

Probability 2017 Homework 3

March 20, 2017

1 Q1 (10 points)

Suppose that X is nonnegative continuous random variable. Prove that $E[X] = \int_0^{+\infty} P(X > x) dx$.

Solution:

Suppose $f(x)$ is the density function of X .

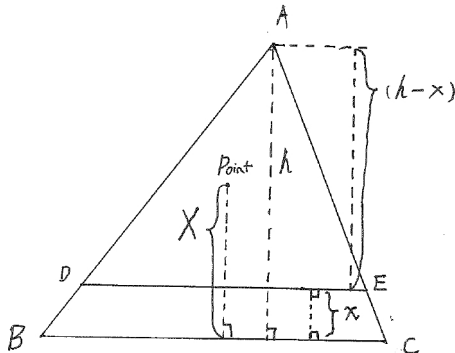
$$\begin{aligned}
 \int_0^{+\infty} P(X > x) dx &= \int_0^{+\infty} \int_x^{+\infty} f(y) dy dx && (\text{by } P(X > x) = \int_x^{+\infty} f(y) dy) \\
 &= \int_0^{+\infty} \int_0^y f(y) dx dy && (\text{by changing the order of integral}) \\
 &= \int_0^{+\infty} f(y) \int_0^y dx dy \\
 &= \int_0^{+\infty} f(y) y dy \\
 &= \int_{-\infty}^{+\infty} f(y) y dy && (\text{by the nonnegativity of } X) \\
 &= E[X]
 \end{aligned} \tag{1}$$

2 Q2 (10 points)

Consider a triangle and a point chosen within the triangle according to the uniform probability law. Let X be the distance from the point to the base of the triangle. Given the height of the triangle, find the CDF and the PDF of X .

Solution:

Suppose $\triangle ABC$ is the considered triangle. Let h be the height of $\triangle ABC$, S be the area of $\triangle ABC$. Draw a line parallel to the base within the triangle, whose distance to the base is x . The intersect of this line and the waist of the triangle is D and E respectively.



Then for $0 < x < h$:

$$\begin{aligned} P(X > x) &= P(\text{the point falls in } \triangle ADE) \\ &= S_{\triangle ADE}/S \quad (\text{by uniform probability law}) \\ &= (h-x)^2/h^2 \quad (\text{by the property of similar triangles}) \end{aligned} \quad (2)$$

Thus the CDF of X is:

$$\begin{aligned} P(X \leq x) &= 1 - P(X > x) \\ &= \begin{cases} 1 - (h-x)^2/h^2, & 0 < x < h \\ 0, & x \leq 0 \\ 1, & x \geq h \end{cases} \end{aligned} \quad (3)$$

The PDF is obtained by differentiating CDF:

$$\begin{aligned} f(x) &= \frac{d}{dx}[P(X \leq x)] \\ &= \begin{cases} 2(h-x)/h^2, & 0 < x < h \\ 0, & \text{otherwise} \end{cases} \end{aligned} \quad (4)$$

3 Q3 (10 points)

Please write down the PDF of a normal random variable with mean μ and standard deviation σ .

Solution:

Suppose X is a normal random variable with mean μ and standard deviation σ . Then the PDF of X is:

$$f(x) = \frac{1}{\sqrt{2\pi}\sigma} e^{-\frac{(x-\mu)^2}{2\sigma^2}} \quad (x \in \mathbb{R})$$

.

4 Q4 (10 points)

A point is chosen at random (according to a uniform PDF) within a semicircle of the form $\{(x, y) : x^2 + y^2 < r^2, y > 0\}$, for some given $r > 0$.

- Find the joint PDF of the coordinates X and Y of the chosen point.
- Find the marginal PDF of Y and use it to find $E[Y]$.

Solution:

- Let $B = \{(x, y) : x^2 + y^2 < r^2, y > 0\}$. the joint PDF of the coordinates X and Y is:

$$f(x, y) = \begin{cases} 2/\pi r^2, & (x, y) \in B \\ 0, & \text{otherwise} \end{cases} \quad (5)$$

- The marginal density of Y on $(0, r)$ is:

$$\begin{aligned} f_Y(y) &= \int_{-\infty}^{+\infty} f(x, y) dx \\ &= \int_{-\sqrt{r^2-y^2}}^{+\sqrt{r^2-y^2}} \frac{2}{\pi r^2} dx \\ &= \frac{4}{\pi r^2} \sqrt{r^2 - y^2} \end{aligned} \quad (6)$$

Thus, the marginal density of Y is:

$$f_Y(y) = \begin{cases} \frac{4}{\pi r^2} \sqrt{r^2 - y^2}, & y \in (0, r) \\ 0, & \text{otherwise} \end{cases} \quad (7)$$

The expectation of Y is:

$$\begin{aligned}
 E[Y] &= \int_{-\infty}^{+\infty} y f_Y(y) dy \\
 &= \int_0^r y \frac{4}{\pi r^2} \sqrt{r^2 - y^2} dy \\
 &= \frac{4r}{3\pi}
 \end{aligned} \tag{8}$$

5 Q5 (10 points)

Let X_1, X_2, \dots, X_n be independent random variables. We know that variance has linearity for independent random variables, namely

$$\text{Var}[X_1 + X_2 + \dots + X_n] = \text{Var}[X_1] + \text{Var}[X_2] + \dots + \text{Var}[X_n]$$

How about the variance of the product? Can you express

$$\text{Var}[X_1 X_2 \dots X_n]$$

in terms of $\text{Var}[X_i]$ and $E[X_i^2]$?

Solution:

First note two facts:

Fact 1: if X_1, X_2, \dots, X_n are independent, then $X_1^2, X_2^2, \dots, X_n^2$ are also independent.

Fact 2: if X and Y are independent, then $E[XY] = E[X]E[Y]$.

Then,

$$\begin{aligned}
 \text{Var}[X_1 X_2 \dots X_n] &\equiv E[(X_1 X_2 \dots X_n - E[X_1 X_2 \dots X_n])^2] \\
 &= E[(X_1 X_2 \dots X_n)^2 - 2EX_1 X_2 \dots X_n + E(X_1 X_2 \dots X_n)^2] \\
 &= E[(X_1 X_2 \dots X_n)^2] - 2E[X_1 X_2 \dots X_n]^2 + E[X_1 X_2 \dots X_n]^2 \\
 &= E[(X_1 X_2 \dots X_n)^2] - E[X_1 X_2 \dots X_n]^2 \\
 &= E[(X_1^2 X_2^2 \dots X_n^2)] - E[X_1 X_2 \dots X_n]^2 \\
 &= E[X_1^2]E[X_2^2] \dots E[X_n^2] - E[X_1 X_2 \dots X_n]^2 \quad (\text{by Fact 1 and Fact 2}) \\
 &= E[X_1^2]E[X_2^2] \dots E[X_n^2] - E[X_1]^2 E[X_2]^2 \dots E[X_n]^2 \quad (\text{by Fact 2}) \\
 &= E[X_1^2]E[X_2^2] \dots E[X_n^2] - [E[X_1^2] - \text{Var}(X_1)][E[X_2^2] - \text{Var}(X_2)] \dots [E[X_n^2] - \text{Var}(X_n)] \\
 &= \prod_{i=1}^n E[X_i^2] - \prod_{i=1}^n [E[X_i^2] - \text{Var}(X_i)]
 \end{aligned} \tag{9}$$