

Project 1: Wrangling, Exploration, Visualization

SDS322E

Data Wrangling, Exploration, Visualization

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Introduction Paragraph or two introducing your datasets and variables, why they are interesting to you, etc.

```
# read your datasets in here, e.g., with read_csv()
library(tidyverse)
library(readr)
library(fivethirtyeight)
data1 <- bad_drivers
data2 <- read_csv("~/project1/2021 County Health Rankings Data - v1.csv",
  skip = 1)
```

I chose the bad_drivers dataset from the fivethirtyeight package and the county health ratings from Countyhealthratings.org (under the Rankings data and document page). I wanted to look at these datasets because I am very big on keeping up with the news. One thing that I always see are news articles on drunk driving, speeding, or just not paying attention lead to people losing their lives due to these accidents. I know that these are just a few factors; therefore, I wanted to bring in the County health rankings data because the data provides more insight into the health aspect of these various counties in each state of the United States.

Some of the variables in my first dataset of “bad_drivers” include number of drivers involved in fatal collisions, percentage of those speeding, alcohol-impaired, not distracted, no prior accidents before this collision, cost of car insurance, and losses experienced by insurance companies across all 50 states. Furthermore, some of the variables for my second dataset of “County_Health_Rankings” include life expectancy, deaths, percent of ethnicities, number of people insured, and more across counties in the 50 states in the United States. I would expect to see that if there were more traffic accidents the location and how populated the state is, could impact the number of traffic fatalities. Furthermore, if there was an increase in the number of people uninsured, this could influence how the insurance premium will come out to be.

```
# Joining the datasets
joineddata <- full_join(data1, data2, by = c(state = "State"))

# Finding which is missing from each dataset
not_in_data1 <- anti_join(data1, data2, by = c(state = "State"))
not_in_data2 <- anti_join(data2, data1, by = c(State = "state"))
```

Joining/Merging I combined both of my datasets by the common variable of “states.” To do this, I utilized the full_join function to retain all of my rows, which is why there was 3,193 rows, one row for each county in each state. Furthermore, the two datasets were joined by the common ID of “state.” When fully joined, there were 3,193 rows with 72 columns. Before the full join, the bad_drivers dataset had 51 rows of the 50 states in the U.S. as well as D.C. with 8 variables. Also before the join, the County health ratings data had 3,193 rows, for each county in the each state of the U.S. along with 65 variables. There were 51 unique

IDs, or states, for both the `bad_drivers` and county health rating datasets. Since both had only unique IDs of the 51 states, there were no IDs that were different from each other in both datasets. Since both datasets were joined by the common function of states, all 50 states in the United States including D.C. were included in both datasets. No observations were dropped and were actually retained by the full join function as well as the common ID being the same for each dataset. The NAs that are located in the full join and extra columns will be tidied. which will be shown above in the tidying section.

Tidying: Reshaping If your datasets are tidy already, demonstrate that you can reshape data with pivot wider/longer here (e.g., untidy and then retidy). Alternatively, it may be easier to wait until the wrangling section so you can reshape your summary statistics. Note here if you are going to do this.

```
# your tidying code (if applicable; can also wait until
# wrangling section)
clean_joineddata <- joineddata %>% select(state, num_drivers,
  perc_alcohol, insurance_premiums, County, `Life Expectancy`,
  `# Deaths`, `% Frequent Physical Distress`, `# Uninsured`,
  `Median Household Income`, `Homicide Rate`, `# Asian`, `# American Indian & Alaska Native`,
  `# Native Hawaiian/Other Pacific Islander`, `# Hispanic`,
  `# Black`, `# Non-Hispanic White`, `% Rural`)

clean_joineddata <- clean_joineddata %>% rename(DriverDeaths = num_drivers,
  AlcoholRelated = perc_alcohol, LE = `Life Expectancy`, Deaths = `# Deaths`,
  PhysicalDistress = `% Frequent Physical Distress`, Uninsured = `# Uninsured`,
  HouseholdIncome = `Median Household Income`, Homicide = `Homicide Rate`,
  Asian = `# Asian`, AmericanIndianAlaskaNative = `# American Indian & Alaska Native`,
  PacificIslander = `# Native Hawaiian/Other Pacific Islander`,
  Hispanic = `# Hispanic`, Black = `# Black`, White = `# Non-Hispanic White`,
  PercRural = `% Rural`, Premiums = insurance_premiums)

clean_joineddata <- clean_joineddata %>% group_by(state) %>%
  summarize_at(c("LE", "Deaths", "DriverDeaths", "AlcoholRelated",
    "PhysicalDistress", "Uninsured", "HouseholdIncome", "Homicide",
    "PercRural", "Premiums", "Asian", "AmericanIndianAlaskaNative",
    "PacificIslander", "Hispanic", "Black", "White"), .funs = list(mean = mean))

clean_joineddata2 <- clean_joineddata %>% pivot_longer(-1) %>%
  separate(name, into = c("name", "stat")) %>% pivot_wider(names_from = "name",
    values_from = "value") %>% select(-stat) %>% mutate(Rural = ifelse(PercRural >
    50, "yes", "no"))
```

Since there were so many different variables, I wanted to just select 10 different unique variables and then variables of different ethnicities. I wanted to untidy and tidy my data first before I started on the wrangling portion, and I just wanted to clean up the names before moving on. In this case, I had to first select the columns I wanted and then remove the extraneous ones like percent of the population over 65, graduation rate, GPA, child mortality rate, infant mortality, etc. There were over 72 columns with the data joined. With this, I made a cleaner version of the `joineddata` (`clean_joineddata`). I then wanted to just make to create a new categorical variable of whether one state would be considered rural or not. Since the variable `Rural` was in percentages, I mutated the variable with the `ifelse` function to say that if the average `'PercRural'` was greater than 50%, then we can say that it is considered rural. If not, then it wasn't considered rural. I then renamed all of the columns because I was having difficulty when it came to tidying the data with the special characters like %, __, and # signs. I then used the `pivot_longer` and `pivot_wider` functions to show that I know how to tidy. Since there were so many repeats of states because of the various counties, I went ahead and actually made new columns using `pivot_longer` and `pivot_wider` to reshape my data to show the averages across the states instead. With these, I found the means of each state with my numeric

variables, including one where I mutated the function ‘% Rural’ to give me a yes if the percentage of rural areas was greater than 50%, and no if otherwise. Through this, I was able to get better visualizations of my data because the plots looked really messy otherwise.

```
# Counting NAs
clean_joineddata2 %>% summarize_all(function(x) sum(is.na(x)))
```

Wrangling

```
## # A tibble: 1 x 18
##   state    LE Deaths DriverDeaths AlcoholRelated PhysicalDistress Uninsured
##   <int> <int>  <int>         <int>          <int>          <int>      <int>
## 1     0    17    14             0              0              0          1
## # ... with 11 more variables: HouseholdIncome <int>, Homicide <int>,
## #   PercRural <int>, Premiums <int>, Asian <int>,
## #   AmericanIndianAlaskaNative <int>, PacificIslander <int>, Hispanic <int>,
## #   Black <int>, White <int>, Rural <int>
```

When I used the summary function, I found that there was 17 NAs in Life Expectancy, 14 NAs in Deaths, 48 NAs in homicide rates, 2 NAs in percent of counties that are rural, 1 NA in number of people uninsured and median household income, 0 NAs in state, Driver fatalities, Alcohol related traffic fatalities, physical distress, insurance premiums, number of Asians, American Indian & Alaska Native, Pacific Islander, Hispanic, Black, and White.

```
# % of BIPOC in each county
clean_joineddata2 <- clean_joineddata2 %>% group_by(state) %>%
  mutate(`% BIPOC` = (sum(Asian, AmericanIndianAlaskaNative,
    PacificIslander, Hispanic, Black))/(sum(Asian, AmericanIndianAlaskaNative,
    PacificIslander, Hispanic, Black, White)) * 100)

clean_joineddata2 <- clean_joineddata2 %>% mutate(Diverse = ifelse(`% BIPOC` >
  50, "high", "low"))
```

First, I wanted to create a new variable that would tell us the percentage of many people of color make up the population in each state. To do this, I utilized the mutate function to sum up BIPOC populations, dividing it by the sum of the “BIPOC” populations and “White” populations, and then multiplying by 100. This gave me the percentage of BIPOC in each state within the U.S. I then utilized mutate again and the ifelse function to tell me if these states were considered diverse. If their percentages were greater than 50% then I considered it as high.

```
# Summary statistics for 10 numeric variables
clean_joineddata2 %>% summarize_at(c("DriverDeaths", "AlcoholRelated",
  "Premiums", "LE", "Deaths", "PhysicalDistress", "Uninsured",
  "HouseholdIncome", "Homicide", "PercRural", "Asian", "AmericanIndianAlaskaNative",
  "PacificIslander", "Hispanic", "Black", "White"), na.rm = TRUE,
  .funs = list(mean = mean, sd = sd, max = max, min = min,
    n = n_distinct)) %>% pivot_longer(contains("_")) %>%
  separate(name, into = c("variable", "stat")) %>% pivot_wider(names_from = "variable",
    values_from = "value") %>% knitr::kable()
```

state	stat	DriverDeaths	AlcoholRelated	Premiums	LE	Deaths	PhysicalDistress	Uninsured	HouseholdIncome	Homicide	PercRural	Asian	AmericanIndianAlaskaNative	PacificIslander	Hispanic	Black	White
Alabama	mean	18.8	30	784.574	69.7069	0.8572	205.062	2.44622	9.16	NA	66.6925	68.0882	3.5000	150.294	40.267	0.88169	8.51292
Alabama	sd	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA
Alabama	max	18.8	30	784.574	69.7069	0.8572	205.062	2.44622	9.16	-	66.6925	68.0882	3.5000	150.294	40.267	0.88169	8.51292
										Inf							

state	stat	Driver	Deaths	Physical	Disse	Hed	Blm	Pol	R	American	Ind	Al	His	Blk	White										
Alabama	min	18.8	30	784.574	697	0.085	27	205	0.062	44	229	1	Inf	66.692	53.082	23.5000	150.294	63	67.088	169.841	22.00				
Alabama	n	1.0	1	1.00	1.0000	0.0000	1.0000	1.0000	1.00	0.0000	0.0000	0.0000	1.0000	1.0000	0.000	1.0000	1.000	1.000	1.000	1.00					
Alaska	meal	18.1	25	1053.88	NA	NA	14.3333	381	24.06	67	24.17	NA	NA	NA	3186.67	06.8667	699.866	67	47.46	12.62	38.53				
Alaska	sd	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA				
Alaska	max	18.1	25	1053.48	-	-	14.3333	381	24.06	67	24.17	-	-	-	3186.67	06.8667	699.866	67	47.46	12.62	38.53				
				Inf	Inf							Inf	Inf												
Alaska	min	18.1	25	1053.88	Inf	Inf	14.3333	381	24.06	67	24.17	NA	Inf	Inf	3186.67	06.8667	699.866	67	47.46	12.62	38.53				
Alaska	n	1.0	1	1.00	0.0000	0.0000	1.0000	1.0000	1.00	0.0000	0.0000	0.0000	1.0000	1.0000	0.000	1.0000	1.000	1.000	1.000	1.00					
Arizona	meal	18.6	28	899.478	662	0.050	47	508	750	881.52	004.12	NA	NA	32.613	375	90.48	237.875	025	21.12	388	23	075	22.42	24.61	25
Arizona	sd	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA
Arizona	max	18.6	28	899.478	662	0.050	47	508	750	881.52	004.12	-	-	32.613	375	90.48	237.875	025	21.12	388	23	075	22.42	24.61	25
				Inf								Inf													
Arizona	min	18.6	28	899.478	662	0.050	47	508	750	881.52	004.12	NA	Inf	32.613	375	90.48	237.875	025	21.12	388	23	075	22.42	24.61	25
Arizona	n	1.0	1	1.00	1.0000	0.0000	1.0000	1.0000	1.00	0.0000	0.0000	0.0000	1.0000	1.0000	0.000	1.0000	1.000	1.000	1.000	1.00					
Arkansas	meal	22.4	26	827.375	015	769.81	68	184	28.34	441	155.22	NA	NA	64.598	826.39	08.3421	310.157	62	27.1	22	42.3	42	06.53		
Arkansas	sd	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA
Arkansas	max	22.4	26	827.375	015	769.81	68	184	28.34	441	155.22	-	-	64.598	826.39	08.3421	310.157	62	27.1	22	42.3	42	06.53		
				Inf								Inf													
Arkansas	min	22.4	26	827.375	015	769.81	68	184	28.34	441	155.22	NA	Inf	64.598	826.39	08.3421	310.157	62	27.1	22	42.3	42	06.53		
Arkansas	n	1.0	1	1.00	1.0000	0.0000	1.0000	1.0000	1.00	0.0000	0.0000	0.0000	1.0000	1.0000	0.000	1.0000	1.000	1.000	1.000	1.00					
California	meal	12.0	28	878.4	NA	11885.13	25	768	0038.81	1367.44	NA	NA	28.292	631	50	220	203	6773.05	62	796	7	53	40.48	97	40.61
California	sd	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA
California	max	12.0	28	878.41	-	11885.13	25	768	0038.81	1367.44	-	-	28.292	631	50	220	203	6773.05	62	796	7	53	40.48	97	40.61
				Inf								Inf													
California	min	12.0	28	878.4	Inf	11885.13	25	768	0038.81	1367.44	Inf	Inf	28.292	631	50	220	203	6773.05	62	796	7	53	40.48	97	40.61
California	n	1.0	1	1.00	0.0000	0.0000	1.0000	1.0000	1.00	0.0000	0.0000	0.0000	1.0000	1.0000	0.000	1.0000	1.000	1.000	1.000	1.00					
Colorado	meal	13.6	28	835.5	NA	NA	11.1538	5922.86	2882.68	NA	NA	NA	57.996	735.82	50.9846	348.584	32	673.96	85.32	38	80.09				
Colorado	sd	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA
Colorado	max	13.6	28	835.50	-	-	11.1538	5922.86	2882.68	-	-	-	57.996	735.82	50.9846	348.584	32	673.96	85.32	38	80.09				
				Inf	Inf							Inf													
Colorado	min	13.6	28	835.5	Inf	Inf	11.1538	5922.86	2882.68	Inf	Inf	57.996	735.82	50.9846	348.584	32	673.96	85.32	38	80.09					
Colorado	n	1.0	1	1.00	0.0000	0.0000	1.0000	1.0000	1.00	0.0000	0.0000	0.0000	1.0000	1.0000	0.000	1.0000	1.000	1.000	1.000	1.00					
Connecticut	meal	10.8	36	1068.8	0.56	703.6	66	222	2362.22	2251.22	333	323	1391	394	778.4444	862.000	0	354	8	596	3	522	249.56		
Connecticut	sd	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA
Connecticut	max	10.8	36	1068.8	0.56	703.6	66	222	2362.22	2251.22	333	323	1391	394	778.4444	862.000	0	354	8	596	3	522	249.56		
Connecticut	min	10.8	36	1068.8	0.56	703.6	66	222	2362.22	2251.22	333	323	1391	394	778.4444	862.000	0	354	8	596	3	522	249.56		
Connecticut	n	1.0	1	1.00	1.0000	0.0000	1.0000	1.0000	1.00	1.0000	0.0000	0.0000	1.0000	1.0000	0.000	1.0000	1.000	1.000	1.000	1.00					
Delaware	meal	16.2	30	1137.88	356	0028.51	02	500	831.56	0516.75	250	22	41	000	31.32	82.0000	529.500	06	695.50	73	29	3	000	04.50	
Delaware	sd	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA
Delaware	max	16.2	30	1137.88	356	0028.51	02	500	831.56	0516.75	250	22	41	000	31.32	82.0000	529.500	06	695.50	73	29	3	000	04.50	
Delaware	min	16.2	30	1137.88	356	0028.51	02	500	831.56	0516.75	250	22	41	000	31.32	82.0000	529.500	06	695.50	73	29	3	000	04.50	
Delaware	n	1.0	1	1.00	1.0000	0.0000	1.0000	1.0000	1.00	1.0000	0.0000	0.0000	1.0000	1.0000	0.000	1.0000	1.000	1.000	1.000	1.00					
District of Columbia	meal	5.9	27	1273.89	008	280.00	0000	0220.00	00	895.01	8.000	000	030	502.01	30.0000	923.000	0	70	477.30	32	29	26	000.00		
District of Columbia	sd	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA
District of Columbia	max	5.9	27	1273.89	008	280.00	0000	0220.00	00	895.01	8.000	000	030	502.01	30.0000	923.000	0	70	477.30	32	29	26	000.00		

state	stat	Driver	Deaths	Physical	Disse	Hes	Holm	Pol	Asian	American	Ind	Pacific	Al	Hispan	Black	White
District of Columbia	min	5.9	27	1273.89	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00
District of Columbia	n	1.0	1	1.00	1.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00
Florida	mean	17.9	29	1160.18	1.37	292.85	254419	740.64	7362.15	NaN	37.07	98671.32	24.41	118	724.73	55658.48
Florida	sd	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA
Florida	max	17.9	29	1160.18	1.37	292.85	254419	740.64	7362.15	-	37.07	98671.32	24.41	118	724.73	55658.48
Florida	min	17.9	29	1160.18	1.37	292.85	254419	740.64	7362.15	Inf	37.07	98671.32	24.41	118	724.73	55658.48
Florida	n	1.0	1	1.00	1.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00
Georgia	mean	15.6	25	913.15	NaN	1663.56	257812	2583.02	2582.28	NaN	60.26	7709.57	750.10	100	155.13	753109.45
Georgia	sd	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA
Georgia	max	15.6	25	913.15	-	1663.56	257812	2583.02	2582.28	-	60.26	7709.57	750.10	100	155.13	753109.45
Georgia	min	15.6	25	913.15	Inf	1663.56	257812	2583.02	2582.28	Inf	60.26	7709.57	750.10	100	155.13	753109.45
Georgia	n	1.0	1	1.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00
Hawaii	mean	17.5	41	861.18	NaN	NaN	10.83	333	NaN	NaN	29.08	373396.85	67.66	67	47849.65	6028.92
Hawaii	sd	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA
Hawaii	max	17.5	41	861.18	-	-	10.83	333	-	-	Inf	29.08	373396.85	67.66	67	47849.65
Hawaii	min	17.5	41	861.18	Inf	Inf	10.83	333	Inf	Inf	Inf	29.08	373396.85	67.66	67	47849.65
Hawaii	n	1.0	1	1.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00
Idaho	mean	15.3	29	641.96	NaN	NaN	13.00	0.00	248.45	58955.73	NaN	60.75	3231.60	383.11	111	175.28
Idaho	sd	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA
Idaho	max	15.3	29	641.96	-	-	13.00	0.00	248.45	58955.73	-	60.75	3231.60	383.11	111	175.28
Idaho	min	15.3	29	641.96	Inf	Inf	13.00	0.00	248.45	58955.73	Inf	60.75	3231.60	383.11	111	175.28
Idaho	n	1.0	1	1.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00
Illinois	mean	12.8	34	803.17	7.99	2300.60	295631	927.25	59944.76	NaN	50.70	68018.25	601.01	94	162.81	543104.30
Illinois	sd	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA
Illinois	max	12.8	34	803.17	7.99	2300.60	295631	927.25	59944.76	-	50.70	68018.25	601.01	94	162.81	543104.30
Illinois	min	12.8	34	803.17	7.99	2300.60	295631	927.25	59944.76	Inf	50.70	68018.25	601.01	94	162.81	543104.30
Illinois	n	1.0	1	1.00	1.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00
Indiana	mean	14.5	29	710.46	6.99	7853.46	240432	286.75	59735.41	NaN	54.16	3752.06	49.46	24	97.61	290523.72
Indiana	sd	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA
Indiana	max	14.5	29	710.46	6.99	7853.46	240432	286.75	59735.41	-	54.16	3752.06	49.46	24	97.61	290523.72
Indiana	min	14.5	29	710.46	6.99	7853.46	240432	286.75	59735.41	Inf	54.16	3752.06	49.46	24	97.61	290523.72
Indiana	n	1.0	1	1.00	1.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00
Iowa	mean	15.7	25	649.06	9.13	7005.58	0.24	0.00	0.82	59755.50	NaN	61.06	682.33	40.20	96.08	0.00
Iowa	sd	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA
Iowa	max	15.7	25	649.06	9.13	7005.58	0.24	0.00	0.82	59755.50	-	61.06	682.33	40.20	96.08	0.00
Iowa	min	15.7	25	649.06	9.13	7005.58	0.24	0.00	0.82	59755.50	Inf	61.06	682.33	40.20	96.08	0.00
Iowa	n	1.0	1	1.00	1.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00
Kansas	mean	17.8	24	780.45	NaN	NaN	12.09	43485.41	51039.05	NaN	66.50	17892.16	62.56	60	69.56	600971.02
Kansas	sd	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA
Kansas	max	17.8	24	780.45	-	-	12.09	43485.41	51039.05	-	66.50	17892.16	62.56	60	69.56	600971.02
Kansas	min	17.8	24	780.45	Inf	Inf	12.09	43485.41	51039.05	Inf	66.50	17892.16	62.56	60	69.56	600971.02

state	stat	Driver	Deaths	Pedestrian	Infants	Deaths	Physical	Disability	Health	Home	Police	Asians	American	Indian	Pacific	Hispanic	Black	White
Kansas	min	17.8	24	780.4	Inf	Inf	12.0943	485.41	51039.0	Inf	66.5011	892.16	62.5660	69.5660	6718.33	57.07	1550.25	
Kansas	n	1.0	1	1.00	0.0000	0.0000	1.0000	1.0000	1.00	0.0000	0.0000	0.0000	1.0000	1.0000	0.0000	1.0000	1.00	
Kentucky	mean	21.4	23	872.5	714.63	228.7	76.24	79639.83	47490.4	NaN	71.4000	81.82	23.5537	69.2892	2887.70	80.97	239.88	
Kentucky	sd	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	
Kentucky	max	21.4	23	872.5	714.63	228.7	76.24	79639.83	47490.4	-	71.4000	81.82	23.5537	69.2892	2887.70	80.97	239.88	
Inf																		
Kentucky	min	21.4	23	872.5	714.63	228.7	76.24	79639.83	47490.4	Inf	71.4000	81.82	23.5537	69.2892	2887.70	80.97	239.88	
Kentucky	n	1.0	1	1.00	1.0000	0.0000	1.0000	1.0000	1.00	0.0000	0.0000	0.0000	1.0000	1.0000	0.0000	1.0000	1.00	
Louisiana	mean	20.5	33	1281.7	55.54	2328.4	3083077	148.03	826.60	NaN	50.7123	90.52	36.0308	87.8153	7599.14	184.83	57.14	
Louisiana	sd	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	
Louisiana	max	20.5	33	1281.7	55.54	2328.4	3083077	148.03	826.60	-	50.7123	90.52	36.0308	87.8153	7599.14	184.83	57.14	
Inf																		
Louisiana	min	20.5	33	1281.7	55.54	2328.4	3083077	148.03	826.60	Inf	50.7123	90.52	36.0308	87.8153	7599.14	184.83	57.14	
Louisiana	n	1.0	1	1.00	1.0000	0.0000	1.0000	1.0000	1.00	0.0000	0.0000	0.0000	1.0000	1.0000	0.0000	1.0000	1.00	
Maine	mean	15.1	30	661.8	88.37	3178.8	239411	8715.5	20490.1	NaN	73.4322	99.41	149.0588	53.5294	2788.23	35.76	47011.41	
Maine	sd	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	
Maine	max	15.1	30	661.8	88.37	3178.8	239411	8715.5	20490.1	-	73.4322	99.41	149.0588	53.5294	2788.23	35.76	47011.41	
Inf																		
Maine	min	15.1	30	661.8	88.37	3178.8	239411	8715.5	20490.1	Inf	73.4322	99.41	149.0588	53.5294	2788.23	35.76	47011.41	
Maine	n	1.0	1	1.00	1.0000	0.0000	1.0000	1.0000	1.00	0.0000	0.0000	0.0000	1.0000	1.0000	0.0000	1.0000	1.00	
Maryland	mean	12.5	32	1048.7	8.51	2601.8	402000	3542.5	60016.0	NaN	33.7322	504.2	90.2000	536.4000	505.7	6082.2	32062.48	
Maryland	sd	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	
Maryland	max	12.5	32	1048.7	8.51	2601.8	402000	3542.5	60016.0	-	33.7322	504.2	90.2000	536.4000	505.7	6082.2	32062.48	
Inf																		
Maryland	min	12.5	32	1048.7	8.51	2601.8	402000	3542.5	60016.0	Inf	33.7322	504.2	90.2000	536.4000	505.7	6082.2	32062.48	
Maryland	n	1.0	1	1.00	1.0000	0.0000	1.0000	1.0000	1.00	0.0000	0.0000	0.0000	1.0000	1.0000	0.0000	1.0000	1.00	
Massachusetts	mean	18.2	35	1011.8	44.40	6679.7	3333260.1	83379.0	7	NaN	16.1660	835.4	500.6667	988.0000	103987	6004.2	53740.00	
Massachusetts	sd	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	
Massachusetts	max	18.2	35	1011.8	44.40	6679.7	3333260.1	83379.0	7	-	16.1660	835.4	500.6667	988.0000	103987	6004.2	53740.00	
Inf																		
Massachusetts	min	18.2	35	1011.8	44.40	6679.7	3333260.1	83379.0	7	Inf	16.1660	835.4	500.6667	988.0000	103987	6004.2	53740.00	
Massachusetts	n	1.0	1	1.00	1.0000	0.0000	1.0000	1.0000	1.00	0.0000	0.0000	0.0000	1.0000	1.0000	0.0000	1.0000	1.00	
Michigan	mean	14.1	28	1110.7	3.10	3008.1	140904	8779.6	93695.4	NaN	61.2189	909.1	1738.6905	99.6666	72576.3	27035.3	7730.05	
Michigan	sd	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	
Michigan	max	14.1	28	1110.7	3.10	3008.1	140904	8779.6	93695.4	-	61.2189	909.1	1738.6905	99.6666	72576.3	27035.3	7730.05	
Inf																		
Michigan	min	14.1	28	1110.7	3.10	3008.1	140904	8779.6	93695.4	Inf	61.2189	909.1	1738.6905	99.6666	72576.3	27035.3	7730.05	
Michigan	n	1.0	1	1.00	1.0000	0.0000	1.0000	1.0000	1.00	0.0000	0.0000	0.0000	1.0000	1.0000	0.0000	1.0000	1.00	
Minnesota	mean	9.6	29	777.1	80.18	684.6	507500	997.5	969448.9	NaN	60.9464	57.8	8760.8864	95.7727	3162.0	8695.9	11367.02	
Minnesota	sd	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	
Minnesota	max	9.6	29	777.1	80.18	684.6	507500	997.5	969448.9	-	60.9464	57.8	8760.8864	95.7727	3162.0	8695.9	11367.02	
Inf																		
Minnesota	min	9.6	29	777.1	80.18	684.6	507500	997.5	969448.9	Inf	60.9464	57.8	8760.8864	95.7727	3162.0	8695.9	11367.02	
Minnesota	n	1.0	1	1.00	1.0000	0.0000	1.0000	1.0000	1.00	0.0000	0.0000	0.0000	1.0000	1.0000	0.0000	1.0000	1.00	
Mississippi	mean	17.6	31	896.0	7	NaN	1242.6	265349	4813.03	61753.6	NaN	70.3867	95.9515	0.7229	43.5180	7412.2	28934.4	
Mississippi	sd	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	
Mississippi	max	17.6	31	896.0	7	-	1242.6	265349	4813.03	61753.6	-	70.3867	95.9515	0.7229	43.5180	7412.2	28934.4	
Inf																		
Mississippi	min	17.6	31	896.0	7	Inf	1242.6	265349	4813.03	61753.6	Inf	70.3867	95.9515	0.7229	43.5180	7412.2	28934.4	
Mississippi	n	1.0	1	1.00	0.0000	0.0000	1.0000	1.0000	1.00	0.0000	0.0000	0.0000	1.0000	1.0000	0.0000	1.0000	1.00	
Missouri	mean	16.1	34	790.3	26.76	6381.3	743310	8404.4	63391.3	NaN	66.4822	228.4	628.9138	169.9482	32.8	97270.8	3450.21	
Missouri	sd	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	

state	stat	Driver	Deaths	Physical	Disability	Health	Blacks	Hispanic	American Indian	Pacific Islander	Black	White					
Missouri	max	6.1	34	790.326	766.538	1.374	3310	404.465	391.31	-	66.482	222.461	78.9138	169.948	2632.807	270.737	10.21
										Inf							
Missouri	min	16.1	34	790.326	766.538	1.374	3310	404.465	391.31	Inf	66.482	222.461	78.9138	169.948	2632.807	270.737	10.21
Missouri	n	1.0	1	1.00	1.0000	1.0000	1.0000	1.0000	1.00	0.0000	1.0000	1.0000	1.0000	1.0000	1.0000	1.0000	1.00
Montana	mean	21.4	44	816.2	NaN	NaN	12.263	1510.455	1629.2	NaN	74.573	468.017	494.5965	32.2807	1518.9	222.313	200.39
Montana	sd	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA
Montana	max	21.4	44	816.21	-	-	12.263	1510.455	1629.21	-	74.573	468.017	494.5965	32.2807	1518.9	222.313	200.39
					Inf	Inf				Inf							
Montana	min	21.4	44	816.2	Inf	Inf	12.263	1510.455	1629.2	Inf	74.573	468.017	494.5965	32.2807	1518.9	222.313	200.39
Montana	n	1.0	1	1.00	0.0000	1.0000	1.0000	1.0000	1.00	0.0000	1.0000	1.0000	1.0000	1.0000	1.0000	1.0000	1.00
Nebraska	mean	14.9	35	732.2	NaN	NaN	10.553	1927.873	105.5	NaN	74.141	631.66	83.0851	49.9787	4673.298	17.632	95.15
Nebraska	sd	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA
Nebraska	max	14.9	35	732.28	-	-	10.553	1927.873	105.59	-	74.141	631.66	83.0851	49.9787	4673.298	17.632	95.15
					Inf	Inf				Inf							
Nebraska	min	14.9	35	732.2	Inf	Inf	10.553	1927.873	105.5	Inf	74.141	631.66	83.0851	49.9787	4673.298	17.632	95.15
Nebraska	n	1.0	1	1.00	0.0000	1.0000	1.0000	1.0000	1.00	0.0000	1.0000	1.0000	1.0000	1.0000	1.0000	1.0000	1.00
Nevada	mean	14.7	32	1029.8	NaN	NaN	14.0000	6360.66	4713.1	NaN	47.072	22296.372	22.7778	2731.66	107063.687	9.554	81.44
Nevada	sd	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA
Nevada	max	14.7	32	1029.87	-	-	14.0000	6360.66	4713.17	-	47.072	22296.372	22.7778	2731.66	107063.687	9.554	81.44
					Inf	Inf				Inf							
Nevada	min	14.7	32	1029.87	Inf	Inf	14.0000	6360.66	4713.17	Inf	47.072	22296.372	22.7778	2731.66	107063.687	9.554	81.44
Nevada	n	1.0	1	1.00	0.0000	1.0000	1.0000	1.0000	1.00	0.0000	1.0000	1.0000	1.0000	1.0000	1.0000	1.0000	1.00
New Hampshire	mean	11.6	30	746.57	9.263	88.909	90909	142.000	50.0	NaN	53.947	337.27	48.0000	122.181	8925.276	46.121	897.64
New Hampshire	sd	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA
New Hampshire	max	11.6	30	746.57	9.263	88.909	90909	142.000	50.0	-	53.947	337.27	48.0000	122.181	8925.276	46.121	897.64
										Inf							
New Hampshire	min	11.6	30	746.57	9.263	88.909	90909	142.000	50.0	Inf	53.947	337.27	48.0000	122.181	8925.276	46.121	897.64
New Hampshire	n	1.0	1	1.00	1.0000	1.0000	1.0000	1.0000	1.00	0.0000	1.0000	1.0000	1.0000	1.0000	1.0000	1.0000	1.00
New Jersey	mean	11.2	28	1301.72	908335.909	90909	1980.363	192.68	NaN	12.428	77619.593	9.0000	930.909	168804.000	140099.55		
New Jersey	sd	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA
New Jersey	max	11.2	28	1301.72	908335.909	90909	1980.363	192.68	-	12.428	77619.593	9.0000	930.909	168804.000	140099.55		
									Inf								
New Jersey	min	11.2	28	1301.72	908335.909	90909	1980.363	192.68	Inf	12.428	77619.593	9.0000	930.909	168804.000	140099.55		
New Jersey	n	1.0	1	1.00	1.0000	1.0000	1.0000	1.0000	1.00	0.0000	1.0000	1.0000	1.0000	1.0000	1.0000	1.0000	1.00
New Mexico	mean	18.4	27	869.8	NaN	NaN	14.588	21598.823	632.9	NaN	47.672	2338.823	517.294	1196.529	60761.232	8.414	48.82
New Mexico	sd	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA
New Mexico	max	18.4	27	869.85	-	-	14.588	21598.823	632.91	-	47.672	2338.823	517.294	1196.529	60761.232	8.414	48.82
					Inf	Inf				Inf							
New Mexico	min	18.4	27	869.8	Inf	Inf	14.588	21598.823	632.9	Inf	47.672	2338.823	517.294	1196.529	60761.232	8.414	48.82
New Mexico	n	1.0	1	1.00	0.0000	1.0000	1.0000	1.0000	1.00	0.0000	1.0000	1.0000	1.0000	1.0000	1.0000	1.0000	1.00

state	stat	Driver	Deaths	Physical	Deaths	Physical	Deaths	Physical	Deaths	Physical	Deaths	Physical	Deaths	Physical	Deaths	Physical	Deaths	Physical	Deaths
New York	mean	2.3	29	1234.79	670369.68	2158732	278.63	4962.84	43.78	45657.00	23.74	6023.74	60	858.22	222908	82906.32	27041.90		
New York	sd	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	
New York	max	2.3	29	1234.79	670369.68	2158732	278.63	4962.84	43.78	45657.00	23.74	6023.74	60	858.22	222908	82906.32	27041.90		
New York	min	2.3	29	1234.79	670369.68	2158732	278.63	4962.84	43.78	45657.00	23.74	6023.74	60	858.22	222908	82906.32	27041.90		
New York	n	1.0	1	1.00	1.000000	0.000001	0.000001	0.000001	0.000001	0.000000	0.000000	0.000000	1.000000	1.000000	0.000001	0.000001	0.000001	0.00	
North Carolina	mean	6.8	31	708.27	7082601.52	1873227	527.62	3898.40	60.93	3689.73	279.14	85	260.00	200313	46368.43	35041.62			
North Carolina	sd	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	
North Carolina	max	6.8	31	708.27	7082601.52	1873227	527.62	3898.40	60.93	3689.73	279.14	85	260.00	200313	46368.43	35041.62			
North Carolina	min	6.8	31	708.27	7082601.52	1873227	527.62	3898.40	60.93	3689.73	279.14	85	260.00	200313	46368.43	35041.62			
North Carolina	n	1.0	1	1.00	1.000000	0.000001	0.000001	0.000001	0.000001	0.000000	0.000000	0.000000	1.000000	1.000000	0.000001	0.000001	0.000001	0.00	
North Dakota	mean	23.9	42	688.75	NA	NA	10.88	88429.29	63698.94	81.77	493.29	6372.77	778	22.77	778167.85	20.96	23611.59		
North Dakota	sd	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	
North Dakota	max	23.9	42	688.75	-	-	10.88	88429.29	63698.94	81.77	493.29	6372.77	778	22.77	778167.85	20.96	23611.59		
North Dakota	min	23.9	42	688.75	Inf	Inf	10.88	88429.29	63698.94	81.77	493.29	6372.77	778	22.77	778167.85	20.96	23611.59		
North Dakota	n	1.0	1	1.00	0.000000	0.000001	0.000001	0.000001	0.000001	0.000000	0.000000	0.000000	1.000000	1.000000	0.000001	0.000001	0.000001	0.00	
Ohio	mean	4.1	34	697.76	833799.93	2628098	77.65	2945.24	47.93	36710.07	74.26	97	158.53	93572.33	288.20	9047.01			
Ohio	sd	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	
Ohio	max	4.1	34	697.76	833799.93	2628098	77.65	2945.24	47.93	36710.07	74.26	97	158.53	93572.33	288.20	9047.01			
Ohio	min	4.1	34	697.76	833799.93	2628098	77.65	2945.24	47.93	36710.07	74.26	97	158.53	93572.33	288.20	9047.01			
Ohio	n	1.0	1	1.00	1.000000	0.000001	0.000001	0.000001	0.000001	0.000000	0.000000	0.000000	1.000000	1.000000	0.000001	0.000001	0.000001	0.00	
Oklahoma	mean	19.9	29	881.57	5.28	0570.43	538462	109.54	2900.10	63.25	2416.79	1365.48	72	220.02	51233.59	24.56	1958.67		
Oklahoma	sd	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	
Oklahoma	max	19.9	29	881.57	5.28	0570.43	538462	109.54	2900.10	63.25	2416.79	1365.48	72	220.02	51233.59	24.56	1958.67		
Oklahoma	min	19.9	29	881.57	5.28	0570.43	538462	109.54	2900.10	63.25	2416.79	1365.48	72	220.02	51233.59	24.56	1958.67		
Oklahoma	n	1.0	1	1.00	1.000000	0.000001	0.000001	0.000001	0.000001	0.000000	0.000000	0.000000	1.000000	1.000000	0.000001	0.000001	0.000001	0.00	
Oregon	mean	12.8	26	804.71	NA	NA	14.70	270639.67	5568.57	43.68	1066.97	68.37	84	1040.27	32740.47	85.40	5107.24		
Oregon	sd	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	
Oregon	max	12.8	26	804.71	-	-	14.70	270639.67	5568.57	43.68	1066.97	68.37	84	1040.27	32740.47	85.40	5107.24		
Oregon	min	12.8	26	804.71	Inf	Inf	14.70	270639.67	5568.57	43.68	1066.97	68.37	84	1040.27	32740.47	85.40	5107.24		
Oregon	n	1.0	1	1.00	0.000000	0.000001	0.000001	0.000001	0.000001	0.000000	0.000000	0.000000	1.000000	1.000000	0.000001	0.000001	0.000001	0.00	
Pennsylvania	mean	18.2	31	905.98	282856.47	262053	39.93	3240.63	47.05	1472.84	29.79	41	293.94	128416.47	18.23	5105.24			
Pennsylvania	sd	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	
Pennsylvania	max	18.2	31	905.98	282856.47	262053	39.93	3240.63	47.05	1472.84	29.79	41	293.94	128416.47	18.23	5105.24			
Pennsylvania	min	18.2	31	905.98	282856.47	262053	39.93	3240.63	47.05	1472.84	29.79	41	293.94	128416.47	18.23	5105.24			
Pennsylvania	n	1.0	1	1.00	1.000000	0.000001	0.000001	0.000001	0.000001	0.000000	0.000000	0.000000	1.000000	1.000000	0.000001	0.000001	0.000001	0.00	

state	stat	Driver	Deaths	Physical	Disturb	Hes	Holm	Pol	Russian	American	Ind	Pacific	Hispanic	Black	White		
Pennsylvania	mean	18.2	31	905.97	8.28	2856.47	72072053	9539.93	3240.63	Inf	47.0514	772.84	29.7941	293.9412	3416.47	18.3251	105.24
Pennsylvania	sd	1.0	1	1.00	1.00	0.00	0.00	0.00	1.00	0.00	0.00	0.00	1.00	1.00	0.00	1.00	1.00
Rhode Island	mean	1.1	38	1148.90	4.13	8874.38	36667489.37	3302.67	NaN	11.13	33878.38	67.6667	708.33	33548.20	655.23	3377.00	
Rhode Island	sd	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA
Rhode Island	max	1.1	38	1148.90	4.13	8874.38	36667489.37	3302.67	-	11.13	33878.38	67.6667	708.33	33548.20	655.23	3377.00	
Rhode Island	min	1.1	38	1148.90	4.13	8874.38	36667489.37	3302.67	Inf	11.13	33878.38	67.6667	708.33	33548.20	655.23	3377.00	
Rhode Island	n	1.0	1	1.00	1.00	0.00	0.00	0.00	1.00	0.00	0.00	0.00	1.00	1.00	0.00	1.00	1.00
South Carolina	mean	23.9	41	858.97	5.61	3848.12	2727662322.34	40452.64	NaN	54.52	40794.72	291.5319	215.5319	191068.57	86.83	369.87	
South Carolina	sd	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA
South Carolina	max	23.9	41	858.97	5.61	3848.12	2727662322.34	40452.64	-	54.52	40794.72	291.5319	215.5319	191068.57	86.83	369.87	
South Carolina	min	23.9	41	858.97	5.61	3848.12	2727662322.34	40452.64	Inf	54.52	40794.72	291.5319	215.5319	191068.57	86.83	369.87	
South Carolina	n	1.0	1	1.00	1.00	0.00	0.00	0.00	1.00	0.00	0.00	0.00	1.00	1.00	0.00	1.00	1.00
South Dakota	mean	19.4	33	669.31	NaN	NaN	11.5074	392.05	957070.91	NaN	NaN	408.92	2388.1791	23.4925	4114.95	580.507	75523.97
South Dakota	sd	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA
South Dakota	max	19.4	33	669.31	-	-	11.5074	392.05	957070.91	-	-	408.92	2388.1791	23.4925	4114.95	580.507	75523.97
South Dakota	min	19.4	33	669.31	Inf	Inf	11.5074	392.05	957070.91	Inf	Inf	408.92	2388.1791	23.4925	4114.95	580.507	75523.97
South Dakota	n	1.0	1	1.00	0.00	0.00	0.00	0.00	1.00	0.00	0.00	0.00	1.00	1.00	0.00	1.00	1.00
Tennessee	mean	19.5	29	767.91	4.94	2792.81	6239583641.60	445.72	NaN	66.09	2383.26	80.4375	137.0418	1753.72	787.29	4573.75	
Tennessee	sd	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA
Tennessee	max	19.5	29	767.91	4.94	2792.81	6239583641.60	445.72	-	66.09	2383.26	80.4375	137.0418	1753.72	787.29	4573.75	
Tennessee	min	19.5	29	767.91	4.94	2792.81	6239583641.60	445.72	Inf	66.09	2383.26	80.4375	137.0418	1753.72	787.29	4573.75	
Tennessee	n	1.0	1	1.00	1.00	0.00	0.00	0.00	1.00	0.00	0.00	0.00	1.00	1.00	0.00	1.00	1.00
Texas	mean	19.4	38	1004.75	NaN	NaN	13.9803	9708.25	6793.81	NaN	55.36	1846.23	32.9569	338.9179	95396.27	4063.93	781.56
Texas	sd	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA
Texas	max	19.4	38	1004.75	-	-	13.9803	9708.25	6793.81	-	55.36	1846.23	32.9569	338.9179	95396.27	4063.93	781.56
Texas	min	19.4	38	1004.75	Inf	Inf	13.9803	9708.25	6793.81	Inf	55.36	1846.23	32.9569	338.9179	95396.27	4063.93	781.56
Texas	n	1.0	1	1.00	0.00	0.00	0.00	0.00	1.00	0.00	0.00	0.00	1.00	1.00	0.00	1.00	1.00
Utah	mean	1.3	16	809.38	NaN	NaN	12.1000	633.36	3396.87	NaN	47.05	5005.13	334.6667	2265.26	60803.25	37.0667	250.60
Utah	sd	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA
Utah	max	1.3	16	809.38	-	-	12.1000	633.36	3396.87	-	47.05	5005.13	334.6667	2265.26	60803.25	37.0667	250.60
Utah	min	1.3	16	809.38	Inf	Inf	12.1000	633.36	3396.87	Inf	47.05	5005.13	334.6667	2265.26	60803.25	37.0667	250.60
Utah	n	1.0	1	1.00	0.00	0.00	0.00	0.00	1.00	0.00	0.00	0.00	1.00	1.00	0.00	1.00	1.00
Vermont	mean	13.6	30	716.20	9.43	3387.73	36666748.13	6085.87	NaN	74.41	335.29	25.4667	32.93	3339.86	95.86	786.93	305.20
Vermont	sd	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA	NA

state	stat	DriverDeaths	AlcoholRelated	PhysicalDistress	Uninsured	HouseholdIncome	Asian	AmericanIndian	AlaskaNative	Black	White
Vermont	max	3.6	30	716.279.43983.73BB6666748.1360085.87-	Inf	74.418335.2905.4667	32.93333695.86786.973305.20				
Vermont	min	13.6	30	716.279.43983.73BB6666748.1360085.87-	Inf	74.418335.2905.4667	32.93333695.86786.973305.20				
Vermont	tn	1.0	1	1.0000000001.000001.00001.00	0.0000000000000	1.0000	1.000001.000	1.00001.00			
Virginia	mean	12.7	27	768.977.408223.01298507432.492534.67-	Inf	52.158881.6495.5522	150.62687454.26361.78228.42				
Virginia	sd	NA	NA	NA NA NA NA NA NA NA NA NA	NA NA NA NA NA NA NA NA NA	NA NA NA NA NA NA NA NA NA	NA NA NA NA NA NA NA NA NA				
Virginia	max	12.7	27	768.977.408223.01298507432.492534.67-	Inf	52.158881.6495.5522	150.62687454.26361.78228.42				
Virginia	min	12.7	27	768.977.408223.01298507432.492534.67-	Inf	52.158881.6495.5522	150.62687454.26361.78228.42				
Virginia	n	1.0	1	1.0000000001.000001.00001.00	0.0000000000000	1.0000	1.000001.000	1.00001.00			
Washington	mean	10.6	33	890.079.993804.55020002405.603075.4-	Inf	44.0735899.7330.9500	3017.4540586.15211.2570029.45				
Washington	sd	NA	NA	NA NA NA NA NA NA NA NA NA	NA NA NA NA NA NA NA NA NA	NA NA NA NA NA NA NA NA NA	NA NA NA NA NA NA NA NA NA				
Washington	max	10.6	33	890.079.993804.55020002405.603075.4-	Inf	44.0735899.7330.9500	3017.4540586.15211.2570029.45				
Washington	min	10.6	33	890.079.993804.55020002405.603075.4-	Inf	44.0735899.7330.9500	3017.4540586.15211.2570029.45				
Washington	n	1.0	1	1.0000000001.000001.00001.00	0.0000000000000	1.0000	1.000001.000	1.00001.00			
West Virginia	mean	23.8	28	992.675.175238.8929178648.824672.48-	Inf	68.88725.10713.7500	18.92857112.92241.954875.43				
West Virginia	sd	NA	NA	NA NA NA NA NA NA NA NA NA	NA NA NA NA NA NA NA NA NA	NA NA NA NA NA NA NA NA NA	NA NA NA NA NA NA NA NA NA				
West Virginia	max	23.8	28	992.675.175238.8929178648.824672.48-	Inf	68.88725.10713.7500	18.92857112.92241.954875.43				
West Virginia	min	23.8	28	992.675.175238.8929178648.824672.48-	Inf	68.88725.10713.7500	18.92857112.92241.954875.43				
West Virginia	n	1.0	1	1.0000000001.000001.00001.00	0.0000000000000	1.0000	1.000001.000	1.00001.00			
Wisconsin	mean	13.8	33	670.379.312733.8320548389.260084.95-	Inf	59.694802.16840.2192	93.671231320.76799.220315.48				
Wisconsin	sd	NA	NA	NA NA NA NA NA NA NA NA NA	NA NA NA NA NA NA NA NA NA	NA NA NA NA NA NA NA NA NA	NA NA NA NA NA NA NA NA NA				
Wisconsin	max	13.8	33	670.379.312733.8320548389.260084.95-	Inf	59.694802.16840.2192	93.671231320.76799.220315.48				
Wisconsin	min	13.8	33	670.379.312733.8320548389.260084.95-	Inf	59.694802.16840.2192	93.671231320.76799.220315.48				
Wisconsin	n	1.0	1	1.0000000001.000001.00001.00	0.0000000000000	1.0000	1.000001.000	1.00001.00			
Wyoming	mean	17.4	32	791.179.37580.66575833789.5864330.75-	Inf	49.88347.583314.8333	49.66667884.0543.3343365.00				
Wyoming	sd	NA	NA	NA NA NA NA NA NA NA NA NA	NA NA NA NA NA NA NA NA NA	NA NA NA NA NA NA NA NA NA	NA NA NA NA NA NA NA NA NA				
Wyoming	max	17.4	32	791.179.37580.66575833789.5864330.75-	Inf	49.88347.583314.8333	49.66667884.0543.3343365.00				
Wyoming	min	17.4	32	791.179.37580.66575833789.5864330.75-	Inf	49.88347.583314.8333	49.66667884.0543.3343365.00				
Wyoming	n	1.0	1	1.0000000001.000001.00001.00	0.0000000000000	1.0000	1.000001.000	1.00001.00			

```
clean_joineddata2 %>% summary(is.numeric)
```

```
##      state              LE              Deaths              DriverDeaths
## Length:51             Min.   :74.63         Min.    : 587.7         Min.    : 5.90
## Class :character      1st Qu.:76.78         1st Qu.: 1501.4        1st Qu.:12.75
## Mode  :character      Median :78.32         Median : 2691.5        Median :15.60
##                               Mean  :77.91         Mean   : 3793.2        Mean   :15.79
##                               3rd Qu.:79.36        3rd Qu.: 5671.8        3rd Qu.:18.50
## AlcoholRelated PhysicalDistress Uninsured HouseholdIncome
## Min.   :16.00         Min.    : 9.00         Min.    : 248.1         Min.    :41754
## 1st Qu.:28.00         1st Qu.:11.59        1st Qu.: 1013.8        1st Qu.:51647
## Median :30.00         Median :12.99        Median : 1678.6        Median :58552
## Mean   :30.69         Mean   :13.11        Mean   : 2677.5        Mean   :59654
```

```
## 3rd Qu.:33.00 3rd Qu.:14.57 3rd Qu.: 2804.0 3rd Qu.:64647
## Homicide PercRural Premiums Asian
## Min. : 2.333 Min. : 0.00 Min. : 642.0 Min. : 345
## 1st Qu.: 3.792 1st Qu.:43.78 1st Qu.: 768.4 1st Qu.: 1901
## Median : 5.250 Median :54.16 Median : 859.0 Median : 5800
## Mean : 8.528 Mean :51.04 Mean : 887.0 Mean : 19619
## 3rd Qu.:11.625 3rd Qu.:64.60 3rd Qu.:1007.9 3rd Qu.: 16595
## AmericanIndianAlaskaNative PacificIslander Hispanic
## Min. : 163.8 Min. : 18.93 Min. : 1113
## 1st Qu.: 756.1 1st Qu.: 94.72 1st Qu.: 5556
## Median : 1760.9 Median : 175.29 Median : 12454
## Mean : 3917.9 Mean : 1540.42 Mean : 48860
## 3rd Qu.: 3973.8 3rd Qu.: 716.53 3rd Qu.: 50897
## Black White Rural % BIPOC
## Min. : 202.3 Min. : 21524 Length:51 Min. : 5.512
## 1st Qu.: 2847.1 1st Qu.: 63476 Class :character 1st Qu.:19.323
## Median : 15211.2 Median :113526 Mode :character Median :27.949
## Mean : 34948.7 Mean :166065 Mean :31.110
## 3rd Qu.: 41505.8 3rd Qu.:231980 3rd Qu.:41.914
## Diverse
## Length:51
## Class :character
## Mode :character
##
##
```

```
## [ reached getOption("max.print") -- omitted 2 rows ]
```

```
clean_joineddata2 %>% group_by(state) %>% summarize(count = n())
```

```
## # A tibble: 51 x 2
## state count
## <chr> <int>
## 1 Alabama 1
## 2 Alaska 1
## 3 Arizona 1
## 4 Arkansas 1
## 5 California 1
## 6 Colorado 1
## 7 Connecticut 1
## 8 Delaware 1
## 9 District of Columbia 1
## 10 Florida 1
## # ... with 41 more rows
```

```
clean_joineddata2 %>% group_by(Rural) %>% summarize(count = n())
```

```
## # A tibble: 3 x 2
## Rural count
## <chr> <int>
## 1 no 20
## 2 yes 29
## 3 <NA> 2
```

```
clean_joineddata2 %>% group_by(Diverse) %>% summarize(count = n())
```

```
## # A tibble: 2 x 2
```

```
##   Diverse count
##   <chr>   <int>
## 1 high       6
## 2 low       45
```

This is a table of my numeric variables, which shows the mean, standard deviation, maximum, minimum, and distinct counts for each state in the United States. This is one big table that prints out these values for the states. I then found the counts for each of my categorical variables. As expected, there is only one count for each of the states. Surprisingly, the amount of states that were considered to be rural were almost even (with 20 being no and 31 being yes). Furthermore, very surprising to see that there were only 6 states that were considered to have high diversity (having a percentage of BIPOC greater than 50).

```
# Some more summary statistics for my variables
clean_joineddata2 %>% group_by(state, Diverse) %>% summarize(HouseholdIncome = mean(HouseholdIncome)) %>%
  arrange(desc((HouseholdIncome)))
```

```
## # A tibble: 51 x 3
## # Groups:   state [51]
##   state          Diverse HouseholdIncome
##   <chr>          <chr>          <dbl>
## 1 District of Columbia high          90395
## 2 New Jersey      low           85193.
## 3 Massachusetts   low           80279.
## 4 Maryland         low           79316.
## 5 Connecticut      low           79151.
## 6 Rhode Island     low           77003.
## 7 California       high          71197.
## 8 New Hampshire    low           70350
## 9 Alaska           low           67924.
## 10 Delaware        low           67517.
## # ... with 41 more rows
```

I wanted to compute some extra summary statistics for some of my variables. I grouped by state and Diverse, summarized to find the mean household incomes within each state, and then arranged by descending income. It was very interesting to see that the mean income in D.C. was \$90395.00 and the diversity was considered to be high. The lowest average household income was in Mississippi with \$41753.00 and the diversity level was considered low.

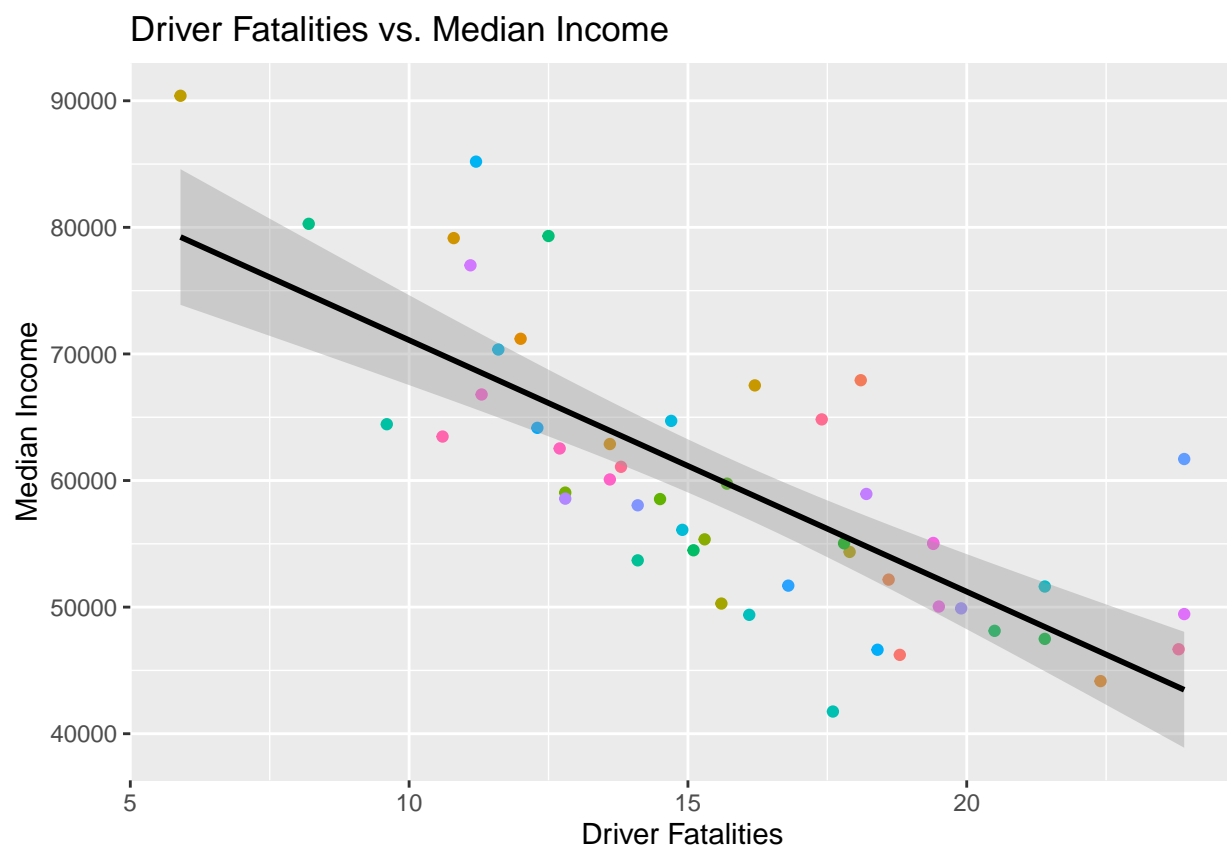
```
clean_joineddata2 %>% filter(Rural == "yes") %>% filter(str_detect(state,
  "[aeiouAEIOU]"))
```

```
## # A tibble: 7 x 20
## # Groups:   state [7]
##   state    LE Deaths DriverDeaths AlcoholRelated PhysicalDistress Uninsured
##   <chr> <dbl> <dbl>          <dbl>          <dbl>          <dbl> <dbl>
## 1 Alab~  74.7  2419.          18.8            30            15.7  1062.
## 2 Arka~  75.0  1270.          22.4            26            16.6   828.
## 3 Idaho  NA      NA            15.3            29            13     1248.
## 4 Illi~  78.0  2801.          12.8            34            12.5  1927.
## 5 Indi~  77.0  1983.          14.5            29            13.2  2287.
## 6 Iowa   79.1   716.          15.7            25            10.2   408.
## 7 Okla~  75.3  1560.          19.9            29            15.4  2110.
## # ... with 13 more variables: HouseholdIncome <dbl>, Homicide <dbl>,
## #   PercRural <dbl>, Premiums <dbl>, Asian <dbl>,
## #   AmericanIndianAlaskaNative <dbl>, PacificIslander <dbl>, Hispanic <dbl>,
## #   Black <dbl>, White <dbl>, Rural <chr>, `% BIPOC` <dbl>, Diverse <chr>
```

Lastly, I wanted to filter and find the states that were considered to be rural. This gave me a value of 31, which meant out of the 50 states, we could say that around 31 were considered to be rural with an average percentage of their counties being greater than 50%. I then used the `str_detect` function to find the states out of the 31 that started with a vowel. This led me to Alabama, Alaska, Arkansas, Colorado, Georgia, Idaho, Illinois, Indiana, Iowa, and Kansas.

```
# Plot of driver fatalities vs. median household income
clean_joineddata2 %>% ggplot(aes(x = DriverDeaths, y = HouseholdIncome)) +
  geom_point(aes(color = state)) + geom_smooth(method = "lm",
  color = 1) + scale_x_continuous(breaks = seq(0, 25, 5)) +
  scale_y_continuous(breaks = seq(0, 1e+05, 10000)) + ggtitle("Driver Fatalities vs. Median Income") +
  xlab("Driver Fatalities") + ylab("Median Income") + theme(legend.position = "none")
```

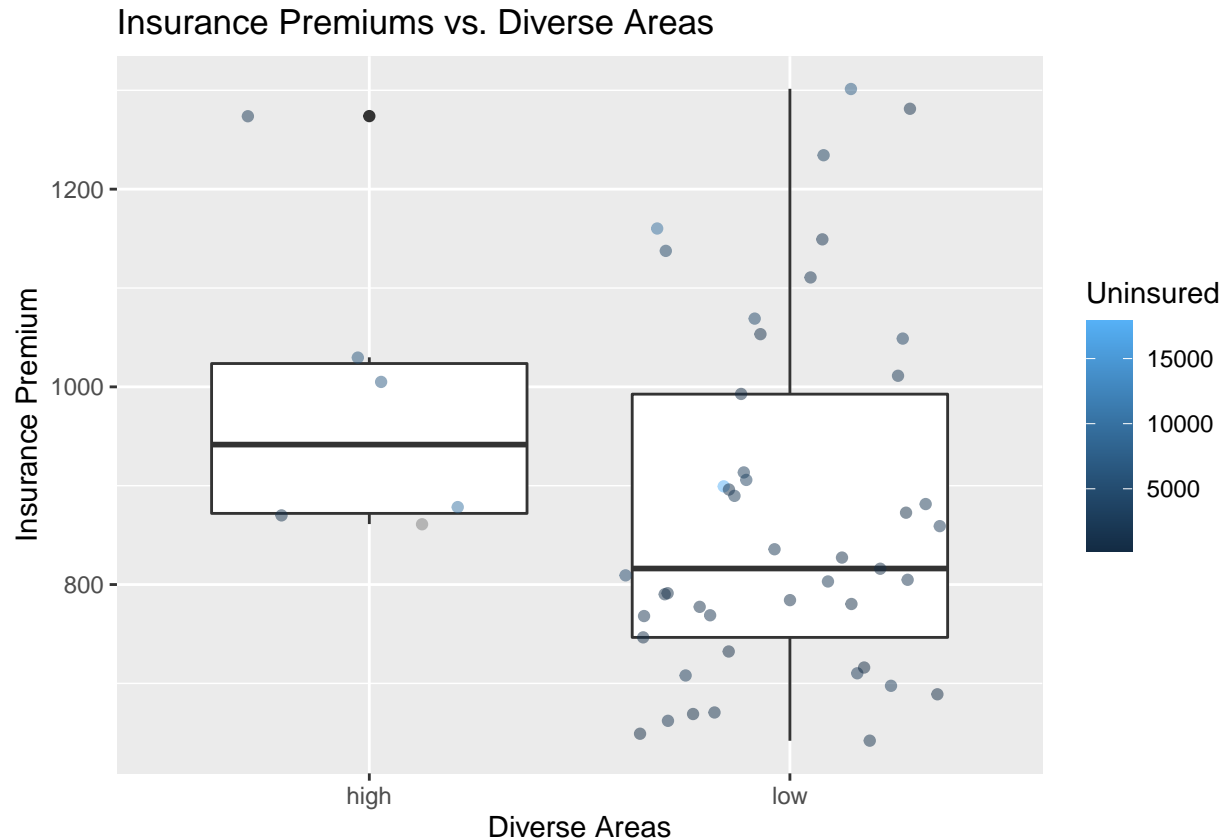
Visualizing



For the first ggplot, we're looking at the relationship between the number of drivers that have died from a car accident and median household income in each of the 50 U.S. states. Furthermore, I added points that are colored by state to help visualize the differences in each state for driver fatalities and median income. When looking at this plot, we can tell that there seems to be a relatively strong relationship between median household income and driver fatalities. As the median household income in each state decreased, there was a greater number of traffic fatalities that occurred. This could be from various different factors, but I thought it was interesting to see this possible relationship.

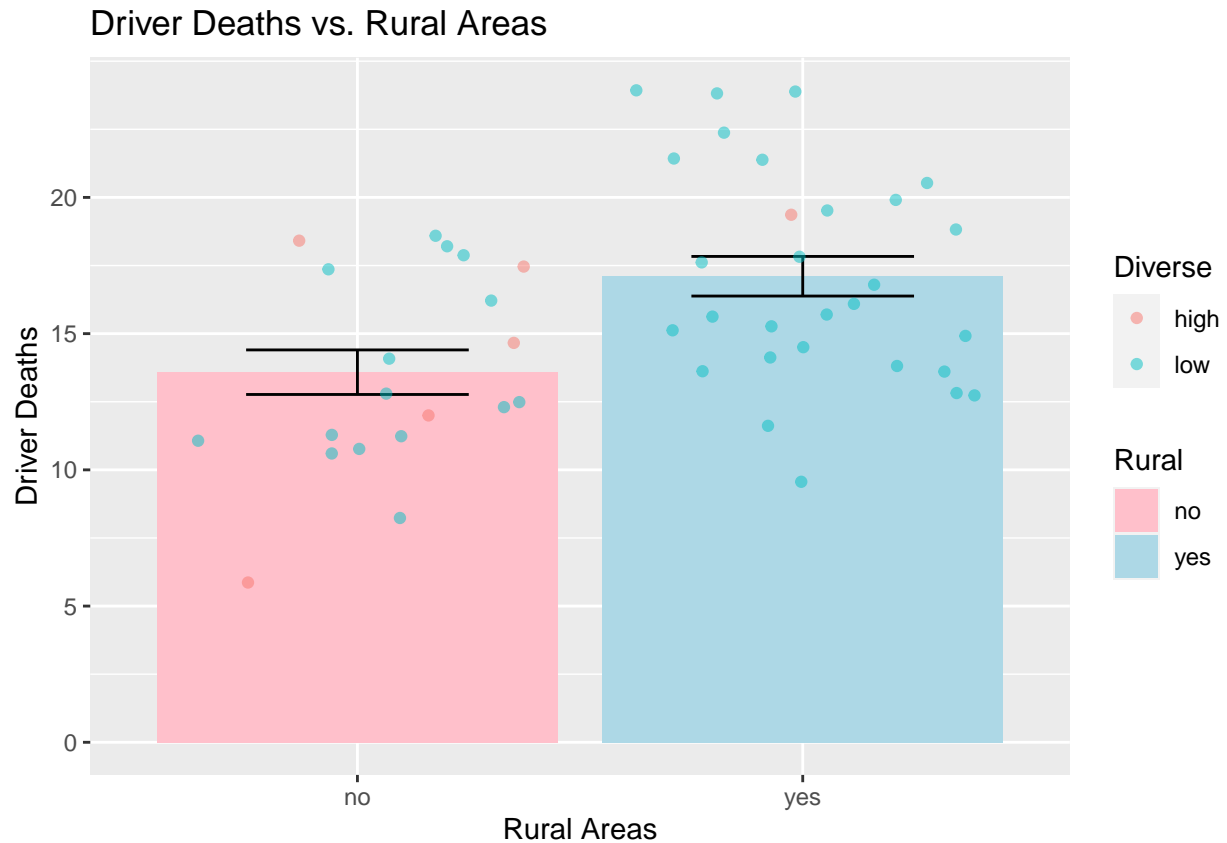
```
# Plot of driver fatalities in each state in the United
# States
clean_joineddata2 %>% ggplot(aes(x = Diverse, y = Premiums)) +
```

```
geom_boxplot() + scale_y_continuous(breaks = seq(0, 1400,
200)) + ggtitle("Insurance Premiums vs. Diverse Areas") +
xlab("Diverse Areas") + ylab("Insurance Premium") + geom_jitter(alpha = 0.5,
aes(color = Uninsured))
```



For the next plot, I wanted to do a boxplot looking at the relationship between the diversity of the areas and the costs of insurance premiums across the state. I also wanted to overlay the points with number of people uninsured to look at how these are all related to each other. There seems to be a positive skew for this plot. In states where there was low diversity, there seemed to be a range of various values of people uninsured and lower median insurance premium than those with high diversity. This could be due to a variety of reasons. In this case, we are only comparing the diversity areas based off of the percentages calculated in the wrangling section; therefore, this could be different if we looked at it from a county point of view. Just as in other boxplots, this plot gives us insight into the visualization of the minimum, first quartile, median, third quartile, and maximum, essentially the summary of the distribution.

```
# your plot 3
clean_joineddata2 %>% filter(!is.na(Rural)) %>% ggplot(aes(x = Rural,
y = DriverDeaths)) + geom_bar(aes(fill = Rural), stat = "summary",
fun = mean) + ggtitle("Driver Deaths vs. Rural Areas") +
geom_errorbar(stat = "summary", width = 0.5) + xlab("Rural Areas") +
ylab("Driver Deaths") + scale_fill_manual(values = c("pink",
"light blue")) + scale_y_continuous(breaks = seq(0, 20, 5)) +
geom_jitter(alpha = 0.5, aes(color = Diverse))
```



For my last plot, I wanted to take a look at the number of driver deaths in rural areas with an overlay of Diversity on the points. This was very interesting because we see that in cases of rural states, there is very lower diversity. Both rural and not rural have relatively similar error bars. In this case, rural areas seemed to have a higher amount of driver fatalities than states that weren't considered to be rural. In cases that aren't rural, there is still more states that are not as diverse. This could also be because states in the United States in general aren't super diverse.

Concluding Remarks I really enjoyed looking at these datasets. Keeping up with the news and learning about these incidents in various ways is informing. Before looking at these datasets, I really only focused on data that I saw locally and maybe a couple of nationwide data. I hope that I can continue to be conscious of the ways that I think about the news and how I view them across the states as well. Furthermore, I moved my tidying section to be after the joining because my knitting process wouldn't work otherwise.