Econometrics: Regression Analysis With Qualitative Information

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- ▶ In the regression model, there are often variables of interest that are qualitative and can not be measured as a quantitative variable.
- ► These variables, called "dummy", or "binary" variables, measure some qualitative characteristics such as:
 - Gender male or female;
 - Immigration status: immigrant or not;
 - Marital status: married or not;
 - Residence status reside in a particular city or not;
 - Sector of a company: manufacturing or service sector;
 - Size of a company: big or small;
 - Month of the year, and so on.



Dummy Variables

- ► Using dummy variables, we can measure the effect of the qualitative factor on our dependent variable
- ▶ Typically, the dummy variables take value 1 in a category and value 0 "otherwise". "Otherwise" can represent one or more other categories. For example:

$$Female = \begin{cases} 1 & \text{if the individual is female} \\ 0 & \text{if the individual is male} \end{cases}$$

$$Male = \begin{cases} 1 & \text{if the individual is male} \\ 0 & \text{if the individual is female} \end{cases}$$



Dummy Variables

$$Small = \begin{cases} 1 & \text{if the firm is small} \\ 0 & \text{otherwise} \end{cases}$$

$$Medium = \begin{cases} 1 & \text{if the firm is medium size} \\ 0 & \text{otherwise} \end{cases}$$

$$Big = \begin{cases} 1 & \text{if the firm is big} \\ 0 & \text{otherwise} \end{cases}$$



Dummy Variables

- ▶ Dummy variables help us with two different aspects
 - Additive dummy variables measure differences in groups with respect to the intercept term
 - **Interaction** dummy variables measure differences in groups with respect to the slope term
- Dummy variable trap: Suppose you have a set of multiple dummy variables for multiple categories and every observation falls in one and only one category. Then, if you include all these dummy variables and a constant term (β₀), you will have perfect multicollinearity. Also known as dummy variable trap.



- Additive dummy variables result in different intercepts for different populations.
- ► So, we have $E[Y|X_{1i}, X_{2i}] = \beta_0 + \beta_1 X_{1i} + \beta_2 X_{2i}$. This implies
 - For females: $E[Y|X_{1i}, X_{2i} = 1] = (\beta_0 + \beta_2) + \beta_1 X_{1i}$
 - For males: $E[Y|X_{1i}, X_{2i} = 0] = \beta_0 + \beta_1 X_{1i}$



- ▶ $\beta_2 = E[Y|X_{1i}, \text{ female}] E[Y|X_{1i}, \text{ male}]$ is the average difference between a women and a man for a given level of education.
- Assuming that $\beta_2 < 0$, graphically we have:



► There are two alternative formulations for this model: 1. $Y_i = \alpha_0 + \alpha_1 X_{1i} + \alpha_2 X_{3i} + \varepsilon \ i = 1, ..., n$, where: $X_{3i} = \begin{cases} 1 & \text{if the individual is male} \\ 0 & \text{if the individual is female} \end{cases}$ 2. $Y_i = \delta_1 X_{1i} + \delta_2 X_{2i} + \delta_3 X_{3i} + \varepsilon \ i = 1, ..., n$



Additive Dummy Variables: Alternative Model (1)

- $Y_i = \alpha_0 + \alpha_1 X_{1i} + \alpha_2 X_{3i} + \varepsilon_i \ i = 1, \dots, n$. Now we have:
- E[Y|X_{1i}, X₃] = α₀ + α₁X₁ + α₃X₃, hence
 E[Y|X_{1i}, female] = E[Y|X_{1i}, X_{3i} = 0] = α₀ + α₁X₁,
 E[Y|X_{1i}, male] = E[Y|X_{1i}, X_{3i} = 1] = (α₀ + α₂) + α₁X₁,
 α₂ = E[Y|X_{1i}, male] E[Y|X_{1i}, female] is the average difference between a women and a man for a given level of education.
 - Therefore our model should satisfy:

$$\alpha_1 = \beta_1$$
$$\alpha_0 = \beta_0 + \beta_2$$
$$\alpha_0 + \alpha_2 = \beta_0$$



Additive Dummy Variables: Alternative Model (2)

- $Y_i = \delta_1 X_{1i} + \delta_2 X_{2i} + \delta_3 X_{3i} + \varepsilon_i \ i = 1, \dots, n$. Now we have:
- $E[Y|X_{1i}, X_2, X_3] = \delta_1 X_{1i} + \delta_2 X_{2i} + \delta_3 X_{3i}$, hence
 - $E[Y|X_{1i}, \text{ female}] = E[Y|X_{1i}, X_{2i} = 1, X_{3i} = 0] = \delta_2 + \delta_1 X_{1i},$
 - $E[Y|X_{1i}, \text{ male}] = E[Y|X_{1i}, X_{2i} = 0, X_{3i} = 1] = \delta_3 + \delta_1 X_{1i},$
 - $\delta_3 \delta_2 = E[Y|X_{1i}, \text{ male}] E[Y|X_{1i}, \text{ female}]$ is the average difference between a women and a man for a given level of education.
 - Therefore our model should satisfy:

$$\delta_1 = \alpha_1 = \beta_1$$
$$\delta_2 = \alpha_0 = \beta_0 + \beta_2$$
$$\delta_3 = \alpha_0 + \alpha_2 = \beta_0$$



► However, notice that a model like $Y_i = \delta_0 + \delta_1 X_{1i} + \delta_2 X_{2i} + \delta_3 X_{3i} + \varepsilon_i \ i = 1, ..., n$ Would **not** be valid due to multicollinearity (Recall problem 2 of set 3)



- ▶ How would we test if there are significant differences between the two groups: male and female?
 - For model $Y_i = \beta_0 + \beta_1 X_{1i} + \beta_2 X_{2i} + \varepsilon_i \Rightarrow H_0 : \beta_2 = 0$ vs. $H_1 : \beta_2 \neq 0$
 - For model $Y_i = \alpha_0 + \alpha_1 X_{1i} + \alpha_3 X_{3i} + \varepsilon_i \Rightarrow H_0 : \alpha_3 = 0$ vs. $H_1 : \alpha_3 \neq 0$
 - For model $Y_i = \delta_1 X_{1i} + \delta_2 X_{2i} + \delta_3 X_{3i} + \varepsilon_i \Rightarrow H_0 : \delta_2 = \delta_3$ vs. $H_1 : \delta_2 \neq \delta_3$



- ▶ We use interaction dummy variables to account for the changes due to the dummy categories, in the effect of the independent variables, i.e., X₁: education,? on Y
- Consider an example with additive and interaction effects: $Y_i = \beta_0 + \beta_1 X_{1i} + \beta_2 X_{2i} + \beta_3 X_{4i} + \varepsilon_i \ i = 1, \dots, n$, where $X_{4i} = X_{1i} \times X_{2i}$.
- In this case, $X_{4i} = \begin{cases} X_{1i} & \text{if the individual is female} \\ 0 & \text{if the individual is male} \end{cases}$
- ► So, we have $E[Y|X_{1i}, X_{2i}, X_{4i}] = \beta_0 + \beta_1 X_{1i} + \beta_2 X_{2i} + \beta_3 X_{4i}$. This implies
 - For females: $E[Y|X_{1i}, \text{ female}] = (\beta_0 + \beta_2) + (\beta_1 + \beta_3)X_{1i}$
 - For males: $E[Y|X_{1i}, X_{2i} = 0] = \beta_0 + \beta_1 X_{1i}$



- β_2 measures the difference in the intercept term between men and women. That is, it is the difference on the mean income of men and women
- ▶ β_3 measures the difference in the slope term between men and women. That is, if education (X_1) increases by 1 year, the on average, the hourly wage increases by:
 - $\beta_1 + \beta_3$ units for women, and
 - β_1 units for men.
 - Thus, measures the differences in the average effect of education on wages due to different genders



► How to test if there are significant differences between genders for the effect of education on the wage rate

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$$\Rightarrow$$
 $H_0: \beta_3 = 0$ vs. $H_1: \beta_3 \neq 0$

► How to test if there are significant differences between genders, on average

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$$\Rightarrow$$
 $H_0: \beta_2 = 0$ vs. $H_1: \beta_2 \neq 0$

▶ How to test if there are any significant difference between men and women

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$$\Rightarrow$$
 $H_0: \beta_2 = \beta_3 = 0$ vs. $H_1: \beta_2 \neq 0$ and/or $\beta_2 \neq 0$



Interaction Dummy Variables: Additional Comments

- ► As in additive dummy variable models, there are alternative specifications for the interaction dummy variable models.
 - For example: $Y_{i} = \alpha_{0} + \alpha_{1}X_{1i} + \alpha_{2}X_{3i} + \alpha_{3}X_{5i} + \varepsilon_{i} \ i = 1, \dots, n, \text{ where}$ $X_{5i} = X_{1i} \times X_{3i}$ • In this case, $X_{3i} = \begin{cases} 1 & \text{if the individual is male} \\ 0 & \text{if the individual is falemale} \end{cases}$ • In this case, $X_{5i} = \begin{cases} X_{1i} & \text{if the individual is male} \\ 0 & \text{if the individual is male} \end{cases}$
 - Alternatively: $Y_i = \delta_1 X_{2i} + \delta_2 X_{3i} + \delta_3 X_{4i} + \delta_4 X_{5i} + \varepsilon_i \ i = 1, \dots, n$



► However, a model like the following will **not** be valid: $Y_i = \gamma_1 X_{1i} + \gamma_2 X_{2i} + \gamma_3 X_{3i} + \gamma_4 X_{4i} + \gamma_5 X_{5i} + \varepsilon_i \ i = 1, ..., n$ since it violates **A4** (no perfect multicollinearity) because $X_{4i} + X_{5i} = X_{1i} \ \forall i \in 1, ..., n$



▶ We may have more than two categories for our dummy variable. For example, assume that firms are divided into three sectors, i.e., services, manufacturing, and agriculture

$$\begin{array}{l} \blacktriangleright V_i = \alpha_0 + \alpha_1 S_{1i} + \alpha_2 S_{2i} + \alpha_3 P_i + \alpha_4 \left(P_i \times S_{1i} \right) + \\ \alpha_5 \left(P_i \times S_{2i} \right) + \varepsilon_i \; i = 1, \ldots, n, \text{ where} \\ \bullet \; V_i = \text{Sales of the company } i \\ \bullet \; P_i = \text{Advertising expenditures of the company } i \\ \bullet \; S_{1i} = \begin{cases} 1 & \text{if the company } i \text{ belongs to sector } 1 \\ 0 & \text{otherwise} \end{cases} \\ \bullet \; S_{2i} = \begin{cases} 1 & \text{if the company } i \text{ belongs to sector } 2 \\ 0 & \text{otherwise} \end{cases}$$



- ► Then:
 - $E[V_i|P_i, \text{ sector } 1] = (\alpha_0 + \alpha_1) + (\alpha_3 + \alpha_4)P_i$
 - $E[V_i|P_i, \text{ sector } 2] = (\alpha_0 + \alpha_2) + (\alpha_3 + \alpha_5)P_i$
 - $E[V_i|P_i, \text{ sector } 3] = \alpha_0 + \alpha_3 P_i$



- In this particular representation of the model, in order to include both the constant term and the variable P_i , we exclude the additive and interaction effects corresponding to sector 3, and only included those of sector 1 ans 2
 - α_0 corresponds to the additive dummy for sector 3 (the constant term for sector 3)
 - α_3 corresponds to the interaction dummy for sector 3 (the effect of advertising on sector 3 sales) which we ignore (Sector 3)
 - The intercept for the other sectors, namely, 1 and 2 are $(\alpha_0 + \alpha_1)$ and $(\alpha_0 + \alpha_2)$, respectively
 - The slopes for the other sectors, namely, 1 and 2 are $(\alpha_3 + \alpha_4)$ and $(\alpha_3 + \alpha_5)$, respectively



- ► There are many alternative representations for this model. One possible way is: $V_i = \delta_1 S_{1i} + \delta_2 S_{2i} + \alpha_3 S_{3i} + \delta_4 (P_i \times S_{1i}) + \delta_5 (P_i \times S_{2i}) + \delta_6 (P_i \times S_{3i}) + \varepsilon_i i = 1, \dots, n$
- Comparing both representation, what are the relationships between α_j's and δ_j's?
- ▶ How would you test for the effects of sector on sales?

