## GEOVIS - Process Crime Data

Consolidate crime data calculating monthly averages for each crime variable, for each county in TX and NY

## 0. Libraries and Variables

```
import arcpy
import arcgis
import os
import numpy as np
import pandas as pd

from arcgis import GIS
from arcgis.features import GeoAccessor, GeoSeriesAccessor
```

```
In [3]: _gis = GIS('Pro')
        _inGdb = r'C:\Users\Ameri\OneDrive - The University of Texas at Dallas\UT Dalla
        _outFolder = r'C:\Users\Ameri\OneDrive - The University of Texas at Dallas\UT [
        _outGdb = 'Crime1.gdb'
        outGdb = os.path.join( outFolder, outGdb)
        d_CrimeTypes = {'aggravated_assault': 'agga',
                         'arson': 'arsn',
                         'burglary': 'burg',
                         'homicide': 'homi',
                         'larceny': 'larc',
                         'motor_vehicle_theft': 'mvt',
                         'rape': 'rape',
                         'robbery': 'rob'}
        1_CrimeTypes = list(sorted(d_CrimeTypes.keys()))
        # county polygons come from processed weather fc
        _fc_ProcessedWeather = r'C:\Users\Ameri\OneDrive - The University of Texas at [
        # city polygons
        _fc_Cities = r'C:\Users\Ameri\OneDrive - The University of Texas at Dallas\UT [
        print(_inGdb, os.path.exists(_inGdb))
        print(_outFolder, os.path.exists(_outFolder))
        print( fc ProcessedWeather, arcpy.Exists( fc ProcessedWeather))
        print(_fc_Cities, arcpy.Exists(_fc_Cities))
```

C:\Users\Ameri\OneDrive - The University of Texas at Dallas\UT Dallas\2024 03
Fall\EPPS 6356 Data Visualization\GeoVis\RawData\Crime\States\_all\monthly\_198
5to2023\_PD19178\_point.gdb True

C:\Users\Ameri\OneDrive - The University of Texas at Dallas\UT Dallas\2024 03
Fall\EPPS 6356 Data Visualization\GeoVis\DataProcess True

C:\Users\Ameri\OneDrive - The University of Texas at Dallas\UT Dallas\2024 03
Fall\EPPS 6356 Data Visualization\GeoVis\DataProcess\Weather3.gdb\Weather\_NY\_
TX v3 True

C:\Users\Ameri\OneDrive - The University of Texas at Dallas\UT Dallas\2024 03
Fall\EPPS 6356 Data Visualization\GeoVis\RawData\AdministrativeBoundaries\Cit
y\_Boundary\city\_TX\_NY.gdb\City\_NY\_TX True

```
In [4]: # Verify all crime feature classes exist

for tag in l_CrimeTypes:
    fc = f'{tag}_monthly_1985to2023'
    print(fc, arcpy.Exists(os.path.join(_inGdb, fc)))

aggravated_assault_monthly_1985to2023 True
    arson_monthly_1985to2023 True
    burglary_monthly_1985to2023 True
    homicide_monthly_1985to2023 True
    larceny_monthly_1985to2023 True
    motor_vehicle_theft_monthly_1985to2023 True
    rape_monthly_1985to2023 True
    robbery_monthly_1985to2023 True
```

### 1. Filter all crime feature classes in Texas & New York

```
In [5]: # Filter all crime feature classes
        d_DFs_Unprocessed = {} # dictionary of crime types dataframes
        # loop all crime types,
              filter NY & TX data,
              save filtered dataframe in dictionary
        for tag in 1 CrimeTypes:
            # get feature class name
            fc = f'{tag}_monthly_1985to2023'
            print(f"~ ... filtering feature class --> {fc}")
            fc = os.path.join(_inGdb, fc)
            # Read feature class
            sdf = pd.DataFrame.spatial.from_featureclass(fc)
            # filter the data # NY & TX
            # state name IN ('New York', 'Texas')
            filtered_df = sdf[sdf['state_name'].isin(['New York', 'Texas'])].copy()
            # save filtered data in dictionary d_DFs for later processing
            d_DFs_Unprocessed[tag] = filtered_df
            ... filtering feature class --> aggravated_assault_monthly_1985to2023
            ... filtering feature class --> arson_monthly_1985to2023
            ... filtering feature class --> burglary_monthly_1985to2023
            ... filtering feature class --> homicide_monthly_1985to2023
            ... filtering feature class --> larceny monthly 1985to2023
            ... filtering feature class --> motor_vehicle_theft_monthly_1985to2023
            ... filtering feature class --> rape_monthly_1985to2023
            ... filtering feature class --> robbery_monthly_1985to2023
```

In [6]: # read weather feature class (processed)
# crime data will be added to weather fc

sdf\_Weather = pd.DataFrame.spatial.from\_featureclass(\_fc\_ProcessedWeather)
sdf\_Weather.head()

Out[6]:	OBJECTID		OBJECTID statefp		countyfp countyns		name	namelsad	Isad	classfp	mtfcc	Cŧ
	0	1	48	327	01383949	48327	Menard	Menard County	06	H1	G4020	
	1	2	48	189	01383880	48189	Hale	Hale County	06	H1	G4020	
	2	3	48	011	01383791	48011	Armstrong	Armstrong County	06	H1	G4020	
	3	4	48	057	01383814	48057	Calhoun	Calhoun County	06	H1	G4020	
	4	5	48	077	01383824	48077	Clay	Clay County	06	H1	G4020	

5 rows × 115 columns

```
In [7]: # Create a temp Weather df with county name as index
# to allow for a quick data retrieval

df_Weather_Temp = sdf_Weather.copy()
df_Weather_Temp['COUNTYID'] = df_Weather_Temp['name'].str.strip().str.upper()
df_Weather_Temp.set_index('COUNTYID', inplace=True)
l_counties = df_Weather_Temp.index.tolist()

print(len(l_counties))
print(len(df_Weather_Temp))
df_Weather_Temp.head()
```

316 316

## Out[7]:

	OBJECTID	statefp	countyfp	countyns	geoid	name	namelsad	Isad	classfp
COUNTYID									
MENARD	1	48	327	01383949	48327	Menard	Menard County	06	H1
HALE	2	48	189	01383880	48189	Hale	Hale County	06	H1
ARMSTRONG	3	48	011	01383791	48011	Armstrong	Armstrong County	06	H1
CALHOUN	4	48	057	01383814	48057	Calhoun	Calhoun County	06	H1
CLAY	5	48	077	01383824	48077	Clay	Clay County	06	H1

5 rows × 115 columns

```
In [9]: # read cities feature class
sdf_Cities = pd.DataFrame.spatial.from_featureclass(_fc_Cities)
sdf_Cities.head()
```

#### Out[9]: **OBJECTID** NAME MUNI\_TYPE MUNITYCODE COUNTY GNIS\_ID FIPS\_CODE **SWIS** 0 1 Albany city 1.0 Albany 978659 3600101000 010100 1 2 Amsterdam city 1.0 Montgomery 978677 3605702066 270100 2 3 Auburn 1.0 3601103078 050100 city Cayuga 978695 3 4 Batavia city 1.0 Genesee 978713 3603704715 180200 5 1.0 978716 3602705100 130200 Beacon city Dutchess

```
In [11]: # Create a temp Cities df with county name as index
# to allow for a quick data retrieval

df_City_Temp = sdf_Cities[['OBJECTID', 'NAME', 'COUNTY', 'SHAPE']].copy()
    df_City_Temp['COUNTYID'] = df_City_Temp['NAME'].str.strip().str.upper()
    df_City_Temp.set_index('COUNTYID', inplace=True)

print(len(df_City_Temp))
    df_City_Temp.head()
```

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Out[11]:		OBJECTID	NAME	COUNTY	SHAPE
	COUNTYID				
	ALBANY	1	Albany	Albany	{"rings": [[[-73.85154879699996, 42.7115402990
	AMSTERDAM	2	Amsterdam	Montgomery	{"rings": [[[-74.20012254099998, 42.9659517100
	AUBURN	3	Auburn	Cayuga	{"rings": [[[-76.59473494199995, 42.9560956210
	BATAVIA	4	Batavia	Genesee	{"rings": [[[-78.19013013599994, 43.0164674340
	BEACON	5	Beacon	Dutchess	{"rings": [[[-73.94261272799997, 41.5150065580

```
In [12]: # create dictionary for months that will be used to process data
         d_month = {}
         d_month[1] = 'Jan'
         d_month[2] = 'Feb'
         d_{month[3]} = 'Mar'
         d_{month[4]} = 'Apr'
         d_month[5] = 'May'
         d_{month[6]} = 'Jun'
         d_{month[7]} = 'Jul'
         d_month[8] = 'Aug'
         d_month[9] = 'Sep'
         d_month[10] = 'Oct'
         d_month[11] = 'Nov'
         d_month[12] = 'Dec'
         l_months = list(d_month.keys())
         print(d_month)
         print(l_months)
         {1: 'Jan', 2: 'Feb', 3: 'Mar', 4: 'Apr', 5: 'May', 6: 'Jun', 7: 'Jul', 8: 'Au
```

```
{1: 'Jan', 2: 'Feb', 3: 'Mar', 4: 'Apr', 5: 'May', 6: 'Jun', 7: 'Jul', 8: 'Aug', 9: 'Sep', 10: 'Oct', 11: 'Nov', 12: 'Dec'}
[1, 2, 3, 4, 5, 6, 7, 8, 9, 10, 11, 12]
```

```
Loop through crime datasets
In [13]: #
         def PreProcessCrimeDataset(df):
             # Create an empty DataFrame with the same columns
             1 columns = list(df.columns)
             # remove all 2024 columns
             1_cols2remove = [x for x in 1_columns if x.startswith(f'y2024')]
             1_columns = [x for x in 1_columns if not x in 1_cols2remove]
             # 1st get all crime data that do NOT need spliting values
             df = df[l_columns].copy()
             new_df = df[~df['counties'].str.contains(',')].copy()
             # Loop through each row in the original DataFrame that needs spliting value
             df_to_duplicate = df[df['counties'].str.contains(',')].copy()
             for index, row in df_to_duplicate.iterrows():
                 l_counties = row['counties'].split(',')
                 # Trim whitespace from each county
                 l_counties = [cty.strip() for cty in l_counties]
                 pct = int(100 / len(l_counties))
                 for cty in l_counties:
                     # Create a new row for each county
                     new row = row.copy()
                     new_row['counties'] = cty
                     for m in 1 months:
                         l_cols2Process = [x for x in l_columns if x.endswith(f'_{m}')]
                         for col2Process in l_cols2Process:
                             if not pd.isnull(new row[col2Process]):
                                 newValue = new_row[col2Process]
                                 if newValue >= len(l_counties):
                                     newValue = int(newValue / len(l counties))
                                 new_row[col2Process] = newValue
                     # Append the new row to the new DataFrame
                     new_df = new_df.append(new_row, ignore_index=True)
             return new df
         for tag in l_CrimeTypes:
             print(f"~ ... Pre-processing dataset: {tag}")
             df = d_DFs_Unprocessed[tag]
             newdf = PreProcessCrimeDataset(df)
             d_DFs_Unprocessed[tag] = newdf
         # df = d_DFs_Unprocessed['arson'].copy()
         # # newdf = df[df['counties'].str.contains(',')]
         # newdf = ProcessCrimeDataset(df)
```

```
# print(len(newdf))
# newdf.head()
```

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## 2. Calculate monthly averages

```
In [19]: # function to calculate monthly averages in dataframe
         def ProcessCrimeDF(df, crimeType):
             sdf = df.copy() # make a copy to preserve original df
             l_columns = list(sdf.columns)
             1_colsProcessed = []
             l_newColumns = []
             # calculate monthly averages
             for m in 1 months:
                  # get columns ending in m (month)
                 l_cols2Process = [x for x in l_columns if x.endswith(f'_{m}')]
                  # add cols2process to list of processed cols
                 1_colsProcessed.extend(1_cols2Process)
                 # create new monthly avg column
                  #colName = f'{weatherType}_avg_{m}_{d_month[m]}'
                  colName = f'c_{d_CrimeTypes[crimeType]}_{d_month[m]}'
                 # calculate mean value in monthly column
                  print(f"~ ... Processing column: {colName}")
                  sdf[colName] = sdf[l_cols2Process].mean(axis=1)
                  # add to list of new columns
                  1_newColumns.append(colName)
             # get updated columns list
             #L_columns = list(sdf.columns)
             # columns excluding processed ones
             #L columns = [x \text{ for } x \text{ in } l \text{ columns if } x \text{ not in } l \text{ colsProcessed}]
             l desiredCols = [x for x in l newColumns]
             l_desiredCols.insert(0, 'counties')
             sdf = sdf[l_desiredCols].copy()
             # group by county and calculate averages
             sdf['COUNTYID'] = sdf['counties'].str.strip().str.upper()
             grouped_df = sdf.groupby('COUNTYID')[l_newColumns].mean()
             grouped df = grouped df.reset index()
             # return updated/grouped sdf
             return grouped df
```

```
In [20]: # process all crime dataframes
         d_DFs_Processed = {} # dictionary of processed crime dataframes
         for tag in l_CrimeTypes:
             # get dataframe to process from dict
             df = d DFs Unprocessed[tag]
             print(f"~ ... calculating monthly averages for crime variable --> {tag}")
             sdf = ProcessCrimeDF(df, tag)
             # add processed df to processed dict
             d DFs Processed[tag] = sdf
                ... Processing column: c_arsn_Jul
                ... Processing column: c_arsn_Aug
                ... Processing column: c_arsn_Sep
                ... Processing column: c_arsn Oct
                ... Processing column: c_arsn_Nov
                ... Processing column: c_arsn_Dec
             ... calculating monthly averages for crime variable --> burglary
                ... Processing column: c_burg_Jan
                ... Processing column: c_burg_Feb
                ... Processing column: c_burg_Mar
                ... Processing column: c_burg_Apr
                ... Processing column: c_burg_May
                ... Processing column: c_burg_Jun
                ... Processing column: c_burg_Jul
                ... Processing column: c burg Aug
                ... Processing column: c_burg_Sep
                ... Processing column: c_burg_Oct
                ... Processing column: c_burg_Nov
                ... Processing column: c_burg_Dec
```

... calculating monthly averages for crime variable --> homicide

# 3. Consolidate Crime Variables Monthly Averages into a single Dataset

```
In [21]: # merge processed columns into a single dataframe
         df_Final = None # this will hold the final dataframe
         for tag in l_CrimeTypes:
             # get processed dataframe
             df = d_DFs_Processed[tag]
             if df Final is None:
                 df_Final = df.copy() # initialize with first dataframe
             else:
                 # append dataframe df to df_Final
                 # get only processed columns
                 1 columns = list(df.columns)
                 1_cols_desired = [x for x in 1_columns if x.startswith(f'c_{d_CrimeType
                 l_cols_desired.insert(0, 'COUNTYID') # keep id for merging purposes
                 # merge to df final
                 df = df[l_cols_desired].copy()
                 df_Final = pd.merge(df_Final, df, on='COUNTYID')
         df_Final.head()
```

## Out[21]:

	COUNTYID	c_agga_Jan	c_agga_Feb	c_agga_Mar	c_agga_Apr	c_agga_May	c_agga_Jun	<b>c</b> _
0	ALBANY	4.228518	3.859037	4.952757	5.110903	5.718672	5.813904	
1	ALLEGANY	0.576763	0.432766	0.544778	0.558405	0.537479	0.466439	
2	ANDERSON	4.405271	4.469373	4.777778	4.839744	5.288462	5.248575	
3	ANDREWS	2.076923	1.250337	2.048920	1.987179	2.162281	1.753711	
4	ANGELINA	1.820441	1.708778	2.013495	1.987916	2.600978	2.271838	

5 rows × 97 columns

```
In [22]: # Merge crime variables to weather variables
         df_Weather_Crime = df_Weather_Temp.reset_index() # weather dataframe
         df_Weather_Crime = pd.merge(df_Weather_Crime, df_Final, on='COUNTYID') # merge
         df_Weather_Crime.head()
```

Out[22]:		COUNTYID	OBJECTID	statefp	countyfp	countyns	geoid	name	namelsad	Isad	clas
	0	MENARD	1	48	327	01383949	48327	Menard	Menard County	06	
	1	HALE	2	48	189	01383880	48189	Hale	Hale County	06	
	2	ARMSTRONG	3	48	011	01383791	48011	Armstrong	Armstrong County	06	
	3	CALHOUN	4	48	057	01383814	48057	Calhoun	Calhoun County	06	
	4	CLAY	5	48	077	01383824	48077	Clay	Clay	06	

5 rows × 212 columns

```
In [23]: # Create filegdb
         if (not os.path.exists(_outGdb)):
             _outGdb = 'Weather_Crime_1.gdb'
             arcpy.management.CreateFileGDB(_outFolder, _outGdb)
             _outGdb = os.path.join(_outFolder, _outGdb)
         print(_outGdb, os.path.exists(_outGdb))
```

C:\Users\Ameri\OneDrive - The University of Texas at Dallas\UT Dallas\2024 03 Fall\EPPS 6356 Data Visualization\GeoVis\DataProcess\Weather\_Crime\_1.gdb True

County

In [24]: # convert to spatial dataframe

sdf\_Weather\_Crime = GeoAccessor.from\_df(df\_Weather\_Crime, geometry\_column='SHAF
sdf\_Weather\_Crime.head()

Out[24]:		COUNTYID	OBJECTID	statefp	countyfp	countyns	geoid	name	namelsad	Isad	clas
	0	MENARD	1	48	327	01383949	48327	Menard	Menard County	06	
	1	HALE	2	48	189	01383880	48189	Hale	Hale County	06	
	2	ARMSTRONG	3	48	011	01383791	48011	Armstrong	Armstrong County	06	
	3	CALHOUN	4	48	057	01383814	48057	Calhoun	Calhoun County	06	
	4	CLAY	5	48	077	01383824	48077	Clay	Clay County	06	

5 rows × 212 columns

In [26]: # save consolidated data to feature class
outWeatherCrimeFC = 'WeatherCrime\_NYTX\_v1'
outWeatherCrimeFC = os.path.join(\_outGdb, outWeatherCrimeFC)
sdf\_Weather\_Crime.spatial.to\_featureclass(outWeatherCrimeFC)

In [27]: print(len(sdf\_Weather\_Crime))

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In [ ]: