

# 2014 Adv Econometrics Midterm: Detailed Solutions

Xiankang Wang

November 7, 2025

## Contents

<b>1 Problem 1: Mixed germination and Bayes posterior</b>	<b>2</b>
1.1 Problem Statement . . . . .	2
1.2 Solution . . . . .	2
<b>2 Problem 2: Functions of i.i.d. standard normals</b>	<b>2</b>
2.1 Problem Statement . . . . .	2
2.2 Solution . . . . .	2
<b>3 Problem 3: Normalizing constant, marginals, and a conditional probability</b>	<b>3</b>
3.1 Problem Statement . . . . .	3
3.2 Solution . . . . .	3
<b>4 Problem 4: Ratio <math>W = X/(X + Y)</math> for independent Gammas</b>	<b>3</b>
4.1 Problem Statement . . . . .	3
4.2 Solution . . . . .	3
<b>5 Problem 5: Triangular support <math>0 &lt; y &lt; x &lt; 1</math></b>	<b>4</b>
5.1 Problem Statement . . . . .	4
5.2 Solution . . . . .	4
<b>6 Problem 6: Thinning a Poisson process</b>	<b>4</b>
6.1 Problem Statement . . . . .	4
6.2 Solution . . . . .	5

## Todo list

# 1 Problem 1: Mixed germination and Bayes posterior

## 1.1 Problem Statement

Rutabaga seeds from growers A, B, C have germination rates 0.90, 0.80, 0.85 respectively. A packet is a mixture drawn with proportions 0.25 (A), 0.35 (B), 0.40 (C). (i) Find the probability a random seed germinates. (ii) Given a seed did not germinate, find the posterior probability it came from grower B.

## 1.2 Solution

Let  $G$  be the event of germination.

(i) Mixture success probability.

$$\Pr(G) = 0.25 \cdot 0.90 + 0.35 \cdot 0.80 + 0.40 \cdot 0.85 = [0.845].$$

(ii) Posterior for source B given failure. Let  $\bar{G}$  denote failure. Then  $\Pr(\bar{G}) = 0.25 \cdot 0.10 + 0.35 \cdot 0.20 + 0.40 \cdot 0.15 = [0.155]$ . Bayes:

$$\Pr(B | \bar{G}) = \frac{\Pr(\bar{G} | B) \Pr(B)}{\Pr(\bar{G})} = \frac{0.20 \cdot 0.35}{0.155} = \left[ \frac{7}{15.5} \approx 0.4516 \right].$$

# 2 Problem 2: Functions of i.i.d. standard normals

## 2.1 Problem Statement

Suppose  $X_i \stackrel{i.i.d.}{\sim} \mathcal{N}(0, 1)$ ,  $i = 1, \dots, k$ . (i) Find the distribution of  $Y = \sum_{i=1}^k X_i^2$ . (ii) For constants  $a_1, \dots, a_k \in \mathbb{R}$ , find the distribution of  $Z = \sum_{i=1}^k a_i X_i$  (in particular, specialize to  $\sum a_i^2 = 1$ ).

## 2.2 Solution

(i)  $\chi^2$  law. By definition  $Y \sim \chi_k^2$ ; its mgf is  $M_Y(t) = (1 - 2t)^{-k/2}$  for  $t < 1/2$ .

(ii) Linear form. Vectorize  $X \sim \mathcal{N}(0, I_k)$ . Then  $Z = a'X \sim \mathcal{N}(0, a'a)$ . In particular, if  $a'a = 1$ , then  $Z \sim \mathcal{N}(0, 1)$ .

### 3 Problem 3: Normalizing constant, marginals, and a conditional probability

#### 3.1 Problem Statement

Let  $X$  and  $Y$  be the fractions of time during a working day when the dining room and drive-thru are busy. The joint density is

$$f_{X,Y}(x,y) = k(x^2 + y^2), \quad 0 < x < 1, 0 < y < 1, 0 \text{ otherwise.}$$

- (i) Find  $k$ . (ii) Find the marginal densities  $f_X, f_Y$ . (iii) Compute  $\Pr(Y > 0.75 | X < 0.10)$ .

#### 3.2 Solution

##### (i) Normalizing constant.

$$1 = \int_0^1 \int_0^1 k(x^2 + y^2) dy dx = k\left(\frac{1}{3} + \frac{1}{3}\right) = \frac{2k}{3} \Rightarrow \boxed{k = \frac{3}{2}}.$$

##### (ii) Marginals.

$$f_X(x) = \int_0^1 \frac{3}{2}(x^2 + y^2) dy = \frac{3}{2}\left(x^2 + \frac{1}{3}\right), \quad 0 < x < 1; \quad f_Y(y) = \frac{3}{2}\left(\frac{1}{3} + y^2\right), \quad 0 < y < 1.$$

##### (iii) Conditional probability.

$$\Pr(Y > 0.75 | X < 0.10) = \frac{\int_0^{0.10} \int_{0.75}^1 \frac{3}{2}(x^2 + y^2) dy dx}{\int_0^{0.10} \int_0^1 \frac{3}{2}(x^2 + y^2) dy dx} = \frac{\frac{3}{2} \left[ \frac{(0.1)^3}{3} \cdot \frac{1}{4} + 0.1 \cdot \frac{1 - 0.75^3}{3} \right]}{\frac{3}{2} \left[ \frac{(0.1)^3}{3} + 0.1 \cdot \frac{1}{3} \right]} = \boxed{\frac{929}{1616} \approx 0.5749}.$$

### 4 Problem 4: Ratio $W = X/(X+Y)$ for independent Gammas

#### 4.1 Problem Statement

Let  $X$  and  $Y$  be independent with densities proportional to  $x^{\alpha-1}e^{-x}$  and  $y^{\beta-1}e^{-y}$  on  $(0, \infty)$ , with  $\alpha, \beta > 0$  (normalizing constants not required). Find the density of  $W = \frac{X}{X+Y}$ .

#### 4.2 Solution

Take the one-to-one transform  $(U, W) = (X+Y, X/(X+Y))$  so that  $(X, Y) = (UW, U(1-W))$  with Jacobian  $|\partial(x, y)/\partial(u, w)| = u$ . The joint density is, up to a constant,

$$f_{U,W}(u, w) \propto (uw)^{\alpha-1} e^{-uw} (u(1-w))^{\beta-1} e^{-u(1-w)} u = u^{\alpha+\beta-1} e^{-u} w^{\alpha-1} (1-w)^{\beta-1}.$$

Integrating out  $u > 0$  yields a finite constant (the Gamma function) times  $w^{\alpha-1}(1-w)^{\beta-1}$  on  $w \in (0, 1)$ . Hence

$$f_W(w) \propto w^{\alpha-1}(1-w)^{\beta-1}, \quad 0 < w < 1 \quad (\text{i.e., } W \sim \text{Beta}(\alpha, \beta)).$$

## 5 Problem 5: Triangular support $0 < y < x < 1$

### 5.1 Problem Statement

Let  $(X, Y)$  have joint density  $f_{X,Y}(x, y) = c$  on  $\{(x, y) : 0 < y < x < 1\}$  and 0 otherwise. Compute:

(i) the conditional pdf  $f_{Y|X}(y | x)$  for  $0 < x < 1$ ; (ii)  $\mathbb{E}[Y | X = x]$ ; (iii)  $\text{Cov}(X, Y)$ .

### 5.2 Solution

The normalizing constant is  $c = 2$  (area of the triangle is  $1/2$ ).

**(i) Conditional pdf.** For fixed  $x \in (0, 1)$ ,

$$f_{Y|X}(y | x) = \frac{f_{X,Y}(x, y)}{f_X(x)} = \frac{2}{\int_0^x 2 dy} = \boxed{\frac{1}{x}, \quad 0 < y < x}.$$

**(ii) Conditional mean.** Since  $Y | X = x \sim \text{Unif}(0, x)$ ,  $\mathbb{E}[Y | X = x] = \boxed{x/2}$ .

**(iii) Covariance.** With  $f_X(x) = \int_0^x 2 dy = 2x$ ,  $\mathbb{E}[X] = \int_0^1 x 2x dx = 2/3$ , and  $\mathbb{E}[Y] = \mathbb{E}[\mathbb{E}(Y | X)] = \mathbb{E}[X/2] = 1/3$ . Moreover,

$$\mathbb{E}[XY] = \mathbb{E}[X \mathbb{E}(Y | X)] = \mathbb{E}[X \cdot X/2] = \frac{1}{2} \mathbb{E}[X^2] = \frac{1}{2} \int_0^1 x^2 2x dx = \frac{1}{4}.$$

Thus  $\text{Cov}(X, Y) = \mathbb{E}[XY] - \mathbb{E}[X] \mathbb{E}[Y] = \boxed{\frac{1}{36}}$ .

## 6 Problem 6: Thinning a Poisson process

### 6.1 Problem Statement

Suppose the number of events  $N \sim \text{Pois}(\lambda)$ . Each event is independently counted with probability  $p$ . Show that the counted and uncounted numbers are independent Poisson with means  $\lambda p$  and  $\lambda(1-p)$ .

## 6.2 Solution

Conditional on  $N = n$ , the counted number  $C \mid N = n \sim \text{Binom}(n, p)$  and the uncounted  $U = n - C$ . For  $c, u \in \mathbb{N}$ ,

$$\begin{aligned}
\Pr(C = c, U = u) &= \sum_{n=c+u}^{\infty} \Pr(N = n) \Pr(C = c \mid N = n) \\
&= \sum_{n=c+u}^{\infty} e^{-\lambda} \frac{\lambda^n}{n!} \binom{n}{c} p^c (1-p)^{n-c} \\
&= e^{-\lambda} \frac{(\lambda p)^c}{c!} \frac{(\lambda(1-p))^u}{u!} \sum_{n=c+u}^{\infty} \frac{\lambda^{n-c-u}}{(n-c-u)!} \\
&= e^{-\lambda} \frac{(\lambda p)^c}{c!} \frac{(\lambda(1-p))^u}{u!} e^\lambda \\
&= \underbrace{e^{-\lambda p} \frac{(\lambda p)^c}{c!}}_{\text{Pois}(\lambda p)} \underbrace{e^{-\lambda(1-p)} \frac{(\lambda(1-p))^u}{u!}}_{\text{Pois}(\lambda(1-p))}.
\end{aligned}$$

Thus  $C \sim \text{Pois}(\lambda p)$ ,  $U \sim \text{Pois}(\lambda(1-p))$ , and the factorization shows  $C \perp U$ .