#### MATH 320: Probability

# Lecture 13: Functions of Random Variables

Chapter 5: Distributions of Functions of Random Variables (5.1)

### Deductibles and caps: Expected value of a function of a random variable

Expected value of a loss or claim

- These examples are in an insurance applications, but are just expected value of a function of a random variable problems.
- Insurance loss.
  - Example: (a) The amount of a single loss X for an insurance policy is exponential, with density function

$$f(x) = 0.002e^{-0.002x}, \quad x \ge 0 \implies X \sim \text{Exp}(\lambda = 0.0002)$$

So the (base) expected value of a single loss is:  $E(X) = \frac{1}{\lambda} = 500$ 

- Insurance with a deductible.
  - Continuing example: (b) Suppose now the insurance policy has a deductible of \$100 for each loss. Find the expected value of a single claim.
    - \*\* Now loss amount ≠ claim amount



- STRATEGY: We need to write a new function g(X) that represents the new claim amount taking into account the deductible.



g(X) will be a piecewise function. So think about the values g(X) takes in cases based on the range of X.

*NOTE:* We are thinking about the values of the claim from the insurance company's perspective.

## claim amount

$$g_{1}(x) = \begin{cases} 0 & 0 \le x \le 100 \\ x - 100 & x > 100 \end{cases}$$

policyholder pays first 100

$$E[g,(x)] = \int_{0}^{\infty} g(x) f(x) dx$$

$$= \int_{0}^{\infty} o f(x) dx + \int_{100}^{\infty} (x-100)[a \circ 0 \circ e^{-0.002x}] dx$$

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- Insurance with a deductible and a cap.
  - Continuing example: (c) Now suppose the insurance policy has a deductible of \$100 per claim AND a restriction that the largest amount paid on any claim will be \$700.
  - Strategy: Use the same strategy as before for the first case, then just need to take into account the cap.

claim amount
$$\int \{\{1, 1, 1\}\} = \int_{0}^{\infty} g_{1}(x) \frac{f(x)}{f(x)} dx$$

$$\int (x) = \begin{cases}
0 & 0 \le x \le 100 \\
x - 100 & (00 \le x \le 800 \\
700 & x > 800
\end{cases}$$

$$\Rightarrow \text{ payments are capped at } 700$$

$$\Rightarrow \text{ loss > 800 recieves a payment of } 800-100 = 700$$

$$\Rightarrow \begin{cases}
167, 09 & + (43.33) = 8(9.01)
\end{cases}$$

- Another example: The amount of a single loss X for an insurance policy has the density function f(x) for  $x \ge 0$  with deductible of \$150 and cap of \$900.
  - (a) Find a function g(X) for the amount paid (claim amount) for a loss x.
  - (b) Write the integral to solve for the expected claim amount.

4) 
$$g(x) = \begin{cases} 0 & 0 < x \le 150 \\ x - 150 & 150 < x \le 1050 \end{cases}$$

6)  $E[g(x)] = \begin{cases} 150 & 0 < x \le 1050 \\ 0 & 0 < x \le 1050 \end{cases}$ 

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• In general, if  $\log x$  with deductible d and cap c, we have

$$d(x) = \begin{cases} c & x > q + c \\ x - q & q + x \neq q + c \end{cases}$$

# The distribution Y = g(X)

Transformations so far

- We have already seen simple methods for finding E[g(X)] and V[g(X)] for any type of variable.
- Example: The monthly maintenance cost for a machine  $X \sim$  Exponential ( $\lambda = 0.01$ ). Next year costs will be increased 5% due to inflation. Thus next year's monthly cost is Y = g(X) = 1.05X.

Find E(Y).

$$E(x): \frac{1}{1} = 100 \implies E(y) = E(1.05 \times): 1.05 E(x): 105$$

- Note we did not need to to know the distribution of Y for this calculation. However, the mean and variance alone are not sufficient to enable us to calculate probabilities for Y = g(X), we need the actual distribution function f(Y).
- Discrete example: Same X with a new (discrete) model and inflation costs Y=g(X)=1.05X:
  - (a) Find the distribution of Y = g(X).
  - (b) Find P(Y < 100).

x	f(x)	y = 1.05x	f(y)
0	0.28	0	0.78
50	0.43	52.5	0.43
100	0.20	105	0.20
150	0.09	157.5	0.09

$$P(Y \ 2 \ 100) = P(Y = 0) + P(Y = 52.5)$$

$$= 0.28 + 0.43$$

$$= 0.71$$

• For the original continuous model, it is not as simple to find the new distribution.

Continuous transformations example

- Continuing example: Using the original  $X \sim \text{Exponential}(\lambda = 0.01) \text{ model...}$
- Find P(Y < 100).

 $\not A$  GOAL: Get the probability statement to be with with respect to X.

 $\nearrow$  STRATEGY: "Indirectly" find the probability for Y based on the known cdf of X and using some simple algebra. Note that this is the same strategy we used to find lognormal probabilities based on the normal cdf.

• Find the cdf  $F_Y(y)$ .

STRATEGY: Use the same reasoning as above, just for a general y:  $P(Y \le 100) = F_Y(100) \longrightarrow P(Y \le y) = F_Y(y)$  for any value  $y \ge 0$ .

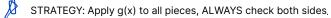
$$F_{y}(y) = P_{y}(yy)$$

$$= P_{x}(1.05 \times 5 y)$$

$$= P_{x}(x \le \frac{y}{1.05})$$

$$= F_{x}(\frac{y}{1.05}) = 1 - e$$
, 420

• Note that the range of X is the interval  $[0, \infty)$ . The range for Y = 1.05X is the same interval. This will not always be the case for transformations g(X).



ply g(x) to all pieces, ALWAYS check both sides.

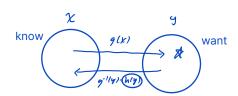
$$\Rightarrow \qquad \qquad \times : \qquad 0 \qquad ! \qquad \times \qquad ! \qquad \infty$$

$$\Rightarrow \qquad \qquad Y = g(x) = 1.05 \times : \qquad 1.05 \cdot (0) \cdot ! \qquad (.05 \times ) \cdot ! \qquad (.05 \cdot (\infty))$$

$$\qquad \qquad \qquad 0 \qquad ! \qquad \forall \qquad ! \qquad \forall \qquad \emptyset$$

Inverses

- Finding the distribution of Y = g(X) like we did above is much simpler when the transformation function g(X) has an inverse.
- Recall that the function g(X) defines a mapping from the original to a range of Y. That is,

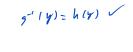


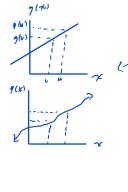
- \*\* We do not know stuff (pdf, cdf, etc.); so we have to use the inverse function to go backwards.  $\mathcal{Y}$  is completely determined by  $\mathcal{X}$ .
- When do inverse functions exist?

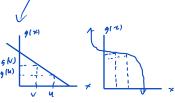
If the function g(x) is strictly **monotone**  $\implies$  one-to-one  $\iff$  inverse exists.

$$u>v\Rightarrow g(u)>g(v)$$
 strictly increasing  $u>v\Rightarrow g(u)< g(v)$  strictly decreasing











• Summary and results:

For a function g(x) that strictly increasing or strictly decreasing on the range of X, we can find an inverse function h(y) defined on the range of Y. Thus we have:

\*\* Strategy when problem solving:



1. Draw a figure of the transformation.

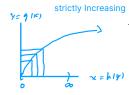
If transformation is strictly increasing or strictly decreasing over  $\mathcal{X}$ , then use the methods described next.



2. Check range of Y (i.e. ALSO transform range of X to range of Y).

Using  $F_X(x)$  to find  $F_Y(y)$  for Y = g(X)

- We will only generalize the methods for when g(X) has an inverse. If this is true, then there are two cases.
- Case 1: g(x) is strictly increasing on the range of X
  - Let h(y) be the inverse function of g(x). The function h(y) will also be strictly increasing. In this case, we can find  $F_Y(y)$  as follows:



- Example: Let  $X \sim$  Exponential ( $\lambda = 3$ ). Find the cdf of  $Y = \sqrt{X}$ .  $\longrightarrow$   $f_{\times}(x) = 1 - e^{-3x}$ ,  $\approx 70$ . There are two ways that we can solve this.

Short way

Long way

$$F_{\gamma}(y): \rho(y \in y)$$

$$= \rho(\int x \in y)$$

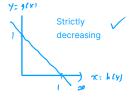
$$= \rho(x \in y)$$

$$= F_{\chi}(y^{2}) = 1 - e^{-3(y^{2})}$$

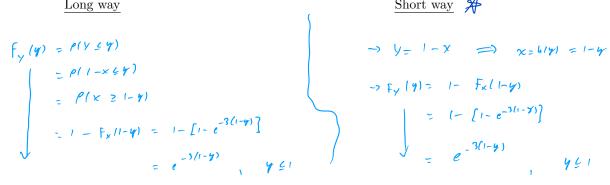
$$\Rightarrow y = \int x = \int$$

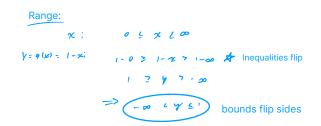


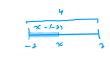
- Case 2: g(x) is strictly decreasing on the range of X
  - Let h(y) be the inverse function of g(x). The function h(y) will also be strictly decreasing. In this case, we can find  $F_Y(y)$  as follows:

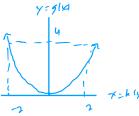


– Example: Let  $X \sim \text{Exponential} (\lambda = 3)$ . Find the cdf of Y = 1 - X. Again, we can do the long ("derivation") way or short way (skip to end result).









Not monotone (one-to-one)

no h(y) over 2

- If g(x) does NOT have an inverse
  - Example: Let  $X \sim \text{Uniform } (a = -2, b = 2)$ . Find the cdf of  $Y = X^2$ .

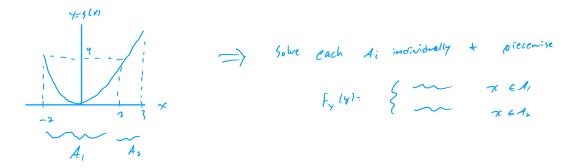
$$\Rightarrow f_{y}(y) : \rho(y \in y)$$

$$= \rho(x) = f_{x}(-\sqrt{y}) = \sqrt{y} + 2$$

$$= f_{x}(\sqrt{y}) - f_{x}(-\sqrt{y}) = \sqrt{y} + 2$$

$$= \sqrt{y}$$

- It can be even more complicate if there isn't a "balanced" range of Y. Example: Let  $X \sim U_n$  form (a:-3, b:3), find the cdf of  $Y: X^2$ .



- Both of these cases will be left for grad school:)

Finding the density function  $f_Y(y)$  for Y = g(X)

• Finding  $F_Y(y)$  gives us all the information that is needed to calculate probabilities for Y, as shown below:

$$P(Y \le y) = \lceil y \rceil^{y} \qquad \qquad P(Y \ge y) = \lceil - \lceil y \rceil^{y} \qquad \qquad P(a \le Y \le \mathbf{b}) = \lceil - \lceil y \rceil^{a} \rceil^{a}$$

Thus there is no real need to find the density function  $f_Y(y)$ . If the density function is required, it can be found by differentiating the cdf:

$$f_Y(y) = \frac{\mathrm{d}}{\mathrm{d}y} F_Y(y)$$

- If X is continuous, it is usually easier to find the cdf of Y and then the pdf of Y (rather than skipping straight to the pdf). But we will learn both methods, which we shall name:
  - 1. Cdf method  $\rho \delta f \times \longrightarrow c \delta f \times \longrightarrow$
  - 2. Pdf method (aka change of variable technique)
- Again when working in situations when g(x) has an inverse, there are two cases:
- Case 1: g(x) is strictly increasing on the range of X
  - Setup: h(y) is the inverse of g(x) and h(y) is strictly increasing.
  - Previous results:  $F_Y(y) = F_X(h(y))$
  - We can find the pdf  $f_Y(y)$  as follows:

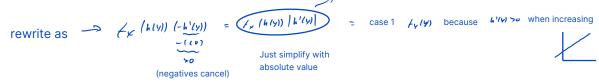
$$f_{y}(y) = \frac{d}{dy} f_{y}(y) = \frac{d}{dy} f_{x}(h(y)) = f_{x}'(h(y)) h'(y)$$

$$chain rule \qquad simplify$$

- Case 2: g(x) is strictly decreasing on the range of X
  - Setup: h(y) is the inverse of g(x) and h(y) is strictly decreasing.
  - Previous results:  $F_Y(y) = 1 F_X(h(y))$
  - We can find the pdf  $f_Y(y)$  as follows:

(Same steps)
$$\oint_{Y} (Y) = \frac{1}{dy} \int_{Y} (y) = \frac{1}{dy} \int_{Y} (y) = -f_{x}(h(y)) h'(y) = -f_{x}(h(y)) h'(y)$$

- Since h(y) is decreasing, its derivative is negative. Thus the final expression above is actually positive.



- Theorem: Let X have cdf  $F_X(x)$  with range  $\mathcal{X}, Y = g(X)$  with and range  $\mathcal{Y}$  and inverse h(y).
  - If g(x) is strictly increasing on  $\mathcal{X} \longrightarrow F_Y(y) = F_X(h(y))$  for  $y \in \mathcal{Y}$ .
  - If g(x) is strictly decreasing on  $\mathcal{X} \longrightarrow F_Y(y) = 1 F_X(h(y))$  for  $y \in \mathcal{Y}$ .
  - If g(x) is strictly increasing or strictly decreasing on  $\mathcal{X}$ , then

$$f_Y(y) = f_X(h(y)) \cdot |h'(y)|$$
 for  $y \in \mathcal{Y}$ 

- Return to previous examples: Let  $X \sim \text{Exponential}(\lambda = 3)$ .
  - (a) Find the pdf of  $Y = \sqrt{X}$ .

Cdf method

$$\frac{1}{2} f_{y}(y) = \frac{d}{dy} f_{y}(y) = \frac{d}{dy} \left[ 1 - e^{-3y^{2}} \right]$$

$$= -e^{3y^{2}} \left[ -5(1)y \right]$$

$$= 6ye^{-3y^{2}}, \quad y \ge 0$$

$$\frac{\text{Cut include}}{\text{on previously showed}} \quad F_{V}(Y) = 1 - e^{-3y^{2}}$$

$$\frac{1}{2} \quad F_{Y}(Y) = \frac{1}{2} \quad F_{Y$$

(ranges shown previously)

(b) Find the pdf of Y = 1 - X.

Cdf method

① previously showed 
$$f_{y}(y) = e^{-3(l-y)}$$

$$\frac{1}{2} f_{y}(y) = \frac{d}{dy} f_{y}(y) = \frac{d}{dy} \left[ e^{-3(1-y)} \right] \\
= e^{-3(1-y)} f_{y} \\
= 3 e^{-3(1-y)} f_{y}$$

Pdf method

$$\Rightarrow f_{y}(y) = (-y) \Rightarrow h'(y) = -1$$

$$\Rightarrow f_{y}(y) = f_{y}(1-y) |h'(y)|$$

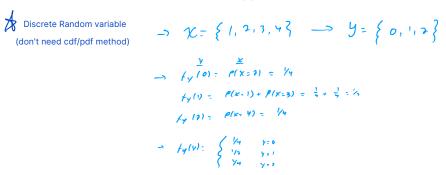
$$= 3e^{-3(1-y)} |-1|$$

$$= 3e^{-3(1-y)} |y = 1$$

(ranges shown previously)

#### More examples

1. Let X be the outcome when you roll a fair four sided die. If you get Y = |X - 2| dollars based on your roll, find  $f_Y(y)$ .



2. Let  $X \sim \text{Poisson}(\lambda = 4)$ . If  $Y = X^2$ , find the property of Y.