DATA 602 PROJECT PROPOSAL

Covid Rates and Income

Research Question

Is there a relationship between Covid rates (cases, hospitalizations, death) and Income?

Justification

This question has relevance across industries. Understanding Covid susceptibility allows for better risk assessment and contingency planning for future public health events.

Data Sources

This analysis will focus on three datasets from two data sources:

- IRS Data by Zip Code 2019 (source: US Dept of Treasury)
- Provisional COVID-19 Death Counts in the United States by County (source: CDC)
- United States COVID-19 Community Levels by County (source: CDC)

Libraries

This project will utilize the following libraries:

- Pandas
- NumPy
- Matplotlib
- Seaborn
- Additional libraries as needed

EDA & Summary Statistics

IRS DATA

In [25]:	ir	s.head()														
Out[25]:		STATEFIPS	STATE	zipcode	agi_stub	N1	mars1	MARS2	MARS4	ELF	CPREP	 N85300	A85300	N11901	A11901	N119
	0	1	AL	0	1	778210.0	491030.0	84770.0	189600.0	712890.0	30670.0	 0.0	0.0	62720.0	51936.0	671860
	1	1	AL	0	2	525940.0	247140.0	123910.0	139860.0	481760.0	18960.0	 0.0	0.0	85860.0	122569.0	438020
	2	1	AL	0	3	285700.0	105140.0	128140.0	44560.0	260570.0	10670.0	 0.0	0.0	73980.0	154932.0	212040
	3	1	AL	0	4	179070.0	38820.0	123110.0	13740.0	164300.0	5020.0	 0.0	0.0	51330.0	139065.0	126850
	4	1	AL	0	5	257010.0	28180.0	216740.0	7150.0	236850.0	8400.0	 90.0	141.0	104290.0	460071.0	152790

5 rows × 152 columns

In [26]:	irs.d	escribe()									
Out[26]:		STATEFIPS	zipcode	agi_stub	N1	mars1	MARS2	MARS4	ELF	CPREP	
	count	166159.000000	166159.000000	166159.000000	1.661590e+05	1.661590e+05	1.661590e+05	166159.00000	1.661590e+05	166159.000000	1.6615
	mean	29.666885	48859.485553	3.499949	1.860508e+03	9.127834e+02	6.478571e+02	258.03676	1.689549e+03	75.771821	9.5219
	std	15.121486	27167.679271	1.707871	3.722335e+04	2.224999e+04	1.200080e+04	6336.06430	3.347049e+04	1866.726495	1.9368
	min	1.000000	0.000000	1.000000	0.000000e+00	0.000000e+00	0.000000e+00	0.00000	0.000000e+00	0.000000	0.0000
	25%	18.000000	27020.000000	2.000000	7.000000e+01	0.000000e+00	4.000000e+01	0.00000	7.000000e+01	0.000000	5.0000
	50%	29.000000	48843.000000	3.000000	2.600000e+02	8.000000e+01	1.100000e+02	30.00000	2.400000e+02	0.000000	1.5000
	75%	42.000000	70652.500000	5.000000	1.080000e+03	3.900000e+02	3.800000e+02	100.00000	9.900000e+02	40.000000	5.6000
	max	56.000000	99999.000000	6.000000	5.506120e+06	4.069770e+06	1.818210e+06	945490.00000	4.827070e+06	338290.000000	3.0225
	8 rows	x 151 columns									

8 rows × 151 columns

<class 'pandas.core.frame.DataFrame'>
RangeIndex: 166159 entries, 0 to 166158
Columns: 152 entries, STATEFIPS to A12000
dtypes: float64(148), int64(3), object(1)
memory usage: 192.7+ MB

The key for this data has been loaded into the project GitHub repository. The original IRS dataset contains over 166k observations across 152 columns, and must be read in locally as it is 192.7mb in size. The data will transformed to its necessary components for this analysis, written to .csv, and uploaded to the project GitHub repository.

CDC Community Levels By County

In [24]: cdc_comm.head() Out[24]: state county_population health_service_area_number health_service_area health_service_area_population covid_inpatic county county_fips Lincoln Marathon (Wausau). 0 55069 Wisconsin 27593.0 282 291401.0 WI - Wood, WI County Sheboygan Manitowoc 55071 Wisconsin 78981.0 355 (Sheboygan), WI -244410.0 County Manitowoc, WI Marathon Marathon (Wausau) 55073 Wisconsin 135692.0 282 291401.0 WI - Wood, WI County La Crosse (La Monroe 55081 Wisconsin 46253.0 290 Crosse), WI -257027.0 County Monroe, WI Portage 55097 Wisconsin 70772.0 400 70772.0 Portage, WI County In [28]: cdc_comm.describe() county_fips county_population health_service_area_number health_service_area_population covid_inpatient_bed_utilization covid_hospital_adm Out[28]: 112836.000000 112648.00000 count 112836.00000 1.128350e+05 1.128290e+05 mean 31438.02789 1.029200e+05 400.462033 5.808604e+05 3.25508 16331.50567 3.293638e+05 9.952625e+05 2.66225 243.444960 std 0.00000 1001.00000 8.600000e+01 1.000000 2.274000e+03 min 25% 19033.00000 1.113100e+04 186.000000 9.021200e+04 1.30000 30027.00000 2.611800e+04 409.000000 2.80000 50% 2.249140e+05 75% 46111.00000 6.721500e+04 587.000000 5.545570e+05 4.50000 78000.00000 1.003911e+07 905.000000 1.321480e+07 36.00000 In [29]: cdc_comm.info() <class 'pandas.core.frame.DataFrame'> RangeIndex: 112836 entries, 0 to 112835 Data columns (total 12 columns): Non-Null Count 0 county 112836 non-null 112836 non-null county_fips int64 state 112836 non-null object 112835 non-null county_population float64 112836 non-null health_service_area_number int64 112836 non-null health_service_area object 112829 non-null health_service_area_population float64 covid_inpatient_bed_utilization 112648 non-null float64 112778 non-null covid_hospital_admissions_per_100k float64 covid_cases_per_100k 112836 non-null float64 covid-19_community_level 112782 non-null object date_updated 112836 non-null object

This dataset contains over 112k observations across 12 columns containing information about Covid-19 cases and hospitalizations.

CDC Provisional Death Counts By County

dtypes: float64(5), int64(2), object(5)

memory usage: 10.3+ MB

In [22]: cdc_prov.head()

Out[22]:		Date as of	Start Date	End Date	State	County nan	FIPS ne County Code	Urban Rural Code	Deaths involving COVID-19	Deaths from All Causes	Footnote
	0	10/19/2022	01/01/2020	10/15/2022	AK	Aleutians Ea Borouç		Noncore	NaN	22.0	One or more data cells have counts between 1-9
	1	10/19/2022	01/01/2020	10/15/2022	AK	Anchoraç Municipali		Medium metro	734.0	7081.0	NaN
	2	10/19/2022	01/01/2020	10/15/2022	AK	Bethel Censo Are		Noncore	39.0	317.0	NaN
	3	10/19/2022	01/01/2020	10/15/2022	AK	Denali Borouç	jh 2068	Noncore	NaN	24.0	One or more data cells have counts between 1-9
	4	10/19/2022	01/01/2020	10/15/2022	AK	Dillingha Census Are		Noncore	NaN	96.0	One or more data cells have counts between 1-9
In [38]:	cdo	_prov.de	scribe()								
Out (38):		FIPS Co	ounty Code	Deaths invo	lving CO	VID-19 Death	s from All Causes	S			
	count 3085.000000		2706.000000			3084.000000)				
	me	an 30	357.156240		391.	218404	3027.57328	1			
	s	td 15′	162.540083		1179.	109564	8527.317813	3			
	m	nin 10	001.000000		10.0	000000	14.000000)			
	25	25 % 18175.000000		29.000000		000000	304.000000)			
	50	50% 29147.000000		75.000000		000000	717.500000)			
	75	75 % 45075.000000			297.500000		2120.500000)			
	m	max 56045.000000			31094.000000		223502.000000)			
In [31]:	cdo	_prov.in	fo()								
	Ran	geIndex:	das.core.f 3085 entr s (total 1	ies, 0 to	3084	ull Count)†vne				
	0 1	<pre>0 Date as of 1 Start Date</pre>		3085 non-null 3085 non-null			object object				
	2	2 End Date		3085 non-null		non-null	object				
	3 4	3 State 4 County name			3085 non-null 3085 non-null		object object				
	5	, , , , ,			3085 non-null		int64				
	6 7		Rural Code				object				
	8		involving from All				float64 float64				
	9	Footno	te		379 n	on-null	object				
	d±v	nes floa	at64(2). i	n+64(1)	hiect(7)					

 $This \ dataset \ contains \ 3085 \ observations \ across \ 10 \ columns, \ containing \ total \ Covid \ deaths \ by \ State \ and \ County.$

Combining the Datasets

memory usage: 241.1+ KB

dtypes: float64(2), int64(1), object(7)

These datasets will be transformed and combined to create a master dataframe containing Zip code, Income, and Covid rate information.