

## In-Class Exercise: Descriptive Statistics Review

### Part 1: Median, Average, and Outliers

Consider the following are the incomes of 22 households from two neighborhoods.

Neighborhood A	Neighborhood B
\$22,000	\$32,000
\$30,000	\$33,000
\$35,000	\$35,000
\$38,000	\$36,000
\$40,000	\$40,000
\$42,000	\$42,000
\$55,000	\$45,000
\$62,000	\$60,000
\$65,000	\$70,000
\$250,000	\$74,000
\$350,000	\$75,000

- a) What is the average income of each neighborhood? Which neighborhood is higher?

**A is higher: \$89,909 vs. \$49,273**

- b) What is the median income of each neighborhood?

**\$42,000 for both**

- c) Now remove the top two incomes from each group. Excluding those households, which neighborhood has the highest average income?

**Neighborhood B: \$43,667 vs. \$43,222**

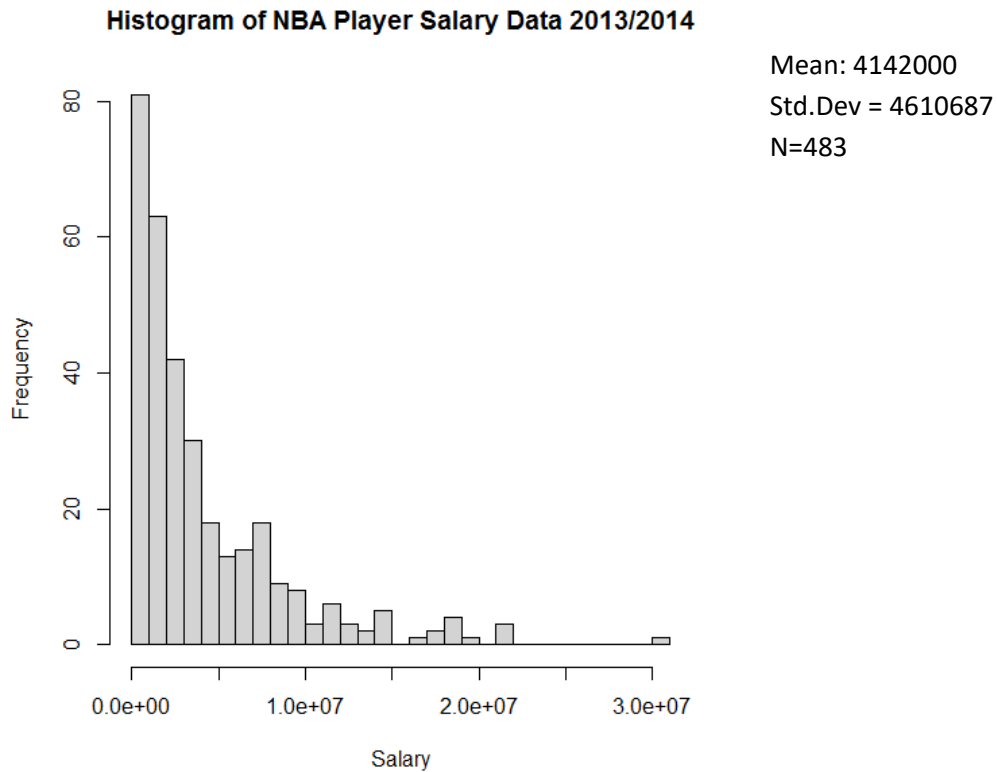
- d) If you include all the data, which measure (median or average) better explains the income distribution of the two neighborhoods? Why?

**Median is better because the two outliers in neighborhood A skew the results**

**Part 2: Interpreting a Histogram**

The following is a histogram of salaries for the 483 players in the NBA along with some summary statistics, all provided by R.

Source: *draftexpress.com*



a) What is the mean player salary?

**\$4,142,000**

b) Do most players make more or less than the mean? Explain.

**Most make less than the mean. You know this because the largest bars are to the left of the peak of the normal curve.**

c) Are player salaries normally distributed? Explain.

**No, they aren't. The histogram doesn't follow the shape of the normal curve.**

d) What do you learn about player salaries based on the standard deviation being greater than the mean?

**This means there probably are a lot of outliers (a lot of players making low salaries and some making really high salaries).**

### Part 3: Interpreting Statistical Tests

The following is the output of statistically testing the average NBA salaries for point guards versus shooting guards (*Source: draftexpress.com*):

Point Guards:	\$4,076,414.56	(n=110)
Shooting Guards:	\$4,158,783.82	(n=115)

*F-value = 0.017, p-value = 0.895*

From this, do you conclude that the two player groups have a statistically significant difference in their salaries? Why?

**No, because the p-value is much larger than 0.05 (or even 0.01).**

Now let's look at the output of statistically testing the difference in average salary of the 50 highest paid basketball players versus the 50 highest paid baseball players (*Source: draftexpress.com and newsday.com*):

Baseball:	\$18,538,001.90	(n=50)
Basketball:	\$15,458,543.62	(n=50)

*F-value = 17.509, p-value = 0.000*

From this, do you conclude that the two sports have a statistically significant difference in their salaries? Why?

**Yes, because the p-value is much smaller than 0.05 (or even 0.01).**

**Part 3: Probability**

Consider flipping a “fair” coin (50% chance of heads, 50% chance of tails).

- a) What’s the probability of getting “tails” two times in a row?

$$\mathbf{0.5 * 0.5 = 0.25 \text{ (or 25%)}}$$

- b) What’s the probability of getting “tails” three times in a row?

$$\mathbf{0.5 * 0.5 * 0.5 = 0.125 \text{ (or 12.5%)}}$$

Now imagine there is a bag with four red marbles and one green marble.

- c) What’s the chance of drawing a red marble?

$$\mathbf{4 \text{ out of } 5 \rightarrow 4/5 = 0.8 \text{ (or 80%)}}$$

- d) Let’s say you get that red marble. Now what’s the chance of drawing another red marble from the remaining marbles?

$$\mathbf{3 \text{ out of } 4 \rightarrow 3/4 = 0.75 \text{ (or 75%)}}$$