

Lab 3: Panel Models

US Traffic Fatalities: 1980 - 2004

Contents

1	U.S. traffic fatalities: 1980-2004	3
2	(30 points, total) Build and Describe the Data	3
2.1	(5 points) Load the data and produce useful features.	3
2.1.0.1	Status complete: TODO Check for completeness	4
2.2	(5 points) Provide a description of the basic structure of the dataset.	6
2.2.0.1	Status not-started	6
2.3	(20 points) Conduct a very thorough EDA	6
2.3.0.1	Status Complete: Check for completeness	6
2.3.1	Average and state wide trends	6
2.3.1.1	Status Incomplete: Refer to graphs and add descriptive text. Does the dependent variable require a transformation?	6
2.3.2	Description of variables	10
2.3.2.1	Status incomplete: TODO: Add figure number references to each graph . . .	10
2.3.2.2	Factor variables	10
2.3.2.3	Description of continuous variables	11
3	(15 points) Preliminary Model	12
3.0.0.1	Status complete: TODO Check if anything is missing	12
4	(15 points) Expanded Model	13
4.0.0.1	Status complete: TODO Check if anything is missing	13
4.1	Transformation(treatment) of variables	13
4.1.1	Pooled OLS model creation	14
4.2	Interpretation of results	15
4.2.1	How are the blood alcohol variables defined? Interpret the coefficients that you estimate for this concept.	15
4.2.2	Do <i>per se laws</i> have a negative effect on the fatality rate?	16
4.2.3	Does having a primary seat belt law reduce fatality rates?	16

5	(15 points) State-Level Fixed Effects	16
5.0.0.1	Status complete: TODO Check if anything is missing	16
5.1	Re-estimate the Expanded Model using fixed effects at the state level.	16
5.2	Interpretation of model	17
5.2.1	What do you estimate for coefficients on the blood alcohol variables? How do the coefficients on the blood alcohol variables change, if at all?	18
5.2.2	What do you estimate for coefficients on per se laws? How do the coefficients on per se laws change, if at all?	18
5.2.3	What do you estimate for coefficients on primary seat-belt laws? How do the coefficients on primary seatbelt laws change, if at all?	18
5.3	Model assumptions	19
6	(10 points) Consider a Random Effects Model	19
6.0.0.1	Status incomplete	19
6.0.1	Random effect model assumptions	19
6.0.2	Estimating the random effect model	19
6.0.3	We conduct a Hausman test for random vs. fixed effects using phptest.	21
7	(10 points) Model Forecasts	21
7.0.0.1	Status not-started	21
8	(5 points) Evaluate Error	22
8.0.1	Analysis of residuals	22
8.0.1.1	Status incomplete	22

1 U.S. traffic fatalities: 1980-2004

In this lab, we are asking you to answer the following **causal** question:

“Do changes in traffic laws affect traffic fatalities?”

To answer this question, please complete the tasks specified below using the data provided in `data/driving.Rdata`. This data includes 25 years of data that cover changes in various state drunk driving, seat belt, and speed limit laws.

Specifically, this data set contains data for the 48 continental U.S. states from 1980 through 2004. Various driving laws are indicated in the data set, such as the alcohol level at which drivers are considered legally intoxicated. There are also indicators for “per se” laws—where licenses can be revoked without a trial—and seat belt laws. A few economics and demographic variables are also included. The description of the each of the variables in the dataset is also provided in the dataset.

```
load(file="./data/driving.RData")

## please comment these calls in your work
#glimpse(data)
#desc
```

2 (30 points, total) Build and Describe the Data

2.1 (5 points) Load the data and produce useful features.

```
# For the fractions, we are taking the majority as a speed limit
# We skipped year_of_observation since there a year column which aligns with dx
df <- data %>%
  mutate(speed_limit = ifelse(sl55 >= 0.5, '55',
                              ifelse(sl65 >= 0.5, '65',
                              ifelse(sl70 >= 0.5, '70',
                              ifelse(sl75 >= 0.5, '75',
                              ifelse(slnone >= 0.5, 'none', '0')
                              ))))) %>%
  mutate(speed_limit=factor(speed_limit,
                             levels=c('55', '65', '70', '75', 'none')),
         blood_alcohol_limit_10 = ifelse(bac10 >= 0.5, 1, 0),
         blood_alcohol_limit_08 = ifelse(bac08 >= 0.5, 1, 0)) %>%
  mutate(bac=ifelse(blood_alcohol_limit_10==1, '10',
                    ifelse(blood_alcohol_limit_08==1, '8', 'none')) %>%
  mutate(bac=factor(bac, levels=c('none', '10', '8'))) %>%
  select(!c((sl55:slnone), (d80:d04), bac10, bac08)) %>% # Excluding
  rename(minimum_drinking_age = minage, zero_tolerance_law = zerotol,
         graduated_drivers_license_law = gdl, per_se_law = perse,
         total_fatalities = totfat, nighttime_fatalities = nghtfat,
         weekend_fatalities = wkndfat, total_fatalities_per_100M_miles = totfatpvm,
         nighttime_fatalities_per_100M_miles = nghtfatpvm,
         weekend_fatalities_per_100M_miles = wkndfatpvm,
         state_population = statepop, total_fatalities_rate = totfatrte,
```

```

nighttime_fatalities_rate = nghtfatrte,
weekend_fatalities_rate = wkndfatrte,
vehicle_miles_traveled = vehicmiles, unemployment_rate = unem,
population_aged_14_to_24_rate = perc14_24,
speed_limit_70_plus = sl70plus,
seat_belt = seatbelt,
primary_seatbelt_law = sbprim, secondary_seatbelt_law = sbsecon,
miles_driven_per_capita = vehicmilespc) %>%
mutate(speed_limit_70_plus =
  ifelse(speed_limit_70_plus>0.5, 1, 0)
) %>%
mutate(seat_belt_law =
  ifelse(seat_belt==0, 'none',
  ifelse(seat_belt==2, 'secondary',
  ifelse(seat_belt==1, 'primary', 'na')))) %>%
mutate(seat_belt_law=factor(seat_belt_law,
  levels=c('none', 'secondary', 'primary')),
) %>%
mutate(per_se_law=round(per_se_law, 0)) %>%
mutate(per_se_law=factor(per_se_law, levels=c(0, 1))) %>%
mutate(log_total_fatalities_rate = log10(total_fatalities_rate))

# Adding states to the dataframe
state_df <- data.frame("index" = 1:51,
  "state_name" = sort(c(state.name, "District of Columbia")))
main_df <- merge(df, state_df, by.x = 'state', by.y = 'index')

pdata <- pdata.frame(main_df, index=c("state", "year"))
head(main_df)

```

2.1.0.1 Status complete: TODO Check for completeness

```

## state year seat_belt minimum_drinking_age zero_tolerance_law
## 1 1 1980 0 18 0
## 2 1 1981 0 18 0
## 3 1 1982 0 18 0
## 4 1 1983 0 18 0
## 5 1 1984 0 18 0
## 6 1 1985 0 20 0
## graduated_drivers_license_law per_se_law total_fatalities
## 1 0 0 940
## 2 0 0 933
## 3 0 0 839
## 4 0 0 930
## 5 0 0 932
## 6 0 0 882
## nighttime_fatalities weekend_fatalities total_fatalities_per_100M_miles
## 1 422 236 3.20
## 2 434 248 3.35
## 3 376 224 2.81
## 4 397 223 3.00
## 5 421 237 2.83

```

## 6	358	224	2.51
##	nighttime_fatalities_per_100M_miles	weekend_fatalities_per_100M_miles	
## 1	1.437	0.803	
## 2	1.558	0.890	
## 3	1.259	0.750	
## 4	1.281	0.719	
## 5	1.278	0.720	
## 6	1.019	0.637	
##	state_population	total_fatalities_rate	nighttime_fatalities_rate
## 1	3893888	24.14	10.84
## 2	3918520	24.07	11.08
## 3	3925218	21.37	9.58
## 4	3934109	23.64	10.09
## 5	3951834	23.58	10.65
## 6	3972527	22.20	9.01
##	weekend_fatalities_rate	vehicle_miles_traveled	unemployment_rate
## 1	6.06	29.37500	8.8
## 2	6.33	27.85200	10.7
## 3	5.71	29.85765	14.4
## 4	5.67	31.00000	13.7
## 5	6.00	32.93286	11.1
## 6	5.64	35.13944	8.9
##	population_aged_14_to_24_rate	speed_limit_70_plus	primary_seatbelt_law
## 1	18.9	0	0
## 2	18.7	0	0
## 3	18.4	0	0
## 4	18.0	0	0
## 5	17.6	0	0
## 6	17.3	0	0
##	secondary_seatbelt_law	miles_driven_per_capita	speed_limit
## 1	0	7543.874	55
## 2	0	7107.785	55
## 3	0	7606.622	55
## 4	0	7879.802	55
## 5	0	8333.562	55
## 6	0	8845.614	55
##	blood_alcohol_limit_10	blood_alcohol_limit_08	bac seat_belt_law
## 1	1	0 10	none
## 2	1	0 10	none
## 3	1	0 10	none
## 4	1	0 10	none
## 5	1	0 10	none
## 6	1	0 10	none
##	log_total_fatalities_rate	state_name	
## 1	1.382737	Alabama	
## 2	1.381476	Alabama	
## 3	1.329805	Alabama	
## 4	1.373647	Alabama	
## 5	1.372544	Alabama	
## 6	1.346353	Alabama	

2.2 (5 points) Provide a description of the basic structure of the dataset.

2.2.0.1 Status not-started

TODO TODO

2.3 (20 points) Conduct a very thorough EDA

2.3.0.1 Status Complete: Check for completeness

TODO TODO

2.3.1 Average and state wide trends

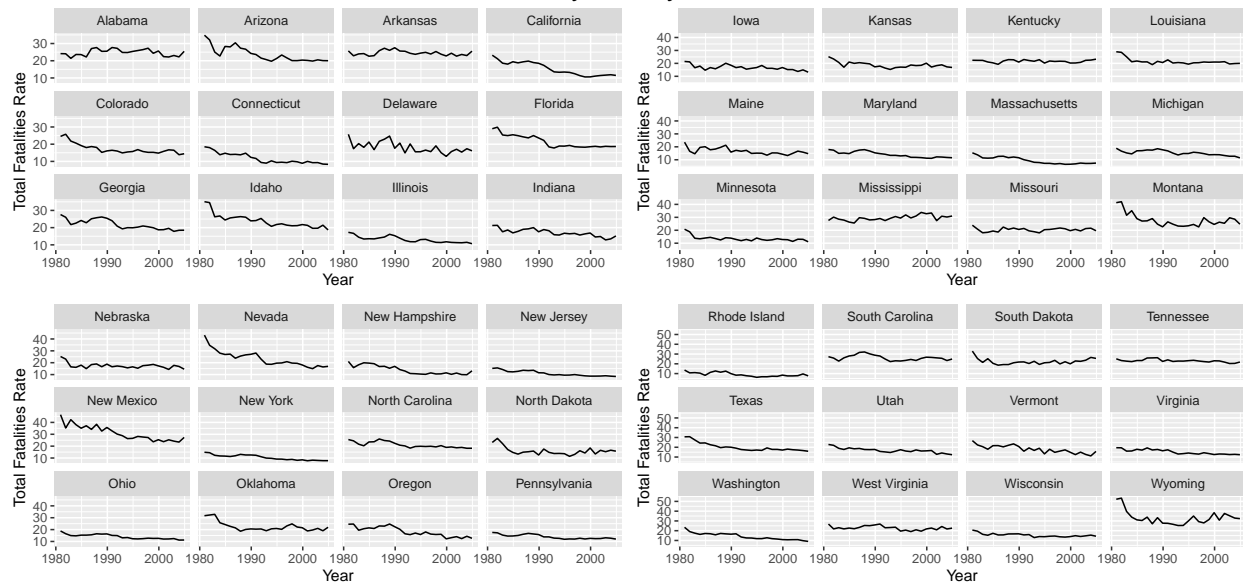
2.3.1.1 Status Incomplete: Refer to graphs and add descriptive text. Does the dependent variable require a transformation?

Average mean fatality rate across US

Fatality rate is going down



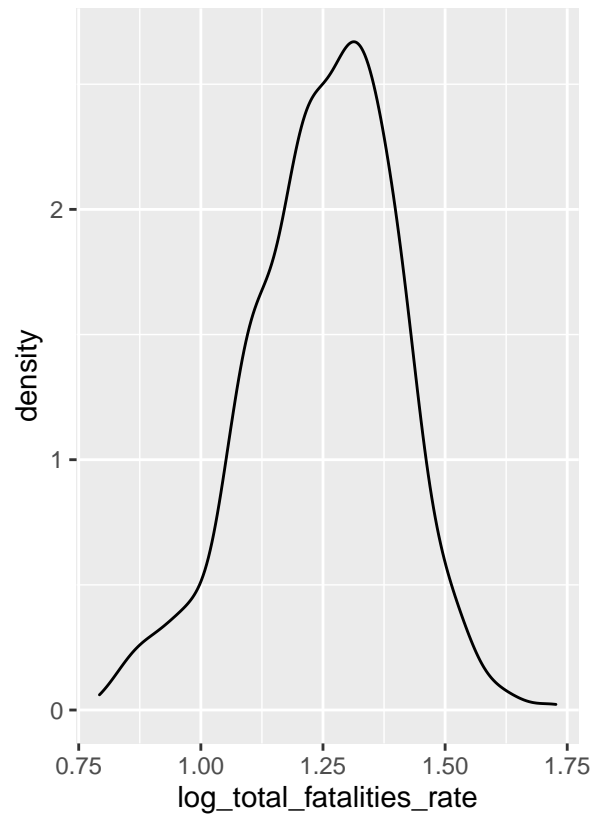
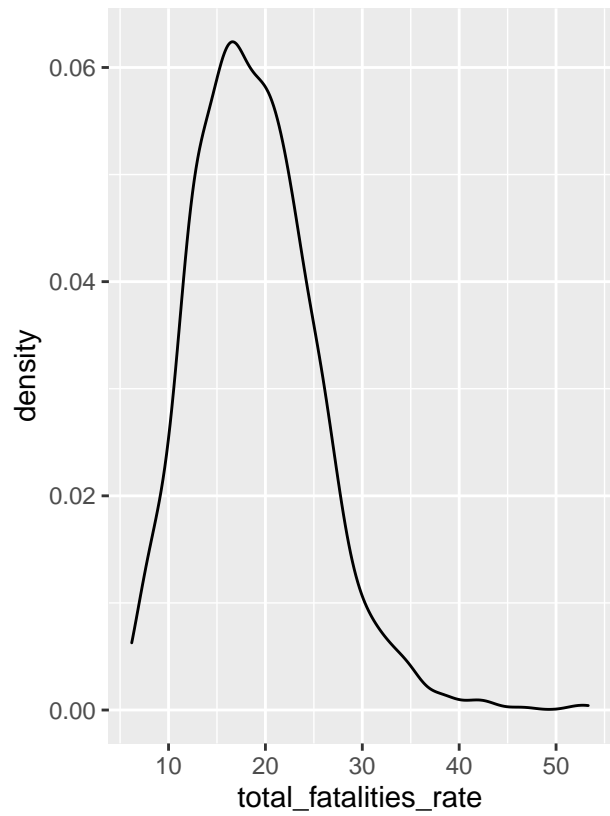
Fatality rate by states



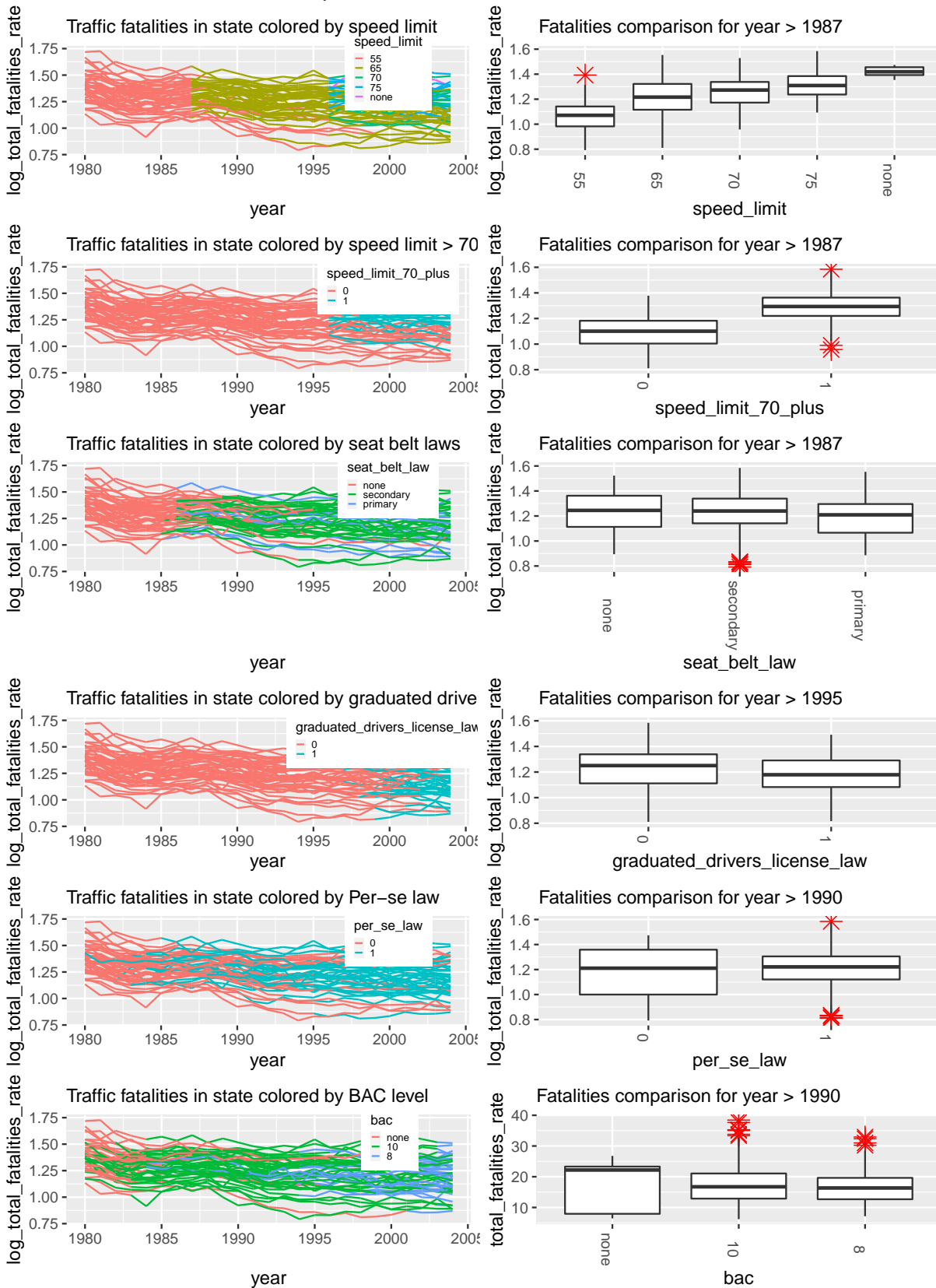
‘For most states, fatality rates go down over the years, but some states like Alabama and Arkansas do not show many changes. Surprisingly, Mississippi has an increase in the fatality rate.’

```
without.tran <- main_df %>% ggplot(aes(x=total_fatalities_rate)) +  
  geom_density()
```

```
with.tran <- main_df %>% ggplot(aes(x=log_total_fatalities_rate)) +  
  geom_density()  
  
without.tran | with.tran
```



Time series and distribution comparisons of factor variables



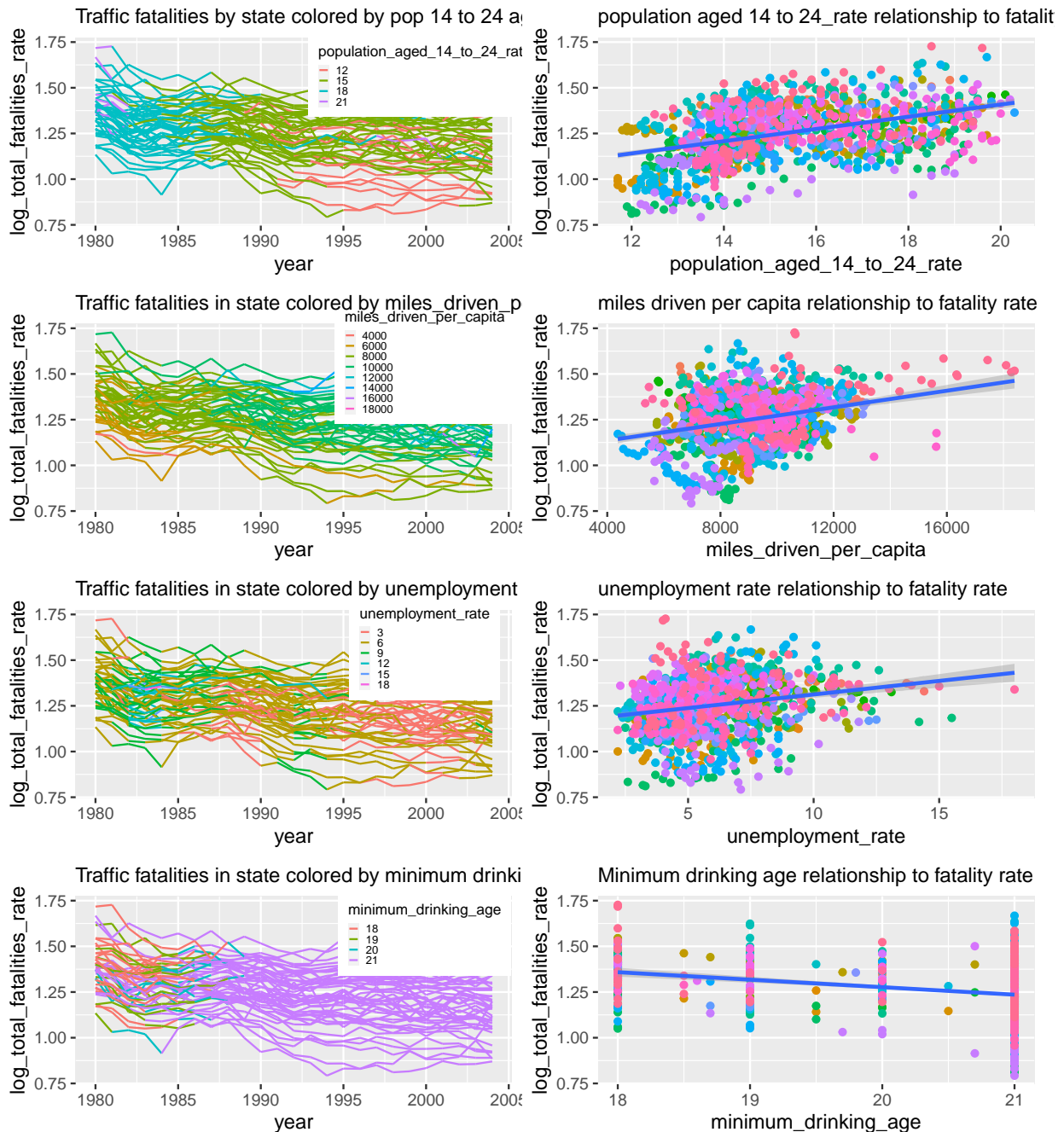
2.3.2 Description of variables

2.3.2.1 Status incomplete: TODO: Add figure enumber references to each graph

2.3.2.2 Factor variables

- The highway speed limit was uniformly 55mph across all states before 1987. Since then, different states have adopted different speed limits. Especially in 1997, there was a significant increase in highway speeds across multiple states. The box plot compares the fatality rate across different speed limits filtered for years greater than 1987. We see that increasing speed limits are associated with increased fatality rates. As there are states with no speed limit, this variable has been treated as a factor.
- Now thresholding speed limits for greater or lower than 70mph shows a similar pattern of higher speeds associated with a higher fatality rate.
- Seat belts started to become mandatory starting mid to late 80s and today there is only one state which does not have it as mandatory. Primary laws are the strictest and allow police to ticket drivers and passengers who are not wearing a proper safety restraint, even if that is the only traffic violation they are committing. Secondary seat belt laws, on the other hand, do not grant law enforcement officials the right to ticket drivers or passengers for failing to wear a safety restraint unless another traffic violation has occurred. There are 15 states with secondary seat belt laws. Source: <https://www.cooper-law-firm.com/what-is-the-difference-between-primary-and-secondary-seat-belt-laws/>.
- The graduated drivers licence law was started to be introduced in the late 90's. The box plot, which has been filtered for years greater than 1995, suggests that even for that time frame, there is a reduction in fatality rate between the two groups.
- Some states had Per-Se laws before the start of the data in 1980 and some still did not have Per-Se laws in 2004. There is a gradual increase in the adoption of the law from 1980 to about the 2000s. Surprisingly, there is an increase in fatality rates in comparison of data with PerSe law as compared to without.
- Most states had adopted a BAC limit by the mid 80s with two states choosing a limit only in 2002.

Time series and correlation comparison of continuous variables



2.3.2.3 Description of continuous variables

- There is a decrease in the percentage of 14 to 24 year olds in the population over time. This is correlated to the decrease in the fatalities during that time period.
- Miles driven per capita also has a positive relationship with fatality rate. An increase in miles driven is associated with an increase in fatality rate
- There is an increase in failure rate with an increase in unemployment rate.
- The minimum drinking age has been 21 in most states since the late 80s. There is a general decrease

in fatality rate with an increased minimum drinking age but there also has been a general decrease in fatality rates during the time period when the age limits were changed.

```
# traffic laws that we are exploring are seat_belt, minimum_drinking_age,
# zero_tolerance_law, graduated_drivers_license_law, per_se_law, speed_limit,
# speed_limit_70_plus, primary_seatbelt_law, secondary_seatbelt_law,
# blood_alcohol_limit_10, blood_alcohol_limit_08
```

3 (15 points) Preliminary Model

```
# Pooled OLS model
pooled_ols <- plm(log_total_fatalities_rate ~ year, data = pdata,
                 index = c("state", "year"),
                 effect = "individual", model = "pooling")
summary(pooled_ols)
```

3.0.0.1 Status complete: TODO Check if anything is missing

```
## Pooling Model
##
## Call:
## plm(formula = log_total_fatalities_rate ~ year, data = pdata,
##      effect = "individual", model = "pooling", index = c("state",
##      "year"))
##
## Balanced Panel: n = 48, T = 25, N = 1200
##
## Residuals:
##      Min.      1st Qu.      Median      3rd Qu.      Max.
## -0.4183307 -0.0961280  0.0043659  0.1008470  0.3770988
##
## Coefficients:
##              Estimate Std. Error t-value Pr(>|t|)
## (Intercept)  1.387907   0.020400  68.0351 < 2.2e-16 ***
## year1981    -0.034213   0.028850  -1.1859  0.2359037
## year1982    -0.086671   0.028850  -3.0042  0.0027191 **
## year1983    -0.102159   0.028850  -3.5411  0.0004141 ***
## year1984    -0.098087   0.028850  -3.3999  0.0006966 ***
## year1985    -0.105539   0.028850  -3.6582  0.0002652 ***
## year1986    -0.085474   0.028850  -2.9627  0.0031107 **
## year1987    -0.086297   0.028850  -2.9912  0.0028363 **
## year1988    -0.082018   0.028850  -2.8429  0.0045473 **
## year1989    -0.107768   0.028850  -3.7355  0.0001963 ***
## year1990    -0.116324   0.028850  -4.0320  5.886e-05 ***
## year1991    -0.149274   0.028850  -5.1742  2.690e-07 ***
## year1992    -0.174714   0.028850  -6.0560  1.875e-09 ***
## year1993    -0.174832   0.028850  -6.0601  1.830e-09 ***
## year1994    -0.177185   0.028850  -6.1416  1.116e-09 ***
## year1995    -0.167169   0.028850  -5.7945  8.794e-09 ***
## year1996    -0.173498   0.028850  -6.0138  2.416e-09 ***
```

```
## year1997    -0.167622    0.028850 -5.8102 8.028e-09 ***
## year1998    -0.177863    0.028850 -6.1651 9.666e-10 ***
## year1999    -0.180014    0.028850 -6.2397 6.108e-10 ***
## year2000    -0.189762    0.028850 -6.5776 7.181e-11 ***
## year2001    -0.189010    0.028850 -6.5515 8.500e-11 ***
## year2002    -0.185324    0.028850 -6.4238 1.927e-10 ***
## year2003    -0.190996    0.028850 -6.6204 5.437e-11 ***
## year2004    -0.194794    0.028850 -6.7520 2.286e-11 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Total Sum of Squares:    26.854
## Residual Sum of Squares: 23.471
## R-Squared:    0.12598
## Adj. R-Squared: 0.10813
## F-statistic: 7.05692 on 24 and 1175 DF, p-value: < 2.22e-16
```

- ‘Starting from a linear model will give us an easy and intuitively clearer overall pattern over the years via the coefficients on the year dummy variables (or instead by adding group ids as dummy variables) using controlling for any variable of interest, if any. Later on, when we perform panel data analysis, we can compare it against the linear model, justify the result and keep building the intuition on top of the linear model.’
- ‘This model explains the log of fatality rate over different years compared to the baseline year, which is 1980. All coefficients except for the 1981 year are statistically significant. From this model, we learn that in all years following the baseline up to 2004, the fatality rate goes down compared to the baseline year 1980. The decline increases as the year increases but not consistently, as seen, for example, between 1986 - 1988, where the fatality rate increases compared to previous years. However, it returns to the decline track/trend in 1989. The higher, in absolute value, coefficients over the years means that driving becomes safer as the fatality rate keeps decreasing. For example, the fatality rate, which was 25.5 as of 1980, became 19.5 in 1990, 16.8 in 2000, and 16.7 in 2004, showing that driving over the years has become safer. In other words, around 8.8 fewer people are predicted to get traffic fatalities out of 100,000 people in 2004 than in 1980.’
- ‘We are ignoring unobserved Heterogeneity and the group structure by taking each entry as a separate observation. Because of that, residuals generally correlate across time and have heteroskedasticity across and/or within groups. Heteroscedastic residuals are a violation of the OLS Homoscedasticity assumption, which will make it difficult to trust the standard error. As a result, the confidence interval can not be trusted as it can be too wide or narrow. Also, the independence assumption (no autocorrelation) is violated since we did not accommodate the lag/trend component, which makes the OLS estimates to be unreliable; in other words, our OLS estimator is not the Best Linear Unbiased Estimator.’

4 (15 points) Expanded Model

4.0.0.1 Status complete: TODO Check if anything is missing

4.1 Transformation(treatment) of variables

- As described in a previous section, a log transformation was performed on the response variable to make the distribution closer to normal.

- States where highway speeds were made over 70mph during the middle of the year contained a fractional value for that year. The fraction was threshold at 0.5 to make this a binary variable. As there is no meaningful interpretation for this variable as a continuous value, this threshold was necessary to convert it into a factor.
- A state can have a primary, secondary or no seat belt laws. The `seat_belt_law` reflects these three factors combined into one variable.
- Graduated drivers licence law and per-se law parameters both have fractional value for years where the law was implemented mid-year. The fractions was threshold at 0.5 to make this a binary variable. As there is no meaningful interpretation for these variables as a continuous value, this threshold was necessary to convert them into a factor.
- Although Blood alcohol content (BAC) levels of 8PPM or 10PPM lends itself to a numeric interpretation, there is no numeric value associated with no BAC limit. For this reason BAC has been treated as a factor variable with levels none, 10 and 8.
- None of the continuous variables, namely, unemployment rate, miles drive per-capita, rate of 14 to 24 aged people in population required any transformation. This can be seen in the correlation plots in Figure y where their relationship to log fatality rates appear linear.

4.1.1 Pooled OLS model creation

```
expanded.ols.data <- main_df %>% select(c(log_total_fatalities_rate, bac, per_se_law,
    seat_belt_law, graduated_drivers_license_law,
    population_aged_14_to_24_rate, minimum_drinking_age,
    unemployment_rate, speed_limit_70_plus,
    miles_driven_per_capita, year, state
))

main_p <- pdata.frame(expanded.ols.data, index=c("state", "year"))
expanded.ols <- plm(log_total_fatalities_rate ~ year + bac +
    population_aged_14_to_24_rate + miles_driven_per_capita +
    unemployment_rate + speed_limit_70_plus + per_se_law +
    seat_belt_law + graduated_drivers_license_law,
    data = main_p,
    index = c("state", "year"),
    effect = "individual", model = "pooling")

summary(expanded.ols)
```

```
## Pooling Model
##
## Call:
## plm(formula = log_total_fatalities_rate ~ year + bac + population_aged_14_to_24_rate +
##     miles_driven_per_capita + unemployment_rate + speed_limit_70_plus +
##     per_se_law + seat_belt_law + graduated_drivers_license_law,
##     data = main_p, effect = "individual", model = "pooling",
##     index = c("state", "year"))
##
## Balanced Panel: n = 48, T = 25, N = 1200
##
## Residuals:
##      Min.      1st Qu.      Median      3rd Qu.      Max.
## -0.34758836 -0.05686774  0.00052907  0.06279643  0.25888532
##
```

```
## Coefficients:
##               Estimate Std. Error t-value Pr(>|t|)
## (Intercept)    6.5776e-01 5.5569e-02 11.8367 < 2.2e-16 ***
## year1981      -4.0821e-02 1.8599e-02 -2.1948 0.02838 *
## year1982      -1.3080e-01 1.9183e-02 -6.8182 1.477e-11 ***
## year1983      -1.5099e-01 1.9581e-02 -7.7113 2.659e-14 ***
## year1984      -1.2006e-01 1.9720e-02 -6.0885 1.545e-09 ***
## year1985      -1.3316e-01 2.0105e-02 -6.6233 5.354e-11 ***
## year1986      -1.2207e-01 2.0907e-02 -5.8386 6.819e-09 ***
## year1987      -1.3467e-01 2.1723e-02 -6.1997 7.839e-10 ***
## year1988      -1.3907e-01 2.2769e-02 -6.1076 1.376e-09 ***
## year1989      -1.7282e-01 2.3648e-02 -7.3079 5.026e-13 ***
## year1990      -1.9557e-01 2.4190e-02 -8.0846 1.550e-15 ***
## year1991      -2.4396e-01 2.4741e-02 -9.8605 < 2.2e-16 ***
## year1992      -2.8991e-01 2.5218e-02 -11.4964 < 2.2e-16 ***
## year1993      -2.8707e-01 2.5525e-02 -11.2464 < 2.2e-16 ***
## year1994      -2.8368e-01 2.5976e-02 -10.9207 < 2.2e-16 ***
## year1995      -2.7658e-01 2.6580e-02 -10.4057 < 2.2e-16 ***
## year1996      -3.3148e-01 2.7527e-02 -12.0418 < 2.2e-16 ***
## year1997      -3.4014e-01 2.7968e-02 -12.1615 < 2.2e-16 ***
## year1998      -3.6607e-01 2.8402e-02 -12.8887 < 2.2e-16 ***
## year1999      -3.7110e-01 2.8800e-02 -12.8855 < 2.2e-16 ***
## year2000      -3.8060e-01 2.9269e-02 -13.0035 < 2.2e-16 ***
## year2001      -4.0014e-01 2.9859e-02 -13.4009 < 2.2e-16 ***
## year2002      -4.1553e-01 3.0151e-02 -13.7815 < 2.2e-16 ***
## year2003      -4.2378e-01 3.0308e-02 -13.9825 < 2.2e-16 ***
## year2004      -4.2116e-01 3.0978e-02 -13.5954 < 2.2e-16 ***
## bac10         -9.9381e-03 8.7474e-03 -1.1361 0.25614
## bac8          -2.6054e-02 1.1804e-02 -2.2073 0.02749 *
## population_aged_14_to_24_rate 6.8653e-03 2.7564e-03 2.4906 0.01289 *
## miles_driven_per_capita    6.8781e-05 2.1301e-06 32.2903 < 2.2e-16 ***
## unemployment_rate         1.6720e-02 1.7501e-03 9.5533 < 2.2e-16 ***
## speed_limit_70_plus        9.6254e-02 9.7589e-03 9.8632 < 2.2e-16 ***
## per_se_law1               -6.8721e-03 6.6160e-03 -1.0387 0.29915
## seat_belt_lawsecondary      9.1036e-03 9.6452e-03 0.9438 0.34545
## seat_belt_lawprimary       -8.6245e-04 1.1029e-02 -0.0782 0.93768
## graduated_drivers_license_law -1.2209e-03 1.1840e-02 -0.1031 0.91789
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Total Sum of Squares:    26.854
## Residual Sum of Squares: 9.6296
## R-Squared:              0.64141
## Adj. R-Squared: 0.63095
## F-statistic: 61.29 on 34 and 1165 DF, p-value: < 2.22e-16
```

4.2 Interpretation of results

4.2.1 How are the blood alcohol variables defined? Interpret the coefficients that you estimate for this concept.

- In an earlier section we have defined our treatment of the BAC variable as a factor with levels 'none', '10' and '8'. In this model we note that the base level is no blood alcohol limit. We note that bac

value of 0.08% is statistically significant and 0.1% is not statistically significant. The model suggests that setting a blood alcohol limit of 0.1% is associated with a 0.98 times decrease in fatality rate as compared to no BAC limit. The model suggests that a setting a blood alcohol limit of 0.08% is associated with a 0.94 times decrease in fatality rate as compared to no BAC limit.

4.2.2 Do *per se* laws have a negative effect on the fatality rate?

- We note that per-se law is not a statistically significant parameter. The model suggests that havng a per-se law is associated with a 0.98 times decrease in fatality rate as compared to not having per-se law.

4.2.3 Does having a primary seat belt law reduce fatality rates?

- We note that the seat belt law factors are not statistically significant. We also note that the implementation of the primary law leads to a 0.998 times decrease in fatality rate which is practically insignificant.

5 (15 points) State-Level Fixed Effects

5.0.0.1 Status complete: TODO Check if anything is missing

5.1 Re-estimate the Expanded Model using fixed effects at the state level.

Model estimation for a fixed effect(within) model.

```
expanded.within <- plm(log_total_fatalities_rate ~ bac + year +
  population_aged_14_to_24_rate + miles_driven_per_capita +
  unemployment_rate + speed_limit_70_plus + per_se_law +
  seat_belt_law + graduated_drivers_license_law,
  data = main_p,
  index = c("state", "year"),
  effect = "individual", model = "within")

summary(expanded.within)

## Oneway (individual) effect Within Model
##
## Call:
## plm(formula = log_total_fatalities_rate ~ bac + year + population_aged_14_to_24_rate +
##     miles_driven_per_capita + unemployment_rate + speed_limit_70_plus +
##     per_se_law + seat_belt_law + graduated_drivers_license_law,
##     data = main_p, effect = "individual", model = "within", index = c("state",
##     "year"))
##
## Balanced Panel: n = 48, T = 25, N = 1200
##
## Residuals:
##      Min.      1st Qu.      Median      3rd Qu.      Max.
## -0.1835098 -0.0223288  0.0020551  0.0227251  0.1279559
##
```



```
## Coefficients:
##
## Estimate Std. Error t-value Pr(>|t|)
## bac10 -1.1165e-02 4.9828e-03 -2.2407 0.025241 *
## bac8 -1.6135e-02 7.1737e-03 -2.2492 0.024694 *
## year1981 -2.5783e-02 7.9252e-03 -3.2533 0.001175 **
## year1982 -5.0669e-02 8.4888e-03 -5.9689 3.204e-09 ***
## year1983 -6.1384e-02 8.7911e-03 -6.9825 4.964e-12 ***
## year1984 -7.8231e-02 8.9205e-03 -8.7698 < 2.2e-16 ***
## year1985 -8.6982e-02 9.2924e-03 -9.3605 < 2.2e-16 ***
## year1986 -6.6274e-02 9.9105e-03 -6.6872 3.585e-11 ***
## year1987 -8.1819e-02 1.0632e-02 -7.6954 3.088e-14 ***
## year1988 -9.1423e-02 1.1516e-02 -7.9387 4.941e-15 ***
## year1989 -1.2221e-01 1.2261e-02 -9.9670 < 2.2e-16 ***
## year1990 -1.2726e-01 1.2730e-02 -9.9971 < 2.2e-16 ***
## year1991 -1.4334e-01 1.3067e-02 -10.9697 < 2.2e-16 ***
## year1992 -1.6845e-01 1.3465e-02 -12.5101 < 2.2e-16 ***
## year1993 -1.7549e-01 1.3715e-02 -12.7957 < 2.2e-16 ***
## year1994 -1.8929e-01 1.4049e-02 -13.4739 < 2.2e-16 ***
## year1995 -1.8602e-01 1.4479e-02 -12.8476 < 2.2e-16 ***
## year1996 -2.0517e-01 1.5272e-02 -13.4345 < 2.2e-16 ***
## year1997 -2.1113e-01 1.5630e-02 -13.5084 < 2.2e-16 ***
## year1998 -2.3185e-01 1.5945e-02 -14.5404 < 2.2e-16 ***
## year1999 -2.3797e-01 1.6125e-02 -14.7580 < 2.2e-16 ***
## year2000 -2.5039e-01 1.6352e-02 -15.3122 < 2.2e-16 ***
## year2001 -2.4326e-01 1.6634e-02 -14.6247 < 2.2e-16 ***
## year2002 -2.2996e-01 1.6803e-02 -13.6862 < 2.2e-16 ***
## year2003 -2.3124e-01 1.6894e-02 -13.6878 < 2.2e-16 ***
## year2004 -2.4447e-01 1.7323e-02 -14.1121 < 2.2e-16 ***
## population_aged_14_to_24_rate 9.0309e-03 1.8219e-03 4.9569 8.270e-07 ***
## miles_driven_per_capita 2.6879e-05 2.1286e-06 12.6277 < 2.2e-16 ***
## unemployment_rate -1.2761e-02 1.1601e-03 -11.0002 < 2.2e-16 ***
## speed_limit_70_plus 2.3830e-02 4.9909e-03 4.7748 2.038e-06 ***
## per_se_law1 -2.3217e-02 4.2933e-03 -5.4078 7.801e-08 ***
## seat_belt_lawsecondary -6.6982e-04 4.8314e-03 -0.1386 0.889762
## seat_belt_lawprimary -2.0458e-02 6.5736e-03 -3.1122 0.001904 **
## graduated_drivers_license_law -7.3527e-03 5.6051e-03 -1.3118 0.189860
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Total Sum of Squares: 6.0213
## Residual Sum of Squares: 1.6682
## R-Squared: 0.72294
## Adj. R-Squared: 0.70287
## F-statistic: 85.8017 on 34 and 1118 DF, p-value: < 2.22e-16
```

5.2 Interpretation of model

```
##
## =====
## log_total_fatalities_rate
## Pooled OLS Within
## (1) (2)
## -----
```

```

## bac10                -0.010                -0.011**
##                      (0.009)                (0.005)
##
## bac8                  -0.026**              -0.016**
##                      (0.012)                (0.007)
##
## per_se_law1           -0.007                -0.023***
##                      (0.007)                (0.004)
##
## seat_belt_lawprimary  -0.001                -0.020***
##                      (0.011)                (0.007)
##
## N                     1,200                 1,200
## R2                    0.641                 0.723
## =====
## Notes:                ***Significant at the 1 percent level.
##                      **Significant at the 5 percent level.
##                      *Significant at the 10 percent level.

```

5.2.1 What do you estimate for coefficients on the blood alcohol variables? How do the coefficients on the blood alcohol variables change, if at all?

We also note that bac value of 0.1% and 0.08% are both statistically significant parameters.

- This model suggests that setting a blood alcohol limit of 0.1% is associated with a 0.97 times decrease in log fatality rate as compared to no BAC limit. Compared to the OLS model which showed that setting a blood alcohol limit of 0.1% is associated with a 0.98 times decrease in fatality rate as compared to no BAC limit.
- This model suggests that a setting a blood alcohol limit of 0.08% is associated with a 0.96 times decrease in fatality rate as compared to no BAC limit. Compared to the OLS model which showed that setting a blood alcohol limit of 0.08% is associated with a 0.96 times decrease in fatality rate as compared to no BAC limit.

5.2.2 What do you estimate for coefficients on per se laws? How do the coefficients on per se laws change, if at all?

We note that per-se law is a statistically significant parameter. This model suggests that having a per-se law is associated with a 0.95 times decrease in fatality rate as compared to not having per-se law. Compared to the OLS model which showed that havng a per-se law is associated with a 0.98 times decrease in fatality rate as compared to not having per-se law

5.2.3 What do you estimate for coefficients on primary seat-belt laws? How do the coefficients on primary seatbelt laws change, if at all?

In this model we note that seat belt primary law is a significant factor. This model suggests that having a primary seat belt law is associated with a 0.95 unit decrease in fatality rate as compared to not having primary seat belt law. Compared to the OLS model for which the parameter was not statistically significant and had a close to 0 value of 1

5.3 Model assumptions

A fundamental assumption in an pooled OLS model is that the data is IID. Let's consider a dataset where a sample of a large population was collected on different years. It is unlikely that a particular individual sample is measured twice. In such a circumstance a pooled OLS model would be applicable. However in this dataset, the individual is the state and the same state is measured multiple times across years. This violates the assumption of IID in the pooled OLS. A fixed effect model is then expected to be a better model in this scenario. We perform a F-test between the pooled and the fixed effect model to check for fixed effects. The Null hypothesis is that there are no fixed effects and the alternate hypothesis is that there are fixed effects. We test against an alpha of 0.05

```
res <- pFtest(expanded.within, expanded.ols)
```

With a p-value of 0 less than 0.05, we reject the null hypothesis of no fixed effects. This means we should include state and/or time fixed effects in our model. Hence the fixed effect model is better for this scenario.

6 (10 points) Consider a Random Effects Model

6.0.0.1 Status incomplete

6.0.1 Random effect model assumptions

Assumption for a random effects model is $cov(x_{it}, a_i) = 0$, i.e. the fixed effect is uncorrelated to the independent variables at any time period.

6.0.2 Estimating the random effect model

```
expanded.re <- plm(log(log_total_fatalities_rate) ~ bac + year +
  population_aged_14_to_24_rate + miles_driven_per_capita +
  unemployment_rate + speed_limit_70_plus + per_se_low +
  seat_belt_low + graduated_drivers_license_low,
  data = main_p,
  index = c("state", "year"),
  model = "random")

summary(expanded.re)

## Oneway (individual) effect Random Effect Model
## (Swamy-Arora's transformation)
##
## Call:
## plm(formula = log(log_total_fatalities_rate) ~ bac + year + population_aged_14_to_24_rate +
## miles_driven_per_capita + unemployment_rate + speed_limit_70_plus +
## per_se_low + seat_belt_low + graduated_drivers_license_low,
## data = main_p, model = "random", index = c("state", "year"))
##
## Balanced Panel: n = 48, T = 25, N = 1200
##
## Effects:
```

```

##               var   std.dev share
## idiosyncratic 0.001039 0.032234 0.224
## individual    0.003590 0.059914 0.776
## theta: 0.893
##
## Residuals:
##      Min.      1st Qu.      Median      3rd Qu.      Max.
## -0.1660054 -0.0184966  0.0024194  0.0204280  0.0990869
##
## Coefficients:
##               Estimate Std. Error z-value Pr(>|z|)
## (Intercept)    2.6777e-02  3.3896e-02   0.7900  0.429542
## bac10          -6.9546e-03  4.2398e-03  -1.6403  0.100940
## bac8           -9.2990e-03  6.0977e-03  -1.5250  0.127259
## year1981       -1.8385e-02  6.7789e-03  -2.7121  0.006685 **
## year1982       -3.6454e-02  7.2529e-03  -5.0261  5.005e-07 ***
## year1983       -4.4786e-02  7.5088e-03  -5.9645  2.454e-09 ***
## year1984       -5.8153e-02  7.6184e-03  -7.6333  2.289e-14 ***
## year1985       -6.4079e-02  7.9313e-03  -8.0792  6.520e-16 ***
## year1986       -4.8582e-02  8.4542e-03  -5.7466  9.107e-09 ***
## year1987       -6.1580e-02  9.0616e-03  -6.7958  1.077e-11 ***
## year1988       -6.9287e-02  9.8064e-03  -7.0655  1.600e-12 ***
## year1989       -9.3947e-02  1.0435e-02  -9.0034 < 2.2e-16 ***
## year1990       -9.9176e-02  1.0828e-02  -9.1589 < 2.2e-16 ***
## year1991       -1.1235e-01  1.1114e-02 -10.1091 < 2.2e-16 ***
## year1992       -1.3484e-01  1.1448e-02 -11.7785 < 2.2e-16 ***
## year1993       -1.4047e-01  1.1659e-02 -12.0487 < 2.2e-16 ***
## year1994       -1.5331e-01  1.1941e-02 -12.8391 < 2.2e-16 ***
## year1995       -1.5171e-01  1.2304e-02 -12.3305 < 2.2e-16 ***
## year1996       -1.7129e-01  1.2974e-02 -13.2019 < 2.2e-16 ***
## year1997       -1.7747e-01  1.3275e-02 -13.3688 < 2.2e-16 ***
## year1998       -1.9703e-01  1.3540e-02 -14.5516 < 2.2e-16 ***
## year1999       -2.0305e-01  1.3691e-02 -14.8304 < 2.2e-16 ***
## year2000       -2.1366e-01  1.3885e-02 -15.3877 < 2.2e-16 ***
## year2001       -2.0870e-01  1.4122e-02 -14.7788 < 2.2e-16 ***
## year2002       -1.9994e-01  1.4262e-02 -14.0195 < 2.2e-16 ***
## year2003       -2.0133e-01  1.4339e-02 -14.0401 < 2.2e-16 ***
## year2004       -2.1385e-01  1.4704e-02 -14.5436 < 2.2e-16 ***
## population_aged_14_to_24_rate 9.6276e-03  1.5455e-03   6.2293  4.684e-10 ***
## miles_driven_per_capita      2.5922e-05  1.7696e-06  14.6491 < 2.2e-16 ***
## unemployment_rate          -9.7906e-03  9.8326e-04  -9.9573 < 2.2e-16 ***
## speed_limit_70_plus         2.7523e-02  4.2568e-03   6.4658  1.008e-10 ***
## per_se_low1                -1.7036e-02  3.6422e-03  -4.6774  2.905e-06 ***
## seat_belt_lowsecondary      1.0944e-03  4.1220e-03   0.2655  0.790617
## seat_belt_lowprimary       -1.4173e-02  5.5867e-03  -2.5369  0.011185 *
## graduated_drivers_license_low -4.0096e-03  4.7872e-03  -0.8376  0.402278
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Total Sum of Squares:    4.3716
## Residual Sum of Squares: 1.2721
## R-Squared:    0.70902
## Adj. R-Squared: 0.70052
## Chisq: 2838.65 on 34 DF, p-value: < 2.22e-16

```

6.0.3 We conduct a Hausman test for random vs. fixed effects using phptest.

We perform this test with an $\alpha = 0.05$

```
res <- phptest(expanded.within, expanded.re)
```

With a p-value of $3.6519259 \times 10^{-51}$ much less than α , we reject the null hypothesis that random effects are appropriate, suggesting that we should use the fixed models. The random effects model is not likely to be consistent in this case.

TODO: Comment about consequences

7 (10 points) Model Forecasts

7.0.0.1 Status not-started The COVID-19 pandemic dramatically changed patterns of driving. Find data (and include this data in your analysis, here) that includes some measure of vehicle miles driven in the US. Your data should at least cover the period from January 2018 to as current as possible. With this data, produce the following statements:

- Comparing monthly miles driven in 2018 to the same months during the pandemic:
 - What month demonstrated the largest decrease in driving? How much, in percentage terms, lower was this driving?
 - What month demonstrated the largest increase in driving? How much, in percentage terms, higher was this driving?

Now, use these changes in driving to make forecasts from your models.

- Suppose that the number of miles driven per capita, increased by as much as the COVID boom. Using the FE estimates, what would the consequences be on the number of traffic fatalities? Please interpret the estimate.
- Suppose that the number of miles driven per capita, decreased by as much as the COVID bust. Using the FE estimates, what would the consequences be on the number of traffic fatalities? Please interpret the estimate.

```
# Source
# U.S. Federal Highway Administration, Vehicle Miles Traveled [TRFVOLUSM227NFWA],
# retrieved from FRED, Federal Reserve Bank of St. Louis;
# https://fred.stlouisfed.org/series/TRFVOLUSM227NFWA, November 29, 2022.

miles.driven <-
  readr::read_csv("./data/TRFVOLUSM227NFWA.csv", show_col_types = FALSE) %>%
  rename(miles = TRFVOLUSM227NFWA, date = DATE)

largest.decrease <- 0
largest.increase <- 0
max <- -Inf
min <- Inf

# Comparing each month miles driven
for (i in 1:12) {
```

```

val.18 <- miles.driven %>%
  filter(date == mdy(paste(i, 1, 2018, sep="/")))

# Comparing 2018 with 2020 and 2021
for (j in 2020:2021) {
  val.covid <- miles.driven %>%
    filter(date == mdy(paste(i, 1, j, sep="/")))

  diff <- val.covid$miles - val.18$miles
  if(diff < min) {
    largest.decrease <- mdy(paste(i, 1, j, sep="/"))
    min = diff
  }
  if(diff > max) {
    largest.increase <- mdy(paste(i, 1, j, sep="/"))
    max = diff
  }
}

# Calculating percentage
prev <- miles.driven %>%
  filter(date == largest.decrease)
lower.percentage <- min/prev$miles

prev <- miles.driven %>%
  filter(date == largest.increase)
higher.percentage <- max/prev$miles

```

‘The largest decrease in driving was on April 2020 and it was lower by -64.14% amount.’

‘The largest increase in driving was on January 2020 and it was higher by 6.18% amount.’

8 (5 points) Evaluate Error

If there were serial correlation or heteroskedasticity in the idiosyncratic errors of the model, what would be the consequences on the estimators and their standard errors? Is there any serial correlation or heteroskedasticity?

8.0.1 Analysis of residuals

8.0.1.1 Status incomplete We test the residuals for serial correlation. The Null hypothesis for the test is that the residuals are not serially correlated and the alternate hypothesis is that the residuals are serially correlated. We check against an α of 0.05.

```

#
res <- pdwtest(expanded.within)

```

With a p-value of $6.5724615 \times 10^{-46}$ less than 0.05, we reject the null hypothesis of no serial correlation. This suggests that we must use robust standard errors for model parameters

Heteroskedasticity test

```
pcdtest(expanded.within, test = "lm")
```

```
##  
## Breusch-Pagan LM test for cross-sectional dependence in panels  
##  
## data: log_total_fatalities_rate ~ bac + year + population_aged_14_to_24_rate + miles_driven_per.  
## chisq = 2826.9, df = 1128, p-value < 2.2e-16  
## alternative hypothesis: cross-sectional dependence
```