

# Lab 3: Panel Models

## US Traffic Fatalities: 1980 - 2004

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## Warning: package 'plm' was built under R version 4.1.3

## 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)
```

```

## Rows: 1,200
## Columns: 56
## $ year      <int> 1980, 1981, 1982, 1983, 1984, 1985, 1986, 1987, 1988, 198~
## $ state     <int> 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, ~
## $ sl55      <dbl> 1.000, 1.000, 1.000, 1.000, 1.000, 1.000, 1.000, 0.542, 0~
## $ sl65      <dbl> 0.000, 0.000, 0.000, 0.000, 0.000, 0.000, 0.000, 0.458, 1~
## $ sl70      <dbl> 0.000, 0.000, 0.000, 0.000, 0.000, 0.000, 0.000, 0.000, 0~
## $ sl75      <dbl> 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, ~
## $ slnone    <dbl> 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, ~
## $ seatbelt  <int> 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 2, 2, 2, 2, 2, 2, ~
## $ minage    <dbl> 18, 18, 18, 18, 18, 20, 21, 21, 21, 21, 21, 21, 21, 21, 21, 2~
## $ zerotol   <dbl> 0.000, 0.000, 0.000, 0.000, 0.000, 0.000, 0.000, 0.000, 0~
## $ gdl       <dbl> 0.00, 0.00, 0.00, 0.00, 0.00, 0.00, 0.00, 0.00, 0.00, 0.0~
## $ bac10     <dbl> 1.000, 1.000, 1.000, 1.000, 1.000, 1.000, 1.000, 1.000, 1~
## $ bac08     <dbl> 0.000, 0.000, 0.000, 0.000, 0.000, 0.000, 0.000, 0.000, 0~
## $ perse     <dbl> 0.000, 0.000, 0.000, 0.000, 0.000, 0.000, 0.000, 0.000, 0~
## $ totfat    <int> 940, 933, 839, 930, 932, 882, 1080, 1111, 1024, 1029, 112~
## $ nghtfat   <int> 422, 434, 376, 397, 421, 358, 500, 499, 423, 418, 466, 47~
## $ wkndfat   <int> 236, 248, 224, 223, 237, 224, 279, 300, 226, 247, 271, 27~
## $ totfatpvm <dbl> 3.200, 3.350, 2.810, 3.000, 2.830, 2.510, 3.177, 2.970, 2~
## $ nghtfatpvm <dbl> 1.437, 1.558, 1.259, 1.281, 1.278, 1.019, 1.471, 1.334, 1~
## $ wkndfatpvm <dbl> 0.803, 0.890, 0.750, 0.719, 0.720, 0.637, 0.821, 0.802, 0~
## $ statepop  <int> 3893888, 3918520, 3925218, 3934109, 3951834, 3972527, 399~
## $ totfatrte <dbl> 24.14, 24.07, 21.37, 23.64, 23.58, 22.20, 27.08, 27.67, 2~
## $ nghtfatrte <dbl> 10.84, 11.08, 9.58, 10.09, 10.65, 9.01, 12.53, 12.43, 10.~
## $ wkndfatrte <dbl> 6.060000, 6.330000, 5.710000, 5.670000, 6.000000, 5.64000~
## $ vehicmiles <dbl> 29.37500, 27.85200, 29.85765, 31.00000, 32.93286, 35.1394~
## $ unem      <dbl> 8.8, 10.7, 14.4, 13.7, 11.1, 8.9, 9.8, 7.8, 7.2, 7.0, 6.9~
## $ perc14_24 <dbl> 18.9, 18.7, 18.4, 18.0, 17.6, 17.3, 17.0, 16.6, 16.2, 15.~
## $ sl70plus  <dbl> 0.000, 0.000, 0.000, 0.000, 0.000, 0.000, 0.000, 0.000, 0~
## $ sbprim    <int> 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, ~
## $ sbsecon   <int> 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 1, 1, 1, 1, 1, 1, ~
## $ d80       <int> 1, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, ~
## $ d81       <int> 0, 1, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, ~
## $ d82       <int> 0, 0, 1, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, ~
## $ d83       <int> 0, 0, 0, 1, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, ~
## $ d84       <int> 0, 0, 0, 0, 1, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, ~
## $ d85       <int> 0, 0, 0, 0, 0, 1, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, ~
## $ d86       <int> 0, 0, 0, 0, 0, 0, 1, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, ~
## $ d87       <int> 0, 0, 0, 0, 0, 0, 0, 1, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, ~
## $ d88       <int> 0, 0, 0, 0, 0, 0, 0, 0, 1, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, ~
## $ d89       <int> 0, 0, 0, 0, 0, 0, 0, 0, 0, 1, 0, 0, 0, 0, 0, 0, 0, 0, 0, ~
## $ d90       <int> 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 1, 0, 0, 0, 0, 0, 0, 0, 0, ~
## $ d91       <int> 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 1, 0, 0, 0, 0, 0, 0, 0, ~
## $ d92       <int> 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 1, 0, 0, 0, 0, 0, 0, ~
## $ d93       <int> 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 1, 0, 0, 0, 0, 0, ~
## $ d94       <int> 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 1, 0, 0, 0, 0, ~
## $ d95       <int> 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 1, 0, 0, 0, ~
## $ d96       <int> 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 1, 0, 0, ~
## $ d97       <int> 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 1, 0, ~
## $ d98       <int> 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 1, ~
## $ d99       <int> 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, ~
## $ d00       <int> 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, ~
## $ d01       <int> 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, ~

```

```
## $ d02          <int> 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, ~
## $ d03          <int> 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, ~
## $ d04          <int> 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, ~
## $ vehicmilespc <dbl> 7543.874, 7107.785, 7606.622, 7879.802, 8333.562, 8845.61~
```

```
desc
```

```
##      variable                                label
## 1      year                                1980 through 2004
## 2      state                        48 continental states, alphabetical
## 3      sl55                                speed limit == 55
## 4      sl65                                speed limit == 65
## 5      sl70                                speed limit == 70
## 6      sl75                                speed limit == 75
## 7      slnone                             no speed limit
## 8      seatbelt      =0 if none, =1 if primary, =2 if secondary
## 9      minage                                minimum drinking age
## 10     zerotol                                zero tolerance law
## 11     gdl                                graduated drivers license law
## 12     bac10                                blood alcohol limit .10
## 13     bac08                                blood alcohol limit .08
## 14     perse administrative license revocation (per se law)
## 15     totfat                                total traffic fatalities
## 16     nghtfat                                total nighttime fatalities
## 17     wkndfat                                total weekend fatalities
## 18     totfatpvm        total fatalities per 100 million miles
## 19     nghtfatpvm        nighttime fatalities per 100 million miles
## 20     wkndfatpvm        weekend fatalities per 100 million miles
## 21     statepop                                state population
## 22     totfatrte        total fatalities per 100,000 population
## 23     nghtfatrte        nighttime fatalities per 100,000 population
## 24     wkndfatrte        weekend accidents per 100,000 population
## 25     vehicmiles        vehicle miles traveled, billions
## 26     unem                                unemployment rate, percent
## 27     perc14_24        percent population aged 14 through 24
## 28     sl70plus                                sl70 + sl75 + slnone
## 29     sbprim                                =1 if primary seatbelt law
## 30     sbsecon                                =1 if secondary seatbelt law
## 31     d80                                =1 if year == 1980
## 32     d81
## 33     d82
## 34     d83
## 35     d84
## 36     d85
## 37     d86
## 38     d87
## 39     d88
## 40     d89
## 41     d90
## 42     d91
## 43     d92
## 44     d93
## 45     d94
## 46     d95
```

```
## 47          d96
## 48          d97
## 49          d98
## 50          d99
## 51          d00
## 52          d01
## 53          d02
## 54          d03
## 55          d04                      =1 if year == 2004
## 56 vehicmilespc
```

```
head(desc)
```

```
##   variable                                label
## 1    year                                1980 through 2004
## 2    state 48 continental states, alphabetical
## 3    sl55                                speed limit == 55
## 4    sl65                                speed limit == 65
## 5    sl70                                speed limit == 70
## 6    sl75                                speed limit == 75
```

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

1. (5 points) Load the data and produce useful features. Specifically:

- Produce a new variable, called `speed_limit` that re-encodes the data that is in `sl55`, `sl65`, `sl70`, `sl75`, and `slnone`;

```
data[data$sl55 == 0.5, 4:6] <- 0
data[data$sl65 == 0.5, 5:6] <- 0

data <- data %>%
  pivot_longer(col = sl55:slnone, names_to="speed_limit", names_prefix="sl") %>%
  filter(value >= 0.5) %>%
  subset(select = -c(value))
```

- Produce a new variable, called `year_of_observation` that re-encodes the data that is in `d80`, `d81`, ..., `d04`.

```
data <- data[-c(which(colnames(data)=="d80"):which(colnames(data)=="d04"))]

data <- data %>%
  rename("year_of_observation" = "year")
```

- Produce a new variable for each of the other variables that are one-hot encoded (i.e. `bac*` variable series).

```

data <- add_column(data, bacnone = 0, .after = "bac08")

data <- data %>%
  mutate(
    bacnone = ifelse(bac10 == 0 & bac08 == 0, 1, 0),
    bac10 = ifelse(bac10 > 0 & bac08 == 0, 1, bac10)
  )

data[data$bac08 == 0.5, "bac10"] <- 0

data <- data %>%
  pivot_longer(col = bac10:bacnone, names_to="blood_alcohol_level", names_prefix="bac") %>%
  filter(value >= 0.5) %>%
  subset(select = -c(value))

```

- Rename these variables to sensible names that are legible to a reader of your analysis. For example, the dependent variable as provided is called, `totfatrte`. Pick something more sensible, like, `total_fatalities_rate`. There are few enough of these variables to change, that you should change them for all the variables in the data. (You will thank yourself later.)

```

col_name_lookup <- c(
  "min_drink_age" = "minage",
  "zero_tol_law" = "zerotol",
  "grad_driver_law" = "gdl",
  "per_se_law" = "perse",
  "total_fatalities" = "totfat",
  "night_fatalities" = "nghtfat",
  "weekend_fatalities" = "wkndfat",
  "total_fatalities_per_100mil_miles" = "totfatpvm",
  "night_fatalities_per_100mil_miles" = "nghtfatpvm",
  "weekend_fatalities_per_100mil_miles" = "wkndfatpvm",
  "state_population" = "statepop",
  "total_fatalities_rate" = "totfatrte",
  "night_fatalities_rate" = "nghtfatrte",
  "weekend_fatalities_rate" = "wkndfatrte",
  "vehicle_miles" = "vehicmiles",
  "unemployment_rate" = "unem",
  "percent_age_14_24" = "perc14_24",
  "primary_seatbelt" = "sbprim",
  "secondary_seatbelt" = "sbsecon"
)

data <- data %>%
  rename(any_of(col_name_lookup))

```

2. (5 points) Provide a description of the basic structure of the dataset. What is this data? How, where, and when is it collected? Is the data generated through a survey or some other method? Is the data that is presented a sample from the population, or is it a *census* that represents the entire population? Minimally, this should include:

- How is the our dependent variable of interest `total_fatalities_rate` defined?

```
pdriving <- pdata.frame(
  data,
  index=c("state", "year_of_observation")
)

pdim(pdriving)
```

```
## Balanced Panel: n = 48, T = 25, N = 1200
```

```
# Renaming States
replacement_state <- c("AL", "AZ", "AR", "CA", "CO", "CT", "DE", "FL",
  "GA", "ID", "IL", "IN", "IA", "KS", "KY", "LA",
  "ME", "MD", "MA", "MI", "MN", "MS", "MO", "MT",
  "NE", "NV", "NH", "NJ", "NM", "NY", "NC", "ND",
  "OH", "OK", "OR", "PA", "RI", "SC", "SD", "TN",
  "TX", "UT", "VT", "VA", "WA", "WV", "WI", "WY")

levels(pdriving$state) <- replacement_state
```

3. (20 points) Conduct a very thorough EDA, which should include both graphical and tabular techniques, on the dataset, including both the dependent variable `total_fatalities_rate` and the potential explanatory variables. Minimally, this should include:

- How is the our dependent variable of interest `total_fatalities_rate` defined?
- What is the average of `total_fatalities_rate` in each of the years in the time period covered in this dataset?

```
sapply(pdriving, function(x)sum(is.na(x)))
```

```
##          year_of_observation          state
##                0                0
##          seatbelt          min_drink_age
##                0                0
##          zero_tol_law          grad_driver_law
##                0                0
##          per_se_law          total_fatalities
##                0                0
##          night_fatalities          weekend_fatalities
##                0                0
## total_fatalities_per_100mil_miles  night_fatalities_per_100mil_miles
##                0                0
## weekend_fatalities_per_100mil_miles          state_population
##                0                0
##          total_fatalities_rate          night_fatalities_rate
##                0                0
##          weekend_fatalities_rate          vehicle_miles
##                0                0
##          unemployment_rate          percent_age_14_24
##                0                0
##          sl70plus          primary_seatbelt
##                0                0
##          secondary_seatbelt          vehicmilespc
```

```
##                                0                                0
##                                speed_limit                        blood_alcohol_level
##                                0                                0
```

```
table(pdriving$year_of_observation)
```

```
##
## 1980 1981 1982 1983 1984 1985 1986 1987 1988 1989 1990 1991 1992 1993 1994 1995
##   48   48   48   48   48   48   48   48   48   48   48   48   48   48   48   48
## 1996 1997 1998 1999 2000 2001 2002 2003 2004
##   48   48   48   48   48   48   48   48   48
```

```
table(pdriving$state)
```

```
##
## AL AZ AR CA CO CT DE FL GA ID IL IN IA KS KY LA ME MD MA MI MN MS MO MT NE NV
## 25 25 25 25 25 25 25 25 25 25 25 25 25 25 25 25 25 25 25 25 25 25 25 25
## NH NJ NM NY NC ND OH OK OR PA RI SC SD TN TX UT VT VA WA WV WI WY
## 25 25 25 25 25 25 25 25 25 25 25 25 25 25 25 25 25 25 25 25 25 25 25 25
```

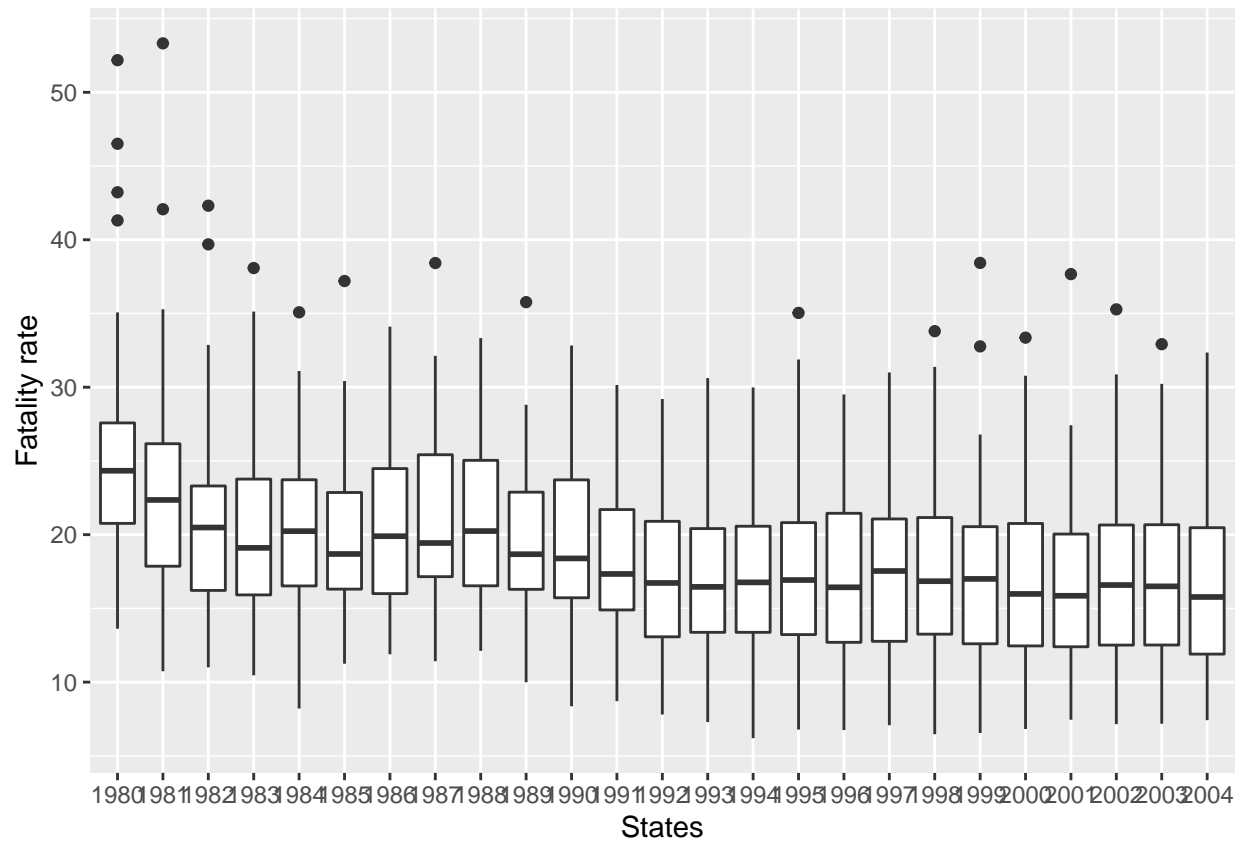
```
pdriving %>%
  is.pconsecutive()
```

```
##    1    3    4    5    6    7    8   10   11   13   14   15   16   17   18   19
## TRUE TRUE TRUE TRUE TRUE TRUE TRUE TRUE TRUE TRUE TRUE TRUE TRUE TRUE TRUE
##  20  21  22  23  24  25  26  27  28  29  30  31  32  33  34  35
## TRUE TRUE TRUE TRUE TRUE TRUE TRUE TRUE TRUE TRUE TRUE TRUE TRUE TRUE TRUE
##  36  37  38  39  40  41  42  43  44  45  46  47  48  49  50  51
## TRUE TRUE TRUE TRUE TRUE TRUE TRUE TRUE TRUE TRUE TRUE TRUE TRUE TRUE TRUE
```

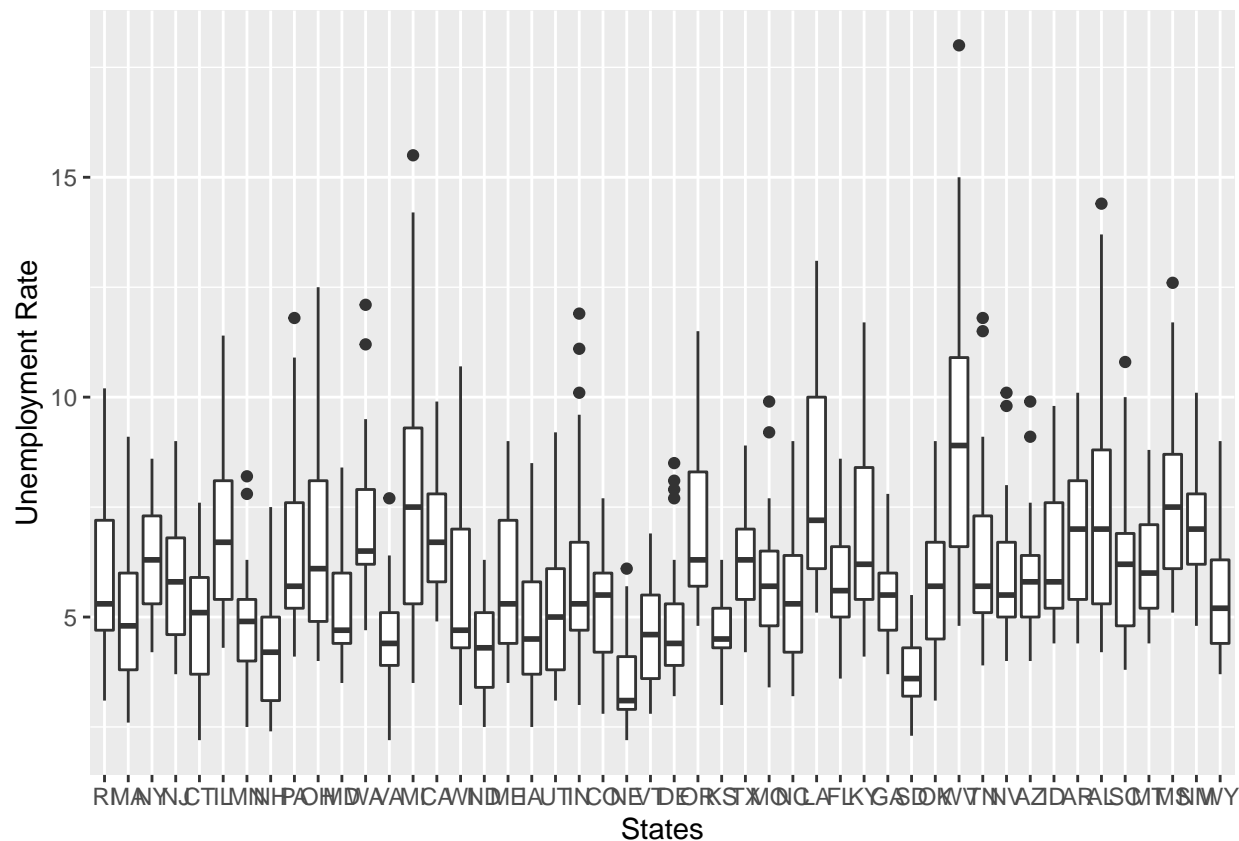
```
pdriving %>%
  group_by(state) %>%
  ggplot(
    aes(x = reorder(state,total_fatalities_rate),
         y = total_fatalities_rate)) +
  geom_boxplot() +
  labs(
    x = "States",
    y = "Fatality rate"
  )
```







```
pdriving %>%
  group_by(state) %>%
  ggplot(aes(x = reorder(state,total_fatalities_rate), y = unemployment_rate)) +
  geom_boxplot() +
  labs(
    x = "States",
    y = "Unemployment Rate",
  )
```



As with every EDA this semester, the goal of this EDA is not to document your own process of discovery – save that for an exploration notebook – but instead it is to bring a reader that is new to the data to a full understanding of the important features of your data as quickly as possible. In order to do this, your EDA should include a detailed, orderly narrative description of what you want your reader to know. Do not include any output – tables, plots, or statistics – that you do not intend to write about.

### 3 (15 points) Preliminary Model

Estimate a linear regression model of *totfatrtte* on a set of dummy variables for the years 1981 through 2004 and interpret what you observe. In this section, you should address the following tasks:

```
mod.prel <- lm(
  formula = total_fatalities_rate ~ year_of_observation,
  data = pdriving
)
```

```
summary(mod.prel)
```

```
##
## Call:
## lm(formula = total_fatalities_rate ~ year_of_observation, data = pdriving)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
```

```
## -12.9302 -4.3468 -0.7305 3.7488 29.6498
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)      25.4946    0.8671  29.401 < 2e-16 ***
## year_of_observation1981 -1.8244    1.2263  -1.488 0.137094
## year_of_observation1982 -4.5521    1.2263  -3.712 0.000215 ***
## year_of_observation1983 -5.3417    1.2263  -4.356 1.44e-05 ***
## year_of_observation1984 -5.2271    1.2263  -4.263 2.18e-05 ***
## year_of_observation1985 -5.6431    1.2263  -4.602 4.64e-06 ***
## year_of_observation1986 -4.6942    1.2263  -3.828 0.000136 ***
## year_of_observation1987 -4.7198    1.2263  -3.849 0.000125 ***
## year_of_observation1988 -4.6029    1.2263  -3.754 0.000183 ***
## year_of_observation1989 -5.7223    1.2263  -4.666 3.42e-06 ***
## year_of_observation1990 -5.9894    1.2263  -4.884 1.18e-06 ***
## year_of_observation1991 -7.3998    1.2263  -6.034 2.14e-09 ***
## year_of_observation1992 -8.3367    1.2263  -6.798 1.68e-11 ***
## year_of_observation1993 -8.3669    1.2263  -6.823 1.43e-11 ***
## year_of_observation1994 -8.3394    1.2263  -6.800 1.66e-11 ***
## year_of_observation1995 -7.8260    1.2263  -6.382 2.51e-10 ***
## year_of_observation1996 -8.1252    1.2263  -6.626 5.25e-11 ***
## year_of_observation1997 -7.8840    1.2263  -6.429 1.86e-10 ***
## year_of_observation1998 -8.2292    1.2263  -6.711 3.01e-11 ***
## year_of_observation1999 -8.2442    1.2263  -6.723 2.77e-11 ***
## year_of_observation2000 -8.6690    1.2263  -7.069 2.67e-12 ***
## year_of_observation2001 -8.7019    1.2263  -7.096 2.21e-12 ***
## year_of_observation2002 -8.4650    1.2263  -6.903 8.32e-12 ***
## year_of_observation2003 -8.7310    1.2263  -7.120 1.88e-12 ***
## year_of_observation2004 -8.7656    1.2263  -7.148 1.54e-12 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 6.008 on 1175 degrees of freedom
## Multiple R-squared:  0.1276, Adjusted R-squared:  0.1098
## F-statistic: 7.164 on 24 and 1175 DF, p-value: < 2.2e-16
```

```
mod.prel <- plm(
  formula = total_fatalities_rate ~ year_of_observation,
  data = pdriving,
  model = "pooling"
)
```

```
summary(mod.prel)
```

```
## Pooling Model
##
## Call:
## plm(formula = total_fatalities_rate ~ year_of_observation, data = pdriving,
##      model = "pooling")
##
## Balanced Panel: n = 48, T = 25, N = 1200
##
## Residuals:
##      Min.      1st Qu.      Median      3rd Qu.      Max.
```

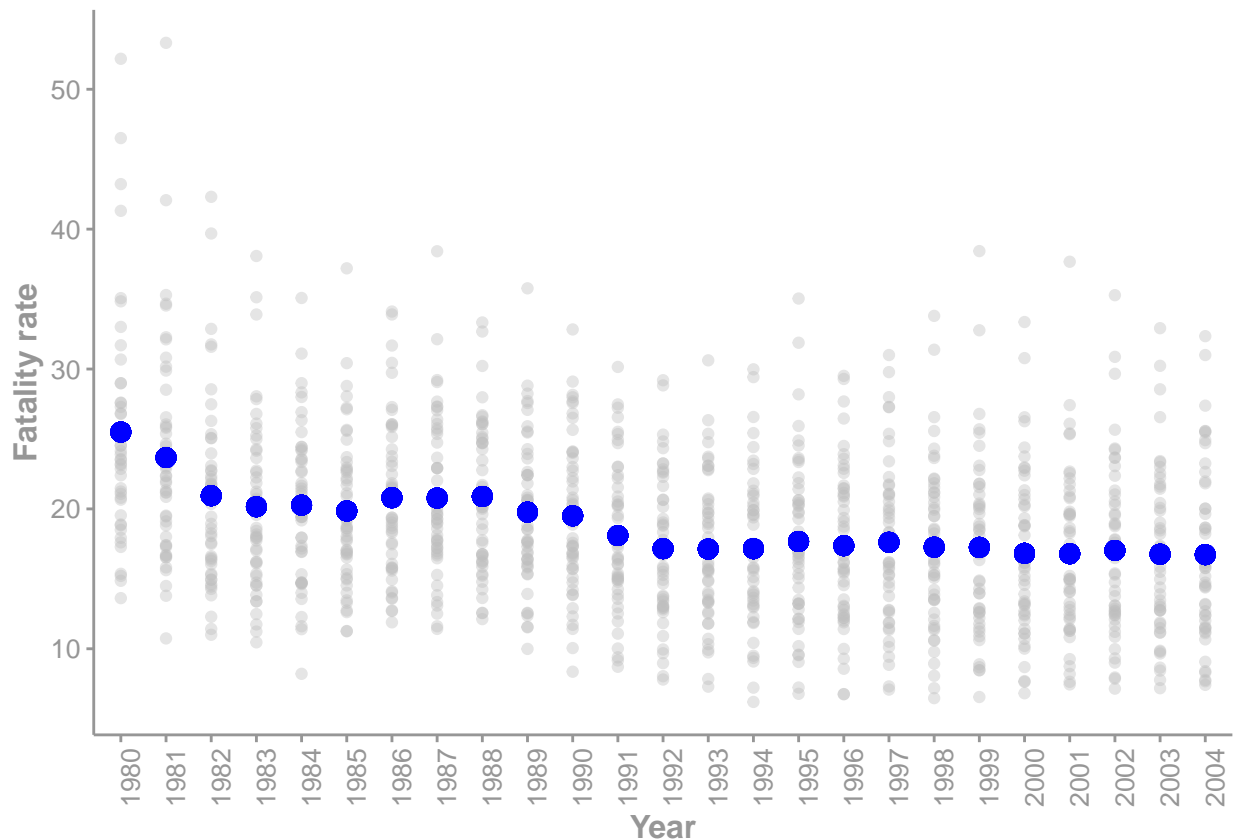
```
## -12.93021 -4.34682 -0.73052 3.74875 29.64979
##
## Coefficients:
##              Estimate Std. Error t-value Pr(>|t|)
## (Intercept)    25.49458    0.86712  29.4015 < 2.2e-16 ***
## year_of_observation1981 -1.82438    1.22629  -1.4877 0.1370936
## year_of_observation1982 -4.55208    1.22629  -3.7121 0.0002152 ***
## year_of_observation1983 -5.34167    1.22629  -4.3560 1.440e-05 ***
## year_of_observation1984 -5.22708    1.22629  -4.2625 2.183e-05 ***
## year_of_observation1985 -5.64313    1.22629  -4.6018 4.644e-06 ***
## year_of_observation1986 -4.69417    1.22629  -3.8279 0.0001360 ***
## year_of_observation1987 -4.71979    1.22629  -3.8488 0.0001251 ***
## year_of_observation1988 -4.60292    1.22629  -3.7535 0.0001829 ***
## year_of_observation1989 -5.72229    1.22629  -4.6663 3.418e-06 ***
## year_of_observation1990 -5.98938    1.22629  -4.8841 1.182e-06 ***
## year_of_observation1991 -7.39979    1.22629  -6.0343 2.137e-09 ***
## year_of_observation1992 -8.33667    1.22629  -6.7983 1.681e-11 ***
## year_of_observation1993 -8.36688    1.22629  -6.8229 1.425e-11 ***
## year_of_observation1994 -8.33938    1.22629  -6.8005 1.656e-11 ***
## year_of_observation1995 -7.82604    1.22629  -6.3819 2.512e-10 ***
## year_of_observation1996 -8.12521    1.22629  -6.6258 5.246e-11 ***
## year_of_observation1997 -7.88396    1.22629  -6.4291 1.863e-10 ***
## year_of_observation1998 -8.22917    1.22629  -6.7106 3.007e-11 ***
## year_of_observation1999 -8.24417    1.22629  -6.7228 2.774e-11 ***
## year_of_observation2000 -8.66896    1.22629  -7.0692 2.666e-12 ***
## year_of_observation2001 -8.70188    1.22629  -7.0961 2.214e-12 ***
## year_of_observation2002 -8.46500    1.22629  -6.9029 8.316e-12 ***
## year_of_observation2003 -8.73104    1.22629  -7.1199 1.877e-12 ***
## year_of_observation2004 -8.76563    1.22629  -7.1481 1.542e-12 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Total Sum of Squares:    48612
## Residual Sum of Squares: 42407
## R-Squared:    0.12765
## Adj. R-Squared: 0.10983
## F-statistic: 7.16387 on 24 and 1175 DF, p-value: < 2.22e-16
```

```
pdriving%>%
  ggplot(
    aes(
      x = year_of_observation,
      y = total_fatalities_rate
    )
  ) +
  geom_point(
    color = "gray",
    alpha = 0.4
  ) +
  geom_point(
    data = broom::augment(mod.prel),
    aes(x = year_of_observation, y = .fitted),
    colour = "blue", size = 3) +
  labs(
```

```

    x = "Year",
    y = "Fatality rate"
  ) +
  theme_classic() +
  theme(
    plot.title = element_text(color = "#0099F8",
                              size = 14,
                              face = "bold"),
    plot.subtitle = element_text(color="#969696",
                                 size = 12,
                                 face = "italic"),
    axis.title = element_text(color = "#969696",
                              size = 12,
                              face = "bold"),
    axis.text = element_text(color = "#969696", size = 10),
    axis.text.x = element_text(angle = 90),
    axis.line = element_line(color = "#969696"),
    axis.ticks = element_line(color = "#969696"),
    legend.position = "none"
  )

```



- Why is fitting a linear model a sensible starting place?
- What does this model explain, and what do you find in this model?
- Did driving become safer over this period? Please provide a detailed explanation.
- What, if any, are the limitation of this model. In answering this, please consider **at least**:

- Are the parameter estimates reliable, unbiased estimates of the truth? Or, are they biased due to the way that the data is structured?
- Are the uncertainty estimate reliable, unbiased estimates of sampling based variability? Or, are they biased due to the way that the data is structured?

## 4 (15 points) Expanded Model

Expand the **Preliminary Model** by adding variables related to the following concepts:

```
mod.exp <- plm(
  formula = total_fatalities_rate ~ year_of_observation +
    blood_alcohol_level + per_se_law + primary_seatbelt +
    secondary_seatbelt + sl70plus + grad_driver_law +
    percent_age_14_24 + unemployment_rate + vehicle_miles,
  data = pdriving,
  model = "pooling"
)

summary(mod.exp)
```

```
## Pooling Model
##
## Call:
## plm(formula = total_fatalities_rate ~ year_of_observation + blood_alcohol_level +
##     per_se_law + primary_seatbelt + secondary_seatbelt + sl70plus +
##     grad_driver_law + percent_age_14_24 + unemployment_rate +
##     vehicle_miles, data = pdriving, model = "pooling")
##
## Balanced Panel: n = 48, T = 25, N = 1200
##
## Residuals:
##      Min.    1st Qu.    Median    3rd Qu.    Max.
## -14.68802  -3.51217  -0.23065   3.10572  30.01097
##
## Coefficients:
##              Estimate Std. Error t-value Pr(>|t|)
## (Intercept)      8.5551252   3.2766935   2.6109 0.0091463 **
## year_of_observation1981 -1.7955110   1.0837542  -1.6568 0.0978391 .
## year_of_observation1982 -5.0673484   1.1164450  -4.5388 6.243e-06 ***
## year_of_observation1983 -5.0740375   1.1440586  -4.4351 1.007e-05 ***
## year_of_observation1984 -3.4475090   1.1451898  -3.0104 0.0026650 **
## year_of_observation1985 -3.4294095   1.1665564  -2.9398 0.0033493 **
## year_of_observation1986 -2.2308516   1.2068120  -1.8485 0.0647762 .
## year_of_observation1987 -1.5209009   1.2460821  -1.2205 0.2225048
## year_of_observation1988 -0.6487868   1.2992242  -0.4994 0.6176167
## year_of_observation1989 -1.2425109   1.3430081  -0.9252 0.3550690
## year_of_observation1990 -1.4168492   1.3684303  -1.0354 0.3007050
## year_of_observation1991 -3.1764258   1.3964649  -2.2746 0.0231099 *
## year_of_observation1992 -4.1359683   1.4149554  -2.9230 0.0035333 **
## year_of_observation1993 -3.8138479   1.4321744  -2.6630 0.0078520 **
## year_of_observation1994 -3.3146562   1.4594016  -2.2712 0.0233141 *
## year_of_observation1995 -2.5915868   1.4909434  -1.7382 0.0824364 .
```

```

## year_of_observation1996 -5.2356191 1.5535719 -3.3701 0.0007762 ***
## year_of_observation1997 -5.6907257 1.5896298 -3.5799 0.0003579 ***
## year_of_observation1998 -6.0548285 1.6049358 -3.7726 0.0001697 ***
## year_of_observation1999 -5.6609841 1.6241503 -3.4855 0.0005095 ***
## year_of_observation2000 -5.7192064 1.6488312 -3.4686 0.0005422 ***
## year_of_observation2001 -5.6886169 1.6781335 -3.3898 0.0007227 ***
## year_of_observation2002 -5.7245484 1.6928836 -3.3815 0.0007447 ***
## year_of_observation2003 -6.0681892 1.7062106 -3.5565 0.0003909 ***
## year_of_observation2004 -5.7213821 1.7434371 -3.2817 0.0010625 **
## blood_alcohol_level10 0.1052716 0.4889762 0.2153 0.8295790
## blood_alcohol_levelnone 1.4628108 0.6968367 2.0992 0.0360126 *
## per_se_law 0.1195992 0.3930198 0.3043 0.7609474
## primary_seatbelt 0.5352343 0.6866053 0.7795 0.4358217
## secondary_seatbelt 0.2625729 0.5691865 0.4613 0.6446605
## sl70plus 6.6797904 0.5742115 11.6330 < 2.2e-16 ***
## grad_driver_law -1.6132134 0.6915411 -2.3328 0.0198292 *
## percent_age_14_24 0.7120920 0.1586494 4.4885 7.885e-06 ***
## unemployment_rate 0.5068911 0.1021488 4.9623 8.003e-07 ***
## vehicle_miles -0.0316637 0.0037593 -8.4228 < 2.2e-16 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Total Sum of Squares: 48612
## Residual Sum of Squares: 32703
## R-Squared: 0.32726
## Adj. R-Squared: 0.30763
## F-statistic: 16.6687 on 34 and 1165 DF, p-value: < 2.22e-16

```

- Blood alcohol levels
- Per se laws
- Primary seat belt laws (Note that if a law was enacted sometime within a year the fraction of the year is recorded in place of the zero-one indicator.)
- Secondary seat belt laws
- Speed limits faster than 70
- Graduated drivers licenses
- Percent of the population between 14 and 24 years old
- Unemployment rate
- Vehicle miles driven per capita.

If it is appropriate, include transformations of these variables. Please carefully explain carefully your rationale, which should be based on your EDA, behind any transformation you made. If no transformation is made, explain why transformation is not needed.

- How are the blood alcohol variables defined? Interpret the coefficients that you estimate for this concept.
- Do *per se laws* have a negative effect on the fatality rate?
- Does having a primary seat belt law?

## 5 (15 points) State-Level Fixed Effects

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

```

mod.fix <- plm(
  formula = total_fatalities_rate ~ state + year_of_observation +
    blood_alcohol_level + per_se_law + primary_seatbelt +
    secondary_seatbelt + sl70plus + grad_driver_law +
    percent_age_14_24 + unemployment_rate + vehicle_miles - 1,
  data = pdriving,
  model = "pooling"
)

summary(mod.fix)

```

```

## Pooling Model
##
## Call:
## plm(formula = total_fatalities_rate ~ state + year_of_observation +
##      blood_alcohol_level + per_se_law + primary_seatbelt + secondary_seatbelt +
##      sl70plus + grad_driver_law + percent_age_14_24 + unemployment_rate +
##      vehicle_miles - 1, data = pdriving, model = "pooling")
##
## Balanced Panel: n = 48, T = 25, N = 1200
##
## Residuals:
##      Min.      1st Qu.      Median      3rd Qu.      Max.
## -9.516829 -1.015138 -0.040218  1.037569  13.641505
##
## Coefficients:
##
##              Estimate Std. Error t-value Pr(>|t|)
## stateAL             31.4583740   2.1053145  14.9424 < 2.2e-16 ***
## stateAZ             29.4119832   2.0385252  14.4281 < 2.2e-16 ***
## stateAR             30.4310326   2.0362274  14.9448 < 2.2e-16 ***
## stateCA             24.0285667   2.6470916   9.0773 < 2.2e-16 ***
## stateCO             22.8876506   2.0253871  11.3004 < 2.2e-16 ***
## stateCT             17.6216667   1.9557770   9.0101 < 2.2e-16 ***
## stateDE             23.4895875   2.0675661  11.3610 < 2.2e-16 ***
## stateFL             28.4312262   2.0145125  14.1132 < 2.2e-16 ***
## stateGA             27.5594818   2.1404408  12.8756 < 2.2e-16 ***
## stateID             29.4316325   2.0853765  14.1133 < 2.2e-16 ***
## stateIL             20.4058697   2.1507783   9.4877 < 2.2e-16 ***
## stateIN             23.6519685   2.1366061  11.0699 < 2.2e-16 ***
## stateIA             22.9361715   2.0603444  11.1322 < 2.2e-16 ***
## stateKS             23.5342417   2.0012596  11.7597 < 2.2e-16 ***
## stateKY             26.9650263   2.1338853  12.6366 < 2.2e-16 ***
## stateLA             29.0260916   2.2238155  13.0524 < 2.2e-16 ***
## stateME             22.8014601   1.9887823  11.4650 < 2.2e-16 ***
## stateMD             19.5654702   2.0305675   9.6355 < 2.2e-16 ***
## stateMA             14.1342121   2.0774172   6.8037 1.658e-11 ***
## stateMI             22.2615369   2.1649593  10.2827 < 2.2e-16 ***
## stateMN             18.9724322   2.0426343   9.2882 < 2.2e-16 ***
## stateMS             36.4054746   2.2137126  16.4454 < 2.2e-16 ***
## stateMO             26.6429954   2.0353690  13.0900 < 2.2e-16 ***
## stateMT             32.7713001   1.9831585  16.5248 < 2.2e-16 ***
## stateNE             21.3167009   1.9894796  10.7147 < 2.2e-16 ***
## stateNV             29.5351437   1.9242026  15.3493 < 2.2e-16 ***

```



## stateNH	18.4196245	1.9670768	9.3640	< 2.2e-16	***
## stateNJ	16.7208760	1.9913664	8.3967	< 2.2e-16	***
## stateNM	38.3660325	2.1010050	18.2608	< 2.2e-16	***
## stateNY	17.2301997	2.1463935	8.0275	2.501e-15	***
## stateNC	27.6846063	2.1007850	13.1782	< 2.2e-16	***
## stateND	20.2483595	2.0986566	9.6482	< 2.2e-16	***
## stateOH	20.6148655	2.1641985	9.5254	< 2.2e-16	***
## stateOK	28.5020995	2.0496949	13.9055	< 2.2e-16	***
## stateOR	26.0316002	1.9751646	13.1795	< 2.2e-16	***
## statePA	19.9812459	2.0904016	9.5586	< 2.2e-16	***
## stateRI	14.0124321	2.0374604	6.8774	1.012e-11	***
## stateSC	30.9955467	2.1529454	14.3968	< 2.2e-16	***
## stateSD	25.6321476	2.0200443	12.6889	< 2.2e-16	***
## stateTN	28.1982226	2.0835904	13.5335	< 2.2e-16	***
## stateTX	27.2401232	2.4040331	11.3310	< 2.2e-16	***
## stateUT	21.5922110	2.2766492	9.4842	< 2.2e-16	***
## stateVT	22.7993780	2.0282579	11.2409	< 2.2e-16	***
## stateVA	19.9419704	2.0946943	9.5202	< 2.2e-16	***
## stateWA	20.9359335	2.0126089	10.4024	< 2.2e-16	***
## stateWV	31.0953312	2.1029006	14.7869	< 2.2e-16	***
## stateWI	21.0947503	2.1071614	10.0110	< 2.2e-16	***
## stateWY	38.2324774	2.0561778	18.5940	< 2.2e-16	***
## year_of_observation1981	-1.4047214	0.4267015	-3.2920	0.0010257	**
## year_of_observation1982	-2.5186230	0.4537352	-5.5509	3.549e-08	***
## year_of_observation1983	-2.8360202	0.4728127	-5.9982	2.691e-09	***
## year_of_observation1984	-3.7743299	0.4774450	-7.9053	6.376e-15	***
## year_of_observation1985	-4.0923152	0.4962825	-8.2459	4.558e-16	***
## year_of_observation1986	-2.8858488	0.5246101	-5.5009	4.681e-08	***
## year_of_observation1987	-3.2314320	0.5565017	-5.8067	8.296e-09	***
## year_of_observation1988	-3.4200752	0.5976286	-5.7227	1.345e-08	***
## year_of_observation1989	-4.5667501	0.6318074	-7.2281	9.058e-13	***
## year_of_observation1990	-4.4395099	0.6499698	-6.8303	1.388e-11	***
## year_of_observation1991	-4.9879391	0.6635662	-7.5169	1.147e-13	***
## year_of_observation1992	-5.5835400	0.6754673	-8.2662	3.885e-16	***
## year_of_observation1993	-5.8673028	0.6886845	-8.5196	< 2.2e-16	***
## year_of_observation1994	-6.2602443	0.7079263	-8.8431	< 2.2e-16	***
## year_of_observation1995	-5.9144495	0.7279900	-8.1244	1.182e-15	***
## year_of_observation1996	-6.2034475	0.7684654	-8.0725	1.767e-15	***
## year_of_observation1997	-6.2335368	0.7900639	-7.8899	7.164e-15	***
## year_of_observation1998	-6.7303628	0.7976212	-8.4380	< 2.2e-16	***
## year_of_observation1999	-6.7627554	0.8029432	-8.4225	< 2.2e-16	***
## year_of_observation2000	-7.2287115	0.8134468	-8.8865	< 2.2e-16	***
## year_of_observation2001	-6.6491671	0.8188523	-8.1201	1.221e-15	***
## year_of_observation2002	-5.7271889	0.8189238	-6.9936	4.604e-12	***
## year_of_observation2003	-5.7483727	0.8241253	-6.9751	5.221e-12	***
## year_of_observation2004	-6.1513452	0.8464757	-7.2670	6.884e-13	***
## blood_alcohol_level10	0.1537963	0.2521919	0.6098	0.5420927	
## blood_alcohol_levelnone	0.7707947	0.3871136	1.9911	0.0467093	*
## per_se_low	-1.1673580	0.2409474	-4.8449	1.445e-06	***
## primary_seatbelt	-1.2097495	0.3567874	-3.3907	0.0007216	***
## secondary_seatbelt	-0.2584527	0.2604218	-0.9924	0.3211983	
## sl70plus	0.2016010	0.2805965	0.7185	0.4726158	
## grad_driver_low	-0.5331449	0.3028770	-1.7603	0.0786357	.
## percent_age_14_24	0.2684700	0.0979019	2.7422	0.0061997	**

```
## unemployment_rate      -0.6960810  0.0608561 -11.4381 < 2.2e-16 ***
## vehicle_miles          -0.0062848  0.0065981  -0.9525  0.3410406
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Total Sum of Squares:    48612
## Residual Sum of Squares: 4840.4
## R-Squared:              0.90043
## Adj. R-Squared: 0.89321
## F-statistic: 1333.08 on 82 and 1118 DF, p-value: < 2.22e-16
```

```
mod.fix <- plm(
  formula = total_fatalities_rate ~ state + year_of_observation +
    blood_alcohol_level + per_se_law + primary_seatbelt +
    secondary_seatbelt + sl70plus + grad_driver_law +
    percent_age_14_24 + unemployment_rate + vehicle_miles - 1,
  data = pdriving,
  effect = "individual",
  model = "pooling"
)

summary(mod.fix)
```

```
## Pooling Model
##
## Call:
## plm(formula = total_fatalities_rate ~ state + year_of_observation +
##      blood_alcohol_level + per_se_law + primary_seatbelt + secondary_seatbelt +
##      sl70plus + grad_driver_law + percent_age_14_24 + unemployment_rate +
##      vehicle_miles - 1, data = pdriving, effect = "individual",
##      model = "pooling")
##
## Balanced Panel: n = 48, T = 25, N = 1200
##
## Residuals:
##      Min.      1st Qu.      Median      3rd Qu.      Max.
## -9.516829 -1.015138 -0.040218  1.037569 13.641505
##
## Coefficients:
##              Estimate Std. Error t-value Pr(>|t|)
## stateAL          31.4583740   2.1053145  14.9424 < 2.2e-16 ***
## stateAZ          29.4119832   2.0385252  14.4281 < 2.2e-16 ***
## stateAR          30.4310326   2.0362274  14.9448 < 2.2e-16 ***
## stateCA          24.0285667   2.6470916   9.0773 < 2.2e-16 ***
## stateCO          22.8876506   2.0253871  11.3004 < 2.2e-16 ***
## stateCT          17.6216667   1.9557770   9.0101 < 2.2e-16 ***
## stateDE          23.4895875   2.0675661  11.3610 < 2.2e-16 ***
## stateFL          28.4312262   2.0145125  14.1132 < 2.2e-16 ***
## stateGA          27.5594818   2.1404408  12.8756 < 2.2e-16 ***
## stateID          29.4316325   2.0853765  14.1133 < 2.2e-16 ***
## stateIL          20.4058697   2.1507783   9.4877 < 2.2e-16 ***
## stateIN          23.6519685   2.1366061  11.0699 < 2.2e-16 ***
## stateIA          22.9361715   2.0603444  11.1322 < 2.2e-16 ***
## stateKS          23.5342417   2.0012596  11.7597 < 2.2e-16 ***
```

## stateKY	26.9650263	2.1338853	12.6366	< 2.2e-16	***
## stateLA	29.0260916	2.2238155	13.0524	< 2.2e-16	***
## stateME	22.8014601	1.9887823	11.4650	< 2.2e-16	***
## stateMD	19.5654702	2.0305675	9.6355	< 2.2e-16	***
## stateMA	14.1342121	2.0774172	6.8037	1.658e-11	***
## stateMI	22.2615369	2.1649593	10.2827	< 2.2e-16	***
## stateMN	18.9724322	2.0426343	9.2882	< 2.2e-16	***
## stateMS	36.4054746	2.2137126	16.4454	< 2.2e-16	***
## stateMO	26.6429954	2.0353690	13.0900	< 2.2e-16	***
## stateMT	32.7713001	1.9831585	16.5248	< 2.2e-16	***
## stateNE	21.3167009	1.9894796	10.7147	< 2.2e-16	***
## stateNV	29.5351437	1.9242026	15.3493	< 2.2e-16	***
## stateNH	18.4196245	1.9670768	9.3640	< 2.2e-16	***
## stateNJ	16.7208760	1.9913664	8.3967	< 2.2e-16	***
## stateNM	38.3660325	2.1010050	18.2608	< 2.2e-16	***
## stateNY	17.2301997	2.1463935	8.0275	2.501e-15	***
## stateNC	27.6846063	2.1007850	13.1782	< 2.2e-16	***
## stateND	20.2483595	2.0986566	9.6482	< 2.2e-16	***
## stateOH	20.6148655	2.1641985	9.5254	< 2.2e-16	***
## stateOK	28.5020995	2.0496949	13.9055	< 2.2e-16	***
## stateOR	26.0316002	1.9751646	13.1795	< 2.2e-16	***
## statePA	19.9812459	2.0904016	9.5586	< 2.2e-16	***
## stateRI	14.0124321	2.0374604	6.8774	1.012e-11	***
## stateSC	30.9955467	2.1529454	14.3968	< 2.2e-16	***
## stateSD	25.6321476	2.0200443	12.6889	< 2.2e-16	***
## stateTN	28.1982226	2.0835904	13.5335	< 2.2e-16	***
## stateTX	27.2401232	2.4040331	11.3310	< 2.2e-16	***
## stateUT	21.5922110	2.2766492	9.4842	< 2.2e-16	***
## stateVT	22.7993780	2.0282579	11.2409	< 2.2e-16	***
## stateVA	19.9419704	2.0946943	9.5202	< 2.2e-16	***
## stateWA	20.9359335	2.0126089	10.4024	< 2.2e-16	***
## stateWV	31.0953312	2.1029006	14.7869	< 2.2e-16	***
## stateWI	21.0947503	2.1071614	10.0110	< 2.2e-16	***
## stateWY	38.2324774	2.0561778	18.5940	< 2.2e-16	***
## year_of_observation1981	-1.4047214	0.4267015	-3.2920	0.0010257	**
## year_of_observation1982	-2.5186230	0.4537352	-5.5509	3.549e-08	***
## year_of_observation1983	-2.8360202	0.4728127	-5.9982	2.691e-09	***
## year_of_observation1984	-3.7743299	0.4774450	-7.9053	6.376e-15	***
## year_of_observation1985	-4.0923152	0.4962825	-8.2459	4.558e-16	***
## year_of_observation1986	-2.8858488	0.5246101	-5.5009	4.681e-08	***
## year_of_observation1987	-3.2314320	0.5565017	-5.8067	8.296e-09	***
## year_of_observation1988	-3.4200752	0.5976286	-5.7227	1.345e-08	***
## year_of_observation1989	-4.5667501	0.6318074	-7.2281	9.058e-13	***
## year_of_observation1990	-4.4395099	0.6499698	-6.8303	1.388e-11	***
## year_of_observation1991	-4.9879391	0.6635662	-7.5169	1.147e-13	***
## year_of_observation1992	-5.5835400	0.6754673	-8.2662	3.885e-16	***
## year_of_observation1993	-5.8673028	0.6886845	-8.5196	< 2.2e-16	***
## year_of_observation1994	-6.2602443	0.7079263	-8.8431	< 2.2e-16	***
## year_of_observation1995	-5.9144495	0.7279900	-8.1244	1.182e-15	***
## year_of_observation1996	-6.2034475	0.7684654	-8.0725	1.767e-15	***
## year_of_observation1997	-6.2335368	0.7900639	-7.8899	7.164e-15	***
## year_of_observation1998	-6.7303628	0.7976212	-8.4380	< 2.2e-16	***
## year_of_observation1999	-6.7627554	0.8029432	-8.4225	< 2.2e-16	***
## year_of_observation2000	-7.2287115	0.8134468	-8.8865	< 2.2e-16	***

```

## year_of_observation2001 -6.6491671 0.8188523 -8.1201 1.221e-15 ***
## year_of_observation2002 -5.7271889 0.8189238 -6.9936 4.604e-12 ***
## year_of_observation2003 -5.7483727 0.8241253 -6.9751 5.221e-12 ***
## year_of_observation2004 -6.1513452 0.8464757 -7.2670 6.884e-13 ***
## blood_alcohol_level10 0.1537963 0.2521919 0.6098 0.5420927
## blood_alcohol_levelnone 0.7707947 0.3871136 1.9911 0.0467093 *
## per_se_law -1.1673580 0.2409474 -4.8449 1.445e-06 ***
## primary_seatbelt -1.2097495 0.3567874 -3.3907 0.0007216 ***
## secondary_seatbelt -0.2584527 0.2604218 -0.9924 0.3211983
## sl70plus 0.2016010 0.2805965 0.7185 0.4726158
## grad_driver_law -0.5331449 0.3028770 -1.7603 0.0786357 .
## percent_age_14_24 0.2684700 0.0979019 2.7422 0.0061997 **
## unemployment_rate -0.6960810 0.0608561 -11.4381 < 2.2e-16 ***
## vehicle_miles -0.0062848 0.0065981 -0.9525 0.3410406
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Total Sum of Squares: 48612
## Residual Sum of Squares: 4840.4
## R-Squared: 0.90043
## Adj. R-Squared: 0.89321
## F-statistic: 1333.08 on 82 and 1118 DF, p-value: < 2.22e-16

```

- What do you estimate for coefficients on the blood alcohol variables? How do the coefficients on the blood alcohol variables change, if at all?
- What do you estimate for coefficients on per se laws? How do the coefficients on per se laws change, if at all?
- What do you estimate for coefficients on primary seat-belt laws? How do the coefficients on primary seatbelt laws change, if at all?

Which set of estimates do you think is more reliable? Why do you think this?

- What assumptions are needed in each of these models?
- Are these assumptions reasonable in the current context?

## 6 (10 points) Consider a Random Effects Model

Instead of estimating a fixed effects model, should you have estimated a random effects model?

- Please state the assumptions of a random effects model, and evaluate whether these assumptions are met in the data.
- If the assumptions are, in fact, met in the data, then estimate a random effects model and interpret the coefficients of this model. Comment on how, if at all, the estimates from this model have changed compared to the fixed effects model.
- If the assumptions are **not** met, then do not estimate the data. But, also comment on what the consequences would be if you were to *inappropriately* estimate a random effects model. Would your coefficient estimates be biased or not? Would your standard error estimates be biased or not? Or, would there be some other problem that might arise?

## 7 (10 points) Model Forecasts

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.

## 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?