REGRESSION

Data Analysis for Journalism and Political Communication (Fall 2025)

Prof. Bell

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- But often, a t-test is too restrictive for the analysis we want to conduct:
 - ▶ What if we have two continuous variables? Income, age, and years of education are common variables that we may not want to force into two discrete categories.
 - ▶ What about **confounders**? For example, we found that Americans allocate less money to welfare applicants who are rated as "poor" workers compared to "excellent" workers. What are some possible confounders we need to consider?

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- But often, a t-test is too restrictive for the analysis we want to conduct:
 - What if we have two continuous variables? Income, age, and years of education are common variables that we may not want to force into two discrete categories.
 - What about confounders? For example, we found that Americans allocate less money to welfare applicants who are rated as "poor" workers compared to "excellent" workers. What are some possible confounders we need to consider?
- Regression is a set of tools that allow us to efficiently evaluate the correlation between two variables while accounting for potential confounders

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Introduction to Linear Regression

- We will focus on the simplest regression method called Ordinary Least Squares (OLS), or linear regression
- Linear regression is used to estimate the effect of a change in the **independent** (explanatory) variable on the mean of the **dependent** (outcome) variable
- Why is it called "linear" regression? In practice, we are just running a best fit line through a scatter plot of two variables.

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ESTIMATING THE REGRESSION (BEST FIT) LINE

The linear regression equation is:

$$Y = \beta_0 + \beta_1 X + \epsilon$$

Y = dependent variable β_0 = intercept β_1 = slope, also called a coefficient X = independent variable ϵ = error

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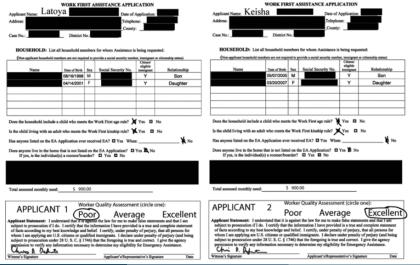
In-class exercise



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"Working Twice as Hard to Get Half as Far: Race, Work Ethic, and America's Deserving Poor"

Christopher D. DeSante, Am. Journal of Political Science vol. 57 iss. 2 (2013)



You want to estimate the effect of income on the allocation to welfare applicants. The linear regression equation is:

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So the coefficient (β_1) is the effect of a **one-unit** change in income (thousands) on the mean allocation.



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Recall that the linear regression equation is:

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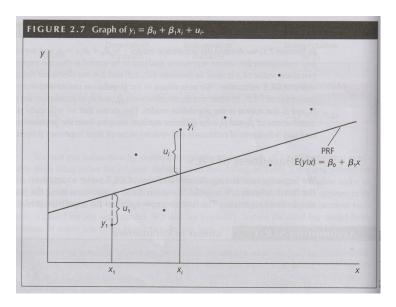


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- Error (ϵ or u) is also called the residual (left over from $Y = \beta_0 + \beta_1 X$, our best fit line)
- Our goal in regression is to fit the best line that minimizes the error
- However, we can never get the ϵ = 0, and often we don't even get close. We just do the best we can to make the "best" best fit line.

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The multiple linear regression equation is:

$$Y = \beta_0 + \beta_1 X_1 + \beta_2 X_2 + \epsilon$$

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What is the interpretation of β_1 ?

 β_1 is the effect of a one-unit change in X_1 on the mean of Y, holding X_2 constant (the independent effect of X_1 on Y)



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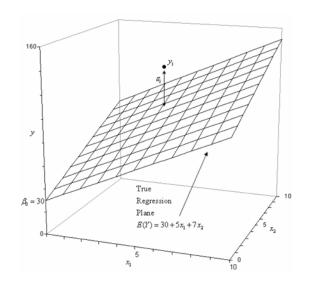
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What is the interpretation of β_2 ?

 β_2 is the effect of a one-unit change in X_2 on the mean of Y, holding X_1 constant (the independent effect of X_2 on Y)



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$$Allocation = \hat{\beta_0} + \hat{\beta_1} Race + \hat{\beta_2} WorkEthic + \hat{\beta_3} Age + \hat{\beta_4} Gender + \hat{\beta_5} PoliticalParty + \hat{\beta_6} Income$$



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- R^2 ranges from 0 to 1 and represents the proportion of change in Y explained by X_i

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- One way to evaluate how well you are explaining change in Y is through the \mathbb{R}^2 , technically called the "coefficient of determination"
- R^2 ranges from 0 to 1 and represents the proportion of change in Y explained by X_i
- My pet peeve is over-relying on the \mathbb{R}^2 theory should drive your modeling decisions, not a summary statistic

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- This assumption is usually fine when we are working with continuous dependent variables like income
- What about categorical dependent variables, like party ID? The linear regression model is not well suited for these.
- What about binary dependent variables, like support for a policy?

We call this a linear probability model because it estimates the effect of a one-unit change in X on the *probability* (percent chance) of a value of "1" for the dependent variable