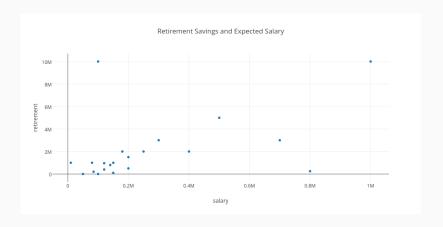
### Econ 103 – Statistics for Economists

Chapter 2 feat. Z Scores and OLS

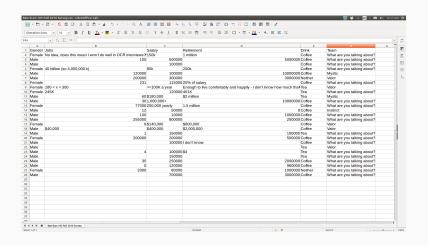
Mallick Hossain

University of Pennsylvania

## Survey Results



### **Problems with Surveys**

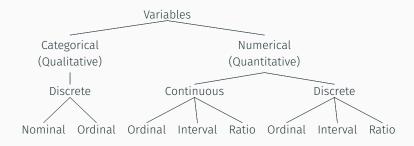


# Types of Variables

#### Discussion!

- · What are the differences between the following variables?
  - "Age" and "gender"
  - "Gender" and "class standing"
  - "SAT score" and "job creation"

### A Few Definitions: A Taxonomy of Variables



#### **Definitions**

- · Discrete: takes a countable number of values
- · Continuous: takes on any value
- Nominal: no order to the categories
- · Ordinal: categories with natural order
- · Interval: only differences meaningful, no natural zero
- · Ratio: differences and ratios meaningful, natural zero

# Summary Statistics

#### **Definitions**

- 1. Measures of Central Tendency
  - Mean: the average ("balance point")

$$\bar{x} = \frac{1}{n} \sum_{i=1}^{n} x_i$$

 Median: the middle observation (if data has even number of observations, take the mean of the middle two observations)

#### **Definitions**

#### 2. Percentiles

 $P^{th}$ Percentile = Value in  $(P/100) \cdot (n+1)^{th}$ Ordered Position

Note: this is not the only definition of percentiles, but it is the one we will use for this course!

### Percentile Example: n = 12

```
60 63 65 67 70 72 75 75 80 82 84 85

Q_1 = value in the 0.25(n+1)^{th} ordered position

= value in the 3.25^{th} ordered position

= 0.75 * 65 + 0.25 * 67

= 65.5
```

#### **Definitions**

- 3. Measures of Spread
  - · Variance: the spread from the mean

$$s^{2} = \frac{1}{n-1} \sum_{i=1}^{n} (x_{i} - \bar{x})^{2}$$

· Standard Deviation: another way to measure the spread

$$S = \sqrt{S^2}$$

· Range: the distance between the highest and lowest value

$$Range = |x_{max} - x_{min}|$$

• Interquartile Range (IQR): the distance between the upper and lower quartiles

$$IQR = |X_{75\%} - X_{25\%}|$$

### Why Is Variance Squared?

What happens if we do not square the deviations?

$$\frac{1}{n-1} \sum_{i=1}^{n} (x_i - \bar{x}) = \frac{1}{n-1} \left[ \sum_{i=1}^{n} x_i - \sum_{i=1}^{n} \bar{x} \right] = \frac{1}{n-1} \left[ \sum_{i=1}^{n} x_i - n\bar{x} \right]$$
$$= \frac{1}{n-1} \left[ \sum_{i=1}^{n} x_i - n \cdot \frac{1}{n} \sum_{i=1}^{n} x_i \right]$$
$$= \frac{1}{n-1} \left[ \sum_{i=1}^{n} x_i - \sum_{i=1}^{n} x_i \right] = 0$$

#### **Definitions**

- 4. Measure of Symmetry
  - Skewness: a measure of symmetry, positive values means the right tail is longer and vice versa

Skewness = 
$$\frac{1}{n} \frac{\sum_{i=1}^{n} (x_i - \bar{x})^3}{s^3}$$

### Skewness – A Measure of Symmetry

Skewness = 
$$\frac{1}{n} \frac{\sum_{i=1}^{n} (x_i - \bar{x})^3}{s^3}$$

#### What do the values indicate?

 $Zero \Rightarrow symmetry$ , positive right-skewed, negative left-skewed.

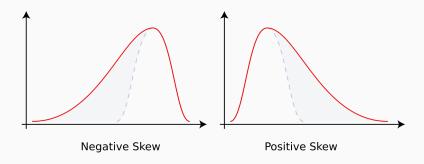
### Why cubed?

To get the desired sign.

### Why divide by s<sup>3</sup>?

So that skewness is unitless

### Skewness – A Measure of Symmetry



#### **Definitions**

- 5. Relationship between variables
  - · Covariance: how two variables vary together
  - Correlation: normalized version of covariance (ranges between -1 and +1)
  - · Regression: an estimation of how two variables are related

### Charts

#### Some Data Visualization Quotes

- 1. "Overload, clutter, and confusion are not attributes of information, they are failures of design" –Edward Tufte
- 2. "...few people will appreciate the music if I just show them the notes. Most of us need to listen to the music to understand how beautiful it is. But often, that's how we present statistics; we just show the notes we don't play the music." –Hans Rosling

## Simplicity



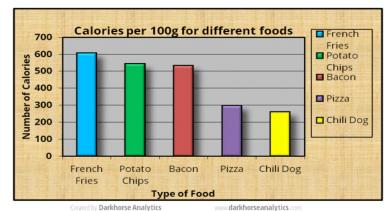
Simplicity is about subtracting the obvious and adding the meaningful

### An Illustration

https://i.imgur.com/W4BKCVU.gif

#### Before and After

### Before



,

#### Before and After

# After Calories per 100g French Potato Bacon Pizza Chili Dog Fries Chips

www.darkhorseanalytics.com

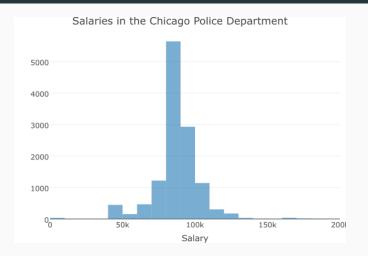
Created by Darkhorse Analytics

### Illustrating with Charts (Box and Whisker Chart)



What summary statistics can you infer from this chart?

### Illustrating with Charts (Histogram)

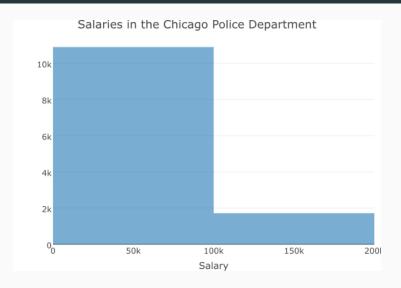


What summary statistics can you infer from this chart?

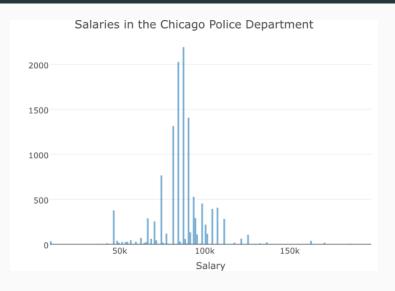
### Histograms are Really Important

- 1. Histograms show the frequency of different observations
- 2. Important Choice: How many bins?

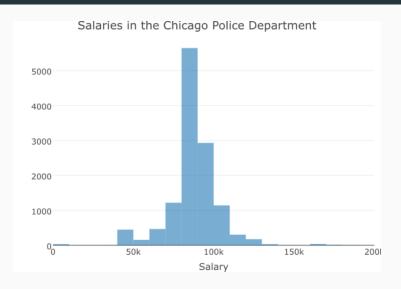
### Too Few Bins (Oversmoothing)



### Too Many Bins (Undersmoothing)



### Just Right! (Usually around 20 bins or so)



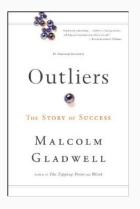
### Questions to Ask Yourself about Each Summary Statistic

- 1. What does it measure?
- 2. What are its units compared to those of the data?
- 3. How do its units change if those of the data change?
- 4. What are the benefits and drawbacks of this statistic?

Some of the information regarding items 2 and 3 is on the homework rather than in the slides because working it out for yourself is a good way to check your understanding.

### Outliers

### What is an Outlier?



**Outlier:** A very unusual observation relative to the other observations in the dataset (i.e. very small or very big).

### Which Summary Stats are Sensitive to Outliers?

- Assume our data is 1, 2, 3, 4, 5, 6, 7, 8, 9. What are our summary stats (mean, median, variance, range, IQR)?
- What will be affected if the data includes an outlier and becomes 1, 2, 3, 4, 5, 6, 7, 8, 63?

### Which Summary Stats are Sensitive to Outliers?

- · Mean changes from 5 to 11
- · Median remains at 5
- Variance changes from 7.5 to 385.5
- · Range changes from 8 to 62
- IQR remains at 5
- When Does the Median Change? IQR?
  - · Ranks would have to change.

### Summary of Sensitivity

#### Variance

Essentially the average squared distance from the mean. Sensitive to both skewness and outliers.

#### Standard Deviation

√Variance, but more convenient since same units as data

#### Range

Difference between larges and smallest observations. *Very* sensitive to outliers.

### Interquartile Range

Range of middle 50% of the data. Insensitive to outliers, skewness.

Sample vs. Population

## Essential Distinction: Sample vs. Population

For now, you can think of the population as a list of N objects:

Population:  $x_1, x_2, \ldots, x_N$ 

from which we draw a sample of size n < N objects:

Sample:  $x_1, x_2, \ldots, x_n$ 

#### **Important Point:**

Later in the course we'll be more formal by considering probability models that represent the *act of sampling* from a population rather than thinking of a population as a list of objects. Once we do this we will no longer use the notation *N* as the population will be *conceptually infinite*.

#### Essential Distinction: Parameter vs. Statistic

N individuals in the Population, n individuals in the Sample:

	Parameter (Population)	Statistic (Sample)
Mean	$\mu = \frac{1}{N} \sum_{i=1}^{N} x_i$	$\bar{x} = \frac{1}{n} \sum_{i=1}^{n} x_i$
Var.	$\sigma^2 = \frac{1}{N} \sum_{i=1}^{N} (x_i - \mu)^2$	$s^{2} = \frac{1}{n-1} \sum_{i=1}^{n} (x_{i} - \bar{x})^{2}$
S.D.	$\sigma = \sqrt{\sigma^2}$	$S = \sqrt{S^2}$

#### **Notation Matters!**

We use a sample  $x_1, ..., x_n$  to calculate statistics (e.g.  $\bar{x}$ ,  $s^2$ , s) that serve as estimates of the corresponding population parameters (e.g.  $\mu$ ,  $\sigma^2$ ,  $\sigma$ ).

## Why Do Sample Variance and Std. Dev. Divide by n-1?

$$\sigma^{2} = \frac{1}{N} \sum_{i=1}^{N} (x_{i} - \mu)^{2}$$

$$s^{2} = \frac{1}{n-1} \sum_{i=1}^{n} (x_{i} - \bar{x})^{2}$$

There is an important reason for this, but explaining it requires some concepts we haven't learned yet.

### Some Intuition

- Intuition 1: If we only had one data point, what would be the sample variance? Would it even be defined?
- Intuition 2: We know that the deviations from the sample mean sum to zero (see discussion of why variance is squared). Hence, we only need to know n-1 of the deviations since the last one will be whatever it takes to make the sum of them equal to 0. Hence, it would be proper to divide by n-1 instead of n

## **Z** Scores

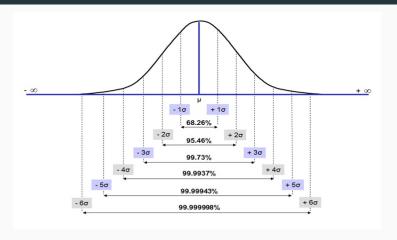
## Why Mean and Variance (and Std. Dev. )?

### **Empirical Rule**

For large populations that are approximately bell-shaped, standard deviation tells where most observations will be relative to the mean:

- $\cdot \approx$  68% of observations are in the interval  $\mu \pm \sigma$
- $\cdot pprox$  95% of observations are in the interval  $\mu \pm 2\sigma$
- Almost all of observations are in the interval  $\mu \pm 3\sigma$

## Standard Deviations



Physics uses a five-sigma rule (i.e. this could only happen normally 0.00057% of the time!)

## Z-scores: How many standard deviations from the mean?

$$z_i = \frac{x_i - \bar{x}}{s}$$

#### Unitless

Allows comparison of variables with different units.

#### **Detecting Outliers**

Measures how "extreme" one observation is relative to the others.

Linear Transformation of x

## What is the sample mean of the z-scores?

$$\bar{Z} = \frac{1}{n} \sum_{i=1}^{n} z_i = \frac{1}{n} \sum_{i=1}^{n} \frac{x_i - \bar{x}}{s} = \frac{1}{n \cdot s} \left[ \sum_{i=1}^{n} x_i - \sum_{i=1}^{n} \bar{x} \right]$$

$$= \frac{1}{n \cdot s} \left[ \sum_{i=1}^{n} x_i - n\bar{x} \right] = \frac{1}{n \cdot s} \left[ \sum_{i=1}^{n} x_i - n \cdot \frac{1}{n} \sum_{i=1}^{n} x_i \right]$$

$$= \frac{1}{n \cdot s} \left[ \sum_{i=1}^{n} x_i - \sum_{i=1}^{n} x_i \right] = 0$$

#### What is the variance of the z-scores?

$$S_z^2 = \frac{1}{n-1} \sum_{i=1}^n (z_i - \bar{z})^2 = \frac{1}{n-1} \sum_{i=1}^n z_i^2 = \frac{1}{n-1} \sum_{i=1}^n \left( \frac{x_i - \bar{x}}{s_x} \right)^2$$
$$= \frac{1}{s_x^2} \left[ \frac{1}{n-1} \sum_{i=1}^n (x_i - \bar{x})^2 \right] = \frac{s_x^2}{s_x^2} = 1$$

So what is the standard deviation of the z-scores?

## Population Z-scores and the Empirical Rule: $\mu \pm 2\sigma$

If we knew the population mean  $\mu$  and standard deviation  $\sigma$  we could create a *population version* of a z-score. This leads to an important way of rewriting the Empirical Rule:

Bell-shaped population  $\Rightarrow$  approx. 95% of observations  $x_i$  satisfy

$$\mu - 2\sigma \le x_i \le \mu + 2\sigma$$
$$-2\sigma \le x_i - \mu \le 2\sigma$$
$$-2 \le \frac{x_i - \mu}{\sigma} \le 2$$

Covariance and Correlation

## Covariance and Correlation: Linear Dependence Measures

#### Two Samples of Numeric Data

 $x_1, \ldots, x_n$  and  $y_1, \ldots, y_n$ 

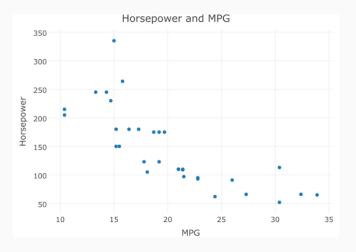
#### Dependence

Do x and y both tend to be large (or small) at the same time?

#### **Key Point**

Use the idea of centering and standardizing to decide what "big" or "small" means in this context.

## Are Engine Cylinders and Horsepower Related?



How do we quantify this relationship?

#### Covariance

$$s_{xy} = \frac{1}{n-1} \sum_{i=1}^{n} (x_i - \bar{x})(y_i - \bar{y})$$

- · Centers each observation around its mean and multiplies.
- · Zero ⇒ no linear dependence
- Positive ⇒ positive linear dependence
- · Negative ⇒ negative linear dependence
- Population parameter:  $\sigma_{xy}$
- · Units?
- Protip: If you know this, you know variance!

#### Correlation

$$r_{xy} = \frac{1}{n-1} \sum_{i=1}^{n} \left( \frac{x_i - \overline{x}}{s_x} \right) \left( \frac{y_i - \overline{y}}{s_y} \right) = \frac{s_{xy}}{s_x s_y}$$

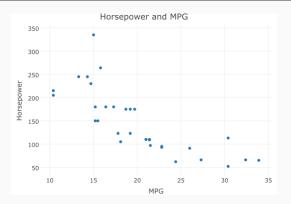
- · Centers and standardizes each observation
- Bounded between -1 and 1
- · Zero ⇒ no linear dependence
- Positive ⇒ positive linear dependence
- Negative ⇒ negative linear dependence
- Population parameter:  $ho_{xy}$
- Unitless

#### Game!

guessthecorrelation.com

Introduction to Regression

## Least Squares Regression – Predict Using a Line



- In order to fit a line through this, we need to estimate  $\hat{y}_i = a + bx_i$
- How do we find a and b?

## Finding a and b

Given our data (x<sub>i</sub>, y<sub>i</sub>), linear regression chooses the slope
 (b) and intercept (a) that minimize the sum of squared
 vertical deviations

$$\min \sum_{i=1}^{n} (y_i - \hat{y}_i)^2 = \min_{a,b} \sum_{i=1}^{n} (y_i - a - bx_i)^2$$

- · Why do we square the deviations?
- Note: Once we have our model, we can make predictions for any x and y!

## Solving for a

## Differentiate with respect to a

$$-2\sum_{i=1}^{n} (y_i - a - bx_i) = 0$$

$$\sum_{i=1}^{n} y_i - \sum_{i=1}^{n} a - b\sum_{i=1}^{n} x_i = 0$$

$$\frac{1}{n} \sum_{i=1}^{n} y_i - \frac{na}{n} - \frac{b}{n} \sum_{i=1}^{n} x_i = 0$$

$$\bar{y} - a - b\bar{x} = 0$$

$$\bar{y} - b\bar{x} = a$$

## Regression Line Goes Through the Means!

$$\bar{y} = a + b\bar{x}$$

## Solving for b

#### Substitute for a

$$\sum_{i=1}^{n} (y_i - a - bx_i)^2 = \sum_{i=1}^{n} (y_i - \bar{y} + b\bar{x} - bx_i)^2$$
$$= \sum_{i=1}^{n} [(y_i - \bar{y}) - b(x_i - \bar{x})]^2$$

#### FOC wrt b

$$-2\sum_{i=1}^{n} [(y_i - \bar{y}) - b(x_i - \bar{x})](x_i - \bar{x}) = 0$$

$$\sum_{i=1}^{n} (y_i - \bar{y})(x_i - \bar{x}) - b\sum_{i=1}^{n} (x_i - \bar{x})^2 = 0$$

$$b = \frac{\sum_{i=1}^{n} (y_i - \bar{y})(x_i - \bar{x})}{\sum_{i=1}^{n} (x_i - \bar{x})^2}$$

## Simple Linear Regression

#### Problem

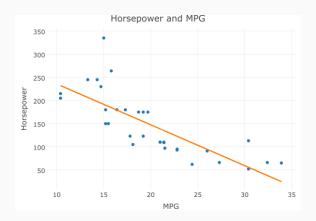
$$\min_{a,b} \sum_{i=1}^n (y_i - a - bx_i)^2$$

#### Solution

$$b = \frac{\sum_{i=1}^{n} (y_i - \bar{y}) (x_i - \bar{x})}{\sum_{i=1}^{n} (x_i - \bar{x})^2}$$

$$a = \bar{y} - b\bar{x}$$

## Fitting a Line



$$\widehat{hp} = 324.08 - 8.83mpg$$

## Relating Regression to Covariance and Correlation

$$b = \frac{\sum_{i=1}^{n} (y_i - \bar{y}) (x_i - \bar{x})}{\sum_{i=1}^{n} (x_i - \bar{x})^2} = \frac{\frac{1}{n-1} \sum_{i=1}^{n} (y_i - \bar{y}) (x_i - \bar{x})}{\frac{1}{n-1} \sum_{i=1}^{n} (x_i - \bar{x})^2} = \frac{s_{xy}}{s_x^2}$$
$$r = \frac{s_{xy}}{s_x s_y} = b \frac{s_x}{s_y}$$

## Comparing Regression, Correlation and Covariance

#### Units

Correlation is unitless, covariance and regression coefficients (a, b) are not. (What are the units of these?)

#### Symmetry

Correlation and covariance are symmetric, regression isn't. (Switching *x* and *y* axes changes the slope and intercept.)

## **Checking Our Results**

Model: hp = a + b \* mpg

Facts:

$$s_{mpg,hp} = -321$$
  $s_{mpg} = 6$   $s_{hp} = 69$   $m\bar{p}g = 20$ ,  $h\bar{p} = 147$ 

What is the sample correlation between MPG (x) and horsepower (y)?

$$r = \frac{s_{xy}}{s_x s_y} = \frac{-321}{6 \times 69} \approx -0.78$$

## **Checking our Results**

Model: 
$$hp = a + b * mpg$$

Facts:

$$s_{mpg,hp} = -321$$
  $s_{mpg} = 6$   $s_{hp} = 69$   $m\bar{p}g = 20$ ,  $h\bar{p} = 147$ 

What is the value of b for the regression?

$$b = \frac{s_{xy}}{s_x^2} = \frac{-321}{6^2} \approx -8.9$$

## Checking our Results

Model: 
$$hp = a + b * mpg$$

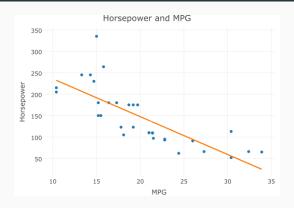
#### Facts:

$$s_{mpg,hp} = -321$$
  $s_{mpg} = 6$   $s_{hp} = 69$   $m\bar{p}g = 20$ ,  $h\bar{p} = 147$ 

What is the value of a for the regression (b = -8.9)?

$$a = \bar{y} - b\bar{x} = 147 - (-8.9) \times 20 \approx 325$$

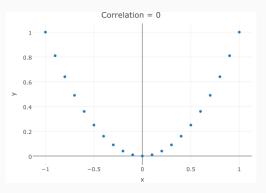
## **Checking our Results**



$$a \approx 325$$
  $b \approx -8.9$   $\widehat{hp} = 324.08 - 8.83 mpg$ 

#### **EXTREMELY IMPORTANT**

- Regression, Covariance, and Correlation: linear association.
- Linear association  $\neq$  causation.
- · Linear is not the only kind of association!



# Review

## Essential Distinction: Parameter vs. Statistic

*N* individuals in the Population, *n* individuals in the Sample:

	Parameter (Population)	Statistic (Sample)
Mean	$\mu_{X} = \frac{1}{N} \sum_{i=1}^{N} X_{i}$	$\bar{x} = \frac{1}{n} \sum_{i=1}^{n} x_i$
Var.	$\sigma_X^2 = \frac{1}{N} \sum_{i=1}^{N-1} (x_i - \mu)^2$ $\sigma_X = \sqrt{\sigma_X^2}$	$s_{x}^{2} = \frac{1}{n-1} \sum_{i=1}^{n} (x_{i} - \bar{x})^{2}$ $s_{x} = \sqrt{s^{2}}$
S.D.	$\sigma_{\rm X} = \sqrt{\sigma_{\rm X}^2}$	$S_X = \sqrt{S^2}$
Cov.	$\sigma_{xy} = \frac{\sum_{i=1}^{N} (x_i - \mu_x)(y_i - \mu_y)}{N}$ $\rho = \frac{\sigma_{xy}}{\sigma_x \sigma_y}$	$s_{xy} = \frac{\sum_{i=1}^{n} (x_i - \overline{x})(y_i - \overline{y})}{n-1}$ $r = \frac{s_{xy}}{s_x s_y}$

## **Related Reading**

- Wonnacott: Chapter 2, Section 4-5 A+B, Section 5-3, 11-1, 11-2, Appendices to 2-2, 2-5, 11-1, and 11-2
- · How to Lie with Statistics: Chapters 2, 5, and 6
- If you're rusty on permutations, combinations, etc., read the "Permutations and Combinations" document on the course page (mallickhossain.com/econ-103)

#### Homework

- · Chapter 2 Problems
- · Additional Problems
- R Tutorial 2