Homework 4 Solutions

Gross domestic product (GDP) is a measure of the total market value of all goods and services produced in a given country in a given year. The percentage growth rate of GDP in year t is

$$100 \times \left(\frac{GDP_{t+1} - GDP_t}{GDP_t}\right) - 100$$

An important claim in economics is that the rate of GDP growth is closely related to the level of government debt, specifically with the ratio of the government's debt to the GDP. The file debt.csv on the class website contains measurements of GDP growth and of the debt-to-GDP ratio for twenty countries around the world, from the 1940s to 2010. Note that not every country has data for the same years, and some years in the middle of the period are missing data for some countries but not others. Throughout, use 3 significant digits for numerical answers!! (That is, signif(mydat,3) is your friend).

```
debt <- read.csv("debt.csv", as.is = TRUE)</pre>
dim(debt)
## [1] 1171
               4
head(debt)
##
       Country Year
                       growth
## 1 Australia 1946 -3.557951 190.41908
## 2 Australia 1947 2.459475 177.32137
## 3 Australia 1948
                     6.437534 148.92981
## 4 Australia 1949
                     6.611994 125.82870
## 5 Australia 1950
                     6.920201 109.80940
## 6 Australia 1951 4.272612 87.09448
```

- 1. Calculate the average GDP growth rate for each country (averaging over years). This is a classic split/apply/combine problem, and you will use daply() to solve it.
 - a. Begin by writing a function, mean.growth(), that takes a data frame as its argument and returns the mean of the growth column of that data frame.

```
library(plyr)
mean.growth <- function(country.df) {
   return(signif(mean(country.df$growth), 3))
}

b. Use `daply()` to apply `mean.growth()` to each country in `debt`. You should not need to use a loop country.avgs <- daply(debt, .(Country), mean.growth) country.avgs["Australia"]

## Australia
## 3.72</pre>
```

```
## Netherlands
## 3.03
```

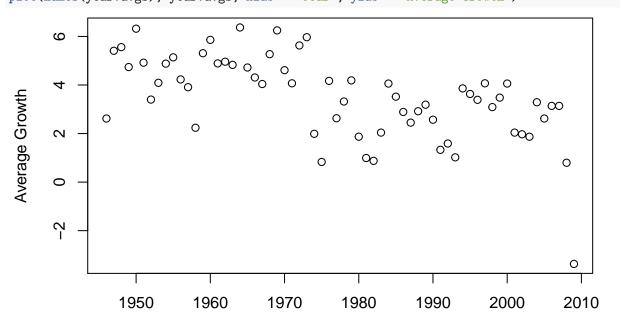
country.avgs["Netherlands"]

2. Using the same instructions as problem 1, calculate the average GDP growth rate for each year (now averaging over countries). (The average growth rates for 1972 and 1989 should be 5.63 and 3.19, respectively.) Make a plot of the growth rates (y-axis) versus the year (x-axis). Make sure the axes are labeled appropriately.

```
year.avgs <- daply(debt, .(Year), mean.growth)
year.avgs["1972"]

## 1972
## 5.63
year.avgs["1989"]

## 1989
## 3.19
plot(names(year.avgs), year.avgs, xlab = "Year", ylab = "Average Growth")</pre>
```



- 3. The function cor(x,y) calculates the correlation coefficient between two vectors x and y.
 - a. Calculate the correlation coefficient between GDP growth and the debt ratio over the whole data set (all countries, all years). Your answer should be -0.1995.

Year

```
set (all countries, all years). Your answer should be -0.1995.

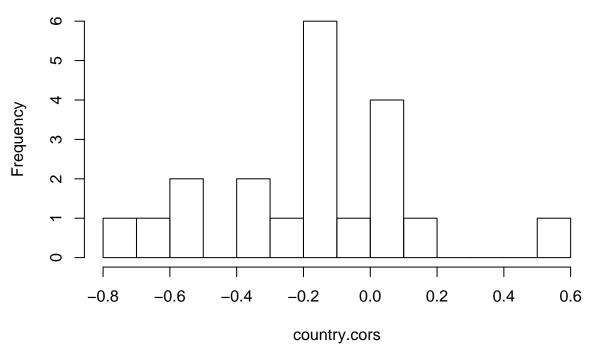
signif(cor(debt$growth, debt$ratio), 3)

## [1] -0.199

b. Compute the correlation coefficient separately for each country, and plot a histogram of these coeff cor.calc <- function(country.df) {
   return(signif(cor(country.df$growth, country.df$ratio), 3))
} country.cors <- daply(debt, .(Country), cor.calc)
mean(country.cors)

## [1] -0.177766
hist(country.cors,breaks=10)</pre>
```

Histogram of country.cors

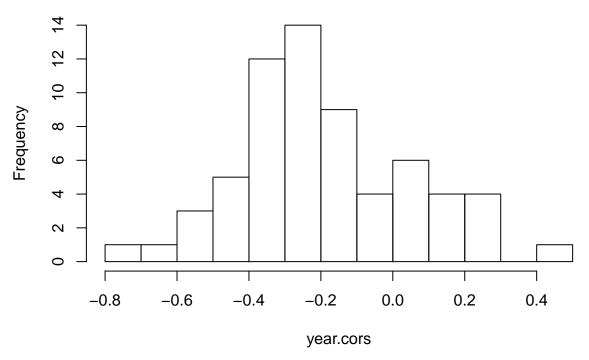


c. Calculate the correlation coefficient separately for each year, and plot a histogram of these coeffi
year.cors <- daply(debt, .(Year), cor.calc)
mean(year.cors)</pre>

[1] -0.190525

hist(year.cors,breaks=10)

Histogram of year.cors



d. Are there any countries or years where the correlation goes against the general trend? sort(country.cors)

```
Portugal
                                                                              US
##
         Japan
                      Italy
                                 Germany
                                               France
##
     -0.702000
                  -0.645000
                               -0.576000
                                            -0.502000
                                                         -0.352000
                                                                      -0.341000
##
       Austria Netherlands
                                                            Sweden
                                                                        Ireland
                                 Belgium
                                              Denmark
##
     -0.253000
                  -0.199000
                               -0.192000
                                            -0.168000
                                                         -0.161000
                                                                      -0.140000
##
             UK
                     Greece
                                 Finland
                                            Australia
                                                                           Spain
                                                            Canada
##
     -0.137000
                  -0.093500
                                0.000581
                                             0.025200
                                                          0.075000
                                                                       0.081400
## New Zealand
                     Norway
##
      0.161000
                   0.563000
```

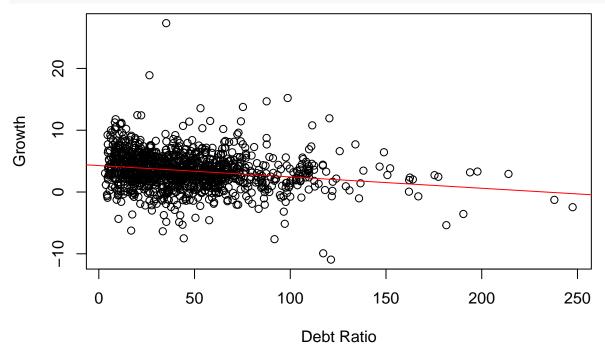
sort(year.cors)

1960 1957 1946 1961 1970 1956 1958 1985 -0.75500 -0.62000 -0.53900 -0.51200 -0.50400 -0.45800 -0.45400 -0.44900## 1993 1983 1964 1986 ## 1979 1951 1962 1996 ## -0.42900 -0.41600 -0.38300 -0.37200 -0.36200 -0.36100 -0.35800 -0.35700 ## 2002 2007 1948 2005 1965 1966 1959 1967 ## -0.34900 -0.34400 -0.34000 -0.31400 -0.31100 -0.31100 -0.28500 -0.27800 ## 1952 1954 1947 1998 1999 1969 2001 ## $-0.27700 \ -0.27500 \ -0.27400 \ -0.26500 \ -0.25800 \ -0.25000 \ -0.23800 \ -0.22700$ ## 1994 1953 2009 1949 1972 2006 1968 -0.22400 -0.20500 -0.20500 -0.20000 -0.19600 -0.19600 -0.18100 -0.17100 ## ## 2004 1984 2000 1980 1997 2008 1987 2003 ## -0.17100 -0.15600 -0.13400 -0.12700 -0.11100 -0.09450 -0.06890 -0.06790 1992 1971 1981 1950 1995 1989 1988 ## 1973 0.07970 ## -0.00222 0.00872 0.03040 0.03980 0.05190 0.06640 0.11400 1974 ## 1963 1990 1977 1991 1982 1975 1978 0.12800 0.15600 0.16400 0.20200 0.23900 0.26000 0.27100 0.43100 *Solution* Norway stands out for having a particularly large, positive correlation. 1978 seems to be p

4. Fit a linear model of overall growth on the debt ratio, using lm(). Report the intercept and slope. Make a scatter-plot of overall GDP growth (vertical) against the overall debt ratio (horizontal). Add a line to your scatterplot from question 5 showing the fitted regression line.

```
plot(debt$ratio, debt$growth, xlab = "Debt Ratio", ylab = "Growth")
lm0 <- lm(debt$growth ~ debt$ratio)
lm0$coef

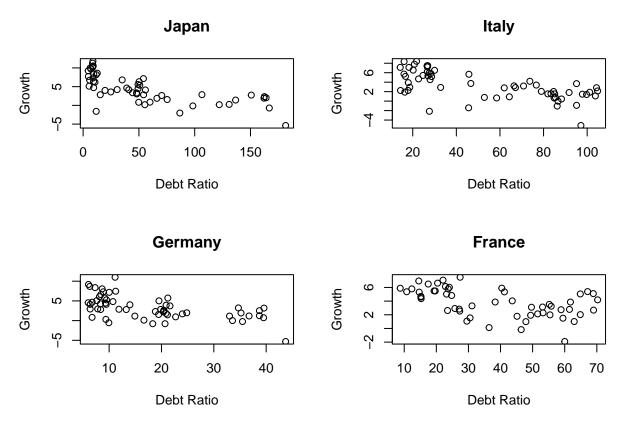
## (Intercept) debt$ratio
## 4.27929049 -0.01835518
abline(lm0, col = "red")</pre>
```



5. There should be four countries with a correlation smaller than -0.5. Separately, plot GDP growth versus debt ratio from each of these four countries and put the country names in the titles. This should be four plots. Call par(mfrow=c(2,2)) before plotting so all four plots will appear in the same figure.

(Think about what this shows: individual relationships at the country level are sometimes concealed or "smudged out" when data is aggregated over *all* groups (countries). This conveys the importance of careful analysis at a more granular group level, when such groupings are available!)

```
par(mfrow=c(2,2))
four.countries <- names(sort(country.cors))[1:4]
for (i in 1:4) {
   plot(debt$ratio[debt$Country == four.countries[i]], debt$growth[debt$Country == four.countries[i]], x
}</pre>
```



- 6. Some economists claim that high levels of government debt cause slower growth. Other economists claim that low economic growth leads to higher levels of government debt. The data file, as given, lets us relate this year's debt to this year's growth rate; to check these claims, we need to relate current debt to future growth.
 - a. Create a new dataframe which just contains the rows of debt for France, but contains all those rows. It should have 54 rows and 4 columns. Note that some years are missing from the middle of this data set.

```
debt.fr <- debt[debt$Country == "France", ]
dim(debt.fr)</pre>
```

[1] 54 4

b. Create a new column in your dataframe for France, `next.growth`, which gives next year's growth _if_

```
next.growth <- function(year, country.df) {
   if(any(country.df$Year == (year + 1))) {
      return(country.df$growth[country.df$Year == (year + 1)])
   } else {
      return(NA)
   }
}
debt.fr$next.growth <- sapply(debt.fr$Year, next.growth, debt.fr)
debt.fr$next.growth[debt.fr$Year == 1971]</pre>
```

[1] 5.885827

```
debt.fr$next.growth[debt.fr$Year == 1972]
## [1] NA
You can also use a loop
next.growth.loop <- function(country.df) {</pre>
  my.years <- country.df$Year</pre>
  next.growth <- rep(NA,length(my.years))</pre>
  counter <- 1
  for (year in my.years) {
    if (any(my.years==year+1)) {next.growth[counter] <- country.df$growth[counter+1]}</pre>
    else (next.growth[counter] <- NA)</pre>
    counter <- counter+1
  }
  country.df$next.growth <- next.growth</pre>
  return(country.df)
}
# Test the loop on France
debt.fr <- debt[debt$Country == "France", ]</pre>
debt.fr <- next.growth.loop(debt.fr)</pre>
debt.fr$next.growth[debt.fr$Year == 1971]
## [1] 5.885827
debt.fr$next.growth[debt.fr$Year == 1972]
```

[1] NA

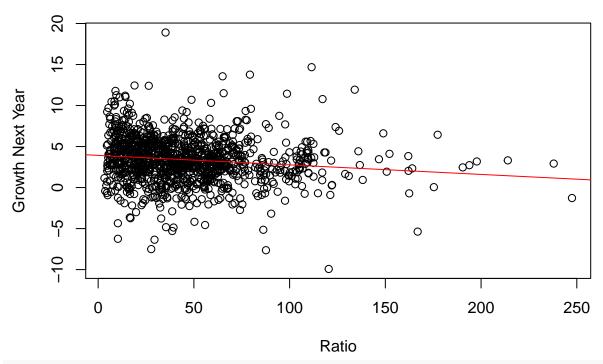
7. Add a next.growth column, as in question 4, to the *whole* of the debt data frame. Make sure that you do not accidentally put the first growth value for one country as the next.growth value for another. (The next.growth for France in 2009 should be NA, not 9.167.) *Hints*: Write a function to encapsulate what you did in question 4, and apply it using ddply().

```
next.growth.all <- function(country.df) {
   country.df$next.growth <- sapply(country.df$Year, next.growth, country.df)
   return(country.df)
}
debt <- ddply(debt, .(Country), next.growth.all)
debt$next.growth[debt$Country == "France" & debt$Year == 2009]</pre>
```

[1] NA

8. Make a scatter-plot of next year's GDP growth against this year's debt ratio. Linearly regress next year's growth rate on the current year's debt ratio, and add the line to the plot. Report the intercept and slope to reasonable precision. How do they compare to the regression of the current year's growth on the current year's debt ratio?

```
plot(debt$ratio, debt$next.growth, xlab = "Ratio", ylab = "Growth Next Year")
lm1 <- lm(debt$next.growth ~ debt$ratio)
abline(lm1, col = "red")</pre>
```



lm0\$coef

```
## (Intercept) debt$ratio
## 4.27929049 -0.01835518
```

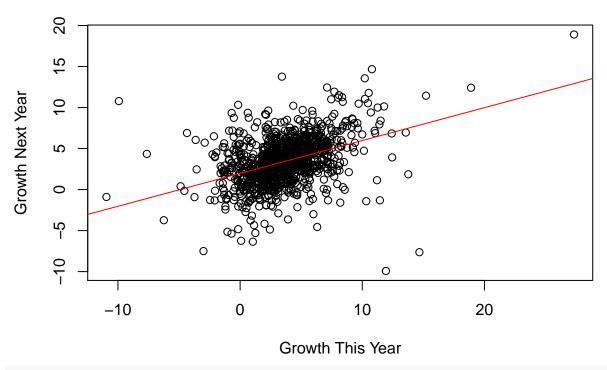
lm1\$coef

```
## (Intercept) debt$ratio
## 3.92472155 -0.01160842
```

The two regressions are similar. Both have a slightly negative slope with an intercept somewhere around 4.

9. Make a scatter-plot of next year's GDP growth against the current year's GDP growth. Linearly regress next year's growth on this year's growth, and add the line to the plot. Report the coefficients. Can you tell, from comparing these two simple regressions (from the current question, and question 9), whether current growth or current debt is a better predictor of future growth?

```
plot(debt$growth, debt$next.growth, xlab = "Growth This Year", ylab = "Growth Next Year")
lm2 <- lm(debt$next.growth ~ debt$growth)
abline(lm2, col = "red")</pre>
```



lm2\$coef

(Intercept) debt\$growth ## 1.9710648 0.4006517

I can't tell! Though this regression has a slightly larger value of \mathbb{R}^2 and slightly more significant coefficients.