

US Counties - Income, Education Analysis

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Loading packages & data

```
setwd("~/Data Science/Datasets")
data <- read.csv("~/Data Science/Datasets/uscountydata.csv")
pop_data <- read.csv("~/Data Science/Datasets/us_pop_by_state.csv")
library(tidyverse)
```

```
## Warning: package 'tidyverse' was built under R version 4.1.3
```

```
## -- Attaching packages ----- tidyverse 1.3.2 --
## v ggplot2 3.4.0      v purrr   1.0.1
## v tibble  3.1.8      v dplyr  1.0.10
## v tidyr   1.2.1      v stringr 1.5.0
## v readr   2.1.3      v forcats 0.5.2
```

```
## Warning: package 'ggplot2' was built under R version 4.1.3
```

```
## Warning: package 'tibble' was built under R version 4.1.3
```

```
## Warning: package 'tidyr' was built under R version 4.1.3
```

```
## Warning: package 'readr' was built under R version 4.1.3
```

```
## Warning: package 'purrr' was built under R version 4.1.3
```

```
## Warning: package 'dplyr' was built under R version 4.1.3
```

```
## Warning: package 'stringr' was built under R version 4.1.3
```

```
## Warning: package 'forcats' was built under R version 4.1.3
```

```
## -- Conflicts ----- tidyverse_conflicts() --
## x dplyr::filter() masks stats::filter()
## x dplyr::lag()    masks stats::lag()
```

Previewing Data in different ways

```
head(data)
```

```
##   county_FIPS state      county per_capita_personal_income_2019
## 1      51013   VA  Arlington, VA                97629
## 2      35028   NM  Los Alamos, NM                72366
## 3       8013   CO    Boulder, CO                79698
## 4      24027   MD    Howard, MD                78013
## 5      36061   NY   New York, NY               173525
## 6      51107   VA    Loudoun, VA                82681
##   per_capita_personal_income_2020 per_capita_personal_income_2021
## 1                      100687                      107603
## 2                      75949                      81306
## 3                      83173                      89593
## 4                      82041                      86380
## 5                     175327                     195543
## 6                      85568                      90254
##   associate_degree_numbers_2016_2020 bachelor_degree_numbers_2016_2020
## 1                      19573                      132394
## 2                      2766                      9098
## 3                     45834                     135876
## 4                     42538                     136792
## 5                     167960                     777483
## 6                      52006                     160853
##   associate_degree_percentage_2016_2020 bachelor_degree_percentage_2015_2019
## 1                      11.21                      75.84
## 2                      20.54                      67.56
## 3                      21.24                      62.97
## 4                      19.49                      62.67
## 5                      13.43                      62.18
## 6                      19.92                      61.62
```

```
str(data)
```

```
## 'data.frame': 3006 obs. of 10 variables:
## $ county_FIPS : int 51013 35028 8013 24027 36061 51107 8097 371
35 47187 6041 ...
## $ state : chr "VA" "NM" "CO" "MD" ...
## $ county : chr "Arlington, VA" "Los Alamos, NM" "Boulder,
CO" "Howard, MD" ...
## $ per_capita_personal_income_2019 : int 97629 72366 79698 78013 173525 82681 183241
66970 95806 139891 ...
## $ per_capita_personal_income_2020 : int 100687 75949 83173 82041 175327 85568 18759
5 69593 99408 148419 ...
## $ per_capita_personal_income_2021 : int 107603 81306 89593 86380 195543 90254 19893
9 74994 107698 164118 ...
## $ associate_degree_numbers_2016_2020 : int 19573 2766 45834 42538 167960 52006 2932 17
907 32995 44228 ...
## $ bachelor_degree_numbers_2016_2020 : int 132394 9098 135876 136792 777483 160853 845
9 55545 91780 114604 ...
## $ associate_degree_percentage_2016_2020: num 11.2 20.5 21.2 19.5 13.4 ...
## $ bachelor_degree_percentage_2015_2019 : num 75.8 67.6 63 62.7 62.2 ...
```

View(data)

Gaining simple information about the data set

```
n_distinct(data$state)
```

```
## [1] 49
```

```
n_distinct(data$county)
```

```
## [1] 3006
```

Simple dplyr manipulation

```
average_income_data <-
  data %>%
  transmute(state, county, av_income = rowMeans(select(data, starts_with('per')))) %>%
  arrange(desc(av_income)) %>%
  filter(state %in% c('FL', 'CA')) %>%
  slice_max(av_income, n=3)
```

Group & Summarise data

```
average_state_income <-
  data %>%
  select(state, per_capita_personal_income_2021) %>%
  group_by(state) %>%
  summarize(mean_income = mean(per_capita_personal_income_2021)) %>%
  arrange(desc(mean_income))
```

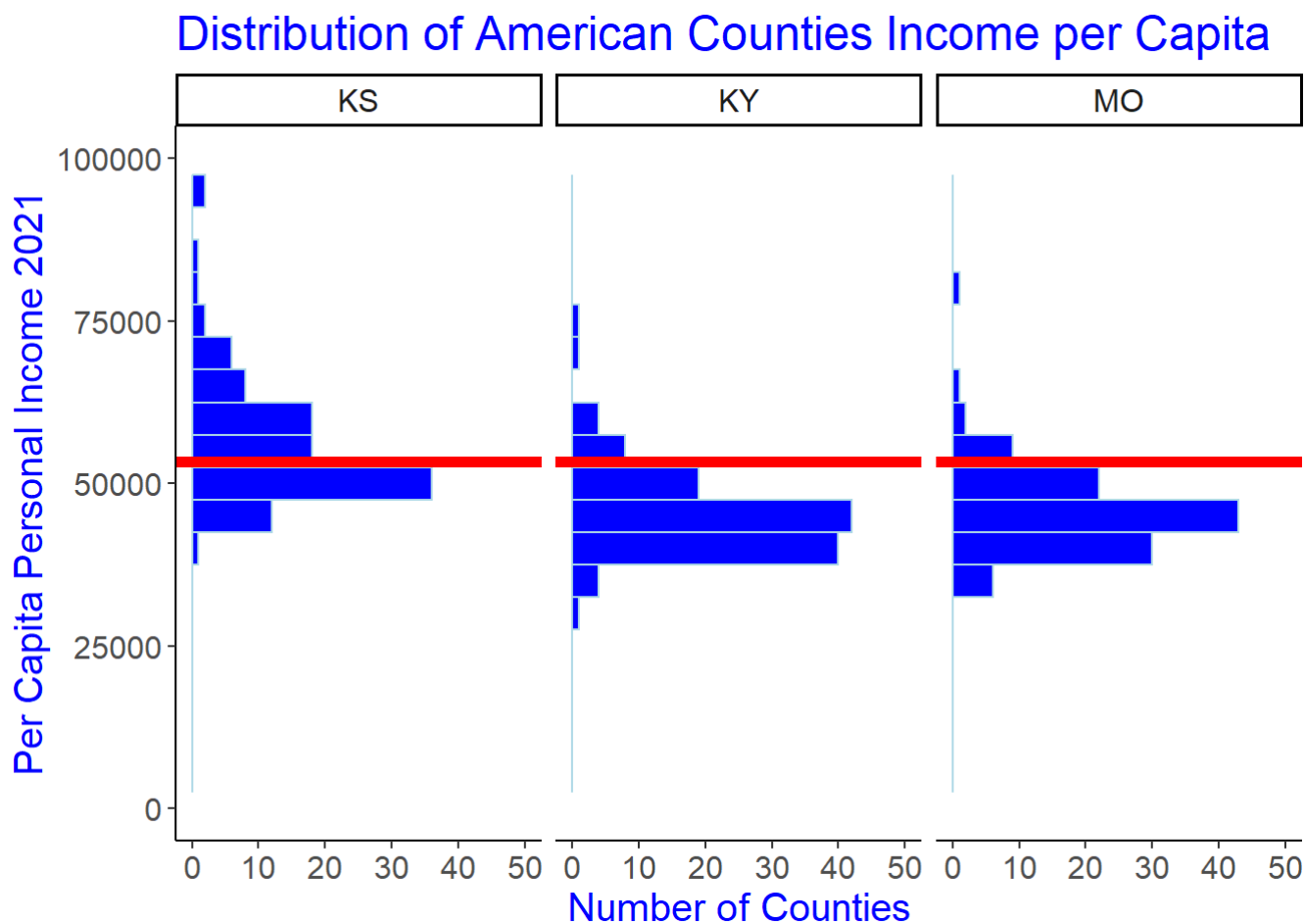
Country average statistic

```
US_av_inc <- data %>%
  pull(per_capita_personal_income_2021) %>%
  mean()
```

ggplot example

```
Three_States <- data %>% filter(state %in% c('KY', 'MO', 'KS'))

# ggplot2
ggplot(Three_States, aes(per_capita_personal_income_2021))+
  geom_histogram(binwidth = 5000, color = 'lightblue', fill = 'blue')+
  xlim(0,100000)+
  ylim(0,50)+
  geom_vline(xintercept = US_av_inc, color = 'red', size = 2)+
  labs(x='Per Capita Personal Income 2021', y = 'Number of Counties', title ='Distribution of
American Counties Income per Capita')+
  annotate('text',x=30,y=60000,label='Country_Average')+
  theme_classic()+
  coord_flip()+
  facet_wrap(~state)+
  theme(text = element_text(family = 'arial', size = 15, color = 'blue'))
```



Second ggplot2 example

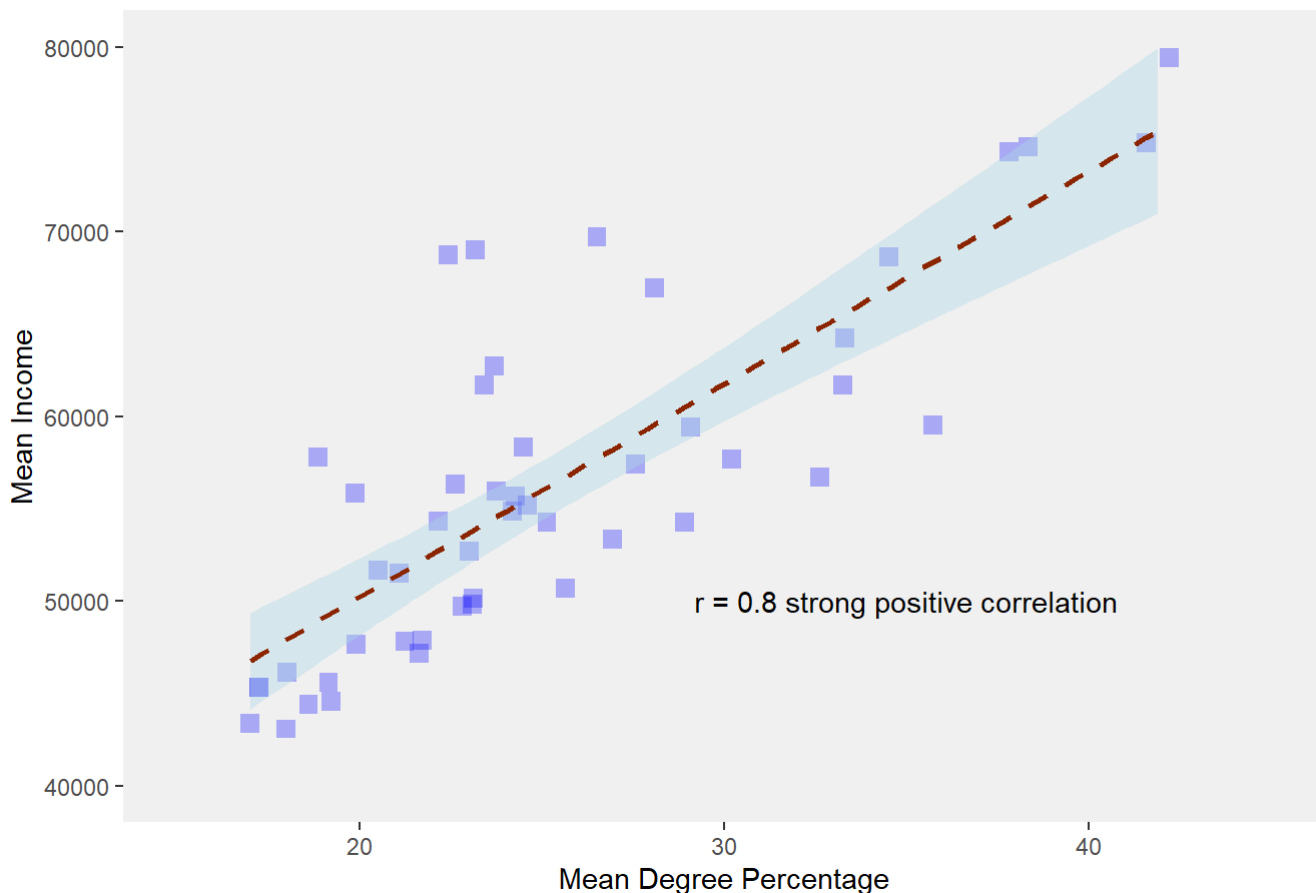
```
degree_vs_income <- data %>%
  select(state, per_capita_personal_income_2021, bachelor_degree_percentage_2015_2019) %>%

  group_by(state) %>%
  summarize(mean_deg_perc = mean(bachelor_degree_percentage_2015_2019),
            mean_income = mean(per_capita_personal_income_2021))

ggplot(degree_vs_income, aes(x=mean_deg_perc, y=mean_income))+
  geom_jitter(size=3,alpha=0.3, color='blue', shape = 15)+
  geom_smooth(method=lm, linetype = 'dashed', color = 'orangered4', fill = 'lightblue')+
  xlim(15,45)+
  ylim(40000,80000)+
  annotate('text',x=35,y=50000,label='r = 0.8 strong positive correlation')+
  labs(x='Mean Degree Percentage',y='Mean Income', title='American State income dependency on
education level')+
  theme(panel.background = element_rect(fill = 'grey94'),
        panel.grid = element_blank()
        )
```

```
## `geom_smooth()` using formula = 'y ~ x'
```

American State income dependency on education level



```
cor(degree_vs_income$mean_deg_perc, degree_vs_income$mean_income)
```

```
## [1] 0.8024959
```

Joining data

```
joined_pop_income <- data %>%  
  left_join(pop_data, by = c('state' = 'state_code')) %>%  
  select(-percent_of_total, -rank)
```

Finding outliers

```
IQR <- quantile(joined_pop_income$per_capita_personal_income_2021, 0.75) - quantile(joined_pop_income$per_capita_personal_income_2021, 0.25)  
Lower_threshold <- quantile(joined_pop_income$per_capita_personal_income_2021, 0.25) - 1.5 * IQR  
Upper_threshold <- quantile(joined_pop_income$per_capita_personal_income_2021, 0.75) + 1.5 * IQR  
  
outliers <- joined_pop_income %>%  
  filter(per_capita_personal_income_2021 < Lower_threshold | per_capita_personal_income_2021 > Upper_threshold) %>%  
  select(county, state.y, per_capita_personal_income_2021, X2020_census) %>%  
  arrange(desc(per_capita_personal_income_2021))  
  
outliers
```

##	county	state.y	per_capita_personal_income_2021
## 1	Teton, WY	Wyoming	318297
## 2	Pitkin, CO	Colorado	198939
## 3	New York, NY	New York	195543
## 4	Summit, UT	Utah	183972
## 5	Marin, CA	California	164118
## 6	San Francisco, CA	California	160749
## 7	San Mateo, CA	California	160485
## 8	Bristol Bay Borough, AK	Alaska	155155
## 9	Santa Clara, CA	California	138724
## 10	Union, SD	South Dakota	134811
## 11	Blaine, ID	Idaho	134722
## 12	Fairfield, CT	Connecticut	127391
## 13	Midland, TX	Texas	126738
## 14	Glasscock, TX	Texas	124963
## 15	Sherman, TX	Texas	122945
## 16	Westchester, NY	New York	119705
## 17	Collier, FL	Florida	117984
## 18	Loving, TX	Texas	115158
## 19	Somerset, NJ	New Jersey	113975
## 20	Goochland, VA	Virginia	110781
## 21	McMullen, TX	Texas	110273
## 22	San Miguel, CO	Colorado	109613
## 23	Monroe, FL	Florida	109136
## 24	King, WA	Washington	108212
## 25	Denali Borough, AK	Alaska	107916
## 26	Morris, NJ	New Jersey	107767
## 27	Williamson, TN	Tennessee	107698
## 28	Arlington, VA	Virginia	107603
## 29	Sully, SD	South Dakota	106241
## 30	Norfolk, MA	Massachusetts	104059
## 31	Lipscomb, TX	Texas	102569
## 32	Martin, FL	Florida	102273
## 33	Fulton, GA	Georgia	102074
## 34	Nantucket, MA	Massachusetts	101902
## 35	Borden, TX	Texas	101470
## 36	Palm Beach, FL	Florida	100627
## 37	Hunterdon, NJ	New Jersey	100288
## 38	Alameda, CA	California	99746
## 39	Nassau, NY	New York	99597
## 40	Contra Costa, CA	California	99312
## 41	Denver, CO	Colorado	99133
## 42	Suffolk, MA	Massachusetts	98644
## 43	Middlesex, MA	Massachusetts	98523
## 44	Routt, CO	Colorado	98371
## 45	Bergen, NJ	New Jersey	97343
## 46	Eagle, CO	Colorado	97255
## 47	Kendall, TX	Texas	96628
## 48	Chester, PA	Pennsylvania	95483
## 49	Indian River, FL	Florida	95109
## 50	Stanton, KS	Kansas	94353
## 51	Dundy, NE	Nebraska	93788
## 52	Greeley, KS	Kansas	93667
## 53	Shackelford, TX	Texas	93075
## 54	Montgomery, MD	Maryland	92740

## 55	Ozaukee, WI	Wisconsin	92719
## 56	Billings, ND	North Dakota	92571
## 57	Monmouth, NJ	New Jersey	92119
## 58	King, TX	Texas	91930
## 59	Benton, AR	Arkansas	91687
## 60	Napa, CA	California	90608
## 61	Loudoun, VA	Virginia	90254
## 62	Hayes, NE	Nebraska	90144
## 63	New Kent, VA	Virginia	90086
## 64	Boulder, CO	Colorado	89593
## 65	Lake, IL	Illinois	89025
## 66	Cavalier, ND	North Dakota	88700
## 67	Montgomery, PA	Pennsylvania	88671
## 68	Rockingham, NH	New Hampshire	88303
## 69	Haines Borough, AK	Alaska	88152
## 70	Perkins, NE	Nebraska	87847
## 71	Douglas, CO	Colorado	87841
## 72	Bristol, RI	Rhode Island	87810
## 73	Douglas, NV	Nevada	87214
## 74	San Juan, WA	Washington	87038
## 75	Potter, SD	South Dakota	87031
## 76	Summit, CO	Colorado	86390
## 77	Howard, MD	Maryland	86380
## 78	Dukes, MA	Massachusetts	86341
## 79	Jones, SD	South Dakota	86307
## 80	Steele, ND	North Dakota	86280
## 81	Hamilton, IN	Indiana	85814
## 82	Santa Cruz, CA	California	85554
## 83	Hennepin, MN	Minnesota	85505
## 84	DuPage, IL	Illinois	85498
## 85	Boone, IN	Indiana	84985
## 86	Talbot, MD	Maryland	84695
## 87	Johnson, KS	Kansas	84535
## 88	Martin, TX	Texas	84121
## 89	Newport, RI	Rhode Island	84054
## 90	Cumberland, IL	Illinois	84035
## 91	Wheeler, NE	Nebraska	83722
## 92	Delaware, OH	Ohio	83603
## 93	Barnstable, MA	Massachusetts	83191
## 94	Bucks, PA	Pennsylvania	83011
## 95	Wells, ND	North Dakota	82108
## 96	Davidson, TN	Tennessee	82087
## 97	Waukesha, WI	Wisconsin	82032
## 98	Castro, TX	Texas	81900
## 99	St. Louis, MO	Missouri	81829
## 100	Travis, TX	Texas	81708
## 101	Suffolk, NY	New York	81309
## 102	Los Alamos, NM	New Mexico	81306
## 103	Orange, CA	California	81034
## 104	Sonoma, CA	California	81006
## 105	Dunn, ND	North Dakota	80976
## 106	Oakland, MI	Michigan	80962
## 107	McIntosh, ND	North Dakota	80675
## 108	Towner, ND	North Dakota	80665
## 109	Carver, MN	Minnesota	80562
## 110	Rock, NE	Nebraska	80527

## 111	Rawlins, KS	Kansas	80389
## 112	Slope, ND	North Dakota	80270
## 113	Faulk, SD	South Dakota	80152
## 114	St. Johns, FL	Florida	80062
## 115	Plymouth, MA	Massachusetts	79817
## 116	Campbell, SD	South Dakota	79723
## 117	Irion, TX	Texas	78988
##	X2020_census		
## 1	576851		
## 2	5773714		
## 3	20201249		
## 4	3205958		
## 5	39538223		
## 6	39538223		
## 7	39538223		
## 8	733391		
## 9	39538223		
## 10	886667		
## 11	1839106		
## 12	3605944		
## 13	29145505		
## 14	29145505		
## 15	29145505		
## 16	20201249		
## 17	21538187		
## 18	29145505		
## 19	9288994		
## 20	8631393		
## 21	29145505		
## 22	5773714		
## 23	21538187		
## 24	7705281		
## 25	733391		
## 26	9288994		
## 27	6910840		
## 28	8631393		
## 29	886667		
## 30	7029917		
## 31	29145505		
## 32	21538187		
## 33	10711908		
## 34	7029917		
## 35	29145505		
## 36	21538187		
## 37	9288994		
## 38	39538223		
## 39	20201249		
## 40	39538223		
## 41	5773714		
## 42	7029917		
## 43	7029917		
## 44	5773714		
## 45	9288994		
## 46	5773714		
## 47	29145505		
## 48	13002700		

## 49	21538187
## 50	2937880
## 51	1961504
## 52	2937880
## 53	29145505
## 54	6177224
## 55	5893718
## 56	779094
## 57	9288994
## 58	29145505
## 59	3011524
## 60	39538223
## 61	8631393
## 62	1961504
## 63	8631393
## 64	5773714
## 65	12801989
## 66	779094
## 67	13002700
## 68	1377529
## 69	733391
## 70	1961504
## 71	5773714
## 72	1097379
## 73	3104614
## 74	7705281
## 75	886667
## 76	5773714
## 77	6177224
## 78	7029917
## 79	886667
## 80	779094
## 81	6785528
## 82	39538223
## 83	5706494
## 84	12801989
## 85	6785528
## 86	6177224
## 87	2937880
## 88	29145505
## 89	1097379
## 90	12801989
## 91	1961504
## 92	11799448
## 93	7029917
## 94	13002700
## 95	779094
## 96	6910840
## 97	5893718
## 98	29145505
## 99	6154913
## 100	29145505
## 101	20201249
## 102	2117522
## 103	39538223
## 104	39538223

## 105	779094
## 106	10077331
## 107	779094
## 108	779094
## 109	5706494
## 110	1961504
## 111	2937880
## 112	779094
## 113	886667
## 114	21538187
## 115	7029917
## 116	886667
## 117	29145505