# Operations on random variables and limit theorems

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#### Transformations of a random variable

Suppose that we are given a distribution on X, and a function g(x). What is the distribution of g(X)? If Y = g(X), then the density of Y is given by:

$$f_Y(y) = \frac{f_X(x)}{|g'(x)|} \tag{1}$$

## Transformations of a random variable

Example

Suppose X is standard normal, and  $Y = e^X$ . What is  $f_Y$ ?

$$f_Y(y) = \frac{1}{\sqrt{2\pi}} \frac{1}{\log y} e^{-\frac{1}{2}(\log y)^2}$$
 (2)

Suppose X and Y are discrete random variables. Can we characterise X + Y?

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If X and Y are continuous (and independent):

$$f_{X+Y}(z) = \int_{-\infty}^{+\infty} f_X(t) f_Y(z-t) dt$$
 (5)

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Example: Poisson

Suppose  $X \sim \mathsf{Poisson}(\lambda_1)$  and  $Y \sim \mathsf{Poisson}(\lambda_2)$ . Let us compute the distribution of X + Y. We will see  $X + Y \sim \mathsf{Poisson}(\lambda_1 + \lambda_2)$ .

Example: Uniform

Suppose  $X \sim \mathcal{U}([0,1])$  and  $Y \sim \mathcal{U}([0,1])$ Let us compute the distribution of X + Y.

## Algebraic properties of expectation

Compute expectation of complex quantities from those of simpler quantities.

## Expectation of a constant

$$\mathbb{E}\,1=1\tag{6}$$

related to the fact that a distribution is normalized.

## Expectations of sums

For any two random variables X and Y, we can derive the distribution of X + Y.

Can also compute  $\mathbb{E} X + Y$  directly:

$$\mathbb{E}[X+Y] = \mathbb{E}X + \mathbb{E}Y \tag{7}$$

X and Y need not be independent!

## Linearity of expectation

In general, if a is real constant, and X, Y are random variables, have:

$$\mathbb{E}[aX + Y] = a \mathbb{E} X + \mathbb{E} Y \tag{8}$$

## Linearity of expectation

Example: variance

Let us obtain an alternative formula for the variance.

$$\sigma^2 = \mathbb{E}(X - \mathbb{E}X)^{=} \mathbb{E}X^2 - (\mathbb{E}X)^2$$
 (9)

## Expectation of products

Let X and Y be independent random variables. We have that:

$$\mathbb{E} XY = \mathbb{E} X \mathbb{E} Y \tag{10}$$

## Expectation of products

Example: variance of sum

Let X and Y be random variables. Let us compute the variance of X+Y. Have:

$$\sigma_{X+Y}^2 = \sigma_X^2 + \sigma_Y^2 + 2\sigma_{XY} \tag{11}$$

#### Covariance

The covariance of X and Y is given by:

$$\sigma_{XY} = \mathbb{E}(X - \mathbb{E}X)(Y - \mathbb{E}Y) \tag{12}$$

If X and Y be independent random variables, then:

$$\sigma_{XY} = 0 \tag{13}$$

However, the converse does not hold.

#### Limit theorems

Want to understand the behaviour of averages:

$$\frac{1}{n} \sum_{i=1}^{n} X_i \tag{14}$$

where  $X_i$  is a sample of i.i.d. random variables.

## Law of large numbers

As  $n \to \infty$ , we expect to get the "right answer".

$$\lim_{n \to \infty} \frac{1}{n} \sum_{i=1}^{n} X_i = \mathbb{E} X \tag{15}$$

#### Central limit theorem

Quantify deviations of the sample mean.

$$\lim_{n \to \infty} \sqrt{n} \left( \frac{1}{n} \sum_{i=1}^{n} X_i - \mathbb{E} X \right) = \mathcal{N}(0, \sigma^2)$$
 (16)

where  $\sigma^2$  is the variance of X.

#### Central limit theorem

Example

Suppose that a store has 200 customers per day on average. We wish to compute the probability that there are more than 6150 customers in 30 days.

Let  $X_i$  be the number of customers on day i, suppose  $X_i \sim \text{Poisson}(200)$ .

Wish to compute:

$$P(\sum_{i} X_i \ge 6150) \tag{17}$$