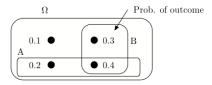
Pr[A|B] =



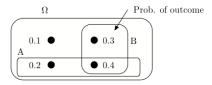
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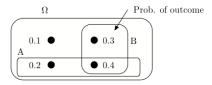


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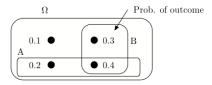


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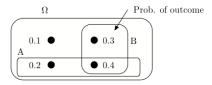
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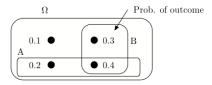
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- 3. Are A and B positively correlated?

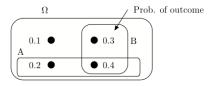


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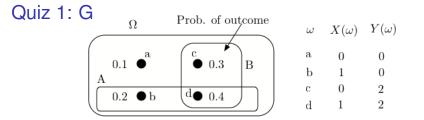
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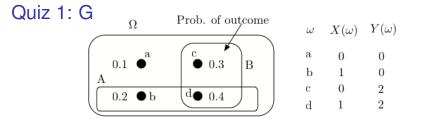
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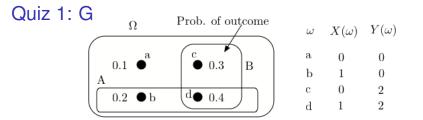
No. $Pr[A \cap B] = 0.4 < Pr[A]Pr[B] = 0.6 \times 0.7$.

Quiz 1: G Prob. of outcome Ω $Y(\omega)$ $X(\omega)$ ω ¥ $ullet^{\mathrm{a}}$ с а 0 0 0.10.3 В b 1 0 А $\frac{2}{2}$ с 0 d● 0.4 0.2 **•** b d 1

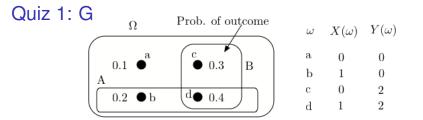




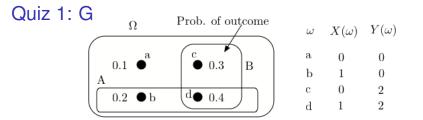
E[Y|X = 0] =



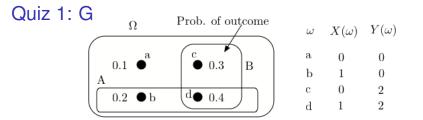
 $E[Y|X=0] = 0 \times Pr[Y=0|X=0] + 2 \times Pr[Y=2|X=0]$



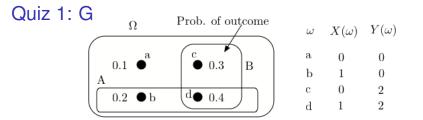
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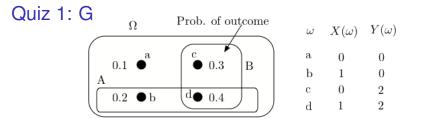
 $E[Y|X=0] = 0 \times Pr[Y=0|X=0] + 2 \times Pr[Y=2|X=0]$ = $2 \times \frac{0.3}{0.4} =$



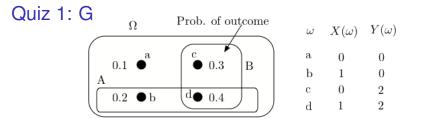
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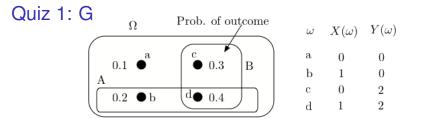
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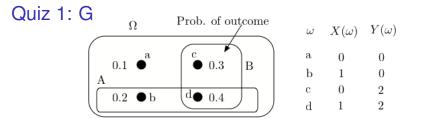
=

4. What is E[Y|X]?

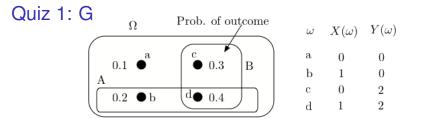
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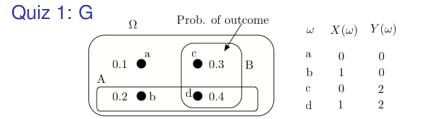


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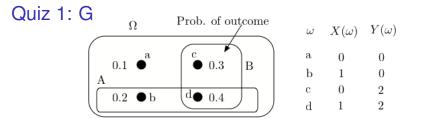
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5. What is cov(X, Y)?



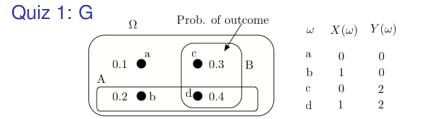
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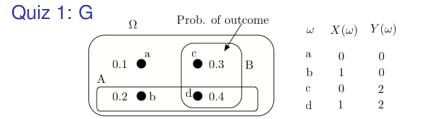
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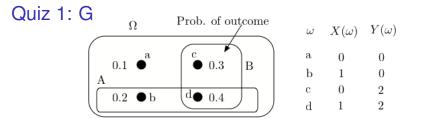
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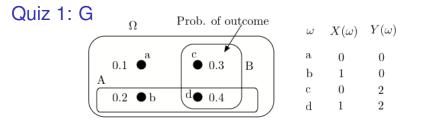


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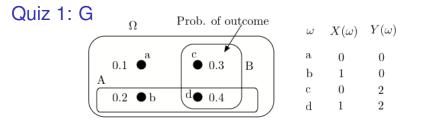
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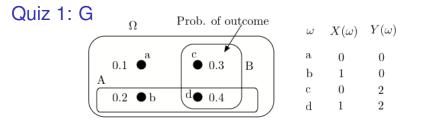
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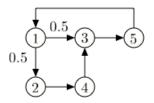
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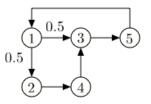


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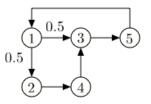


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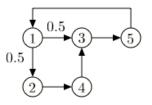




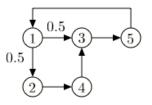
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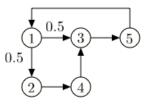


- 7. Is this Markov chains irreducible? Yes.
- 8. Is this Markov chain periodic?



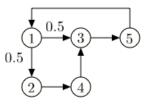
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No.

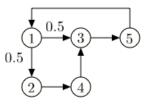


- 7. Is this Markov chains irreducible? Yes.
- 8. Is this Markov chain periodic?

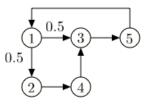
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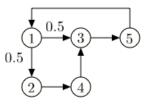
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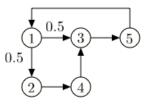
9. Does π_n converge to a value independent of π_0 ?



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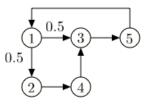
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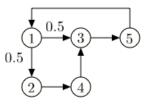
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- 10. Does $\frac{1}{n}\sum_{m=1}^{n-1} 1\{X_m = 3\}$ converge as $n \to \infty$?



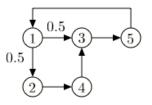
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- 11. Calculate π .

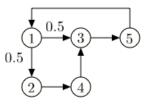


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Let $a = \pi(1)$.

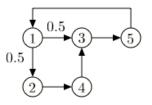


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Let $a = \pi(1)$. Then $a = \pi(5)$,

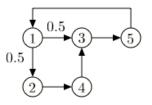


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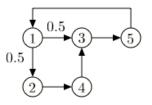


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Let $a = \pi(1)$. Then $a = \pi(5), \pi(2) = 0.5a, \pi(4) = \pi(2) = 0.5a$,

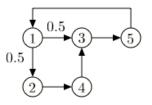


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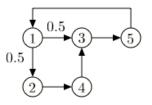


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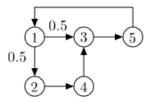


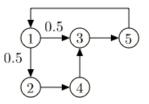
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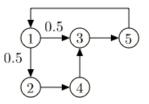
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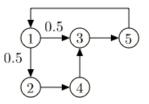


12. Write the first step equations for calculating the mean time from 1 to 4.



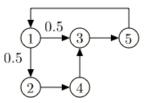
12. Write the first step equations for calculating the mean time from 1 to 4.

 $\beta(1) = 1 + 0.5\beta(2) + 0.5\beta(3)$



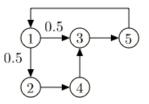
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 $\beta(1) = 1 + 0.5\beta(2) + 0.5\beta(3)$ $\beta(2) = 1$



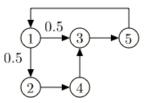
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eta(1) = 1 + 0.5eta(2) + 0.5eta(3)eta(2) = 1eta(3) = 1 + eta(5)



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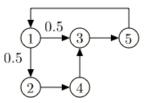
$$\begin{split} \beta(1) &= 1 + 0.5\beta(2) + 0.5\beta(3) \\ \beta(2) &= 1 \\ \beta(3) &= 1 + \beta(5) \\ \beta(5) &= 1 + \beta(1). \end{split}$$



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13. Solve these equations.

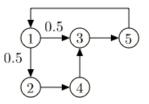


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13. Solve these equations.

 $\beta(1) = 1 + 0.5 \times 1 + 0.5 \times (1 + (1 + \beta(1)))$

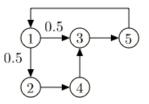


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13. Solve these equations.

$$\begin{aligned} \beta(1) &= 1 + 0.5 \times 1 + 0.5 \times (1 + (1 + \beta(1))) \\ &= 2.5 + 0.5\beta(1). \end{aligned}$$



12. Write the first step equations for calculating the mean time from 1 to 4.

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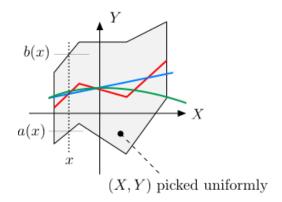
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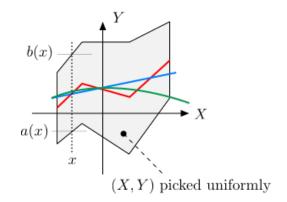
Hence, $\beta(1) = 5$.

14. Which is E[Y|X]? Blue, red or green?

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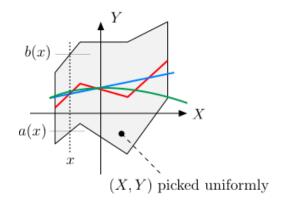


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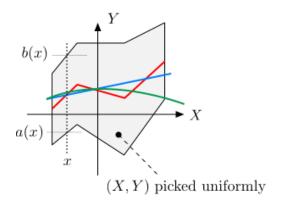
Answer: Red.

14. Which is E[Y|X]? Blue, red or green?



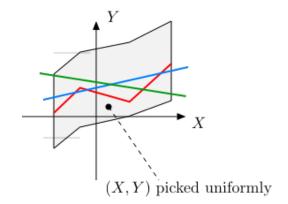
Answer: Red. Given X = x, Y = U[a(x), b(x)].

14. Which is E[Y|X]? Blue, red or green?

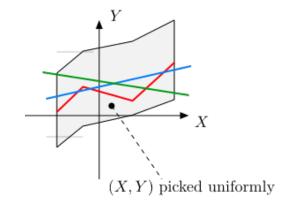


Answer: Red. Given X = x, Y = U[a(x), b(x)]. Thus, $E[Y|X = x] = \frac{a(x)+b(x)}{2}$.

15. Which is L[Y|X]? Blue, red or green?

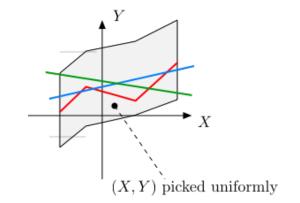


15. Which is L[Y|X]? Blue, red or green?



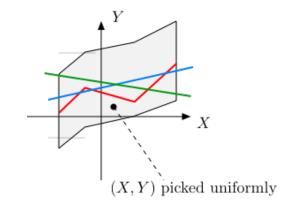
Answer: Blue.

15. Which is L[Y|X]? Blue, red or green?

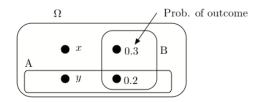


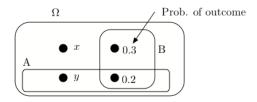
Answer: Blue. Cannot be red (not a straight line).

15. Which is L[Y|X]? Blue, red or green?

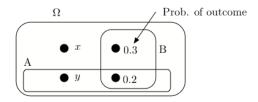


Answer: Blue. Cannot be red (not a straight line). Cannot be green: *X* and *Y* are clearly positively correlated.



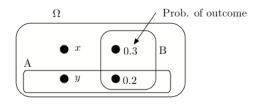


1. Find (x, y) so that A and B are independent.



 Find (x,y) so that A and B are independent. We need

 $Pr[A \cap B] = Pr[A]Pr[B]$

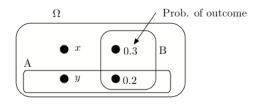


1. Find (*x*, *y*) so that *A* and *B* are independent. We need

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That is,

 $0.2 = (y + 0.2) \times 0.5 =$

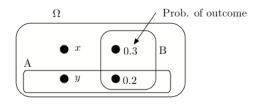


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 $Pr[A \cap B] = Pr[A]Pr[B]$

That is,

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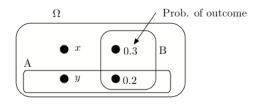
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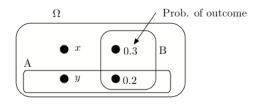
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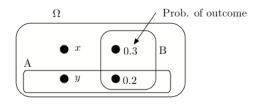
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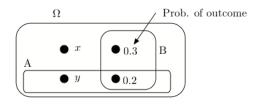
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 and $x = 0.3$.

2. Find the value of x that maximizes Pr[B|A].



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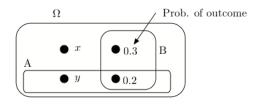
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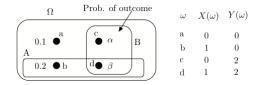
That is,

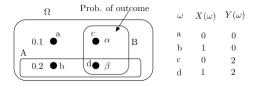
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Hence,

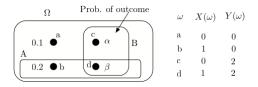
$$y = 0.2$$
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2. Find the value of x that maximizes Pr[B|A]. Obviously, it is x = 0.5.



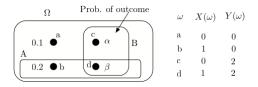


3. Find α so that X and Y are independent.



Find α so that X and Y are independent.
 We need

$$Pr[X = 0, Y = 0] = Pr[X = 0]Pr[Y = 0]$$

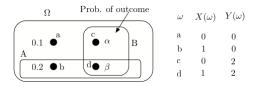


Find α so that X and Y are independent.
 We need

$$Pr[X = 0, Y = 0] = Pr[X = 0]Pr[Y = 0]$$

That is,

 $0.1 = (0.1 + \alpha) \times (0.1 + 0.2) =$

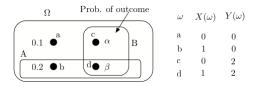


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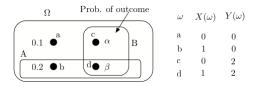
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4. A CS70 student is great w.p. 0.3 and good w.p. 0.7.

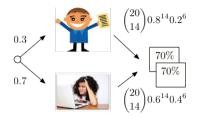
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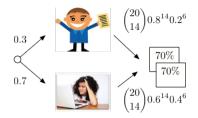
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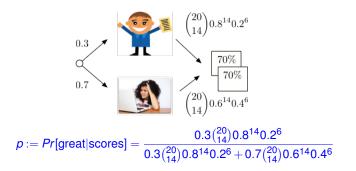
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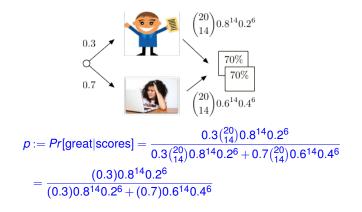


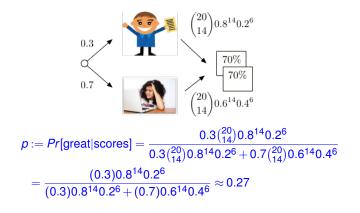
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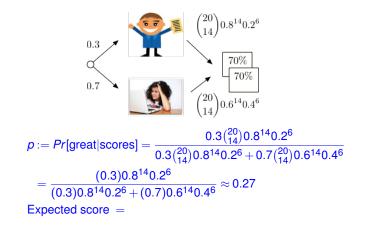


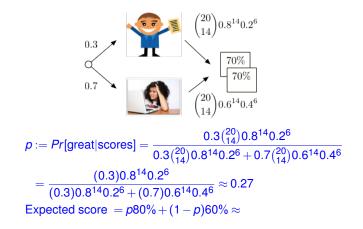
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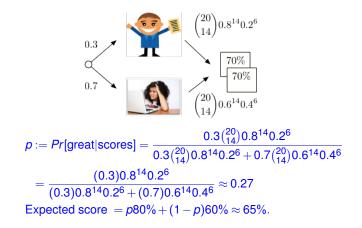












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Hint: If $X = Expo(\lambda)$, $f_X(x) = \lambda e^{-\lambda x} \mathbb{1}\{x > 0\}$, $E[X] = 1/\lambda$.

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, $f_X(x) = \lambda e^{-\lambda x} \mathbb{1}\{x > 0\}$, $E[X] = 1/\lambda$.

(a)
$$p := Pr[G|X \in (0.6, 0.6 + \delta)]$$

 $= \frac{0.5Pr[X \in (0.6, 0.6 + \delta)|G]}{0.5Pr[X \in (0.6, 0.6 + \delta)|G] + 0.5Pr[X \in (0.6, 0.6 + \delta)|D]}$
 $= \frac{e^{-0.6}\delta}{e^{-0.6}\delta + (0.8)^{-1}e^{-(0.8)^{-1}0.6}\delta} \approx 0.488.$
(b) $E[$ lifespan of other bulb $] =$

 The lifespans of good lightbulbs are exponentially distributed with mean 1 year. Those of defective bulbs are exponentially distributed with mean 0.8. All the bulbs in one batch are equally likely to be good or defective. You test one bulb and note that it burns out after 0.6 year. (a) What is the probability you got a batch of good bulbs? (b) What is the expected lifespan of another bulb in that batch?

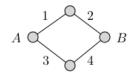
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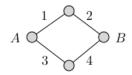
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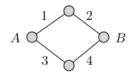
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(b)
$$E[\text{ lifespan of other bulb }] = p \times 1 + (1-p) \times 0.8 \approx 0.9.$$



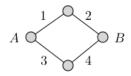


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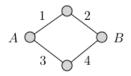
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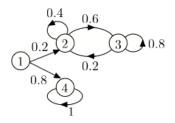
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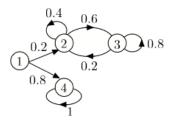
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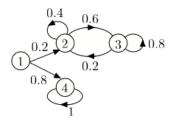
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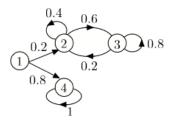




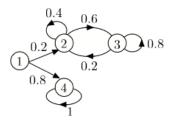
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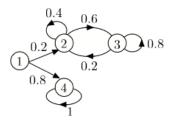
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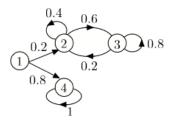
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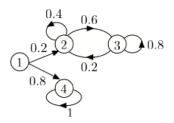


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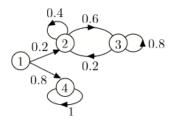
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