

STAT206B Homework #1

Due: 01/20 (Th) - Tentative.

1. Let $X \sim \text{Exp}(\lambda)$, where $E(X) = 1/\lambda$. What is the pmf (probability mass function) of $Y = \lfloor X \rfloor$ (the floor of X)? Do you recognize it as a distribution that you have studied in the past?
2. Let X_1 and X_2 be two independent random variables such that $X_i \sim \text{Ga}(a_i, b)$ for any $a_1, a_2, b > 0$. Define $Y = X_1/(X_1 + X_2)$ and $Z = (X_1 + X_2)$.
 - (a) Find the joint pdf for Y and Z and show that these two random variables are independent.
 - (b) Find the marginal pdf of Z . Do you recognize this pdf as belonging to some family that you know?
 - (c) Find the marginal pdf of Y . Do you recognize this pdf as belonging to some family that you know?
 - (d) Compute $E(Y^k)$ for any $k > 0$.
 - (e) What does this result imply if $a_i = b = 1$?
3. Consider three independent random variables X_1, X_2 and X_3 such that $X_i \stackrel{\text{indep}}{\sim} \text{Gamma}(a_i, b)$, $i = 1, 2, 3$. Let

$$\mathbf{Y} = (Y_1, Y_2, Y_3) = \left(\frac{X_1}{X_1 + X_2 + X_3}, \frac{X_2}{X_1 + X_2 + X_3}, \frac{X_3}{X_1 + X_2 + X_3} \right).$$

- (a) Show that $\mathbf{Y} \sim \text{Dirichlet}(a_1, a_2, a_3)$, a Dirichlet distribution.
 - (b) How can this result be used to generate random variables according to a Dirichlet distribution? Write a simple function in R or Matlab (your choice) that takes as inputs n , the number of trivariate vectors to be generated, and $\mathbf{a} = (a_1, a_2, a_3)$ and generates a matrix of size $n \times 3$ whose rows correspond to independent samples from a Dirichlet distribution with parameter (a_1, a_2, a_3) .
Use each of $\mathbf{a} = (0.01, 0.01, 0.01)$, $(100, 100, 100)$, and $(3, 5, 10)$ and comment how the density of \mathbf{Y} changes over \mathbf{a} .
4. Y follows an inverse Gamma distribution with shape parameter a and scale parameter b ($Y \sim \text{IG}(a, b)$) if $Y = 1/X$ with $X \sim \text{Gamma}(a, b)$ (assume the Gamma distribution is parameterized such that $E(X) = ab$).
 - (a) Find the density of Y .
 - (b) Compute $E(Y^k)$. Do you need to impose any constrain on the problem for this expectation to exists?
 - (c) Compare $E(Y^k)$ to $1/E(X^k)$ (hint: look at the ratio of the two quantities).
 5. Y follows a log normal distribution with parameters μ and σ^2 (denotes as $Y \sim \text{Log-N}(\mu, \sigma^2)$) if $Y = \exp(X)$ where $X \sim \text{N}(\mu, \sigma^2)$.
 - (a) Find the density of Y .

(b) Compute the mean and the variance of Y .

6. Let $\mathbf{X} = (X_1, X_2, \dots, X_p)$ with $X \sim N_p(\boldsymbol{\mu}, \Sigma)$ and set $\mathbf{Z}_1 = (X_1, \dots, X_q)$ and $\mathbf{Z}_2 = (X_{q+1}, \dots, X_p)$ with $1 < q < p$. Show that

$$\mathbf{Z}_1 | \mathbf{Z}_2 \sim N_q(\boldsymbol{\mu}_1 + \Sigma_{12}\Sigma_{22}^{-1}(\mathbf{Z}_2 - \boldsymbol{\mu}_2), \Sigma_{11} - \Sigma_{12}\Sigma_{22}^{-1}\Sigma_{21}),$$

where $\boldsymbol{\mu}_k$ and $\Sigma_{k\ell}$ denote the blocks of $\boldsymbol{\mu}$ and Σ where the rows correspond to the variables in \mathbf{Z}_k and the columns to the variables in \mathbf{Z}_ℓ .

7. Show that if $X \sim \text{Exp}(\beta)$, then

- (a) $Y = X^{1/\gamma}$ has a Weibull distribution with parameters γ and β with $\gamma > 0$ a constant.
(b) $Y = (2X/\beta)^{1/2}$ has the Rayleigh distribution.

For both parts, derive the form of the pdf, verify that is a pdf, and calculate the mean and the variance.

8. Let $Y | X \sim \text{Poisson}(X)$ and let $X \sim \text{Exp}(\lambda)$. What is the marginal distribution of Y ?
9. (Robert) If $y \sim \text{Binomial}(n, \theta)$ and $x \sim \text{Binomial}(m, \theta)$, and $\theta \sim \text{Beta}(\alpha, \beta)$. Find the predictive distribution of y given x .
10. (Robert) Give the posterior and the marginal distributions in the following cases:
- (a) $x | \sigma^2 \sim N(0, \sigma^2)$ and $1/\sigma^2 \sim \text{Gamma}(1, 2)$.
(b) $x | p \sim \text{Negative-Binomial}(10, p)$ and $p \sim \text{Beta}(1/2, 1/2)$.
11. Assume that an observation, x_1, \dots, x_n are iid from $N(\theta, \sigma^2)$, where μ and σ^2 are unknown. Consider $\tilde{\pi}(\theta, \sigma^2) \propto 1/\sigma^2$ (not a probability density, i.e., improper, Jeffreys prior).
- (a) Find the joint posterior distribution.
(b) Find the posterior distributions $\pi(\theta | \bar{x}, s^2, \sigma^2)$ and $\pi(\sigma^2 | \bar{x}, s^2)$.
(c) Find the marginal posterior distribution of θ , $\pi(\theta | \bar{x}, s^2)$.
12. Consider $\mathbf{x}_i | \boldsymbol{\theta}, \Sigma \stackrel{iid}{\sim} N_p(\boldsymbol{\theta}, \Sigma)$, $i = 1, \dots, n$, where $N_p(\boldsymbol{\theta}, \Sigma)$ represents the p -dimensional normal distribution with mean vector $\boldsymbol{\theta} \in \mathbb{R}^p$ and covariance matrix Σ ($p \times p$ positive definite matrix). Suppose $\boldsymbol{\theta}$ and Σ are unknown. Consider the following conjugate prior distributions;

$$\boldsymbol{\theta} | \Sigma \sim N_p(\boldsymbol{\mu}, 1/n_0 \Sigma), \text{ and } \Sigma^{-1} \sim \text{Wishart}_p(\alpha, W).$$

The Wishart distribution is described in Robert Exercise #3.21. Note that if $\Sigma^{-1} \sim \text{Wishart}_p(\alpha, W)$, $\Sigma \sim \text{inverse-Wishart}_p(\alpha, W^{-1})$.

fact: Given n observations $\mathbf{x}_1, \dots, \mathbf{x}_n$ of $N_p(\boldsymbol{\theta}, \Sigma)$, a sufficient statistic is

$$\bar{\mathbf{x}} = \frac{1}{n} \sum_{i=1}^n \mathbf{x}_i, \text{ and } S = \sum_{i=1}^n (\mathbf{x}_i - \bar{\mathbf{x}})(\mathbf{x}_i - \bar{\mathbf{x}})^t.$$

Tip: Read Robert §4.4.2 (and/or BDA §3.6).

- (a) Find an expression of the joint posterior distribution as follows;

$$\pi(\boldsymbol{\theta}, \Sigma^{-1} \mid \bar{\mathbf{x}}, S) = \pi_1(\boldsymbol{\theta} \mid \Sigma, \bar{\mathbf{x}}, S) \pi_2(\Sigma^{-1} \mid \bar{\mathbf{x}}, S).$$

Also, identify $\pi_1(\boldsymbol{\theta} \mid \Sigma, \bar{\mathbf{x}}, S)$ and $\pi_2(\Sigma^{-1} \mid \bar{\mathbf{x}}, S)$

- (b) Is the prior conjugate? Explain.