WHEN TO USE CONTINUOUS VS. ORDINAL

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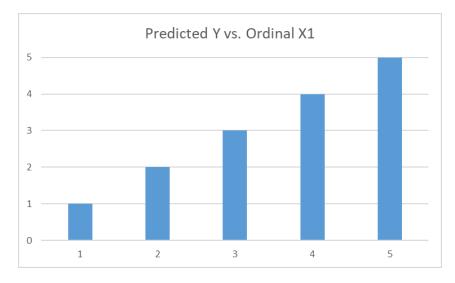
WHAT'S THE PROBLEM?

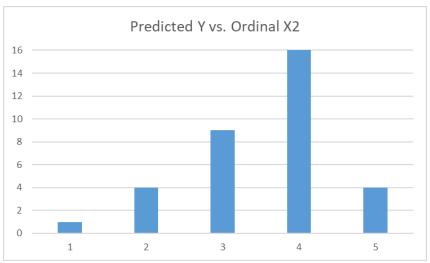
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- Typically the question is asked as:
 - How many levels before its continuous?
 - Have rules of thumb used in industry...
 "Anything fewer than 10 is probably ordinal..."
 - "Anything more than 20 is probably continuous..."
 - "Anything in between... well... it depends... ©"

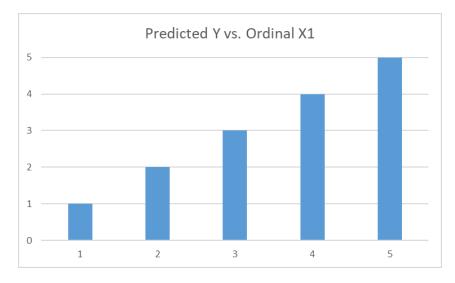
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- Typically the question is asked as:
 - How many levels before its continuous?
- This is actually the wrong question!

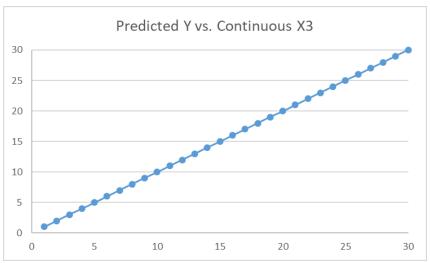
- Must understand the relationship between ordinal and continuous variables.
- Let's compare visually:



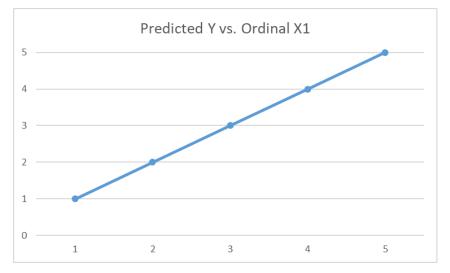


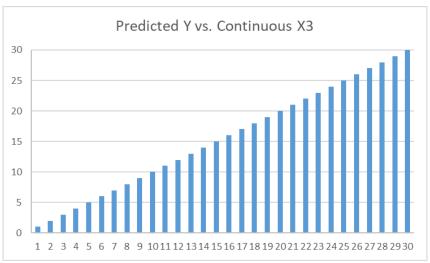
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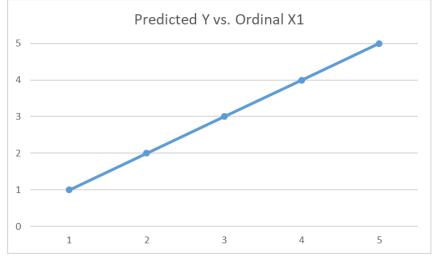


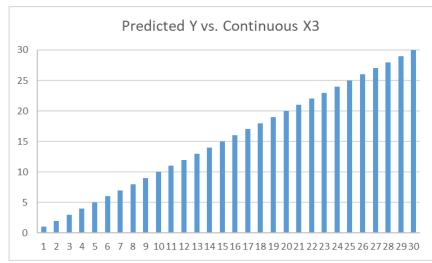
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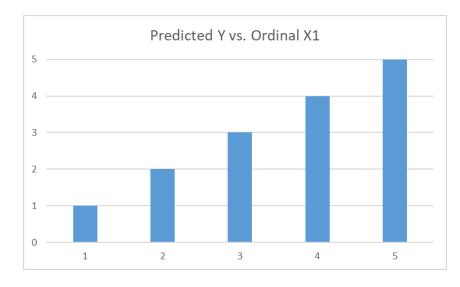


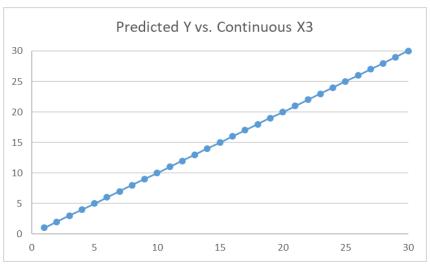
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- The jumps (in y) between points (values of x) are the same.





- What does the straight line imply?
- The jumps (in y) between points (categories of x) are the same.





- It is always a question of whether to treat a variable as an ordinal variable or continuous variable in a regression model.
- Typically the question is asked as:
 - How many levels before its continuous?
- Are you willing to assume the jumps (slope) between each value of x (category) are the same?

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- More mathematically... are you willing to substitute equation 1 for equation 2?

$$\hat{y} = \hat{\beta}_0 + \hat{\beta}_1 x_1$$

$$\hat{y} = \hat{\beta}_0 + \hat{\beta}_1 D_1 + \hat{\beta}_2 D_2 + \dots + \hat{\beta}_{k-1} D_{k-1}$$

Design variables for categories of x_1



WHAT'S THE SOLUTION?

Comparison of Models

- Let the data decide if it should be modeled as continuous or categorical.
- Compare the following models:

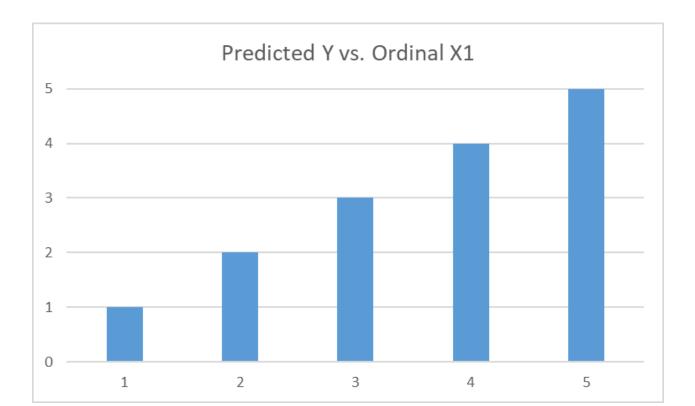
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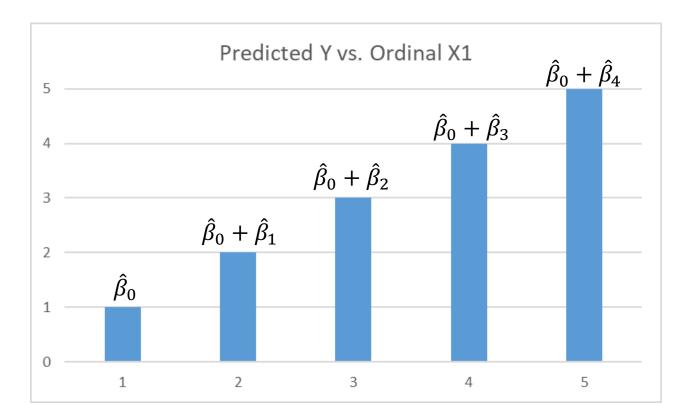
• The first model is a special case (nested within) of the second model!

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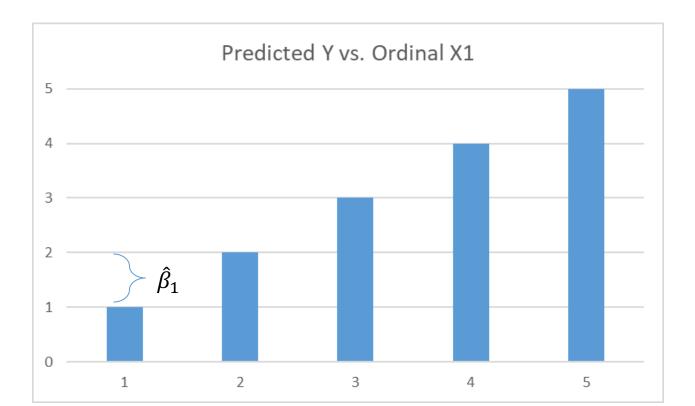
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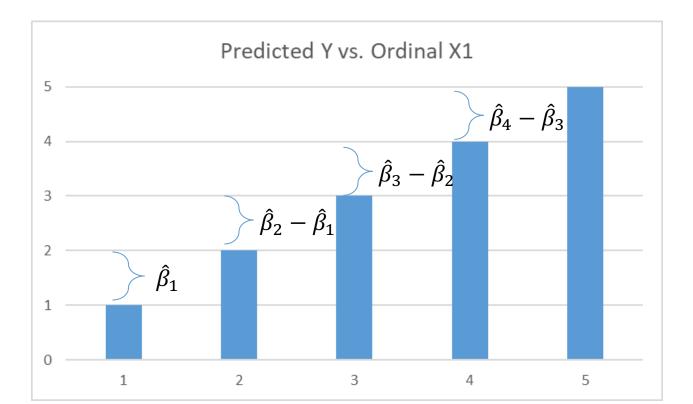
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 Let's examine that second model a little closer to see how the first one could be a special case.

$$\hat{y} = \hat{\beta}_0 + \hat{\beta}_1 D_1 + \hat{\beta}_2 D_2 + \hat{\beta}_3 D_3 + \hat{\beta}_4 D_4$$

What if the following were true?

$$\hat{\beta}_1 = \hat{\beta}_2 - \hat{\beta}_1 = \hat{\beta}_3 - \hat{\beta}_2 = \hat{\beta}_4 - \hat{\beta}_3$$

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Only need to really estimate $\hat{\beta}_1!!!$

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- Let the data decide if it should be modeled as continuous or categorical.
- Compare the following models:

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• The first model is a special case (nested within) of the second model!

- Many different statistical tests can compare two models if they are nested.
- Linear regression primarily uses the Nested F Test.
- Logistic regression primarily uses the Likelihood Ratio Test (LRT).



HOW TO IMPLEMENT SOLUTION?

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OR

$$H_0: \hat{\beta}_1 = \hat{\beta}_2 - \hat{\beta}_1 = \hat{\beta}_3 - \hat{\beta}_2 = \hat{\beta}_4 - \hat{\beta}_3$$

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 If models are the same USE THE SIMPLER MODEL (aka treat the variable as continuous)!

Likelihood Ratio Test Example – R

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```
## Analysis of Deviance Table
##
## Model 1: INS ~ factor(CCPURC)
## Model 2: INS ~ CCPURC
## Resid. Df Resid. Dev Df Deviance Pr(>Chi)
## 1 7415 9611.7
## 2 7418 9628.2 -3 -16.514 0.0008896 ***
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
```

