# Stat/CS 187: Homework 4

yourname

# Spring 2022

# Contents

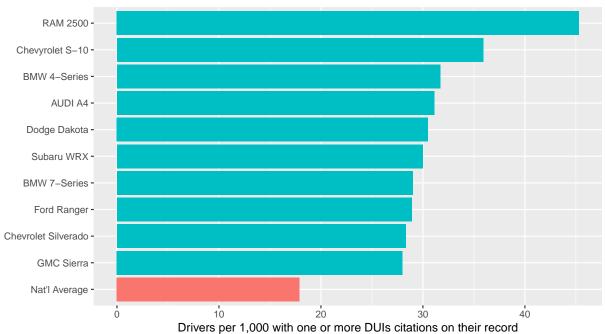
```
## Set the default size of figures
knitr::opts_chunk$set(echo = TRUE)
knitr::opts_chunk$set(echo = TRUE,
                   fig.width=8,
                   fig.height=5)
# Load the packages we will be using
pacman::p_load(hexmode,ggthemes)
## Warning: package 'hexmode' is not available for this version of R
## A version of this package for your version of R might be available elsewhere,
## see the ideas at
## https://cran.r-project.org/doc/manuals/r-patched/R-admin.html#Installing-packages
## Warning: unable to access index for repository http://www.stats.ox.ac.uk/pub/RWin/bin/windows/contri
    cannot open URL 'http://www.stats.ox.ac.uk/pub/RWin/bin/windows/contrib/4.1/PACKAGES'
## Warning in p_install(package, character.only = TRUE, ...):
## Warning in library(package, lib.loc = lib.loc, character.only = TRUE,
## logical.return = TRUE, : there is no package called 'hexmode'
## Warning in pacman::p_load(hexmode, ggthemes): Failed to install/load:
## hexmode
library(dplyr)
library(ggplot2)
library(ggthemes)
library(readr)
```

## 1. Insurify DUI Graph

```
31.1,
"AUDI A4",
                                  "Yes",
 "Dodge Dakota",
                     30.5,
                                  "Yes",
 "Subaru WRX",
                                  "Yes",
                      30.0,
 "BMW 7-Series",
                     29.0,
                                 "Yes",
 "Ford Ranger",
                     28.9,
                                 "Yes",
                                  "Yes",
 "Chevrolet Silverado", 28.3,
 "GMC Sierra", 28.0,
                                  "Yes",
 "Nat'l Average",
                                  "No"
                     17.9,
```

#### Part 1a)

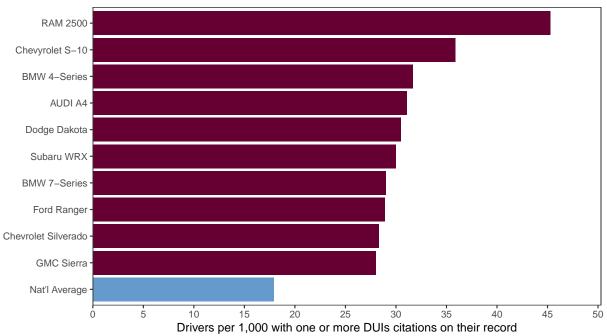
ISURIFY
Car Models with the Most DUIs



Source: Insurify analysis of over 2.7 million car insuarance applications

# Part 1b)

ISURIFY
Car Models with the Most DUIs



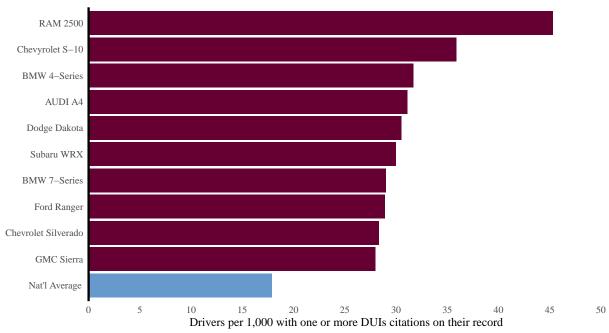
Source: Insurify analysis of over 2.7 million car insuarance applications

# Part 1c)

```
# change theme for non-data ink options
gg_dui2 + theme_tufte(ticks=FALSE)+
# add vertical line at x=0
    theme(axis.line.y = element_line(size=1,colour="black",linetype=1),
# change color & size of subtitle and size & position of subtitle
        plot.title= element_text(size=12,colour="#FF6633"),
        plot.subtitle = element_text(size=20,hjust=0.5))
```

#### **ISURIFY**

# Car Models with the Most DUIs



Source: Insurify analysis of over 2.7 million car insuarance applications

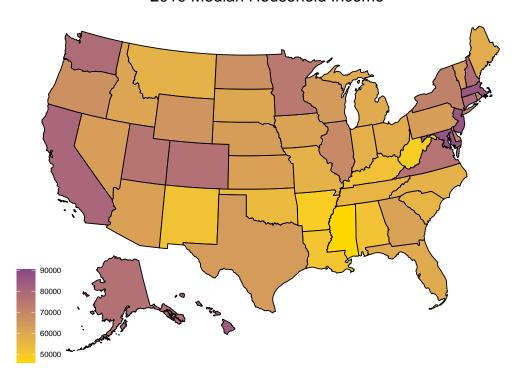
# 2. Mapping income and unemployment

```
# Read in data files:
US_2019_C <- read_csv("USDA_2019_County.csv")</pre>
## Rows: 3144 Columns: 9
## -- Column specification
## Delimiter: ","
## chr (2): stabb, state
## dbl (7): FIPS, Civilian_labor_force_2019, Employed_2019, Unemployed_2019, Un...
## i Use `spec()` to retrieve the full column specification for this data.
## i Specify the column types or set `show_col_types = FALSE` to quiet this message.
US_2019_S <- read_csv("USDA_2019_State.csv")</pre>
## Rows: 51 Columns: 8
## -- Column specification
## Delimiter: ","
## chr (3): FIPS, StAbb, state
## dbl (5): Civilian_labor_force_2019, Employed_2019, Unemployed_2019, Unemploy...
## i Use `spec()` to retrieve the full column specification for this data.
## i Specify the column types or set `show_col_types = FALSE` to quiet this message.
## Set the default size of figures
knitr::opts_chunk$set(fig.width=8, fig.height=5)
# Changing the default theme
theme_set(theme_bw())
```

```
# Load the libraries we will be using
pacman::p_load(gapminder,socviz, tidyverse, grid,usmap, maps, statebins, viridis, leaflet)
# A map theme provided by our book:
theme_map <- function(base_size=9, base_family="") {</pre>
   require(grid)
   theme bw(base size=base size, base family=base family) %+replace%
        theme(axis.line=element blank(),
              axis.text=element_blank(),
              axis.ticks=element_blank(),
              axis.title=element_blank(),
              panel.background=element_blank(),
              panel.border=element_blank(),
              panel.grid=element_blank(),
              panel.spacing=unit(0, "lines"),
              plot.background=element_blank(),
              legend.justification = c(0,0),
              legend.position = c(0,0)
              )
}
```

## 2a) Simple US Map

# 2019 Median Household Income

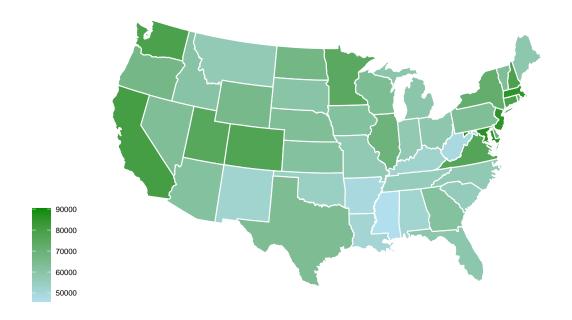


#### 2b) For more serious maps: Merge state lines info with USDA data.

```
# create state_lines using the map_data function like we did in class
state_lines <- map_data("state")</pre>
head(state_lines)
##
                    lat group order region subregion
          long
## 1 -87.46201 30.38968
                        1 1 alabama
                                                 <NA>
## 2 -87.48493 30.37249
                                 2 alabama
                                                 <NA>
                           1
## 3 -87.52503 30.37249
                           1
                                 3 alabama
                                                 <NA>
## 4 -87.53076 30.33239
                                4 alabama
                                                 <NA>
                          1
## 5 -87.57087 30.32665
                                  5 alabama
                                                 <NA>
## 6 -87.58806 30.32665
                                  6 alabama
                                                 <NA>
                            1
# clean data and change region to state for merge
state_lines<- state_lines%>%
  mutate(state=region)%>%
  select(-region,-subregion)
# change state to all lowercase
US_2019_S$state<-tolower(US_2019_S$state)</pre>
# merge data
US <- US_2019_S %>%
 left_join(y = state_lines,
           by = 'state')
```

### 2c) Map a Numeric variable by State: Median Income in 2019

## 2019 Median Household Income



## 2d) Now map data at a county level. First, the data, merged with county map info

```
# county_map is a data frame in socviz -- do help(county_map) to see more
?county_map
```

## starting httpd help server ... done

```
#create a new data frame using county_map data
new_cmap<-county_map%>%
  select all()%>%
# create a variable called FIPS
 mutate(FIPS = id)%>%
# drop id
 select(-id)
# change FIPS to numeric values
new_cmap$FIPS <- as.numeric(new_cmap$FIPS)</pre>
# merge the two together
US2 <- US_2019_C %>%
 left_join(y = new_cmap,
            by = 'FIPS')
# Print the first 6 rows of the resulting data set
head(US2)
## # A tibble: 6 x 15
```

```
##
     FIPS stabb state
                                  Civilian_labor_f~ Employed_2019 Unemployed_2019
## <dbl> <chr> <chr>
                                              <dbl>
                                                           <dbl>
                                                                           <dbl>
## 1 1001 AL
                                                           25458
                                                                             714
               Autauga County, AL
                                              26172
## 2 1001 AL Autauga County, AL
                                              26172
                                                           25458
                                                                             714
## 3 1001 AL Autauga County, AL
                                              26172
                                                           25458
                                                                             714
## 4 1001 AL Autauga County, AL
                                              26172
                                                           25458
                                                                             714
## 5 1001 AL Autauga County, AL
                                              26172
                                                           25458
                                                                             714
               Autauga County, AL
## 6 1001 AL
                                              26172
                                                           25458
                                                                             714
## # ... with 9 more variables: Unemployment_rate_2019 <dbl>,
## # Median_Household_Income_2019 <dbl>,
      Med_HH_Income_Percent_of_State_Total_2019 <dbl>, long <dbl>, lat <dbl>,
## #
      order <int>, hole <lgl>, piece <fct>, group <fct>
```

### 2e) Map Median Household Income by county:

Map income by county, matching the map below, in terms of colors (pretty close is fine), title, legend label, font size, theme, etc.



