Probability Review

- 1. True or False
- 2. Some Key Results

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- 3. Quiz 1: G

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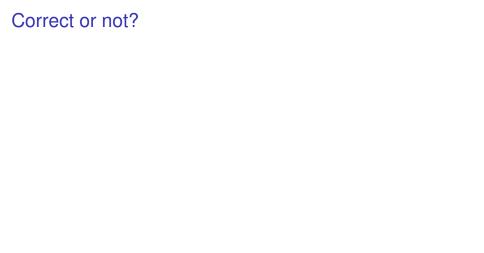
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. [5] $E[Y] + \frac{cov(X, Y)}{var(X)}(X - E[X])$. [2] $Pr[|X - E[X]| > a] \le \frac{var[X]}{a^2}$ [6] $\sum_{y} y Pr[Y = y | X = x]$ [7] $Pr[|X_1 + \dots + X_n - E[X_1]| \ge \varepsilon] \to 0$, [4] $g(\cdot)$ convex $\Rightarrow E[g(X)] \ge g(E[X])$ [8] $E[(Y - E[Y|X])h(X)] = 0$.

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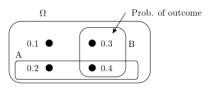
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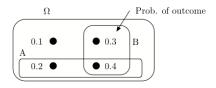
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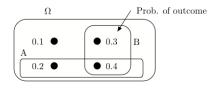
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- ▶ LLSE (5)
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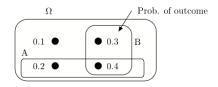
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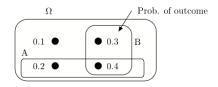




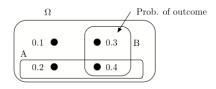
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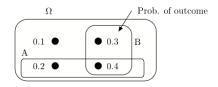


$$Pr[A|B] = \frac{Pr[A \cap B]}{Pr[B]} = \frac{0.4}{0.7}$$



1. What is P[A|B]?

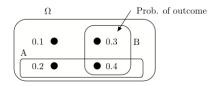
$$Pr[A|B] = \frac{Pr[A \cap B]}{Pr[B]} = \frac{0.4}{0.7}$$



1. What is P[A|B]?

$$Pr[A|B] = \frac{Pr[A \cap B]}{Pr[B]} = \frac{0.4}{0.7}$$

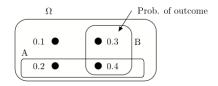
$$Pr[B|A] =$$



1. What is P[A|B]?

$$Pr[A|B] = \frac{Pr[A \cap B]}{Pr[B]} = \frac{0.4}{0.7}$$

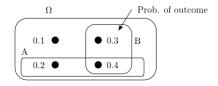
$$Pr[B|A] = \frac{Pr[A \cap B]}{Pr[A]} =$$



1. What is P[A|B]?

$$Pr[A|B] = \frac{Pr[A \cap B]}{Pr[B]} = \frac{0.4}{0.7}$$

$$Pr[B|A] = \frac{Pr[A \cap B]}{Pr[A]} = \frac{0.4}{0.6}$$



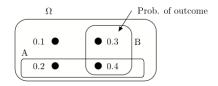
1. What is P[A|B]?

$$Pr[A|B] = \frac{Pr[A \cap B]}{Pr[B]} = \frac{0.4}{0.7}$$

2. What is Pr[B|A]?

$$Pr[B|A] = \frac{Pr[A \cap B]}{Pr[A]} = \frac{0.4}{0.6}$$

3. Are A and B positively correlated?



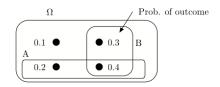
1. What is P[A|B]?

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Are A and B positively correlated?No.



1. What is P[A|B]?

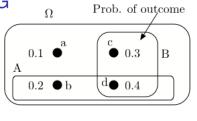
$$Pr[A|B] = \frac{Pr[A \cap B]}{Pr[B]} = \frac{0.4}{0.7}$$

What is Pr[B|A]?

$$Pr[B|A] = \frac{Pr[A \cap B]}{Pr[A]} = \frac{0.4}{0.6}$$

3. Are A and B positively correlated?

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



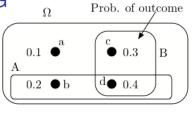
 $\begin{array}{ccccc} \omega & X(\omega) & Y(\omega) \\ & & 0 & 0 \\ & b & 1 & 0 \\ & c & 0 & 2 \\ & d & 1 & 2 \end{array}$

d● 0.4

4. What is E[Y|X]?

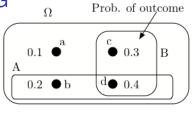
0.2 ● b





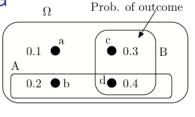
$$\begin{array}{cccc} \omega & X(\omega) & Y(\omega) \\ & & 0 & 0 \\ b & 1 & 0 \\ c & 0 & 2 \end{array}$$

$$E[Y|X=0] =$$

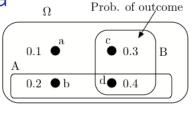


$$\omega \quad X(\omega) \quad Y(\omega)$$
a 0 0
b 1 0
c 0 2

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



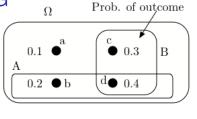
$$E[Y|X=0] = 0 \times Pr[Y=0|X=0] + 2 \times Pr[Y=2|X=0]$$



$$\omega \quad X(\omega) \quad Y(\omega)$$
a 0 0
b 1 0
c 0 2

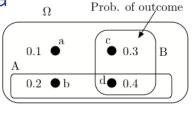
$$E[Y|X=0] = 0 \times Pr[Y=0|X=0] + 2 \times Pr[Y=2|X=0]$$

= $2 \times \frac{0.3}{0.4}$ =



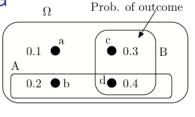
$$E[Y|X=0] = 0 \times Pr[Y=0|X=0] + 2 \times Pr[Y=2|X=0]$$

= $2 \times \frac{0.3}{0.4} = 1.5$



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

= $2 \times \frac{0.3}{0.4} = 1.5$
 $E[Y|X=1] =$

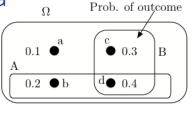


$$\omega \quad X(\omega) \quad Y(\omega)$$
a 0 0
b 1 0
c 0 2

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

$$= 2 \times \frac{0.3}{0.4} = 1.5$$

$$E[Y|X=1] = 0 \times Pr[Y=0|X=1] + 2 \times Pr[Y=2|X=1]$$

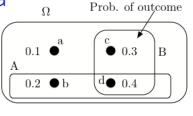


$$E[Y|X=0] = 0 \times Pr[Y=0|X=0] + 2 \times Pr[Y=2|X=0]$$

$$= 2 \times \frac{0.3}{0.4} = 1.5$$

$$E[Y|X=1] = 0 \times Pr[Y=0|X=1] + 2 \times Pr[Y=2|X=1]$$

$$=$$



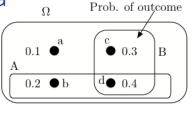
$$\omega \quad X(\omega) \quad Y(\omega)$$
a 0 0
b 1 0
c 0 2

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

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$$E[Y|X=1] = 0 \times Pr[Y=0|X=1] + 2 \times Pr[Y=2|X=1]$$

$$= 2 \times \frac{0.4}{0.6} =$$



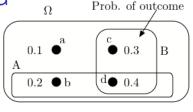
$$\omega \quad X(\omega) \quad Y(\omega)$$
a 0 0
b 1 0
c 0 2

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

$$= 2 \times \frac{0.3}{0.4} = 1.5$$

$$E[Y|X=1] = 0 \times Pr[Y=0|X=1] + 2 \times Pr[Y=2|X=1]$$

$$= 2 \times \frac{0.4}{0.6} = 1.33$$



$$\omega \quad X(\omega) \quad Y(\omega)$$
a 0 0
b 1 0
c 0 2

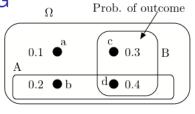
4. What is E[Y|X]?

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

$$= 2 \times \frac{0.3}{0.4} = 1.5$$

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$$= 2 \times \frac{0.4}{0.6} = 1.33$$



$$\omega \quad X(\omega) \quad Y(\omega)$$
a 0 0
b 1 0
c 0 2

4. What is E[Y|X]?

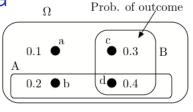
$$E[Y|X=0] = 0 \times Pr[Y=0|X=0] + 2 \times Pr[Y=2|X=0]$$

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$$E[Y|X=1] = 0 \times Pr[Y=0|X=1] + 2 \times Pr[Y=2|X=1]$$

$$= 2 \times \frac{0.4}{0.6} = 1.33$$

5. What is cov(X, Y)? cov(X, Y) =



$$\omega \quad X(\omega) \quad Y(\omega)$$
a 0 0
b 1 0
c 0 2

4. What is E[Y|X]?

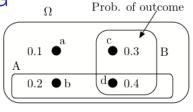
$$E[Y|X=0] = 0 \times Pr[Y=0|X=0] + 2 \times Pr[Y=2|X=0]$$

$$= 2 \times \frac{0.3}{0.4} = 1.5$$

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$$= 2 \times \frac{0.4}{0.6} = 1.33$$

$$cov(X, Y) = E[XY] - E[X]E[Y] =$$



$$\omega \quad X(\omega) \quad Y(\omega)$$
a 0 0
b 1 0
c 0 2

4. What is E[Y|X]?

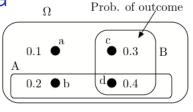
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$$cov(X, Y) = E[XY] - E[X]E[Y] = 0.8 - 0.6 \times 1.4 =$$



4. What is E[Y|X]?

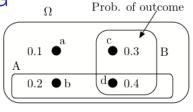
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$$= 2 \times \frac{0.4}{0.6} = 1.33$$

$$cov(X, Y) = E[XY] - E[X]E[Y] = 0.8 - 0.6 \times 1.4 = -0.04$$



$$\omega \quad X(\omega) \quad Y(\omega)$$
a 0 0
b 1 0
c 0 2

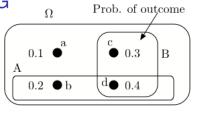
$$E[Y|X=0] = 0 \times Pr[Y=0|X=0] + 2 \times Pr[Y=2|X=0]$$

$$= 2 \times \frac{0.3}{0.4} = 1.5$$

$$E[Y|X=1] = 0 \times Pr[Y=0|X=1] + 2 \times Pr[Y=2|X=1]$$

$$= 2 \times \frac{0.4}{0.6} = 1.33$$

- 5. What is cov(X, Y)? $cov(X, Y) = E[XY] E[X]E[Y] = 0.8 0.6 \times 1.4 = -0.04$
- 6. What is L[Y|X]?



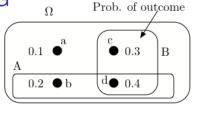
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$$\omega \quad X(\omega) \quad Y(\omega)$$
a 0 0
b 1 0
c 0 2

4. What is E[Y|X]?

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

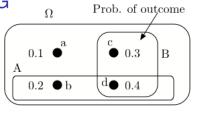
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$$E[Y|X=1] = 0 \times Pr[Y=0|X=1] + 2 \times Pr[Y=2|X=1]$$

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6. What is
$$L[Y|X]$$
?
 $L[Y|X] = E[Y] + \frac{cov(X,Y)}{var(X)}(X - E[X]) =$



4. What is E[Y|X]?

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

$$= 2 \times \frac{0.3}{0.4} = 1.5$$

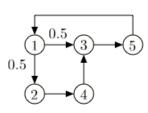
$$E[Y|X=1] = 0 \times Pr[Y=0|X=1] + 2 \times Pr[Y=2|X=1]$$

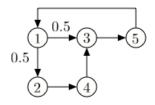
$$= 2 \times \frac{0.4}{0.6} = 1.33$$

5. What is cov(X, Y)?

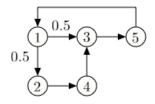
$$cov(X, Y) = E[XY] - E[X]E[Y] = 0.8 - 0.6 \times 1.4 = -0.04$$

$$L[Y|X] = E[Y] + \frac{cov(X,Y)}{var(X)}(X - E[X]) = 1.4 + \frac{-0.04}{0.6 \times 0.4}(X - 0.6)$$

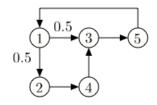




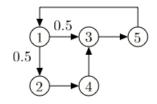
7. Is this Markov chains irreducible?



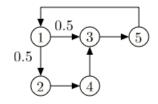
7. Is this Markov chains irreducible? Yes.



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

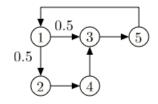


- 7. Is this Markov chains irreducible? Yes.
- Is this Markov chain periodic?No.

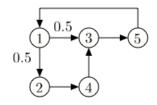


- 7. Is this Markov chains irreducible? Yes.
- 8. Is this Markov chain periodic?

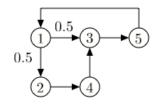
No. The return times to 3 are



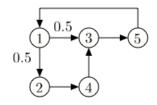
- 7. Is this Markov chains irreducible? Yes.
- 8. Is this Markov chain periodic?



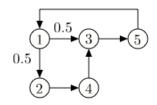
- 7. Is this Markov chains irreducible? Yes.
- 8. Is this Markov chain periodic?



- 7. Is this Markov chains irreducible? Yes.
- 8. Is this Markov chain periodic?
 - No. The return times to 3 are $\{3,5,..\}$: coprime!
- 9. Does π_n converge to a value independent of π_0 ?

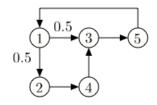


- 7. Is this Markov chains irreducible? Yes.
- 8. Is this Markov chain periodic?
 - No. The return times to 3 are $\{3,5,..\}$: coprime!
- 9. Does π_n converge to a value independent of π_0 ? Yes!



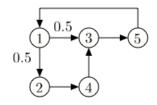
- 7. Is this Markov chains irreducible? Yes.
- 8. Is this Markov chain periodic?

- 9. Does π_n converge to a value independent of π_0 ? Yes!
- 10. Does $\frac{1}{n} \sum_{m=1}^{n-1} 1\{X_m = 3\}$ converge as $n \to \infty$?



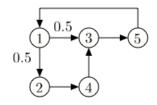
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- 7. Is this Markov chains irreducible? Yes.
- 8. Is this Markov chain periodic?

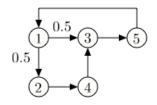
- 9. Does π_n converge to a value independent of π_0 ? Yes!
- 10. Does $\frac{1}{n}\sum_{m=1}^{n-1} 1\{X_m = 3\}$ converge as $n \to \infty$? Yes!
- 11. Calculate π .



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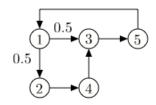
Let
$$a = \pi(1)$$
.



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

- 9. Does π_n converge to a value independent of π_0 ? Yes!
- 10. Does $\frac{1}{n}\sum_{m=1}^{n-1} 1\{X_m = 3\}$ converge as $n \to \infty$? Yes!
- 11. Calculate π .

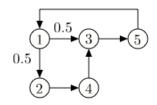
Let
$$a = \pi(1)$$
. Then $a = \pi(5)$,



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

- 9. Does π_n converge to a value independent of π_0 ? Yes!
- 10. Does $\frac{1}{n}\sum_{m=1}^{n-1} 1\{X_m = 3\}$ converge as $n \to \infty$? Yes!
- 11. Calculate π .

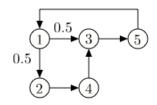
Let
$$a = \pi(1)$$
. Then $a = \pi(5), \pi(2) = 0.5a$,



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

- 9. Does π_n converge to a value independent of π_0 ? Yes!
- 10. Does $\frac{1}{n}\sum_{m=1}^{n-1} 1\{X_m = 3\}$ converge as $n \to \infty$? Yes!
- 11. Calculate π .

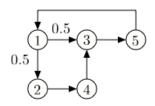
Let
$$a = \pi(1)$$
. Then $a = \pi(5), \pi(2) = 0.5a, \pi(4) = \pi(2) = 0.5a$,



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

- 9. Does π_n converge to a value independent of π_0 ? Yes!
- 10. Does $\frac{1}{n}\sum_{m=1}^{n-1} 1\{X_m = 3\}$ converge as $n \to \infty$? Yes!
- 11. Calculate π .

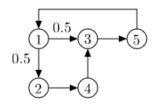
Let
$$a=\pi(1)$$
. Then $a=\pi(5), \pi(2)=0.5a, \pi(4)=\pi(2)=0.5a, \pi(3)=0.5\pi(1)+\pi(4)=a$.



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

- 9. Does π_n converge to a value independent of π_0 ? Yes!
- 10. Does $\frac{1}{n}\sum_{m=1}^{n-1} 1\{X_m = 3\}$ converge as $n \to \infty$? Yes!
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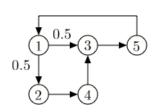
Let
$$a=\pi(1)$$
. Then $a=\pi(5), \pi(2)=0.5a, \pi(4)=\pi(2)=0.5a, \pi(3)=0.5\pi(1)+\pi(4)=a$. Thus, $\pi=[a,0.5a,a,0.5a,a]=[1,0.5,1,0.5,1]a$, so $a=$

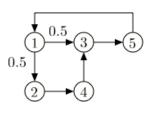


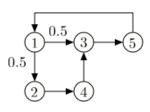
- 7. Is this Markov chains irreducible? Yes.
- 8. Is this Markov chain periodic?

- 9. Does π_n converge to a value independent of π_0 ? Yes!
- 10. Does $\frac{1}{n}\sum_{m=1}^{n-1} 1\{X_m = 3\}$ converge as $n \to \infty$? Yes!
- 11. Calculate π .

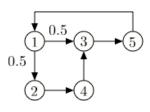
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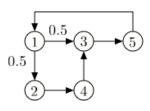


$$\beta(1) = 1 + 0.5\beta(2) + 0.5\beta(3)$$



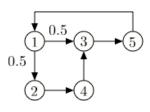
$$\beta(1) = 1 + 0.5\beta(2) + 0.5\beta(3)$$

 $\beta(2) = 1$



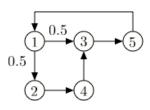
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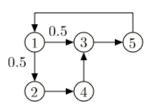


12. Write the first step equations for calculating the mean time from 1 to 4.

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

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13. Solve these equations.



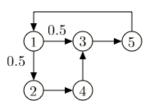
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$$\beta(1) = 1 + 0.5 \times 1 + 0.5 \times (1 + (1 + \beta(1)))$$



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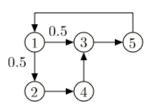
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= 2.5 + 0.5\beta(1).



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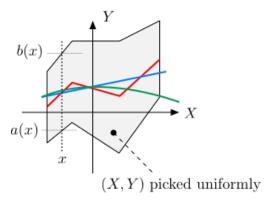
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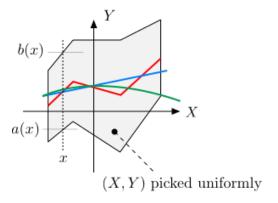
Hence, $\beta(1) = 5$.

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

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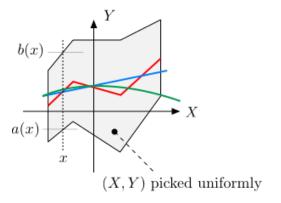


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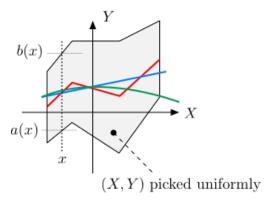
Answer: Red.

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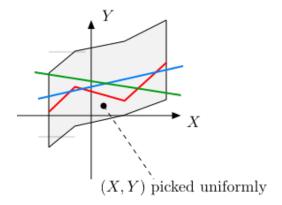
Answer: Red. Given X = x, Y = U[a(x), b(x)].

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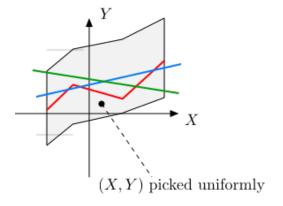


Answer: Red. Given X = x, Y = U[a(x), b(x)]. Thus, $E[Y|X = x] = \frac{a(x) + b(x)}{2}$.

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

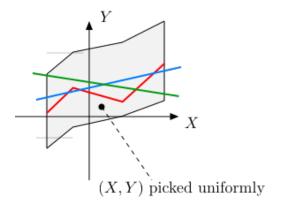


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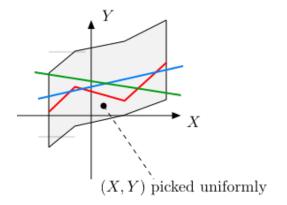
Answer: Blue.

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



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

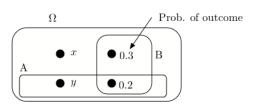
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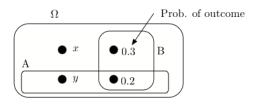


Answer: Blue.

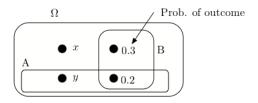
Cannot be red (not a straight line).

Cannot be green: *X* and *Y* are clearly positively correlated.



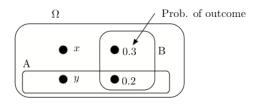


1. Find (x, y) so that A and B are independent.



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 We need

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



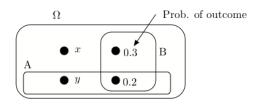
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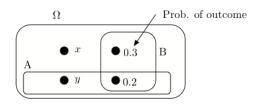
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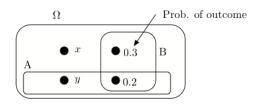
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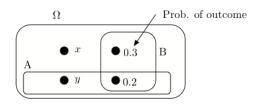
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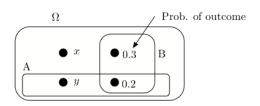
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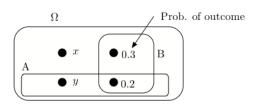
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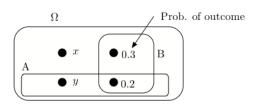
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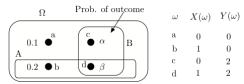
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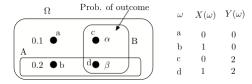
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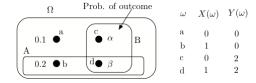
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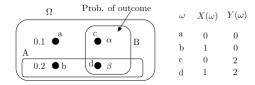


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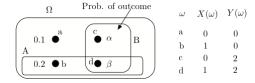
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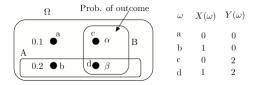
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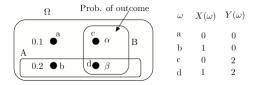
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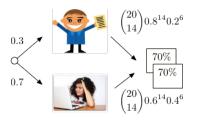
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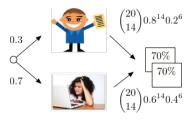
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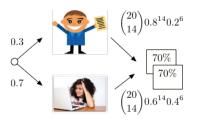
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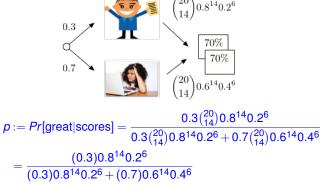
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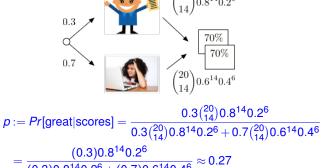


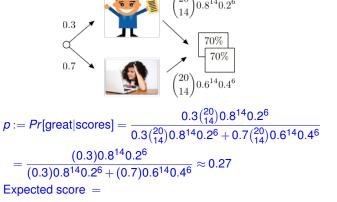
p := Pr[great|scores] =

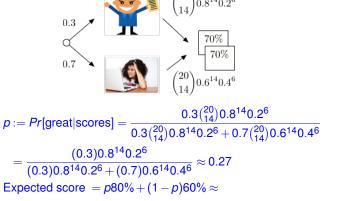


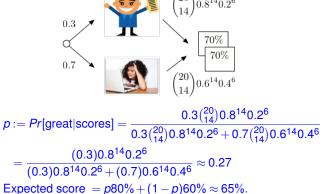
$$p := Pr[\text{great}|\text{scores}] = \frac{0.3\binom{20}{14}0.8^{14}0.2^{6}}{0.3\binom{20}{14}0.8^{14}0.2^{6} + 0.7\binom{20}{14}0.6^{14}0.4^{6}}$$











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 $\le \frac{var(X)}{15^2}.$

Now,

$$var(X) = 20var(X_1) = 20 \times 2.9 = 58.$$

You roll a balanced six-sided die 20 times. Use Chebyshev to upper-bound the probability that the total number of dots exceeds 85.

Let $X = X_1 + \dots + X_{20}$ be the total number of dots. Then

$$Pr[X > 85] = Pr[X - 70 > 15] \le Pr[|X - 70| > 15]$$

 $\le \frac{var(X)}{15^2}.$

Now,

$$var(X) = 20var(X_1) = 20 \times 2.9 = 58.$$

Hence,

$$Pr[X > 85] \le \frac{58}{15^2} \approx 0.26.$$

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Thus,

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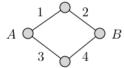
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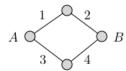
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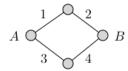
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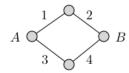


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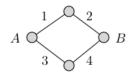
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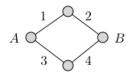
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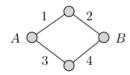
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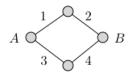
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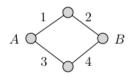
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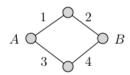


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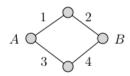


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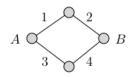
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$$= (2t - t^2)^2$$



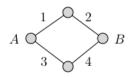
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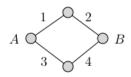
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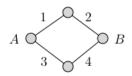
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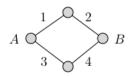
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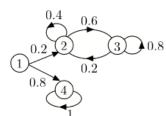
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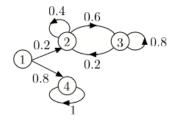
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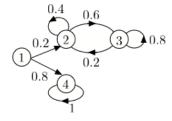
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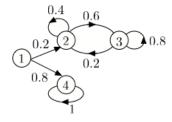




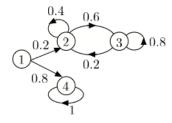
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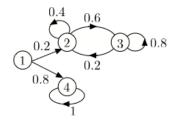
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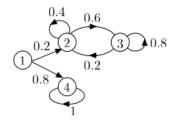
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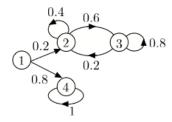
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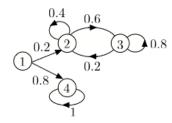
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Quiz 3: R

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Thus,

$$E[Y|X] = (2\alpha - 1)X = -\frac{1}{2n-1}X.$$

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