
Midterm 1

Last Name	First Name	SID
Left Neighbor First and Last Name		Right Neighbor First and Last Name

Rules.

- Unless otherwise stated, all your answers need to be justified and your work must be shown. Answers without sufficient justification will get no credit.
- All work you want graded should be on the fronts of the sheets in the space provided. Back sides may be used for scratch work, but will not be scanned/graded.
- You have 10 minutes to read the exam and 70 minutes to complete the exam. (DSP students with $X\%$ time accommodation should spend $10 \cdot X\%$ time on reading and $70 \cdot X\%$ time on completing the exam).
- This exam is closed-book. You may reference one double-sided handwritten sheet of paper. No calculator or phones allowed.
- Collaboration with others is strictly prohibited. If you are caught cheating, you may fail the course and face disciplinary consequences.

Problem	out of
SID	1
Problem 1	72
Problem 2	30
Problem 3	15
Problem 4	8
Total	126

1 Similarity of Vectors [72 points]

Let \mathcal{V} be an arbitrary set, and $N \in \mathbb{N}$ be a positive integer. Let $\mathbf{x} = (x_1, x_2, \dots, x_N) \in \mathcal{V}^N$ and $\mathbf{y} = (y_1, y_2, \dots, y_N) \in \mathcal{V}^N$ be two vectors where each element $x_i, y_i \in \mathcal{V}, \forall i \in [N] \triangleq \{1, 2, \dots, N\}$. We are further given a function $f : \mathcal{V} \times \mathcal{V} \rightarrow \mathbb{R}$, and thus define the similarity of two vectors \mathbf{x} and \mathbf{y} , denoted as $s(\mathbf{x}, \mathbf{y})$, as

$$s(\mathbf{x}, \mathbf{y}) \triangleq \sum_{i=1}^N f(x_i, y_i).$$

For each subpart, please carefully check the definition of \mathcal{V} and $f(x, y)$ before you start solving problems.

1.1 Warm up: Bernoulli distribution [16 points]

Let $\mathcal{V} = \{0, 1\}$. Let $x_i \stackrel{\text{i.i.d.}}{\sim} \text{Bernoulli}(p_1)$ and $y_i \stackrel{\text{i.i.d.}}{\sim} \text{Bernoulli}(p_2)$ where $p_1, p_2 \in [0, 1]$ and \mathbf{x} and \mathbf{y} are independent. Let $f(x, y) = xy$.

- If $N = 1$, what is the distribution of $s(\mathbf{x}, \mathbf{y})$, and what is $\mathbb{E}[s(\mathbf{x}, \mathbf{y})]$ and $\text{Var}(s(\mathbf{x}, \mathbf{y}))$? [4 points]
- For a general fixed N and $p_2 = 1$, what is the distribution of $s(\mathbf{x}, \mathbf{y})$, and what is $\mathbb{E}[s(\mathbf{x}, \mathbf{y})]$ and $\text{Var}(s(\mathbf{x}, \mathbf{y}))$? [4 points]
- For a general fixed N and $p_1, p_2 \in (0, 1)$, what is the conditional probability $\mathbb{P}(x_1 = 1 | s(\mathbf{x}, \mathbf{y}) = 1)$? [8 points]

(a) When $N = 1$, $s(\mathbf{x}, \mathbf{y}) = x_1 y_1$ is the product of two Bernoulli random variables. By direct calculation, $s(\mathbf{x}, \mathbf{y}) \sim \text{Bernoulli}(p_1 p_2)$. [2 points] Therefore, $\mathbb{E}[s(\mathbf{x}, \mathbf{y})] = p_1 p_2$ [1 point] and $\text{Var}(s(\mathbf{x}, \mathbf{y})) = p_1 p_2 (1 - p_1 p_2)$. [1 point]

(b) When $p_2 = 1$, we have $y_i = 1$ almost surely. Therefore, $s(\mathbf{x}, \mathbf{y}) = \sum_{i=1}^N x_i$, which is the summation of N i.i.d. Bernoulli random variables. Therefore, $s(\mathbf{x}, \mathbf{y}) \sim \text{Binomial}(N, p_1)$. [2 points] Therefore, $\mathbb{E}[s(\mathbf{x}, \mathbf{y})] = N p_1$ [1 point] and $\text{Var}(s(\mathbf{x}, \mathbf{y})) = N p_1 (1 - p_1)$. [1 point]

(c) Applying Bayes' rule, we have

$$\mathbb{P}(x_1 = 1 | s(\mathbf{x}, \mathbf{y}) = 1) = \frac{\mathbb{P}(x_1 = 1) \mathbb{P}(s(\mathbf{x}, \mathbf{y}) = 1 | x_1 = 1)}{\mathbb{P}(s(\mathbf{x}, \mathbf{y}) = 1)} \text{ [2 points].}$$

Since x_1 is Bernoulli and $s(\mathbf{x}, \mathbf{y})$ is Binomial by the extension of (b), we can directly obtain that $\mathbb{P}(x_1 = 1) = p_1$ [1 point] and $\mathbb{P}(s(\mathbf{x}, \mathbf{y}) = 1) = N p_1 p_2 (1 - p_1 p_2)^{N-1}$. [1 point] To calculate $\mathbb{P}(s(\mathbf{x}, \mathbf{y}) = 1 | x_1 = 1)$, note that if $s(\mathbf{x}, \mathbf{y}) = 1$, either $x_1 y_1 = 1$ and $\sum_{i=2}^N x_i y_i = 0$,

or $x_1 y_1 = 0$ and $\sum_{i=2}^N x_i y_i = 1$. Let $u = \sum_{i=2}^N x_i y_i \sim \text{Binomial}(N-1, p_1 p_2)$. Therefore,

$$\begin{aligned}
 \mathbb{P}(s(\mathbf{x}, \mathbf{y}) = 1 | x_1 = 1) &= \mathbb{P}(x_1 y_1 = 1, u = 0 | x_1 = 1) + \mathbb{P}(x_1 y_1 = 0, u = 1 | x_1 = 1) \quad [1 \text{ point}] \\
 &= \mathbb{P}(y_1 = 1, u = 0 | x_1 = 1) + \mathbb{P}(y_1 = 0, u = 1 | x_1 = 1) \\
 &= \mathbb{P}(y_1 = 1, u = 0) + \mathbb{P}(y_1 = 0, u = 1) \\
 &= \mathbb{P}(y_1 = 1) \mathbb{P}(u = 0) + \mathbb{P}(y_1 = 0) \mathbb{P}(u = 1) \quad [1 \text{ point}] \\
 &= p_2 (1 - p_1 p_2)^{N-1} + (1 - p_2) (N-1) p_1 p_2 (1 - p_1 p_2)^{N-2}. \quad [1 \text{ point}]
 \end{aligned}$$

Plugging all three probabilities in, we can finally get

$$\mathbb{P}(x_1 = 1 | s(\mathbf{x}, \mathbf{y}) = 1) = \frac{p_1 p_2 (1 - p_1 p_2)^{N-1} + (1 - p_2) (N-1) p_1^2 p_2 (1 - p_1 p_2)^{N-2}}{N p_1 p_2 (1 - p_1 p_2)^{N-1}}. \quad [1 \text{ point}]$$

1.2 Text similarity: Uniform distribution [20 points]

Let $\mathcal{V} = \{'a', 'b', \dots, 'z', 'A', 'B', \dots, 'Z'\}$ with $|\mathcal{V}| = 52$ which contains all 26 upper case letters and all 26 lower case letters. Let $f(x, y) = \mathbb{1}\{x = y\}, \forall x, y \in \mathcal{V}$, i.e., $f(x, y) = 1$ if and only if x and y are the same letter (case-sensitive). For example, $f('a', 'C') = 0$, $f('B', 'B') = 1$, $f('Z', 'z') = 0$. For a vector $\mathbf{x} \in \mathcal{V}^N$, it is essentially a text, and we will equivalently represent it using a string. For example, if $N = 5$, and $\mathbf{x} = (x_1, x_2, \dots, x_5) = ('A', 'p', 'p', 'l', 'e')$, we will also write $\mathbf{x} = \text{'Apple'}$ for convenience.

- (a) If $N = 7$, and \mathbf{x} is the first name of Prof. Jiantao Jiao, and \mathbf{y} is the first name of head TA Tianhao Wu, what is x_2 and y_4 ? [4 points]
- (b) If $N = 7$, $\mathbf{x} = \text{'Tianhao'}$ and $\mathbf{y} = \text{'Jiantao'}$, what is the similarity between \mathbf{x} and \mathbf{y} , i.e., $s(\mathbf{x}, \mathbf{y})$? If $N = 11$, $\mathbf{x} = \text{'TATianhaoWu'}$ and $\mathbf{y} = \text{'probability'}$, what is $s(\mathbf{x}, \mathbf{y})$? [4 points]
- (c) Assume each x_i and y_i are i.i.d. uniformly sampled from \mathcal{V} . For a general fixed N , please calculate $\mathbb{E}[s(\mathbf{x}, \mathbf{y})]$ and $\text{Var}(s(\mathbf{x}, \mathbf{y}))$. [4 points]
- (d) Assume each x_i and y_i are still i.i.d. uniformly sampled from \mathcal{V} . Additionally, N follows a uniform distribution over $\{1, 2, 3, 4, 5\}$ and N , \mathbf{x} , \mathbf{y} are mutually independent. Please calculate $\mathbb{E}[s(\mathbf{x}, \mathbf{y})]$ and $\text{Var}(s(\mathbf{x}, \mathbf{y}))$. You may denote $q = 1/52$ and leave q in your final answer. [8 points]

(a) Since $\mathbf{x} = \text{'Jiantao'}$ and $\mathbf{y} = \text{'Tianhao'}$, we can obtain that $x_2 = \text{'i'}$ [2 points] and $y_4 = \text{'n'}$ [2 points].

(b) When $\mathbf{x} = \text{'Tianhao'}$ and $\mathbf{y} = \text{'Jiantao'}$, by comparing letters one by one, we have $s(\mathbf{x}, \mathbf{y}) = 5$. [2 points] When $\mathbf{x} = \text{'TATianhaoWu'}$ and $\mathbf{y} = \text{'probability'}$, by careful comparison, we have $s(\mathbf{x}, \mathbf{y}) = 1$. [2 points]

(c) First, observe that $\mathbb{P}(f(x_i, y_i) = 1) = \mathbb{P}(x_i = y_i) = 1/52$. We denote $q = 1/52$. Then $s(\mathbf{x}, \mathbf{y}) \sim \text{Binomial}(N, q)$, from which we can conclude $\mathbb{E}[s(\mathbf{x}, \mathbf{y})] = Nq = \frac{N}{52}$ [2 points] and $\text{Var}(s(\mathbf{x}, \mathbf{y})) = Nq(1 - q) = \frac{51N}{52^2}$. [2 points]

(d) By the law of iterated expectation,

$$\begin{aligned} \mathbb{E}[s(\mathbf{x}, \mathbf{y})] &= \mathbb{E}[\mathbb{E}[s(\mathbf{x}, \mathbf{y})|N]] \text{ [1 point]} \\ &= \mathbb{E}[Nq] \text{ [1 point]} \\ &= q\mathbb{E}[N] = 3q. \text{ [1 point]} \end{aligned}$$

To calculate $\text{Var}(s(\mathbf{x}, \mathbf{y}))$, we need to calculate $\mathbb{E}[\text{Var}(s(\mathbf{x}, \mathbf{y})|N)]$ and $\text{Var}(\mathbb{E}[s(\mathbf{x}, \mathbf{y})|N])$. For the first term,

$$\mathbb{E}[\text{Var}(s(\mathbf{x}, \mathbf{y})|N)] = \mathbb{E}[Nq(1 - q)] = q(1 - q)\mathbb{E}[N] = 3q(1 - q). \text{ [2 points]}$$

For the second term,

$$\begin{aligned}\text{Var}(\mathbb{E}[s(\mathbf{x}, \mathbf{y})|N]) &= \mathbb{E}[\mathbb{E}^2[s(\mathbf{x}, \mathbf{y})|N]] - \mathbb{E}^2[\mathbb{E}[s(\mathbf{x}, \mathbf{y})|N]] \\ &= \mathbb{E}[N^2q^2] - \mathbb{E}^2[Nq] \text{ [1 point]} \\ &= 11q^2 - 9q^2 = 2q^2. \text{ [1 point]}\end{aligned}$$

Therefore, by the law of total variance, we can obtain that

$$\text{Var}(s(\mathbf{x}, \mathbf{y})) = \mathbb{E}[\text{Var}(s(\mathbf{x}, \mathbf{y})|N)] + \text{Var}(\mathbb{E}[s(\mathbf{x}, \mathbf{y})|N]) = 3q(1 - q) + 2q^2 = q(3 - q). \text{ [1 point]}$$

1.3 Exponential fun: Exponential distribution [20 points]

Let $\mathcal{V} = \mathbb{R}$ be the set of all real numbers. Let $f(x, y) = xy$ be the product of x and y . In this case, $s(\mathbf{x}, \mathbf{y})$ coincides with the inner product of two real-valued vectors. Also, we assume \mathbf{x} is independent of N and \mathbf{y} . Moreover, $x_i \stackrel{\text{i.i.d.}}{\sim} \exp(\lambda)$ for all $i \in [N]$ for some fixed $\lambda > 0$. Therefore, the p.d.f. of x_i is $f_{x_i}(t) = \frac{1}{\lambda}e^{-\lambda t}$ for $t \geq 0$.

- (a) What is the MGF (moment generating function) of x_i ? You only need to provide the answer without justification. Be careful about the domain of the MGF. [4 points]
- (b) Let $y_i = 1$ for all $i = 1, 2, \dots$. For a fixed N , what is the MGF of $s(\mathbf{x}, \mathbf{y})$? [4 points]
- (c) Let $y_i = 1$ for all $i = 1, 2, \dots$. Let $N \sim \text{Geo}(p)$ which is a geometric distribution with head probability $p \in (0, 1)$. What is the MGF of $s(\mathbf{x}, \mathbf{y})$? From the MGF of $s(\mathbf{x}, \mathbf{y})$, what is the distribution of $s(\mathbf{x}, \mathbf{y})$? *Hint: Consider applying the law of iterated expectation. Again be careful about the domain of the MGF.* [8 points]
- (d) Let $y_i \stackrel{\text{i.i.d.}}{\sim} \text{Bernoulli}(p)$ for some fixed $p \in (0, 1)$ for $i = 1, 2, \dots$. Let $N = \min\{n \geq 1 \mid y_n = 0\}$. What is the distribution of N and $s(\mathbf{x}, \mathbf{y})$? You only need to provide the answer without justification. [4 points]

(a) $M_{x_i}(t) = \frac{\lambda}{\lambda - t}$ [2 points] for $t < \lambda$ [1 point].

(b) Since $s(\mathbf{x}, \mathbf{y}) = \sum_{i=1}^N x_i$ where x_i are i.i.d. exponential random variable, [1 point] we have

$$M_{s(\mathbf{x}, \mathbf{y})}(t) = \prod_{i=1}^N M_{x_i}(t) = \left(\frac{\lambda}{\lambda - t} \right)^N. \text{ [2 points]}$$

(c) By law of iterated expectation, we have

$$M_{s(\mathbf{x}, \mathbf{y})}(t) = \mathbb{E}[e^{s(\mathbf{x}, \mathbf{y})t}] = \mathbb{E}[\mathbb{E}[e^{s(\mathbf{x}, \mathbf{y})t} | N]] = \mathbb{E} \left[\left(\frac{\lambda}{\lambda - t} \right)^N \right]. \text{ [2 points]}$$

Since $N \sim \text{Geo}(p)$, we have

$$\begin{aligned} \mathbb{E} \left[\left(\frac{\lambda}{\lambda - t} \right)^N \right] &= \sum_{i=1}^{\infty} \mathbb{P}(N = i) \left(\frac{\lambda}{\lambda - t} \right)^i \\ &= \sum_{i=1}^{\infty} (1-p)^{i-1} p \cdot \left(\frac{\lambda}{\lambda - t} \right)^i \\ &= \frac{p\lambda}{\lambda - t} \sum_{i=0}^{\infty} \left(\frac{(1-p)\lambda}{\lambda - t} \right)^i. \text{ [2 points]} \end{aligned}$$

When $\frac{(1-p)\lambda}{\lambda-t} \in (0, 1)$, which is $t < p\lambda$ [1 point], we have

$$\sum_{i=0}^{\infty} \left(\frac{(1-p)\lambda}{\lambda-t} \right)^i = \frac{1}{1 - \frac{(1-p)\lambda}{\lambda-t}} = \frac{\lambda-t}{p\lambda-t}. \quad [1 \text{ point}]$$

Therefore, we have $M_{s(\mathbf{x}, \mathbf{y})}(t) = \frac{p\lambda}{p\lambda-t}$ where $t < p\lambda$. [1 point]

From the MGF of $s(\mathbf{x}, \mathbf{y})$, we have $s(\mathbf{x}, \mathbf{y}) \sim \exp(p\lambda)$. [1 point]

- (d) $N \sim \text{Geo}(1-p)$. [1 point]. $s(\mathbf{x}, \mathbf{y})$ follows a mixed distribution, where with probability $1-p$ it is 0, and with probability p it follows $\exp((1-p)\lambda)$. [2 points]

1.4 Concentration: Gaussian distribution [16 points]

Let $\mathcal{V} = \mathbb{R}$ be the set of all real numbers, and let $f(x, y) = xy$ be the product of x and y . Also, we assume \mathbf{x} , \mathbf{y} and N are jointly independent. Moreover, $x_i \stackrel{\text{i.i.d.}}{\sim} \mathcal{N}(0, \sigma^2)$ for all $i \in [N]$ for some fixed $\sigma > 0$.

- (a) Let $N = 126$ and $y_i = 2^{\frac{i-1}{2}}$ for all $i \in [N]$. What is the distribution of $s(\mathbf{x}, \mathbf{y})$? You only need to provide your answer without justification. [4 points]
- (b) Let $N = 1$ and $y_1 = 1$. Please use Chebyshev's inequality to prove that $\mathbb{P}(s(\mathbf{x}, \mathbf{y}) \geq k\sigma) \leq \frac{1}{2k^2}$ for $k \geq 1$. [4 points]
- (c) Let $N = 1$ and $y_1 = 1$. Now, we want to get a tighter bound for large k . We could apply Markov's inequality to $e^{t \cdot s(\mathbf{x}, \mathbf{y})}$ (which is known as Chernoff bound), to obtain that

$$\mathbb{P}(s(\mathbf{x}, \mathbf{y}) \geq k\sigma) = \mathbb{P}(e^{t \cdot s(\mathbf{x}, \mathbf{y})} \geq e^{t \cdot k\sigma}) \leq \frac{\mathbb{E}[e^{t \cdot s(\mathbf{x}, \mathbf{y})}]}{e^{t \cdot k\sigma}}, \forall t > 0.$$

By choosing an appropriate t , one can get a tighter bound than we have in (b) for large k . Now please prove

$$\mathbb{P}(s(\mathbf{x}, \mathbf{y}) \geq k\sigma) \leq e^{-\frac{k^2}{2}}.$$

by choosing an appropriate t . You need to solve t explicitly. *Hint: You may use without proof that the MGF of $Z \sim \mathcal{N}(\mu, \sigma^2)$ is $M_Z(t) = \exp(\mu t + \frac{1}{2}\sigma^2 t^2)$.* [8 points]

(a) $s(\mathbf{x}, \mathbf{y}) \sim \mathcal{N}(0, (2^{126} - 1)\sigma^2)$. [3 points]

(b) Since $s(\mathbf{x}, \mathbf{y}) = x_1 \sim \mathcal{N}(0, \sigma^2)$, we have $\mathbb{E}[s(\mathbf{x}, \mathbf{y})] = 0$ and $\text{Var}(s(\mathbf{x}, \mathbf{y})) = \sigma^2$. [1 point] By Chebyshev's inequality, we have

$$\mathbb{P}(|s(\mathbf{x}, \mathbf{y})| \geq k\sigma) \leq \frac{1}{k^2}. [1 point]$$

By symmetry of the Gaussian distribution, we have $\mathbb{P}(s(\mathbf{x}, \mathbf{y}) \geq k\sigma) \leq \frac{1}{2k^2}$. [1 point]

(c) Since $\mathbb{E}[e^{t \cdot s(\mathbf{x}, \mathbf{y})}] = \exp(\frac{1}{2}\sigma^2 t^2)$, [1 point] we have

$$\mathbb{P}(s(\mathbf{x}, \mathbf{y}) \geq k\sigma) \leq \exp(\frac{1}{2}\sigma^2 t^2 - k\sigma t). [1 point]$$

By minimizing $\frac{1}{2}\sigma^2 t^2 - k\sigma t$, we have can solve $t = \frac{k}{\sigma}$. [2 points]

2 Card Game [30 points]

x club ♣ cards and y heart ♥ cards are stacked face down in a uniformly random order. Two players Alice and Bob, knowing x and y but not knowing the order, take turns to choose to either deal the top card to their opponent or themselves, after which the distributed card is immediately revealed. Dealing a ♥ to oneself gives one another turn (i.e. skips one's opponent's next turn). Whenever one player is distributed a ♣, the game ends and the other player wins. Alice deals first.

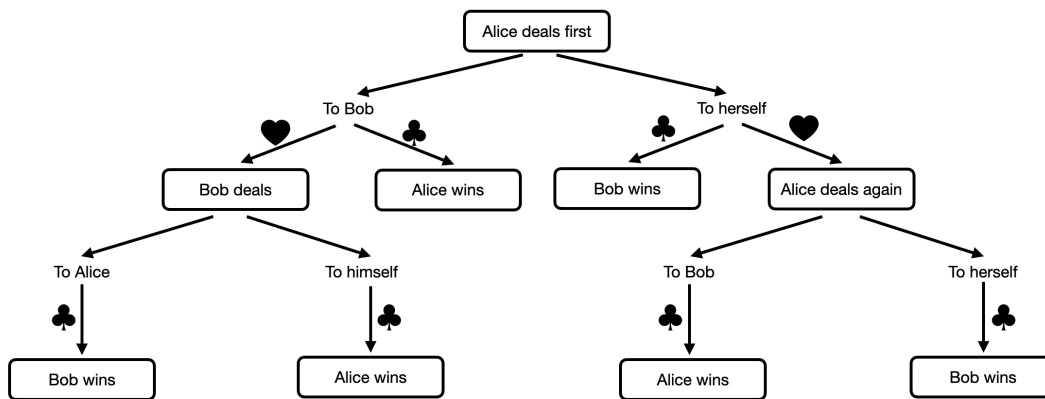


Figure 1: Example flow of the game when $x = y = 1$

- (a) (15 points) Consider the setting where $x = 2, y = 3$. If both player always deal the card to their opponent, find:

$$\mathbb{P}(\text{the third card is } \spadesuit \mid \text{Alice wins}).$$

- (b) (15 points) Consider the setting where $x = 1, y = 2$. A strategy is defined as a mapping from the numbers of remaining ♣ and ♥ cards in the pile to the choice of dealing (possibly random). Describe Alice's optimal strategy and find her probability to win under this strategy.

(a) Define:

$$S = \{\text{the third card is } \spadesuit\}$$

$$W = \{\text{Alice wins}\}.$$

There are only 4 orders out of $\binom{5}{2} = 10$ total orderings in which Alice losses:

(a) ♥, ♣, ♣, ♥, ♥;

(b) ♥, ♣, ♥, ♣, ♥;

(c) ♥, ♣, ♥, ♥, ♣;

(d) ♥, ♥, ♥, ♣, ♣;

It follows that $\mathbb{P}(W) = 1 - \frac{4}{10} = 6/10$ (3 points). Notice that $\mathbb{P}(S) = 2/5$ (3 points), $\mathbb{P}(W|S) = 3/4$ (3 points) since Alice losses only when the second card is ♣ given S . By

Bayes rule,

$$\begin{aligned}\mathbb{P}(S|W) &= \frac{\mathbb{P}(W|S)\mathbb{P}(S)}{\mathbb{P}(W)} \text{ (4 points)} \\ &= \frac{\frac{3}{4} \cdot \frac{2}{5}}{\frac{6}{10}} \\ &= 1/2 \text{ (2 points)}.\end{aligned}$$

(b) Alice's best strategy is to always deal to Bob (5 points). Let

$$\begin{aligned}W &= \{\text{Alice wins}\} \\ T &= \{\text{the first card is } \clubsuit\}.\end{aligned}$$

From Figure 1, it is obvious that the winning probability is always $1/2$ for both player when $x = y = 1$, whatever strategy they use (3 points). By law of total probability,

$$\begin{aligned}\mathbb{P}(\text{Alice wins}) &= \mathbb{P}(T) \cdot \mathbb{P}(W|T) + \mathbb{P}(T^c) \cdot \mathbb{P}(W|T^c) \text{ (3 points)} \\ &= \frac{1}{3} \cdot 1 + \frac{2}{3} \cdot \frac{1}{2} \text{ (3 points)} \\ &= 2/3 \text{ (1 points)}.\end{aligned}$$

3 Independence [15 points]

Random variable (ξ, η) has joint density

$$p(x, y) = \begin{cases} \frac{1+xy}{4}, & |x| < 1, |y| < 1 \\ 0, & \text{otherwise} \end{cases}.$$

Prove that ξ and η are not independent but ξ^2 and η^2 are independent.

We have

$$f_{\xi}(x) = \begin{cases} \frac{1}{2}, & |x| < 1 \\ 0, & \text{otherwise} \end{cases}$$

$$g_{\eta}(y) = \begin{cases} \frac{1}{2}, & |y| < 1 \\ 0, & \text{otherwise} \end{cases}.$$

(3 points)

Since $p(x, y) \neq f_{\xi}(x)g_{\eta}(y)$ (3 points), ξ and η are not independent.

Notice for any $u \in [0, 1]$,

$$\mathbb{P}(\xi^2 < u) = \int_{x^2 < u} f_{\xi}(x) dx = \sqrt{u}$$

$$\mathbb{P}(\eta^2 < v) = \sqrt{v}.$$

(3 points)

Furthermore,

$$\mathbb{P}(\xi^2 < u, \eta^2 < v) = \int \int_{x^2 < u, y^2 < v} p(x, y) dx dy = \sqrt{uv} \text{ (3 points)}.$$

Since $\mathbb{P}(\xi^2 < u, \eta^2 < v) = \mathbb{P}(\xi^2 < u)\mathbb{P}(\eta^2 < v)$ (3 points), it follows that ξ^2 and η^2 are independent.

4 Concentration Inequalities [8 points]

Prove that for any random variable X ,

$$\mathbb{P}(X = 0) \leq \frac{\text{Var}(X)}{\mathbb{E}[X]^2}.$$

By Chebyshev's inequality,

$$\begin{aligned} \mathbb{P}(X = 0) &\leq \mathbb{P}(|X - \mathbb{E}[X]| \geq |\mathbb{E}[X]|) \text{ (4 points)} \\ &\leq \frac{\text{Var}(X)}{\mathbb{E}[X]^2} \cdot \text{(4 points)} \end{aligned}$$