

Pol Sci 630: Problem Set 6 Solutions: Dummy Variables and Interactions

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Due Date: Wednesday, Oct 12, 2016 (Beginning of class)

1 Merging data (8 points)

Insert your comments on the assignment that you are grading above the solution in bold and red text. For example write: "GRADER COMMENT: everything is correct! - 8/8 Points" Also briefly point out which, if any, problems were not solved correctly and what the mistake was. See below for more examples.

The most common merging task in political science is to merge datasets based on country-year. The biggest obstacle is that country codes can come in many forms (country name, World Bank code, COW code, ISO2, ISO3, etc.)

This exercise will let you dip your toes in the sea of pain that is merging real world data. You're expected to Google and read help files to figure out two packages: 1) `countrycode`, which converts between different types of country codes, and 2) `psData`, a package that automates the downloading of many common Political Science dataset.

This exercise is not technically hard, just requires you to figure out things on your own.

1.1 Download WDI data

Download GDP per capita ('NY.GDP.PCAP.CD') and FDI ('BX.KLT.DINV.CD.WD') from WDI, 2007-2009, `extra = FALSE`. What country indicators are there?

Note: There should be 792 rows

Solution

```
library(WDI)

## Loading required package: RJSONIO

d_wdi <- WDI(indicator = c("NY.GDP.PCAP.CD", 'BX.KLT.DINV.CD.WD'),
             start = 2007, end = 2009, extra = FALSE)

names(d_wdi)
```

```
## [1] "iso2c"          "country"          "year"
## [4] "NY.GDP.PCAP.CD"    "BX.KLT.DINV.CD.WD"

nrow(d_wdi)

## [1] 792
```

The two indicators are 'country' and 'iso2c'

1.2 Download Polity data

Use `PolityGet()` in package `psData` to download Polity data. Download the 'polity2' variable (*not* the entire dataset). Use 'iso3c' as the format for the country code.

What country indicators are there?

Note: There should be 16351 rows

Solution

```
library(psData)
d_polity <- PolityGet(vars = 'polity2', OutCountryID = 'iso3c')
names(d_polity)

## [1] "iso3c"          "standardized_country" "country"
## [4] "year"           "polity2"

nrow(d_polity)

## [1] 16351
```

The country indicators are 'country' and 'iso3c'

1.3 Convert country code

To merge WDI and Polity data we must first create a common country ID. (We can't use country name, because there's no guarantee they will be the same). Use package `countrycode` to convert the country code in WDI data from 'iso2c' to 'iso3c'. Store this newly created country code in the WDI data frame.

Solution

```
library(countrycode)

##
## Attaching package: 'countrycode'
## The following object is masked from 'package:psData':
##
##   countrycode_data
```

```
d_wdi$iso3c <- countrycode(d_wdi$iso2c,
                           origin = 'iso2c', destination = 'iso3c')
```

1.4 Merge

Merge the WDI and the Polity data based on 'iso3c' and 'year' (Note: There should be 492 rows).

There are two variables showing country names in the merged dataset. Why? Clean them up so we only have 1 country name variable in the merged dataset.

Solution

```
d_merged <- merge(d_wdi, d_polity, by = c("iso3c", "year"))
nrow(d_merged)

## [1] 492

head(d_merged)

##   iso3c year iso2c  country.x NY.GDP.PCAP.CD BX.KLT.DINV.CD.WD
## 1   AFG 2007   AF Afghanistan      380.4010      188690000
## 2   AFG 2008   AF Afghanistan      384.1317      46033740
## 3   AFG 2009   AF Afghanistan      458.9558      197512727
## 4   AGO 2007   AO      Angola      3151.0224     -893342152
## 5   AGO 2008   AO      Angola      4242.3631      1678971010
## 6   AGO 2009   AO      Angola      3678.9477      2205298180
##   standardized_country  country.y polity2
## 1             Afghanistan Afghanistan      NaN
## 2             Afghanistan Afghanistan      NaN
## 3             Afghanistan Afghanistan      NaN
## 4                 Angola      Angola      -2
## 5                 Angola      Angola      -2
## 6                 Angola      Angola      -2

# Drop extra country variable and clean up
d_merged$country.y <- NULL
colnames(d_merged)[colnames(d_merged) == "country.x"] <- "country"
```

1.5 Check merged result

(Optional) Figure out which country years appear in WDI data but not in Polity data. Note: There should be 300 unmatched records.

In real research, this is useful to check that you are not throwing away data erroneously. There are more than one way to do this and should require some Googling.

Solution

```
# My favorite way
library(dplyr)

##
## Attaching package: 'dplyr'
## The following objects are masked from 'package:stats':
##
## filter, lag
## The following objects are masked from 'package:base':
##
## intersect, setdiff, setequal, union

d_unmatched <- anti_join(d_wdi, d_polity, by = c("iso3c", "year"))
nrow(d_unmatched)

## [1] 300
```

2 Factors and Regression with Factors (8 points)

"GRADER COMMENT: everything is correct! - 8/8 Points"

2.1 Dichotomize a continuous variable

Create a new factor variable in your merged dataset, called `polity2_binary` that is 1 (labeled 'democracy') when `polity2` ≥ 0 , and 0 (labeled 'dictatorship') otherwise.

Solution

```
d_merged$polity2_binary <- ifelse(d_merged$polity2 >= 0, 1, 0)

d_merged$polity2_binary <- factor(d_merged$polity2_binary,
                                levels = c(0, 1),
                                labels = c("dictatorship", "democracy"))
```

2.2 Regression with one binary variable

Regress FDI on the binary variable `polity2_binary`. From the regression result, report the average amount of FDI that democracy and dictatorship gets.

Note: You should know this from the regression result, not from running `mean()`

Solution

```

# Fancy way (dplyr) to rename variables that you'll learn one day
d_merged <- d_merged %>%
  rename(gdppc = NY.GDP.PCAP.CD, fdi = BX.KLT.DINV.CD.WD)

m_3a <- lm(fdi ~ polity2_binary, data = d_merged)
summary(m_3a)

##
## Call:
## lm(formula = fdi ~ polity2_binary, data = d_merged)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -4.583e+10 -1.550e+10 -9.131e+09 -4.598e+09  7.179e+11
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)    6.396e+09  4.047e+09   1.580   0.1147
## polity2_binarydemocracy 9.759e+09  4.816e+09   2.026   0.0433 *
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 4.806e+10 on 478 degrees of freedom
## (12 observations deleted due to missingness)
## Multiple R-squared:  0.008518, Adjusted R-squared:  0.006443
## F-statistic: 4.106 on 1 and 478 DF, p-value: 0.04328

```

The average FDI for dictatorship is 6.3956174×10^9 USD.

The average FDI for democracy is 1.6154234×10^{10} USD.

2.3 Regression with interaction and interpretation

Regress FDI against `polity2_binary`, `gdppc`, and their interaction term.

I want to plot FDI against `gdppc` with two lines, one representing democracy, the other representing dictatorship (similar to the last plot in the lab). What would be the intercept and slope of these two lines?

Solution

Run regression

```

m_3b <- lm(fdi ~ polity2_binary + gdppc + polity2_binary * gdppc,
  data = d_merged)
summary(m_3b)

##
## Call:
## lm(formula = fdi ~ polity2_binary + gdppc + polity2_binary *

```

```
##      gdppc, data = d_merged)
##
## Residuals:
##      Min      1Q    Median      3Q      Max
## -1.372e+11 -6.150e+09 -4.277e+09 -2.075e+09  6.804e+11
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)      6.055e+09  4.535e+09   1.335  0.18250
## polity2_binarydemocracy -4.080e+09  5.460e+09  -0.747  0.45534
## gdppc              8.200e+04  2.716e+05   0.302  0.76283
## polity2_binarydemocracy:gdppc  9.251e+05  2.983e+05   3.102  0.00204 **
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 4.554e+10 on 465 degrees of freedom
## (23 observations deleted due to missingness)
## Multiple R-squared:  0.1325, Adjusted R-squared:  0.1269
## F-statistic: 23.67 on 3 and 465 DF,  p-value: 2.859e-14
```

Intercept of dictatorship line: 6.0550101×10^9 Intercept of democracy line: 1.9751839×10^9

Slope of dictatorship line: 8.1996599×10^4 Slope of democracy line: 1.0070961×10^6

2.4 Demonstrating substantive meaning of coefficients

In research, we usually have to demonstrate the substantive meaning of our regression result. A common way to do it is to give the estimated outcome for a “typical” country, varying one important factor.

For example, imagine that we have a country with median gdppc. What would be its FDI if it were a 1) dictatorship and 2) democracy, holding gdppc at the median value?

Hint: You could either calculate using regression formula, or feed `newdata` to `predict`

Solution

FDI for democracy with median gdppc

```
newdata <- data.frame(polity2_binary = factor("democracy"),
                      gdppc = median(d_merged$gdppc, na.rm=TRUE))

# Using regression formula
m_3b$coefficients['(Intercept)'] + m_3b$coefficients['polity2_binarydemocracy'] +
  m_3b$coefficients['gdppc'] * median(d_merged$gdppc, na.rm=TRUE) +
  m_3b$coefficients['polity2_binarydemocracy:gdppc'] * median(d_merged$gdppc, na.rm=TRUE)
```

```
## (Intercept)
## 6067766617

# Using predict
predict(m_3b, newdata = newdata)

##          1
## 6067766617
```

FDI for dictatorship with median gdppc

```
newdata <- data.frame(polity2_binary = factor("dictatorship"),
                      gdppc = median(d_merged$gdppc, na.rm=TRUE))
predict(m_3b, newdata = newdata)

##          1
## 6388223468
```