Continuous Random Variables 2 STAT-S520

Arturo Valdivia

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Exercise 1:

Let X be a continuous random variables and the function f() given by:

$$f(x) = \begin{cases} 2k & 0 \le x < 3 \\ k & 4 \le x < 8 \\ 0 & otherwise \end{cases}$$

- 1. Find k so f() is a PDF of X2. $P(X < 2) = \int_{-\infty}^{2} f(x) dx = 0.4$ 3. $P(3 < X \le 4) = 0$ 4. EX1. Find k so f() is a PDF of X1. EX1. e1. e

Exercise 1: (continued)

$$f(4) = 0.1$$

$$P(X = 4) = 0$$

$$f(4) \neq P(X = 4)$$

$$f(4) = 0.1$$

$$P(X = 4) = 0$$

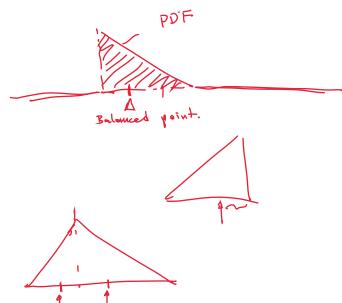
$$f(4) \neq P(X = 4)$$

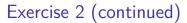
$$f(4) = 0.1$$

$$f(5) = 0.1$$

$$f(7) = 0.$$

Exercise 2: ISI Section 5.6. Exercise 5





Exercise 3: ISI Section 5.6. Exercise 4

$$P(A) = P(X \le 0.5)$$

$$P(A) = Area(B)$$

$$P$$

$$P(\Delta) = \frac{P(X \le 0.5)}{P(\Delta)} = P(\Delta) = 0.25$$

$$P(\Delta) = \frac{Area(\Delta)}{T} = \frac{2Y \cdot 0.5^{2}}{P(\Delta)} = 0.25$$

$$P(A) = \frac{Area(A)}{TT} = \frac{10.5^{2}}{7^{10}} = 0.25$$

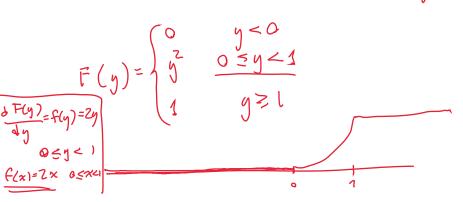
$$(0.5 < 10.7)$$

$$= P(10.7) - P(10.5)$$

$$= 0.7^{2} \times 10^{-10} = 0.49 - 0.75 = 0.24$$

Exercise 3: (continued)

Recall that FCy) = P(X ≤ y) For all y ∈ IR



Normal distribution

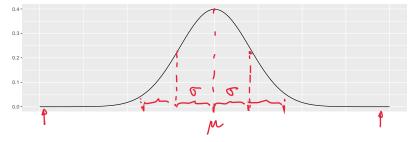
We say that X follows the Normal distribution if its PDF is given by

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

for $x \in \mathbb{R}$ where $\mu = EX$ and $\sigma^2 = VarX$. We then write

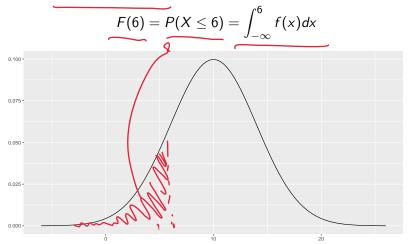
$$X \sim \text{Normal}(\mu, \sigma^2)$$

The graph of the PDF is:



CDF for the Normal

- ▶ Recall that the CDF and PDF are related by $F(y) = \int_{-\infty}^{y} f(x) dx$. This is just the area under the curve for $(-\infty, y]$.
- ▶ Let $X \sim \text{Normal}(10, 16)$, let's graph



Standard normal

▶ Let $X \sim \mathsf{Normal}(\mu, \sigma^2)$ and

$$Z = \frac{X - \mu}{\sigma}$$

then $Z \sim \text{Normal}(0, 1)$ is called a standard normal random variable.

Exercise: Use properties of the expected value and variance to show that EZ = 0 and VarZ = 1.

Sums and differences of Normals

If $X \sim \text{Normal}(\mu_X, \sigma_X^2)$ and $Y \sim \text{Normal}(\mu_Y, \sigma_Y^2)$ are two independent normal random variables and let S = X + Y and D = X - Y then

$$S \sim \text{Normal}(\mu_x + \mu_y, \sigma_x^2 + \sigma_y^2)$$

and

$$D \sim \text{Normal}(\mu_x - \mu_y, \sigma_x^2 + \sigma_y^2)$$