REPORT OF COVID19 DATA ANALYSIS PROJECT STAGE ONE (Data and Project Understanding)

1. Group One Members

The names of the group members are:

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2. Describing the Dataset and Data Type - Variable Dictionary

The description of the various data sets are presented in Table I, Table II, Table III, and Table IV. Table I contains the description of the Population dataset. It comprises the following fields: CountyFIPS, County Name, State, and Population. Table II represents the description of the COVID19 cases dataset. The dataset comprises the number of infections with covid 19 virus. The table has the following fields: CountyFIPS, County Name, State, StateFIPS, and Dates from 1/22/2020 till 7/23/2023

In a similar vein, Table III represents the variable description of COVID19 death dataset. The Covid19 deaths dataset comprises the number of deaths arising from covid 19 infections. The table has the following fields: CountyFIPS, County Name, State, StateFIPS, and Dates from 1/22/2020 till 7/23/2023.

Table IV indicates the description of ACS Social, Economic, and Housing

Table I: Description of Population Dataset

Name	Definition	Data Type	Possible Values	Required?
CountyFIPS	Unique County ID (unknown counties have an ID of 0)	Integer	0,1001,1003,10 41,1027	Yes
County Name	Name of County	Text	Baldwin County, Blount County, Conecuh County, Coffee County, Dallas County	Yes
State	Name of State	Text	AL, CA, CO, GA, FL	Yes
Population	Number of population.	Integer	292256,156714, 16116, 132235, 149910	Yes

Table II: Description of COVID19 Cases Dataset

Name	Definition	Data Type	Possible Values	Required?
CountyFIPS	Unique County ID (unknown counties have an ID of 0)	Integer	1001, 1003, 1011, 1013, 56045	YES
County Name	Name of County	Text	Crook County, Vilas County, Weston County, Goshen County, Park County	YES
State	Name of State	Text	TX, PA, WI, TN, SD	YES
StateFIPS	FIPS Code for a State	Text	01, 02, 03, 04, 56	YES
Date	Number of cases per day from 1/22/2020 to 7/23/2023	Integer	0, 1, 2, 3, 4,5	YES

 Table III:
 Description of COVID19 Deaths Dataset

Name	Definition	Data Type	Possible Values	Required?
countyFIPS	Unique County ID (unknown counties have an ID of 0)	Integer	0, 1001, 1003, 51047, 56045	Yes
County Name	Name of County	Text	Wake County, Orange County	Yes
State	Name of State	Text	TX, PA, WI, TN, SD	Yes
StateFIPS	State ID	Text	01, 02, 10, 12, 56	Yes
Deaths (all date columns)	Number of Death per day from 1/22/2020 to 7/23/2023	Integer	0, 26, 1310, 6472	Yes

Table IV: Description of ACS Social, Economic, and Housing - Kevin Hayes

Name	Definition	Data Type	Possible Values	Required?
Geography	A unique number associated with each county. This will not be used because I am only looking at the US, so the names of states and counties are sufficient for identification.	Text	0500000US01003 0500000US01015 0500000US01043 0500000US01049 0500000US01051 0500000US01055 0500000US01069	Yes.
Geographic Area Name	The name of the county, followed by a comma and the full name of its state.	Text	Baldwin County, Alabama Calhoun County, Alabama Cullman County, Alabama DeKalb County, Alabama Elmore County, Alabama Etowah County, Alabama	Yes
Estimate	The estimated value of a given label	Number (Either Integer or Float. Depends on row)	327,167,439, 97.0, 38.2, 52,423,114, 79.9	Yes
Margin of Error	Absolute expected deviation from a given estimate.	Number (Either Integer or Float. Depends on row)	27,812, 27,815, 0.1, 0.2, 34,697	No
Percent	Percentage of a given sublabel. (Note: base level labels have either their initial value or (X) for this. This will be fixed in data cleaning)	Float	49.2, 50.8, 6.0, 6.1, 6.5,	No
Percent	relative expected	float	0.4	No

Margin of Error	deviation from a given estimate (Note: labels where this does not apply have either their initial value or (X) for this. This will be fixed in data cleaning)		0.3 0.6 0.9 0.5	
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- We can join this to our main dataframe by taking a counties state and name and using that to generate its Geographic Area Name, which we can use as a foreign key to map between the two tables.
- All of these are followed by !!<label>!!<sublabel> to denote where the value applies to.
 These labels are SEX AND AGE, RACE, Race alone or in combination with one or more
 other races, HISPANIC OR LATINO AND RACE, and CITIZEN, VOTING AGE
 POPULATION. The sublabels refer to various races or qualifiers to the main label. In
 practice I plan to break up each label into its own dataframe.
- This enrichment data can help analyze the spread of COVID-19 by allowing us to see if areas with people of given ages or races were more or less likely to have people infected with or killed by COVID-19.
- Potential hypothesis questions:
 - Is there a correlation between mean age in a county and the number of COVID-19 deaths in said county?
 - Is there a correlation between mean age in a county and the number of COVID-19 infections in said county?
 - Is there a correlation between the percentage of people of age 85 or over in a county and the number of COVID-19 deaths in said county?
 - Is there a correlation between the percentage of people in a county that are hispanic or latino and deaths from COVID-19?

Table V: Description of Employment Variables

Name	Definition	Data Type	Possible Values	Required?
area_fips	County FIPS	Text	05123, 01041, 56045	Yes
own_code	1-character Ownership code	Text	0, 1, 2, 3, 4, 5, 8, 9	No
industry_cod e	NAICS industry code	Text	49111, 92, 522298	No
agglvl_code	Code for aggregation level	Text	74, 77, 78	No
size_code	Establishment size code	Text	0	No
year	Census year	Text	2020, 2021, 2022	Yes
qtr	Fiscal quarter (always A for annual)	Text	А	No
disclosure_c ode	1-character disclosure code	Text	""(blank), N	No
area_title	County name (based on area_fips)	Text	St. Francis County, Arkansas; Crenshaw County, Alabama; Weston County, Wyoming	No
own_title	Type of ownership (based on own_code)	Text	Federal Government, State Government, Private	No
industry_title	Industry title (based on industry_code)	Text	NAICS 49111 Postal service; NAICS 92 Public administration; NAICS 522298 All other nondepository credit intermediation	No
agglvl_title	Title of level of aggregation (based on agglvl_code)	Text	County, NAICS Sector by ownership sector; County, NAICS	No

			5-digit by ownership sector; County, NAICS 6-digit by ownership sector	
size_title	Size title (based on size_code)	Text	All establishment sizes	No
annual_avg_ estabs	Annual average of quarterly establishment counts for a given year*	Numeric	385, 109, 254	No
annual_avg_ emplvl	Annual average of monthly employment levels for a given year*	Numeric	4756, 1231, 2149	Yes
total_annual _wages	Total wages for the fiscal year	Numeric	217856056, 82305381, 101598481	Yes
taxable_ann ual_wages	Sum of the four quarterly total taxable wage totals for a given year*	Numeric	36984510, 10206971, 39354459	Yes
annual_contr ibutions	Sum of the four quarterly contribution totals for a given year*	Numeric	567653, 100344, 946941	No
annual_avg_ wkly_wage	Average weekly wage based on the 12-monthly employment levels and total annual wage levels*	Numeric	881, 1286, 909	Yes
avg_annual_ pay	Average annual pay based on employment and wage levels for a given year*	Numeric	45809, 66865, 47270	Yes
lq_disclosure _code	1-character location-quotient disclosure code*	Text	""(blank), N	No

lq_annual_a vg_estabs	Location quotient of annual average establishment count relative to the U.S. (Rounded to hundredths place)*	Numeric	1	No
lq_annual_a vg_emplvl	Location quotient of annual average employment relative to the U.S. (Rounded to hundredths place)*	Numeric	1	No
lq_total_ann ual_wages	Location quotient of total annual wages relative to the U.S. (Rounded to hundredths place)*	Numeric	1	No
lq_taxable_a nnual_wages	Location quotient of taxable annual wages relative to the U.S. (Rounded to hundredths place)*	Numeric	1	No
lq_annual_c ontributions	Location quotient of total annual contributions relative to the U.S. (Rounded to hundredths place)*	Numeric	1	No
lq_annual_a vg_wkly_wa ge	Location quotient of annual average weekly wage relative to the U.S. (Rounded to hundredths place)*	Numeric	1	No
lq_avg_annu al_pay	Location quotient of annual average pay relative to the U.S. (Rounded to hundredths place)*	Numeric	1	No
oty_disclosur e_code	1-character over-the-year disclosure code (either ' '(blank) or 'N' not	Text	""(blank), N	No

	disclosed)*			
oty_annual_ avg_estabs_ chg	Over-the-year change in annual average establishments for a given year*	Numeric	15, -8	
oty_annual_ avg_estabs_ pct_chg	Over-the-year percent change in annual average establishments for a given year (Rounded to the tenths place)*	Numeric	4.1, -6.8, -3.1	No
oty_annual_ avg_emplvl_ chg	Over-the-year change in annual average employment for a given year*	Numeric	103, -58, -17	No
oty_annual_ avg_emplvl_ pct_chg	Over-the-year percent change in annual average employment for a given year (Rounded to the tenths place)*	Numeric	2.2, -4.5, -0.8	Yes
oty_total_an nual_wages_ chg	Over-the-year change in the total annual wages for a given year*	Numeric	3927403, 3743942, 2850032	No
oty_total_an nual_wages_ pct_chg	Over-the-year percent change in total annual wages for a given year (Rounded to the tenths place)*	Numeric	1.8, 4.8, 2.9	No
oty_taxable_ annual_wag es_chg	Over-the-year change in taxable annual wages for a given year*	Numeric	66827, -603070, 3088805	No
oty_taxable_ annual_wag es_pct_chg	Over-the-year percent change in taxable annual wages for a given year (Rounded to the tenths place)*	Numeric	0.2, -5.6, 8.5	No

oty_annual_ contributions _chg	Over-the-year change in annual contributions for a given year*	Numeric	19847, -16026, 143365	No
oty_annual_ contributions _pct_chg	Over-the-year percent change in annual contributions for a given year (Rounded to the tenths place)*	Numeric	3.6, -13.8, 17.8	No
oty_annual_ avg_wkly_w age_chg	Over-the-year change in annual average weekly wage for a given year*	Numeric	-3, 114, 32	No
oty_annual_ avg_wkly_w age_pct_chg	Over-the-year percent change in annual average weekly wage for a given year (Rounded to the tenths place)*	Numeric	-0.3, 9.7, 3.6	Yes
oty_avg_ann ual_pay_chg	Over-the-year change in average annual pay for a given year*	Numeric	-165, 5933, 1676	No
oty_avg_ann ual_pay_pct _chg	Over-the-year percent change in average annual pay for a given year (Rounded to the tenths place)*	Numeric	-0.4, 9.7, 3.7	Yes

^{*} This definition is taken from the U.S. Department of Labor Statistics QCEW annual field layout

- We can use the area FIPS variable to merge data with the COVID dataset.
- We could look at the correlation between Employment levels and wages and COVID infection and death rates
- Hypothesis questions:
 - Is there a correlation between employment levels and COVID infection/death rates?
 - o Is there a correlation between wages and COVID infection/death rates?

Table VI: Description of Presidential Dataset

Name	Definition	Data Type	Possible Values	Required?
State	Name of State	Text	Florida, Geaogia Hawaii, Idaho, Illinois	Yes
County	Name of county	Text	Kent County, Ward 2, Duval County, Hardee County, Highlands County	Yes
Candidate	Name of Contestant	Text	Joe Biden, Donald Trump, Jo Jorgensen, Howie Hawkins	Yes
Party	Name of Political Party	Text	DEM, REP, LIB, GRN	Yes
Total-Vote	Vote Count	Integer	420, 1282, 1003, 259, 56682	Yes
Won	Winner in the Polls	Boolean	True, False	Yes

- Presidential dataset is the table containing the outcomes of the 2020 presidential poll in the USA. This dataset provides detailed information on the 2020 USA presidential election results. The table has six fields namely State, County, Candidate, Political Party, Total votes per candidate, and the winner per county.
- To join the two tables, a common variable that exists in both tables is essential. In this case, the common variables are "State" and "County" since they are present in both the presidential results table and the COVID-19 cases table. So, the variable that can be used to join the two tables would be a combination of "State" and "County". When performing the join operation, the rows in the presidential results table will be matched with the corresponding rows in the COVID-19 cases table based on the values in these two columns.
- Hypothesis

- Is there a correlation between the political affiliation and the spread of Covid-19 cases/deaths
- Is there a correlation between the victory of a candidate and the spread of the covid-19 cases/deaths
- Is there a correlation between total votes cast and the COVID 19 cases/deaths