

Final Project

March 3, 2025

1 Data Exploration and Analysis - Final Project

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Class: Data Exploration and Analysis

Prof: Metzger

1.0.1 Part 1 - Importing data and Libraries

```
[47]: import pandas as pd
import matplotlib.pyplot as plt
import numpy as np
```

```
[48]: data = pd.read_excel('finaldata.xlsx')
```

```
[49]: data
```

[49]:	state	abv.	county_fips	county	m_housing	m_food	\
0		AL	1001	Autauga County	944.584583	338.420067	
1		AL	1003	Baldwin County	1051.000000	382.602658	
2		AL	1005	Barbour County	652.000000	325.259275	
3		AL	1007	Bibb County	691.599283	309.278317	
4		AL	1009	Blount County	640.721025	304.578042	
...	
3138		WY	56037	Sweetwater County	896.000000	338.420067	
3139		WY	56039	Teton County	1317.000000	475.668175	
3140		WY	56041	Uinta County	684.000000	326.199342	
3141		WY	56043	Washakie County	684.000000	354.401000	
3142		WY	56045	Weston County	801.000000	350.640792	
	m_transport		m_health	m_other	m_tax	m_total	...
0	1124.548500		525.053017	431.675333	667.752767	4032.034167	...
1	1128.868000		495.053017	482.345008	708.367350	4248.236000	...
2	1072.752583		485.199867	328.805300	548.314408	3412.331417	...
3	1190.644917		501.053017	336.751833	587.831750	3617.159167	...

4	1165.808083	501.053017	318.052067	564.144975	3494.357083	...
...
3138	1218.016000	674.583208	415.328733	564.086425	4106.434583	...
3139	1252.413583	674.583208	603.154992	774.953450	5097.773417	...
3140	1191.231917	674.583208	339.888183	499.248617	3715.151333	...
3141	1140.456833	674.583208	349.376833	496.646892	3699.464833	...
3142	1298.514667	674.583208	387.477092	558.107058	4070.322917	...

	st_cost_rank	st_med_aff_rank	st_income_rank	min_wage	m_income	\
0	8	48.0	11.0	7.25	1256.666667	
1	2	59.0	4.0	7.25	1256.666667	
2	63	19.0	55.0	7.25	1256.666667	
3	18	20.0	39.0	7.25	1256.666667	
4	24	46.0	18.0	7.25	1256.666667	
...	
3138	12	19.0	6.0	7.25	1256.666667	
3139	1	13.0	1.0	7.25	1256.666667	
3140	19	20.0	10.0	7.25	1256.666667	
3141	18	8.0	18.0	7.25	1256.666667	
3142	7	9.0	12.0	7.25	1256.666667	

	a_min_income	m_ratio	a_min_ratio	median_household_income	\
0	15080.0	0.311671	0.311671	70148	
1	15080.0	0.295809	0.295809	71704	
2	15080.0	0.368272	0.368272	41151	
3	15080.0	0.347418	0.347418	54309	
4	15080.0	0.359627	0.359627	60553	
...	
3138	15080.0	0.306024	0.306024	75779	
3139	15080.0	0.246513	0.246513	127677	
3140	15080.0	0.338255	0.338255	73072	
3141	15080.0	0.339689	0.339689	60699	
3142	15080.0	0.308739	0.308739	67677	

	median_ratio
0	1.449806
1	1.406545
2	1.004958
3	1.251189
4	1.444066
...	...
3138	1.537810
3139	2.087137
3140	1.639054
3141	1.367292
3142	1.385578

[3143 rows x 29 columns]

1.0.2 Part 2 - Calculating Descriptive Statistics and Outliers

```
[112]: data.mean(numeric_only=True)
```

```
[112]: county_fips          30374.576519
      m_housing            803.242426
      m_food              337.483889
      m_transport         1144.573051
      m_health            490.475464
      m_other             383.804850
      m_tax               589.548597
      m_total             3749.128282
      a_housing           9638.909116
      a_food              4049.806665
      a_transport        13734.876612
      a_health            5885.705567
      a_other             4605.658203
      a_taxes             7074.583165
      a_total_cost        44989.539389
      median_family_income 82714.867135
      num_counties_in_st   96.394209
      st_cost_rank         48.697105
      st_med_aff_rank      48.688279
      st_income_rank       48.688279
      min_wage             9.711429
      m_income            1683.314286
      a_min_income        20199.771429
      m_ratio              0.449504
      a_min_ratio          0.449504
      median_household_income 63240.506841
      median_ratio         1.398116
      dtype: float64
```

```
[113]: data.std(numeric_only=True)
```

```
[113]: county_fips          15168.624870
      m_housing            274.255576
      m_food              36.104752
      m_transport         90.963024
      m_health            87.105507
      m_other             99.454344
      m_tax               122.434463
      m_total             491.214497
      a_housing           3291.066914
      a_food              433.257020
```

a_transport	1091.556288
a_health	1045.266081
a_other	1193.452125
a_taxes	1469.213556
a_total_cost	5894.573965
median_family_income	20074.548603
num_counties_in_st	57.025363
st_cost_rank	43.109713
st_med_aff_rank	42.814481
st_income_rank	42.814462
min_wage	2.910342
m_income	504.459227
a_min_income	6053.510718
m_ratio	0.123340
a_min_ratio	0.123340
median_household_income	16248.920157
median_ratio	0.257360
dtype:	float64

```
[112]: data.mean(numeric_only=True)
```

county_fips	30374.576519
m_housing	803.242426
m_food	337.483889
m_transport	1144.573051
m_health	490.475464
m_other	383.804850
m_tax	589.548597
m_total	3749.128282
a_housing	9638.909116
a_food	4049.806665
a_transport	13734.876612
a_health	5885.705567
a_other	4605.658203
a_taxes	7074.583165
a_total_cost	44989.539389
median_family_income	82714.867135
num_counties_in_st	96.394209
st_cost_rank	48.697105
st_med_aff_rank	48.688279
st_income_rank	48.688279
min_wage	9.711429
m_income	1683.314286
a_min_income	20199.771429
m_ratio	0.449504
a_min_ratio	0.449504
median_household_income	63240.506841

```
median_ratio          1.398116
dtype: float64
```

```
[113]: data.std(numeric_only=True)
```

```
[113]: county_fips          15168.624870
      m_housing           274.255576
      m_food              36.104752
      m_transport        90.963024
      m_health           87.105507
      m_other            99.454344
      m_tax              122.434463
      m_total            491.214497
      a_housing          3291.066914
      a_food             433.257020
      a_transport        1091.556288
      a_health           1045.266081
      a_other            1193.452125
      a_taxes            1469.213556
      a_total_cost       5894.573965
      median_family_income 20074.548603
      num_counties_in_st   57.025363
      st_cost_rank        43.109713
      st_med_aff_rank     42.814481
      st_income_rank      42.814462
      min_wage            2.910342
      m_income            504.459227
      a_min_income        6053.510718
      m_ratio             0.123340
      a_min_ratio         0.123340
      median_household_income 16248.920157
      median_ratio        0.257360
      dtype: float64
```

```
[224]: data.min(numeric_only=True)
```

```
[224]: county_fips          1001.000000
      m_housing           412.534217
      m_food              256.635217
      m_transport        531.612633
      m_health           320.842792
      m_other            237.797950
      m_tax              397.818642
      m_total            2994.983750
      a_housing          4950.410600
      a_food             3079.622600
      a_transport        6379.351600
```

a_health	3850.113500
a_other	2853.575400
a_taxes	4773.823700
a_total_cost	35939.805000
median_family_income	34894.000000
num_counties_in_st	1.000000
st_cost_rank	1.000000
st_med_aff_rank	1.000000
st_income_rank	1.000000
min_wage	7.250000
m_income	1256.666667
a_min_income	15080.000000
m_ratio	0.209606
a_min_ratio	0.209606
median_household_income	28972.000000
median_ratio	0.652809
dtype:	float64

```
[225]: data.max(numeric_only=True)
```

```
[225]: county_fips      56045.000000
m_housing      2849.000000
m_food        741.703942
m_transport    1767.560583
m_health      814.621342
m_other       1130.125750
m_tax         1669.890333
m_total       7727.068333
a_housing     34188.000000
a_food        8900.447300
a_transport    21210.727000
a_health      9775.456100
a_other       13561.509000
a_taxes       20038.684000
a_total_cost   92724.820000
median_family_income  211445.000000
num_counties_in_st    254.000000
st_cost_rank    254.000000
st_med_aff_rank  252.000000
st_income_rank   252.000000
min_wage        17.500000
m_income        3033.333333
a_min_income    36400.000000
m_ratio         0.852108
a_min_ratio     0.852108
median_household_income  167605.000000
median_ratio     2.711813
```

dtype: float64

```
[122]: datadf = pd.DataFrame(data)
datadfoutlier = datadf[datadf['min_wage']>=17]
datadfoutlier
```

```
[122]:
```

	state abv.	county_fips	county	m_housing	m_food	\
205	CA	6037	Los Angeles County	1777.000000	364.741617	
227	CA	6081	San Mateo County	2550.637333	436.185833	
320	DC	11001	District of Columbia	1687.667833	445.586425	
1208	MD	24031	Montgomery County	1803.139833	424.905192	

	m_transport	m_health	m_other	m_tax	m_total	...	\
205	984.808167	373.777833	720.603025	908.883500	5129.814167	...	
227	1169.506750	584.777833	1004.936167	1539.703000	7285.746750	...	
320	843.867833	542.053017	717.747400	965.791083	5202.713583	...	
1208	1001.162750	352.053017	749.640383	989.722417	5320.623667	...	

	st_cost_rank	st_med_aff_rank	st_income_rank	min_wage	m_income	\
205	22	26.0	28.0	17.27	2993.466667	
227	1	46.0	2.0	17.06	2957.066667	
320	1	1.0	1.0	17.50	3033.333333	
1208	2	19.0	2.0	17.15	2972.666667	

	a_min_income	m_ratio	a_min_ratio	median_household_income	\
205	35921.6	0.583543	0.583543	82455	
227	35484.8	0.405870	0.405870	143795	
320	36400.0	0.583029	0.583029	99897	
1208	35672.0	0.558706	0.558706	118020	

	median_ratio
205	1.339473
227	1.644707
320	1.600078
1208	1.848468

[4 rows x 29 columns]

```
[126]: datadfoutlier3 = datadf[datadf['a_total_cost']<=27305]
datadfoutlier3
```

```
[126]: Empty DataFrame
Columns: [state abv., county_fips, county, m_housing, m_food, m_transport,
m_health, m_other , m_tax, m_total, a_housing, a_food, a_transport, a_health,
a_other, a_taxes, a_total_cost, median_family_income, num_counties_in_st,
st_cost_rank, st_med_aff_rank, st_income_rank, min_wage, m_income, a_min_income,
m_ratio, a_min_ratio, median_household_income, median_ratio]
```

```
[0 rows x 29 columns]
```

$[111]$:

[111]: 2.909878666684371

```
[125]: datadfoutlier2 = datadf[datadf['a_total_cost']>=62673.4]
```

```
datadfoutlier2
```

[125]:	state	abv.	county	fips	county	m_housing	m_food	\
	187		CA	6001	Alameda County	1823.915167	420.204917	
	193		CA	6013	Contra Costa County	1827.062667	418.324792	
	207		CA	6041	Marin County	2278.204917	494.469317	
	213		CA	6053	Monterey County	2340.000000	370.381958	
	214		CA	6055	Napa County	1819.000000	440.886108	
	
	2925		VA	51600	Fairfax city	1884.858583	413.624508	
	2926		VA	51610	Falls Church city	1958.583000	433.365683	
	2936		VA	51685	Manassas Park city	1931.935583	358.161217	
	2970		WA	53033	King County	2252.594083	423.965125	
	2984		WA	53061	Snohomish County	2065.523583	397.643558	
		m_transport		m_health	m_other	m_tax	m_total	...
	187	1063.247417		532.777833	755.048908	1057.811583	5653.005833	...
	193	1126.897750		537.777833	755.475342	1087.298500	5752.836583	...
	207	1096.417667		529.777833	932.884417	1372.521333	6704.275417	...
	213	1239.690083		510.777833	911.925750	1389.984250	6762.759750	...
	214	1194.407417		529.777833	760.353517	1121.111833	5865.536500	...
	
	2925	1004.960750		366.053017	773.339683	1059.659083	5502.495417	...
	2926	870.210667		366.053017	804.786783	1056.068167	5489.067750	...
	2936	1244.308750		366.053017	770.518067	1143.825000	5814.801417	...
	2970	955.555167		376.777833	900.545917	956.448000	5865.886083	...
	2984	1124.091417		367.777833	828.748783	916.859250	5700.644500	...
		st_cost_rank		st_med_aff_rank	st_income_rank	min_wage	m_income	\
	187		6		49.0	5.0	16.00	2773.333333
	193		7		39.0	6.0	16.00	2773.333333
	207		2		53.0	1.0	16.00	2773.333333
	213		16		11.0	25.0	16.00	2773.333333
	214		10		41.0	9.0	16.00	2773.333333
	
	2925		7		126.0	5.0	12.00	2080.000000
	2926		3		133.0	1.0	12.00	2080.000000

2936	4	7.0	35.0	12.00	2080.000000
2970	1	38.0	1.0	16.28	2821.866667
2984	2	31.0	2.0	16.28	2821.866667

	a_min_income	m_ratio	a_min_ratio	median_household_income	\
187	33280.0	0.490594	0.490594	121190	
193	33280.0	0.482081	0.482081	119667	
207	33280.0	0.413666	0.413666	135960	
213	33280.0	0.410089	0.410089	91450	
214	33280.0	0.472818	0.472818	98580	
...	
2925	24960.0	0.378010	0.378010	122790	
2926	24960.0	0.378935	0.378935	142513	
2936	24960.0	0.357708	0.357708	90816	
2970	33862.4	0.481064	0.481064	116044	
2984	33862.4	0.495008	0.495008	101440	

	median_ratio
187	1.786513
193	1.733449
207	1.689966
213	1.126882
214	1.400554
...	...
2925	1.859611
2926	2.163588
2936	1.301506
2970	1.648572
2984	1.482873

[72 rows x 29 columns]

```
[128]: datadfoutlier4 = datadf[datadf['a_min_ratio']>=.758]
datadfoutlier4
```

```
[128]:
```

	state	abv.	county_fips	county	m_housing	m_food	\
1194	MD		24001	Allegany County	642.328408	336.539933	
2954	WA		53001	Adams County	754.000000	305.518108	
2955	WA		53003	Asotin County	819.946042	347.820600	
2958	WA		53009	Clallam County	847.000000	381.662642	
2960	WA		53013	Columbia County	735.000000	359.101275	
2961	WA		53015	Cowlitz County	909.000000	356.281125	
2963	WA		53019	Ferry County	640.000000	342.180300	
2965	WA		53023	Garfield County	676.000000	347.820600	
2967	WA		53027	Grays Harbor County	775.000000	355.341067	
2973	WA		53039	Klickitat County	836.000000	354.401000	
2974	WA		53041	Lewis County	843.000000	353.460942	

2975	WA	53043	Lincoln County	731.000000	368.501833
2976	WA	53045	Mason County	866.000000	363.801558
2977	WA	53047	Okanogan County	712.000000	323.379167
2978	WA	53049	Pacific County	754.000000	337.480000
2979	WA	53051	Pend Oreille County	702.000000	343.120317
2985	WA	53063	Spokane County	890.000000	368.501833
2986	WA	53065	Stevens County	751.000000	342.180300
2988	WA	53069	Wahkiakum County	727.000000	356.281125
2991	WA	53075	Whitman County	794.000000	376.022300

	m_transport	m_health	m_other	m_tax	m_total	...	\
1194	1045.776667	352.053017	329.346683	509.913333	3215.958000	...	
2954	1212.068583	392.583208	356.481808	460.397258	3481.049167	...	
2955	1080.206333	362.777833	392.902750	457.011967	3460.665333	...	
2958	1052.130083	359.583208	413.391600	466.991208	3520.758750	...	
2960	1198.619833	354.583208	368.117550	459.355675	3474.777667	...	
2961	1071.424417	455.777833	425.712117	499.704550	3717.900083	...	
2963	1209.159500	357.583208	330.461000	432.246458	3311.630500	...	
2965	1225.840083	354.583208	344.471150	446.067058	3394.782250	...	
2967	1090.622750	359.583208	380.310667	448.486408	3409.344083	...	
2973	1146.176750	336.583233	400.518233	470.954633	3544.633750	...	
2974	1149.899250	359.583208	402.557125	477.885133	3586.385750	...	
2975	1220.424417	357.583208	369.934608	465.732100	3513.176083	...	
2976	1160.333833	361.583208	413.774825	489.224125	3654.717417	...	
2977	1188.224000	392.583208	348.360108	449.221317	3413.767917	...	
2978	1158.775250	359.583208	367.235592	451.717200	3428.791333	...	
2979	1209.340917	357.583208	351.637575	449.049317	3412.731417	...	
2985	995.041333	365.777833	423.431192	464.797733	3507.549833	...	
2986	1248.310083	365.777833	367.807658	471.233117	3546.308917	...	
2988	1171.299667	359.583208	364.477050	452.029175	3430.670250	...	
2991	1056.086000	354.583208	393.661658	451.175533	3425.528667	...	

	st_cost_rank	st_med_aff_rank	st_income_rank	min_wage	m_income	...	\
1194	22	3.0	23.0	15.00	2600.000000	...	
2954	38	2.0	38.0	16.28	2821.866667	...	
2955	30	15.0	30.0	16.28	2821.866667	...	
2958	21	13.0	27.0	16.28	2821.866667	...	
2960	28	39.0	12.0	16.28	2821.866667	...	
2961	11	5.0	23.0	16.28	2821.866667	...	
2963	39	6.0	36.0	16.28	2821.866667	...	
2965	37	30.0	24.0	16.28	2821.866667	...	
2967	27	7.0	32.0	16.28	2821.866667	...	
2973	23	21.0	25.0	16.28	2821.866667	...	
2974	20	14.0	26.0	16.28	2821.866667	...	
2975	32	24.0	28.0	16.28	2821.866667	...	
2976	19	26.0	18.0	16.28	2821.866667	...	
2977	36	3.0	37.0	16.28	2821.866667	...	

2978	26	17.0	29.0	16.28	2821.866667
2979	35	8.0	34.0	16.28	2821.866667
2985	16	28.0	16.0	16.28	2821.866667
2986	31	16.0	31.0	16.28	2821.866667
2988	25	1.0	39.0	16.28	2821.866667
2991	34	35.0	19.0	16.28	2821.866667

	a_min_income	m_ratio	a_min_ratio	median_household_income	\
1194	31200.0	0.808468	0.808468	50582	
2954	33862.4	0.810637	0.810637	62410	
2955	33862.4	0.815412	0.815412	65625	
2958	33862.4	0.801494	0.801494	66693	
2960	33862.4	0.812100	0.812100	62935	
2961	33862.4	0.758995	0.758995	71847	
2963	33862.4	0.852108	0.852108	57064	
2965	33862.4	0.831236	0.831236	64413	
2967	33862.4	0.827686	0.827686	62258	
2973	33862.4	0.796095	0.796095	70371	
2974	33862.4	0.786827	0.786827	67954	
2975	33862.4	0.803224	0.803224	68318	
2976	33862.4	0.772116	0.772116	78029	
2977	33862.4	0.826614	0.826614	58661	
2978	33862.4	0.822992	0.822992	55615	
2979	33862.4	0.826865	0.826865	59332	
2985	33862.4	0.804512	0.804512	69079	
2986	33862.4	0.795719	0.795719	65748	
2988	33862.4	0.822541	0.822541	70099	
2991	33862.4	0.823776	0.823776	54412	

	median_ratio
1194	1.310703
2954	1.494042
2955	1.580260
2958	1.578566
2960	1.509329
2961	1.610385
2963	1.435949
2965	1.581177
2967	1.521749
2973	1.654402
2974	1.578981
2975	1.620518
2976	1.779185
2977	1.431971
2978	1.351667
2979	1.448791
2985	1.641198

```

2986      1.544987
2988      1.702753
2991      1.323689

```

[20 rows x 29 columns]

```
[129]: datadfoutlier5 = datadf[datadf['a_min_ratio']<=.142]
      datadfoutlier5
```

```
[129]: Empty DataFrame
      Columns: [state abv., county_fips, county, m_housing, m_food, m_transport,
      m_health, m_other , m_tax, m_total, a_housing, a_food, a_transport, a_health,
      a_other, a_taxes, a_total_cost, median_family_income, num_counties_in_st,
      st_cost_rank, st_med_aff_rank, st_income_rank, min_wage, m_income, a_min_income,
      m_ratio, a_min_ratio, median_household_income, median_ratio]
      Index: []
```

[0 rows x 29 columns]

```
[131]: datadfoutlier6 = datadf[datadf['median_household_income']>=111987]
      datadfoutlier6
```

```
[131]:
```

	state abv.	county_fips	county	m_housing \
68	AK	2016	Aleutians West Census Area	1349.000000
187	CA	6001	Alameda County	1823.915167
193	CA	6013	Contra Costa County	1827.062667
207	CA	6041	Marin County	2278.204917
224	CA	6075	San Francisco County	2132.731250
227	CA	6081	San Mateo County	2550.637333
229	CA	6085	Santa Clara County	2383.000000
252	CO	8014	Broomfield County	1906.903000
263	CO	8035	Douglas County	1926.211083
265	CO	8039	Elbert County	1320.305000
445	GA	13117	Forsyth County	1974.428583
495	GA	13219	Oconee County	1190.128000
726	IN	18057	Hamilton County	1230.887167
1086	KY	21185	Oldham County	1033.193833
1195	MD	24003	Anne Arundel County	1732.200167
1197	MD	24009	Calvert County	1510.906917
1201	MD	24017	Charles County	1685.003083
1203	MD	24021	Frederick County	1515.348167
1206	MD	24027	Howard County	1773.981917
1208	MD	24031	Montgomery County	1803.139833
1211	MD	24037	St. Mary's County	1470.000000
1226	MA	25017	Middlesex County	2161.855333
1228	MA	25021	Norfolk County	2376.973667
1324	MN	27019	Carver County	1207.509583

1384	MN	27139	Scott County	1241.794833
1776	NJ	34003	Bergen County	1513.822083
1784	NJ	34019	Hunterdon County	1500.925583
1787	NJ	34025	Monmouth County	1413.447750
1788	NJ	34027	Morris County	1628.298833
1792	NJ	34035	Somerset County	1689.090833
1811	NM	35028	Los Alamos County	995.000000
1858	NY	36059	Nassau County	1710.035000
1868	NY	36079	Putnam County	2394.327000
1880	NY	36103	Suffolk County	1706.139667
2064	OH	39041	Delaware County	1145.778167
2259	PA	42029	Chester County	1550.335917
2522	TN	47187	Williamson County	1832.267750
2566	TX	48085	Collin County	1648.510750
2653	TX	48259	Kendall County	1229.000000
2722	TX	48397	Rockwall County	1746.943000
2792	UT	49029	Morgan County	1079.666250
2799	UT	49043	Summit County	1383.000000
2803	UT	49051	Wasatch County	981.000000
2827	VA	51013	Arlington County	2020.760250
2849	VA	51059	Fairfax County	1980.789083
2850	VA	51061	Fauquier County	1386.552417
2857	VA	51075	Goochland County	1488.633667
2873	VA	51107	Loudoun County	2058.066583
2882	VA	51127	New Kent County	1570.332833
2893	VA	51153	Prince William County	1778.269083
2906	VA	51179	Stafford County	1709.874000
2925	VA	51600	Fairfax city	1884.858583
2926	VA	51610	Falls Church city	1958.583000
2970	WA	53033	King County	2252.594083
3139	WY	56039	Teton County	1317.000000

	m_food	m_transport	m_health	m_other	m_tax \
68	449.346642	1025.811917	728.583250	605.065508	725.969400
187	420.204917	1063.247417	532.777833	755.048908	1057.811583
193	418.324792	1126.897750	537.777833	755.475342	1087.298500
207	494.469317	1096.417667	529.777833	932.884417	1372.521333
224	483.188642	711.081950	574.777833	880.143417	1137.178833
227	436.185833	1169.506750	584.777833	1004.936167	1539.703000
229	413.624508	1117.580167	453.777833	940.942667	1362.795917
252	388.243000	1037.334500	404.777833	772.216958	1016.534667
263	401.403808	1149.467500	404.777833	783.141275	1073.220333
265	365.681683	1354.404333	404.777833	567.261275	840.198000
445	390.123125	1189.301583	454.053017	795.568925	1191.887750
495	368.501833	1209.641833	486.053017	524.411217	813.644775
726	368.501833	1091.121250	439.677733	538.124917	703.860592
1086	334.659833	1240.897583	436.053017	460.223350	730.045083

1195	381.662642	1114.850750	352.053017	711.222983	975.627917
1197	392.943275	1297.733750	352.053017	640.562867	940.403917
1201	360.981408	1202.138750	352.053017	688.384925	974.388000
1203	376.962325	1156.409000	352.053017	636.680258	884.557667
1206	409.864300	1095.976750	352.053017	734.769367	1002.710250
1208	424.905192	1001.162750	352.053017	749.640383	989.722417
1211	362.861492	1234.711667	352.053017	616.678267	884.151583
1226	420.204917	1041.497583	393.842775	868.751083	1189.015250
1228	432.425617	1091.623250	389.842775	945.240833	1317.078083
1324	438.065958	1207.299667	346.677733	553.664717	764.247150
1384	392.943275	1235.801500	346.677733	550.018392	752.868900
1776	403.283892	1020.270833	462.842775	645.022908	800.355717
1784	383.542725	1187.359417	462.842775	634.041708	850.352417
1787	403.283892	1083.700833	455.842775	611.251300	772.246908
1788	406.104083	1142.305083	462.842775	684.488200	906.075083
1792	396.703492	1140.378167	455.842775	701.779133	927.602500
1811	386.362917	1141.841167	601.583208	464.768592	688.548017
1858	394.823408	1117.987667	656.842775	708.193358	1112.916750
1868	406.104083	1238.034583	649.842775	942.223417	1499.033333
1880	396.703492	1182.969250	656.842775	707.515383	1136.226917
2064	361.921467	1141.394667	450.677733	507.275433	650.373575
2259	412.684450	1124.296333	346.842775	660.471067	852.803917
2522	423.965125	1144.049833	469.053017	759.124350	867.807167
2566	354.401000	1032.091167	433.053017	673.892825	721.306400
2653	328.079450	1227.222583	453.053017	523.889608	611.830117
2722	346.880575	1139.969417	433.053017	704.480717	789.509117
2792	344.060383	1399.862667	443.777833	479.022175	794.006675
2799	480.368450	1261.528250	418.583208	626.942542	952.193250
2803	363.801558	1233.820250	418.583208	452.467283	698.090658
2827	471.907958	864.354750	366.053017	838.674583	1103.238833
2849	433.365683	1032.460750	366.053017	812.258217	1126.515000
2850	377.902383	1244.002833	366.053017	593.662392	884.338000
2857	356.281125	1247.267500	379.053017	620.733683	930.045000
2873	417.384725	1155.098250	366.053017	832.881833	1203.570750
2882	355.341067	1294.734417	379.053017	647.905600	988.713667
2893	384.482792	1174.547917	366.053017	727.672042	1055.347583
2906	373.202108	1225.552583	366.053017	700.864667	1035.137500
2925	413.624508	1004.960750	366.053017	773.339683	1059.659083
2926	433.365683	870.210667	366.053017	804.786783	1056.068167
2970	423.965125	955.555167	376.777833	900.545917	956.448000
3139	475.668175	1252.413583	674.583208	603.154992	774.953450

	m_total	...	st_cost_rank	st_med_aff_rank	st_income_rank	\
68	4883.776667	...	5	27.0	2.0	
187	5653.005833	...	6	49.0	5.0	
193	5752.836583	...	7	39.0	6.0	
207	6704.275417	...	2	53.0	1.0	

224	5919.102250	...	3	43.0	4.0
227	7285.746750	...	1	46.0	2.0
229	6671.721333	...	4	55.0	3.0
252	5526.009750	...	11	62.0	2.0
263	5738.222000	...	7	61.0	1.0
265	4852.628250	...	15	59.0	5.0
445	5995.363250	...	1	157.0	1.0
495	4592.380833	...	19	159.0	2.0
726	4372.173500	...	1	92.0	1.0
1086	4235.072583	...	7	120.0	1.0
1195	5267.617500	...	3	12.0	4.0
1197	5134.603500	...	7	24.0	3.0
1201	5262.949250	...	5	17.0	7.0
1203	4922.010417	...	8	21.0	5.0
1206	5369.355500	...	1	20.0	1.0
1208	5320.623667	...	2	19.0	2.0
1211	4920.456083	...	12	23.0	8.0
1226	6075.166667	...	5	14.0	2.0
1228	6553.184250	...	1	13.0	1.0
1324	4517.464833	...	2	87.0	1.0
1384	4520.104833	...	3	76.0	2.0
1776	4845.598333	...	6	17.0	5.0
1784	5019.064417	...	3	21.0	3.0
1787	4739.773417	...	7	19.0	4.0
1788	5230.114250	...	2	20.0	1.0
1792	5311.396833	...	1	18.0	2.0
1811	4278.103833	...	1	33.0	1.0
1858	5700.798833	...	5	62.0	1.0
1868	7129.565083	...	1	44.0	4.0
1880	5786.397167	...	4	56.0	3.0
2064	4257.420917	...	1	88.0	1.0
2259	4947.434583	...	1	67.0	1.0
2522	5496.266917	...	1	95.0	1.0
2566	4863.255250	...	4	249.0	1.0
2653	4373.074833	...	9	248.0	4.0
2722	5160.835917	...	1	243.0	3.0
2792	4540.396167	...	4	28.0	2.0
2799	5122.615917	...	1	29.0	1.0
2803	4147.763000	...	9	27.0	3.0
2827	5664.989583	...	1	120.0	3.0
2849	5751.442083	...	6	128.0	4.0
2850	4852.511083	...	22	131.0	7.0
2857	5022.014000	...	15	125.0	10.0
2873	6033.055333	...	5	132.0	2.0
2882	5236.080417	...	19	117.0	18.0
2893	5486.372417	...	8	114.0	9.0
2906	5410.683917	...	10	127.0	8.0

2925	5502.495417	...	7	126.0	5.0
2926	5489.067750	...	3	133.0	1.0
2970	5865.886083	...	1	38.0	1.0
3139	5097.773417	...	1	13.0	1.0

	min_wage	m_income	a_min_income	m_ratio	a_min_ratio	\
68	11.73	2033.200000	24398.4	0.416317	0.416317	
187	16.00	2773.333333	33280.0	0.490594	0.490594	
193	16.00	2773.333333	33280.0	0.482081	0.482081	
207	16.00	2773.333333	33280.0	0.413666	0.413666	
224	16.00	2773.333333	33280.0	0.468540	0.468540	
227	17.06	2957.066667	35484.8	0.405870	0.405870	
229	16.00	2773.333333	33280.0	0.415685	0.415685	
252	14.42	2499.466667	29993.6	0.452309	0.452309	
263	14.42	2499.466667	29993.6	0.435582	0.435582	
265	14.42	2499.466667	29993.6	0.515075	0.515075	
445	7.25	1256.666667	15080.0	0.209606	0.209606	
495	7.25	1256.666667	15080.0	0.273642	0.273642	
726	7.25	1256.666667	15080.0	0.287424	0.287424	
1086	7.25	1256.666667	15080.0	0.296728	0.296728	
1195	15.00	2600.000000	31200.0	0.493582	0.493582	
1197	15.00	2600.000000	31200.0	0.506368	0.506368	
1201	15.00	2600.000000	31200.0	0.494020	0.494020	
1203	15.00	2600.000000	31200.0	0.528239	0.528239	
1206	15.00	2600.000000	31200.0	0.484229	0.484229	
1208	17.15	2972.666667	35672.0	0.558706	0.558706	
1211	15.00	2600.000000	31200.0	0.528406	0.528406	
1226	15.00	2600.000000	31200.0	0.427972	0.427972	
1228	15.00	2600.000000	31200.0	0.396754	0.396754	
1324	10.85	1880.666667	22568.0	0.416310	0.416310	
1384	10.85	1880.666667	22568.0	0.416067	0.416067	
1776	15.13	2622.533333	31470.4	0.541220	0.541220	
1784	15.13	2622.533333	31470.4	0.522514	0.522514	
1787	15.13	2622.533333	31470.4	0.553304	0.553304	
1788	15.13	2622.533333	31470.4	0.501429	0.501429	
1792	15.13	2622.533333	31470.4	0.493756	0.493756	
1811	12.00	2080.000000	24960.0	0.486197	0.486197	
1858	16.00	2773.333333	33280.0	0.486482	0.486482	
1868	15.00	2600.000000	31200.0	0.364679	0.364679	
1880	16.00	2773.333333	33280.0	0.479285	0.479285	
2064	10.45	1811.333333	21736.0	0.425453	0.425453	
2259	7.25	1256.666667	15080.0	0.254004	0.254004	
2522	7.25	1256.666667	15080.0	0.228640	0.228640	
2566	7.25	1256.666667	15080.0	0.258400	0.258400	
2653	7.25	1256.666667	15080.0	0.287365	0.287365	
2722	7.25	1256.666667	15080.0	0.243501	0.243501	
2792	7.25	1256.666667	15080.0	0.276775	0.276775	

2799	7.25	1256.666667	15080.0	0.245317	0.245317
2803	7.25	1256.666667	15080.0	0.302975	0.302975
2827	12.00	2080.000000	24960.0	0.367167	0.367167
2849	12.00	2080.000000	24960.0	0.361648	0.361648
2850	12.00	2080.000000	24960.0	0.428644	0.428644
2857	12.00	2080.000000	24960.0	0.414176	0.414176
2873	12.00	2080.000000	24960.0	0.344767	0.344767
2882	12.00	2080.000000	24960.0	0.397244	0.397244
2893	12.00	2080.000000	24960.0	0.379121	0.379121
2906	12.00	2080.000000	24960.0	0.384425	0.384425
2925	12.00	2080.000000	24960.0	0.378010	0.378010
2926	12.00	2080.000000	24960.0	0.378935	0.378935
2970	16.28	2821.866667	33862.4	0.481064	0.481064
3139	7.25	1256.666667	15080.0	0.246513	0.246513

	median_household_income	median_ratio
68	112218	1.914809
187	121190	1.786513
193	119667	1.733449
207	135960	1.689966
224	135366	1.905779
227	143795	1.644707
229	150502	1.879850
252	114746	1.730393
263	140768	2.044303
265	120394	2.067505
445	130909	1.819587
495	114758	2.082398
726	117068	2.231308
1086	116173	2.285931
1195	112525	1.780137
1197	118110	1.916896
1201	112932	1.788161
1203	116796	1.977444
1206	133068	2.065239
1208	118020	1.848468
1211	112154	1.899451
1226	118494	1.625388
1228	115770	1.472185
1324	117602	2.169395
1384	115687	2.132823
1776	113885	1.958564
1784	137334	2.280206
1787	117699	2.069350
1788	131562	2.096226
1792	134735	2.113929
1811	139217	2.711813

1858	135528	1.981126
1868	112167	1.311055
1880	119253	1.717433
2064	121528	2.378748
2259	117326	1.976209
2522	133486	2.023889
2566	113943	1.952447
2653	120238	2.291256
2722	124291	2.006958
2792	119482	2.192944
2799	132358	2.153164
2803	115793	2.326415
2827	131020	1.927335
2849	144632	2.095590
2850	120301	2.065958
2857	113617	1.885316
2873	167605	2.315093
2882	115627	1.840228
2893	119051	1.808283
2906	134456	2.070841
2925	122790	1.859611
2926	142513	2.163588
2970	116044	1.648572
3139	127677	2.087137

[55 rows x 29 columns]

```
[132]: datadfoutlier6 = datadf[datadf['median_household_income']<=14500]
datadfoutlier6
```

```
[132]: Empty DataFrame
Columns: [state abv., county_fips, county, m_housing, m_food, m_transport,
m_health, m_other , m_tax, m_total, a_housing, a_food, a_transport, a_health,
a_other, a_taxes, a_total_cost, median_family_income, num_counties_in_st,
st_cost_rank, st_med_aff_rank, st_income_rank, min_wage, m_income, a_min_income,
m_ratio, a_min_ratio, median_household_income, median_ratio]
Index: []
```

[0 rows x 29 columns]

```
[133]: datadfoutlier7 = datadf[datadf['median_ratio']>=2.18]
datadfoutlier7
```

```
[133]:      state abv.  county_fips      county  m_housing  m_food \
726      IN      18057  Hamilton County  1230.887167  368.501833
814      IA      19049   Dallas County    921.501833  380.722575
1086     KY      21185   Oldham County   1033.193833  334.659833
```

1784	NJ	34019	Hunterdon County	1500.925583	383.542725
1811	NM	35028	Los Alamos County	995.000000	386.362917
2064	OH	39041	Delaware County	1145.778167	361.921467
2071	OH	39055	Geauga County	822.186442	397.643558
2126	OH	39165	Warren County	994.050333	342.180300
2653	TX	48259	Kendall County	1229.000000	328.079450
2792	UT	49029	Morgan County	1079.666250	344.060383
2803	UT	49051	Wasatch County	981.000000	363.801558
2873	VA	51107	Loudoun County	2058.066583	417.384725

	m_transport	m_health	m_other	m_tax	m_total	...	\
726	1091.121250	439.677733	538.124917	703.860592	4372.173500	...	
814	1142.270833	405.677733	438.141967	653.790850	3942.105833	...	
1086	1240.897583	436.053017	460.223350	730.045083	4235.072583	...	
1784	1187.359417	462.842775	634.041708	850.352417	5019.064417	...	
1811	1141.841167	601.583208	464.768592	688.548017	4278.103833	...	
2064	1141.394667	450.677733	507.275433	650.373575	4257.420917	...	
2071	1194.671833	442.677733	410.419800	572.328692	3839.928083	...	
2126	1112.372917	435.677733	449.583575	587.420250	3921.285167	...	
2653	1227.222583	453.053017	523.889608	611.830117	4373.074833	...	
2792	1399.862667	443.777833	479.022175	794.006675	4540.396167	...	
2803	1233.820250	418.583208	452.467283	698.090658	4147.763000	...	
2873	1155.098250	366.053017	832.881833	1203.570750	6033.055333	...	

	st_cost_rank	st_med_aff_rank	st_income_rank	min_wage	m_income	...	\
726	1	92.0	1.0	7.25	1256.666667		
814	1	99.0	1.0	7.25	1256.666667		
1086	7	120.0	1.0	7.25	1256.666667		
1784	3	21.0	3.0	15.13	2622.533333		
1811	1	33.0	1.0	12.00	2080.000000		
2064	1	88.0	1.0	10.45	1811.333333		
2071	3	84.0	4.0	10.45	1811.333333		
2126	2	86.0	2.0	10.45	1811.333333		
2653	9	248.0	4.0	7.25	1256.666667		
2792	4	28.0	2.0	7.25	1256.666667		
2803	9	27.0	3.0	7.25	1256.666667		
2873	5	132.0	2.0	12.00	2080.000000		

	a_min_income	m_ratio	a_min_ratio	median_household_income	\
726	15080.0	0.287424	0.287424	117068	
814	15080.0	0.318781	0.318781	109138	
1086	15080.0	0.296728	0.296728	116173	
1784	31470.4	0.522514	0.522514	137334	
1811	24960.0	0.486197	0.486197	139217	
2064	21736.0	0.425453	0.425453	121528	
2071	21736.0	0.471710	0.471710	101636	
2126	21736.0	0.461923	0.461923	104523	

2653	15080.0	0.287365	0.287365	120238
2792	15080.0	0.276775	0.276775	119482
2803	15080.0	0.302975	0.302975	115793
2873	24960.0	0.344767	0.344767	167605

	median_ratio
726	2.231308
814	2.307100
1086	2.285931
1784	2.280206
1811	2.711813
2064	2.378748
2071	2.205684
2126	2.221274
2653	2.291256
2792	2.192944
2803	2.326415
2873	2.315093

[12 rows x 29 columns]

```
[135]: datadfoutlier8 = datadf[datadf['median_ratio']<=.77]
datadfoutlier8
```

```
[135]:
```

	state abv.	county_fips	county	m_housing	m_food	\
65	AL	1131	Wilcox County	779.0000	320.559000	
1427	MS	28051	Holmes County	704.0000	322.439108	
1428	MS	28053	Humphreys County	775.0000	322.439108	
1831	NY	36005	Bronx County	1885.0075	378.842450	
2370	SD	46017	Buffalo County	720.0000	344.060383	
3016	WV	54047	McDowell County	668.0000	313.978617	

	m_transport	m_health	m_other	m_tax	m_total	...	\
65	1133.738833	502.199867	369.953817	606.010942	3711.462583	...	
1427	1076.569083	532.053017	345.352175	556.212242	3536.625667	...	
1428	1091.783250	522.199867	369.240600	581.478067	3662.140917	...	
1831	531.612633	695.842775	761.687175	989.748250	5242.740583	...	
2370	1171.214917	585.727500	358.010092	491.912642	3670.925417	...	
3016	1030.745167	755.199867	330.393150	600.056967	3698.373667	...	

	st_cost_rank	st_med_aff_rank	st_income_rank	min_wage	m_income	\
65	15	6.0	60.0	7.25	1256.666667	
1427	20	2.0	80.0	7.25	1256.666667	
1428	68	3.0	81.0	7.25	1256.666667	
1831	14	1.0	62.0	15.00	2600.000000	
2370	35	8.0	59.0	11.20	1941.333333	
3016	40	1.0	55.0	8.75	1516.666667	

	a_min_income	m_ratio	a_min_ratio	median_household_income	\
65	15080.0	0.338591	0.338591		33682
1427	15080.0	0.355329	0.355329		31972
1428	15080.0	0.343151	0.343151		33148
1831	31200.0	0.495924	0.495924		45864
2370	23296.0	0.528840	0.528840		30659
3016	18200.0	0.410090	0.410090		28972

	median_ratio
65	0.756261
1427	0.753355
1428	0.754295
1831	0.729008
2370	0.695987
3016	0.652809

[6 rows x 29 columns]

```
[140]: datadf['a_min_income'].mode
```

```
[140]: <bound method Series.mode of 0      15080.0
1      15080.0
2      15080.0
3      15080.0
4      15080.0
...
3138   15080.0
3139   15080.0
3140   15080.0
3141   15080.0
3142   15080.0
Name: a_min_income, Length: 3143, dtype: float64>
```

```
[141]: datadf['a_total_cost'].mode
```

```
[141]: <bound method Series.mode of 0      48384.410
1      50978.832
2      40947.977
3      43405.910
4      41932.285
...
3138   49277.215
3139   61173.281
3140   44581.816
3141   44393.578
3142   48843.875
```

Name: a_total_cost, Length: 3143, dtype: float64>

```
[142]: datadf['a_min_ratio'].mode
```

```
[142]: <bound method Series.mode of 0      0.311671
1      0.295809
2      0.368272
3      0.347418
4      0.359627
...
3138   0.306024
3139   0.246513
3140   0.338255
3141   0.339689
3142   0.308739
Name: a_min_ratio, Length: 3143, dtype: float64>
```

```
[143]: datadf['median_household_income'].mode
```

```
[143]: <bound method Series.mode of 0      70148
1      71704
2      41151
3      54309
4      60553
...
3138   75779
3139  127677
3140   73072
3141   60699
3142   67677
Name: median_household_income, Length: 3143, dtype: int64>
```

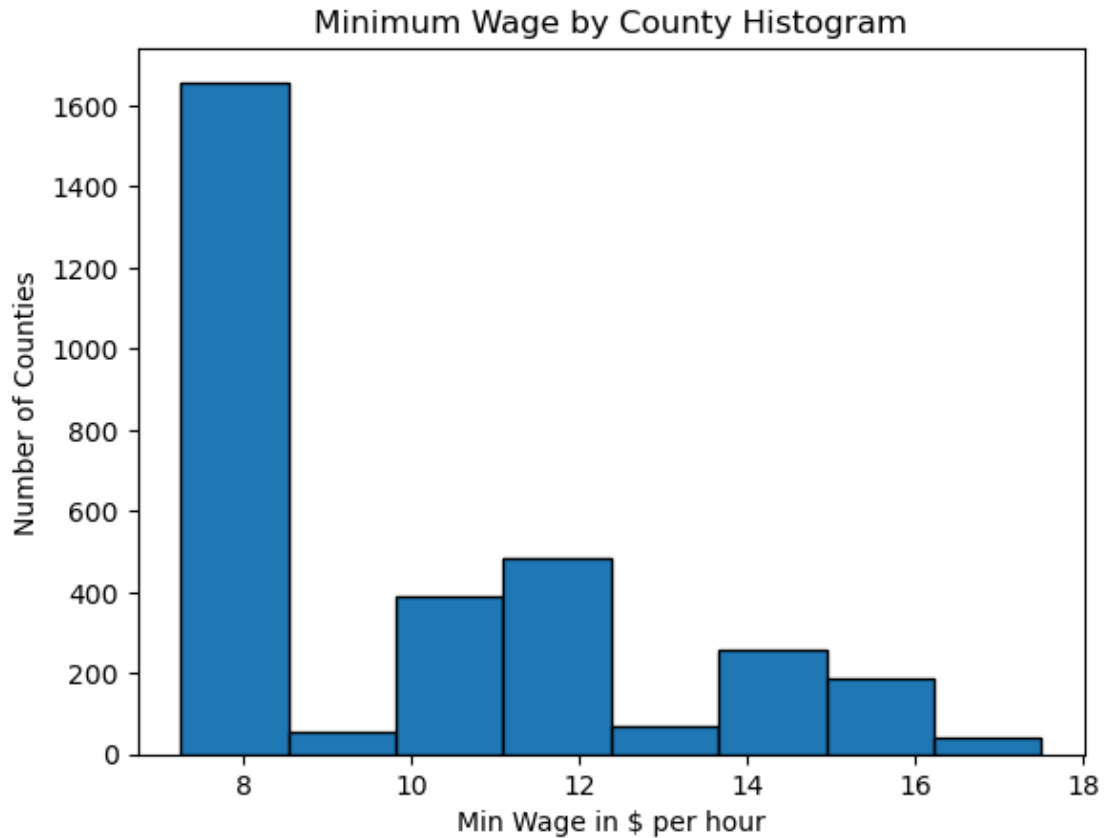
```
[144]: datadf['median_ratio'].mode
```

```
[144]: <bound method Series.mode of 0      1.449806
1      1.406545
2      1.004958
3      1.251189
4      1.444066
...
3138   1.537810
3139   2.087137
3140   1.639054
3141   1.367292
3142   1.385578
Name: median_ratio, Length: 3143, dtype: float64>
```

1.0.3 Part 4 - Creating Histograms

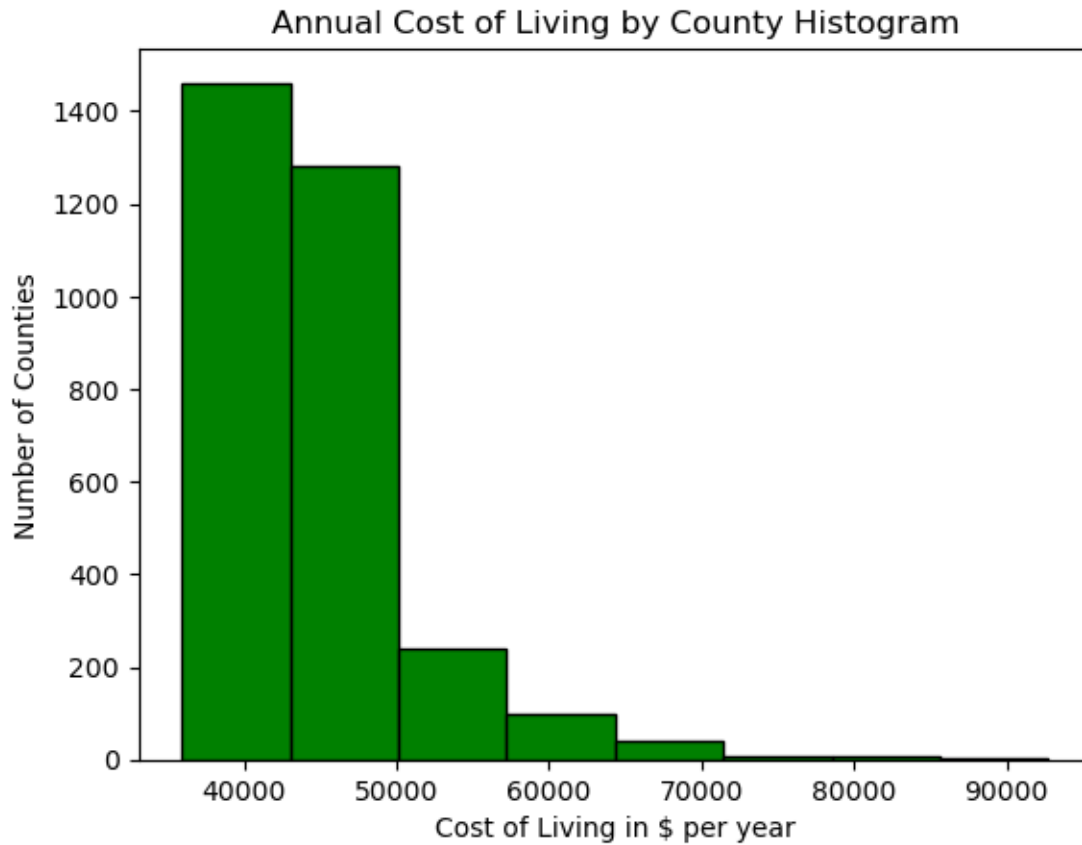
```
[77]: plt.hist(data['min_wage'], bins = 8, edgecolor='black')
plt.xlabel('Min Wage in $ per hour')
plt.ylabel('Number of Counties')
plt.title('Minimum Wage by County Histogram')

min_wage_hist = plt
min_wage_hist.show()
```



```
[108]: plt.hist(data['a_total_cost'], bins = 8, color='green', edgecolor='black')
plt.xlabel('Cost of Living in $ per year')
plt.ylabel('Number of Counties')
plt.title('Annual Cost of Living by County Histogram')

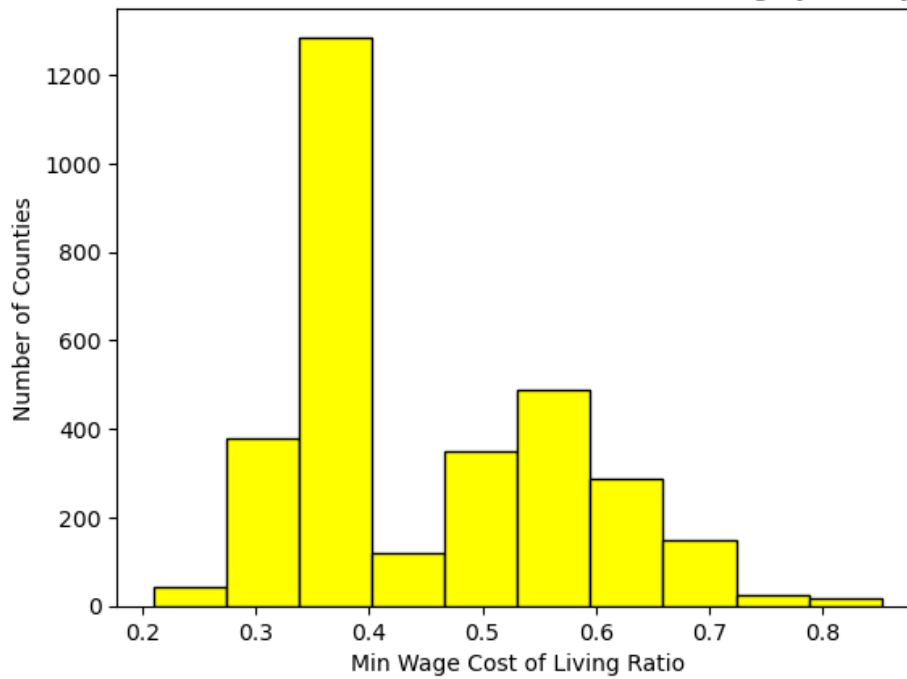
a_total_cost_hist = plt
a_total_cost_hist.show()
```



```
[105]: plt.hist(data['a_min_ratio'], bins = 10, color='yellow', edgecolor='black')
plt.xlabel('Min Wage Cost of Living Ratio')
plt.ylabel('Number of Counties')
plt.title('Ratio of Annual Minimum Income and Annual Cost of Living by County_
↳Histogram')

a_min_ratio_hist = plt
a_min_ratio_hist.show()
```

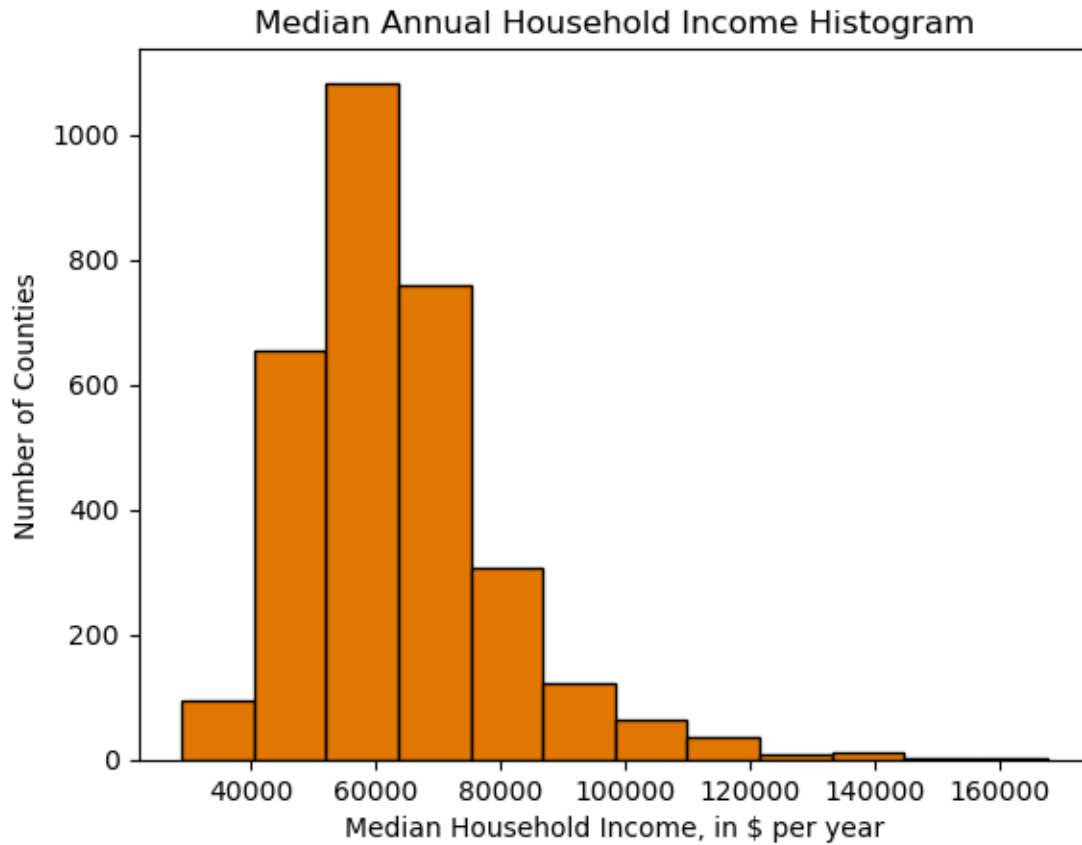

Ratio of Annual Minimum Income and Annual Cost of Living by County Histogram



[]:

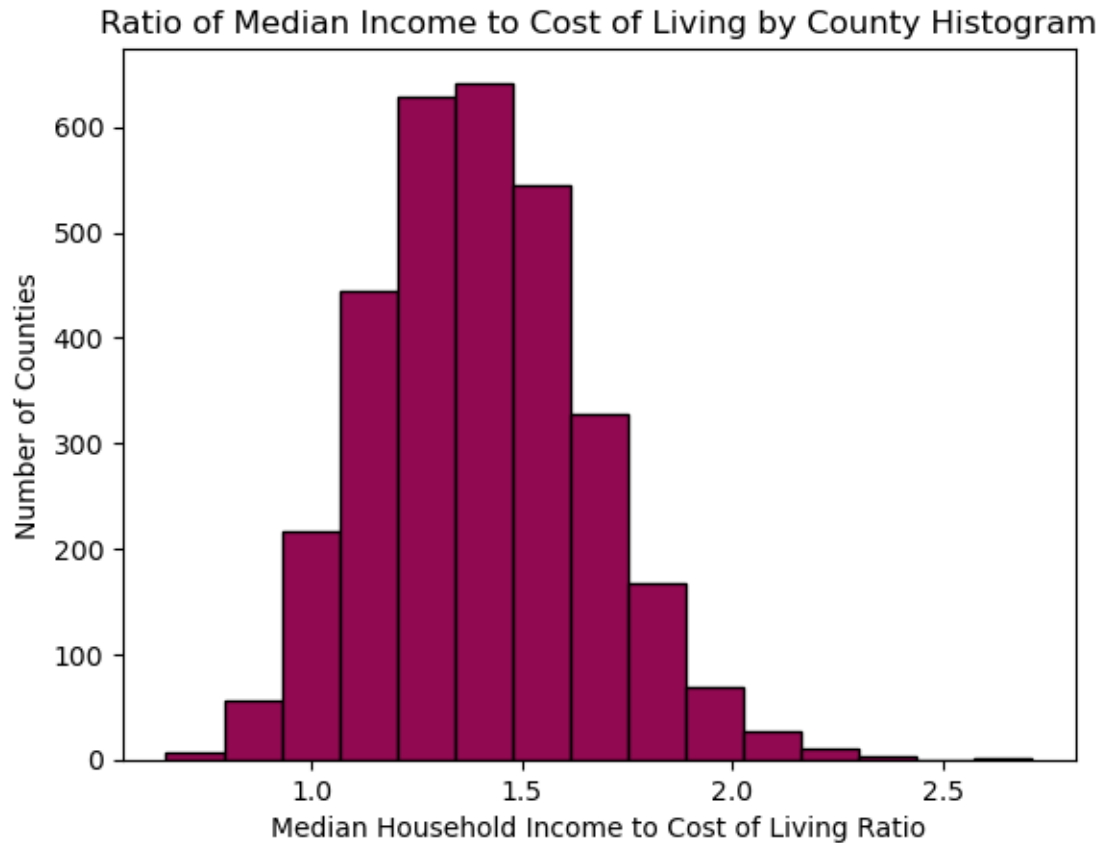
```
[89]: plt.hist(data['median_household_income'], bins = 10, color='orange',
             edgecolor='black')
plt.xlabel('Median Household Income, in $ per year')
plt.ylabel('Number of Counties')
plt.title('Median Annual Household Income Histogram')

median_household_income_hist = plt
median_household_income_hist.show()
```



```
[106]: plt.hist(data['median_ratio'], bins = 15, color='#910951', edgecolor='black')
plt.xlabel('Median Household Income to Cost of Living Ratio')
plt.ylabel('Number of Counties')
plt.title('Ratio of Median Income to Cost of Living by County Histogram')

median_household_income_hist = plt
median_household_income_hist.show()
```



1.0.4 Part 5 - PMF

```
[244]: import thinkplot
import thinkstats2
pmf1data = datadf[datadf['min_wage']>=12]
pmf2data = datadf[datadf['min_wage']<12]
```

```
[ ]:
```

```
[245]: bins = 16
binlabels = [35000, 38250, 41500, 44850, 48000, 51250, 54500, 57750, 61000, 64250, 67500, 70750, 74000, 77250, 80500, 83750]
pmf1bin = pd.cut(pmf1data.a_total_cost, bins=bins, labels=binlabels)
pmf2bin = pd.cut(pmf2data.a_total_cost, bins=bins, labels=binlabels)

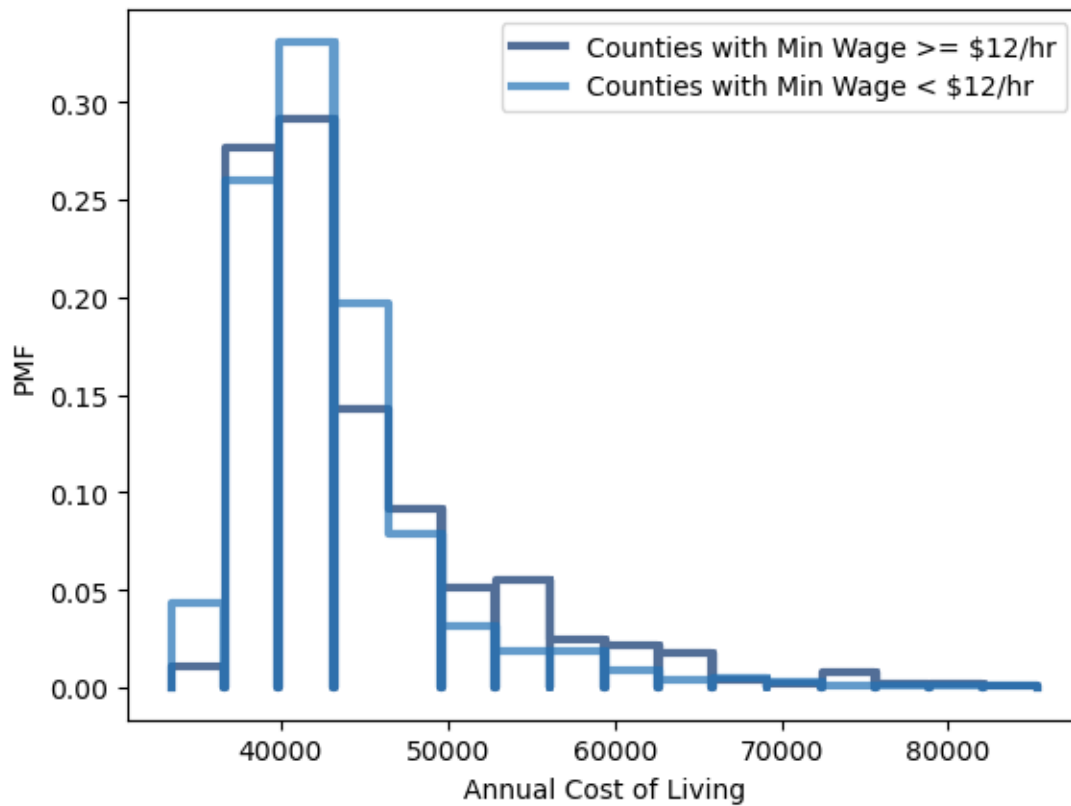
cost1 = thinkstats2.Pmf(pmf1bin, label = "Counties with Min Wage >= $12/hr")
cost2 = thinkstats2.Pmf(pmf2bin, label = "Counties with Min Wage < $12/hr")

axis = [35000, 93000, 0, 0.4]
```

```

first_pmf = thinkplot.Pmf(cost1)
other_pmf = thinkplot.Pmf(cost2)
thinkplot.Config(xlabel="Annual Cost of Living", ylabel="PMF")

```

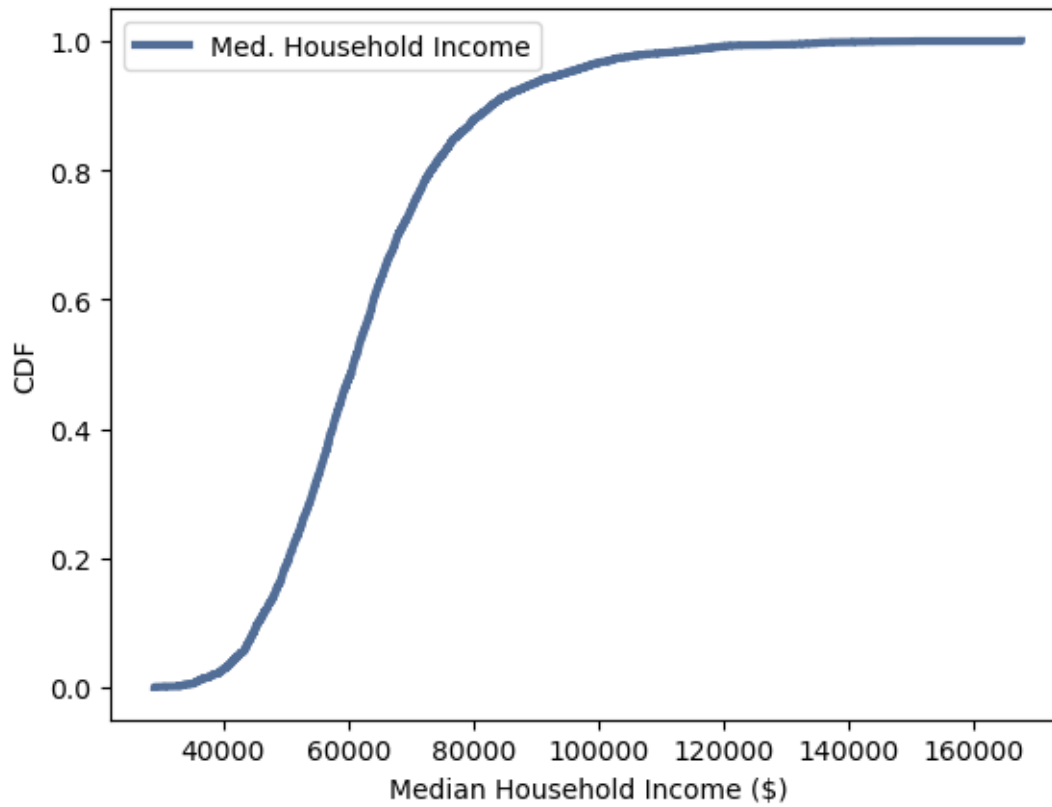


1.0.5 Part 6 - CDF

```

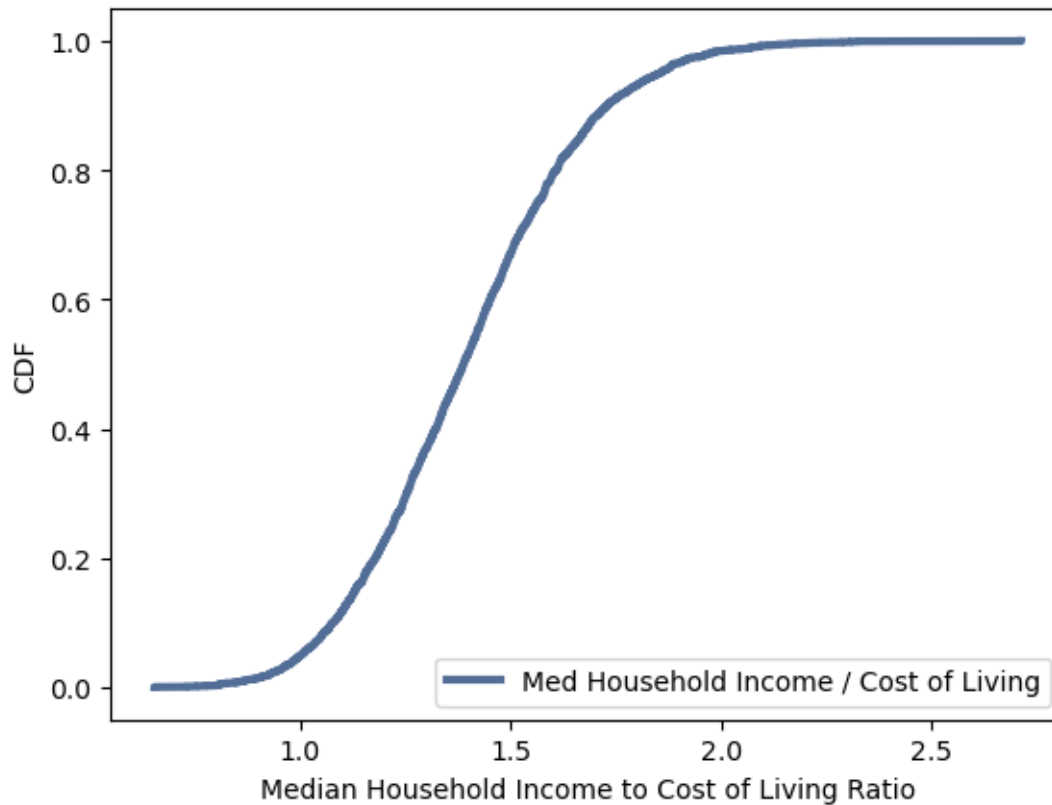
[259]: cdf = thinkstats2.Cdf(datadf.median_household_income, label='Med. Household_
↪Income')
thinkplot.Cdf(cdf)
thinkplot.Show(xlabel='Median Household Income ($)', ylabel="CDF")

```



<Figure size 800x600 with 0 Axes>

```
[256]: cdf2 = thinkstats2.Cdf(datadf.median_ratio, label='Med Household Income / ')
thinkplot.Cdf(cdf2)
thinkplot.Show(xlabel='Median Household Income to Cost of Living Ratio',
               ylabel="CDF")
```



<Figure size 800x600 with 0 Axes>

1.0.6 Part 7 - Analytical Distribution

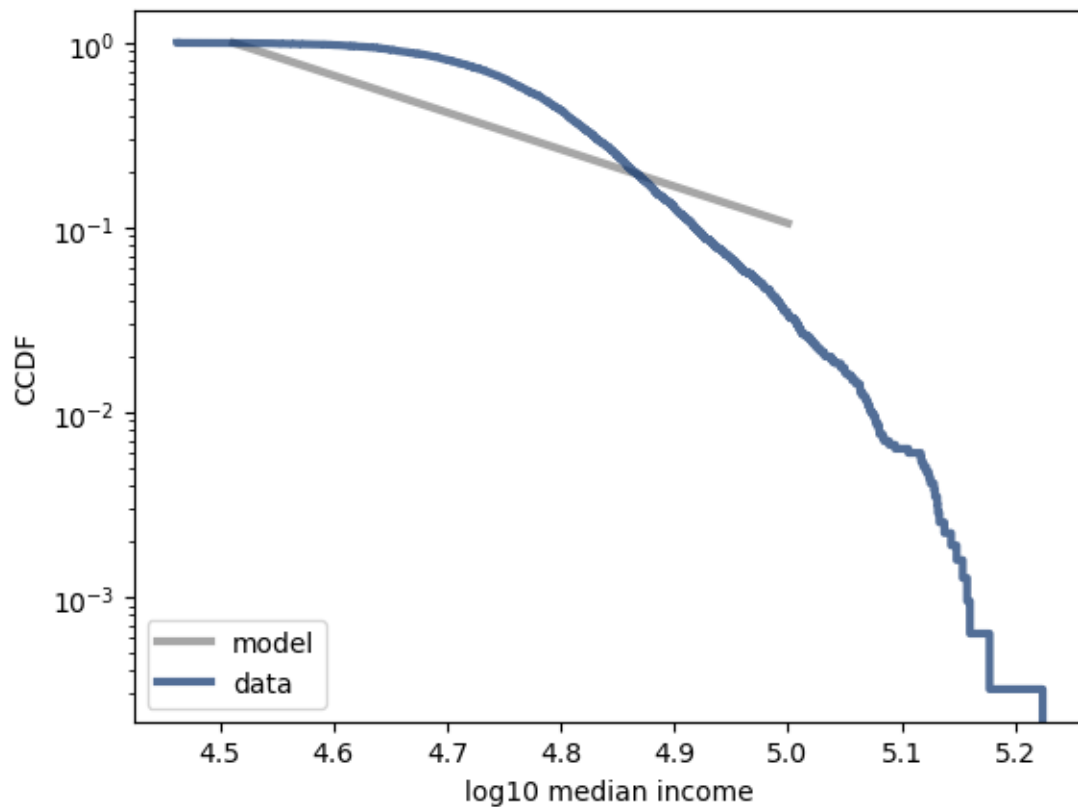
```
[276]: import scipy.stats
```

```
[282]: pareto_data = datadf
```

```
log_income = np.log10(pareto_data.median_household_income)
cdf = thinkstats2.Cdf(pareto_data.median_household_income)
cdf_log = thinkstats2.Cdf(log_income, label="data")
```

```
[319]: xs, ys = thinkstats2.RenderParetoCdf(xmin=32500, alpha=2, low=0, high=1e5)
thinkplot.Plot(np.log10(xs), 1 - ys, label="model", color="0.5")

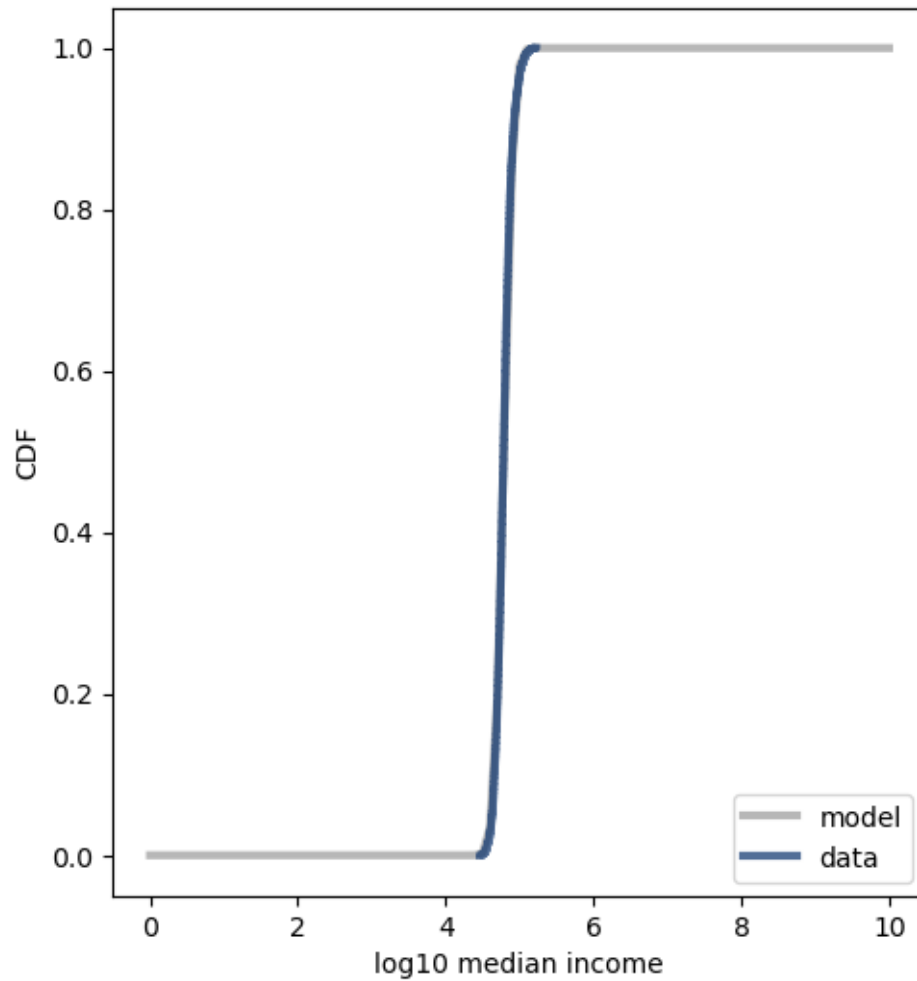
thinkplot.Cdf(cdf_log, complement=True)
thinkplot.Config(
    xlabel="log10 median income", ylabel="CCDF", yscale="log", loc="lower left")
```



```
[325]: thinkplot.PrePlot(cols=2)

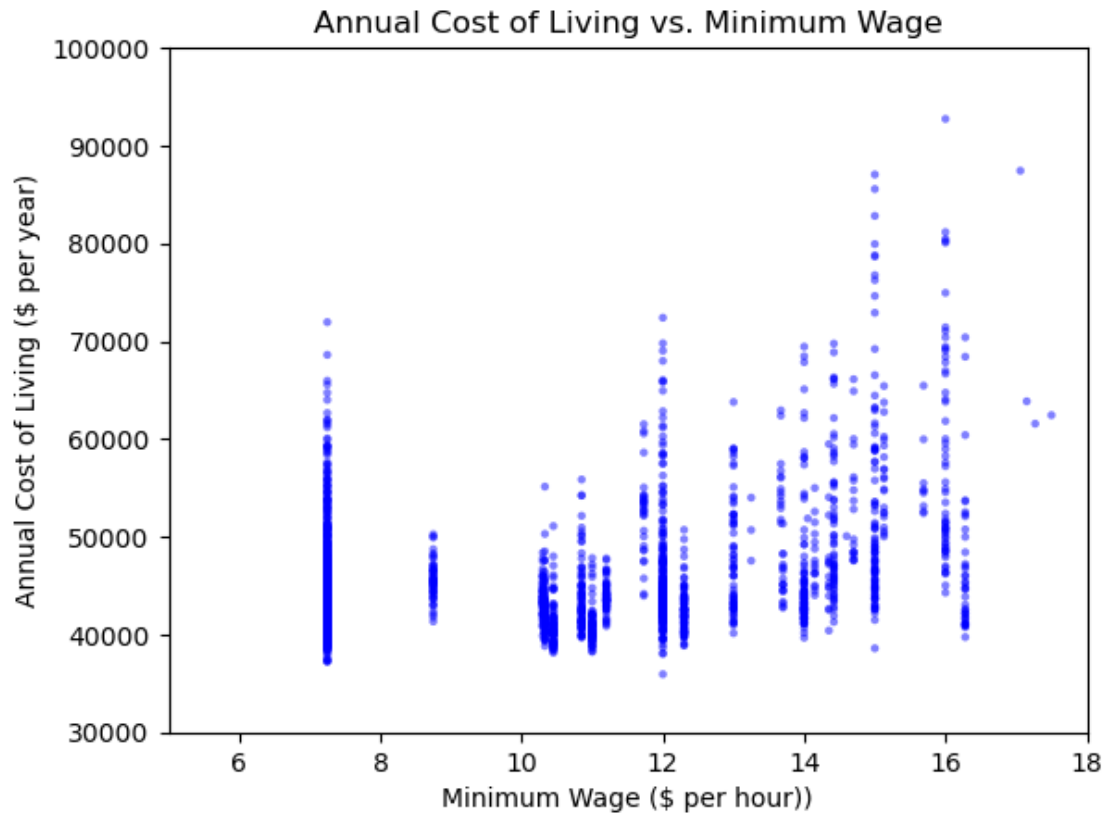
mu, sigma = log_income.mean(), log_income.std()
xs, ps = thinkstats2.RenderNormalCdf(mu, sigma, low=0, high=10)
thinkplot.Plot(xs, ps, label="model", color="0.6")

thinkplot.Cdf(cdf_log)
thinkplot.Config(xlabel="log10 median income", ylabel="CDF", loc="lower right")
```



1.0.7 Part 8 Scatterplots and Correlation

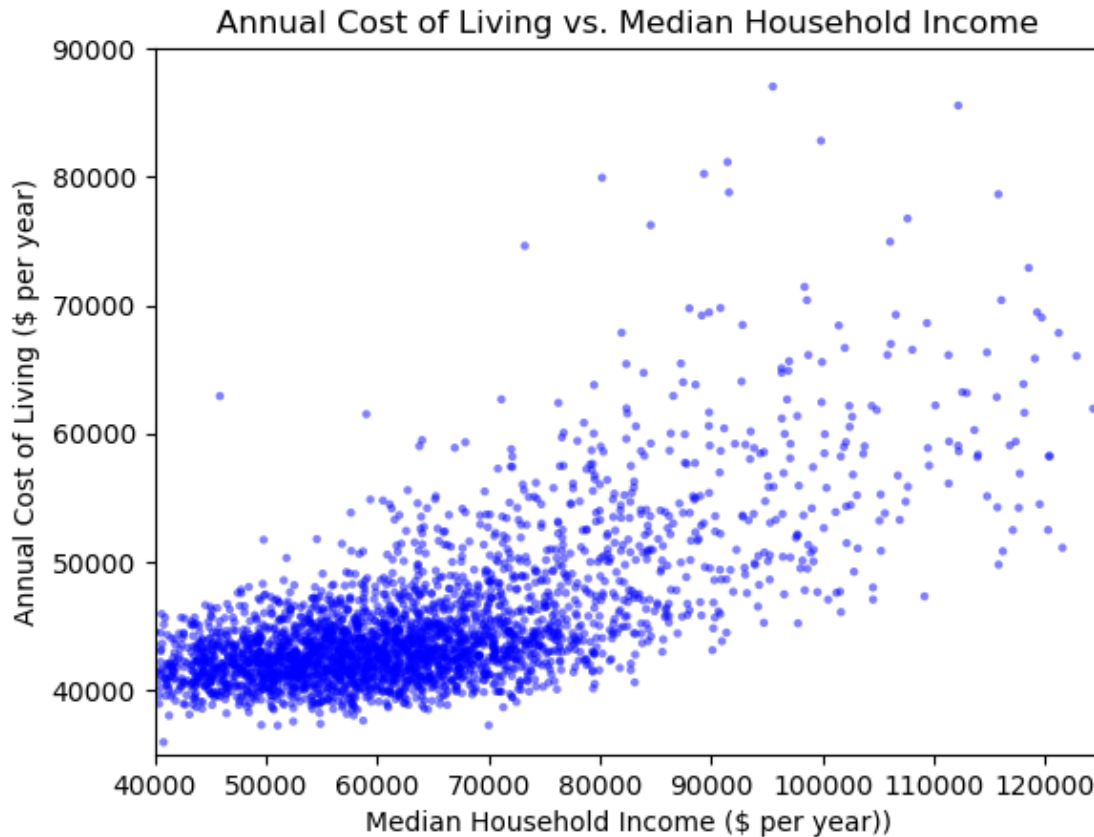
```
[357]: thinkplot.Scatter(datadf.min_wage, datadf.a_total_cost, alpha=.5, s=10)
thinkplot.Config(xlabel='Minimum Wage ($ per hour)',
                  title = "Annual Cost of Living vs. Minimum Wage",
                  ylabel='Annual Cost of Living ($ per year)',
                  xlim=[5, 18],
                  ylim=[30000, 100000],
                  legend=False)
```

```
[367]: print('Corr: ', thinkstats2.Corr(datadf.min_wage, datadf.a_total_cost))
print('SpearmanCorr: ', thinkstats2.SpearmanCorr(datadf.min_wage, datadf.
↪a_total_cost))
```

```
Corr: 0.3918515877955295
SpearmanCorr: 0.30506633739900596
```

```
[364]: thinkplot.Scatter(datadf.median_household_income, datadf.a_total_cost, alpha=.
↪5, s=10)
thinkplot.Config(xlabel='Median Household Income ($ per year)',
                  title = "Annual Cost of Living vs. Median Household_
↪Income",
                  ylabel='Annual Cost of Living ($ per year)',
                  xlim=[40000, 125000],
                  ylim=[35000, 90000],
                  legend=False)
```



```
[368]: print('Corr: ', thinkstats2.Corr(datadf.median_household_income, datadf.
        ↪a_total_cost))
print('SpearmanCorr: ', thinkstats2.SpearmanCorr(datadf.
        ↪median_household_income, datadf.a_total_cost))
```

```
Corr: 0.6967097054309237
SpearmanCorr: 0.5506802578331028
```

1.0.8 Part 8 - Hypothesis Testing

```
[374]: class CorrelationPermute(thinkstats2.HypothesisTest):
        def TestStatistic(self, data):
            xs, ys = data
            test_stat = abs(thinkstats2.Corr(xs, ys))
            return test_stat
        def RunModel(self):
            xs, ys = self.data
            xs = np.random.permutation(xs)
            return xs, ys
```

```
data = datadf.min_wage, datadf.a_total_cost
ht = CorrelationPermute(data)
pvalue = ht.PValue()
pvalue
```

[374]: 0.0

```
[375]: data = datadf.median_household_income, datadf.a_total_cost
ht = CorrelationPermute(data)
pvalue = ht.PValue()
pvalue
```

[375]: 0.0

1.0.9 Part 9 - Regression Analysis

```
[380]: import first
```

```
[381]: median_income = datadf.median_household_income
a_total_cost = datadf.a_total_cost
min_wage = datadf.min_wage
```

```
[382]: from thinkstats2 import Mean, MeanVar, Var, Std, Cov

def LeastSquares(xs, ys):
    meanx, varx = MeanVar(xs)
    meany = Mean(ys)

    slope = Cov(xs, ys, meanx, meany) / varx
    inter = meany - slope * meanx

    return inter, slope
```

```
[402]: inter, slope = LeastSquares(median_income, a_total_cost)
inter, slope
```

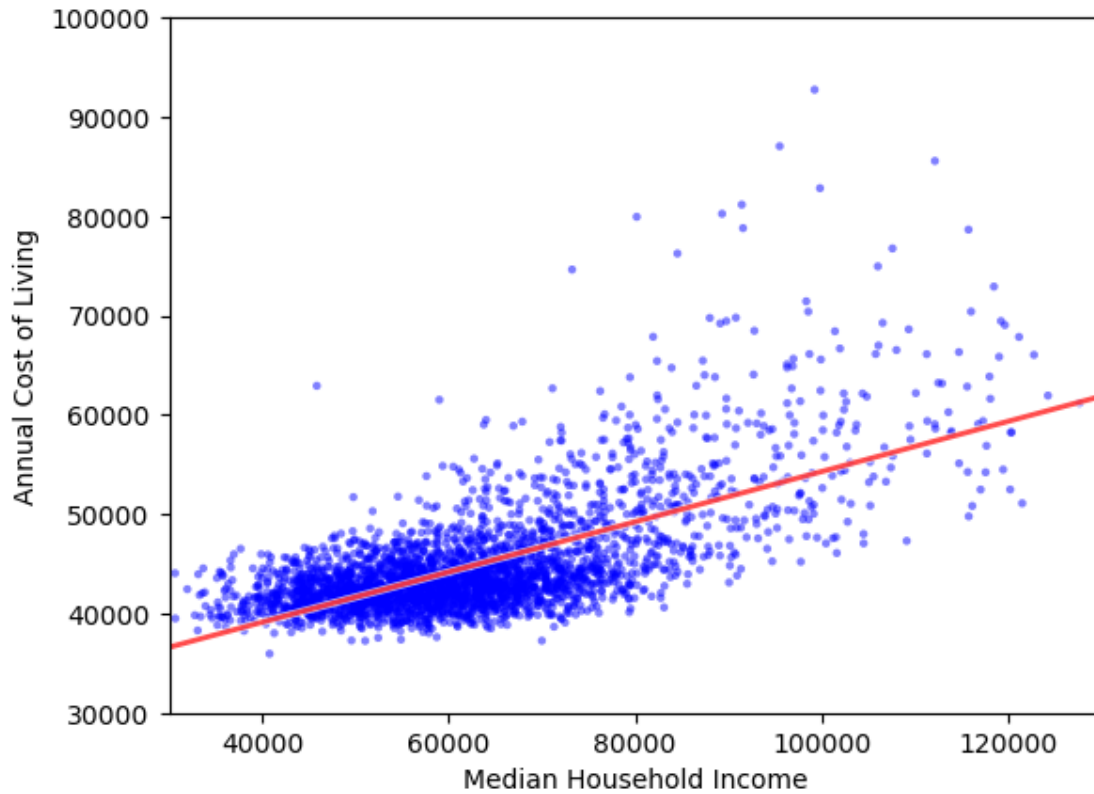
[402]: (29005.920369609048, 0.2527433731692815)

```
[403]: def FitLine(xs, inter, slope):
    fit_xs = np.sort(xs)
    fit_ys = inter + slope * fit_xs
    return fit_xs, fit_ys
```

```
[404]: fit_xs, fit_ys = FitLine(median_income, inter, slope)
```

```
[409]: thinkplot.Scatter(median_income, a_total_cost, color='blue', alpha=0.5, s=10)
thinkplot.Plot(fit_xs, fit_ys, color='white', linewidth=3)
thinkplot.Plot(fit_xs, fit_ys, color='red', linewidth=2)
```

```
thinkplot.Config(xlabel="Median Household Income",
                  ylabel="Annual Cost of Living",
                  axis=[30000, 130000, 30000, 100000],
                  legend=False)
```



[]: