**Probability Review** 

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

- 1. True or False
- 2. Some Key Results
- 3. Quiz 1: G

- 1. True or False
- 2. Some Key Results
- 3. Quiz 1: G (≈40%)

- 1. True or False
- 2. Some Key Results
- 3. Quiz 1: G (≈ 40%)
- 4. Quiz 2: PG

- 1. True or False
- 2. Some Key Results
- 3. Quiz 1: G (≈ 40%)
- 4. Quiz 2: PG (≈ 40%)

- 1. True or False
- 2. Some Key Results
- 3. Quiz 1: G (≈ 40%)
- 4. Quiz 2: PG (≈ 40%)
- 5. Quiz 3: R

- 1. True or False
- 2. Some Key Results
- 3. Quiz 1: G (≈ 40%)
- 4. Quiz 2: PG (≈ 40%)
- 5. Quiz 3: R (≈ 20%)

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- 2. Some Key Results
- 3. Quiz 1: G (≈ 40%)
- 4. Quiz 2: PG (≈ 40%)
- 5. Quiz 3: R (≈ 20%)
- 6. Common Mistakes

- 1. True or False
- 2. Some Key Results
- 3. Quiz 1: G (≈ 40%)
- 4. Quiz 2: PG (≈ 40%)
- 5. Quiz 3: R (≈ 20%)
- 6. Common Mistakes

•  $\Omega$  and A are independent.

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- $Pr[A \setminus B] \ge Pr[A] Pr[B].$

- $\Omega$  and A are independent. True
- ▶  $Pr[A \cap B] = Pr[A] + Pr[B] Pr[A \cup B]$ . True
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- $Pr[A \cap B] = Pr[A] + Pr[B] Pr[A \cup B]$ . True
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$$X_1, \ldots, X_n \text{ i.i.d.} \implies var(\frac{X_1 + \cdots + X_n}{n}) = var(X_1).$$

- $\Omega$  and A are independent. True
- ▶  $Pr[A \cap B] = Pr[A] + Pr[B] Pr[A \cup B]$ . True
- $Pr[A \setminus B] \ge Pr[A] Pr[B]$ . True

► 
$$X_1, \ldots, X_n$$
 i.i.d.  $\implies var(\frac{X_1 + \cdots + X_n}{n}) = var(X_1)$ . False:  $\times \frac{1}{n}$ 

•  $\Omega$  and A are independent. True

▶ 
$$Pr[A \cap B] = Pr[A] + Pr[B] - Pr[A \cup B]$$
. True

► 
$$X_1, \ldots, X_n$$
 i.i.d.  $\implies var(\frac{X_1 + \cdots + X_n}{n}) = var(X_1)$ . False:  $\times \frac{1}{n}$ 

• 
$$Pr[|X-a| \ge b] \le \frac{E[(X-a)^2]}{b^2}$$
.

- $\Omega$  and A are independent. True
- ▶  $Pr[A \cap B] = Pr[A] + Pr[B] Pr[A \cup B]$ . True
- $Pr[A \setminus B] \ge Pr[A] Pr[B]$ . True

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

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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:  $\sqrt{n}$ 

Ω and A are independent. True

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$$Pr[A \cap B] = Pr[A] + Pr[B] - Pr[A \cup B]$$
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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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$$Pr[|X-a| \ge b] \le \frac{E[(X-a)^2]}{b^2}$$
. True

- ►  $X_1, ..., X_n$  i.i.d.  $\implies \frac{X_1 + \dots + X_n n \mathbb{E}[X_1]}{n \sigma(X_1)} \rightarrow \mathcal{N}(0, 1)$ . False:  $\sqrt{n}$
- $X = Expo(\lambda) \implies Pr[X > 5|X > 3] = Pr[X > 2].$

Ω and A are independent. True

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

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- $X = Expo(\lambda) \implies Pr[X > 5 | X > 3] = Pr[X > 2]$ . True:

► 
$$Pr[A \cap B] = Pr[A] + Pr[B] - Pr[A \cup B]$$
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$$Pr[A \setminus B] \ge Pr[A] - Pr[B]$$
. True

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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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. True

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

$$X = Expo(\lambda) \implies Pr[X > 5 | X > 3] = Pr[X > 2].$$
True:  
$$\frac{\exp\{-\lambda 5\}}{\exp\{-\lambda 3\}} = \exp\{-\lambda 2\}.$$

• 
$$[A_n - 2\sigma \frac{1}{n}, A_n + 2\sigma \frac{1}{n}] = 95\%$$
-Cl for  $\mu$ .

• 
$$[A_n - 2\sigma \frac{1}{n}, A_n + 2\sigma \frac{1}{n}] = 95\%$$
-Cl for  $\mu$ . No

• 
$$[A_n - 2\sigma \frac{1}{n}, A_n + 2\sigma \frac{1}{n}] = 95\%$$
-CI for  $\mu$ . No

$$\blacktriangleright [A_n - 2\sigma \frac{1}{\sqrt{n}}, A_n + 2\sigma \frac{1}{\sqrt{n}}] = 95\%$$
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$$[A_n - 2\sigma \frac{1}{n}, A_n + 2\sigma \frac{1}{n}] = 95\%$$
-Cl for  $\mu$ . No  
•  $[A_n - 2\sigma \frac{1}{\sqrt{n}}, A_n + 2\sigma \frac{1}{\sqrt{n}}] = 95\%$ -Cl for  $\mu$ . Yes  
• If  $0.3 < \sigma < 3$ , then  
 $[A_n - 0.6\frac{1}{\sqrt{n}}, A_n + 0.6\frac{1}{\sqrt{n}}] = 95\%$ -Cl for  $\mu$ .

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 $[A_n - 0.6 \frac{1}{\sqrt{n}}, A_n + 0.6 \frac{1}{\sqrt{n}}] = 95\%$ -Cl for  $\mu$ . No  
• If  $0.3 < \sigma < 3$ , then  
 $[A_n - 6 \frac{1}{\sqrt{n}}, A_n + 6 \frac{1}{\sqrt{n}}] = 95\%$ -Cl for  $\mu$ .
#### Correct or not?

When  $n \gg 1$ , one has

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• If  $0.3 < \sigma < 3$ , then  
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$$\begin{array}{ll} [1] \ Pr[X \ge a] \le \frac{E[f(X)]}{f(a)}, \\ [2] \ Pr[|X - E[X]| > a] \le \frac{var[X]}{a^2} \\ [3] \ Pr[X \ge a] \le \min_{\theta > 0} \frac{E[e^{\theta X}]}{e^{\theta a}} \\ [4] \ g(\cdot) \ \text{convex} \ \Rightarrow E[g(X)] \ge g(E[X]) \\ [6] \ \sum_{y} yPr[Y = y|X = x] \\ [7] \ Pr[|\frac{X_1 + \dots + X_n}{n} - E[X_1]| \ge \varepsilon] \rightarrow 0, \\ [8] \ E[(Y - E[Y|X])h(X)] = 0. \end{array}$$

- ► WLLN (7)
- MMSE

$$\begin{array}{ll} [1] \ Pr[X \ge a] \le \frac{E[f(X)]}{f(a)}, \\ [2] \ Pr[|X - E[X]| > a] \le \frac{var[X]}{a^2} \\ [3] \ Pr[X \ge a] \le \min_{\theta > 0} \frac{E[e^{\theta X}]}{e^{\theta a}} \\ [4] \ g(\cdot) \ \text{convex} \ \Rightarrow E[g(X)] \ge g(E[X]) \\ [6] \ \sum_{y} y Pr[Y = y|X = x] \\ [7] \ Pr[|\frac{X_1 + \dots + X_n}{n} - E[X_1]| \ge \varepsilon] \rightarrow 0, \\ [8] \ E[(Y - E[Y|X])h(X)] = 0. \end{array}$$

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- Projection property (8)

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

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

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

- WLLN (7)
- MMSE (6)
- Projection property (8)
- Chebyshev (2)
- LLSE (5)

- WLLN (7)
- MMSE (6)
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- LLSE (5)
- Markov's inequality

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- MMSE (6)
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- LLSE (5)
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1. What is *P*[*A*|*B*]?



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Pr[A|B] =



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Pr[B|A] =



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- 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?



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- 3. Are *A* and *B* positively correlated?

#### No.



1. What is *P*[*A*|*B*]?

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

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?

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

#### Quiz 1: G Prob. of outcome $\Omega$ $Y(\omega)$ $X(\omega)$ 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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 $E[Y|X=0] = 0 \times Pr[Y=0|X=0] + 2 \times Pr[Y=2|X=0]$ =  $2 \times \frac{0.3}{0.4} =$ 



 $E[Y|X=0] = 0 \times Pr[Y=0|X=0] + 2 \times Pr[Y=2|X=0]$ =  $2 \times \frac{0.3}{0.4} = 1.5$


 $E[Y|X = 0] = 0 \times Pr[Y = 0|X = 0] + 2 \times Pr[Y = 2|X = 0]$ =  $2 \times \frac{0.3}{0.4} = 1.5$ E[Y|X = 1] =





=

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)?



 $E[Y|X = 0] = 0 \times Pr[Y = 0|X = 0] + 2 \times Pr[Y = 2|X = 0]$ =  $2 \times \frac{0.3}{0.4} = 1.5$  $E[Y|X = 1] = 0 \times Pr[Y = 0|X = 1] + 2 \times Pr[Y = 2|X = 1]$ =  $2 \times \frac{0.4}{0.6} = 1.33$ 

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



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



 $E[Y|X = 0] = 0 \times Pr[Y = 0|X = 0] + 2 \times Pr[Y = 2|X = 0]$ = 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)?  $cov(X, Y) = E[XY] - E[X]E[Y] = 0.8 - 0.6 \times 1.4 =$ 



 $E[Y|X = 0] = 0 \times Pr[Y = 0|X = 0] + 2 \times Pr[Y = 2|X = 0]$ = 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)?  $cov(X, Y) = E[XY] - E[X]E[Y] = 0.8 - 0.6 \times 1.4 = -0.04$ 



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



- 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]? L[Y|X] =



- 5. What is cov(X, Y)?  $cov(X, Y) = E[XY] - E[X]E[Y] = 0.8 - 0.6 \times 1.4 = -0.04$
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- 6. What is L[Y|X]?  $L[Y|X] = E[Y] + \frac{cov(X,Y)}{var(X)}(X - E[X]) = 1.4 + \frac{-0.04}{0.6 \times 0.4}(X - 0.6)$





7. Is this Markov chains irreducible?



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No. The return times to 3 are



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No. The return times to 3 are  $\{3, 5, ..\}$ : coprime!

9. Does  $\pi_n$  converge to a value independent of  $\pi_0$ ?



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- 9. Does  $\pi_n$  converge to a value independent of  $\pi_0$ ? Yes!
- 10. Does  $\frac{1}{n}\sum_{m=1}^{n-1} 1\{X_m = 3\}$  converge as  $n \to \infty$ ?



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- 11. Calculate  $\pi$ .



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Let  $a = \pi(1)$ .



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



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No. The return times to 3 are  $\{3, 5, ..\}$ : coprime!

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- 10. Does  $\frac{1}{n}\sum_{m=1}^{n-1} 1\{X_m = 3\}$  converge as  $n \to \infty$ ? Yes!
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13. Solve these equations.

 $\beta(1) = 1 + 0.5 \times 1 + 0.5 \times (1 + (1 + \beta(1)))$ = 2.5 + 0.5 \beta(1).

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

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

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

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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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3. Find  $\alpha$  so that X and Y are independent.



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That is,

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p := Pr[great|scores] =













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Hence,

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

7. Let *X*, *Y*, *Z* be i.i.d. *Expo*(1).

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# Quiz 3: R

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