ES1004 Econometrics by Example

Lecture 3: Qualitative Explanatory Variables

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Gujarati textbook, second edition

14th May 2016



Basic Idea I

- sometimes cannot obtain set of numerical values for all variables to use in a model
- because some variables cannot be quantified easily
- examples
 - gender may play a role in determining salary levels
 - different ethnic groups may follow different consumption patterns
 - educational levels can affect earnings from employment



Basic Idea II

- qualitative variables as regressors
 - to include in a regression we define dummy variables
 - nominal scale variables which have no particular numerical values
 - usually in cross-sectional models, but can appear in time series as well
- more examples [in times series]
 - changes in political regime may affect production
 - war can impact on economic activities
 - certain days in week or certain months in year can have different effects on the fluctuation of stock prices
 - seasonal effects often observed in demand of various products



Basic Idea III

- note that dummy variables are also called
 - indicator variables
 - categorical variables, and
 - qualitative variables





Including Dummy Variables I

• consider following cross-sectional model

$$wage_i = \beta_1 + \beta_2 exper_i + u_i$$

- this model assumes that the constant β_1 is the same for all the observations in our dataset
- what if we have two different subgroups
 - male and female, for example





Including Dummy Variables II

 we convert such qualitative information into a quantitative variable by creating a dummy variable

$$D = \begin{cases} 1 & \text{if female} \\ 0 & \text{if male} \end{cases}$$

- note that
 - i the choice of which of the two different outcomes is to be assigned the value of 1 does not alter the results
 - ii the 0 classification is often referred to as the benchmark, or control category



Including Dummy Variables III

$$wage_i = \beta_1 + \beta_2 exper_i + \beta_3 D_i + u_i$$

- now we have two cases
 - when D = 0 (male)

$$Y_i = \beta_1 + \beta_2 X_i + u_i$$

② when D = 1 (female)

$$Y_i = (\beta_1 + \beta_3) + \beta_2 X_i + u_i$$

• two groups [male & female] but included only one dummy D_i





Dummy Variable Trap

- if an intercept is included and we have a qualitative variable with m categories, then introduce only (m-1) dummy variables
 - consider a self-reported health as a choice among excellent, good, and poor
 - we can have at most two dummy variables to represent three categories
- ullet not following this rule o dummy variable trap o perfect collinearity





Reference Category

- the subgroup that gets value of 0 is called the
 - reference category,
 - benchmark, or
 - comparison category
- all comparisons are made in relation to the reference category
 - if there are several dummy variables, you must keep track of the reference category





Modelling Wages: Data

- we want to study what factors determine hourly wage (in dollars)
- table 1.1 data of 1289 individuals interviewed in March 1995
 - wage hourly wage in dollars [dependent variable]
 - female 🕼 gender, coded 1 for female, 0 for male
 - nonwhite race, coded 1 for nonwhite, 0 for white workers
 - union is union status, coded 1 if in a union job, 0 otherwise
 - education education in years
 - exper 🕼 potential work experience in years





Modelling Wages: Estimation

Dependent Variable: WAGE Method: Least Squares Date: 05/14/16 Time: 09:39

Sample: 1 1289

Variable	Coefficient	Std. Error	t-Statistic	Prob.
C FEMALE NONWHITE UNION EDUCATION EXPER	-7.183338	1.015788	-7.071691	0.0000
	-3.074875	0.364616	-8.433184	0.0000
	-1.565313	0.509188	-3.074139	0.0022
	1.095976	0.506078	2.165626	0.0305
	1.370301	0.065904	20.79231	0.0000
	0.166607	0.016048	10.38205	0.0000





Refining the Wage Function I

- we found that the average salary of a
 - female worker is lower than that of her male counterpart
 - nonwhite worker is lower than that of his white counterpart
- what about a female nonwhite?
- we need to include an interactive dummy





Refining the Wage Function I

Dependent Variable: WAGE Method: Least Squares Date: 05/14/16 Time: 10:19 Sample: 1 1289

Variable	Coefficient	Std. Error	t-Statistic	Prob.
C FEMALE NONWHITE UNION EDUCATION EXPER FEMALE*NONWHITE	-7.088725	1.019482	-6.953264	0.0000
	-3.240148	0.395328	-8.196106	0.0000
	-2.158525	0.748426	-2.884087	0.0040
	1.115044	0.506352	2.202113	0.0278
	1.370113	0.065900	20.79076	0.0000
	0.165856	0.016061	10.32631	0.0000
	1.095371	1.012897	1.081424	0.2797





Refining the Wage Function II

- we implicitly assumed that slope coefficients of quantitative regressors remain the same between
 - male and female
 - white and nonwhite
- however, we do not need to we can include differential slope dummies





Refining the Wage Function II

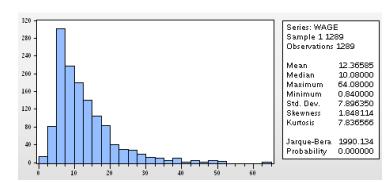
Dependent Variable: WAGE Method: Least Squares Date: 05/14/16 Time: 10:23 Sample: 1 1289

Variable	Coefficient	Std. Error	t-Statistic	Prob.
C	-11.09129	1.421846	-7.800623	0.0000
FEMALE	3.174158	1.966465	1.614144	0.1067
NONWHITE	2.909129	2.780066	1.046424	0.2956
UNION	4.454212	2.973494	1.497972	0.1344
EDUCATION	1.587125	0.093819	16.91682	0.0000
EXPER	0.220912	0.025107	8.798919	0.0000
FEMALE*EDUCATION	-0.336888	0.131993	-2.552314	0.0108
FEMALE*EXPER	-0.096125	0.031813	-3.021530	0.0026
NONWHITE*EDUCATI	-0.321855	0.195348	-1.647595	0.0997
NONWHITE*EXPER	-0.022041	0.044376	-0.496700	0.6195
UNION*EDUCATION	-0.198323	0.191373	-1.036318	0.3003
<u> </u>	-0.033454	0.046054	-0.726410	0.4677



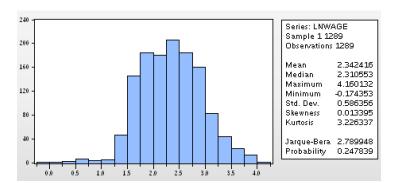


Functional Form





Functional Form





Functional Form

Dependent Variable: LNWAGE

Method: Least Squares Date: 05/14/16 Time: 10:34

Sample: 1 1289

Variable	Coefficient	Std. Error	t-Statistic	Prob.
C FEMALE NONWHITE UNION EDUCATION EXPER	0.905504 -0.249154 -0.133535 0.180204 0.099870 0.012760	0.074175 0.026625 0.037182 0.036955 0.004812 0.001172	12.20768 -9.357891 -3.591399 4.876316 20.75244 10.88907	0.0000 0.0000 0.0003 0.0000 0.0000







