

IMD0033 - Probabilidade

Aula 16 - Média

Ivanovitch Silva
Maio, 2019



Agenda

- Média
- A média como um ponto de equilíbrio
- Definindo a média algebricamente
- Estimando a média da população
- Estimando a média da população a partir de amostras pequenas

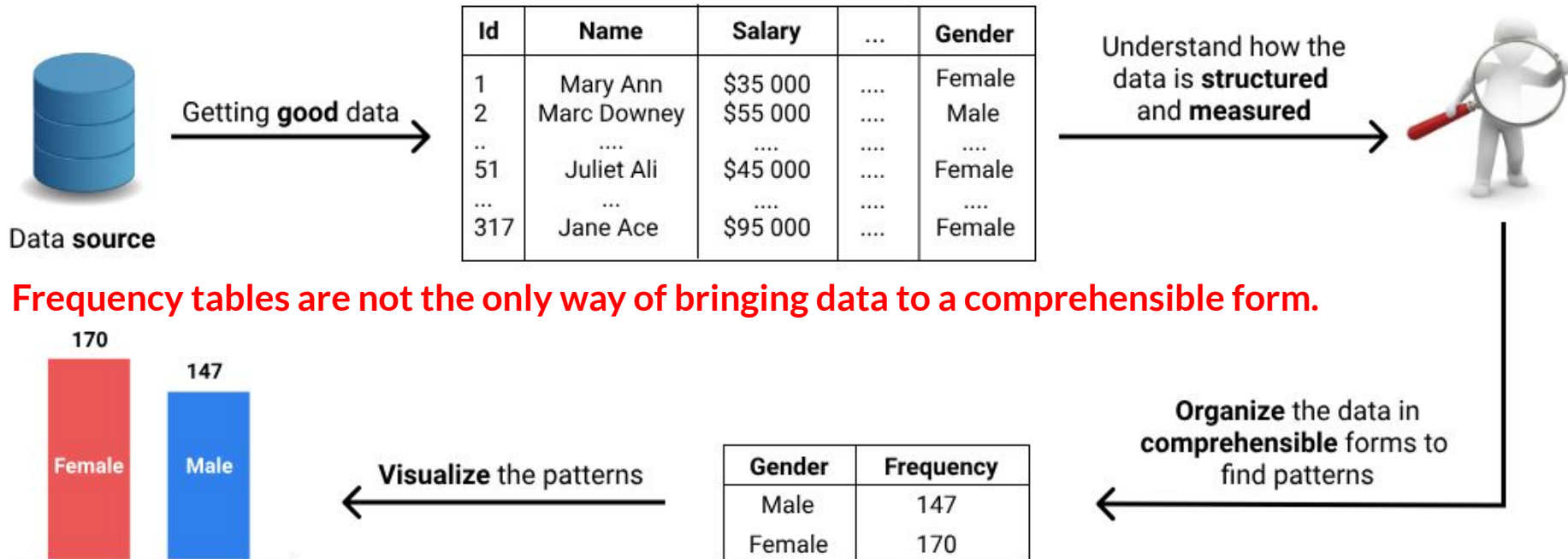
Atualizar o repositório

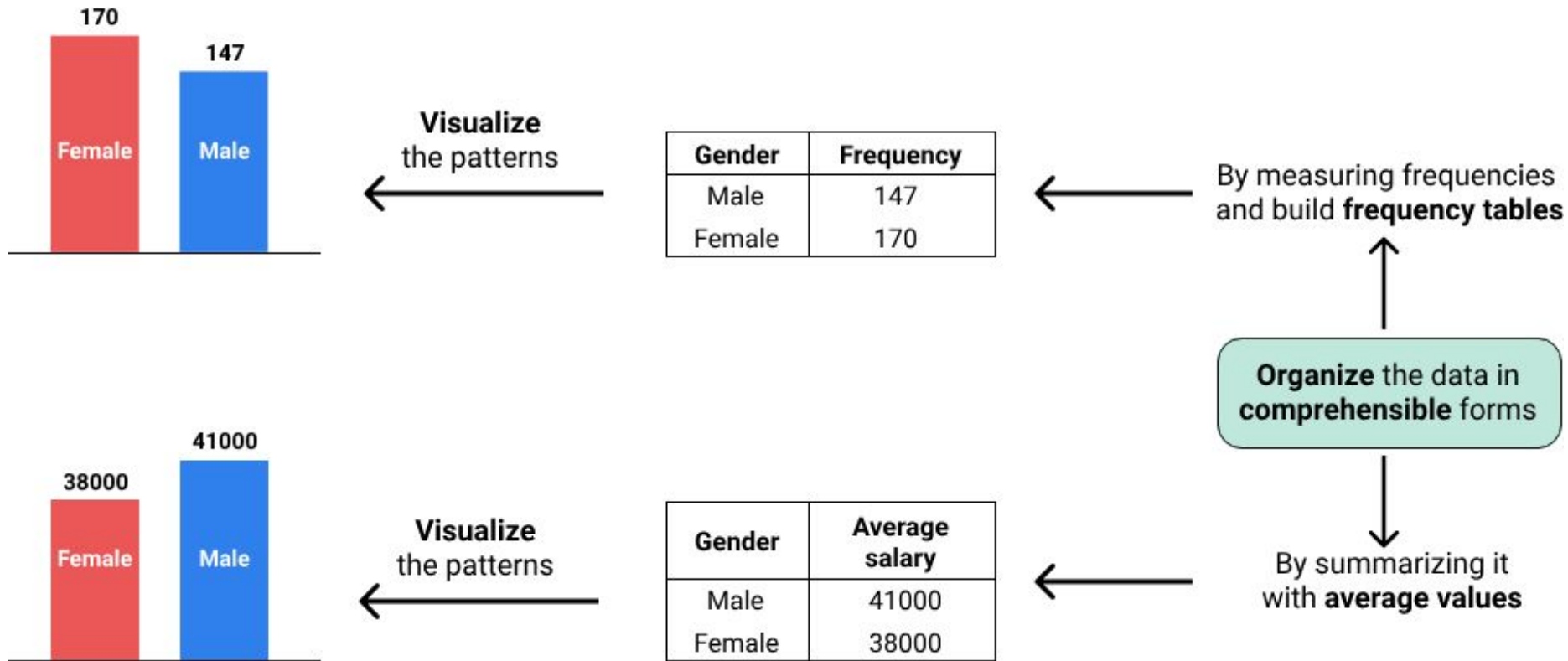
```
git clone https://github.com/ivanovitchm/imd0033_2019_1.git
```

Ou

```
git pull
```

PREVIOUSLY ON...





Depending on the particular characteristics of a distribution, we'll see that we can summarize it using the **mean**, the **weighted mean**, the **median**, or the **mode**.

We'll also learn to measure the **variability** in a distribution

$$A = [3, 3, 3, 3]$$

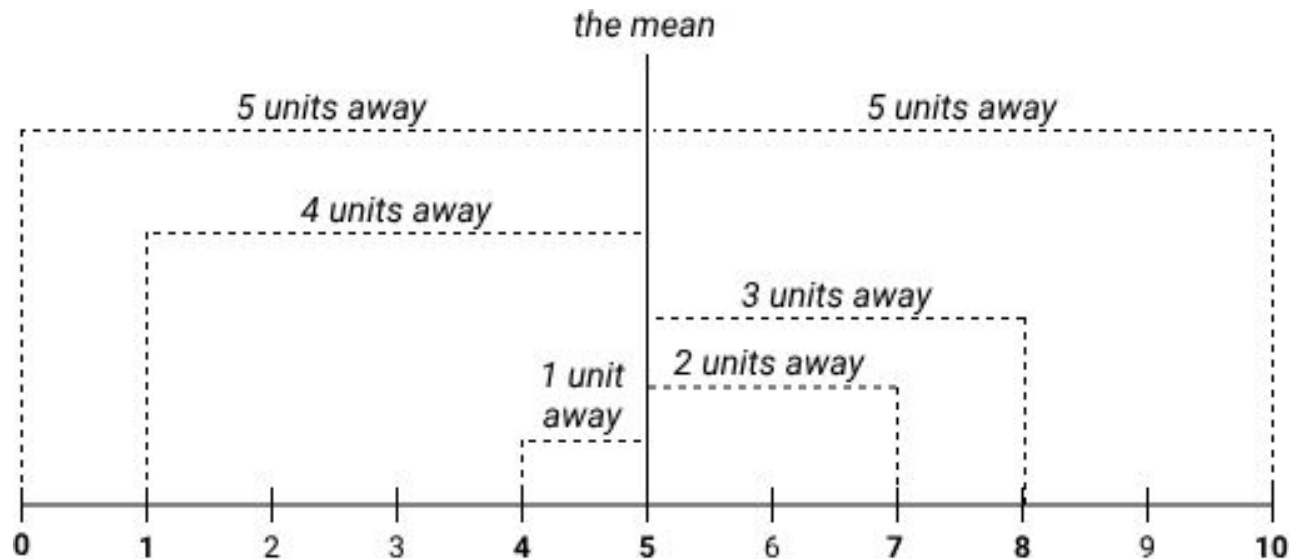
$$B = [30, 1, 15, 43]$$

We can clearly see that there's much more variability (diversity) in B. We'll learn to quantify variability using measures like **variance**, **standard deviation** and **z-scores**.

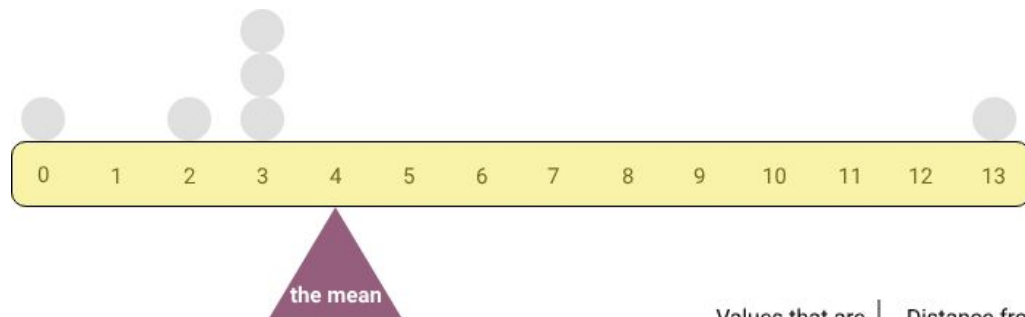
The Mean

[0, 1, 4, 7, 8, 10]

$$\frac{0 + 1 + 4 + 7 + 8 + 10}{6} = \frac{30}{6} = 5$$



The Mean as a Balance Point



[0,2,3,3,3,13]

Values that are below the mean	Distance from the mean
0	4 units
2	2 units
3	1 unit
3	1 unit
3	1 unit
<hr/>	
Total distance: 9 units	

Values that are above the mean	Distance from the mean
13	9 units
<hr/>	
Total distance: 9 units	


```
1 from numpy.random import randint, seed
2 equal_distances = 0
3
4 for i in range(1000):
5     seed(i)
6     # generate ten random numbers between 0 and 1000
7     distribution = randint(0,1000,10)
8     mean = sum(distribution) / len(distribution)
9
10    above = []
11    below = []
12    for value in distribution:
13        if value == mean:
14            continue # continue with the next iteration because the distance is 0
15        if value < mean:
16            below.append(mean - value)
17        if value > mean:
18            above.append(value - mean)
19
20    sum_above = round(sum(above),1)
21    sum_below = round(sum(below),1)
22    if (sum_above == sum_below):
23        equal_distances += 1
24
25 print(equal_distances)
```

Defining the mean algebraically

$$\mu = \frac{X_1 + X_2 + \dots + X_N}{N}$$

$N = 7$
 $\left[\begin{array}{ccccccc} 0 & 2 & 3 & 3 & 3 & 4 & 13 \\ \vdots & \vdots & \vdots & \vdots & \vdots & \vdots & \vdots \\ X_1 & X_2 & X_3 & X_4 & X_5 & X_6 & X_7 \end{array} \right]$

$$\mu = \frac{0 + 2 + 3 + 3 + 3 + 4 + 13}{7} = \frac{28}{7} = 4$$

Population Mean

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

$n = 3$
 $\left[\begin{array}{ccc} 2 & 3 & 4 \\ \vdots & \vdots & \vdots \\ X_1 & X_2 & X_3 \end{array} \right]$

$$\bar{X} = \frac{2 + 3 + 4}{3} = \frac{9}{3} = 3$$

Sample Mean

	Population	Sample
Mean	μ	$\bar{X}, \bar{X}_n, \bar{X}, M$
Number of values	N	n

Defining the mean algebraically

$$\mu = \frac{x_1 + x_2 + \dots + x_N}{N} = \frac{\sum X}{N} = \frac{\sum_{i=1}^N x_i}{N}$$

Population Mean

$$\bar{x} = \frac{x_1 + x_2 + \dots + x_n}{n} = \frac{\sum X}{n} = \frac{\sum_{i=1}^n x_i}{n}$$

Sample Mean



Ames, Iowa: Alternative to the Boston Housing Data as an End of Semester Regression Project

[Dean De Cock](#)

Truman State University

Journal of Statistics Education Volume 19, Number 3(2011),
www.amstat.org/publications/jse/v19n3/decock.pdf

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Key Words: Multiple Regression; Linear Models; Assessed Value; Group Project.

Abstract

This paper presents a data set describing the sale of individual residential property in Ames, Iowa from 2006 to 2010. The data set contains 2930 observations and a large number of explanatory variables (23 nominal, 23 ordinal, 14 discrete, and 20 continuous) involved in assessing home values. I will discuss my previous use of the Boston Housing Data Set and I will suggest methods for incorporating this new data set as a final project in an undergraduate regression course.



```

1 import pandas as pd
2 pd.set_option('display.max_columns', 500)
3
4 houses = pd.read_csv("AmesHousing_1.txt", sep='\t')
5 houses.shape

```

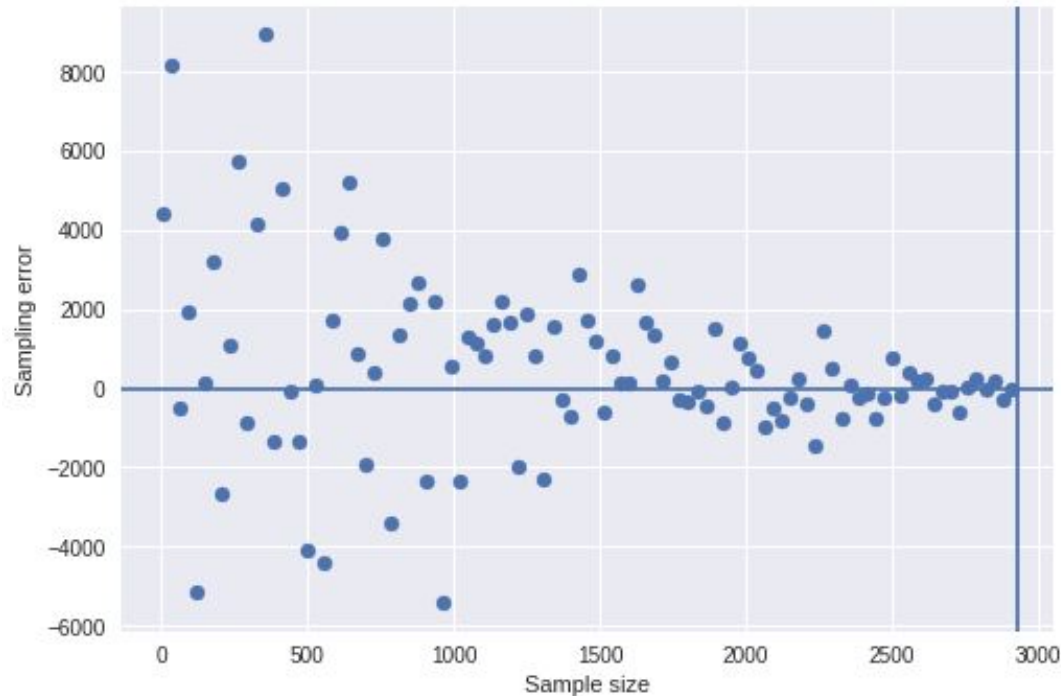
__	Order	PID	MS SubClass	MS Zoning	Lot Frontage	Lot Area	Street	Alley	Mo Sold	Yr Sold	Sale Type	Sale Condition	SalePrice
0	1	526301100	20	RL	141.0	131770	Pave	0	5	2010	WD	Normal	215000
1	2	526350040	20	RH	80.0	11622	Pave	0	6	2010	WD	Normal	105000
2	3	526351010	20	RL	81.0	14267	Pave	12500	6	2010	WD	Normal	172000
3	4	526353030	20	RL	93.0	11160	Pave	0	4	2010	WD	Normal	244000
4	5	527105010	60	RL	74.0	13830	Pave	0	3	2010	WD	Normal	189900

Estimating the population mean

In practice, **we almost always work with samples.**
But most of the times we're not interested in answering questions about samples — **we want to answer questions about populations.**

- What is the mean amount of money our customers spent last year on our website?
- What is the mean amount of time customers spent daily the first week after the promotion we ran?
- What is the mean sale price of a house in Ames, Iowa for the period 2006-2010?

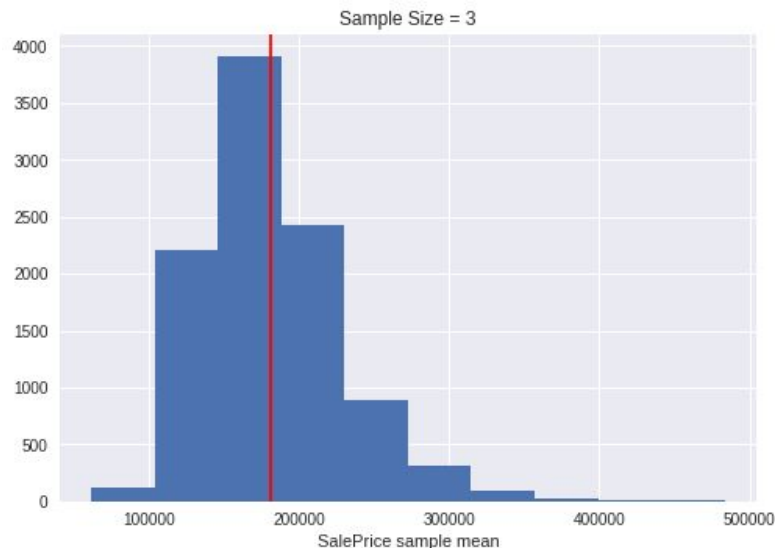
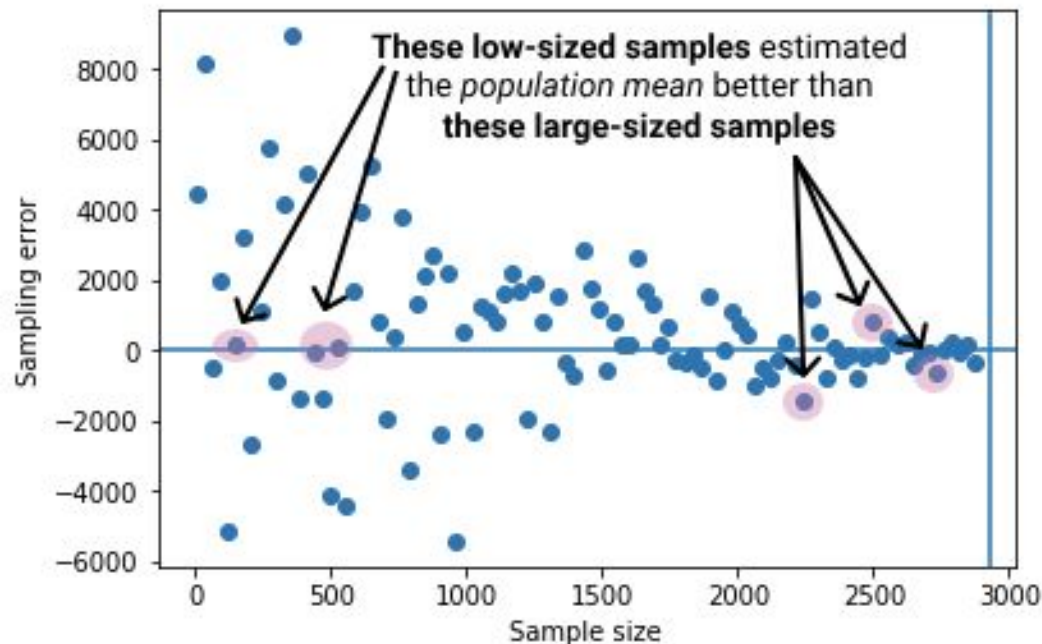
Estimating the population mean

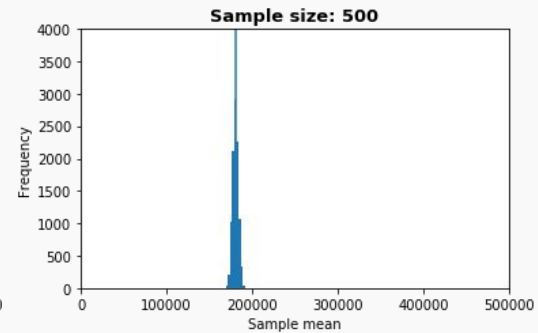
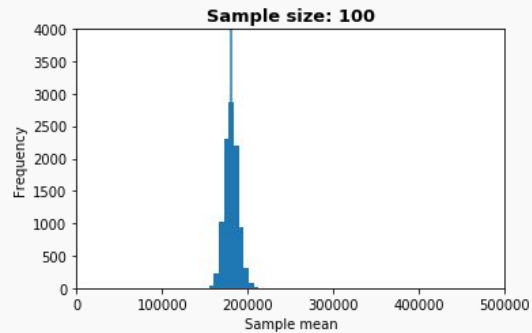
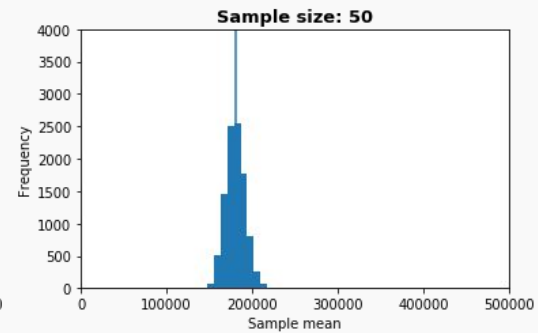
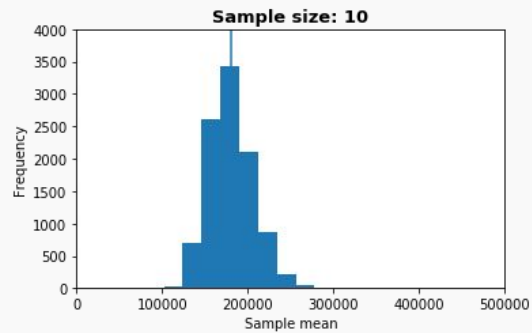
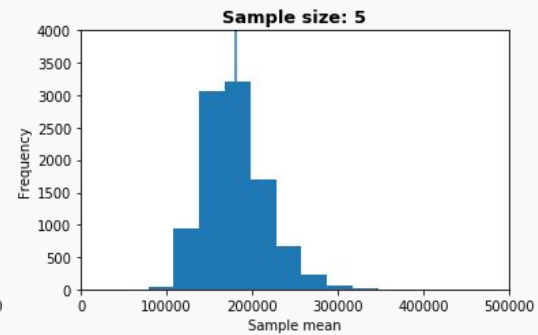
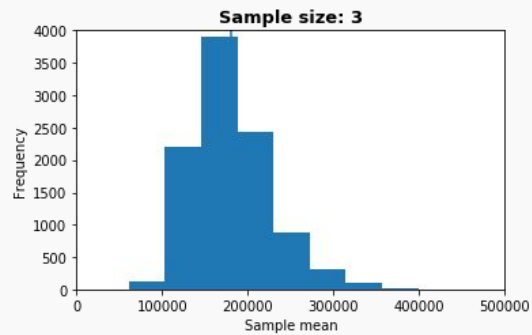


The general tendency for the sampling error is to decrease as the sample size increases.

$$\text{sampling error} = \mu - \bar{x}$$

Estimates from low-sized samples





Next Steps

In the next mission, we'll explore a few edge cases where it's either impossible to compute the mean, or it's possible but not theoretically sound.



```
index.js
import React, { useState } from 'react';
import './index.css';

function App() {
  const [contacts, setContacts] = useState([
    { name: 'John Doe', phone: '123-456-7890' },
    { name: 'Jane Smith', phone: '987-654-3210' }
  ]);

  const handleClick = () => {
    // TODO: Implement the logic to handle the click
  };

  return (
    <div>
      <h1>React App</h1>
      <button onClick={handleClick}>Click Me</button>
    </div>
  );
}

export default App;
```

```
index.html
<!DOCTYPE html>
<html>
  <head>
    <script src="index.js"></script>
  </head>
  <body>
    <div>
      <h1>React App</h1>
      <button>Click Me</button>
    </div>
  </body>
</html>
```