

# Arizona Fatal Shootings

2023-08-27

```
## load libraries
library(ggplot2)
library(tidyverse)
```

```
## -- Attaching core tidyverse packages ----- tidyverse 2.0.0 --
## v dplyr      1.1.3      v readr      2.1.4
## v forcats    1.0.0      v stringr   1.5.0
## v lubridate  1.9.2      v tibble    3.2.1
## v purrr      1.0.2      v tidyr     1.3.0
## -- Conflicts ----- tidyverse_conflicts() --
## x dplyr::filter() masks stats::filter()
## x dplyr::lag()     masks stats::lag()
## i Use the conflicted package (<http://conflicted.r-lib.org/>) to force all conflicts to become errors
```

```
library(dplyr)
library(stringr)
## load data
shooting.df <- read.csv("~/Documents/Stats Seminar/fatal-police-shootings-data.csv", header=TRUE)
summary(shooting.df)
```

```
##           id           date      threat_type      flee_status
## Min.      : 3   Length:8721      Length:8721      Length:8721
## 1st Qu.:2447   Class :character   Class :character   Class :character
## Median :4833   Mode  :character   Mode  :character   Mode  :character
## Mean      :4804
## 3rd Qu.:7156
## Max.      :9472
##
## armed_with      city      county      state
## Length:8721     Length:8721     Length:8721     Length:8721
## Class :character   Class :character   Class :character   Class :character
## Mode  :character   Mode  :character   Mode  :character   Mode  :character
##
##
##
## latitude      longitude      location_precision      name
## Min.      :19.50   Min.      :-160.01   Length:8721     Length:8721
## 1st Qu.:33.47     1st Qu.: -112.02   Class :character   Class :character
## Median :36.09     Median :  -94.32   Mode  :character   Mode  :character
## Mean      :36.64     Mean      : -97.04
## 3rd Qu.:40.00     3rd Qu.:  -83.19
## Max.      :71.30     Max.      :  -67.87
```

```
## NA's :980 NA's :980
## age gender race race_source
## Min. : 2.00 Length:8721 Length:8721 Length:8721
## 1st Qu.:27.00 Class :character Class :character Class :character
## Median :35.00 Mode :character Mode :character Mode :character
## Mean :37.31
## 3rd Qu.:45.00
## Max. :92.00
## NA's :600
## was_mental_illness_related body_camera agency_ids
## Length:8721 Length:8721 Length:8721
## Class :character Class :character Class :character
## Mode :character Mode :character Mode :character
##
##
##
##
```

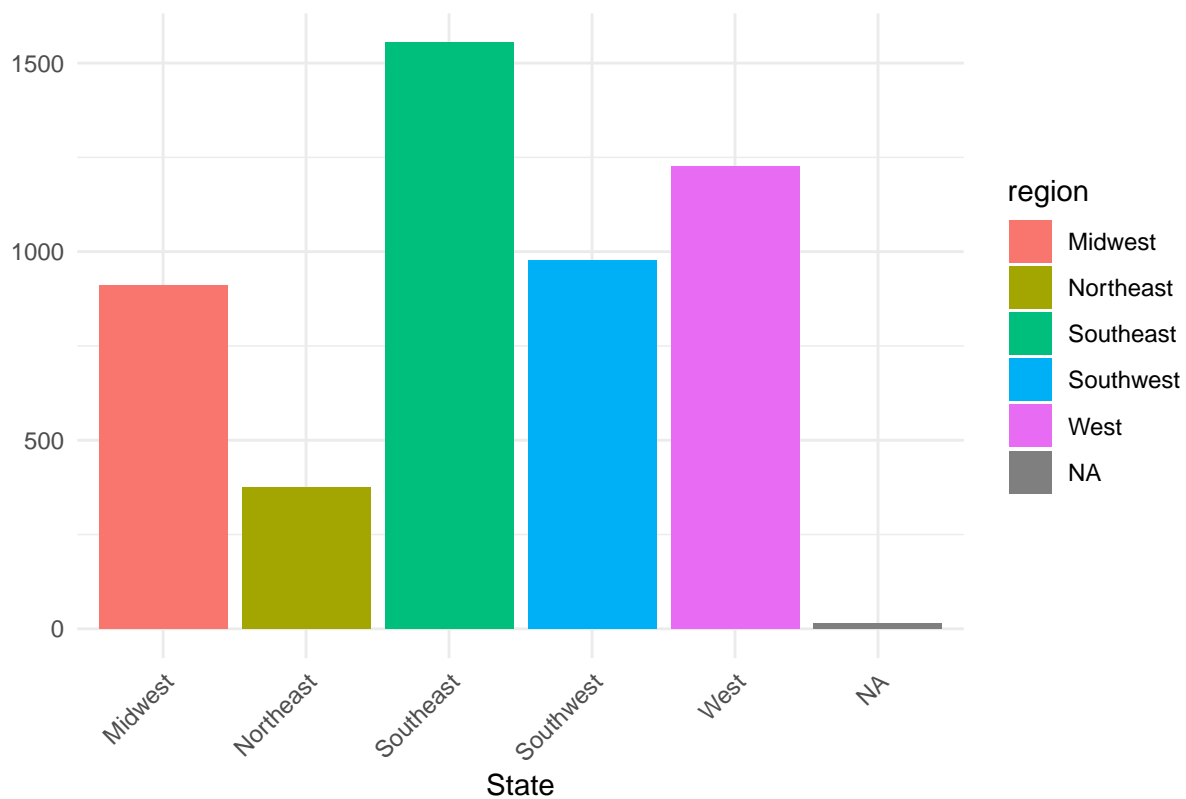
```
head(shooting.df)
```

```
## id date threat_type flee_status armed_with city county
## 1 3 2015-01-02 point not gun Shelton Mason
## 2 4 2015-01-02 point not gun Aloha Washington
## 3 5 2015-01-03 move not unarmed Wichita Sedgwick
## 4 8 2015-01-04 point not replica San Francisco San Francisco
## 5 9 2015-01-04 point not other Evans Weld
## 6 11 2015-01-04 attack not gun Guthrie Logan
## state latitude longitude location_precision name age gender
## 1 WA 47.24683 -123.12159 not_available Tim Elliot 53 male
## 2 OR 45.48742 -122.89170 not_available Lewis Lee Lembke 47 male
## 3 KS 37.69477 -97.28055 not_available John Paul Quintero 23 male
## 4 CA 37.76291 -122.42200 not_available Matthew Hoffman 32 male
## 5 CO 40.38394 -104.69226 not_available Michael Rodriguez 39 male
## 6 OK 35.87699 -97.42345 not_available Kenneth Joe Brown 18 male
## race race_source was_mental_illness_related body_camera agency_ids
## 1 A not_available True False 73
## 2 W not_available False False 70
## 3 H not_available False False 238
## 4 W not_available True False 196
## 5 H not_available False False 473
## 6 W not_available False False 101
```

```
gun_incidents <- shooting.df %>%
  filter(armed_with == "gun")
ggplot(gun_incidents) +
  aes(x = state, fill = state) +
  geom_bar() +
  theme(axis.text.x = element_text(angle = 90))
```

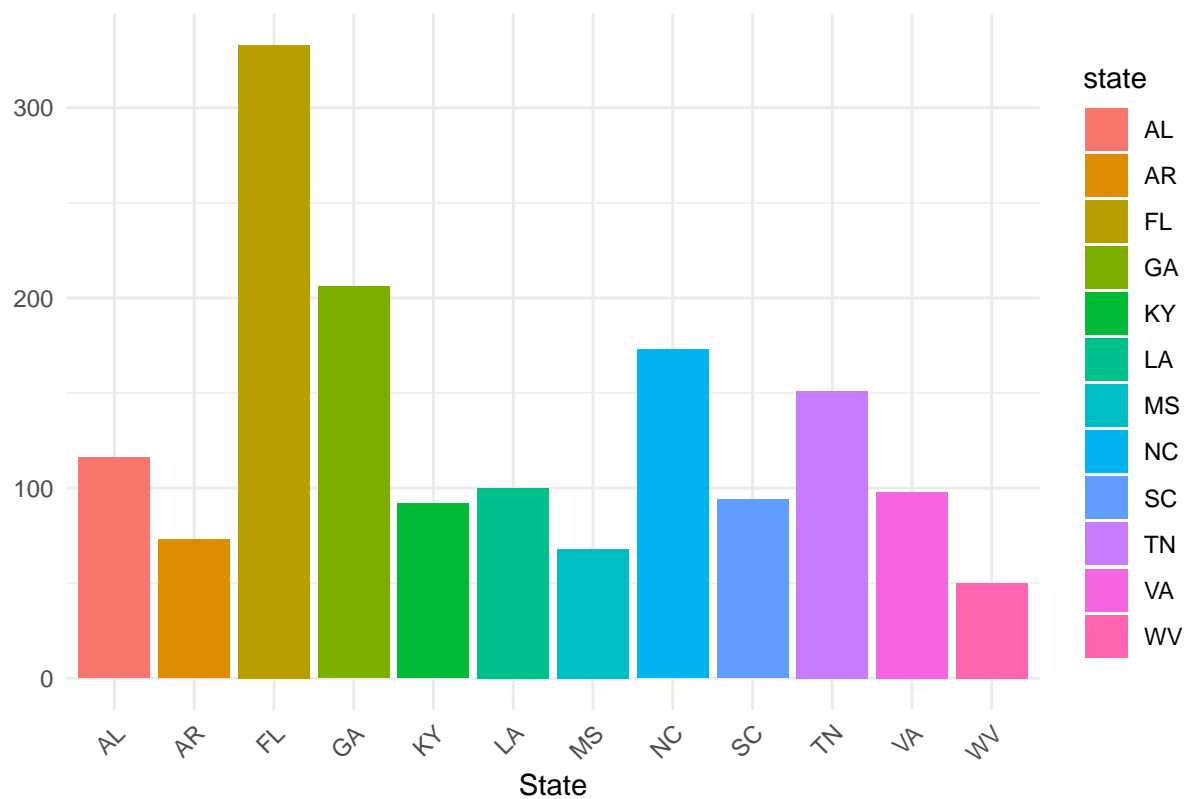


# Distribution of Police Shootings Involving Armed Victims by Region



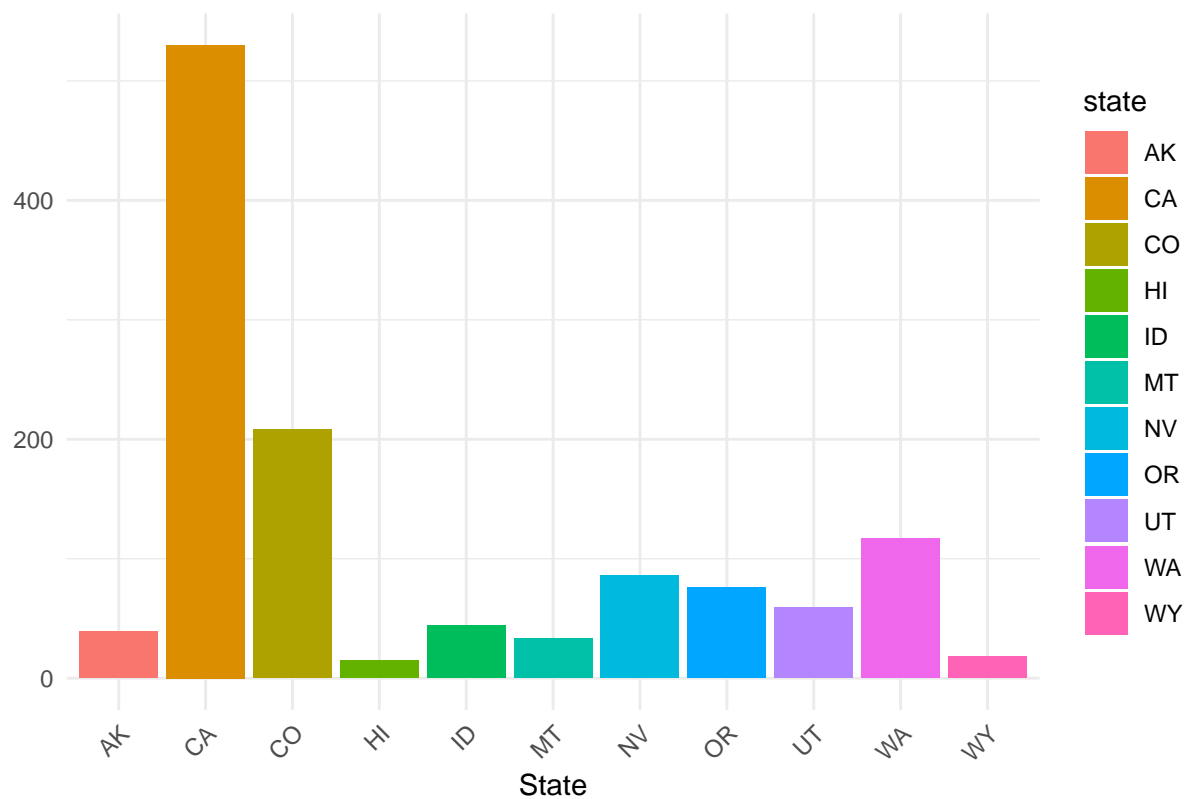
```
gun_incidents_southeast <- gun_incidents_with_regions %>%
  filter(region == "Southeast")
ggplot(gun_incidents_southeast) +
  aes(x = state, fill = state) +
  geom_bar() +
  labs(title = "Distribution of Police Shootings Involving Armed Victims in the Southeast by State", x = "State") +
  theme_minimal() +
  theme(axis.text.x = element_text(angle = 45, hjust = 1))
```

Distribution of Police Shootings Involving Armed Victims in the Southeast b



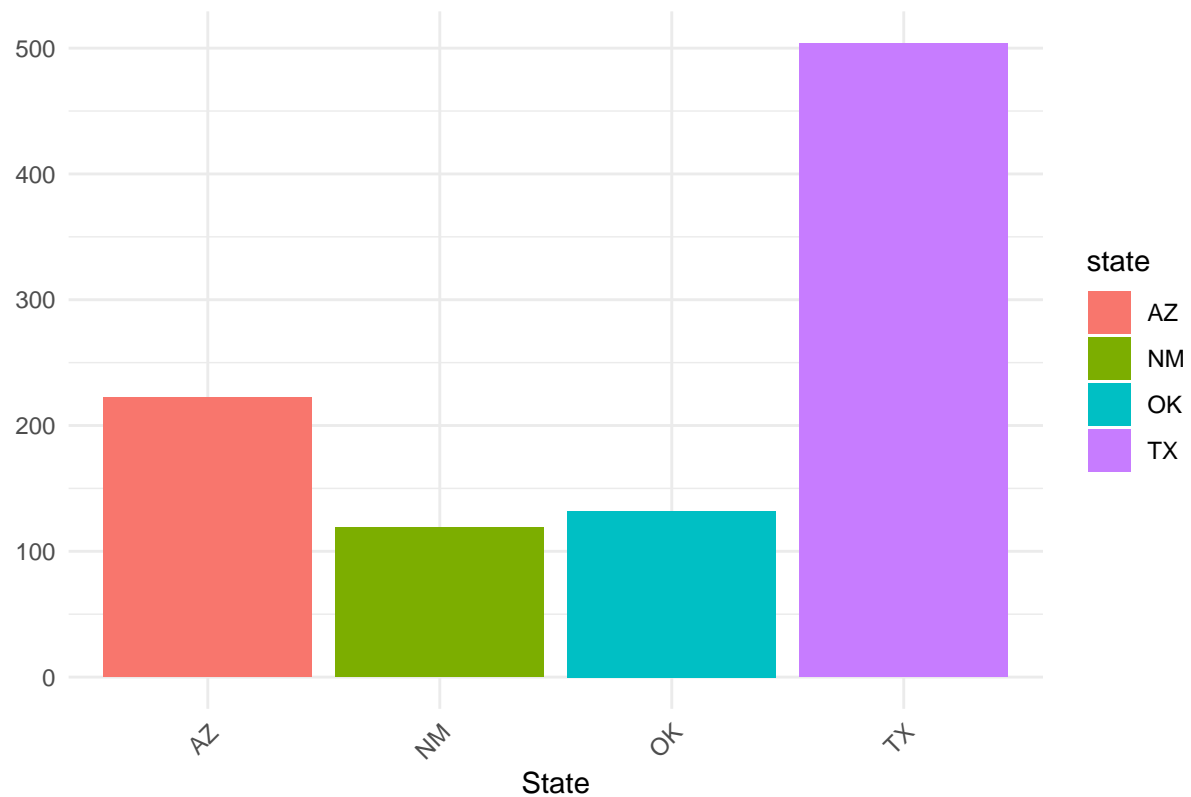
```
gun_incidents_west <- gun_incidents_with_regions %>%
  filter(region == "West")
ggplot(gun_incidents_west) +
  aes(x = state, fill = state) +
  geom_bar() +
  labs(title = "Distribution of Police Shootings Involving Armed Victims in the West by State", x = "State") +
  theme_minimal() +
  theme(axis.text.x = element_text(angle = 45, hjust = 1))
```

Distribution of Police Shootings Involving Armed Victims in the West by Sta



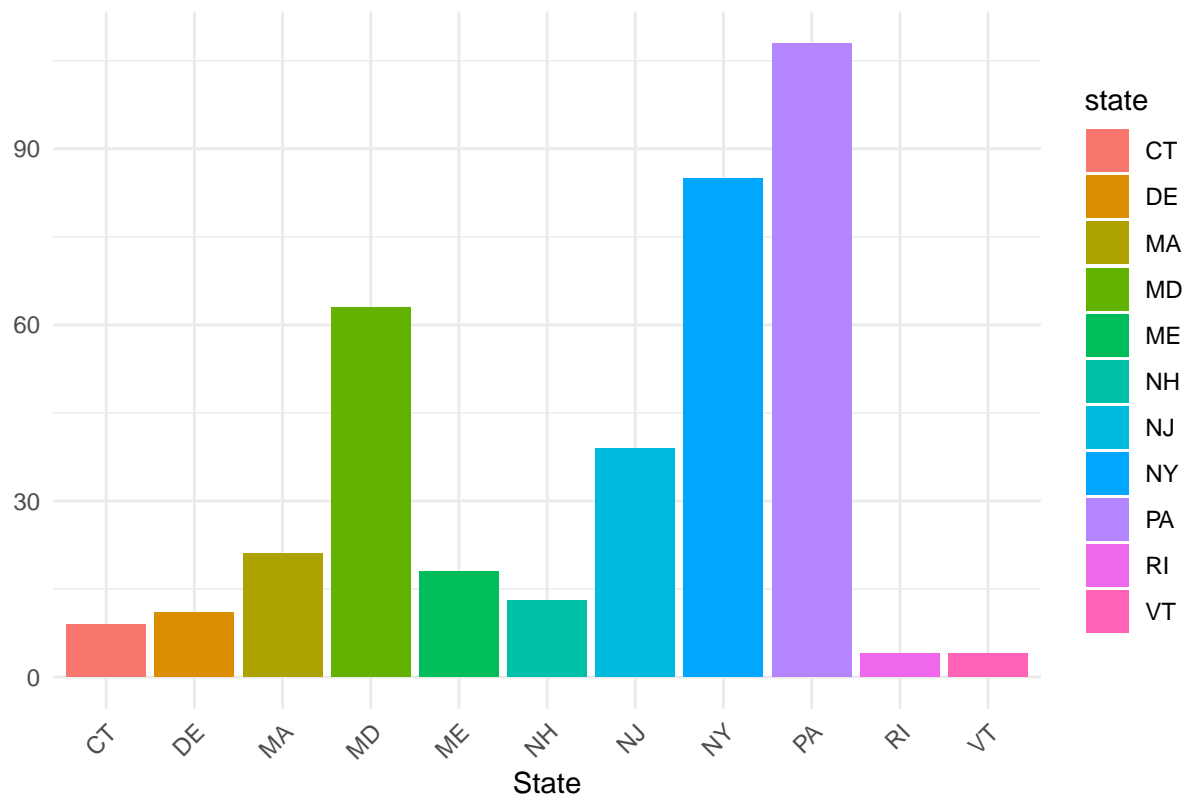
```
gun_incidents_southwest <- gun_incidents_with_regions %>%
  filter(region == "Southwest")
ggplot(gun_incidents_southwest) +
  aes(x = state, fill = state) +
  geom_bar() +
  labs(title = "Distribution of Police Shootings Involving Armed Victims in the Southwest by State", x = state) +
  theme_minimal() +
  theme(axis.text.x = element_text(angle = 45, hjust = 1))
```

Distribution of Police Shootings Involving Armed Victims in the Southwest b



```
gun_incidents_northeast <- gun_incidents_with_regions %>%
  filter(region == "Northeast")
ggplot(gun_incidents_northeast) +
  aes(x = state, fill = state) +
  geom_bar() +
  labs(title = "Distribution of Police Shootings Involving Armed Victims in the Northeast by State", x = "State") +
  theme_minimal() +
  theme(axis.text.x = element_text(angle = 45, hjust = 1))
```

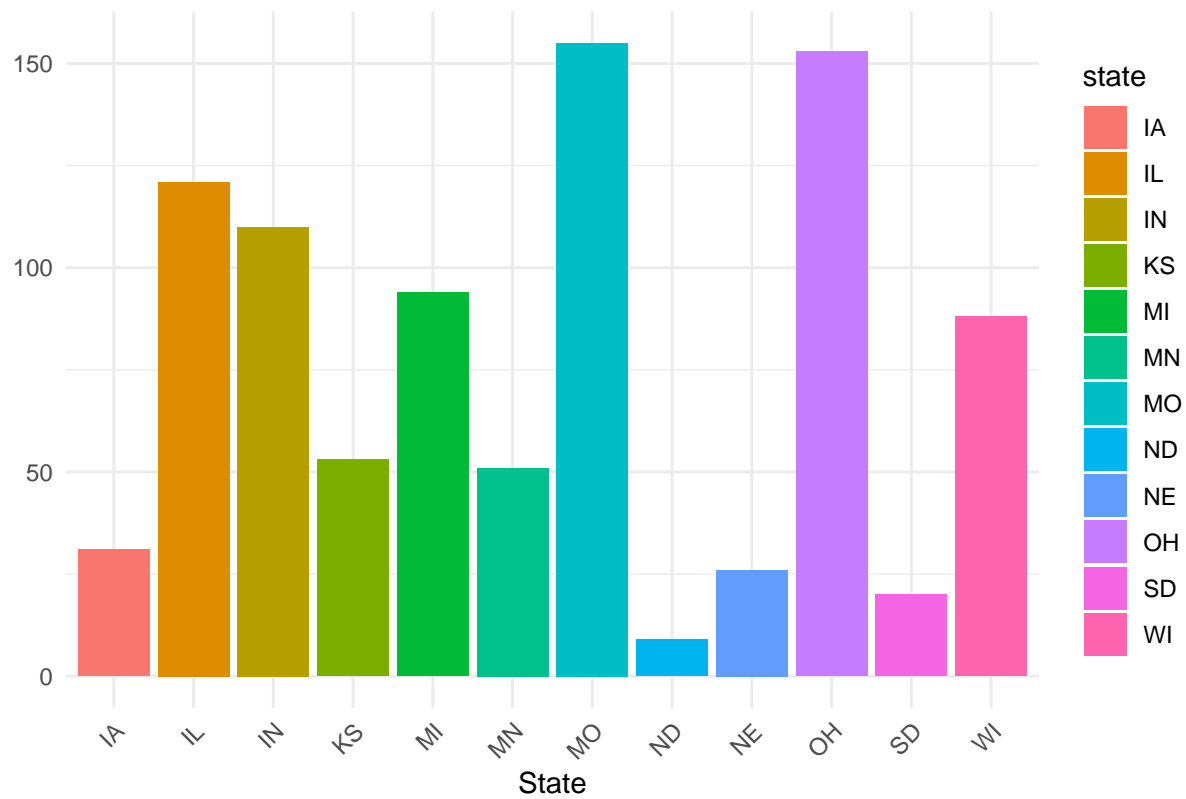
Distribution of Police Shootings Involving Armed Victims in the Northeast by



```
gun_incidents_midwest <- gun_incidents_with_regions %>%
  filter(region == "Midwest")
ggplot(gun_incidents_midwest) +
  aes(x = state, fill = state) +
  geom_bar() +
  labs(title = "Distribution of Police Shootings Involving Armed Victims in the Midwest by State", x =
  theme_minimal() +
  theme(axis.text.x = element_text(angle = 45, hjust = 1))
```



Distribution of Police Shootings Involving Armed Victims in the Midwest by State



```
ggplot(gun_incidents_southwest) +  
  aes(x = date, y = state)
```

TX -

OK -

state

NM -

AZ -

date

```
geom_point()
```

```
## geom_point: na.rm = FALSE
## stat_identity: na.rm = FALSE
## position_identity
```

```
#Clean the data
```

```
shootings_cleaned <- subset(shooting.df, age!="NA" & gender == "male" & race %in% c("A", "H", "B", "W"))
```

```
arizona_shootings_cleaned <- subset(shooting.df, age!="NA" & gender == "male" & race %in% c("A", "H", "I"))
```

```
arizona_pop <- read.csv("~/Documents/Stats Seminar/Arizona Population Data.csv", header=TRUE)
```

```
arizona_pop
```

```
##   Year Population Growth.Rate
## 1 2022   7,359,197      1.30%
## 2 2021   7,264,877      1.18%
## 3 2020   7,179,943     -1.53%
## 4 2019   7,291,843      1.78%
## 5 2018   7,164,228      1.65%
## 6 2017   7,048,088      1.49%
## 7 2016   6,944,767      1.64%
## 8 2015   6,832,810      1.48%
```

```

library(lubridate)

# Create a new empty column named 'year'
arizona_shootings_cleaned$year <- NA

# Populate the 'year' column with extracted year values
for (i in 1:nrow(arizona_shootings_cleaned)) {
  specific_date <- arizona_shootings_cleaned$date[i]

  if (!is.na(specific_date)) {
    year <- year(specific_date) # Extract year using lubridate
    arizona_shootings_cleaned$year[i] <- year
  }
}

# Loop through each row in the shootings dataframe
for (i in 1:nrow(arizona_shootings_cleaned)) {
  year <- arizona_shootings_cleaned$year[i] # Extract the year from the shootings dataframe

  # Match the year to the corresponding row in the population dataframe
  matched_row <- arizona_pop[arizona_pop$Year == year, ]

  # If a match is found, assign the population value to the 'population' column in the shootings dataframe
  if (nrow(matched_row) > 0) {
    arizona_shootings_cleaned$population[i] <- matched_row$Population
  }
}

# Print the updated shootings dataframe
arizona_shootings_cleaned <- subset(arizona_shootings_cleaned, year!= "2023")
print(arizona_shootings_cleaned)

```

##	id	date	threat_type	flee_status	armed_with
## 7	13	2015-01-05	shoot	car	gun
## 43	80	2015-01-18	point	not	gun
## 46	85	2015-01-20	threat	car	
## 67	349	2015-01-27	shoot	not	gun
## 86	124	2015-02-04	point	car	gun
## 87	125	2015-02-04	accident	not	unarmed
## 110	154	2015-02-13	threat	not	knife
## 116	156	2015-02-16	point	not	gun
## 159	198	2015-03-03	attack	not	knife
## 168	271	2015-03-07	point	not	gun
## 183	281	2015-03-11	point	not	replica
## 215	235	2015-03-21	shoot	not	gun
## 229	242	2015-03-25	move	not	gun
## 257	301	2015-04-05	threat	not	knife
## 291	376	2015-04-18	shoot	not	gun
## 301	394	2015-04-22	threat	not	knife
## 323	416	2015-04-29	point	not	gun
## 345	439	2015-05-08	attack	not	knife
## 379	474	2015-05-23	shoot	car	gun
## 398	500	2015-05-30	attack	foot	unarmed

##	421	526	2015-06-08	point	not	gun
##	491	609	2015-07-07	threat	not	blunt_object
##	501	623	2015-07-09	point	foot	gun
##	535	652	2015-07-18	shoot	not	gun
##	572	696	2015-08-01	shoot	not	gun
##	602	725	2015-08-10	threat	not	
##	646	775	2015-08-25	point	not	gun
##	663	791	2015-08-30	point	not	gun
##	668	799	2015-09-02	shoot	not	gun
##	736	868	2015-09-25	shoot	not	gun
##	747	888	2015-10-02	shoot	car	gun
##	895	1035	2015-11-25	shoot	not	gun
##	902	1038	2015-11-29	point	not	gun
##	917	1056	2015-12-04	threat	not	blunt_object
##	935	1075	2015-12-11	attack	not	unarmed
##	983	1121	2015-12-26	threat	not	blunt_object
##	988	1132	2015-12-27	threat	not	knife
##	1038	1188	2016-01-17	move	not	unarmed
##	1039	1178	2016-01-18	shoot	not	gun
##	1042	1186	2016-01-18	point	not	gun
##	1083	1236	2016-02-04	threat	not	knife
##	1148	1317	2016-02-23	shoot	not	gun
##	1151	1314	2016-02-24	move	car	
##	1168	1335	2016-03-03	move	car	knife
##	1249	1436	2016-03-30	undetermined	not	gun
##	1307	1486	2016-04-23	shoot	not	gun
##	1326	1507	2016-04-30	attack	car	
##	1365	1546	2016-05-18	shoot	not	gun
##	1400	1575	2016-05-30	point	not	gun
##	1454	1638	2016-06-20	threat	not	knife
##	1514	1700	2016-07-08	threat	not	knife
##	1523	1708	2016-07-13	undetermined	foot	gun
##	1549	1736	2016-07-27	attack	foot	unarmed
##	1550	1739	2016-07-27	threat	foot	knife
##	1557	1743	2016-07-28	shoot	other	gun
##	1594	1782	2016-08-12	threat	car	undetermined
##	1597	1781	2016-08-13	attack	not	knife
##	1604	1790	2016-08-16	threat	not	replica
##	1606	1792	2016-08-16	attack	not	unarmed
##	1618	1800	2016-08-19	attack	not	gun
##	1633	1820	2016-08-27	point	not	replica
##	1665	1864	2016-09-06	point	car	gun
##	1667	1868	2016-09-06	attack	not	knife
##	1683	1884	2016-09-15	threat	other	gun
##	1714	1941	2016-09-27	threat	car	
##	1717	1921	2016-09-29	shoot	not	gun
##	1745	1954	2016-10-09	threat	car	gun
##	1796	2003	2016-10-30	threat	car	
##	1826	2034	2016-11-09	shoot	not	gun
##	1834	2045	2016-11-12	shoot	car	gun
##	1865	2142	2016-11-25	threat	not	unarmed
##	1866	2077	2016-11-26	attack	car	
##	1976	2205	2017-01-09	move	foot	unarmed
##	2038	2707	2017-01-28	point	not	gun

##	2095	2335	2017-02-13	shoot	not	gun
##	2116	2362	2017-02-19	point	not	gun
##	2149	2397	2017-03-02	point	not	replica
##	2179	2428	2017-03-15	threat	car	
##	2202	2453	2017-03-23	attack	not	knife
##	2243	2499	2017-04-13	threat	not	
##	2261	2518	2017-04-20	threat	car	
##	2267	2526	2017-04-23	shoot	not	gun
##	2322	2579	2017-05-14	shoot	car	gun
##	2331	2587	2017-05-16	threat		knife
##	2343	2601	2017-05-23	shoot	other	gun
##	2352	2616	2017-05-26	point	not	replica
##	2370	2625	2017-06-03	point	not	gun
##	2427	2704	2017-06-23	attack	car	
##	2434	2696	2017-06-25	point	not	gun
##	2447	2727	2017-07-01	threat	not	gun
##	2487	2770	2017-07-11	shoot	foot	gun
##	2491	2774	2017-07-13	threat		gun
##	2493	2776	2017-07-13	attack	not	knife
##	2520	2803	2017-07-25	shoot	not	gun
##	2522	2804	2017-07-26	shoot	other	gun
##	2584	2883	2017-08-17	threat	not	knife
##	2643	2935	2017-09-10	attack	not	gun
##	2651	2948	2017-09-13	shoot	car	gun
##	2676	2980	2017-09-22	point	not	gun
##	2705	3021	2017-10-04	threat	not	gun
##	2722	3033	2017-10-08	shoot	not	gun
##	2725	3027	2017-10-11	threat	not	gun
##	2728	3031	2017-10-12	shoot	not	gun
##	2736	3041	2017-10-16	threat	not	gun
##	2739	3044	2017-10-18	shoot	foot	gun
##	2809	3126	2017-11-10	attack	not	knife
##	2846	3166	2017-11-24	threat	not	unknown
##	2874	3190	2017-12-07	attack	car	
##	2961	3277	2018-01-07	move	not	unarmed
##	2992	4244	2018-01-17	threat	not	knife
##	3061	3402	2018-02-09	threat	not	gun
##	3065	3404	2018-02-10	point	not	gun
##	3078	3408	2018-02-15	point	not	gun
##	3097	3432	2018-02-22	threat	foot	gun
##	3105	3438	2018-02-26	point		gun
##	3115	3466	2018-03-01	threat	not	unknown
##	3117	3454	2018-03-02	shoot	foot	gun
##	3118	3458	2018-03-02	threat	not	knife
##	3120	3460	2018-03-02	threat	not	gun
##	3147	3492	2018-03-09	shoot	foot	gun
##	3165	3514	2018-03-15	attack	not	unarmed
##	3194	3540	2018-03-23	attack	not	knife
##	3196	3542	2018-03-23	threat	not	other
##	3207	3554	2018-03-27	point	not	gun
##	3213	3550	2018-03-28	threat	not	knife
##	3230	4255	2018-04-01	threat	foot	gun
##	3237	3584	2018-04-04	shoot	foot	gun
##	3276	3618	2018-04-14	point	other	gun

##	3278	3620	2018-04-14	shoot	not	gun
##	3427	3766	2018-06-12	threat	not	unarmed
##	3478	3818	2018-06-29	shoot	not	gun
##	3505	3831	2018-07-08	threat	not	gun
##	3522	3868	2018-07-17	shoot	car	gun
##	3534	3874	2018-07-21	attack	foot	vehicle;gun
##	3535	3878	2018-07-21	attack	not	vehicle
##	3551	3959	2018-07-25	move	not	unarmed
##	3569	3963	2018-07-31	shoot	not	gun
##	3571	3909	2018-08-01	threat	not	gun
##	3592	3933	2018-08-07	threat	other	gun
##	3616	3982	2018-08-18	move	foot	gun
##	3634	3990	2018-08-27	attack	car	gun;vehicle
##	3635	3991	2018-08-27	shoot	not	gun
##	3666	4041	2018-09-14	point	not	gun
##	3690	4082	2018-09-24	shoot	not	gun
##	3738	4100	2018-10-12	shoot	not	gun
##	3776	4137	2018-10-29	shoot	car	gun
##	3814	4175	2018-11-13	attack	car	unknown
##	3892	4304	2018-12-16	point	not	gun
##	3913	4325	2018-12-25	shoot	not	gun
##	3920	4346	2018-12-28	move	not	gun
##	3964	4390	2019-01-11	threat	not	gun
##	3981	4386	2019-01-15	threat	foot	replica
##	3997	4405	2019-01-23	undetermined	foot	unarmed
##	4001	4414	2019-01-23	shoot	foot	gun
##	4003	4514	2019-01-24	move	car	unarmed
##	4019	4511	2019-01-30	point	foot	gun
##	4023	4432	2019-02-02	attack	not	knife
##	4118	4557	2019-03-09	point	not	gun
##	4134	4570	2019-03-13	flee	other	unarmed
##	4177	4611	2019-03-31	attack	not	knife
##	4210	4643	2019-04-12	attack	not	blunt_object
##	4245	4693	2019-04-29	point	foot	replica
##	4262	5417	2019-05-04	point	not	knife
##	4264	4699	2019-05-06	threat	not	gun
##	4271	4703	2019-05-09	point	not	gun
##	4292	4725	2019-05-19	shoot	foot	gun
##	4375	4790	2019-06-16	attack		gun
##	4569	5002	2019-08-30	attack	other	gun;vehicle
##	4599	5027	2019-09-09	point	not	gun;knife
##	4682	5089	2019-10-13	shoot	not	gun
##	4684	5221	2019-10-13	shoot		gun
##	4751	5154	2019-11-03	shoot	not	gun
##	4753	5172	2019-11-04	attack	not	knife
##	4787	5189	2019-11-16	point	not	gun
##	4862	5277	2019-12-14	shoot	not	gun
##	4884	5299	2019-12-20	attack	not	other
##	4894	5317	2019-12-23	shoot	car	gun;vehicle
##	4920	5332	2019-12-31	threat	not	knife
##	4944	5353	2020-01-06	shoot	foot	gun
##	4979	5391	2020-01-20	threat	not	knife
##	4986	5508	2020-01-22	shoot		gun
##	4999	5424	2020-01-27	attack	car	vehicle

##	5003	5440	2020-01-28	attack	not	knife
##	5004	5442	2020-01-28	point	not	gun
##	5085	5579	2020-02-27	threat		knife
##	5113	5609	2020-03-08	point	not	gun
##	5128	5654	2020-03-13	point	not	replica
##	5176	5697	2020-03-29	shoot	not	gun
##	5187	5710	2020-04-02	attack	other	other
##	5200	5728	2020-04-07	attack	not	knife
##	5256	5788	2020-04-28	shoot	car	gun
##	5331	5870	2020-05-21	move	not	gun
##	5341	5841	2020-05-25	move	not	unarmed
##	5409	5927	2020-06-16	shoot	not	gun
##	5455	5969	2020-07-04	undetermined	not	undetermined
##	5459	5972	2020-07-05	point	not	gun
##	5555	6049	2020-08-13	attack	not	gun
##	5575	6103	2020-08-19	shoot	not	gun
##	5593	6121	2020-08-26	point	foot	gun
##	5659	6198	2020-09-25	move	car	unarmed
##	5723	6250	2020-10-20	point	not	gun
##	5727	6334	2020-10-20	threat	not	unarmed
##	5731	6258	2020-10-22	point	car	gun
##	5747	6253	2020-10-25	point	not	replica
##	5756	6273	2020-10-28	point	foot	gun
##	5853	6363	2020-11-30	point	not	replica
##	5861	6396	2020-12-03	shoot	not	gun
##	5866	6428	2020-12-04	shoot		gun
##	5915	6437	2020-12-22	threat	not	blunt_object
##	5921	6455	2020-12-25	attack	car	vehicle
##	5932	6519	2020-12-28	move	foot	unarmed
##	5948	6542	2021-01-02	threat	other	gun
##	5953	6509	2021-01-04	threat	not	knife
##	5970	6681	2021-01-09	shoot	foot	gun
##	5982	6502	2021-01-12	threat	other	gun
##	6001	6539	2021-01-20		car	undetermined
##	6020	6556	2021-01-30	threat	not	gun
##	6025	6553	2021-02-01	attack	other	vehicle
##	6050	6584	2021-02-06	point	not	gun
##	6172	7579	2021-03-22	shoot	foot	gun
##	6220	6743	2021-04-03	point	foot	gun
##	6267	6823	2021-04-17	threat	car	undetermined
##	6284	6800	2021-04-23	threat	not	gun
##	6302	7630	2021-05-03	threat	not	blunt_object
##	6316	7628	2021-05-09	threat	foot	unknown
##	6334	6863	2021-05-17	attack		undetermined
##	6360	6883	2021-05-24	flee	car	undetermined
##	6374	6894	2021-05-31	point	not	gun
##	6389	6910	2021-06-04	threat	not	knife
##	6413	7605	2021-06-10	shoot	car	gun
##	6506	7012	2021-07-08	point	not	replica
##	6530	7395	2021-07-18	shoot	car	gun
##	6536	7068	2021-07-20	point	not	knife
##	6736	7291	2021-09-29	move	not	gun
##	6838	7328	2021-11-03	attack	other	vehicle
##	6901	7390	2021-11-29	move		knife

##	6980	7511	2021-12-26	attack	not	knife
##	6998	7459	2022-01-03	move	not	blunt_object
##	7000	7448	2022-01-04	point	car	gun
##	7044	7717	2022-01-22	threat	not	gun
##	7085	7696	2022-02-07	threat	not	gun
##	7087	7718	2022-02-07	threat	not	knife
##	7089	7694	2022-02-08	point	not	replica
##	7125	7750	2022-02-17	threat	other	gun
##	7178	7782	2022-03-06	undetermined	car	unarmed
##	7192	7790	2022-03-12	attack	not	knife
##	7194	7793	2022-03-12	attack	car	vehicle
##	7242	7876	2022-03-27	threat		gun
##	7247	7882	2022-03-29	point		gun
##	7297	7969	2022-04-09	flee	foot	undetermined
##	7311	7934	2022-04-14	shoot	foot	gun
##	7374	8027	2022-05-10	threat		knife
##	7392	8050	2022-05-18	shoot	not	gun
##	7413	8084	2022-05-23	threat		knife
##	7433	8091	2022-05-29	shoot		gun
##	7552	8207	2022-07-05	move	not	gun
##	7590	8248	2022-07-17	threat		knife
##	7678	8345	2022-08-20	threat		gun
##	7696	8359	2022-08-25	point		gun
##	7737	8397	2022-09-07	threat	foot	knife
##	7769	8439	2022-09-15	flee	foot	unarmed
##	7798	8449	2022-09-25	attack		blunt_object
##	7889	8548	2022-10-27	shoot	car	gun
##	7907	8638	2022-11-02	shoot		gun
##	7939	8586	2022-11-12	threat		knife
##	7974	8677	2022-11-23	threat	foot	gun
##				city	county	state
##	7			Chandler	Maricopa	AZ
##	43			Maricopa	Pinal	AZ
##	46			Scottsdale	Maricopa	AZ
##	67			Morenci	Greenlee	AZ
##	86			Tempe	Maricopa	AZ
##	87			Tempe	Maricopa	AZ
##	110			San Manuel	Pinal	AZ
##	116			Marana	Pima	AZ
##	159			Tempe	Maricopa	AZ
##	168			Phoenix	Maricopa	AZ
##	183			Kingman	Mohave	AZ
##	215			Cottonwood	Yavapai	AZ
##	229			Glendale	Maricopa	AZ
##	257			Phoenix	Maricopa	AZ
##	291			Phoenix	Maricopa	AZ
##	301			Bisbee	Cochise	AZ
##	323			Mesa	Maricopa	AZ
##	345			Kearny	Pinal	AZ
##	379			Eagar	Apache	AZ
##	398			Cottonwood	Yavapai	AZ
##	421			Sun City	Maricopa	AZ
##	491			Mesa	Maricopa	AZ
##	501			Phoenix	Maricopa	AZ



## 535	Phoenix	Maricopa	AZ 33.55312 -112.1516
## 572	Tucson	Pima	AZ 32.27235 -110.9735
## 602	Gilbert	Maricopa	AZ 33.36471 -111.7560
## 646	Phoenix	Maricopa	AZ 33.46595 -112.1345
## 663	Tucson	Pima	AZ 32.14670 -110.9596
## 668	Prescott	Yavapai	AZ 34.71986 -112.5433
## 736	Wilhoit	Yavapai	AZ 34.41560 -112.5894
## 747	Sierra Vista	Cochise	AZ 31.46726 -110.2333
## 895	Eagar	Apache	AZ 34.11615 -109.2929
## 902	Sun City	Maricopa	AZ 33.59327 -112.2980
## 917	Mesa	Maricopa	AZ 33.46634 -111.6725
## 935	Tucson	Pima	AZ 32.30876 -110.8749
## 983	Phoenix	Maricopa	AZ 33.68004 -112.0617
## 988	Tempe	Maricopa	AZ 33.38054 -111.9215
## 1038	Mesa	Maricopa	AZ 33.41466 -111.6846
## 1039	Tucson	Pima	AZ 32.22174 -110.9265
## 1042	Phoenix	Maricopa	AZ 33.58210 -112.0976
## 1083	Mesa	Maricopa	AZ 33.43685 -111.6586
## 1148	Phoenix	Maricopa	AZ 33.59638 -112.1687
## 1151	Phoenix	Maricopa	AZ 33.46574 -112.2206
## 1168	Gilbert	Maricopa	AZ 33.32987 -111.7772
## 1249	Glendale	Maricopa	AZ 33.52343 -112.2248
## 1307	Chandler	Maricopa	AZ 33.25377 -111.8412
## 1326	Florence	Pinal	AZ 33.03145 -111.3873
## 1365	Phoenix	Maricopa	AZ 33.37774 -111.9784
## 1400	Tucson	Pima	AZ 32.23981 -110.9012
## 1454	Lake Havasu City	Mohave	AZ 34.49374 -114.3238
## 1514	Tucson	Pima	AZ 32.24814 -110.9719
## 1523	Flagstaff	Coconino	AZ 35.19369 -111.6343
## 1549	Tempe	Maricopa	AZ 33.36687 -111.9297
## 1550	Scottsdale	Maricopa	AZ 33.58246 -111.8828
## 1557	Tucson	Pima	AZ 32.17780 -110.9685
## 1594	Mesa	Maricopa	AZ 33.45145 -111.8397
## 1597	Phoenix	Maricopa	AZ 33.49495 -112.1517
## 1604	Apache Junction	Pinal	AZ 33.42959 -111.5472
## 1606	Tucson	Pima	AZ 32.18042 -111.1610
## 1618	Phoenix	Maricopa	AZ 33.48023 -112.0388
## 1633	Phoenix	Maricopa	AZ 33.48173 -112.1832
## 1665	Phoenix	Maricopa	AZ 33.44838 -112.0740
## 1667	Phoenix	Maricopa	AZ 33.48395 -112.1621
## 1683	Phoenix	Maricopa	AZ 33.42402 -112.0305
## 1714	Phoenix	Maricopa	AZ 33.67500 -112.0993
## 1717	Kingman	Mohave	AZ 35.21253 -114.0401
## 1745	Scottsdale	Maricopa	AZ 33.66792 -111.9614
## 1796	Kingman	Mohave	AZ 35.23940 -114.0377
## 1826	Show Low	Navajo	AZ 34.15983 -109.9830
## 1834	Tucson	Pima	AZ 31.97196 -110.9690
## 1865	Tucson	Pima	AZ 32.20084 -110.9734
## 1866	Surprise	Maricopa	AZ 33.60797 -112.4400
## 1976	Phoenix	Maricopa	AZ 33.56756 -112.1342
## 2038	Apache Junction	Pinal	AZ 33.42235 -111.5634
## 2095	Golden Shores	Mohave	AZ 34.76724 -114.4959
## 2116	Tucson	Pima	AZ 32.07406 -110.9591
## 2149	Phoenix	Maricopa	AZ 33.45144 -112.0695

## 2179	Phoenix Maricopa	AZ 33.78391 -112.1175
## 2202	Phoenix Maricopa	AZ 33.45154 -112.1981
## 2243	Avondale Maricopa	AZ 33.46071 -112.3409
## 2261	Glendale Maricopa	AZ 33.53845 -112.1707
## 2267	Prescott Yavapai	AZ 34.52693 -112.4779
## 2322	Kingman	AZ 35.18944 -114.0530
## 2331	Tempe Maricopa	AZ 33.39295 -111.9263
## 2343	Glendale Maricopa	AZ 33.55312 -112.1516
## 2352	Tucson Pima	AZ 32.14861 -110.9562
## 2370	Avondale Maricopa	AZ 33.47894 -112.2899
## 2427	Phoenix	AZ 33.45031 -112.0216
## 2434	Phoenix Maricopa	AZ 33.75570 -111.9905
## 2447	Glendale Maricopa	AZ 33.53769 -112.2344
## 2487	Golden Valley	AZ 35.22249 -114.2231
## 2491	Pima Pima	AZ 32.09074 -111.2231
## 2493	Goodyear	AZ NA NA
## 2520	Avondale Maricopa	AZ 33.49335 -112.2985
## 2522	Phoenix	AZ 33.68382 -112.1348
## 2584	Phoenix Maricopa	AZ 33.45858 -112.0291
## 2643	Youngtown	AZ 33.59956 -112.3017
## 2651	Quartzsite La Paz	AZ 33.66153 -114.2295
## 2676	Mesa Maricopa	AZ 33.39313 -111.8498
## 2705	Mesa Maricopa	AZ 33.41518 -111.8315
## 2722	Pinetop-Lakeside Navajo	AZ 34.15480 -109.9749
## 2725	Marana Pima	AZ 32.34980 -111.0979
## 2728	Flagstaff Coconino	AZ 35.19961 -111.6143
## 2736	Picture Rocks Pima	AZ 32.34455 -111.1994
## 2739	Tucson Pima	AZ 32.25115 -110.8254
## 2809	Phoenix Maricopa	AZ 33.46263 -112.0782
## 2846	Yuma Yuma	AZ 32.70247 -114.6675
## 2874	Tucson Pima	AZ 32.23664 -110.8805
## 2961	Sedona Coconino	AZ 34.86974 -111.7610
## 2992	Phoenix	AZ NA NA
## 3061	Flagstaff Coconino	AZ 35.19847 -111.6566
## 3065	Kingman Mohave	AZ 35.22467 -114.0144
## 3078	Mesa Maricopa	AZ 33.41097 -111.7925
## 3097	Phoenix Maricopa	AZ 33.40646 -112.2723
## 3105	Gilbert Maricopa	AZ 33.36937 -111.7048
## 3115	Tucson Pima	AZ 32.17744 -110.9826
## 3117	Glendale Maricopa	AZ 33.56747 -112.1517
## 3118	Mesa Maricopa	AZ 33.37917 -111.6363
## 3120	Scottsdale Maricopa	AZ 33.57711 -111.8863
## 3147	Phoenix Maricopa	AZ 33.48690 -112.1638
## 3165	Casa Grande Pinal	AZ 32.94180 -111.7212
## 3194	Phoenix	AZ 33.49467 -112.2206
## 3196	Douglas	AZ 31.35082 -109.5483
## 3207	Phoenix	AZ 33.45775 -112.1217
## 3213	Surprise	AZ 33.63875 -112.3247
## 3230	Phoenix	AZ NA NA
## 3237	Mesa	AZ 33.37929 -111.7396
## 3276	Phoenix Maricopa	AZ 33.48044 -112.0892
## 3278	Phoenix Maricopa	AZ 33.37775 -112.0733
## 3427	Phoenix Maricopa	AZ 33.48752 -112.0737
## 3478	Phoenix Maricopa	AZ 33.51380 -112.1447

## 3505	Tucson	Pima	AZ 32.22222	-110.7666
## 3522	Tucson	Pima	AZ 32.17805	-110.9264
## 3534	Tucson		AZ 32.13129	-111.0868
## 3535	Mesa		AZ 33.39314	-111.8746
## 3551	Lake Havasu		AZ NA	NA
## 3569	Glendale		AZ 33.37908	-111.7739
## 3571	Tucson		AZ 32.14093	-110.9125
## 3592	Salt River Reservation		AZ 33.51671	-111.7674
## 3616	Phoenix		AZ 33.39224	-112.0905
## 3634	Phoenix		AZ 33.45958	-112.1431
## 3635	Phoenix		AZ 33.63951	-112.1344
## 3666	Mesa		AZ 33.41545	-111.6718
## 3690	Peoria		AZ 33.58119	-112.2290
## 3738	Chandler		AZ 33.33515	-111.8591
## 3776	Phoenix		AZ NA	NA
## 3814	Phoenix		AZ 33.52389	-112.2031
## 3892	Mesa		AZ 33.43686	-111.6153
## 3913	Buckeye		AZ 33.43624	-112.5713
## 3920	Phoenix		AZ 33.61078	-112.1685
## 3964	Phoenix		AZ 33.50880	-112.2550
## 3981	Tempe		AZ NA	NA
## 3997	Flagstaff		AZ 35.18812	-111.6812
## 4001	Flagstaff		AZ 35.19233	-111.6790
## 4003	Ehrenberg		AZ 33.61833	-114.5054
## 4019	Phoenix		AZ 33.54123	-112.0648
## 4023	Flagstaff		AZ 35.20718	-111.6061
## 4118	Phoenix		AZ 33.48042	-111.9955
## 4134	Phoenix		AZ 33.45023	-112.1023
## 4177	Marana		AZ 32.35884	-111.1229
## 4210	Buckeye		AZ 33.33399	-112.5000
## 4245	Phoenix		AZ 33.45866	-112.1345
## 4262	Phoenix		AZ NA	NA
## 4264	Mesa		AZ 33.41548	-111.8357
## 4271	Phoenix		AZ 33.43684	-112.0825
## 4292	Phoenix		AZ 33.46051	-112.0476
## 4375	Tucson		AZ 32.14836	-111.0031
## 4569	Tolleson		AZ 33.45181	-112.2383
## 4599	Phoenix		AZ 33.61082	-112.1512
## 4682	Glendale		AZ 33.57289	-112.2056
## 4684	Buckeye		AZ 33.42129	-112.5442
## 4751	Phoenix		AZ 33.39968	-112.0213
## 4753	Glendale		AZ 33.56756	-112.1517
## 4787	Mesa		AZ 33.41076	-111.8577
## 4862	Tucson		AZ 32.26909	-110.9694
## 4884	Tempe		AZ 33.43398	-111.9263
## 4894	Phoenix		AZ 33.47640	-112.2255
## 4920	Globe		AZ NA	NA
## 4944	Phoenix		AZ 33.44503	-112.0752
## 4979	Sedona		AZ 34.85728	-111.7659
## 4986	Mesa		AZ 33.39315	-111.8487
## 4999	Chandler		AZ NA	NA
## 5003	Scottsdale		AZ 33.48766	-111.9262
## 5004	Tucson		AZ 32.17786	-110.9984
## 5085	Clarkdale		AZ 34.76818	-112.0637

## 5113	Apache Junction	AZ 33.41509 -111.5807
## 5128	Chandler	AZ 33.30538 -111.8358
## 5176	Phoenix	AZ NA NA
## 5187	Mesa	AZ 33.40777 -111.8230
## 5200	El Mirage	AZ 33.60022 -112.3165
## 5256	Buckeye	AZ 33.42119 -112.5564
## 5331	Phoenix	AZ 33.30607 -112.0557
## 5341	Phoenix	AZ 33.67428 -111.9745
## 5409	Phoenix	AZ 33.49502 -112.0999
## 5455	Phoenix	AZ 33.49477 -112.2082
## 5459	Mesa	AZ 33.42279 -111.7191
## 5555	Phoenix	AZ 33.50941 -112.1108
## 5575	Lake Havasu City	AZ 34.50880 -114.2950
## 5593	Tucson	AZ 32.17061 -110.9920
## 5659	Tempe	AZ 33.42224 -111.8834
## 5723	Phoenix	AZ 33.50929 -112.0345
## 5727	Phoenix	AZ 33.50929 -112.0345
## 5731	Casa Grande	AZ 32.89416 -111.7573
## 5747	Phoenix	AZ 33.73417 -112.1187
## 5756	Tucson	AZ 32.13495 -110.9985
## 5853	Phoenix	AZ 33.49476 -112.1121
## 5861	Phoenix	AZ 33.56761 -112.0995
## 5866	Tonopah	AZ 33.49348 -112.9371
## 5915	Glendale	AZ 33.55308 -112.1862
## 5921	Rio Rico	AZ NA NA
## 5932	Phoenix	AZ 33.54586 -112.1170
## 5948	Chandler	AZ 33.31073 -111.8364
## 5953	Payson	AZ 34.24501 -111.3215
## 5970	Phoenix	AZ 33.45131 -112.0129
## 5982	Phoenix	AZ NA NA
## 6001	Tucson	AZ 32.37197 -111.0132
## 6020	Paradise Valley	AZ 33.54065 -111.9282
## 6025	Tucson	AZ 32.18610 -110.9680
## 6050	Mesa	AZ 33.40751 -111.8746
## 6172	Tucson	AZ NA NA
## 6220	Phoenix	AZ 33.49467 -112.2206
## 6267	Kingman	AZ 35.26681 -114.0394
## 6284	Casa Grande	AZ 32.88681 -111.7595
## 6302	Glendale	AZ NA NA
## 6316	Glendale	AZ NA NA
## 6334	Peoria	AZ 33.55291 -112.2704
## 6360	Nogales	AZ 31.36030 -110.9326
## 6374	Tucson	AZ 32.27198 -110.9733
## 6389	Tucson	AZ 32.21568 -110.8151
## 6413	Wilcox	AZ NA NA
## 6506	Phoenix	AZ 33.55309 -112.1170
## 6530	Tucson	AZ NA NA
## 6536	Phoenix	AZ 33.45856 -112.0172
## 6736	Tempe	AZ 33.37841 -111.9372
## 6838	Glendale	AZ 33.53487 -112.1928
## 6901	Tucson	AZ NA NA
## 6980	Buckeye	AZ NA NA
## 6998	Mesa	AZ NA NA
## 7000	Paulden	AZ NA NA

## 7044	Phoenix	AZ	NA	NA
## 7085	Marana	AZ	NA	NA
## 7087	Phoenix	AZ	NA	NA
## 7089	Mesa	AZ	NA	NA
## 7125	Beaver Dam	AZ	NA	NA
## 7178	Peoria	AZ	33.66715	-112.2800
## 7192	Tucson	AZ	32.17574	-111.0786
## 7194	Phoenix	AZ	NA	NA
## 7242	Maricopa	AZ	33.07315	-111.9963
## 7247	Tucson	AZ	32.21592	-110.8095
## 7297	Gilbert	AZ	33.30682	-111.7555
## 7311	Glendale	AZ	33.59927	-112.1859
## 7374	Yuma	AZ	32.71308	-114.6258
## 7392	Guadalupe	AZ	33.36006	-111.9631
## 7413	Arizona City	AZ	32.75259	-111.6803
## 7433	Safford	AZ	32.80234	-109.7078
## 7552	Glendale	AZ	NA	NA
## 7590	Chandler	AZ	33.30605	-111.8762
## 7678	Tucson	AZ	32.17845	-110.9812
## 7696	Tucson	AZ	32.14296	-110.9571
## 7737	Flagstaff	AZ	35.22362	-111.5841
## 7769	Kingman	AZ	35.22486	-114.0362
## 7798	Phoenix	AZ	33.53458	-112.0998
## 7889	Mesa	AZ	33.40765	-111.8567
## 7907	Phoenix	AZ	33.46595	-112.1388
## 7939	Goodyear	AZ	33.33499	-112.4298
## 7974	Chandler	AZ	33.29892	-111.7909
##	location_precision		name	age gender race
## 7	not_available		Kenneth Arnold Buck	22 male H
## 43	not_available		Jonathan Guillory	32 male W
## 46	not_available		Dewayne Carr	42 male B
## 67	not_available		Chris Ingram	29 male W
## 86	not_available		Salvador Muna	28 male H
## 87	not_available		Joaquin Hernandez	28 male H
## 110	not_available		Daniel Mejia	37 male H
## 116	not_available		Lawrence Caldwell	56 male W
## 159	not_available		Matthew Metz	25 male W
## 168	not_available		Adam Reinhart	29 male W
## 183	not_available		Ryan Dean Burgess	31 male W
## 215	not_available		Enoch Gaver	21 male W
## 229	not_available		Joseph Tassinari	63 male W
## 257	not_available		Ken Cockerel	51 male W
## 291	not_available		Erik Tellez	43 male H
## 301	not_available	Carlos Saavedra Ramirez	51 male	H
## 323	not_available		Joshua Deysie	33 male H
## 345	not_available		Shaun Johnson	35 male W
## 379	not_available		Eric Robinson	40 male W
## 398	not_available	Ebin Lamont Proctor	19 male	W
## 421	not_available	Richard Warolf	69 male	W
## 491	not_available	Adam Dujanovic	33 male	W
## 501	not_available	Robert Hammonds	68 male	W
## 535	not_available	Kevin Thomas Snyder	46 male	W
## 572	not_available	Armando Serrano Jr.	29 male	H
## 602	not_available	Richard Tyler Young	24 male	W

## 646	not_available	Shane Rudolph	45	male	W
## 663	not_available	David M. Leon	40	male	H
## 668	not_available	Arthur Bates	45	male	W
## 736	not_available	Jeffrey Blood	45	male	W
## 747	not_available	David M. Diaz	28	male	H
## 895	not_available	Douglas R. Slade	52	male	W
## 902	not_available	Ralph Aguilar	69	male	H
## 917	not_available	Ivan Krstic	47	male	W
## 935	not_available	Steven Wickert	36	male	W
## 983	not_available	Lonnie Niesen	41	male	W
## 988	not_available	Sean Mould	34	male	W
## 1038	not_available	Daniel Shaver	26	male	W
## 1039	not_available	Jordan Szymanski	33	male	W
## 1042	not_available	Levi Gene Wilson	39	male	W
## 1083	not_available	Kayden Clarke	24	male	W
## 1148	not_available	Alex Buckner	26	male	W
## 1151	not_available	Victor Rivera	27	male	H
## 1168	not_available	Sergio Ochoa	27	male	H
## 1249	not_available	Ruben Cossyleon	25	male	H
## 1307	not_available	Mitchell Oakley	24	male	W
## 1326	not_available	Bender Dass	55	male	W
## 1365	not_available	Israel Santos-Banos	20	male	H
## 1400	not_available	Osee Calix	33	male	B
## 1454	not_available	Devin Christopher Scott	20	male	W
## 1514	not_available	Abraham Smith	30	male	W
## 1523	not_available	Donald S. Myers	32	male	W
## 1549	not_available	Dalvin Hollins	19	male	B
## 1550	not_available	Dylan Liberti	24	male	W
## 1557	not_available	Jesus Rael	31	male	H
## 1594	not_available	Jesus Armando Carillo	22	male	H
## 1597	not_available	Ruben Horacio Strand Alvear	38	male	H
## 1604	not_available	Larry Eugene Kurtley Jr.	53	male	W
## 1606	not_available	Marcos Antonio Gastelum	25	male	H
## 1618	not_available	Jorge Cenicerros	22	male	H
## 1633	not_available	Angel Torres Jr.	26	male	H
## 1665	not_available	Steven Del Rio	31	male	H
## 1667	not_available	Dylan Papa	25	male	W
## 1683	not_available	Robert Carrillo	32	male	H
## 1714	not_available	John Ethan Carpentier	26	male	W
## 1717	not_available	Jeffrey Clair Cave	53	male	W
## 1745	not_available	Luis Michael Hoff	40	male	W
## 1796	not_available	Kenneth Jack Kennedy	34	male	W
## 1826	not_available	Daniel Erickson	36	male	W
## 1834	not_available	Richard T. Herrera Jr.	29	male	H
## 1865	not_available	Carlos Valencia	26	male	H
## 1866	not_available	Derek Adame	20	male	H
## 1976	not_available	JR Williams	38	male	B
## 2038	not_available	Bradley Joseph Szacaks	45	male	W
## 2095	not_available	Drey Krause	53	male	W
## 2116	not_available	Manuel Encinas	26	male	H
## 2149	not_available	Michael Cisneros	27	male	H
## 2179	not_available	Todd Munson	25	male	W
## 2202	not_available	Francisco Valdez	24	male	H
## 2243	not_available	Erik Pamias	28	male	W

## 2261	not_available	Brandon Pequeno	25	male	H
## 2267	not_available	Wayne Noel Simard	70	male	W
## 2322	not_available	David Jay Juarez	41	male	H
## 2331	not_available	Tristan Long	25	male	B
## 2343	not_available	Jorge Alberto Fuentes	21	male	H
## 2352	not_available	Joseph Zimmerman	34	male	W
## 2370	not_available	Donald Lee Cramer	44	male	W
## 2427	not_available	Jesus Ramon Deltoro	34	male	H
## 2434	not_available	Nicholas Johnston	47	male	W
## 2447	not_available	Patrick Sanchez	23	male	H
## 2487	not_available	Silas Andrew Smith	54	male	W
## 2491	not_available	Chancey Chamblee	55	male	W
## 2493		Pedro Rubio	42	male	H
## 2520	not_available	Moises Balladares	33	male	H
## 2522	not_available	Andrew Collins	35	male	W
## 2584	not_available	Edgar Ramirez Carreto	28	male	H
## 2643	not_available	Erick Perez	22	male	H
## 2651	not_available	Rodolfo Ballardo	31	male	H
## 2676	not_available	Scott Farnsworth	28	male	W
## 2705	not_available	Michael Ferrell	67	male	W
## 2722	not_available	Glenn Southwood Jr.	46	male	W
## 2725	not_available	Brandon Wade Rucker	43	male	W
## 2728	not_available	Sean D. Brady	29	male	W
## 2736	not_available	Luis David Flores	34	male	H
## 2739	not_available	Daniel Spear	35	male	W
## 2809	not_available	Thomas Aikens	43	male	B
## 2846	not_available	Steve Steenhard	51	male	W
## 2874	not_available	Frederick Douglas Wilburn Jr.	34	male	B
## 2961	not_available	Tyler Miller	51	male	W
## 2992		Jordan Keckhut	22	male	H
## 3061	not_available	John Hamilton	78	male	W
## 3065	not_available	Michael Weber	63	male	W
## 3078	not_available	Arthur Joseph Gonzales	57	male	H
## 3097	not_available	Jacob Uptain	27	male	W
## 3105	not_available	Joshua Kinnard	37	male	W
## 3115	not_available	Joel Jacobo	29	male	H
## 3117	not_available	Stephen Hudak	44	male	W
## 3118	not_available	Christopher Race	36	male	W
## 3120	not_available	Erik Dunham	48	male	W
## 3147	not_available	David Gardea	27	male	H
## 3165	not_available	Cameron Hall	27	male	B
## 3194	not_available	Kevin Robles	23	male	W
## 3196	not_available	Robert Vega	48	male	H
## 3207	not_available	Jose Aaron Gonzalez	44	male	H
## 3213	not_available	George Pappas	61	male	W
## 3230		John Wussler	58	male	W
## 3237	not_available	Kelvin Baldwin	30	male	W
## 3276	not_available	Andre Lavance Rippy	39	male	B
## 3278	not_available	Eduardo Andrade	44	male	H
## 3427	not_available	Alexandre J. Aldrich	34	male	W
## 3478	not_available	Chukwumankpam Mbegbu	19	male	B
## 3505	not_available	Craig Yelton	34	male	W
## 3522	not_available	Juan Manuel Correa-Leyva	27	male	H
## 3534	not_available	Vincent James Ewer II	39	male	W

## 3535	not_available	Anthony C. Lopez	21	male	H
## 3551		Brent C. Bowdon	56	male	W
## 3569	not_available	Skyler Martin	23	male	H
## 3571	not_available	David Judge	53	male	W
## 3592	not_available	Pedro Lucero	42	male	H
## 3616	not_available	Andres Arteaga	28	male	H
## 3634	not_available	Fabian Ortiz Adame	25	male	W
## 3635	not_available	Bryce Schenkenberg	20	male	W
## 3666	not_available	David Huffines Jr.	40	male	W
## 3690	not_available	Michael Stout	20	male	W
## 3738	not_available	Kay Kenniker	84	male	W
## 3776		Arnaldo Caraveo	27	male	H
## 3814	not_available	Timothy Leon	24	male	B
## 3892	not_available	Edward Rudhman	50	male	W
## 3913	not_available	Antonio A. Ramos	34	male	H
## 3920	not_available	Edwin C. Bundy	46	male	W
## 3964	not_available	Jacob Michael Harris	19	male	B
## 3981		Antonio Arce	14	male	H
## 3997	not_available	Preston Oszust	20	male	W
## 4001	not_available	Marcus Gishal	20	male	H
## 4003	not_available	Christian Albarran	26	male	H
## 4019	not_available	Michael Joe Jolls	47	male	W
## 4023	not_available	Henry Harold Russell	47	male	W
## 4118	not_available	Eric Hagstrom	35	male	W
## 4134	not_available	Henry Wayne Rivera	30	male	H
## 4177	not_available	Juan Padilla	24	male	H
## 4210	not_available	Juan Torres	39	male	H
## 4245	not_available	Alejandro Hernandez	26	male	H
## 4262		Jason Demarcus Larkin	35	male	W
## 4264	not_available	Pedro Colazo-villa	33	male	H
## 4271	not_available	Hector Lopez	29	male	H
## 4292	not_available	Robert Rabago	17	male	H
## 4375	not_available	Carlos Vale	31	male	H
## 4569	not_available	Rene Enrique Ruiz	51	male	H
## 4599	not_available	Hector Miranda	47	male	H
## 4682	not_available	Levy Isacc Madueno Santibanez	17	male	H
## 4684	not_available	Jose Eduardo Flores	26	male	H
## 4751	not_available	Michael Austin	32	male	B
## 4753	not_available	Matthew Rasmussen	31	male	W
## 4787	not_available	Garrett Ryberg	27	male	W
## 4862	not_available	Victor A. Parsons	42	male	W
## 4884	not_available	Mohamed Ahemed Al-Hashemi	25	male	B
## 4894	not_available	Richard Ruiz	38	male	H
## 4920		Eric M. Tellez	28	male	W
## 4944	not_available	Xavier Jaime Rovie	28	male	H
## 4979	not_available	Jonathan David Messare	41	male	W
## 4986	not_available	Armando Moreno Garcia	37	male	H
## 4999		Kevin Alan Smallman	32	male	W
## 5003	not_available	Joshua Greenleaf	27	male	W
## 5004	not_available	Robert Cocio	39	male	H
## 5085	not_available	Joshua David Hernandez Lord	33	male	W
## 5113	not_available	Lawson Edward Schaber	84	male	W
## 5128	not_available	Rosario Angel Alvarado	41	male	H
## 5176		Jacob Emry Mcilveen	22	male	W



## 5187	not_available	Jose Moreno	30	male	H
## 5200	not_available	Joshua Bacco	31	male	W
## 5256	not_available	Robert Musser	32	male	W
## 5331	not_available	Ryan Whitaker	40	male	W
## 5341	not_available	Dion Johnson	28	male	B
## 5409	not_available	Donald Ward	27	male	B
## 5455	not_available	James Porter Garcia	28	male	H
## 5459	not_available	Rodney Liveringhouse	70	male	W
## 5555	not_available	Derick Manuel Bonilla	39	male	H
## 5575	not_available	Ronald Stuart Chipman	39	male	H
## 5593	not_available	Simon Pancho	17	male	H
## 5659	not_available	Angel Benitez	21	male	H
## 5723	not_available	George Cocreham	43	male	H
## 5727	not_available	Emmett Cocreham	44	male	H
## 5731	not_available	Francisco Danny Flores	30	male	H
## 5747	not_available	Paul Sarver	53	male	W
## 5756	not_available	Charles Robert Arviso	35	male	H
## 5853	not_available	Ekou Udofia	33	male	B
## 5861	not_available	Maurice Jackson	42	male	B
## 5866	not_available	Leonard Francis Kieren	60	male	W
## 5915	not_available	Christopher Cuevas	45	male	H
## 5921		John Moreno	32	male	H
## 5932	not_available	Jordan Crawford	30	male	W
## 5948	not_available	Anthony Bernal Cano	17	male	H
## 5953	not_available	Michael Romo	28	male	W
## 5970	not_available	Paul Bolden	37	male	B
## 5982		Antonio Carbajal	23	male	H
## 6001	not_available	Bradley Alexander Lewis	19	male	W
## 6020	not_available	Kenneth Michael Dallas	61	male	W
## 6025	not_available	Andrew Scott Kislek	34	male	W
## 6050	not_available	Keenan Sailer	23	male	W
## 6172		Edward Kayer	42	male	W
## 6220	not_available	Jose Arenas	26	male	H
## 6267	not_available	Bradley Michael Rose	29	male	W
## 6284	not_available	Edward Robbesom	52	male	W
## 6302		Mariano Villegas	45	male	H
## 6316		Maxwell Jerome Davis	28	male	W
## 6334	not_available	Garrett Sheppard	32	male	W
## 6360	not_available	Glen Ray Cockrum Jr.	39	male	W
## 6374	not_available	Joshua Lee Moore	33	male	W
## 6389	not_available	Raymond Edwards	45	male	W
## 6413		Josiah Li'Shon Byard	21	male	B
## 6506	not_available	Stanley Howard	64	male	W
## 6530		Leslie Stephen Scarlett	35	male	B
## 6536	not_available	Miguel F. Hernandez-Rodriguez	31	male	H
## 6736	not_available	Stephen Phil Franco	37	male	H
## 6838	not_available	Jose Enrique Cortez	26	male	H
## 6901		Richard Lee Richards	61	male	W
## 6980		Vincent Black	31	male	W
## 6998		James Schild	49	male	W
## 7000		Carl Schaede	53	male	W
## 7044		Dedrick Garcia	37	male	W
## 7085		Anthony Parker	48	male	W
## 7087		Emilio Chamizo	30	male	H

## 7089		Brian W. Schneider	36	male	W
## 7125		Oscar Darwin Alcantara	30	male	H
## 7178	not_available	Jerry William Lott	33	male	W
## 7192	not_available	Zakareya Ibrahim	17	male	B
## 7194		Adam Vespoli	39	male	W
## 7242	not_available	Lawrence Knudsen	56	male	W
## 7247	not_available	Eric Putnam	27	male	W
## 7297	not_available	Jonatan Mosqueda	25	male	H
## 7311	not_available	Anthony Hood-Schaffner	28	male	B
## 7374	not_available	Felipe De Jesus Herrera Jr.	34	male	W
## 7392	not_available	Rolando Abel Rojas	37	male	H
## 7413	not_available	Arthuro Rivera	30	male	H
## 7433	not_available	Steven Paul John	51	male	W
## 7552		Juan Carlos Bojorquez	15	male	H
## 7590	not_available	Tyson W. Cobb	44	male	W
## 7678	not_available	Adum Mahamat	25	male	B
## 7696	not_available	Francisco Javier Galarza	49	male	H
## 7737	not_available	Donald Wayne Henry	51	male	W
## 7769	not_available	Marcus Adam Fuentes	36	male	H
## 7798	not_available	Ali Osman	34	male	B
## 7889	not_available	Breen Reiss	41	male	W
## 7907	not_available	Leontae Kirk	29	male	B
## 7939	not_available	Adam Romero	26	male	H
## 7974	not_available	Cody Allan Smestad	30	male	W
##	race_source	was_mental_illness_related	body_camera	agency_ids	year
## 7	not_available	False	False	195	2015
## 43	not_available	True	False	127	2015
## 46	not_available	False	False	510	2015
## 67	not_available	False	False	174;175	2015
## 86	not_available	False	False	247;195;2267;319	2015
## 87	not_available	False	False	247;195;2267;319	2015
## 110	not_available	True	False	348	2015
## 116	not_available	False	False	215	2015
## 159	not_available	False	False	247	2015
## 168	not_available	False	False	80	2015
## 183	not_available	True	False	596	2015
## 215	not_available	False	False	422	2015
## 229	not_available	True	False	62	2015
## 257	not_available	True	False	80	2015
## 291	not_available	True	False	80	2015
## 301	not_available	False	False	297	2015
## 323	not_available	True	False	319	2015
## 345	not_available	False	False	498	2015
## 379	not_available	False	False	614	2015
## 398	not_available	False	False	763	2015
## 421	not_available	True	False	127	2015
## 491	not_available	False	False	319	2015
## 501	not_available	False	False	80	2015
## 535	not_available	False	False	80	2015
## 572	not_available	False	False	397	2015
## 602	not_available	False	False	710	2015
## 646	not_available	False	False	80	2015
## 663	not_available	False	False	397	2015
## 668	not_available	True	True	763	2015

## 736	not_available	False	False	817;763	2015
## 747	not_available	False	False	834;835	2015
## 895	not_available	False	False	614	2015
## 902	not_available	True	False	944	2015
## 917	not_available	False	False	319	2015
## 935	not_available	False	False	964	2015
## 983	not_available	True	False	80	2015
## 988	not_available	True	True	247	2015
## 1038	not_available	False	True	319	2016
## 1039	not_available	True	False	397	2016
## 1042	not_available	True	False	80	2016
## 1083	not_available	True	True	319	2016
## 1148	not_available	True	False	80	2016
## 1151	not_available	False	False	80	2016
## 1168	not_available	False	False	710;319	2016
## 1249	not_available	False	False	80	2016
## 1307	not_available	True	False	195	2016
## 1326	not_available	False	False	348	2016
## 1365	not_available	False	False	80	2016
## 1400	not_available	False	False	397	2016
## 1454	not_available	False	True	1320	2016
## 1514	not_available	True	False	397	2016
## 1523	not_available	False	True	1268	2016
## 1549	not_available	True	False	247	2016
## 1550	not_available	False	True	510	2016
## 1557	not_available	False	False	215	2016
## 1594	not_available	False	False	319	2016
## 1597	not_available	True	False	80	2016
## 1604	not_available	False	False	944	2016
## 1606	not_available	False	False	964	2016
## 1618	not_available	False	False	80	2016
## 1633	not_available	False	False	80	2016
## 1665	not_available	False	False	80	2016
## 1667	not_available	True	False	80	2016
## 1683	not_available	False	False	80	2016
## 1714	not_available	False	False	80	2016
## 1717	not_available	False	True	596	2016
## 1745	not_available	True	False	817	2016
## 1796	not_available	False	False	1504	2016
## 1826	not_available	False	False	817	2016
## 1834	not_available	False	False	1526	2016
## 1865	not_available	False	True	1577	2016
## 1866	not_available	False	True	1540	2016
## 1976	not_available	False	True	80	2017
## 2038	not_available	False	False	1787	2017
## 2095	not_available	False	False	1320	2017
## 2116	not_available	False	False	964	2017
## 2149	not_available	False	False	80	2017
## 2179	not_available	False	False	80	2017
## 2202	not_available	False	False	80	2017
## 2243	not_available	False	False	80	2017
## 2261	not_available	False	False	62;319	2017
## 2267	not_available	False	False	1725	2017
## 2322	not_available	False	False	1504	2017

## 2331 not_available	False	False	247 2017
## 2343 not_available	False	False	62 2017
## 2352 not_available	True	False	397 2017
## 2370 not_available	False	False	1763 2017
## 2427 not_available	False	False	80 2017
## 2434 not_available	False	False	80 2017
## 2447 not_available	False	True	62 2017
## 2487 not_available	False	False	1504 2017
## 2491 not_available	False	False	964 2017
## 2493 not_available	False	False	944 2017
## 2520 not_available	True	False	1763 2017
## 2522 not_available	False	False	80 2017
## 2584 not_available	True	False	80 2017
## 2643 not_available	False	False	944;1875 2017
## 2651 not_available	False	False	817 2017
## 2676 not_available	True	False	319 2017
## 2705 not_available	False	False	319 2017
## 2722 not_available	False	False	1521 2017
## 2725 not_available	False	False	215 2017
## 2728 not_available	True	True	1268 2017
## 2736 not_available	False	False	964 2017
## 2739 not_available	False	False	397 2017
## 2809 not_available	False	False	80 2017
## 2846 not_available	False	False	1360 2017
## 2874 not_available	False	False	397 2017
## 2961 not_available	True	False	2013 2018
## 2992 not_available	False	False	80 2018
## 3061 not_available	False	True	1268 2018
## 3065 not_available	False	True	596 2018
## 3078 not_available	False	False	319 2018
## 3097 not_available	False	False	80 2018
## 3105 not_available	False	False	710 2018
## 3115 not_available	False	False	397 2018
## 3117 not_available	False	False	80 2018
## 3118 not_available	False	False	319 2018
## 3120 not_available	False	False	2086 2018
## 3147 not_available	False	False	80 2018
## 3165 not_available	False	False	1532 2018
## 3194 not_available	False	False	80 2018
## 3196 not_available	False	False	2123 2018
## 3207 not_available	False	False	80 2018
## 3213 not_available	False	False	1540 2018
## 3230 not_available	False	False	80 2018
## 3237 not_available	False	False	319 2018
## 3276 not_available	False	False	80 2018
## 3278 not_available	False	False	80 2018
## 3427 not_available	False	False	80 2018
## 3478 not_available	False	False	80 2018
## 3505 not_available	False	False	397 2018
## 3522 not_available	False	False	964 2018
## 3534 not_available	False	False	964 2018
## 3535 not_available	False	False	319 2018
## 3551 not_available	True	True	1320 2018
## 3569 not_available	False	False	62 2018

## 3571 not_available	False	False	397 2018
## 3592 not_available	False	False	2267 2018
## 3616 not_available	False	False	80 2018
## 3634 not_available	False	False	80 2018
## 3635 not_available	False	False	80 2018
## 3666 not_available	True	False	319 2018
## 3690 not_available	False	False	2301 2018
## 3738 not_available	True	False	195 2018
## 3776 not_available	False	False	319 2018
## 3814 not_available	False	False	62 2018
## 3892 not_available	True	False	2361 2018
## 3913 not_available	False	False	1342 2018
## 3920 not_available	False	False	80 2018
## 3964 not_available	False	False	80 2019
## 3981 not_available	False	True	247 2019
## 3997 not_available	False	False	817 2019
## 4001 not_available	False	False	1268 2019
## 4003 not_available	False	False	817 2019
## 4019 not_available	False	False	80 2019
## 4023 not_available	True	True	1268 2019
## 4118 not_available	False	False	80 2019
## 4134 not_available	False	False	80 2019
## 4177 not_available	False	False	215 2019
## 4210 not_available	True	True	944 2019
## 4245 not_available	False	False	80 2019
## 4262 not_available	False	False	1874 2019
## 4264 not_available	False	False	944 2019
## 4271 not_available	False	False	80 2019
## 4292 not_available	False	False	80 2019
## 4375 not_available	False	False	397 2019
## 4569 not_available	False	False	80 2019
## 4599 not_available	False	False	80 2019
## 4682 not_available	False	False	62 2019
## 4684 not_available	False	False	944 2019
## 4751 not_available	False	False	80 2019
## 4753 not_available	True	True	62 2019
## 4787 not_available	False	False	319 2019
## 4862 not_available	False	False	397 2019
## 4884 not_available	False	False	817 2019
## 4894 not_available	False	False	80 2019
## 4920 not_available	False	False	2673 2019
## 4944 not_available	False	False	80 2020
## 4979 not_available	True	False	2690 2020
## 4986 not_available	False	False	247 2020
## 4999 not_available	False	False	195 2020
## 5003 not_available	True	False	510 2020
## 5004 not_available	False	False	964;397;215 2020
## 5085 not_available	False	False	2740 2020
## 5113 not_available	False	False	1787 2020
## 5128 not_available	False	False	195 2020
## 5176 not_available	True	False	80 2020
## 5187 not_available	False	False	2361 2020
## 5200 not_available	True	False	1875 2020
## 5256 not_available	False	False	80 2020

## 5331 not_available	False	True	80 2020
## 5341 not_available	False	True	817 2020
## 5409 not_available	False	True	80 2020
## 5455 not_available	False	True	80 2020
## 5459 not_available	True	False	944 2020
## 5555 not_available	False	False	62 2020
## 5575 not_available	False	False	1320 2020
## 5593 not_available	False	True	397 2020
## 5659 not_available	False	True	319 2020
## 5723 not_available	True	True	80 2020
## 5727 not_available	False	True	80 2020
## 5731 not_available	False	False	1532 2020
## 5747 not_available	False	False	80 2020
## 5756 not_available	False	True	397 2020
## 5853 not_available	True	True	80 2020
## 5861 not_available	False	False	80 2020
## 5866 not_available	False	False	2986 2020
## 5915 not_available	False	False	62 2020
## 5921 not_available	False	False	2992 2020
## 5932 not_available	False	True	80 2020
## 5948 not_available	False	True	195 2021
## 5953 not_available	False	False	3001 2021
## 5970 not_available	False	True	80 2021
## 5982 not_available	False	True	80 2021
## 6001 not_available	False	False	964 2021
## 6020 not_available	False	False	944;3015 2021
## 6025 not_available	False	False	397 2021
## 6050 not_available	True	True	319 2021
## 6172 not_available	False	False	763 2021
## 6220 photo	False	False	80 2021
## 6267 photo	False	False	1504 2021
## 6284 not_available	False	False	1532 2021
## 6302 not_available	False	False	62 2021
## 6316 not_available	False	False	62 2021
## 6334 not_available	False	True	2301 2021
## 6360 photo	False	False	2992;3108 2021
## 6374 photo	False	True	397 2021
## 6389 photo	False	True	397 2021
## 6413 photo	False	False	817 2021
## 6506 photo	True	False	80 2021
## 6530 not_available	False	False	397 2021
## 6536 not_available	False	False	80 2021
## 6736 not_available	False	False	319 2021
## 6838 photo	False	False	62 2021
## 6901 photo	False	False	397 2021
## 6980 not_available	False	False	1342 2021
## 6998 not_available	False	False	319 2022
## 7000 not_available	False	False	1873 2022
## 7044 not_available	False	False	80 2022
## 7085 photo	False	False	817 2022
## 7087 not_available	False	False	80 2022
## 7089 not_available	False	False	319 2022
## 7125 photo	False	False	3364;3365 2022
## 7178 photo	False	True	2301 2022

## 7192	photo	True	False	964	2022
## 7194	photo	False	False	80	2022
## 7242	photo	True	False	944	2022
## 7247	public_record	False	True	397	2022
## 7297	photo	False	True	710	2022
## 7311	public_record	False	False	62	2022
## 7374	public_record	False	False	1360	2022
## 7392	public_record	False	False	944	2022
## 7413	public_record	False	False	348	2022
## 7433	photo	False	False	817;3474;3473	2022
## 7552	photo	False	False	62	2022
## 7590	photo	False	True	195	2022
## 7678	photo	False	False	397	2022
## 7696	photo	False	False	397	2022
## 7737	photo	False	False	1268	2022
## 7769	not_available	False	False	596	2022
## 7798	public_record	False	False	80	2022
## 7889	public_record	False	False	247	2022
## 7907	not_available	False	True	80	2022
## 7939	public_record	True	False	3622	2022
## 7974	not_available	False	True	195	2022
##	population				
## 7	6,832,810				
## 43	6,832,810				
## 46	6,832,810				
## 67	6,832,810				
## 86	6,832,810				
## 87	6,832,810				
## 110	6,832,810				
## 116	6,832,810				
## 159	6,832,810				
## 168	6,832,810				
## 183	6,832,810				
## 215	6,832,810				
## 229	6,832,810				
## 257	6,832,810				
## 291	6,832,810				
## 301	6,832,810				
## 323	6,832,810				
## 345	6,832,810				
## 379	6,832,810				
## 398	6,832,810				
## 421	6,832,810				
## 491	6,832,810				
## 501	6,832,810				
## 535	6,832,810				
## 572	6,832,810				
## 602	6,832,810				
## 646	6,832,810				
## 663	6,832,810				
## 668	6,832,810				
## 736	6,832,810				
## 747	6,832,810				
## 895	6,832,810				

## 902 6,832,810  
## 917 6,832,810  
## 935 6,832,810  
## 983 6,832,810  
## 988 6,832,810  
## 1038 6,944,767  
## 1039 6,944,767  
## 1042 6,944,767  
## 1083 6,944,767  
## 1148 6,944,767  
## 1151 6,944,767  
## 1168 6,944,767  
## 1249 6,944,767  
## 1307 6,944,767  
## 1326 6,944,767  
## 1365 6,944,767  
## 1400 6,944,767  
## 1454 6,944,767  
## 1514 6,944,767  
## 1523 6,944,767  
## 1549 6,944,767  
## 1550 6,944,767  
## 1557 6,944,767  
## 1594 6,944,767  
## 1597 6,944,767  
## 1604 6,944,767  
## 1606 6,944,767  
## 1618 6,944,767  
## 1633 6,944,767  
## 1665 6,944,767  
## 1667 6,944,767  
## 1683 6,944,767  
## 1714 6,944,767  
## 1717 6,944,767  
## 1745 6,944,767  
## 1796 6,944,767  
## 1826 6,944,767  
## 1834 6,944,767  
## 1865 6,944,767  
## 1866 6,944,767  
## 1976 7,048,088  
## 2038 7,048,088  
## 2095 7,048,088  
## 2116 7,048,088  
## 2149 7,048,088  
## 2179 7,048,088  
## 2202 7,048,088  
## 2243 7,048,088  
## 2261 7,048,088  
## 2267 7,048,088  
## 2322 7,048,088  
## 2331 7,048,088  
## 2343 7,048,088  
## 2352 7,048,088



## 2370 7,048,088  
## 2427 7,048,088  
## 2434 7,048,088  
## 2447 7,048,088  
## 2487 7,048,088  
## 2491 7,048,088  
## 2493 7,048,088  
## 2520 7,048,088  
## 2522 7,048,088  
## 2584 7,048,088  
## 2643 7,048,088  
## 2651 7,048,088  
## 2676 7,048,088  
## 2705 7,048,088  
## 2722 7,048,088  
## 2725 7,048,088  
## 2728 7,048,088  
## 2736 7,048,088  
## 2739 7,048,088  
## 2809 7,048,088  
## 2846 7,048,088  
## 2874 7,048,088  
## 2961 7,164,228  
## 2992 7,164,228  
## 3061 7,164,228  
## 3065 7,164,228  
## 3078 7,164,228  
## 3097 7,164,228  
## 3105 7,164,228  
## 3115 7,164,228  
## 3117 7,164,228  
## 3118 7,164,228  
## 3120 7,164,228  
## 3147 7,164,228  
## 3165 7,164,228  
## 3194 7,164,228  
## 3196 7,164,228  
## 3207 7,164,228  
## 3213 7,164,228  
## 3230 7,164,228  
## 3237 7,164,228  
## 3276 7,164,228  
## 3278 7,164,228  
## 3427 7,164,228  
## 3478 7,164,228  
## 3505 7,164,228  
## 3522 7,164,228  
## 3534 7,164,228  
## 3535 7,164,228  
## 3551 7,164,228  
## 3569 7,164,228  
## 3571 7,164,228  
## 3592 7,164,228  
## 3616 7,164,228

## 3634 7,164,228  
## 3635 7,164,228  
## 3666 7,164,228  
## 3690 7,164,228  
## 3738 7,164,228  
## 3776 7,164,228  
## 3814 7,164,228  
## 3892 7,164,228  
## 3913 7,164,228  
## 3920 7,164,228  
## 3964 7,291,843  
## 3981 7,291,843  
## 3997 7,291,843  
## 4001 7,291,843  
## 4003 7,291,843  
## 4019 7,291,843  
## 4023 7,291,843  
## 4118 7,291,843  
## 4134 7,291,843  
## 4177 7,291,843  
## 4210 7,291,843  
## 4245 7,291,843  
## 4262 7,291,843  
## 4264 7,291,843  
## 4271 7,291,843  
## 4292 7,291,843  
## 4375 7,291,843  
## 4569 7,291,843  
## 4599 7,291,843  
## 4682 7,291,843  
## 4684 7,291,843  
## 4751 7,291,843  
## 4753 7,291,843  
## 4787 7,291,843  
## 4862 7,291,843  
## 4884 7,291,843  
## 4894 7,291,843  
## 4920 7,291,843  
## 4944 7,179,943  
## 4979 7,179,943  
## 4986 7,179,943  
## 4999 7,179,943  
## 5003 7,179,943  
## 5004 7,179,943  
## 5085 7,179,943  
## 5113 7,179,943  
## 5128 7,179,943  
## 5176 7,179,943  
## 5187 7,179,943  
## 5200 7,179,943  
## 5256 7,179,943  
## 5331 7,179,943  
## 5341 7,179,943  
## 5409 7,179,943

## 5455 7,179,943  
## 5459 7,179,943  
## 5555 7,179,943  
## 5575 7,179,943  
## 5593 7,179,943  
## 5659 7,179,943  
## 5723 7,179,943  
## 5727 7,179,943  
## 5731 7,179,943  
## 5747 7,179,943  
## 5756 7,179,943  
## 5853 7,179,943  
## 5861 7,179,943  
## 5866 7,179,943  
## 5915 7,179,943  
## 5921 7,179,943  
## 5932 7,179,943  
## 5948 7,264,877  
## 5953 7,264,877  
## 5970 7,264,877  
## 5982 7,264,877  
## 6001 7,264,877  
## 6020 7,264,877  
## 6025 7,264,877  
## 6050 7,264,877  
## 6172 7,264,877  
## 6220 7,264,877  
## 6267 7,264,877  
## 6284 7,264,877  
## 6302 7,264,877  
## 6316 7,264,877  
## 6334 7,264,877  
## 6360 7,264,877  
## 6374 7,264,877  
## 6389 7,264,877  
## 6413 7,264,877  
## 6506 7,264,877  
## 6530 7,264,877  
## 6536 7,264,877  
## 6736 7,264,877  
## 6838 7,264,877  
## 6901 7,264,877  
## 6980 7,264,877  
## 6998 7,359,197  
## 7000 7,359,197  
## 7044 7,359,197  
## 7085 7,359,197  
## 7087 7,359,197  
## 7089 7,359,197  
## 7125 7,359,197  
## 7178 7,359,197  
## 7192 7,359,197  
## 7194 7,359,197  
## 7242 7,359,197

```
## 7247 7,359,197
## 7297 7,359,197
## 7311 7,359,197
## 7374 7,359,197
## 7392 7,359,197
## 7413 7,359,197
## 7433 7,359,197
## 7552 7,359,197
## 7590 7,359,197
## 7678 7,359,197
## 7696 7,359,197
## 7737 7,359,197
## 7769 7,359,197
## 7798 7,359,197
## 7889 7,359,197
## 7907 7,359,197
## 7939 7,359,197
## 7974 7,359,197
```

```
arizona_pop
```

```
##   Year Population Growth.Rate
## 1 2022 7,359,197      1.30%
## 2 2021 7,264,877      1.18%
## 3 2020 7,179,943     -1.53%
## 4 2019 7,291,843      1.78%
## 5 2018 7,164,228      1.65%
## 6 2017 7,048,088      1.49%
## 7 2016 6,944,767      1.64%
## 8 2015 6,832,810      1.48%
```

```
value_counts <- table(arizona_shootings_cleaned$year)
arizona_pop$shooting_count <- value_counts
arizona_pop$Population <- as.numeric(gsub(",", "", arizona_pop$Population))
arizona_pop
```

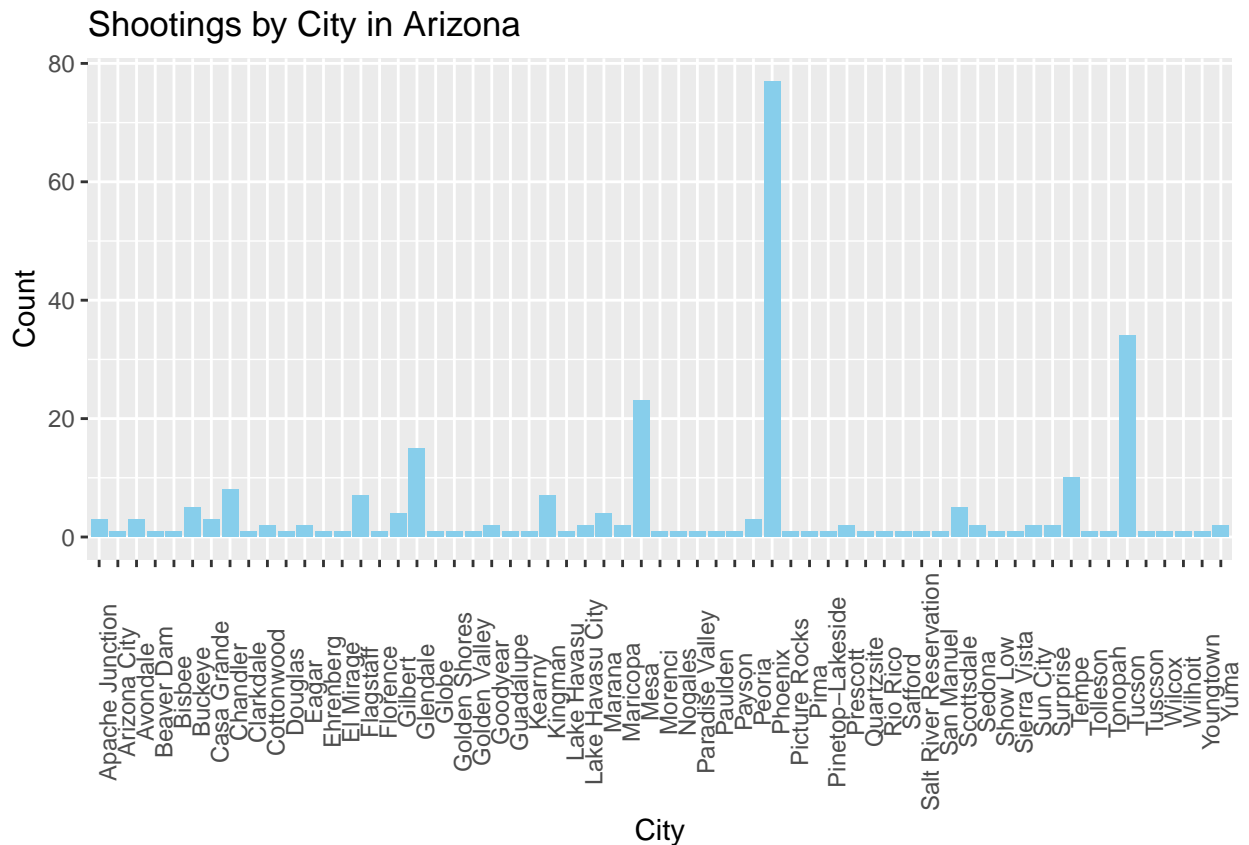
```
##   Year Population Growth.Rate shooting_count
## 1 2022    7359197      1.30%             37
## 2 2021    7264877      1.18%             35
## 3 2020    7179943     -1.53%             36
## 4 2019    7291843      1.78%             42
## 5 2018    7164228      1.65%             28
## 6 2017    7048088      1.49%             33
## 7 2016    6944767      1.64%             26
## 8 2015    6832810      1.48%             29
```

```
rate <- c(arizona_pop$Population / arizona_pop$shooting_count)
rate
```

```
## [1] 198897.2 207567.9 199442.9 173615.3 255865.3 213578.4 267106.4 235614.1
```

```
#city_counts.df <- table(azizona_shootings_cleaned$city)
#city_counts
```

```
ggplot(azizona_shootings_cleaned, aes(x = city)) +
  geom_bar(fill = 'skyblue') +
  labs(title = "Shootings by City in Arizona",
       x = "City",
       y = "Count") +
  theme(axis.text.x = element_text(angle = 90, hjust = 0.5))
```



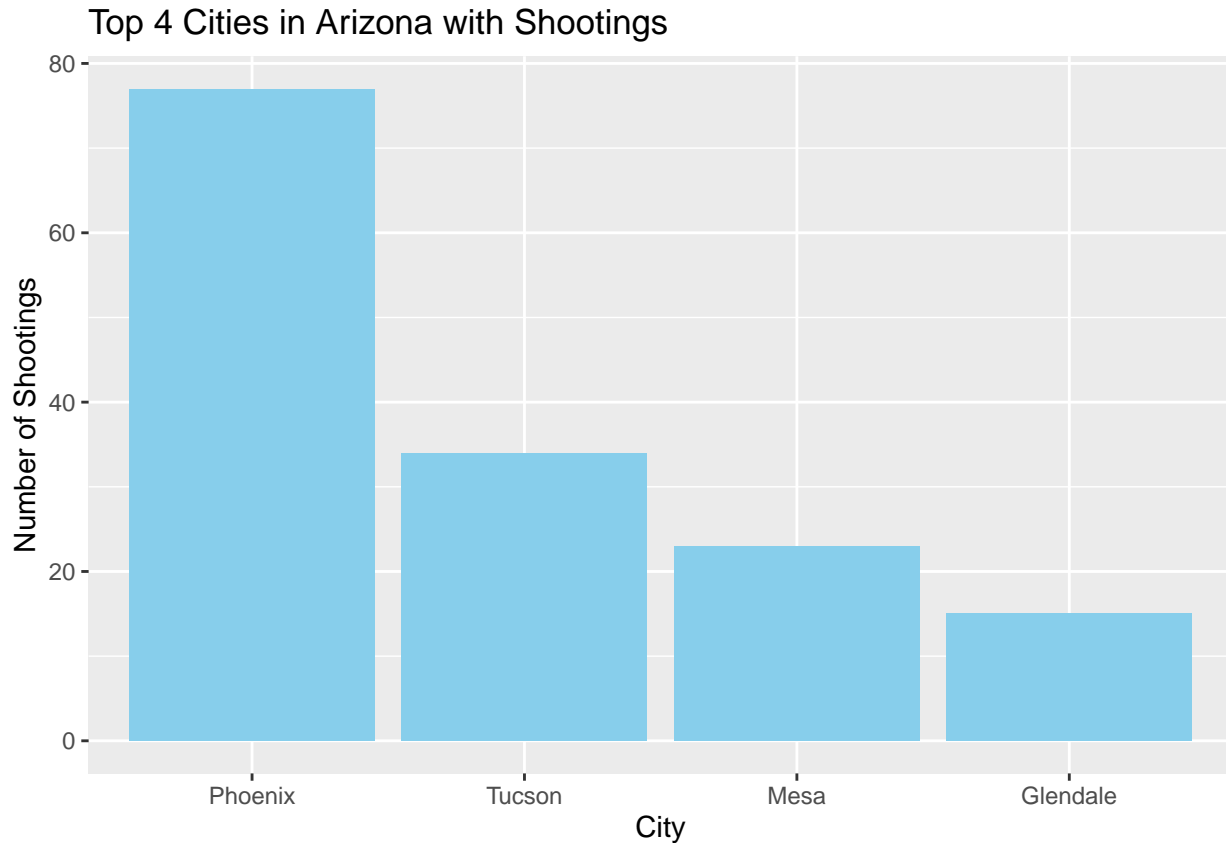
```
print(azizona_shootings_cleaned %>%
      count(city))
```

```
##           city  n
## 1  Apache Junction  3
## 2   Arizona City  1
## 3    Avondale  3
## 4   Beaver Dam  1
## 5     Bisbee  1
## 6    Buckeye  5
## 7   Casa Grande  3
## 8    Chandler  8
## 9   Clarkdale  1
## 10  Cottonwood  2
## 11   Douglas  1
```

```
## 12          Eagar 2
## 13      Ehrenberg 1
## 14      El Mirage 1
## 15      Flagstaff 7
## 16      Florence 1
## 17      Gilbert 4
## 18      Glendale 15
## 19          Globe 1
## 20      Golden Shores 1
## 21      Golden Valley 1
## 22      Goodyear 2
## 23      Guadalupe 1
## 24      Kearny 1
## 25      Kingman 7
## 26      Lake Havasu 1
## 27      Lake Havasu City 2
## 28      Marana 4
## 29      Maricopa 2
## 30          Mesa 23
## 31      Morenci 1
## 32      Nogales 1
## 33      Paradise Valley 1
## 34      Paulden 1
## 35      Payson 1
## 36      Peoria 3
## 37      Phoenix 77
## 38      Picture Rocks 1
## 39          Pima 1
## 40      Pinetop-Lakeside 1
## 41      Prescott 2
## 42      Quartzsite 1
## 43      Rio Rico 1
## 44      Safford 1
## 45      Salt River Reservation 1
## 46      San Manuel 1
## 47      Scottsdale 5
## 48      Sedona 2
## 49      Show Low 1
## 50      Sierra Vista 1
## 51      Sun City 2
## 52      Surprise 2
## 53      Tempe 10
## 54      Tolleson 1
## 55      Tonopah 1
## 56      Tucson 34
## 57      Tuscon 1
## 58      Wilcox 1
## 59      Wilhoit 1
## 60      Youngtown 1
## 61          Yuma 2
```

```
top_4 <- arizona_shootings_cleaned %>%
  count(city) %>%
  filter(n >= 15)
```

```
ggplot(top_4, aes(x = reorder(city, -n), y = n)) +
  geom_bar(stat = "identity", fill = "skyblue") +
  labs(x = "City", y = "Number of Shootings") +
  ggtitle("Top 4 Cities in Arizona with Shootings")
```

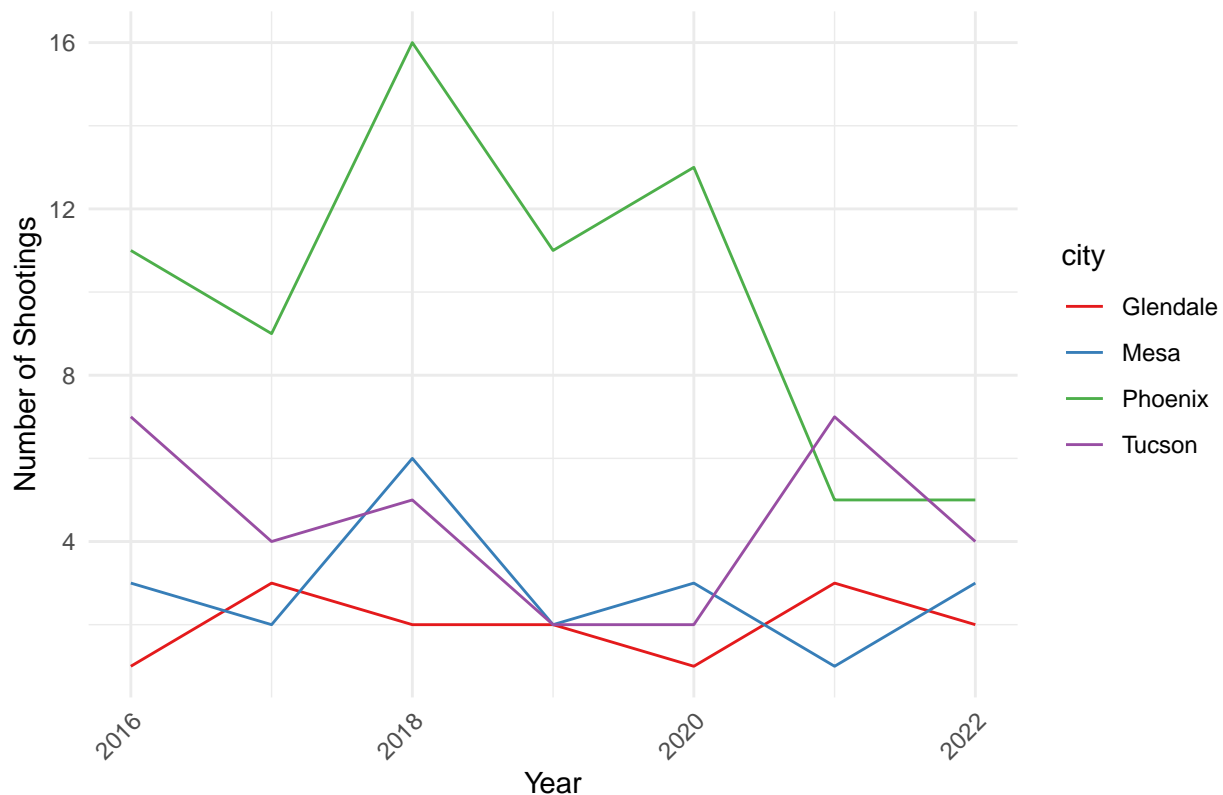


```
# Assuming your dataframe is called 'arizona_shootings', and you have a 'year' column representing the year
# Filter for shootings from 2016 to 2021 and select the top 4 cities
filtered_data <- arizona_shootings_cleaned %>%
  filter(year >= 2016 & year <= 2021 & city %in% c("Phoenix", "Tucson", "Mesa", "Glendale")) %>%
  group_by(city, year) %>%
  summarize(count = n())
```

## 'summarise()' has grouped output by 'city'. You can override using the  
## '.groups' argument.

```
# Create a line plot
ggplot(filtered_data, aes(x = year, y = count, color = city, group = city)) +
  geom_line() +
  labs(x = "Year", y = "Number of Shootings") +
  ggtitle("Trend of Shootings in Top 4 Cities in Arizona (2016-2021)") +
  scale_color_brewer(palette = "Set1") + # Choose a color palette
  theme_minimal() +
  theme(axis.text.x = element_text(angle = 45, hjust = 1)) # Rotate x-axis labels for better readability
```

### Trend of Shootings in Top 4 Cities in Arizona (2016–2021)

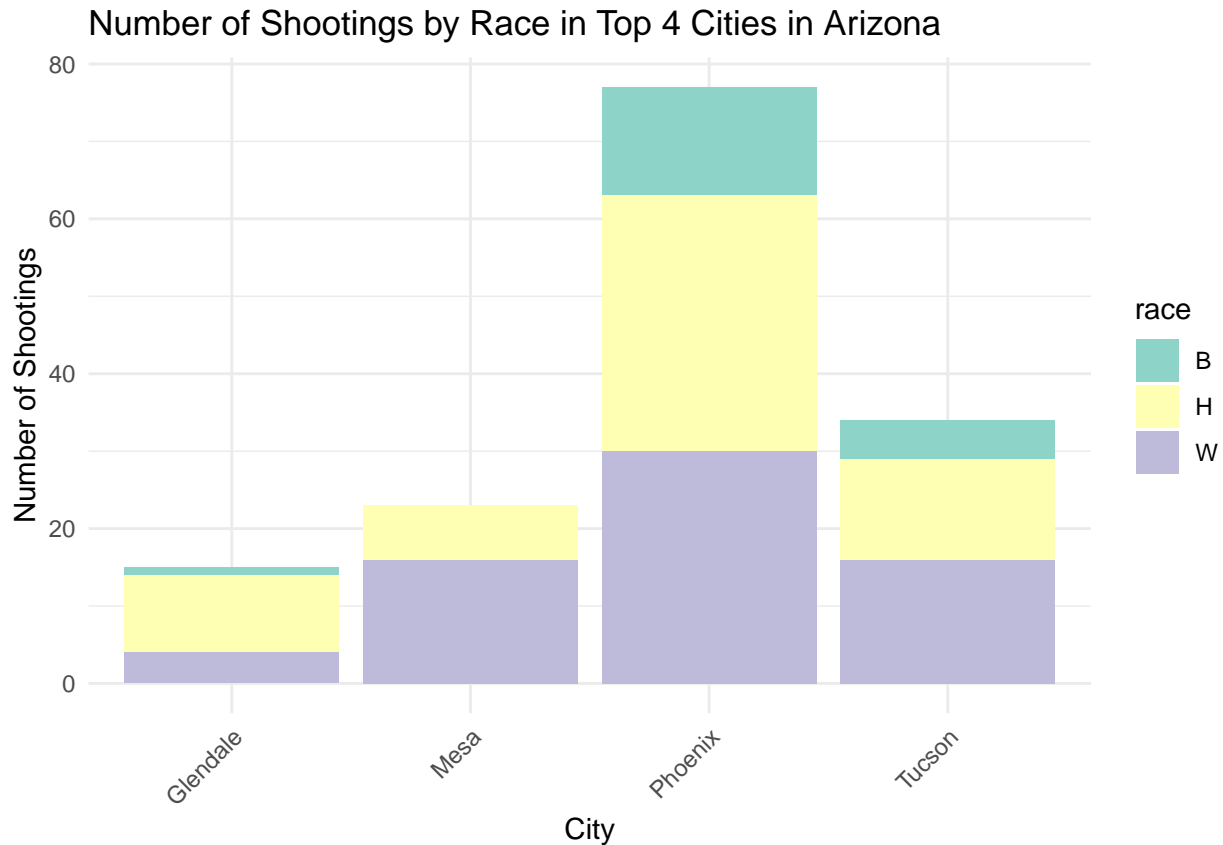


```
# Assuming your dataframe is called 'arizona_shootings' and you have a 'year' column representing the year
# Filter for shootings in the top 4 cities
filtered_data <- arizona_shootings_cleaned %>%
  filter(city %in% c("Phoenix", "Tucson", "Mesa", "Glendale")) %>%
  group_by(city, race) %>%
  summarize(count = n())
```

## 'summarise()' has grouped output by 'city'. You can override using the  
## '.groups' argument.

```
# Create a bar plot
ggplot(filtered_data, aes(x = city, y = count, fill = race)) +
  geom_bar(stat = "identity") +
  labs(x = "City", y = "Number of Shootings") +
  ggtitle("Number of Shootings by Race in Top 4 Cities in Arizona") +
  scale_fill_brewer(palette = "Set3") + # Choose a color palette
  theme_minimal() +
  theme(axis.text.x = element_text(angle = 45, hjust = 1)) # Rotate x-axis labels for better readability
```





```
phoenix_data <- arizona_shootings_cleaned %>%
  filter(year >= 2016 & year <= 2022 & city == "Phoenix") %>%
  group_by(year, race) %>%
  summarize(count = n())
```

## 'summarise()' has grouped output by 'year'. You can override using the  
## '.groups' argument.

```
median_income <- c("", "", "", "", "", "", "", "", "", 50757, 50757, 50757, 54907, 54907, 54907, 58308, 58308, 58308, "")
phoenix_data$median_income <- median_income
phoenix_data$shooting_count <- phoenix_data$count
phoenix_data$count <- NULL
```

```
arrests.df <- read.csv("~/Documents/Stats Seminar/arrests_adult-arrests-details_arrestdetail.csv", header = TRUE)
```

```
arrests_simple.df <- arrests.df[, c("YEAR", "SIMPLE_SUBJ_RE_GRP", "P2DRUG_CHARGES")]
arrests_simple.df <- arrests_simple.df %>%
  filter(SIMPLE_SUBJ_RE_GRP %in% c("Hispanic", "Black", "White") & YEAR != 2023 )
```

```
arrests_simple.df$P2DRUG_CHARGES <- arrests_simple.df$P2DRUG_CHARGES > 0
arrests_clean.df <- arrests_simple.df %>%
  group_by(YEAR, SIMPLE_SUBJ_RE_GRP, P2DRUG_CHARGES) %>%
  summarize(count = n())
```

## 'summarise()' has grouped output by 'YEAR', 'SIMPLE\_SUBJ\_RE\_GRP'. You can

```
## override using the '.groups' argument.
```

```
arrests_clean.df$race <- arrests_clean.df$SIMPLE_SUBJ_RE_GRP
arrests_clean.df$SIMPLE_SUBJ_RE_GRP <- NULL
arrests_clean.df
```

```
## # A tibble: 30 x 4
## # Groups:   YEAR [5]
##   YEAR P2DRUG_CHARGES count race
##   <int> <lgl>         <int> <chr>
## 1  2018 FALSE           7093 Black
## 2  2018 TRUE           1610 Black
## 3  2018 FALSE          11889 Hispanic
## 4  2018 TRUE           3110 Hispanic
## 5  2018 FALSE          13640 White
## 6  2018 TRUE           3318 White
## 7  2019 FALSE           7024 Black
## 8  2019 TRUE           1759 Black
## 9  2019 FALSE          12197 Hispanic
## 10 2019 TRUE           3562 Hispanic
## # i 20 more rows
```

```
total_counts <- arrests_simple.df %>%
  group_by(YEAR, SIMPLE_SUBJ_RE_GRP) %>%
  summarize(count = n())
```

```
## 'summarise()' has grouped output by 'YEAR'. You can override using the
## '.groups' argument.
```

```
total_counts$race <- total_counts$SIMPLE_SUBJ_RE_GRP
total_counts$SIMPLE_SUBJ_RE_GRP <- NULL

# Assuming your dataframe is called 'your_dataframe'
total_counts <- total_counts %>%
  mutate(race0 = case_when(
    race == "Black" ~ "B",
    race == "Hispanic" ~ "H",
    race == "White" ~ "W",
    TRUE ~ race # If none of the above conditions match, keep the original value
  ))
total_counts$race <- NULL
total_counts$race <- total_counts$race0
total_counts$race0 <- NULL
total_counts$arrest_count <- total_counts$count
total_counts$count <- NULL
total_counts$year <- total_counts$YEAR
total_counts$YEAR <- NULL
total_counts
```

```
## # A tibble: 15 x 3
##   race arrest_count year
##   <chr>         <int> <int>
```

```
## 1 B      8703 2018
## 2 H      14999 2018
## 3 W      16958 2018
## 4 B      8783 2019
## 5 H      15759 2019
## 6 W      16285 2019
## 7 B      7334 2020
## 8 H      13559 2020
## 9 W      12017 2020
## 10 B     6050 2021
## 11 H     10875 2021
## 12 W     9989 2021
## 13 B     5500 2022
## 14 H     9132 2022
## 15 W     7759 2022
```

```
phoenix_data
```

```
## # A tibble: 20 x 4
## # Groups:   year [7]
##   year race median_income shooting_count
##   <dbl> <chr> <chr>          <int>
## 1 2016 H     ""              7
## 2 2016 W     ""              4
## 3 2017 B     ""              2
## 4 2017 H     ""              4
## 5 2017 W     ""              3
## 6 2018 B     ""              3
## 7 2018 H     ""              6
## 8 2018 W     ""              7
## 9 2019 B     "50757"         2
## 10 2019 H     "50757"         6
## 11 2019 W     "50757"         3
## 12 2020 B     "54907"         4
## 13 2020 H     "54907"         5
## 14 2020 W     "54907"         4
## 15 2021 B     "58308"         1
## 16 2021 H     "58308"         3
## 17 2021 W     "58308"         1
## 18 2022 B     ""              2
## 19 2022 H     ""              1
## 20 2022 W     ""              2
```

```
# Perform a full join
combined_df <- full_join(total_counts, phoenix_data, by = c("year", "race"))

# Fill missing values with 0
combined_df[is.na(combined_df)] <- 0
combined_df
```

```
## # A tibble: 20 x 5
##   race arrest_count year median_income shooting_count
##   <chr>          <int> <dbl> <chr>          <int>
```

```
## 1 B      8703 2018 ""      3
## 2 H     14999 2018 ""      6
## 3 W     16958 2018 ""      7
## 4 B      8783 2019 "50757"  2
## 5 H     15759 2019 "50757"  6
## 6 W     16285 2019 "50757"  3
## 7 B      7334 2020 "54907"  4
## 8 H     13559 2020 "54907"  5
## 9 W     12017 2020 "54907"  4
## 10 B      6050 2021 "58308"  1
## 11 H     10875 2021 "58308"  3
## 12 W      9989 2021 "58308"  1
## 13 B      5500 2022 ""      2
## 14 H      9132 2022 ""      1
## 15 W      7759 2022 ""      2
## 16 H         0 2016 ""      7
## 17 W         0 2016 ""      4
## 18 B         0 2017 ""      2
## 19 H         0 2017 ""      4
## 20 W         0 2017 ""      3
```

```
#rearrange columns with year as first column and in order
combined_df <- combined_df %>%
  select(year, everything())
combined_df <- combined_df %>%
  arrange(year)
combined_df_clean <- combined_df %>%
  filter(year >= 2018 & year <= 2022)
combined_df_clean
```

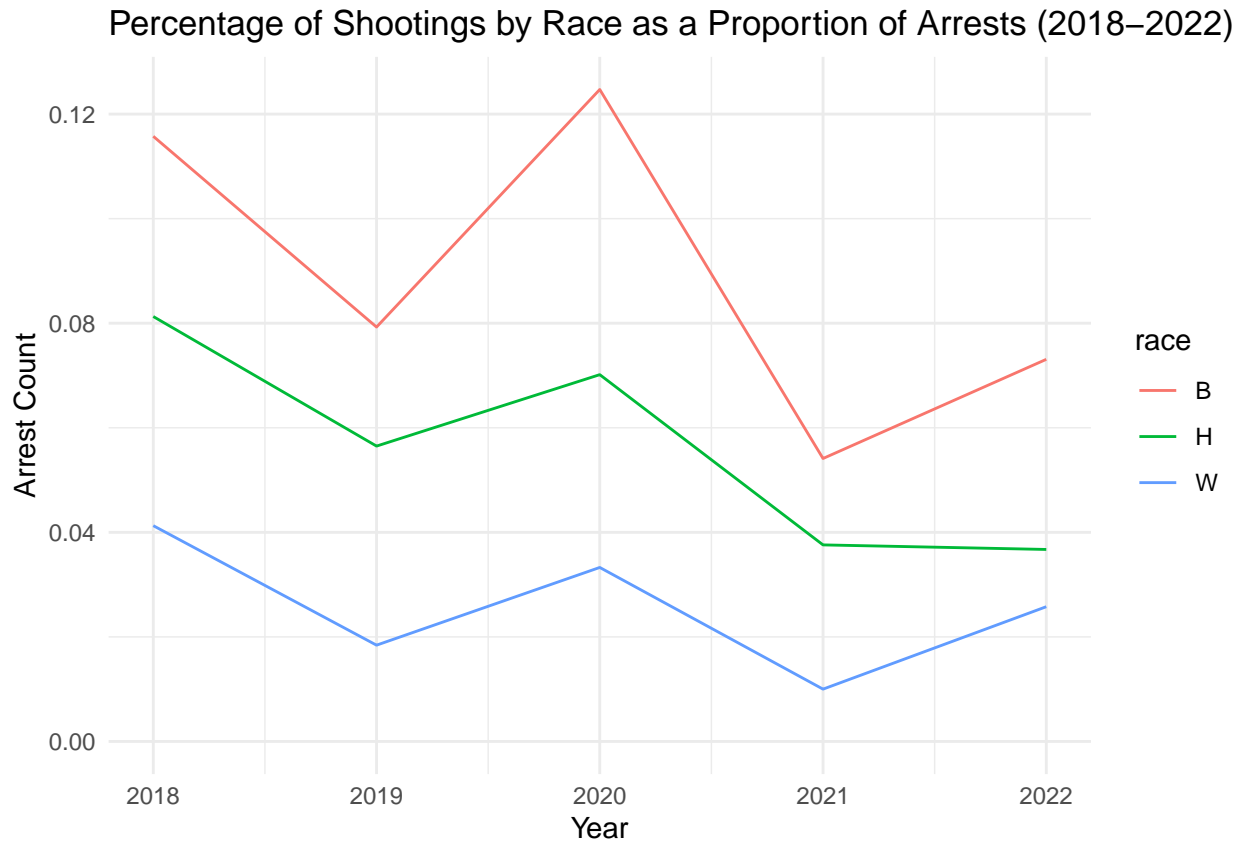
```
## # A tibble: 15 x 5
##   year race arrest_count median_income shooting_count
##   <dbl> <chr>      <int> <chr>      <int>
## 1 2018 B      8703 ""      3
## 2 2018 H     14999 ""      6
## 3 2018 W     16958 ""      7
## 4 2019 B      8783 "50757"  2
## 5 2019 H     15759 "50757"  6
## 6 2019 W     16285 "50757"  3
## 7 2020 B      7334 "54907"  4
## 8 2020 H     13559 "54907"  5
## 9 2020 W     12017 "54907"  4
## 10 2021 B      6050 "58308"  1
## 11 2021 H     10875 "58308"  3
## 12 2021 W      9989 "58308"  1
## 13 2022 B      5500 ""      2
## 14 2022 H      9132 ""      1
## 15 2022 W      7759 ""      2
```

```
ggplot(data = combined_df_clean, aes(x = year, y = (shooting_count/arrest_count) * 100)) +
  geom_line(stat = "identity", position = "stack") +
  aes(colour = race) +
  labs(
```

```

title = "Percentage of Shootings by Race as a Proportion of Arrests (2018-2022)",
x = "Year",
y = "Arrest Count"
) +
scale_fill_manual(values = c("red", "blue", "green")) +
theme_minimal()

```

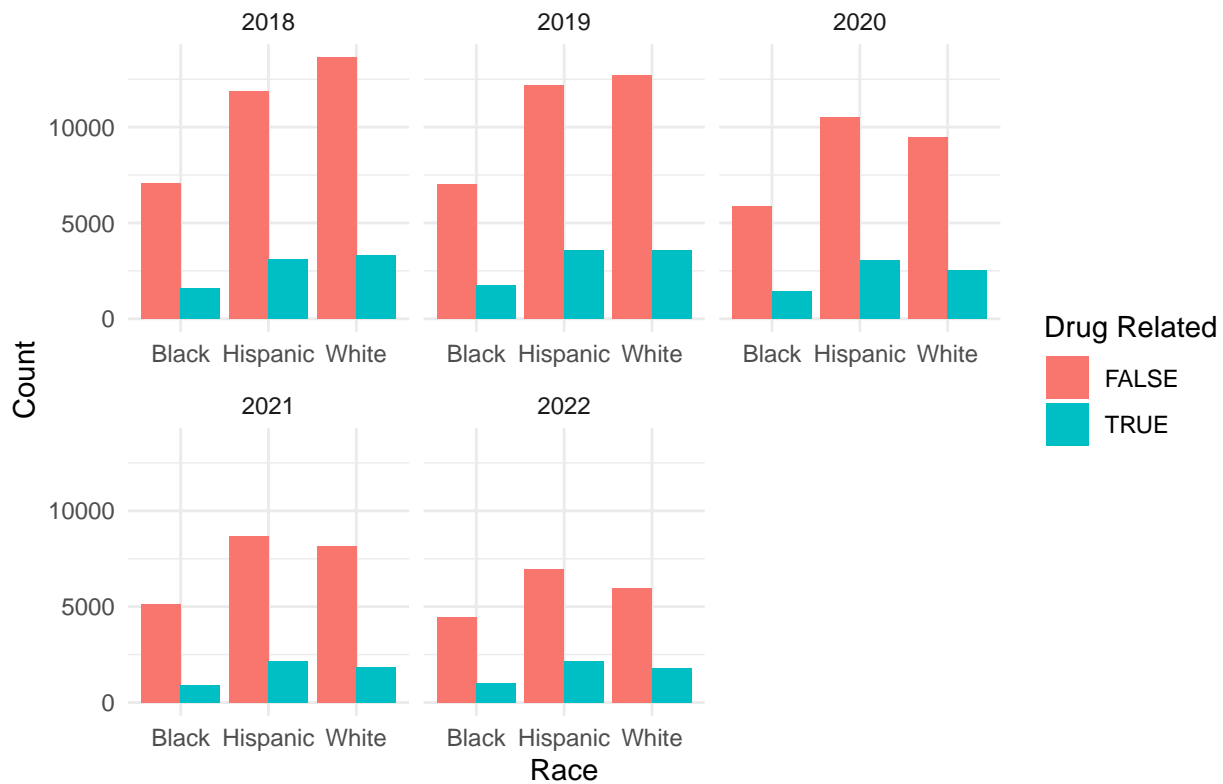


```

# Create a bar plot
ggplot(arrests_clean.df, aes(x = race, y = count, fill = P2DRUG_CHARGES)) +
  geom_bar(stat = "identity", position = "dodge") +
  labs(x = "Race", y = "Count") +
  ggtitle("Counts of Charges by Race (Drug-Related vs. Non-Drug-Related)") +
  scale_fill_discrete(name = "Drug Related") +
  theme_minimal() +
  facet_wrap(~YEAR, scales = "free_x")

```

Counts of Charges by Race (Drug-Related vs. Non-Drug-Related)



```
# Create the top_5 dataframe
top_5 <- data.frame(
  cities = c("Los Angeles", "Miami", "Atlanta", "Houston", "Phoenix"),
  per_capita_income_ranked = c(19, 21, 77, 62, 126),
  state_drug_crime_rating = c("D-", "D+", "D", "D-", "F"),
  city_drug_crime_rating = c("D-", "D+", "F", "C-", "D+"),
  city_overall_crime_rating = c("D-", "F", "C-", "C-", "F"),
  state_overall_crime_rating = c("D-", "C", "C+", "C", "F")
)

# Print the dataframe
print(top_5)
```

```
##      cities per_capita_income_ranked state_drug_crime_rating
## 1 Los Angeles                19          D-
## 2 Miami                    21          D+
## 3 Atlanta                   77           D
## 4 Houston                   62          D-
## 5 Phoenix                   126           F
## city_drug_crime_rating city_overall_crime_rating state_overall_crime_rating
## 1 D-          D-          D-          D-
## 2 D+          F           C           C
## 3 F           C-          C+          C+
## 4 C-          C-          C           C
## 5 D+          F           F           F
```

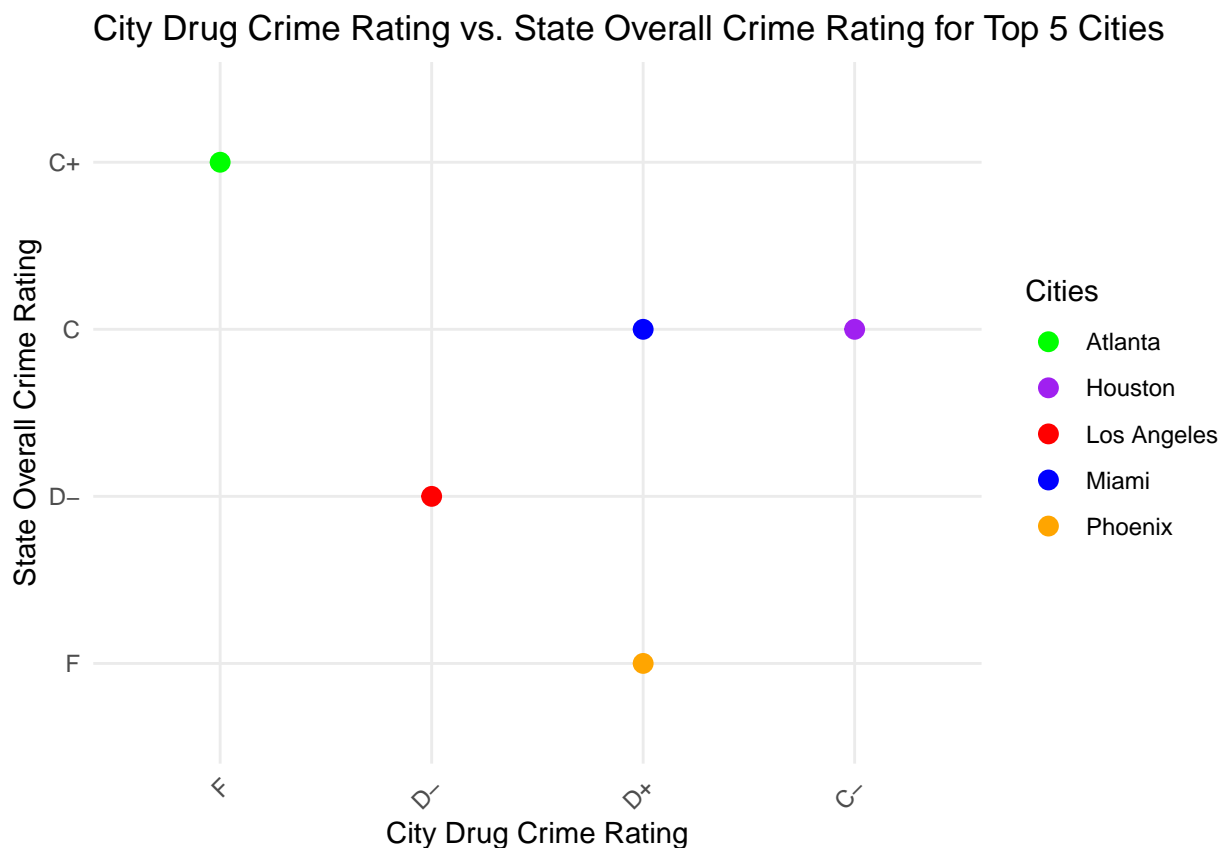
```

# Assuming your dataframe is named top_5
# Define the levels for city_drug_crime_rating in reverse order
possible_grades <- c("F", "D-", "D", "D+", "C-", "C", "C+", "B-", "B", "B+", "A-", "A", "A+")
top_5$city_drug_crime_rating <- factor(top_5$city_drug_crime_rating, levels = rev(possible_grades))

# Define the levels for state_overall_crime_rating in reverse order
state_overall_grades <- c("F", "D-", "D", "D+", "C-", "C", "C+", "B-", "B", "B+", "A-", "A", "A+")
top_5$state_overall_crime_rating <- factor(top_5$state_overall_crime_rating, levels = state_overall_grades)

# Create a scatterplot
ggplot(top_5, aes(x = reorder(city_drug_crime_rating, -as.numeric(city_drug_crime_rating)), y = state_overall_crime_rating)) +
  geom_point(size = 3) +
  labs(x = "City Drug Crime Rating", y = "State Overall Crime Rating", color = "Cities") +
  scale_color_manual(values = c("Los Angeles" = "red", "Miami" = "blue", "Atlanta" = "green", "Houston" = "purple", "Phoenix" = "orange")) +
  theme_minimal() +
  theme(axis.text.x = element_text(angle = 45, hjust = 1)) +
  ggtitle("City Drug Crime Rating vs. State Overall Crime Rating for Top 5 Cities")

```



```

# Assuming your dataframe is named top_5
# Define the levels for city_drug_crime_rating in reverse order
possible_grades <- c("F", "D-", "D", "D+", "C-", "C", "C+", "B-", "B", "B+", "A-", "A", "A+")
top_5$city_drug_crime_rating <- factor(top_5$city_drug_crime_rating, levels = rev(possible_grades))

# Define the levels for state_overall_crime_rating in reverse order
state_overall_grades <- c("F", "D-", "D", "D+", "C-", "C", "C+", "B-", "B", "B+", "A-", "A", "A+")
top_5$state_overall_crime_rating <- factor(top_5$state_overall_crime_rating, levels = state_overall_grades)

```

```
# Create a scatterplot
ggplot(top_5, aes(x = reorder(city_drug_crime_rating, -as.numeric(city_drug_crime_rating)), y = per_capita_income_ranking)) +
  geom_point(size = 3) +
  labs(x = "City Drug Crime Rating", y = "Per Capita Income Ranking", color = "Cities") +
  scale_color_manual(values = c("Los Angeles" = "red", "Miami" = "blue", "Atlanta" = "green", "Houston" = "purple", "Phoenix" = "orange")) +
  theme_minimal() +
  theme(axis.text.x = element_text(angle = 45, hjust = 1)) +
  ggtitle("City Drug Crime Rating vs. Per Capita Income Ranking")
```

