

# Stat 230: Probability

## Lecture 23

Jeremy VanderDoes

University of Waterloo

Monday, June 27th

## Example

Suppose during a storm little Timmy sees lightning strike on average once every 2 minutes (with an exponential distribution). Naturally, he yells for his mommy. What is the probability of no lightning strikes for 3 minutes after she arrives? What if she already waited 2 minutes?

Last time we talked about:

(1) Exponential Distribution

For today:

(1) Normal Distribution

Reading: Chapter 8

# Review

- Today: Quiz 4
- Review timing

# Normal Distribution

## Definition

$X$  is said to have a **normal distribution** (or *Gaussian distribution*) with mean  $\mu$  and variance  $\sigma^2$  if the density of  $X$  is

$$f(x) = \frac{1}{\sqrt{2\pi\sigma^2}} e^{\frac{-(x-\mu)^2}{2\sigma^2}}, \quad x \in \mathbb{R}$$

Shorthand,  $X \sim N(\mu, \sigma^2)$

# Normal Distribution



Figure: Carl Friedrich Gauss, 1777-1855.

# Normal Distribution

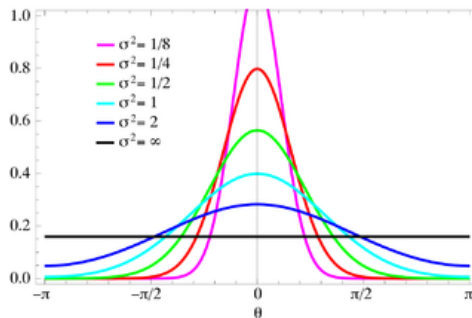


Figure: Normal densities with mean zero and increasing standard deviation.

# Normal Distribution

## Remark (Properties of the Normal Distribution)

(1) *Symmetric about its mean: If  $X \sim N(\mu, \sigma^2)$*

$$P(X \leq \mu + t) = P(X \geq \mu - t)$$

(2) *Density is unimodal: Peak is at  $\mu$ .*

(3) *Mean and Variance are the parameters:  $E(X) = \mu$ , and  $\text{Var}(X) = \sigma^2$ .*



# Normal Distribution

## Theorem

If  $X \sim N(\mu, \sigma^2)$ , then

$$E(X) = \mu, \quad \text{Var}(X) = \sigma^2$$

# Normal Distribution

Many probabilists and statisticians would argue (myself included) that the normal distribution is the most important/useful distribution.

- (1) The main reason for this is the Central Limit Theorem, which we will discuss by the end of this course.
- (2) As described by the Central Limit Theorem, the normal distribution models many different phenomena: Test scores on exams, measurements of heights and weights, error measurements in experiments, etc..

# Normal Distribution

Major problem with the normal random variable: If  $X \sim N(\mu, \sigma^2)$ , then

$$P(a \leq X \leq b) = \int_a^b \frac{1}{\sqrt{2\pi\sigma^2}} e^{-\frac{(x-\mu)^2}{2\sigma^2}} dx = ???$$

functions of the form  $e^{-x^2}$  do not have elementary antiderivatives... Drat!

This is a famous theorem of Liouville in the mid 1800's.

In practice these integrals must be evaluated numerically.

# Normal Distribution

## Definition

We say that  $X$  is a **standard normal** random variable if  $X \sim N(0, 1)$ .

Frequently in probability and statistics literature, the density of the standard normal random variable is denoted

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

and the CDF of a standard normal random variable is denoted

$$\Phi(x) = \int_{-\infty}^x \frac{1}{\sqrt{2\pi}} e^{-\frac{y^2}{2}} dy$$

# Normal Distribution

## Example

Suppose that  $Z \sim N(0, 1)$ . Compute

(1)  $P(Z \leq 2.15)$

(2)  $P(Z < -1.29)$

# Normal Distribution

## Theorem

If  $Y \sim N(\mu, \sigma^2)$ , then if

$$X = \frac{Y - \mu}{\sigma}, \quad X \sim N(0, 1)$$

The operation of centering a random variable and scaling it by the reciprocal of the standard deviation is called “standardization”.

# Normal Distribution

## Example

Suppose the score of a randomly selected student on Midterm 2 follows a normal distribution with mean 81 and variance  $6^2$ .

- (1) If a student is selected at random, compute the probability that the student's score exceeds 75.
- (2) If a student is selected at random, compute the probability that the student's score is less than 10.