

# Mathematical statistics

March 1<sup>st</sup>, 2018

Lecture 8: The distribution of the sample mean

# Chapter 6: Statistics and Sampling Distributions

6.1 Statistics and their distributions

6.2 The distribution of the sample mean

6.3 The distribution of a linear combination

Order 6.1  $\rightarrow$  6.3  $\rightarrow$  6.2

Next class (Monday, 03/04): QUIZ 1

Next Friday (03/08 ): Homework 2 due

# Questions for this chapter

Given a random sample  $X_1, X_2, \dots, X_n$ , and

$$T = a_1X_1 + a_2X_2 + \dots + a_nX_n$$

- If we **know** the distribution of  $X_i$ 's, can we obtain the distribution of  $T$ ?
  - Simple cases
  - If  $X_i$ 's follow normal distribution, then so does  $T$ .
- If we **don't know** the distribution of  $X_i$ 's, can we still obtain/approximate the distribution of  $T$ ?
  - Can we at least compute the mean and the variance?
  - When  $T$  is the sample mean, i.e.  $a_1 = a_2 = \dots = \frac{1}{n}$

## Theorem

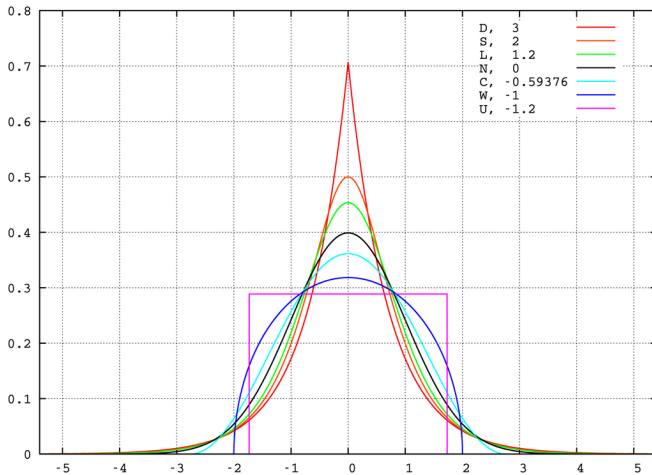
Let  $X_1, X_2, \dots, X_n$  be independent random variables (with possibly different means and/or variances). Define

$$T = a_1X_1 + a_2X_2 + \dots + a_nX_n,$$

then the mean and the standard deviation of  $T$  can be computed by

- $E(T) = a_1E(X_1) + a_2E(X_2) + \dots + a_nE(X_n)$
- $\sigma_T^2 = a_1^2\sigma_{X_1}^2 + a_2^2\sigma_{X_2}^2 + \dots + a_n^2\sigma_{X_n}^2$

# Bad news



In general, the mean and the variance do not define a probability distribution.

## Problem

Given a random sample  $X_1, X_2, \dots, X_n$  from a distribution with mean  $\mu$  and standard deviation  $\sigma$ , the mean is modeled by a random variable  $\bar{X}$ ,

$$\bar{X} = \frac{X_1 + X_2 + \dots + X_n}{n}$$

- Compute  $E(\bar{X})$
- Compute  $\text{Var}(\bar{X})$

# Mean and variance of the sample mean

Let  $X_1, X_2, \dots, X_n$  be a random sample from a distribution with mean value  $\mu$  and standard deviation  $\sigma$ . Then

1.  $E(\bar{X}) = \mu_{\bar{X}} = \mu$

2.  $V(\bar{X}) = \sigma_{\bar{X}}^2 = \sigma^2/n$  and  $\sigma_{\bar{X}} = \sigma/\sqrt{n}$

# The Central Limit Theorem

## Theorem

Let  $X_1, X_2, \dots, X_n$  be a random sample from a distribution with mean  $\mu$  and variance  $\sigma^2$ . Then, in the limit when  $n \rightarrow \infty$ , the standardized version of  $\bar{X}$  have the standard normal distribution

$$\lim_{n \rightarrow \infty} \mathbb{P} \left( \frac{\bar{X} - \mu}{\sigma/\sqrt{n}} \leq z \right) = \mathbb{P}[Z \leq z] = \Phi(z)$$

Rule of Thumb:

If  $n > 30$ , the Central Limit Theorem can be used for computation.



# Example

## Problem

*When a batch of a certain chemical product is prepared, the amount of a particular impurity in the batch is a random variable with mean value 4.0 g and standard deviation 1.5 g.*

*If 50 batches are independently prepared, what is the (approximate) probability that the sample average amount of impurity  $\bar{X}$  is between 3.5 and 3.8 g?*

Hint:

- First, compute  $\mu_{\bar{X}}$  and  $\sigma_{\bar{X}}$
- Note that

$$\frac{\bar{X} - \mu_{\bar{X}}}{\sigma_{\bar{X}}}$$

is (approximately) standard normal.

# Example

## Problem

*The tip percentage at a restaurant has a mean value of 18% and a standard deviation of 6%.*

*What is the approximate probability that the sample mean tip percentage for a random sample of 40 bills is between 16% and 19%?*

# Example

## Problem

*The weight of a certain crab is known to have an expected value of 1.4 kg and standard deviation of 0.4 kg. The distribution is unknown. Consider one net that contains 112 crabs.*

*What is the probability that the total weight of the crabs in the net is less than 150 kg?*