# DataEng: Data Integration Activity

This week you will gain hands-on experience with Data Integration by combining data from two distinct sources into a unified DataFrame for analysis.

**Submit**: Make a copy of this document and use it to record your results. Store a PDF copy of the document in your git repository along with any needed code before submitting for this week.

Your job is to integrate <u>county-level COVID-19 data</u> with the <u>ACS Census Tract data for 2017</u> to build a model that allows you to relate COVID numbers with economic data such as population, per capita income and poverty level. To do this you should build a pandas DataFrame that has a row per USA county (there are more than 3000 counties in the USA) and includes the following columns:

County - name of the county

State - name of the state in which the county resides

TotalCases - total number of COVID cases for this county as of February 20, 2021 Dec2020Cases - number of COVID cases recorded in this county in December of 2020 TotalDeaths - total number of COVID deaths for this county as of February 20, 2021 Dec2020Deaths - number of COVID deaths recorded in this county in December of 2020

Population - population of this county

Poverty - % of people in poverty in this county

PerCapitalncome - per capita personal income for this county

We hope that you make it all the way through to the end. Regardless, use your time wisely to gain python programming experience and learn as much as you can about building integrated multi-source data models using python and pandas.

For this activity you should use whichever environment is convenient for you to develop with python 3 and pandas. You are not required to use GCP, but you can use it if you prefer.

Submit: <u>In-class Activity Submission Form</u>

### A. Aggregate Census Data to County Level

Your integration will use two different dimensions: location (as indicated by state and county) and time. You should greatly simplify your processing and reduce your time by pre-processing your data along each of these dimensions.

The ACS data is separated into "Census Tracts" which are regions within counties that correspond to groups of approximately 4000 people. The Census Bureau defines these

to help organize the actual job of collecting census data, but this grouping can make your Data Engineering job more more challenging. This level of detail is not needed for your county-level analysis, and you can greatly decrease your efforts by aggregating per-tract data to the county level.

Create a python program that produces a one-row-per-county version of the ACS data set. To do this you will need to think about how to properly aggregate Census Tract-level data into County-level summaries.

In this step you can also eliminate unneeded columns from the ACS data.

**Question**: Show your aggregated county-level data rows for the following counties: Loudoun County Virginia, Washington County Oregon, Harlan County Kentucky, Malheur County Oregon

state	county	population	income	poverty_population	income_per_capita	poverty
9 Virginia	Loudoun County	374558	8298861.0	13819.683	8673419.0	0.036896
l Oregon	Washington County	572071	7961909.0	59044.602	8533980.0	0.103212
2 Kentucky	Harlan County	27548	291194.0	9826.229	318742.0	0.356695
3 Oregon	Malheur County	30421	272162.0	7391.763	302583.0	0.242982

### B. Simplify the COVID Data

You can simplify the COVID data along the time dimension. The COVID data set contains day-level resolution data from (approximately) March of 2020 through February of 2021. However, you will only need four data points per county: total cases, total deaths, cases reported during December of 2020 and deaths reported during December 2020.

Create a python program that reduces the COVID data to one line per county.

**Question**: Show your simplified COVID data for the counties listed above.

	state		county	total_cases	total_deaths	dec_cases	dec_deaths	
0	Virginia	Loudoun	County	2496450	35820.0	2496450	35820.0	
1	0regon	Washington	County	2157339	22455.0	2157339	22455.0	
2	Kentucky	Harlan	County	205984	3994.0	205984	3994.0	
3	Oregon	Malheur	County	453634	7770.0	453634	7770.0	

### C. Integrate COVID Data with ACS Data

Create a single pandas DataFrame containing one row per county and using the columns described above. You are free to add additional columns if needed. For example, you might want to normalize all of the COVID data by the population of each county so that you have a consistent "number of cases/deaths per 100000 residents" value for each county.

**Question**: List your integrated data for all counties in the State of Oregon.

		county	state	population	income	poverty_population	income_per_capita	poverty	total_cases	total_deaths	dec_cases	dec_deaths
126			0regon		264799.0		280779.0					
197			0regon	88249	972822.0		1061071.0	0.224212	180225.0		180225.0	
533			0regon						1284402.0	20040.0	1284402.0	20040.0
558								0.121901				
631								0.123153		1363.0		
656						11260.649	631284.0	0.178965		969.0		
697				21717						1134.0		1134.0
718					214926.0		237303.0					
796				175321		21215.415		0.121009				
839					999898.0		1107474.0	0.170260			174952.0	
1068							41741.0					
1115					86283.0			0.136358				
1238						1261.195	92849.0	0.175288	17024.0		17024.0	
1320					232038.0	2780.807		0.121231	107383.0		107383.0	
1415				212070	2021385.0				713288.0		713288.0	
1453	Jefferson				292205.0		314912.0	0.206949				
1493							750930.0	0.186464		2638.0		2638.0
1559						12337.856			224256.0		224256.0	
1608								0.201393		348.0		348.0
1626	Lane			363471	4317975.0			0.192305				
1720							811161.0			3117.0		3117.0
1731				121074	1037043.0	19449.241	1158117.0			5949.0		5949.0
1822				30421	272162.0			0.242982		7770.0		
1848				330453		53297.166	3564175.0		1974030.0			
2070				11153	109814.0		120967.0		139209.0		139209.0	1447.0
2078				788459	11278735.0	129896.001	12067194.0		3374737.0	58787.0	3374737.0	58787.0
2359				79666		12459.729	712726.0		268036.0	5480.0		
2671					42074.0			0.137000	5807.0		5807.0	
2860	Tillamook			25840	360838.0	4008.486		0.155127	34370.0		34370.0	
2919				76736	754218.0	13678.362		0.178252	933975.0		933975.0	
2933				25810	385921.0	4547.360		0.176186	161223.0		161223.0	
2994				6864		943.716		0.137488	13017.0		13017.0	
3021				25687		3511.623		0.136708	121202.0	3039.0	121202.0	3039.0
3046				572071	7961909.0	59044.602	8533980.0		2157339.0			
3109	Wheeler					291.490		0.206000			1454.0	
3195		County	Oregon	102366	1015494.0	14129.229	1117860.0	0.138027	356425.0	6010.0	356425.0	

## D. Analysis

For each of the following, determine the strength of the correlation between each pair of variables. Compute the correlation strength by calculating the Pearson correlation coefficient R for pairs of columns in your DataFrame. For example, if you have a DataFrame df with each row representing a distinct county, and columns named 'TotalCases' and 'Poverty', then you can compute R like this:

```
R = df['TotalCases'].corr(df['Poverty'])
```

For any R that is > 0.5 or < -0.5 also display a scatter plot (see <u>pandas scatterplot</u> and <u>seaborn</u> <u>documentation</u> for information about how to display scatter plots from DataFrame data).

The COVID numbers should be normalized to population (# of cases per 100,000 residents) so that different sized counties are comparable. So for example, "COVID total cases" below really means "((COVID total cases in county \* 100000) / population of county)".

- 1. Across all of the counties in the State of Oregon
  - a. COVID total cases vs. % population in poverty
  - b. COVID total deaths vs. % population in poverty
  - c. COVID total cases vs. Per Capita Income level
  - d. COVID total deaths vs. Per Capita Income level
  - e. COVID cases during December 2020 vs. % population in poverty
  - f. COVID deaths during December 2020 vs. % population in poverty
  - g. COVID cases during December 2020 vs. Per Capita Income level
  - h. COVID cases during December 2020 vs. Per Capita Income level

	corrs_cases_poverty	corrs_cases_ipc	corrs_deaths_poverty	corrs_deaths_ipc	corrs_dec_cases_poverty	corrs_dec_cases_ipc	corrs_dec_deaths_poverty	corrs_dec_deaths_ipc
0	0.192759	-0.003175	0.269679	0.052913	0.192759	-0.003175	0.269679	0.052913

- 2. Across all of the counties in the entire USA
  - a. COVID total cases vs. % population in poverty
  - b. COVID total deaths vs. % population in poverty
  - c. COVID total cases vs. Per Capita Income level
  - d. COVID total deaths vs. Per Capita Income level
  - e. COVID cases during December 2020 vs. % population in poverty
  - f. COVID deaths during December 2020 vs. % population in poverty
  - g. COVID cases during December 2020 vs. Per Capita Income level
  - h. COVID cases during December 2020 vs. Per Capita Income level



Note that this exercise does not constitute a competent, thorough statistical analysis of the relationships between immunological data and demographic data. It is just an illustration of the types of computations that might be accomplished with an integrated data set.