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### 1. Preparing Your Dataset for Cleaning

Before you start data cleaning on the COVID Deaths, COVID Cases, and Population datasets, ensure that the following files have been uploaded to your Jupyter Notebook environment:

1. covid\_deaths\_usafacts.csv
2. covid\_confirmed\_usafacts.csv
3. covid\_county\_population\_usafacts.csv

You can reference these files by their names directly in your notebook.

**If the Files Are Not Uploaded** If you have not uploaded the files yet, please follow these steps:

1. Download the following files:

- covid\_deaths\_usafacts.csv
- covid\_confirmed\_usafacts.csv
- covid\_county\_population\_usafacts.csv

1. Upload the files to Jupyter Notebook:

- Click on the 'Upload' button in the Jupyter Notebook interface.
- Select the dataset files from your local system. Once the files are uploaded, you can reference them by their names in your notebook.

### 2. Data Cleaning: COVID Deaths, COVID Cases, and Population Datasets

As part of the data cleaning process for the COVID Deaths , COVID Cases and Population dataset, the following steps were implemented:

1. Removing Rows with countyFIPS Value of 0

- **Action:** Rows where the countyFIPS column value is 0 were excluded from all three datasets.

- **Reason:** Rows with a `countyFIPS` value of 0 were removed for the following reasons:
  - In the `Population` dataset, counties with a `countyFIPS` value of 0 have a population of 0, which is not meaningful for COVID-19 analysis.
  - It is logical to exclude these rows since counties with zero population would not have COVID-19 cases or deaths.
  - Additionally, the `Enrichment` dataset does not contain any `countyFIPS` codes with a value of 0. Since the datasets need to be merged, it is essential to ensure consistency across all datasets by removing these rows.
  - A countyFIPS code of 0 is not a valid county code in the FIPS standard. As FIPS code for counties in the United States is a 5-digit number where: The first two digits represent the state code and last three digits represent the county code within the state.
- **Result:** This step reduced the number of rows as follows:
  - **COVID Deaths and COVID Cases Datasets:** From 3,193 to 3,142 rows.
  - **Population Dataset:** From 3,195 to 3,144 rows.

### COVID Deaths:

```
In [107... import pandas as pd
deaths_with_zeroCountyCode=pd.read_csv('covid_deaths_usafacts.csv')
#print("(Rows, Columns)=",deaths_with_zeroCountyCode.shape)
deaths=deaths_with_zeroCountyCode[deaths_with_zeroCountyCode['countyFIPS'] != 0]
print("(Rows, Columns)=",deaths.shape)

(Rows, Columns)= (3142, 1269)
```

### COVID Cases:

```
In [108... cases_with_zeroCountyCode=pd.read_csv('covid_confirmed_usafacts.csv')
#print("(Rows, Columns)=",cases_with_zeroCountyCode.shape)
cases=cases_with_zeroCountyCode[cases_with_zeroCountyCode['countyFIPS'] != 0]
print("(Rows, Columns)=",cases.shape)

(Rows, Columns)= (3142, 1269)
```

### Population:

```
In [109... population_with_zeroCountyCode=pd.read_csv('covid_county_population_usafacts.csv')
#print(population_with_zeroCountyCode.shape)
population=population_with_zeroCountyCode[population_with_zeroCountyCode['countyFIPS'] != 0]
print("(Rows, Columns)=",population.shape)

(Rows, Columns)= (3144, 4)
```

## 3. Merging of COVID Deaths, COVID Cases and Population

- **Covid Deaths dataset name:** `deaths`
- **Covid Cases dataset name:** `cases`
- **Population dataset name:** `population`
- **After merging this three dataset name:** `super_covid19_dataframe.csv`

After analyzing the three datasets, we perform a merge on the `countyFIPS` column using an `inner` join.

**Reason for Inner Join:** An inner join was chosen based on the following analysis:

- The `population` dataset contained two rows with `countyFIPS` codes 2270 and 6000, where the population value was 0. These rows are not present in the `deaths` and `cases` datasets.
- Additionally, these `countyFIPS` codes are not included in the enrichment dataset.

Using an inner join ensures that only rows with matching `countyFIPS` codes across all three datasets are included. The exclusion of these two rows, which have no relevant data in the other datasets, will not significantly impact the merged result.

By using an inner join, we effectively consolidate the datasets while maintaining data integrity and relevance.

```
In [110... super_cases_population=pd.merge(cases,population[['countyFIPS','population']],on='countyFIPS',how='inner')
print("super_cases_population:(Rows, Columns)=",super_cases_population.shape)
super_deaths_population=pd.merge(deaths,population[['countyFIPS','population']],on='countyFIPS',how='inner')
print("super_deaths_population:(Rows, Columns)=",super_deaths_population.shape)

# Merge the 'super_cases_population' DataFrame with the 'super_deaths_population' DataFrame
# Drop the 'County Name', 'State', 'StateFIPS', and 'population' columns from 'super_cases_population' and 'super_deaths_population'
# and use suffixes '_cases' and '_deaths' to differentiate between the case and death data
super_covid19_dataframe=pd.merge(super_cases_population,super_deaths_population.drop(['County Name','State','StateFIPS','population'],axis=1),on='countyFIPS',how='inner')
print("super_covid19_dataframe: (Rows, Columns)=",super_covid19_dataframe.shape)

# Save the final DataFrame to a CSV file without the index column
super_covid19_dataframe.to_csv('super_covid19_dataframe.csv', index=False)

super_cases_population:(Rows, Columns)= (3142, 1270)
super_deaths_population:(Rows, Columns)= (3142, 1270)
super_covid19_dataframe: (Rows, Columns)= (3142, 2535)
```

## 4. Upload Merge Dataset

To upload dataset file:

1. First download the `super_covid19_dataframe.csv` file from path: Group-4/super\_covid19\_dataframe.csv
2. After downloading, click on the 'Upload' button in the jupyter Notebook interface.
3. Select the dataset file from your local system.
4. Once uploaded, you can use the file in your notebook by referencing its name.

Below code provides instructions for reading and inspecting a dataset of COVID-19 data using the pandas library.

```
In [111... import pandas as pd

super_covid19=pd.read_csv('super_covid19_dataframe.csv')
print("(Rows, Columns)=",super_covid19.shape)

(Rows, Columns)= (3142, 2535)
```

The below Python code is used to display the first few rows of a Data

In [112...

```
print(super_covid19.head())
```

	countyFIPS	County Name	State	StateFIPS	2020-01-22_cases	\
0	1001	Autauga County	AL	1	0	
1	1003	Baldwin County	AL	1	0	
2	1005	Barbour County	AL	1	0	
3	1007	Bibb County	AL	1	0	
4	1009	Blount County	AL	1	0	

	2020-01-23_cases	2020-01-24_cases	2020-01-25_cases	2020-01-26_cases	\
0	0	0	0	0	
1	0	0	0	0	
2	0	0	0	0	
3	0	0	0	0	
4	0	0	0	0	

	2020-01-27_cases	...	2023-07-14_deaths	2023-07-15_deaths	\
0	0	...	235	235	
1	0	...	731	731	
2	0	...	104	104	
3	0	...	111	111	
4	0	...	261	261	

	2023-07-16_deaths	2023-07-17_deaths	2023-07-18_deaths	2023-07-19_deaths	\
0	235	235	235	235	
1	731	731	731	731	
2	104	104	104	104	
3	111	111	111	111	
4	261	261	261	261	

	2023-07-20_deaths	2023-07-21_deaths	2023-07-22_deaths	2023-07-23_deaths
0	235	235	235	235
1	731	731	731	731
2	104	104	104	104
3	111	111	111	111
4	261	261	261	261

[5 rows x 2535 columns]

## 5. Data for the year 2020

After analyzing the code we need to do two task for required data. **Filter Columns for 2020 Data**: As mentioned in project discription, we need to focus on the year 2020 data so, we take the columns which starts with 2020

In [113...

```
columns_2020 = [col for col in super_covid19.columns if col.startswith('2020')] # Filter columns for 2020
superdata_2020=super_covid19[columns_2020]
additional_columns = super_covid19[['countyFIPS', 'County Name', 'State', 'StateFIPS']]
superdata_2020=pd.concat([additional_columns,superdata_2020], axis=1) # Concatenate additional columns
print("(Rows, Columns)=",superdata_2020.shape)
```

(Rows, Columns)= (3142, 695)

## 6. Calculate COVID-19 data trends for last week

Selected state information:

- **State name:** Arizona

- **State:** AZ
- **StateFIPS:** 4
- **Total county count:** 15

The below code snippet extracts and visualizes COVID-19 case data for the state of Arizona. Specifically, it focuses on the last week of data from the year 2020. For each county within Arizona, the code generates a line plot showing the trend of COVID-19 cases over this week.

In [114...

```
import matplotlib.pyplot as plt

state_data = superdata_2020[superdata_2020['StateFIPS'] == 4] # Filter the data for Arizona
# print(state_data.shape)
date_columns = [col for col in state_data.columns if col.endswith('_cases')] # Extract date columns
date_columns.sort()

filter_last_week_dates = date_columns[-7:] # Select columns for the last week

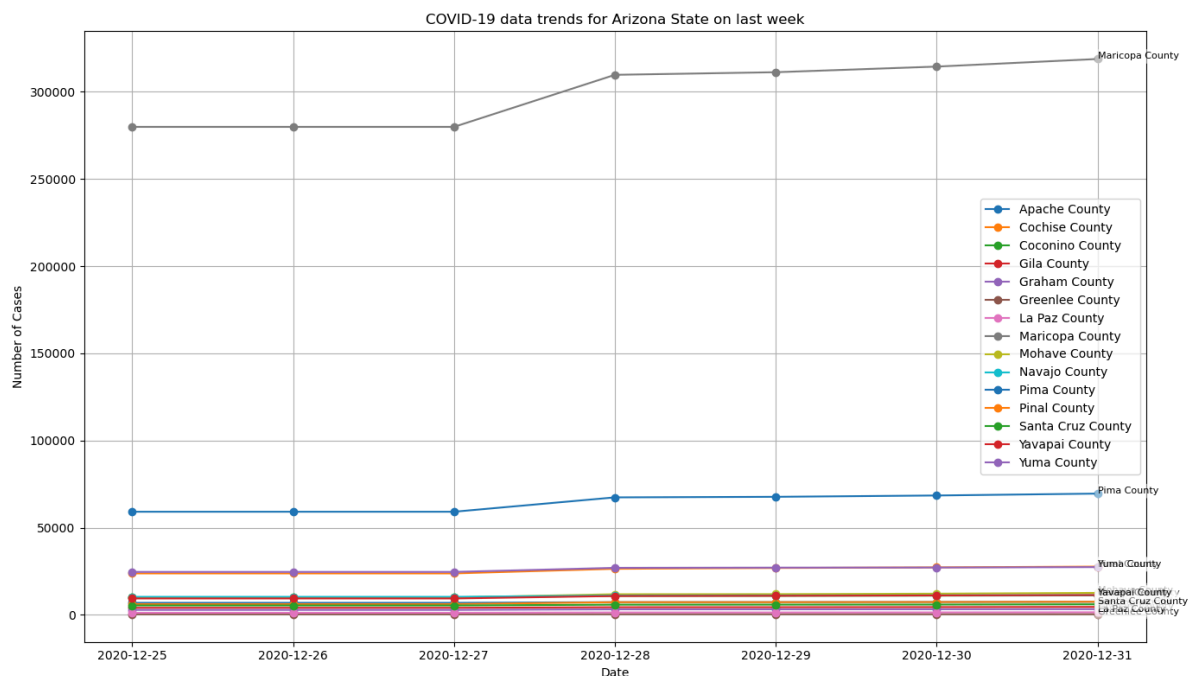
# Extract and convert dates from column names
last_week_dates = [col.split('_')[0] for col in filter_last_week_dates]
last_week_dates = pd.to_datetime(last_week_dates)

#print("Dates for the last week:", last_week_dates)
plt.figure(figsize=(14, 8))
for index, row in state_data.iterrows():
    county_name = row['County Name']
    case_counts = row[filter_last_week_dates].values # Get case counts for the last week
    plt.plot(last_week_dates, case_counts, marker='o', linestyle='-', label=county_name)

    # Annotate the last point of each line with the county name
    plt.text(last_week_dates[-1], case_counts[-1], county_name, fontsize=8, color='red',
             bbox=dict(facecolor='white', alpha=0.5, edgecolor='none', boxstyle='round'))

plt.title('COVID-19 data trends for Arizona State on last week')
plt.xlabel('Date')
plt.ylabel('Number of Cases')
plt.legend()
plt.grid(True)
plt.tight_layout() # Adjust layout to prevent clipping of labels

# Show the plot
plt.show()
```



By analyzing the graph we can see that cases are increasing, decreasing, or stable for each county.

- Apache County: **Stable**
- Cochise County: **Stable**
- Coconino County: **Stable**
- Gila County: **Stable**
- Graham County: **Stable**
- Greenlee County: **Stable**
- La Paz County: **Stable**
- Maricopa County: **Increasing**
- Mohave County: **Stable**
- Navajo County: **Stable**
- Pima County: **Increasing**
- Pinal County: **Increasing**
- Santa Cruz County: **Stable**
- Yavapai County: **Stable**
- Yuma County: **Increasing**

## 7. Enrichment dataset: ACS Demographic and Housing Estimates

To upload dataset file:

1. First download the ACS Demographic and Housing Estimates.csv file from path: Group-4/ACS Demographic and Housing Estimates.csv
2. After downloading, click on the 'Upload' button in the jupyter Notebook interface.
3. Select the dataset file from your local system.
4. Once uploaded, you can use the file in your notebook by referencing its name.

We can also download it from link: [https://data.census.gov/table/ACSDP5Y2020.DP05?q=dp&g=010XX00US\\$0500000](https://data.census.gov/table/ACSDP5Y2020.DP05?q=dp&g=010XX00US$0500000)

Below code provides instructions for reading and display a dataset of COVID-19 data using the pandas library.

```
In [115... ACS=pd.read_csv('ACS Demographic and Housing Estimates.csv',skiprows=1)
print("(Rows, Columns)=",ACS.shape)
print(ACS.head())
print(ACS.dtypes) #Data type of all columns
```



(Rows, Columns)= (3221, 359)

	Geography	Geographic Area Name	\
0	0500000US01001	Autauga County, Alabama	
1	0500000US01003	Baldwin County, Alabama	
2	0500000US01005	Barbour County, Alabama	
3	0500000US01007	Bibb County, Alabama	
4	0500000US01009	Blount County, Alabama	

	Estimate!!SEX AND AGE!!Total population	\
0	55639	
1	218289	
2	25026	
3	22374	
4	57755	

	Margin of Error!!SEX AND AGE!!Total population	\
0	*****	
1	*****	
2	*****	
3	*****	
4	*****	

	Estimate!!SEX AND AGE!!Total population!!Male	\
0	27052	
1	105889	
2	13156	
3	12022	
4	28677	

	Margin of Error!!SEX AND AGE!!Total population!!Male	\
0	167	
1	253	
2	86	
3	170	
4	153	

	Estimate!!SEX AND AGE!!Total population!!Female	\
0	28587	
1	112400	
2	11870	
3	10352	
4	29078	

	Margin of Error!!SEX AND AGE!!Total population!!Female	\
0	167	
1	253	
2	86	
3	170	
4	153	

	Estimate!!SEX AND AGE!!Total population!!Sex ratio (males per 100 females)	\
0	94.6	
1	94.2	
2	110.8	
3	116.1	
4	98.6	

	Margin of Error!!SEX AND AGE!!Total population!!Sex ratio (males per 100 female s)	\
0	1.1	
1	0.4	
2	1.5	
3	3.6	
4	1.0	

... \

0	...
1	...
2	...
3	...
4	...

Percent Margin of Error!!HISPANIC OR LATINO AND RACE!!Total population!!Not Hispanic or Latino!!Two or more races!!Two races excluding Some other race, and Three or more races \

0	0.7
1	0.4
2	0.6
3	0.3
4	0.3

Percent!!Total housing units	Percent Margin of Error!!Total housing units \
0	(X)
1	(X)
2	(X)
3	(X)
4	(X)

Percent!!CITIZEN, VOTING AGE POPULATION!!Citizen, 18 and over population \

0	41959
1	167986
2	19468
3	17583
4	42978

Percent Margin of Error!!CITIZEN, VOTING AGE POPULATION!!Citizen, 18 and over population \

0	(X)
1	(X)
2	(X)
3	(X)
4	(X)

Percent!!CITIZEN, VOTING AGE POPULATION!!Citizen, 18 and over population!!Male \

0	47.6
1	47.8
2	53.2
3	53.6
4	48.9

Percent Margin of Error!!CITIZEN, VOTING AGE POPULATION!!Citizen, 18 and over population!!Male \

0	0.4
1	0.2
2	0.2
3	0.4
4	0.3

Percent!!CITIZEN, VOTING AGE POPULATION!!Citizen, 18 and over population!!Female \

0	52.4
1	52.2
2	46.8
3	46.4
4	51.1

Percent Margin of Error!!CITIZEN, VOTING AGE POPULATION!!Citizen, 18 and over p

```

population!!Female \
0                      0.4
1                      0.2
2                      0.2
3                      0.4
4                      0.3

Unnamed: 358
0          NaN
1          NaN
2          NaN
3          NaN
4          NaN

[5 rows x 359 columns]
Geography
object
Geographic Area Name
object
Estimate!!SEX AND AGE!!Total population
int64
Margin of Error!!SEX AND AGE!!Total population
object
Estimate!!SEX AND AGE!!Total population!!Male
int64

...
Percent!!CITIZEN, VOTING AGE POPULATION!!Citizen, 18 and over population!!Male
float64
Percent Margin of Error!!CITIZEN, VOTING AGE POPULATION!!Citizen, 18 and over popu
lation!!Male      float64
Percent!!CITIZEN, VOTING AGE POPULATION!!Citizen, 18 and over population!!Female
float64
Percent Margin of Error!!CITIZEN, VOTING AGE POPULATION!!Citizen, 18 and over popu
lation!!Female    float64
Unnamed: 358
float64
Length: 359, dtype: object

```

## 8. Data Cleaning: ACS Demographic and Housing Estimates

As part of the data cleaning process for the `ACS Demographic and Housing Estimates.csv` dataset, the following steps were implemented:

### 1. Exclusion of Puerto Rico Data:

- **Action:** Rows where the 'Geography' column values start with '0500000US7' were excluded.
- **Reason:** The 'Geography' column includes data for all counties within the United States as well as Puerto Rico. Since the `superdata_2020` dataset only contains data for counties within the United States (excluding Puerto Rico), it was necessary to remove Puerto Rican data. Analysis revealed that rows beginning with '0500000US7' correspond to Puerto Rico.
- **Result:** This step filters the dataset to include only U.S. counties, making the row count approximately match that of the `superdata_2020` dataset.

### 2. Remove Prefix from 'Geography' Column:

- **Action:** The prefix '0500000US' is removed from the 'Geography' column.

- **Reason:** For merging with the `superdata_2020` dataset on the `countyFIPS` column, a proper county code is required. The county code is derived from the 'Geography' column after removing the prefix '0500000US'.
- **Result:** This step produces a unique 5-digit code for each county, which accurately identifies counties within their respective states.

### 3. Rename the 'Geography' Column to 'countyFIPS' Column

- **Action:** Rename the 'Geography' Column to 'countyFIPS' Column
- **Reason:** For better understanding and future merging.
- **Result:** It change the name of 'Geography' Column to 'countyFIPS' Column in `ACS_filtered` dataset

### 4. Change Datatype of 'countyFIPS' column

- **Action:** Change the Datatype of 'countyFIPS' column from object to int.
- **Reason:** For merging with `superdata_2020` dataset column 'countyFIPS' which is an int datatype. Because merging column should contain same datatype
- **Result:** It change the Datatype of 'countyFIPS' column from object to int.

### 5. Drop the last column wich contain NaN value

- **Action:** Drop the last column
- **Reason:** Last column is unnamed and it's conatin NaN value for all rows
- **Result:** Column count can be 358 from 359.

In [116...

```
ACS_filtered = ACS[~ACS['Geography'].astype(str).str.startswith('0500000US7')].copy()
#print(ACS_filtered.Geography.nunique())

prefix_to_remove = '0500000US'
# Remove the prefix from the 'Geography' column
ACS_filtered['Geography'] = ACS_filtered['Geography'].astype(str).str.replace(prefix_to_remove, '')
ACS_filtered.rename(columns={'Geography': 'countyFIPS'}, inplace=True)
#print(ACS_filtered['countyFIPS'].dtype)
#print(superdata_2020['countyFIPS'].dtype)
ACS_filtered['countyFIPS'] = ACS_filtered['countyFIPS'].astype(int)
ACS_filtered = ACS_filtered.drop(ACS_filtered.columns[-1], axis=1)
print("(Rows, Columns)=", ACS_filtered.shape)
print(ACS_filtered.head())
```

(Rows, Columns)= (3143, 358)

	countyFIPS	Geographic Area Name \
0	1001	Autauga County, Alabama
1	1003	Baldwin County, Alabama
2	1005	Barbour County, Alabama
3	1007	Bibb County, Alabama
4	1009	Blount County, Alabama

	Estimate!!SEX AND AGE!!Total population \
0	55639
1	218289
2	25026
3	22374
4	57755

	Margin of Error!!SEX AND AGE!!Total population \
0	*****
1	*****
2	*****
3	*****
4	*****

	Estimate!!SEX AND AGE!!Total population!!Male \
0	27052
1	105889
2	13156
3	12022
4	28677

	Margin of Error!!SEX AND AGE!!Total population!!Male \
0	167
1	253
2	86
3	170
4	153

	Estimate!!SEX AND AGE!!Total population!!Female \
0	28587
1	112400
2	11870
3	10352
4	29078

	Margin of Error!!SEX AND AGE!!Total population!!Female \
0	167
1	253
2	86
3	170
4	153

	Estimate!!SEX AND AGE!!Total population!!Sex ratio (males per 100 females) \
0	94.6
1	94.2
2	110.8
3	116.1
4	98.6

	Margin of Error!!SEX AND AGE!!Total population!!Sex ratio (males per 100 female s) \
0	1.1
1	0.4
2	1.5
3	3.6
4	1.0

... \

0	...
1	...
2	...
3	...
4	...

Percent!!HISPANIC OR LATINO AND RACE!!Total population!!Not Hispanic or Latino!!Two or more races!!Two races excluding Some other race, and Three or more races

\	
0	1.9
1	1.8
2	1.2
3	0.4
4	1.6

Percent Margin of Error!!HISPANIC OR LATINO AND RACE!!Total population!!Not Hispanic or Latino!!Two or more races!!Two races excluding Some other race, and Three or more races \

0	0.7
1	0.4
2	0.6
3	0.3
4	0.3

Percent!!Total housing units	Percent Margin of Error!!Total housing units \
0	(X)
1	(X)
2	(X)
3	(X)
4	(X)

Percent!!CITIZEN, VOTING AGE POPULATION!!Citizen, 18 and over population \

0	41959
1	167986
2	19468
3	17583
4	42978

Percent Margin of Error!!CITIZEN, VOTING AGE POPULATION!!Citizen, 18 and over population \

0	(X)
1	(X)
2	(X)
3	(X)
4	(X)

Percent!!CITIZEN, VOTING AGE POPULATION!!Citizen, 18 and over population!!Male

\	
0	47.6
1	47.8
2	53.2
3	53.6
4	48.9

Percent Margin of Error!!CITIZEN, VOTING AGE POPULATION!!Citizen, 18 and over population!!Male \

0	0.4
1	0.2
2	0.2
3	0.4
4	0.3

```

Percent!!CITIZEN, VOTING AGE POPULATION!!Citizen, 18 and over population!!Femal
e \
0          52.4
1          52.2
2          46.8
3          46.4
4          51.1

Percent Margin of Error!!CITIZEN, VOTING AGE POPULATION!!Citizen, 18 and over po
pulation!!Female
0          0.4
1          0.2
2          0.2
3          0.4
4          0.3

[5 rows x 358 columns]

```

## 9. Merging the COVID-19 data with Enrichment data

- **Covid-19 data name:** `superdata_2020`
- **Enrichment data name:** `ACS_filtered`
- **After merging dataset name:** `merge_Enrichment_data`

After analyzing the two datasets, we perform a merge on the `countyFIPS` column using an `outer` join.

**Reason for Outer Join:** An outer join is used to ensure that all records from both datasets are included in the merged result. This is necessary because there are some `countyFIPS` codes present in the `ACS_filtered` dataset that are not found in the `superdata_2020` dataset. Examples of such `countyFIPS` codes include 2063, 2066, and 2261.

By using an outer join, we can include these records and ensure a comprehensive dataset that captures all relevant information.

```

In [117... merge_Enrichment_data=pd.merge(superdata_2020,ACS_filtered,on=['countyFIPS'],how='c
print("(Rows, Columns)= ",merge_Enrichment_data.shape)
print(merge_Enrichment_data.head())
merge_Enrichment_data.to_csv('MergeData.csv', index=False)

```

(Rows, Columns)= (3144, 1052)

	countyFIPS	County Name	State	StateFIPS	population	2020-01-22_cases	\
0	1001	Autauga County	AL	1.0	55869.0	0.0	
1	1003	Baldwin County	AL	1.0	223234.0	0.0	
2	1005	Barbour County	AL	1.0	24686.0	0.0	
3	1007	Bibb County	AL	1.0	22394.0	0.0	
4	1009	Blount County	AL	1.0	57826.0	0.0	

	2020-01-23_cases	2020-01-24_cases	2020-01-25_cases	2020-01-26_cases	\
0	0.0	0.0	0.0	0.0	
1	0.0	0.0	0.0	0.0	
2	0.0	0.0	0.0	0.0	
3	0.0	0.0	0.0	0.0	
4	0.0	0.0	0.0	0.0	

... \

0 ...

1 ...

2 ...

3 ...

4 ...

Percent!!HISPANIC OR LATINO AND RACE!!Total population!!Not Hispanic or Latino!!Two or more races!!Two races excluding Some other race, and Three or more races

	\
0	1.9
1	1.8
2	1.2
3	0.4
4	1.6

Percent Margin of Error!!HISPANIC OR LATINO AND RACE!!Total population!!Not Hispanic or Latino!!Two or more races!!Two races excluding Some other race, and Three or more races

	\
0	0.7
1	0.4
2	0.6
3	0.3
4	0.3

	Percent!!Total housing units	Percent Margin of Error!!Total housing units	\
0	(X)	(X)	
1	(X)	(X)	
2	(X)	(X)	
3	(X)	(X)	
4	(X)	(X)	

	Percent!!CITIZEN, VOTING AGE POPULATION!!Citizen, 18 and over population	\
0	41959.0	
1	167986.0	
2	19468.0	
3	17583.0	
4	42978.0	

Percent Margin of Error!!CITIZEN, VOTING AGE POPULATION!!Citizen, 18 and over population

	\
0	(X)
1	(X)
2	(X)
3	(X)
4	(X)

Percent!!CITIZEN, VOTING AGE POPULATION!!Citizen, 18 and over population!!Male

\



0	47.6
1	47.8
2	53.2
3	53.6
4	48.9

Percent Margin of Error!!CITIZEN, VOTING AGE POPULATION!!Citizen, 18 and over population!!Male \

0	0.4
1	0.2
2	0.2
3	0.4
4	0.3

Percent!!CITIZEN, VOTING AGE POPULATION!!Citizen, 18 and over population!!Female \

0	52.4
1	52.2
2	46.8
3	46.4
4	51.1

Percent Margin of Error!!CITIZEN, VOTING AGE POPULATION!!Citizen, 18 and over population!!Female

0	0.4
1	0.2
2	0.2
3	0.4
4	0.3

[5 rows x 1052 columns]

## 10. Enrichment Data's Role in COVID-19 Spread Analysis with Initial Hypothesis Questions

Demographic factors like population density, age distribution, and socioeconomic conditions can affect the transmission rate and mortality of COVID-19 in a region. For instance:

- **Population Density and Housing:** The number of housing units, particularly in relation to the population size, can provide insights into population density and crowding, both of which are factors that increase the likelihood of COVID-19 spread.
- **Age Distribution:** Areas with a higher elderly population might experience higher mortality rates since COVID-19 poses a greater risk to older adults.
- **Sex Ratios and COVID-19:** The dataset provides information about the sex ratio, which can be used to analyze if certain trends in the virus's transmission. For example, men were initially found to have a higher risk of severe outcomes from COVID-19.

**Initial Hypothesis Questions:** The enriched dataset allows us to pose several hypothesis questions for future analysis:

1. Does higher population density correlate with a higher rate of COVID-19 cases?
2. Are counties with a larger elderly population experiencing higher COVID-19 death rates?
3. Does sex ratio influence the COVID-19 death rate?