CMPE 320: Probability, Statistics, and Random **Processes**

Lecture 20: Derived distributions

Spring 2018

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Derived distributions

- Given a RV X and its PDF $f_X(x)$, derive the PDF $f_Y(y)$ of Y = g(X)
 - First calculate the CDF F_{Y} of Y

- Differentiate F_{Y} to obtain the PDF f_{Y}

$$f_{y}(y) = \frac{df_{y}(y)}{dy}$$

Example 4.1. Let
$$X$$
 be uniform on $[0,1]$, and let $Y = \sqrt{X}$.

Compute the PDF of Y .

Note $\{20\}$
 $F_{Y}(Y) = P(Y \subseteq Y) = P(\sqrt{X} \subseteq Y) = P(X \subseteq Y^{2})$
 $= \int_{-\infty}^{y^{2}} f_{X}(x) dx = \begin{cases} y^{2}, & 0 \le y^{2} \le 1 \Rightarrow 0 \le y \le 1 \end{cases}$
 $f_{Y}(Y) = \frac{d}{dy} = \begin{cases} 2y, & 0 \le y \le 1 \end{cases}$
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Therefore

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Example 4.2. John Slow is driving from Boston to the New York area, a distance of 180 miles at a constant speed, whose value is uniformly distributed between 30 and 60 miles per hour. What is the PDF of the duration of the trip?

X: John's speed. Uniform over [30,60] mph =)
$$f_X(x) = \begin{cases} \frac{1}{30} & 30 \le x \le 60 \\ 0 & 0 \end{cases}$$

Fy(9) = P(Y \leq 9) = P(\frac{180}{x} \leq 9) = P(\frac{180}{x} \leq 9) = P(\frac{180}{y} \leq 60 \rightarrow 2) \frac{180}{y} \leq

Linear case

$$F_{y}(y) = P(Y \leq y) = P(aX + b \leq y) = P(X \leq \frac{y - b}{a}), a > 0$$

$$P(X \geq \frac{y - b}{a}), a < 0$$

$$F_{x}(y) = \begin{cases} F_{x}(\frac{y - b}{a}), a < 0 \end{cases}$$

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Example 4.5. A Linear Function of a Normal Random Variable is Normal. Suppose that X is a normal random variable with mean μ and variance σ^2 , and let Y = aX + b, where a and b are scalars, with $a \neq 0$. What is the PDF of Y?

$$f_{x}(x) = \frac{1}{\sqrt{2\pi\sigma^{2}}} e^{-\frac{(x-\mu)^{2}}{2\sigma^{2}}} \qquad f_{y}(y) = \frac{1}{|a|} f_{x}(\frac{y-b}{a})$$

$$f_{y}(x) = \frac{1}{|a|} \frac{1}{\sqrt{2\pi\sigma^{2}}} e^{-\frac{(x-\mu)^{2}}{2\sigma^{2}}} = \frac{1}{\sqrt{2\pi\alpha^{2}\sigma^{2}}} \exp\left(-\frac{(y-(b+a\mu))^{2}}{2\alpha^{2}\sigma^{2}}\right)$$

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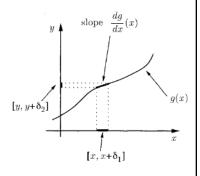
$$\left(\begin{array}{c} E[Y] = E[aX+b] = aE[X]+b = a\mu+b \\ Var(Y) = Var(aX+b) = a^2 ver(X) = a^2 \sigma^2 \end{array}\right)$$

Strictly monotonic case

• Y = g(X), but now g(x) is strictly monotonic

There is an inverse of g such that $J = g(x) \iff x = h(y)$ $= \begin{cases} P(X \le h(y)), & \text{increasing} \\ P(X \ge h(y)), & \text{decreasing} \end{cases} = \begin{cases} F_X(h(y)), & \text{increasing} \\ F_Y(y) = \frac{JF_Y(y)}{J} = \begin{cases} f_X(h(y)), & \text{decreasing} \\ -f_X(h(y)), & \text{decreasing} \end{cases} = \begin{cases} f_X(h(y)), & \text{decreasing} \\ -f_X(h(y)), & \text{decreasing} \end{cases}$

Alternative derivation



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Example 4.2. John Slow is driving from Boston to the New York area, a distance of 180 miles at a constant speed, whose value is uniformly distributed between 30 and 60 miles per hour. What is the PDF of the duration of the trip?

$$Y = \frac{160}{X} = g(x), \qquad h(y) = x = \frac{180}{y}$$

$$f_{Y}(y) = f_{X}(h(y)) \left| \frac{dh(y)}{dy} \right|$$

$$= \int_{30}^{1} \left| -\frac{160}{y^{2}} \right| = \frac{6}{3^{2}}, \quad 30 \le \frac{160}{5} \le 60$$

$$= \int_{0}^{1} \frac{dh(y)}{dy} = \int_{$$

Functions of 2 RVs

Example 4.8. Let X and Y be independent random variables that are uniformly distributed on the interval [0,1]. What is the PDF of the random variable Z=

$$F_{2}(z) = P(Z \le z) = P(\frac{1}{X} \le z) = P(\gamma \le zX)$$

$$= \int_{-2}^{2} \cdot 1 \cdot z = \int_{-2}^{2} \cdot 1 \cdot$$

$$= \begin{cases} \frac{1}{2} \cdot 1 \cdot \frac{1}{2}, & \frac{1}{2} \cdot 1 \\ 1 - \frac{1}{2} \cdot 1 \cdot \frac{1}{2}, & \frac{1}{2} \cdot 1 \end{cases}$$

$$f_{\frac{1}{2}}(z) = \begin{cases} \frac{1}{2} & 0 \le z \le 1 \\ \frac{1}{2}z^2, & z > 1 \end{cases}$$

UMBC CMPE 320 Example 4.9. Romeo and Juliet have a date at a given time, and each, independently, will be late by an amount of time that is exponentially distributed with parameter λ . What is the PDF of the difference between their times of arrival?

$$F_{2}(2) = P(Z \le 2) = P(X - Y \le 2) = P(Y \ge X - 2)$$

$$If -z < 0 \ (or \ z > 0), F_{2}(z) = I - P(Y < X - 2)$$

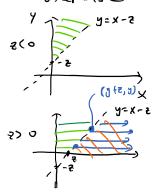
$$= I - \int_{0}^{\infty} \int_{y+2}^{y} (x,y) dx dy = I - \int_{0}^{\infty} \int_{y+2}^{y} e^{-\lambda(x+y)} dx dy$$

$$= I - \int_{0}^{\infty} \left[-\lambda e^{-\lambda(x+y)} \right]_{y+2}^{\infty} dy = I - \int_{0}^{\infty} \lambda e^{-\lambda(x+y)} dy$$

$$= 2 - \int_{0}^{\infty} \left[-\lambda e^{-\lambda(x+y)} \right]_{y+2}^{\infty} dy = I - \int_{0}^{\infty} \lambda e^{-\lambda(x+y)} dy$$

$$= 2 - \int_{0}^{\infty} \left[-\lambda e^{-\lambda(x+y)} \right]_{y+2}^{\infty} dy = I - \int_{0}^{\infty} \lambda e^{-\lambda(x+y)} dy$$

$$= 1 - \left[-\frac{1}{2} e^{-\lambda(2\pi + 2)} \right]_{0}^{2} = 1 - \left(\frac{1}{2} e^{-\lambda z} \right)$$



For
$$2 < 0$$
, can do similarly, or use the symmetry

" $2 = x - y$ and $-Z = y - x$ must have the same distribution

 $F_{2}(z) = P(Z \le z) = P(-Z \ge -z) = P(Z \ge -z) = 1 - F_{2}(-z)$
 $= 1 - (1 - \frac{1}{2}e^{+\lambda z}) = \frac{1}{2}e^{+\lambda z}$ for $z < 0$

$$\Rightarrow F_{2}(z) = \begin{cases} 1 - \frac{1}{2}e^{-\lambda z} & z > 0 \\ \frac{1}{2}e^{-\lambda z} & z < 0 \end{cases}$$

$$f_{2}(z) = \frac{dF_{2}(z)}{dz} = \begin{cases} +\frac{\lambda}{2}e^{-\lambda z} & z > 0 \\ \frac{\lambda}{2}e^{-\lambda z} & z < 0 \end{cases}$$

$$\Rightarrow \frac{\lambda}{2}e^{-\lambda z} = \frac{\lambda}{2}e^{-\lambda z}$$

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Sum of independent RVs: Convolution

X, Y: independent RVs What is PDF of
$$Z = X + Y$$
 lets compute $f_{x,2}(x,z)$ and marginalize $f_{x,2}(x,z) = f_{2|X}(z|x) f_{X}(x)$

To compute $f_{2|X}(z|x)$, consider $f_{2|X}(z|x)$

$$f_{2|X}(z|x) = P(Z \le z \mid X = x) = P(X + Y \le z \mid X = x)$$

$$= P(x + Y \le z \mid X = x) = P(Y \le z - x) = f_{Y}(z - x)$$

$$f_{2|X}(z|x) = \frac{\partial f_{2|X}}{\partial z} = f_{Y}(z - x) \Rightarrow f_{X,R}(x,z) = f_{X}(x)f_{Y}(z - x)$$

$$f_{2|X}(z|x) = \int_{0}^{\infty} f_{X,Z}(x,z) dx = \int_{0}^{\infty} f_{X}(x) f_{Y}(z - x) dx$$

Convolution of $f_{X,Z}(z) = f_{X,Z}(z) = f_{X,Z}(z) = f_{X}(z) = f_{X$

