

Lecture 4.1

Random variables and their distribution: Transformations

Transformations of random variables

We often need to construct new random variables by transforming old ones, e.g.,

- Changing the units of measurement
- Output of deterministic algorithms with random inputs
- Estimators of parameters

We will only consider transformations of a single random variable.

Transformations of discrete random variables

Recall that X is a discrete random variable if it only takes on discrete values.

Let Y = g(X) where g is some arbitrary function. Obviously, Y is also a discrete rv that takes on values $\{y_1, y_2, \ldots\}$ such that $\{x_1, x_2, \ldots\} \stackrel{g}{\mapsto} \{y_1, y_2, \ldots\}$.

So, we can express the pmf of Y, i.e., f_Y , in terms of the pmf of X, i.e., f_X , as

$$f_Y(y_j) = \mathbb{P}(Y = y_j) = \mathbb{P}(g(X) = y_j)$$

$$= \sum_{x:g(x)=y_j} \mathbb{P}(X = x) = \sum_{x:g(x)=y_j} f_X(x).$$

Transformations of continuous random variables

Suppose X is a continuous random variable and let Y = g(X) where g is some arbitrary function.

Similar to the case of the discrete rv, the cdf of Y can be easily calculated as follows

$$F_Y(y) = \mathbb{P}(Y \leq y) = \mathbb{P}(g(X) \leq y) = \int_{\{x: g(x) \leq y\}} f_X(x) dx.$$

Transformations of continuous random variables

Even if g is continuous, Y = g(X) need not be a continuous random variable.

- Necessary Condition: If g(X) is a continuous random variable for any continuous random variable X, then necessarily g is not constant on any interval.
- Sufficient Condition: If g is differentiable and its derivative is zero at only finitely many points (e.g., g is linear or strictly monotone), then Y = g(X) will be a continuous rv.

If Y = g(X) is also a continuous random variable, then we can also calculate the pdf of Y as $f_Y(y) = F_Y'(y)$, at every point y at which F_Y is differentiable.

Transformations of continuous random variables

Suppose the continuous random variable X has pdf

$$f_X(x) = \frac{1}{2} \exp(-|x|), \quad x \in \mathbb{R}.$$

Define $Y = X^2$. What is the pdf of Y?

Answer: Note that Y can only take values in $[0, \infty)$. Also, $g(x) = x^2$ satisfy the sufficient condition on the previous slide. So, for any $y \ge 0$

$$F_{Y}(y) = \mathbb{P}(Y \le y) = \mathbb{P}(X^{2} \le y) = \mathbb{P}(-\sqrt{y} \le X \le \sqrt{y})$$

$$= \mathbb{P}(X \le \sqrt{y}) - \mathbb{P}(X \le -\sqrt{y}) = F_{X}(\sqrt{y}) - F_{X}(-\sqrt{y})$$

$$f_{Y}(y) = F'_{Y}(y) = \frac{1}{2\sqrt{y}} (f_{X}(\sqrt{y}) + f_{X}(-\sqrt{y}))$$

$$= \begin{cases} \frac{1}{2\sqrt{y}} \exp(-\sqrt{y}), & y \in [0, \infty) \\ 0, & \text{else.} \end{cases}$$

Linear transformations of continuous random variables

Suppose X is a continuous rv with cdf F_X and pdf f_X . Define Y=aX+b, where a>0 and $b\in\mathbb{R}$. Then, for any $y\in\mathbb{R}$,

$$F_Y(y) = \mathbb{P}(Y \le y) = \mathbb{P}(aX + b \le y)$$
$$= \mathbb{P}\left(X \le \frac{y - b}{a}\right) = F_X\left(\frac{y - b}{a}\right).$$

We can now find f_Y , the pdf of Y, by differentiating the cdf F_Y :

$$f_Y(y) = F'_Y(y) = \frac{1}{a} f_X\left(\frac{y-b}{a}\right).$$

Linear transformations of continuous random variables

Suppose X has pdf

$$f_X(x) = \frac{1}{2} \exp(-|x|), \quad x \in \mathbb{R}.$$

Define Y = 3X + 2. What is the pdf of Y?

Answer: For any $y \in \mathbb{R}$,

$$F_Y(y) = \mathbb{P}(Y \le y) = \mathbb{P}(3X + 2 \le y) = F_X\left(\frac{y-2}{3}\right)$$

Differentiating the cdf of Y, for all $y \in \mathbb{R}$,

$$f_Y(y) = F_Y'(y) = \frac{1}{3} f_X\left(\frac{y-2}{3}\right) = \frac{1}{6} \exp\left(-\left|\frac{y-2}{3}\right|\right).$$

Linear transformations of continuous random variables

In general, for a continuous random variable X with pdf f_X , the pdf of Y=aX+b, where $a\neq 0$ and $b\in \mathbb{R}$, is given by

$$f_Y(y) = \frac{1}{|a|} f_X\left(\frac{y-b}{a}\right).$$

Family of distributions can be constructed by a linear transformations. Such a family is sometimes called location—scale family.

Monotone transformations of continuous random variables

Suppose Y = g(X), where g is strictly increasing on the range of values that X takes. Then g^{-1} , the inverse of g, is also strictly increasing.

We can express the cdf of Y in terms of F_X using similar reasoning to what we used for the linear transformation.

If the pdf of X is non-zero on (a,b), then the pdf of Y is non-zero on (g(a),g(b)). For all $y\in (g(a),g(b))$,

$$F_Y(y) = \mathbb{P}(g(X) \le y) = \mathbb{P}(X \le g^{-1}(y)) = F_X(g^{-1}(y)).$$

We obtain the pdf of Y by differentiating its cdf (using chain rule).

Monotone transformations of continuous random variables

Suppose X has pdf

$$f_X(x) = \frac{1}{2} \exp(-|x|), \quad x \in \mathbb{R}.$$

Define $Y = e^X$. What is the pdf of Y?

Answer: Note that Y only takes values in $(0, \infty)$. For any y > 0,

$$F_Y(y) = \mathbb{P}(Y \le y) = \mathbb{P}(e^X \le y) = \mathbb{P}(X \le \ln y) = F_X(\ln y).$$

Differentiating the cdf of Y, for y > 0,

$$f_Y(y) = F'_Y(y) = \frac{1}{y} f_X(\ln y) = \frac{1}{2y} \exp(-|\ln y|)$$

and $f_Y(y) = 0$ for $y \le 0$.

Monotone transformations of continuous random variables

If g is strictly decreasing, then its inverse is also strictly decreasing.

If the pdf of X is non-zero on (a,b), then the pdf of Y is non-zero on (g(b),g(a)). Using the same reasoning as in the case of g being strictly increasing then shows that for all $y \in (g(b),g(a))$,

$$F_Y(y) = \mathbb{P}(Y \le y) = \mathbb{P}(X \ge g^{-1}(y)) = 1 - \mathbb{P}(X < g^{-1}(y))$$

= 1 - \mathbb{P}(X \le g^{-1}(y)) = 1 - F_X(g^{-1}(y)).

Again, we obtain the pdf of Y by differentiating its cdf.