

17.871, Political Science Lab
Spring 2012
Problem set # 3

Handed out: March 21

Due: April 5 (not additional day, compared to syllabus)

The purpose of this problem set is to help you understand multiple regression, especially when and why controlling for variables changes results. Submit a paper copy of your answers to each question, including regression output, code for each command, and calculations. Do all the work in one do file and attach that to your problem set when you hand it in. Work in groups. Write up separately.

Death Sentence and Discrimination

In a 1987 Supreme Court case, *McCleskey v. Kemp*, McCleskey's counsel argued that the race of both victim and defendant played a significant role in the decision to impose the death penalty, even after accounting for other facts of the case. On this basis, they argued that capital punishment was being applied in an unconstitutionally discriminatory manner.

In a 5-4 decision, the Court rejected McCleskey's claim. Justice Powell, writing for the majority, argued that statistical evidence of discrimination was not sufficient to overturn McCleskey's conviction. Instead, Powell wrote, McCleskey would have to show strong evidence of racial bias in this particular case. Statistical patterns alone would not be sufficient. (Powell, now retired, is quoted in a recent book as saying he has now changed his mind on this issue and would vote differently if presented with the case today.) While *McCleskey v. Zant* has essentially settled the question for the courts, this issue has not disappeared. In an amendment to the 1994 "Crime Bill" members of the House of Representatives proposed a provision allowing statistical evidence to be used to overturn death penalty convictions. The Senate rejected the amendment and a hot debate between the two chambers ensued. The amendment was removed from the final bill, over the strong objections of many House members, especially the Congressional Black Caucus.

With that background, reanalyze data from the McCleskey case. In the file *death.dta* (in Course Locker, Examples folder) you will find 100 cases used as part of McCleskey's evidence (this is a subset of the cases used by McCleskey).

The variables:

ID	1-100 ID number of each case death
death	1=Death sentence 0=Life sentence
bd	1=Black defendant 0=White defendant
wv	1=One or more white victims 0=No white victims
ac	Number (1-7) of aggravating circumstances
fv	1=Female victim 0=Male victim
vs	1=Victim was a stranger to defendant 0=Victim was known to defendant
v2	1=Two or more victims 0=One victim
ms	1=Multiple stabs 0=No multiple stabs
yv	1=Victim 12 years of age or younger 0=Victim over 12 years of age

Analyze these data.

- (2 points) In sentencing people to death, does the Justice system discriminate against blacks? Use cross tabulations (the *tabulate* command) to check. Show and briefly summarize your findings. (Hint: consider more than one form of discrimination. If you are not finding discrimination, keep looking. Try other variables.)
- (2 points) Regress death on bd, that is, run a regression where death is the dependent variable and bd is the explanatory variable. Report and interpret the coefficient and the constant. (Hint: in regressions with a binary dependent variable, the coefficients can be interpreted as probabilities.)
- (3 points) Now regress death on bd while controlling for wv. Report and interpret the coefficients and the constant. Why does the coefficient on bd change when controlling for wv? That is, why does it change from the bivariate regression in your answer to b? (Hint:

look at the correlation matrix by running the command *corr*, and review the lecture slides about the difference between bivariate in multivariate regression coefficients.)

- d. (4 points) Consider your findings in part a. What variables in the data set might provide alternative explanations for your finding? In other words, what variables in the data set might lead to the appearance (or absence) of discrimination in a. In answering this question, you might find it helpful to look at the results from the correlations in c., but you should primarily rely on your expectations and reasoning about the criminal justice system. In your answer, note the variables and your explanation for why they may constitute alternative explanations for your finding in part a. Discuss at least two control variables.
- e. (2 points) Regress death on all the substantive variables. Before doing so, code *ac* so that it varies from zero to one. Do your bivariate findings from part a. hold? That is, does controlling for the other variables change the bivariate results? Report the regression results and discuss whether your findings from part a. change. In doing so, interpret the coefficient on *wv*.
- f. (3 points) From part e., which variable has the largest effect on the death sentence? By largest, I mean has largest effect if the variable changed from its min to its max. Interpret this variable's coefficient.
- g. (4 points) The O.J. Simpson case presented circumstances that can be applied to these data. For the benefit of those not familiar with the case, here is a brief summary. Simpson is a retired American football player. In 1994, he was charged with murder in the deaths of his former wife and a man she knew. The victims were killed by multiple stab wounds. Simpson obviously knew his ex-wife, but apparently did not know the male victim. Simpson is black. Both victims were white. No children were killed. We do not know how many aggravating circumstances there were, but it seems likely to be at least 1 and perhaps as many as 3 or 4. Use this information in your regression model from part e. to compute predicted probabilities that Simpson would have received the death penalty if he had been convicted of murder based on these data from Georgia. (In fact, the Los Angeles District Attorney decided not to seek the death penalty, and Simpson was acquitted of the murder charges. That does not diminish the interest in asking if someone in a case similar to Simpson's could expect the death penalty.) Some of the explanatory variables are easily applicable. Two, *vs* and *ac*, are not so easily applicable. You decide how to handle these, and justify your decision. Discuss the results. (Hint: We've only briefly covered model-based predictions, but they are (usually) straightforward. You simply need to plug in the hypothetical values for the explanatory variables into the model produced by the regression. Don't forget to add the constant. You can probably do this most quickly by hand or in Excel. You can also use Stata's *margins* command.)