

TA Lecture 13 - Markov Chain & Final Review

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HW Problems

Final Review

Previous Final Exams

Problem 1

Given a coin with the probability p of landing heads. p is unknown and we need to estimate its value through data. In our data collection model, we have n independent tosses, the result of each toss is either Head or Tail. Let X denote the number of heads in the total n tosses. Now we conduct experiments to collect data and find $X = k$. Then we need to find \hat{p} , the estimation of p .

Problem 1 Continued

- (a) Assume p is an unknown constant. Find \hat{p} through the MLE (Maximum Likelihood Estimation) rule.
- (b) Assume p is a random variable with a prior distribution $p \sim \text{Beta}(a, b)$, where a and b are known constants. Find \hat{p} through the MAP (Maximum a Posterior Probability) rule.
- (c) Assume p is a random variable with a prior distribution $p \sim \text{Beta}(a, b)$, where a and b are known constants. Find \hat{p} through the MMSE (Minimal Mean Squared Error) rule.

Problem 1 Solution

Problem 2

Let X be the height of a randomly chosen adult man, and Y be his father's height, where X and Y have been standardized to have mean 0 and standard deviation 1. Suppose that (X, Y) is Bivariate Normal, with $X, Y \sim \mathcal{N}(0, 1)$ and $\text{Corr}(X, Y) = \rho$.

Problem 2 Continued

- (a) Let $y = ax + b$ be the equation of the best line for predicting Y from X (in the sense of minimizing the mean squared error), e.g., if we were to observe $X = 1.3$ then we would predict that Y is $1.3a + b$. Now suppose that we want to use Y to predict X , rather than using X to predict Y . Give and explain an intuitive guess for what the slope is of the best line for predicting X from Y .
- (b) Find a constant c (in terms of ρ) and an r.v. V such that $Y = cX + V$, with V independent of X . Hint: Start by finding c such that $\text{Cov}(X, Y - cX) = 0$.
- (c) Find a constant d (in terms of ρ) and an r.v. W such that $X = dY + W$, with W independent of Y .
- (d) Find $E(Y | X)$ and $E(X | Y)$.
- (e) Reconcile (a) and (d), giving a clear and correct intuitive explanation.

Problem 2 Solution

Problem 3

Two chess players, Vishy and Magnus, play a series of games. Given p , the game results are i.i.d. with probability p of Vishy winning, and probability $q = 1 - p$ of Magnus winning (assume that each game ends in a win for one of the two players). But p is unknown, so we will treat it as an r.v. To reflect our uncertainty about p , we use the prior $p \sim \text{Beta}(a, b)$, where a and b are known positive integers and $a \geq 2$.

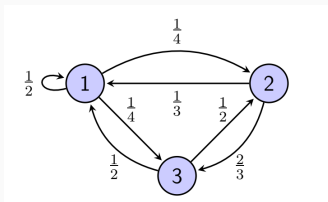
Problem 3 Continued

- (a) Find the expected number of games needed in order for Vishy to win a game (including the win). Simplify fully; your final answer should not use factorials or Γ .
- (b) Explain in terms of independence vs. conditional independence the direction of the inequality between the answer to (a) and $1 + E(G)$ for $G \sim \text{Geom}\left(\frac{a}{a+b}\right)$.
- (c) Find the conditional distribution of p given that Vishy wins exactly 7 out of the first 10 games.

Problem 3 Solution

Problem 4

Given a Markov chain with state-transition diagram shown as follows:



- (a) Is this chain irreducible?
- (b) Is this chain aperiodic?
- (c) Find the stationary distribution of this chain.
- (d) Is this chain reversible?

Definition

A Markov chain with transition matrix Q is *irreducible* if for any two states i and j , it is possible to go from i to j in a finite number of steps (with positive probability). That is, for any states i, j there is some positive integer n such that the (i, j) entry of Q^n is positive. A Markov chain that is not irreducible is called *reducible*.

Problem 4 Solution

Definition

For a Markov chain with transition matrix Q , the *period* of state i , denoted $d(i)$, is the greatest common divisor of the set of possible return times to i . That is,

$$d(i) = \gcd\{n > 0 : Q_{i,i}^n > 0\}.$$

If $d(i) = 1$, state i is said to be *aperiodic*. If the set of return times is empty, set $d(i) = +\infty$.

Definition

A Markov chain is called *periodic* if it is irreducible and all states have period greater than 1.

A Markov chain is called *aperiodic* if it is irreducible and all states have period equal to 1.

Problem 4 Solution

Definition

A row vector $\mathbf{s} = (s_1, \dots, s_M)$ such that $s_i \geq 0$ and $\sum_i s_i = 1$ is a *stationary distribution* for a Markov chain with transition matrix Q if

$$\sum_i s_i q_{i,j} = s_j.$$

for all j , or equivalently,

$$\mathbf{s}Q = \mathbf{s}.$$

Theorem

Given a Markov chain with finite state space.

- *If such Markov chain is irreducible, then it has a unique stationary distribution. In this distribution, every state has positive probability.*
- *If such Markov chain is irreducible and aperiodic, with stationary distribution \mathbf{s} and transition matrix Q , then $P(X_n = i)$ converges to s_i as $n \rightarrow \infty$. In terms of the transition matrix, Q^n converges to a matrix in which each row is \mathbf{s} .*

Problem 4 Solution

Definition

Let $Q = (q_{i,j})$ be the transition matrix of a Markov chain. Suppose there is $\mathbf{s} = (s_1, \dots, s_M)$ with $s_i \geq 0$, $\sum_i s_i = 1$, such that

$$s_i q_{i,j} = s_j q_{j,i}$$

for all states i and j . This equation is called the *reversibility* or *detailed balance* condition, and we say that the chain is *reversible* with respect to \mathbf{s} if it holds.

Theorem

Suppose that $Q = (q_{i,j})$ is a transition matrix of a Markov chain that is reversible with respect to a nonnegative vector $\mathbf{s} = (s_1, \dots, s_M)$ whose components sum to 1. Then \mathbf{s} is a stationary distribution of the chain.

Problem 4 Solution

Theorem

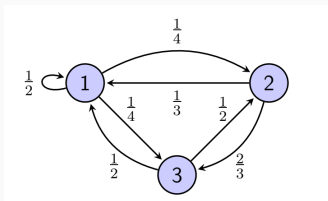
If for an irreducible Markov chain with transition matrix $Q = (q_{i,j})$, there exists a probability solution π to the detailed balance equations

$$\pi_i q_{i,j} = \pi_j q_{j,i}$$

for all pairs of states i, j , then this Markov chain is reversible and the solution π is the unique stationary distribution.

Problem 5

Given a Markov chain with state-transition diagram shown as follows:



- (a) Find $P(X_3 = 3 \mid X_2 = 2)$ and $P(X_4 = 1 \mid X_3 = 2)$.
- (b) If $P(X_0 = 2) = \frac{2}{5}$, find $P(X_0 = 2, X_1 = 3, X_2 = 1)$.
- (c) Find $P(X_2 = 1 \mid X_0 = 2)$, $P(X_2 = 2 \mid X_0 = 2)$, and $P(X_2 = 3 \mid X_0 = 2)$.
- (d) Find $E(X_2 \mid X_0 = 2)$.

Problem 5 Solution

Problem 6

A fair coin is flipped repeatedly. We use H to denote the "Head appeared" and T to denote the "Tail appeared".

- (a) What is the expected number of flips until the pattern HTHT is observed?
- (b) What is the expected number of flips until the pattern THTT is observed?
- (c) What is the probability that pattern HTHT is observed earlier than THTT?

Problem 6 Solution

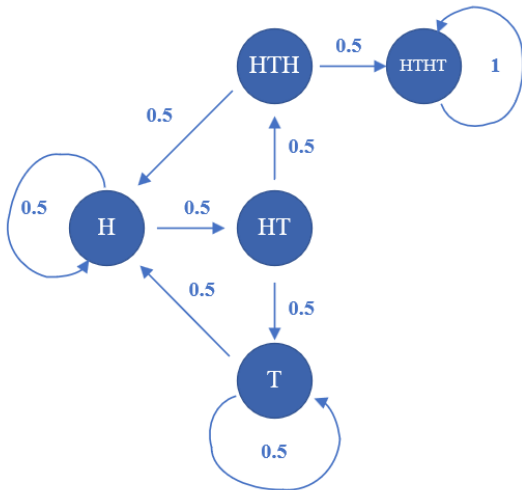


Figure 1: 6(1)

Problem 6 Solution

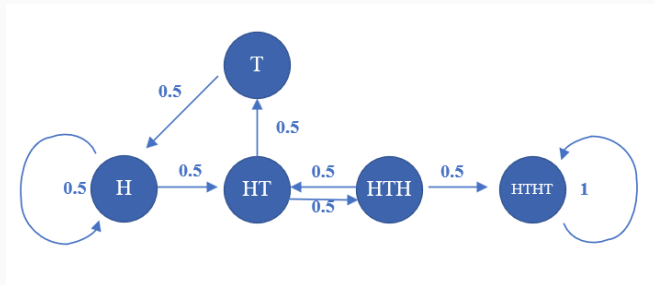


Figure 2: 6(1)

Problem 6 Solution

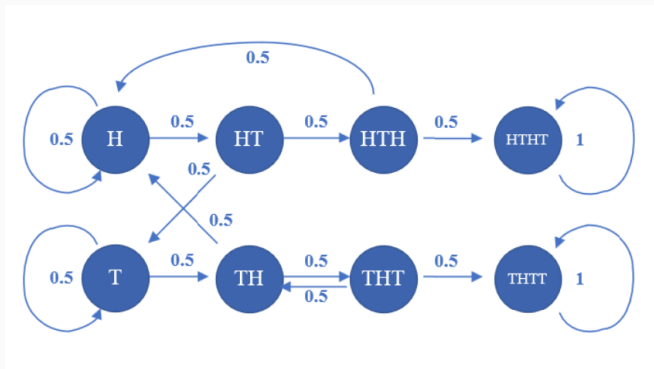


Figure 3: 6(1)

HW Problems

Final Review

Previous Final Exams

Outline of Topics

Part I

- Probability & Counting
- Conditional Probability
- Random Variables
- Expectations
- Continuous Random Variables

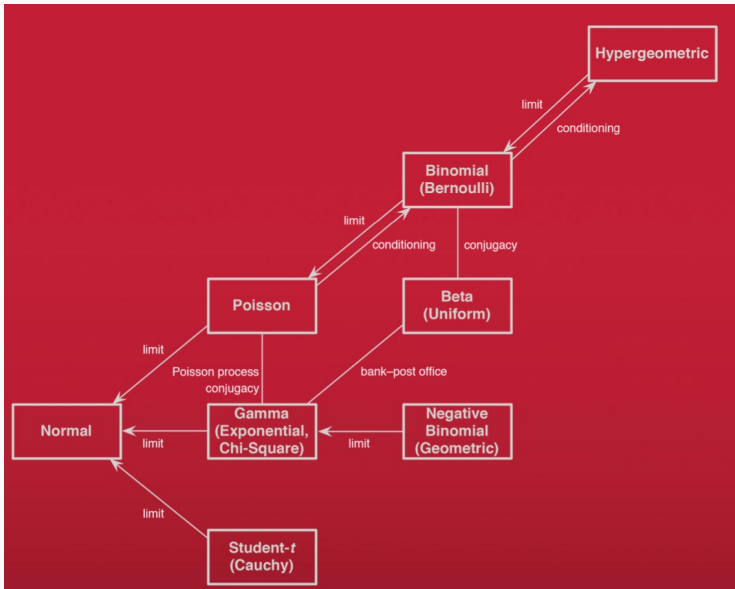
Part II

- Joint Distributions
- Transformations
- Monte Carlo Methods & Concentration Inequalities
- Statistical Inference
- Markov Chains

Random Variables

- First success & Geometric
- Exponential & Poisson & Gamma
- Bernoullis
- Uniform
- Normal & MVN
- Beta & Binomial

Random Variables



Random Variables

- Ordering
 - Order statistics
 - Max & Min operators
- Jointness
 - Independence: pairwise & conditional
 - Correlation & Covariance
- Transformation
 - Change of variables & convolution
 - Inverse transform method
- Relationship
 - Discrete to continuous with δ -step method
 - Conjugacy: Beta-Binomial, Normal-Normal, Gamma-Pois
 - Connection: Beta-Gamma, Uniform-Beta, Binomial-Gamma

Independence

- $P(X, Y) = P(X)P(Y)$
- $P(X|Y) = P(X)$ with $P(Y) \neq 0$
- Factorization of PDF $f_{X,Y}(x,y)$ and MGF $M_{X,Y}(t)$
- $E(XY) = E(X)E(Y)$
- $\text{Corr}(X, Y) = \text{Cov}(X, Y) = 0$

- Bayes' rule, LOTP & LOTE, LOTUS: 1D & 2D
- Indicator & linearity of expectation
- First-step analysis & recursive equations
- Conditional expectation: Adam's & Eve's law
- Generating functions: PGF & MGF
- Symmetry
 - $X + Y, XY, |X - Y|, \frac{X}{Y}, \frac{X}{X+Y}, \frac{Y}{X+Y}$
 - The property of i.i.d. continuous random variables
 - Normal distributions

Model-Based Problems

- Birthday problem: static & dynamic
- Sequence of coin tosses: biased or not
- Gambler's ruin & Random walk
- Coupon collector: given total number or not
- Pattern matching: coin or dice
- Chicken-egg problem & Poisson process
- Bank-post office

Model-Free Problems

- Computation via definitions
 - PMF, PDF, CDF, Joint distribution
 - Expectation, PGF, MGF
 - Markov chains
- Approximation
 - CLT & Law of Large Number
 - Poisson approximation & Law of Small Number
 - Non-asymptotic inequalities
- Estimation
 - MLE & MAP
 - Confidence interval
 - MMSE & LLSE

HW Problems

Final Review

Previous Final Exams

- Please describe the difference and connection between probability and statistics.
- Please describe the pros and cons of Bayesian statistical inference and classical statistical inference.

Q1 Solution

Let X be a discrete r.v. whose distinct possible values are x_0, x_1, \dots , and let $p_k = P(X = x_k)$. The entropy of X is $H(X) = \sum_{k=0}^{\infty} p_k \log(1/p_k)$

- Find $H(X)$ for $X \sim \text{Geom}(p)$.
- Let X and Y be i.i.d. discrete r.v.s. Show that $P(X = Y) \geq 2^{-H(X)}$

Q2 Solution

Let $X_1 \sim \text{Expo}(\lambda_1)$, $X_2 \sim \text{Expo}(\lambda_2)$ and $X_3 \sim \text{Expo}(\lambda_3)$ be independent.

- Find $E(X_1 + X_2 + X_3 | X_1 > 1, X_2 > 2, X_3 > 3)$ in terms of $\lambda_1, \lambda_2, \lambda_3$.
- Find $P(X_1 = \min(X_1, X_2, X_3))$

Q3 Solution

Let $X \sim \text{Gamma}(a, \lambda)$, $Y \sim \text{Gamma}(b, \lambda)$. Assume X and Y are independent.

- Find the joint distribution of $T = X + Y$ and $W = \frac{X}{X+Y}$.
- Find the distribution of T and W respectively.
- Find $E(W)$

Instead of predicting a single value for the parameter, we give an interval that is likely to contain the parameter: A $1 - \delta$ confidence interval for a parameter p is an interval $[\hat{p} - \epsilon, \hat{p} + \epsilon]$ such that $\Pr(p \in [\hat{p} - \epsilon, \hat{p} + \epsilon]) \geq 1 - \delta$. Now we toss a coin with probability p landing heads and probability $1 - p$ landing tails. The parameter p is unknown and we need to estimate its value from experiments results. We toss such coin N times. Let $X_i = 1$ if the i th result is head, otherwise 0. We estimate p by using $\hat{p} = \frac{X_1 + \dots + X_N}{N}$. Find the $1 - \delta$ confidence interval for p , then discuss the impacts of δ and N .

Hint: You can use the following Hoeffding bound: Let the random variables X_1, X_2, \dots, X_n be independent with $E(X_i) = \mu, a \leq X_i \leq b$ for each $i = 1, \dots, n$, where a, b are constants. Then for any $\epsilon \geq 0$,

$$\mathbb{P} \left(\left| \frac{1}{n} \sum_{i=1}^n X_i - \mu \right| \geq \epsilon \right) \leq 2e^{-\frac{2n\epsilon^2}{(b-a)^2}}$$

Q5 Solution

Given a coin with the probability p of landing heads. p is unknown and we need to estimate its value through data. In our data collection model, we have n independent tosses, the result of each toss is either Head or Tail. Let X denote the number of heads in the total n tosses. Now we conduct experiments to collect data and find $X = k$. Then we need to find \hat{p} , the estimation of p .

- Assume p is an unknown constant. Find \hat{p} through the MLE (Maximum Likelihood Estimation) rule.
- Assume p is a random variable with a prior distribution $p \sim \text{Beta}(a, b)$, where a and b are known constants. Find \hat{p} through the MAP (Maximum a Posterior Probability) rule.
- Assume p is a random variable with a prior distribution $p \sim \text{Beta}(a, b)$, where a and b are known constants. Find \hat{p} through the MMSE (Minimal Mean Squared Error) rule.

Q6 Solution

We know that the MMSE of X given Y is given by $g(Y) = E[X | Y]$. We also know that the Linear Least Square Estimate (LLSE) of X given Y , denoted by $L[Y | X]$, is shown as follows:

$$L[Y | X] = E(Y) + \frac{\text{Cov}(X, Y)}{\text{Var}(X)}(X - E(X))$$

Now we wish to estimate the probability of landing heads, denoted by θ , of a biased coin. We model θ as the value of a random variable Θ with a known prior PDF $f_{\Theta} \sim \text{Unif}(0, 1)$. We consider n independent tosses and let X be the number of heads observed.

- Show that $E[(\Theta - E[\Theta | X])h(X)] = 0$ for any real function $h(\cdot)$.
- Find the MMSE $E[\Theta | X]$ and LLSE $L[\Theta | X]$. (Eve's law: $\text{Var}(Y) = E(\text{Var}(Y | X)) + \text{Var}(E(Y | X)).$)

Q7 Solution

Show the following inequalities.

(a) Let $X \sim \text{Pois}(\lambda)$. If there exists a constant $a > \lambda$, then

$$\mathbb{P}(X \geq a) \leq \frac{e^{-\lambda}(e\lambda)^a}{a^a}.$$

(b) Let X be a random variable with finite variance σ^2 . Then for any constant $a > 0$,

$$\mathbb{P}(|X - \mathbb{E}[X]| \geq a) \leq \frac{2\sigma^2}{\sigma^2 + a^2}.$$

Q8 Solution

Let $X \sim \mathcal{N}(0, 1)$, $Y \sim \mathcal{N}(0, 1)$; X and Y are independent. Now let $Z_1 = \sin(X + Y)$, $Z_2 = \cos(X + Y)$.

(a) Find $E(Z_1)$ and $E(Z_2)$.

(b) Find $\text{Var}(Z_1)$ and $\text{Var}(Z_2)$.

- Please describe the difference and connection between probability and statistics.
- Please describe the pros and cons of Bayesian statistical inference and classical statistical inference.

Let X and Y be two continuous random variables with joint PDF

$$f_{X,Y}(x,y) = \begin{cases} x + cy^2 & \text{if } 0 \leq x \leq 1, 0 \leq y \leq 1 \\ 0 & \text{otherwise} \end{cases}$$

- (a) Find the value of constant c .
- (b) Find the joint probability $P(0 \leq X \leq 1/2, 0 \leq Y \leq 1/2)$.

Q2 Solution

Let X and Y be two continuous random variables with joint PDF

$$f_{X,Y}(x,y) = \begin{cases} 6xy & \text{if } 0 \leq x \leq 1, 0 \leq y \leq \sqrt{x} \\ 0 & \text{otherwise} \end{cases}$$

(a) Find the marginal distributions of X and Y . Are X and Y independent?

(b)) Find $E[X \mid Y = y]$ and $\text{Var}[X \mid Y = y]$ for $0 \leq y \leq 1$.

Q3 Solution

Given a coin with the probability p of landing heads. p is unknown and we need to estimate its value through data. In our data collection model, we have n independent tosses, the result of each toss is either Head or Tail. Let X denote the number of heads in the total n tosses. Now we conduct experiments to collect data and find $X = k$. Then we need to find \hat{p} , the estimation of p .

- Assume p is an unknown constant. Find \hat{p} through the MLE (Maximum Likelihood Estimation) rule.
- Assume p is a random variable with a prior distribution $p \sim \text{Beta}(a, b)$, where a and b are known constants. Find \hat{p} through the MAP (Maximum a Posterior Probability) rule.
- Assume p is a random variable with a prior distribution $p \sim \text{Beta}(a, b)$, where a and b are known constants. Find \hat{p} through the MMSE (Minimal Mean Squared Error) rule.

We know that the MMSE of X given Y is given by $g(Y) = E[X | Y]$. We also know that the Linear Least Square Estimate (LLSE) of X given Y , denoted by $L[Y | X]$, is shown as follows:

$$L[Y | X] = E(Y) + \frac{\text{Cov}(X, Y)}{\text{Var}(X)}(X - E(X))$$

Now we wish to estimate the probability of landing heads, denoted by θ , of a biased coin. We model θ as the value of a random variable Θ with a known prior PDF $f_{\Theta} \sim \text{Unif}(0, 1)$. We consider n independent tosses and let X be the number of heads observed.

- Show that $E[(\Theta - E[\Theta | X])h(X)] = 0$ for any real function $h(\cdot)$.
- Find the MMSE $E[\Theta | X]$ and LLSE $L[\Theta | X]$. (Eve's law: $\text{Var}(Y) = E(\text{Var}(Y | X)) + \text{Var}(E(Y | X)).$)

Laplace's law of succession says that if X_1, X_2, \dots, X_{n+1} are conditionally independent Bern (p) r.v.s given p , but p is given a $\text{Unif}(0, 1)$ prior to reflect ignorance about its value, then

$$P(X_{n+1} = 1 \mid X_1 + \dots + X_n = k) = \frac{k+1}{n+2}.$$

As an example, Laplace discussed the problem of predicting whether the sun will rise tomorrow, given that the sun did rise every time for all n days of recorded history; the above formula then gives $(n+1)/(n+2)$ as the probability of the sun rising tomorrow (of course, assuming independent trials with p unchanging over time may be a very unreasonable model for the sunrise problem).

- (a) Find the posterior distribution of p given $X_1 = x_1, \dots, X_n = x_n$, and show that it only depends on the sum of the $x_j, j \in \{1, \dots, n\}$.
- (b) Prove Laplace's law of succession, using a form of LOTP to find

$$P(X_{n+1} = 1 \mid X_1 + \dots + X_n = k)$$

by conditioning on p .

Q6 Solution

A handy rule of thumb in statistics and life is as follows: Conditioning often makes things better. This problem explores how the above rule of thumb applies to estimating unknown parameters. Let θ be an unknown parameter that we wish to estimate based on data X_1, \dots, X_n (these are r.v.s before being observed, and then after the experiment they "crystallize" into data). In this problem, θ is viewed as an unknown constant, and is not treated as an r.v. as in the Bayesian statistical inference. Let T_1 be an estimator for θ (this means that T_1 is a function of X_1, \dots, X_n which is being used to estimate θ). A strategy for improving T_1 (in some problems) is as follows.

Suppose that we have an r.v. R such that $T_2 = E(T_1 | R)$ is a function of X_1, \dots, X_n (in general, $E(T_1 | R)$ might involve unknowns such as θ but then it couldn't be used as an estimator). Also suppose that $P(T_1 = T_2) < 1$, and that $E(T_1^2)$ is finite. (a) Use Jensen's inequality and Adam's Law to show that T_2 is better than T_1 in the sense that the mean squared error is less, i.e.,

$$E[(T_2 - \theta)^2] < E[(T_1 - \theta)^2].$$

(b) The bias of an estimator T for θ is defined to be $b(T) = E(T) - \theta$. An important identity in statistics, a form of the bias-variance trade-off, is that the mean squared error equals the variance plus the squared bias:

$$E[(T - \theta)^2] = \text{Var}(T) + (b(T))^2.$$

Use this identity and Eve's law to give an alternative proof of the result from (a).

(c) Now suppose that X_1, \dots, X_n are i.i.d. with mean θ , and consider the special case $T_1 = X_1, R = \sum_{j=1}^n X_j$. Find T_2 in a simplified form, and check that it has a lower mean squared error than T_1 for $n \geq 2$. Also, explain what happens to T_1 and T_2 as $n \rightarrow \infty$.

Q7 Solution

Let X and Y be two independent random variables satisfying first success distribution $FS(p)$.

(a) (5 points) Define $Z_1 = X - Y$. Find the PMF of Z_1 and $E(Z_1)$.

(b) (5 points) Define $Z_2 = \frac{X}{Y}$. Find the PMF of Z_2 and $E(Z_2)$.

Q8 Solution

A scientist makes two measurements X, Y , considered to be i.i.d. random variables.

(a) If $X, Y \sim \mathcal{N}(0, 1)$. Find the correlation between the larger and smaller of the values, i.e., $\text{Corr}(\max(X, Y), \min(X, Y))$.

(b) If $X, Y \sim \text{Unif}(0, 1)$. Find the correlation between the larger and smaller of the values, i.e., $\text{Corr}(\max(X, Y), \min(X, Y))$.

(5 points) Please describe the pros and cons of **Bayesian statistical inference** and **Classical statistical inference**. Then explain why conjugate priors are important for Bayesian statistical inference.

(10 points) Let X, Y be two continuous random variables with joint PDF

$$f_{X,Y}(x,y) = \begin{cases} cx^2y & \text{if } 0 \leq y \leq x \leq 1, \\ 0 & \text{otherwise.} \end{cases}$$

(a) (5 points) Find the value of constant c .

(b) (5 points) Find the conditional probability $P(Y \leq \frac{X}{4} | Y \leq \frac{X}{2})$.

(15 points) Let X and Y be two integer random variables with joint PMF

$$P_{X,Y}(x,y) = \begin{cases} \frac{1}{6 \cdot 2^{\min(x,y)}} & \text{if } x, y \geq 0, |x - y| \leq 1, \\ 0 & \text{otherwise.} \end{cases}$$

- (a) (5 points) Find the marginal distributions of X and Y . Are X and Y independent?
- (b) (5 points) Find $P(X = Y)$.
- (c) (5 points) Find $E[X|Y = 2]$ and $\text{Var}[X|Y = 2]$.

Q3 solution

(20 points) Let Z_1, Z_2 be two *i.i.d.* random variables satisfying standard normal distributions, *i.e.*, $Z_1, Z_2 \sim \mathcal{N}(0, 1)$. Define

$$X = \sigma_X Z_1 + \mu_X;$$

$$Y = \sigma_Y \left(\rho Z_1 + \sqrt{1 - \rho^2} Z_2 \right) + \mu_Y,$$

where $\sigma_X > 0, \sigma_Y > 0, -1 < \rho < 1$.

- (a) (5 points) Show that X and Y are bivariate normal.
- (b) (5 points) Find the correlation coefficient between X and Y , *i.e.*, $\text{Corr}(X, Y)$.
- (c) (5 points) Find the joint PDF of X and Y .
- (d) (5 points) Find $E[Y|X]$ and $\text{Var}[Y|X]$.

(10 points) Let the random variable $X \sim \mathcal{N}(\mu, \tau^2)$. Given $X = x$, random variables Y_1, Y_2, \dots, Y_n are *i.i.d.* and have the same conditional distribution, *i.e.*, $Y_i|X = x \sim \mathcal{N}(x, \sigma^2)$. Define the sample mean \bar{Y} as follows:

$$\bar{Y} = \frac{Y_1 + \dots + Y_n}{n}.$$

- (a) (4 points) Find the posterior PDF of X given \bar{Y} .
- (b) (3 points) Find the MAP (Maximum a Posterior Probability) estimates of X given \bar{Y} .
- (c) (3 points) Find the MMSE estimates of X given \bar{Y} . (We know that the MMSE of X given Y is given by $g(Y) = E[X|Y]$).

(10 points) Let $X_1 \sim \text{Expo}(\lambda_1)$, $X_2 \sim \text{Expo}(\lambda_2)$ and $X_3 \sim \text{Expo}(\lambda_3)$ be independent.

(a) (5 points) Find $E(X_1 + X_2 + X_3 | X_1 > 1, X_2 > 2, X_3 > 3)$ in terms of $\lambda_1, \lambda_2, \lambda_3$.

(b) (5 points) Find $P(X_1 = \min(X_1, X_2, X_3))$.

(10 points) Let $X \sim \text{Expo}(\lambda)$, $Y \sim \text{Expo}(\lambda)$; X and Y are independent.

(a) (5 points) Find $E(X|X + Y)$.

(b) (5 points) Find $E(X^2|X + Y)$.

(10 points) Instead of predicting a single value for the parameter, we give an interval that is likely to contain the parameter: A $1 - \delta$ confidence interval for a parameter p is an interval $[\hat{p} - \epsilon, \hat{p} + \epsilon]$ such that $P(p \in [\hat{p} - \epsilon, \hat{p} + \epsilon]) \geq 1 - \delta$. Now we toss a coin with probability p landing heads and probability $1 - p$ landing tails. The parameter p is unknown and we need to estimate its value from experiment results. We toss such coin N times. Let $X_i = 1$ if the i th result is head, otherwise 0. We estimate p by using $\hat{p} = \frac{X_1 + \dots + X_N}{N}$. Find the $1 - \delta$ confidence interval for p , then discuss the impacts of δ and N .

Hint: You can use the following Hoeffding bound: Let the random variables X_1, X_2, \dots, X_n be independent with $E(X_i) = \mu, a \leq X_i \leq b$ for each $i = 1, \dots, n$, where a, b are constants. Then for any $\epsilon \geq 0$,

$$P\left(\left|\frac{1}{n} \sum_{i=1}^n X_i - \mu\right| \geq \epsilon\right) \leq 2 \exp\left(-\frac{2n\epsilon^2}{(b-a)^2}\right).$$

Q8 solution

(10 points) Show the following inequalities.

(a) (5 points) Let $X \sim \text{Pois}(\lambda)$. If there exists a constant $a > \lambda$, then

$$P(X \geq a) \leq \frac{e^{-\lambda}(e\lambda)^a}{a^a}.$$

(b) (5 points) Let X be a random variable with finite variance σ^2 . Then for any constant $a > 0$,

$$P(|X - \mathbb{E}[X]| \geq a) \leq \frac{2\sigma^2}{\sigma^2 + a^2}.$$

Please, Check Out the Previous Example Papers.

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Please, Check Out the HWs.

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