

The background of the slide is a complex, abstract fractal pattern. It features swirling, organic shapes in vibrant blue and bright orange, separated by thin, dark lines. The overall effect is reminiscent of a microscopic view of a mineral or a digital fractal generation. A white, scalloped-edged frame is positioned on the left side, containing the title and course information.

Semipartial Correlations & Partial Correlations

PSY 612

Types of Correlations

- ♦ **Zero-order correlation:** the correlation between a predictor variable, X , and an outcome variable, Y , ignoring their relationship with any other variables
- ♦ **Semipartial correlation:** the correlation between an outcome variable, Y , with the part of a predictor variable, X_1 , that is unrelated to a second predictor variable, X_2
- ♦ **Partial correlation:** the correlation between the part of an outcome variable, Y , and the part of a predictor variable, X_1 , that are both unrelated to a second predictor variable, X_2

Example

A researcher is interested in the theory that people who plan their goals are more likely to achieve them. To test this theory, the researcher wants to examine the relationship between **goal achievement** (Y) and **planfulness** (X1).

However, **grit**, or one's perseverance and passion for pursuing long-term goals, has also been found to relate to people's goal achievement.

The current researcher wants to examine the **unique** relationship between **goal achievement** and **planfulness** when *controlling for grit* to demonstrate that planfulness has a relationship with goal achievement **over and above** the relationship that grit has with goal achievement.

Goal achievement Y	Planfulness X1	Grit X2

Example

Goal achievement was measured using the number of days during winter term that students went to the rec center among those who set a new years goal to exercise regularly.

Planfulness was measured on a scale from 1 (strongly disagree) to 5 (strongly agree) using items like, “Developing a clear plan when I have a goal is important to me.”

Grit was measured on a scale from 1 (not at all like me) to 5 (very much like me) using items like, “I finish whatever I begin.”

Goal achievement Y	Planfulness X1	Grit X2
48	5	4
21	4	2
15	3	3
6	1	1
10	2	2

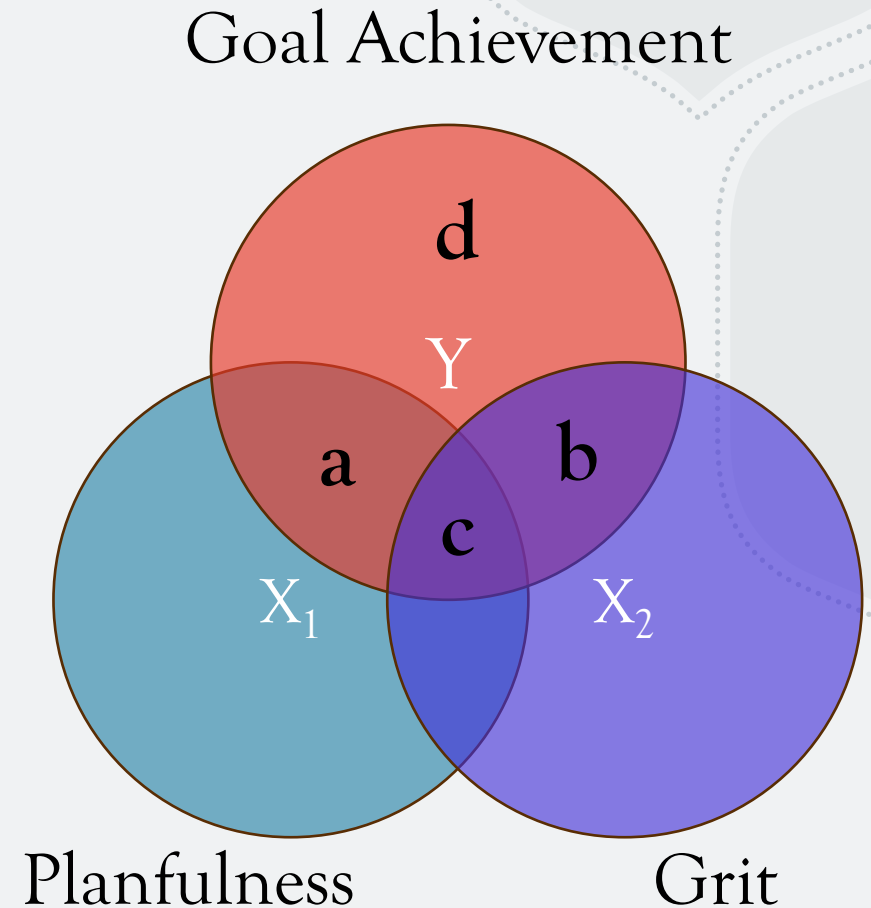
Example

The zero-order correlations between these three variables are shown in the correlation matrix below:

	Goal achievement	Planfulness	Grit
Goal achievement	1	-	-
Planfulness	0.9033	1	-
Grit	0.8571	0.8321	1

Explaining Variance in Y

- The venn diagram represents the variance in goal achievement (Y) that is related to planfulness (X1) and grit (X2)
- Notice that, because planfulness and grit are correlated with each other, there is **overlap** in the variance they explain in goal achievement
- Some of the variance explained by planfulness is **redundant** with the variance explained by grit in goal achievement scores



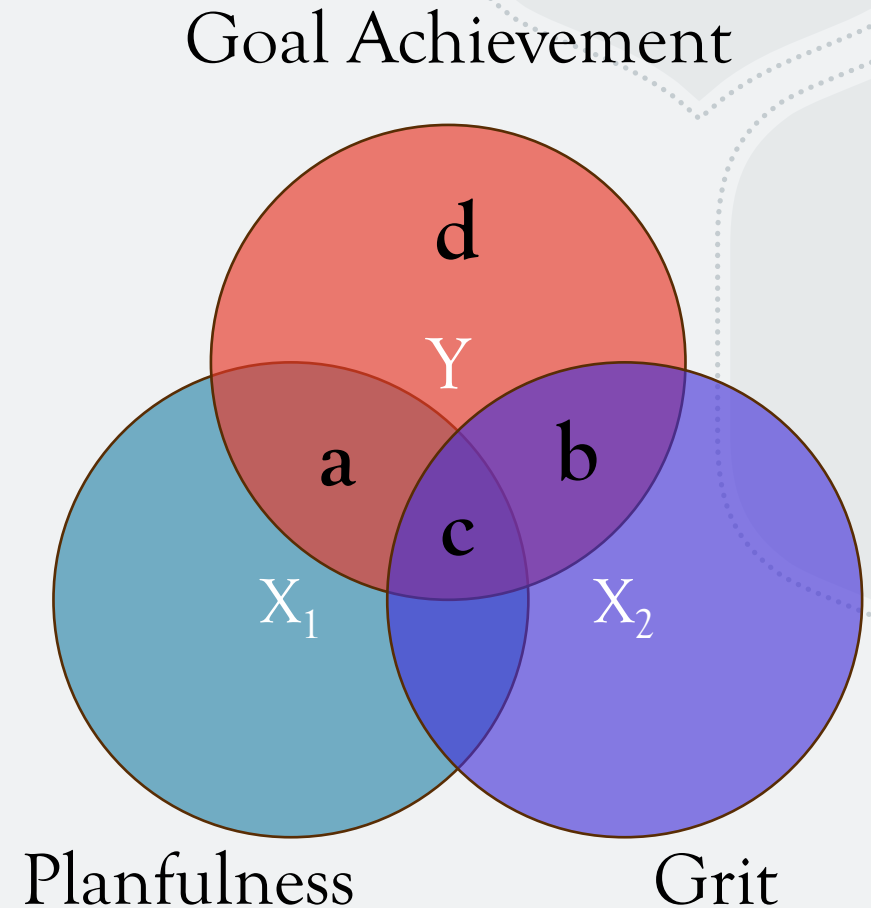
Explaining Variance in Y

a = the variance in goal achievement that is **uniquely** explained by its relationship with planfulness

b = the variance in goal achievement that is **uniquely** explained by its relationship with grit

c = the variance in goal achievement that is **redundantly explained by *both*** planfulness and grit

a + b + c + d = the total variance in goal achievement scores ($SS_{\text{Total}} = SS_Y$)



Semipartial Correlation

Semipartial correlation: the correlation between an outcome variable, Y, with the part of a predictor variable, X₁, that is unrelated to a second predictor variable, X₂

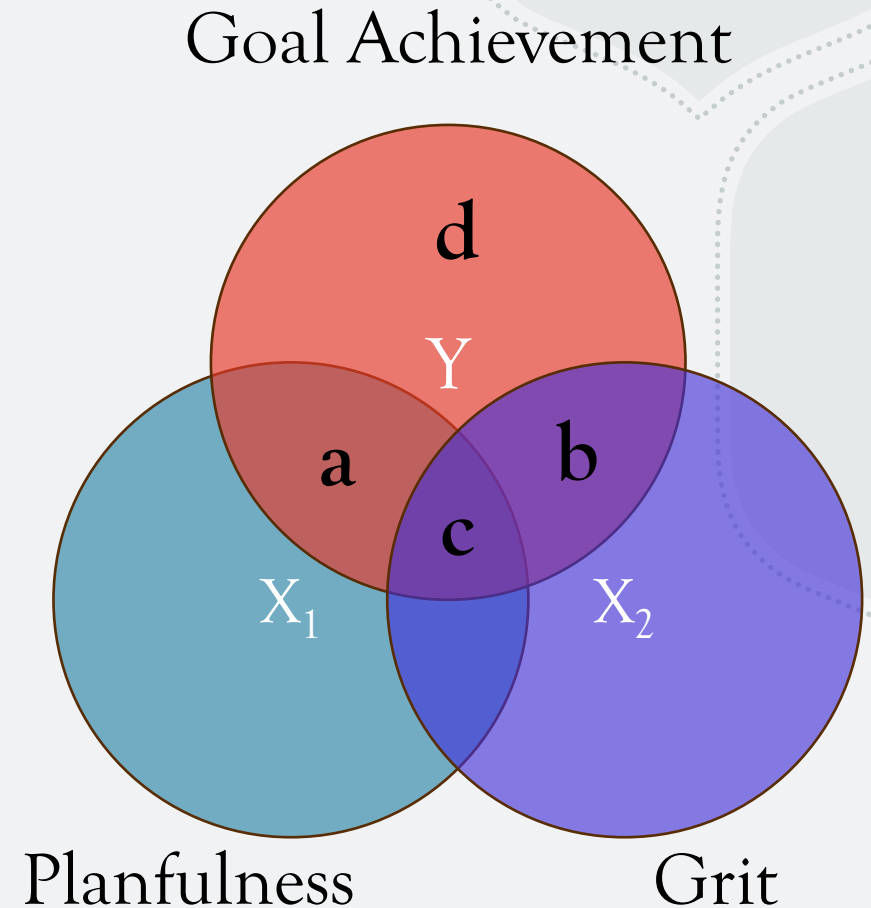
The formula for the semipartial correlation between Y and X₁ is:

$$sr_1 = \frac{r_{Y1} - r_{Y2}r_{12}}{\sqrt{1 - r_{12}^2}}$$

r_{Y1} = the zero-order correlation between Y and X₁

r_{Y2} = the zero-order correlation between Y and X₂

r_{12} = the zero-order correlation between X₁ and X₂

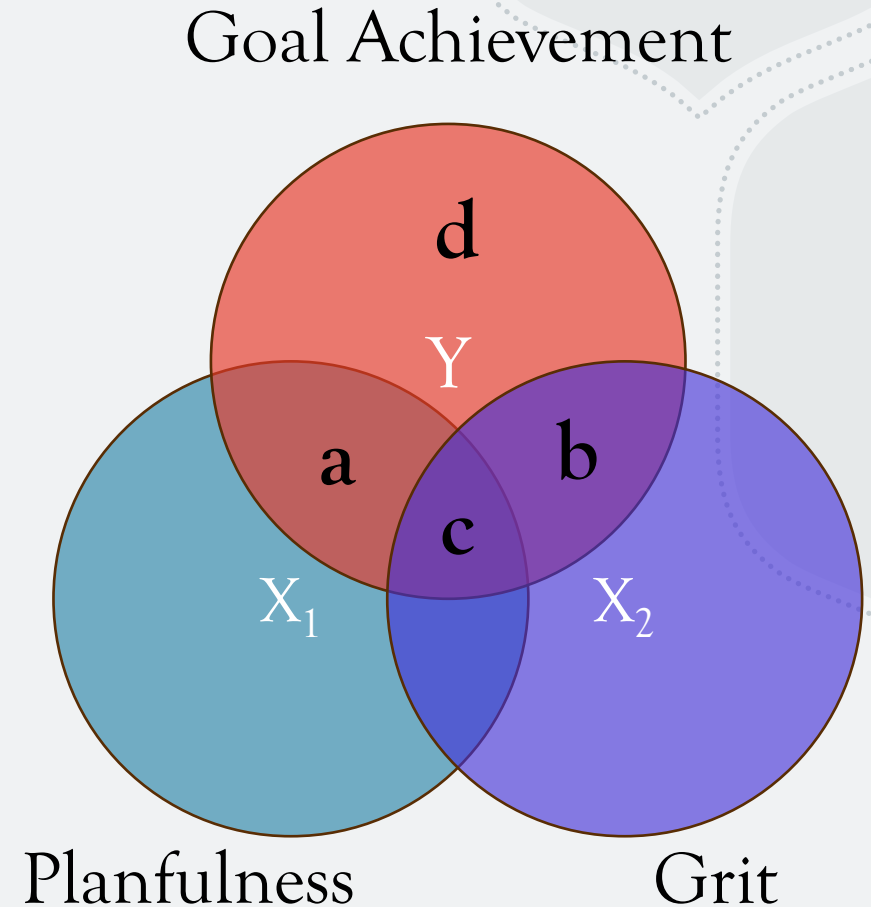


Semipartial Correlation

- When examining the unique relationship between planfulness and goal achievement using a **semipartial correlation**, the variance explained is equal to:

$$sr_1^2 = \frac{a}{a + b + c + d}$$

- The **semipartial correlation** measures the relationship between the part of X1 that is unrelated to X2 and all of Y



Semipartial Correlation

For our example (by hand):

	Goal Ach (Y)	Planfulness (X1)	Grit (X2)
Goal Ach (Y)	1	-	-
Planfulness (X1)	0.9033	1	-
Grit (X2)	0.8571	0.8321	1

$$sr_1 = \frac{0.9033 - (0.8571 * 0.8321)}{\sqrt{1 - 0.8321^2}}$$

$$sr_1 = \frac{0.1901}{\sqrt{0.3076}} = 0.34$$

$$sr_1^2 = (0.34^2) = .1156, \text{ or } \approx 12\%$$

For our example (in R):

```
spcor.test(data$goal_ach, # Y  
            data$planfulness, # X1  
            data$grit) # X2
```

```
estimate p.value statistic n gp Method  
1 0.3428421 0.6571579 0.5161332 5 1 pearson
```

```
> (0.3428)^2  
[1] 0.1175118
```

Planfulness explains approximately 12% of the variance in goal achievement over and above the variance explained by grit.

Partial Correlation

Partial correlation: the correlation between the part of an outcome variable, Y, and the part of a predictor variable, X₁, that are both unrelated to a second predictor variable, X₂

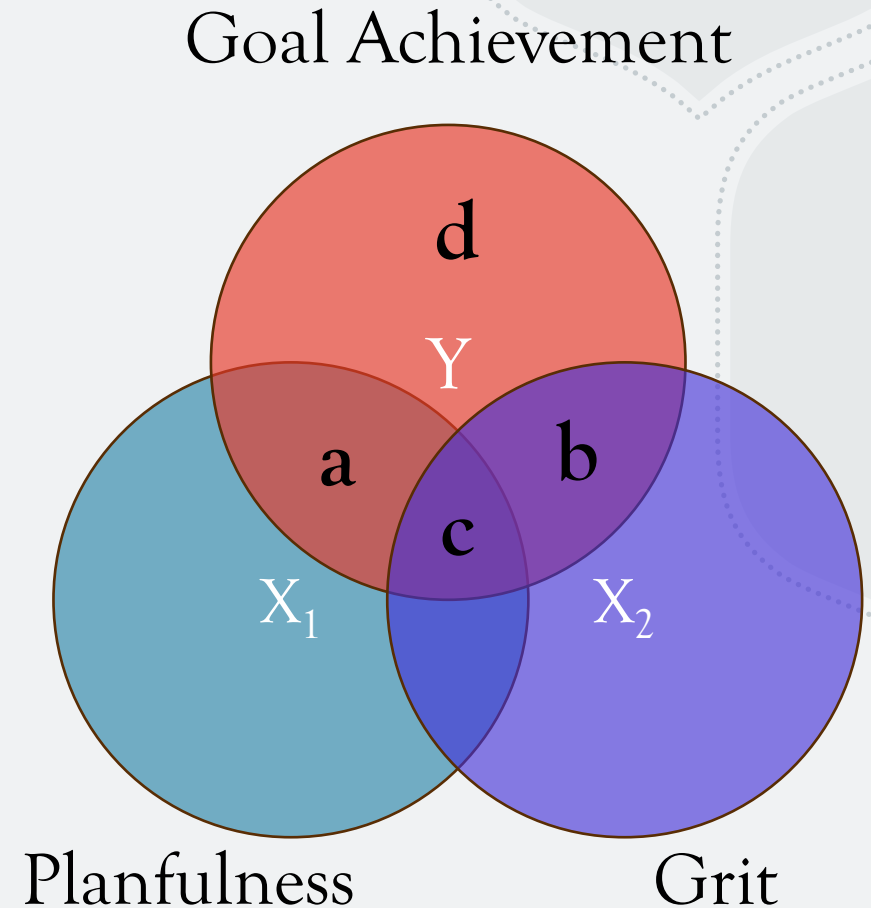
The formula for the partial correlation between Y and X₁ is:

$$pr_1 = \frac{r_{Y1} - r_{Y2}r_{12}}{\sqrt{1 - r_{Y2}^2}\sqrt{1 - r_{12}^2}}$$

r_{Y1} = the zero-order correlation between Y and X₁

r_{Y2} = the zero-order correlation between Y and X₂

r_{12} = the zero-order correlation between X₁ and X₂

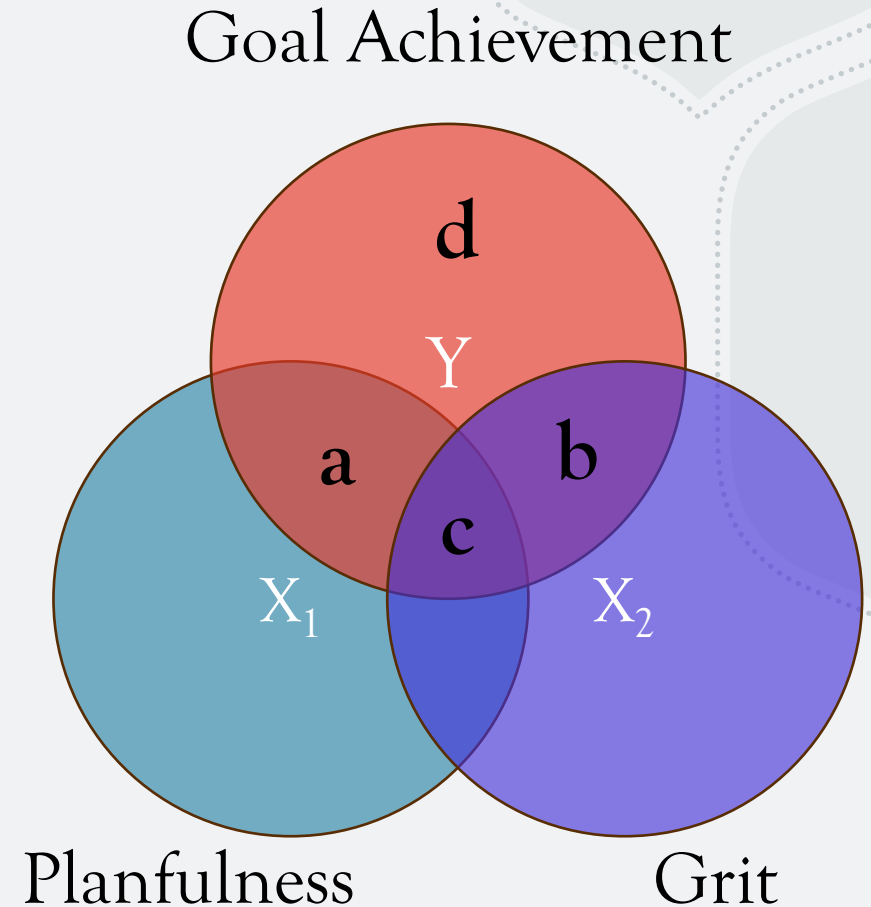


Partial Correlation

- When examining the unique relationship between planfulness and goal achievement using a **partial correlation**, the variance explained is equal to:

$$pr_1^2 = \frac{a}{a + d}$$

- The **partial correlation** measures the relationship between the part of X1 that is unrelated to X2 and the part of Y that is unrelated to X2



Partial Correlation

For our example (by hand):

	Goal Ach (Y)	Planfulness (X1)	Grit (X2)
Goal Ach (Y)	1	-	-
Planfulness (X1)	0.9033	1	-
Grit (X2)	0.8571	0.8321	1

$$pr_1 = \frac{.9033 - (.8571 * .8321)}{\sqrt{1 - .8571^2} \sqrt{1 - .8321^2}}$$

$$pr_1 = \frac{.1901}{\sqrt{.2654} \sqrt{.3076}} = 0.67$$

$$pr_1^2 = (0.67^2) = .4489, \text{ or } \approx 45\%$$

For our example (in R):

```
pcor.test(data$goal_ach, # Y  
          data$planfulness, # X1  
          data$grit) # X2
```

```
estimate p.value statistic n gp Method  
1 0.66553 0.33447 1.261037 5 1 pearson
```

```
> (0.6655)^2  
[1] 0.4428902
```

Planfulness explains approximately 45% of the variance in goal achievement after partialling out the relationship that both goal achievement and planfulness have with grit.

Semipartial Correlations

- Semipartial correlations are **most similar conceptually** to regression coefficients (i.e., the parameter estimates) in a linear model with multiple continuous predictors.

Semipartial correlation between X1 and Y

$$sr_1 = \frac{r_{Y1} - r_{Y2}r_{12}}{\sqrt{1 - r_{12}^2}}$$

Standardized regression coefficient for X1 predicting Y

$$b^*_1 = \frac{r_{Y1} - r_{Y2}r_{12}}{1 - r_{12}^2}$$

- The only difference is the square root in the denominator, which bounds the semipartial correlation to between ± 1 , whereas the standardized regression coefficient can be larger than 1.

Semipartial Correlations

- A semipartial correlation is used to demonstrate that a predictor has a relationship with an outcome variable over and above the relationship that other predictor variables have with the outcome variable
- This is how we will interpret the meaning of the **parameter estimates** (aka, regression coefficients) in a model with **multiple continuous predictors**

Semipartial correlation between X1 and Y

$$sr_1 = \frac{r_{Y1} - r_{Y2}r_{12}}{\sqrt{1 - r_{12}^2}}$$

Standardized regression coefficient for X1 predicting Y

$$b^*_1 = \frac{r_{Y1} - r_{Y2}r_{12}}{1 - r_{12}^2}$$

Question: What is the standardized regression coefficient equal to if there is **no correlation** (i.e., no redundancy) between predictors X1 and X2?

Partial Correlations

- Partial correlations are used to rule out **third variable** arguments by removing X2's relationship from *both* X1 and Y
- Demonstrates whether X1 and Y are still related even after completely removing the relationship between each and a third variable

