

# SOC245: Visualizing Data

## Precept 9: Characterizing Associations

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# Outline

- 1 Linear Regression using OLS
  - Reviewing OLS
  - Computing OLS
- 2 Probability
  - Probability review
- 3 Computing probability of events in R

# Outline

## 1 Linear Regression using OLS

- Reviewing OLS
- Computing OLS

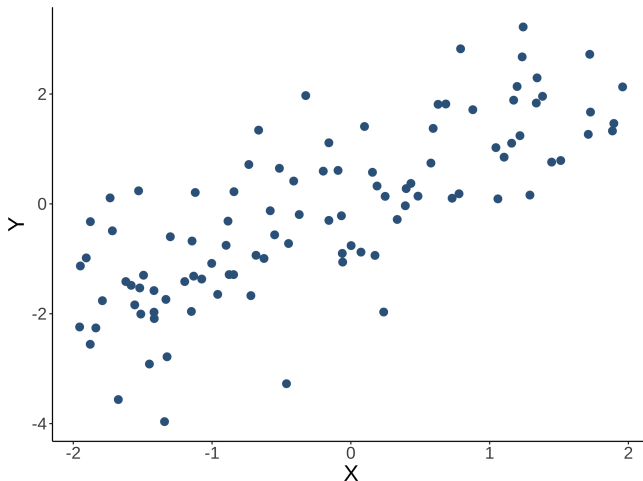
## 2 Probability

- Probability review

## 3 Computing probability of events in R

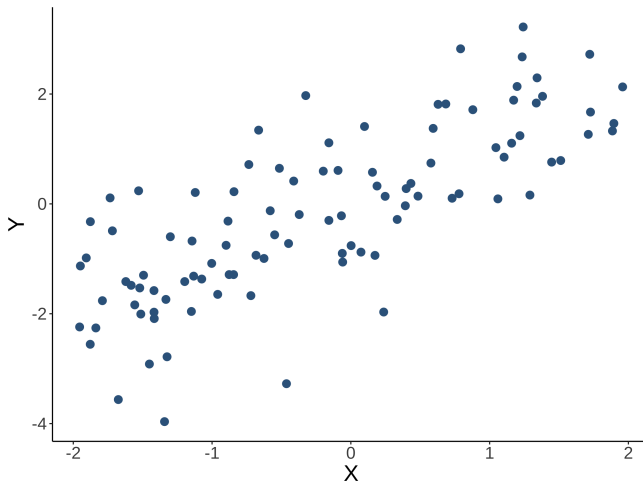
# Reviewing OLS

Let's say we have a data that has a linear association.



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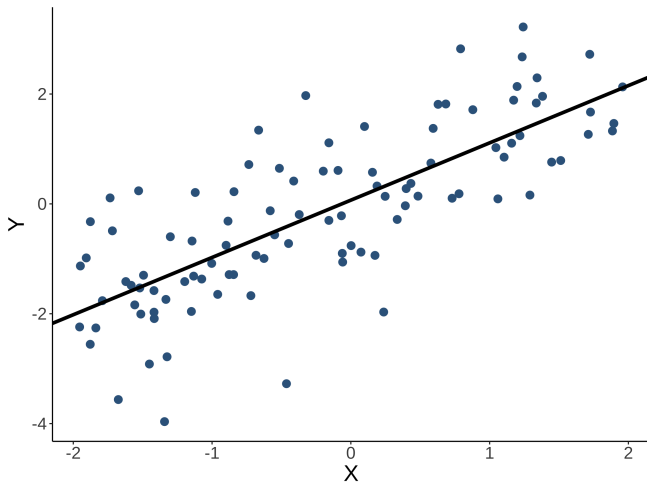
Let's say we have a data that has a linear association.



How do we find a line that best describes this data?

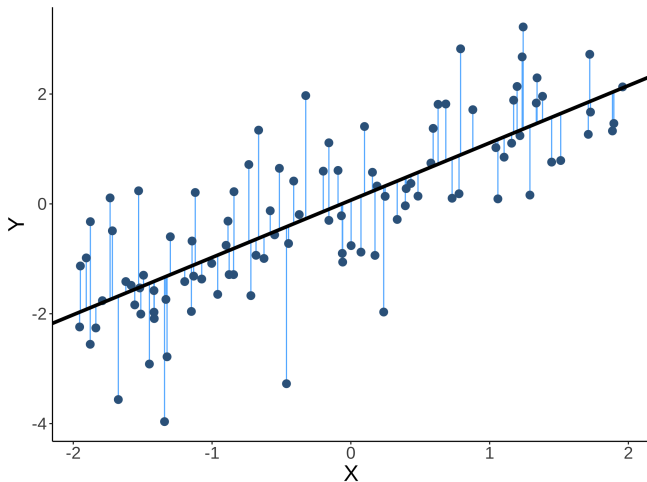
# Reviewing OLS

Want to minimize the **sum of squared residuals (SSR)**.



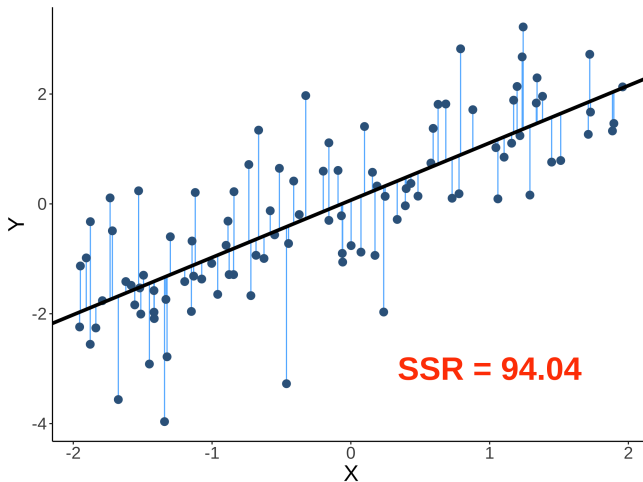
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# Reviewing OLS

- Our line is of the form  $\hat{Y} = a + bX$ .
- Values for  $a$  and  $b$  that minimize the SSR are:

$$b = \frac{\widehat{\text{Cov}}(X, Y)}{\hat{\sigma}_X^2}$$

$$a = \bar{Y} - \frac{\widehat{\text{Cov}}(X, Y)}{\hat{\sigma}_X^2} \bar{X}$$

Let's work with the `oppotunity2.csv` dataset, that contains colleges along with the median income of children who graduate from it and the median income of their parents.

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Load `tidyverse` and read in this dataset.

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- Note: `cov(X, Y)` finds the covariance between variables `X` and `Y`.

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- Note: `cov(X, Y)` finds the covariance between variables `X` and `Y`.
- What's the formula for variance?

## Computing OLS in R

```
b <- cov(col$par_median, col$k_median )/  
      (var(col$par_median))  
a <- mean(col$k_median) - b*mean(col$par_median)
```



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Let's plot this association along with the linear regression

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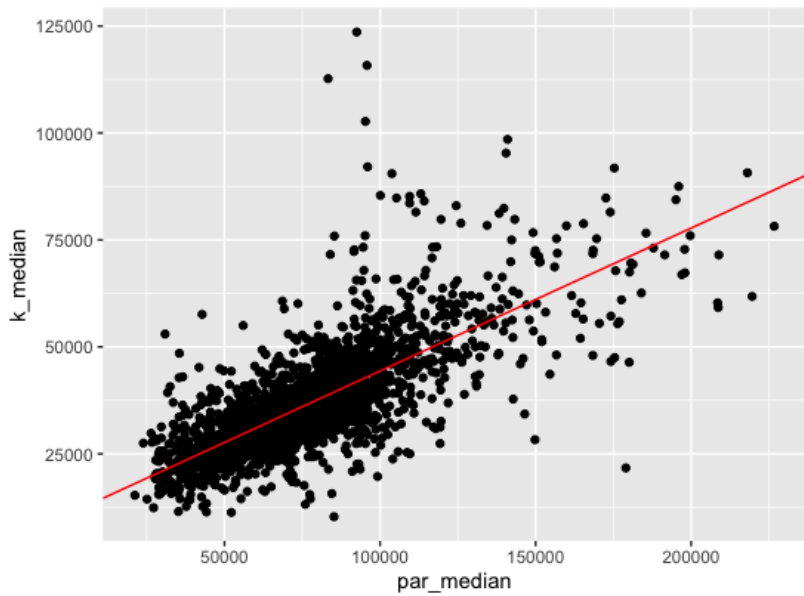
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```
ggplot(col) +  
  geom_point(mapping =aes(x=par_median, y=k_median)) +  
  geom_abline(intercept=a, slope=b, color="red")
```



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Look at `ols`. What does it contain?

```
> ols
```

```
Call:
```

```
lm(formula = k_median ~ par_median, data = opp)
```

```
Coefficients:
```

```
(Intercept)  par_median  
1.095e+04    3.341e-01
```

```
>
```



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We have what the regression model was trained on ...

```
> ols
```

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...and the coefficients.

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...and the coefficients. (Are these what you found before?)

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Residuals:

Min	1Q	Median	3Q	Max
-49051	-4816	-990	3948	81749

Coefficients:

	Estimate	Std. Error	t value	Pr(> t )
(Intercept)	1.095e+04	5.353e+02	20.45	<2e-16 ***
par_median	3.341e-01	6.469e-03	51.65	<2e-16 ***

---

Signif. codes: 0 '\*\*\*' 0.001 '\*\*' 0.01 '\*' 0.05 '.' 0.1 ' ' 1

Residual standard error: 8639 on 2200 degrees of freedom

Multiple R-squared: 0.548, Adjusted R-squared: 0.5478

F-statistic: 2667 on 1 and 2200 DF, p-value: < 2.2e-16

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We also have some statistics about the residuals.

## Accessing the residuals

You can access the residuals using `ols$residuals`.

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How would you compute the SSR of this line?



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We can access it using `summary(ols)$r.squared`.

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# Recall

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- Probabilities indicate our uncertainty about events.
- Intuitively, we measure how likely an event is to occur.

## Examples

Let's say I'm rolling a 6-sided die. What is the probability I get an even number?

## Examples

Suppose I have a bag containing 5 white marbles, 6 red marbles, and 4 black marbles, and I pick a marble without looking. What is the probability I pick a white marble?



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Read in the dataset `masc_raw-responses.csv`.  
This is a dataset containing responses from roughly 6000 men about what they think it means to be a man.

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This is a dataset containing responses from roughly 6000 men about what they think it means to be a man.

The full list of questions is in the `masculinity-survey.pdf`

We'll consider question 7(b) in particular to analyze today.

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This is the question: How often would you say you ask a friend for personal advice?

Answer options:

- 1 Often
- 2 Sometimes
- 3 Rarely
- 4 Never, but open to it
- 5 Never, and not open to it

What is the probability that the survey respondent said “Rarely”?

## Computing probabilities in R

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- We want to compute  $\Pr(R)$
- How can we do this?

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- What should our numerator be?
  - ▶ Number of respondents who replied “Rarely”

```
num <- nrow(filter(masc, q0007_0002 == "Rarely"))
```

You try!

What is the probability that the survey respondent said either “Never, but open to it” or “Never, and not open to it”?



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nrow(filter(masc,  
  q0007_0002 == "Never, but open to it" |  
  q0007_0002 == "Never, and not open to it")) /  
nrow(masc)
```

How can we find

$$\Pr(R | \text{Respondant is over 65})$$