## **Data Processing: ACS Demographics**

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Processes the ACs demographics. Needs access to the files available on the team GitHub folder, which are too large and numerous to upload here.

```
In [1]:
         1 #imports
           2 import numpy as np
           3 import matplotlib.pyplot as plt
           4 import matplotlib.image as mpimg
           5 import seaborn as sns
           6 import numpy as np
           7 import pandas as pd
           8 import pickle
           9 %matplotlib inline
In [2]:
          1 | # dictionary to translate numeric codes to state abbrev. names
           2 state_codes = {
           3 1: "AL", 2: "AK", 4: "AZ", 5: "AR", 6: "CA", 8: "CO", 9: "CT", 10: "DE",
           4 | 11: "DC",12: "FL",13: "GA",15: "HI",16: "ID",17: "IL",18: "IN",19: "IA",
             20: "KS",21: "KY",22: "LA",23: "ME",24: "MD",25: "MA",26: "MI",27: "MN",
28: "MS",29: "MO",30: "MT",31: "NE",32: "NV",33: "NH",34: "NJ",35: "NM",
36: "NY",37: "NC",38: "ND",39: "OH",40: "OK",41: "OR",42: "PA",44: "RI",
45: "SC",46: "SD",47: "TN",48: "TX",49: "UT",50: "VT",51: "VA",53: "WA",
              54: "WV",55: "WI",56: "WY"}
In [3]:
          1 # raw ACS data file paths
           2 path10 = 'Datasets/demographics/ACS_10_1YR_S0201_with_ann.csv'
           3 path12 = 'Datasets/demographics/ACS_12_1YR_S0201_with_ann.csv'
              path14 = 'Datasets/demographics/ACS_14_1YR_S0201_with_ann.csv'
              path16 = 'Datasets/demographics/ACS_16_1YR_S0201_with_ann.csv'
              path17 = 'Datasets/demographics/ACS_17_1YR_S0201_with_ann.csv'
              # read in the raw files
              df10 = pd.read csv(path10, header = 1)
          10 df12 = pd.read csv(path12, header = 1)
          11 df14 = pd.read csv(path14, header = 1)
          12 df16 = pd.read csv(path16, header = 1)
          13 df17 = pd.read csv(path17, header = 1)
```

```
In [4]:
            def clean(df, year):
         1
         3
                # insert year column
         4
                df.insert(0, 'year', year)
         5
         6
                # insert state column (as 2-char code, like: MA)
         7
                df.insert(1, 'state', df.Id2.apply(lambda x: int(x / 100)))
         8
                df.state = df.state.apply(lambda x: state codes[x])
         9
        10
                # insert district column, as an integer
                # districts should start with district 1, not 0
        11
        12
                df.insert(2, 'district', df.Id2.apply(lambda x: int(str(x)[-2:])))
                df.district = df.district.replace(0, 1)
        13
        14
        15
                # filter out Margin of Error columns
                columns to keep = [col for col in df.columns if not 'Margin of Error;' in col
        16
        17
                df = df[columns to keep]
        18
                # create index as: state_district year
        19
                df.index = ['{0} {1:02d} {2}'.format(row['state'], row['district'], row['year
        20
        21
        22
                return df
         1 | acs10 = clean(df10, 2010)
In [5]:
         2 | acs12 = clean(df12, 2012)
            acs14 = clean(df14, 2014)
            acs16 = clean(df16, 2016)
            acs17 = clean(df17, 2017)
In [6]:
         1 # check order of unemployement columns
            [col for col in acs10.columns if 'EMPLOYMENT STATUS - In labor force - Civilian I
Out[6]: ['Estimate; EMPLOYMENT STATUS - In labor force - Civilian labor force - Unemployed
         'Estimate; EMPLOYMENT STATUS - In labor force - Civilian labor force - Unemployed
        - Percent of civilian labor force',
         'Estimate; EMPLOYMENT STATUS - In labor force - Civilian labor force - Unemployed
        .1',
         'Estimate; EMPLOYMENT STATUS - In labor force - Civilian labor force - Unemployed
        - Percent of civilian labor force.1']
In [7]:
        1 len(acs17.columns)
Out[7]: 316
In [8]:
        1 # find the common set of columns between two dataframes
         2 len(set(acs16.columns).intersection(set(acs17.columns)))
Out[8]: 287
```

```
In [9]:
             set(acs10.columns).intersection(set(acs12.columns))
 Out[9]: {'Estimate; CLASS OF WORKER - Civilian employed population 16 years and over',
          'Estimate; CLASS OF WORKER - Government workers',
          'Estimate; CLASS OF WORKER - Private wage and salary workers',
          'Estimate; CLASS OF WORKER - Self-employed workers in own not incorporated busine
          'Estimate; CLASS OF WORKER - Unpaid family workers',
          'Estimate; COMMUTING TO WORK - Car, truck, or van - carpooled',
          'Estimate; COMMUTING TO WORK - Car, truck, or van - drove alone',
          'Estimate; COMMUTING TO WORK - Mean travel time to work (minutes)',
          'Estimate; COMMUTING TO WORK - Other means',
          'Estimate; COMMUTING TO WORK - Public transportation (excluding taxicab)',
          'Estimate; COMMUTING TO WORK - Walked',
          'Estimate; COMMUTING TO WORK - Worked at home',
          'Estimate; COMMUTING TO WORK - Workers 16 years and over',
          'Estimate; DISABILITY STATUS - Civilian noninstitutionalized population 18 to 64
         years',
          'Estimate; DISABILITY STATUS - Civilian noninstitutionalized population 65 years
         and older',
          'Estimate; DISABILITY STATUS - Civilian noninstitutionalized population under 18
In [10]:
          1 #[col for col in df.columns if not 'Margin of Error;' in col]
```

```
In [11]:
          1 # 2010 and 2012
           2 raw_predictor_name_subset_10_12 = [
             "Estimate; SEX AND AGE - Female",
             "Estimate; SEX AND AGE - 18 to 24 years",
             "Estimate; SEX AND AGE - 25 to 34 years",
             "Estimate; SEX AND AGE - Median age (years)",
             "Estimate; RELATIONSHIP - Nonrelatives - Unmarried partner",
             "Estimate; HOUSEHOLDS BY TYPE - Nonfamily households - Male householder - Living
             "Estimate; EDUCATIONAL ATTAINMENT - Bachelor's degree or higher",
             "Estimate; FERTILITY - Women 15 to 50 years who had a birth in the past 12 months
          10
             "Estimate; VETERAN STATUS - Civilian veteran",
             "Estimate; RESIDENCE 1 YEAR AGO - Same house",
             "Estimate; PLACE OF BIRTH, CITIZENSHIP STATUS AND YEAR OF ENTRY - Native",
             "Estimate; PLACE OF BIRTH, CITIZENSHIP STATUS AND YEAR OF ENTRY - Foreign born",
              "Estimate; WORLD REGION OF BIRTH OF FOREIGN BORN - Latin America",
              "Estimate; LANGUAGE SPOKEN AT HOME AND ABILITY TO SPEAK ENGLISH - Language other
              "Estimate; EMPLOYMENT STATUS - In labor force - Civilian labor force - Unemployed
             "Estimate; COMMUTING TO WORK - Public transportation (excluding taxicab)",
              "Estimate; HEALTH INSURANCE COVERAGE - No health insurance coverage",
              "Estimate; POVERTY RATES FOR FAMILIES AND PEOPLE FOR WHOM POVERTY STATUS IS DETEF
          21
              "Estimate; OWNER CHARACTERISTICS - Median value (dollars)"]
          22
          23
             raw_predictor_name_supplement_10 = [
          24
              "Estimate; INCOME IN THE PAST 12 MONTHS (IN 2010 INFLATION-ADJUSTED DOLLARS) - M\epsilon
          25
              "Estimate; INCOME IN THE PAST 12 MONTHS (IN 2010 INFLATION-ADJUSTED DOLLARS) - Wi
          26
          27
             raw_predictor_name_supplement_12 = [
              "Estimate; INCOME IN THE PAST 12 MONTHS (IN 2012 INFLATION-ADJUSTED DOLLARS) - Me
          28
          29
             "Estimate; INCOME IN THE PAST 12 MONTHS (IN 2012 INFLATION-ADJUSTED DOLLARS) - Wi
          30
          31
             raw_predictor_names_10 = raw_predictor_name_subset_10_12 + raw_predictor_name_sup
          32
             raw_predictor_names_12 = raw_predictor_name_subset_10_12 + raw_predictor_name_sup
          33
          34
          35
             # 2014 and beyond
             raw_predictor_name_subset_14_beyond = [
             "Estimate; SEX AND AGE - Total population - Female",
             "Estimate; SEX AND AGE - 18 to 24 years",
             "Estimate; SEX AND AGE - 25 to 34 years",
             "Estimate; SEX AND AGE - Median age (years)",
             "Estimate; RELATIONSHIP - Population in households - Nonrelatives - Unmarried par
             "Estimate; HOUSEHOLDS BY TYPE - Households - Nonfamily households - Male househol
             "Estimate; EDUCATIONAL ATTAINMENT - Bachelor's degree or higher",
          44
             "Estimate; FERTILITY - Women 15 to 50 years - Women 15 to 50 years who had a birt
          45
             "Estimate; VETERAN STATUS - Civilian population 18 years and over - Civilian vete
             "Estimate; RESIDENCE 1 YEAR AGO - Population 1 year and over - Same house",
             "Estimate; PLACE OF BIRTH, CITIZENSHIP STATUS AND YEAR OF ENTRY - Native",
          47
             "Estimate; PLACE OF BIRTH, CITIZENSHIP STATUS AND YEAR OF ENTRY - Foreign born",
             "Estimate; WORLD REGION OF BIRTH OF FOREIGN BORN - Foreign-born population exclud
          49
             "Estimate; LANGUAGE SPOKEN AT HOME AND ABILITY TO SPEAK ENGLISH - Population 5 ye
          51
             "Estimate; EMPLOYMENT STATUS - Population 16 years and over - In labor force - Ci
          52
             "Estimate; COMMUTING TO WORK - Workers 16 years and over - Public transportation
          53
              "Estimate; HEALTH INSURANCE COVERAGE - Civilian noninstitutionalized population -
          54
              "Estimate; POVERTY RATES FOR FAMILIES AND PEOPLE FOR WHOM POVERTY STATUS IS DETER
          55
              "Estimate; OWNER CHARACTERISTICS - Owner-occupied housing units - Median value (
          56
          57
             raw predictor name supplement 14 = [
          58
             "Estimate; INCOME IN THE PAST 12 MONTHS (IN 2014 INFLATION-ADJUSTED DOLLARS) - Ho
          59
              "Estimate; INCOME IN THE PAST 12 MONTHS (IN 2014 INFLATION-ADJUSTED DOLLARS) - Wi
          60
          61
             raw predictor name supplement 16 = [
              "Estimate; INCOME IN THE PAST 12 MONTHS (IN 2016 INFLATION-ADJUSTED DOLLARS) - Ho
          62
          63
              "Estimate; INCOME IN THE PAST 12 MONTHS (IN 2016 INFLATION-ADJUSTED DOLLARS) - Wi
          64
             ..... ..... 12 - 1 - ...... ...... 1 - ..... 1 7
```

```
In [12]:
               renamed predictors = [
                "female_pct",
               "age18_24_pct",
               "age25_34_pct",
               "median_age",
               "unmarried_partner_pct",
               "male_living_alone_pct",
               "bachelors_deg_or_higher_pct",
               "past_year_births_to_unmarried_women_pct",
               "civilian_veteran_pct",
           10
               "live_same_house_past_year_pct",
           12
               "native born population",
                "foreign_born_population",
                "foreign born proportion from LatinAmerica",
                "speak_other_language_at_home_pct",
                "labor force unemployed pct",
                "public transit commuter pct",
           18
                "no health insurance pct",
           19
                "poverty_rate_pct",
           20
                "median_housing_value",
           21
                "median household income",
                "food_stamp_benefits_pct"]
               # check number of predictors
In [13]:
            1
               len(acs17[raw predictor names 17].columns)
Out[13]: 21
In [14]:
            1
               identifier cols = ['state', 'district', 'year']
            2
               acs10 = acs10[['state', 'district', 'year'] + raw_predictor_names_10]
            3
               acs12 = acs12[['state', 'district', 'year'] + raw_predictor_names_12]
acs14 = acs14[['state', 'district', 'year'] + raw_predictor_names_14]
acs16 = acs16[['state', 'district', 'year'] + raw_predictor_names_16]
               acs17 = acs17[['state', 'district', 'year'] + raw_predictor_names_17]
In [15]:
               acs10.columns = identifier cols + renamed predictors
               acs10.head()
Out[15]:
                       state district year female_pct age18_24_pct age25_34_pct median_age unmarried_partner_pct n
           AL_01_2010
                        AL
                                 1 2010
                                              51.0
                                                           9.0
                                                                       12.3
                                                                                  38.4
                                                                                                       1.6
           AL_02_2010
                                2 2010
                                                           10.1
                                                                                  37.7
                        AL
                                              52.1
                                                                       12.6
                                                                                                       1.7
           AL_03_2010
                                3 2010
                                              51.7
                                                           12.5
                                                                       12.1
                                                                                  36.7
                                                                                                       1.6
           AL_04_2010
                                 4 2010
                                              50.6
                                                           8.4
                                                                                  40.2
                                                                                                       1.2
                        AL
                                                                       11.3
           AL_05_2010
                                 5 2010
                                              51.0
                                                           9.8
                                                                       12.4
                                                                                  38.3
                                                                                                       1.2
          5 rows × 24 columns
In [16]:
               # convert foreign / native born population to ratio
               for df in [acs10, acs12, acs14, acs16, acs17]:
                    df.columns = identifier cols + renamed predictors
                    df['foreign to native born ratio'] = df['foreign born population'] / df['nati
                    df.drop(['native born population', 'foreign born population'], axis = 1, inpl
```

```
# use 2017 demographics as 2018 data
In [17]:
             1
                 acs18 = acs17
             2
             3
                 acs18.year = 2018
                 acs18.index = ['\{0\}_{\{1:02d\}_{\{2\}}'.format(row['state'], row['district'], row['year'])
In [18]:
                 acs18
Out[18]:
                         state
                               district year female_pct age18_24_pct age25_34_pct median_age unmarried_partner_pct ma
             AL_01_2018
                           AL
                                    1 2018
                                                   51.8
                                                                 8.2
                                                                              12.9
                                                                                          40.0
                                                                                                                 1.4
             AL_02_2018
                           ΑL
                                    2 2018
                                                   51.2
                                                                 9.7
                                                                              13.0
                                                                                          38.5
                                                                                                                 1.7
             AL_03_2018
                           AL
                                    3 2018
                                                   51.2
                                                                 10.7
                                                                              13.0
                                                                                          38.1
                                                                                                                 2.1
             AL_04_2018
                           AL
                                    4 2018
                                                   51.4
                                                                 8.8
                                                                                          40.7
                                                                                                                 1.3
                                                                              11.7
             AL_05_2018
                           AL
                                    5 2018
                                                   50.8
                                                                 9.3
                                                                              13.2
                                                                                          39.5
                                                                                                                 1.6
             AL_06_2018
                           AL
                                    6 2018
                                                   52.3
                                                                 8.6
                                                                              12.3
                                                                                          39.1
                                                                                                                 1.3
             AL_07_2018
                           AL
                                    7 2018
                                                   52.3
                                                                 12.5
                                                                              14.2
                                                                                          36.3
                                                                                                                 1.6
             AK_01_2018
                                    1 2018
                                                   47.9
                                                                 9.9
                                                                              16.0
                                                                                          34.5
                                                                                                                 2.7
                           ΑK
             AZ_01_2018
                                    1 2018
                                                                              12.2
                           ΑZ
                                                   49.5
                                                                10.7
                                                                                          37.6
                                                                                                                 2.3
             AZ_02_2018
                           ΑZ
                                    2 2018
                                                   50.5
                                                                 9.5
                                                                              12.8
                                                                                          41.4
                                                                                                                 3.0
             AZ_03_2018
                           ΑZ
                                    3 2018
                                                   50.0
                                                                 13.1
                                                                              14.1
                                                                                          31.8
                                                                                                                 2.6
In [19]:
             1
                 # stack 5 dataframes from each year on top of each other
                 demographics data = acs18.append([acs16, acs14, acs12, acs10])
In [20]:
                 demographics data
Out[20]:
                         state
                               district year female_pct age18_24_pct age25_34_pct median_age unmarried_partner_pct ma
             AL_01_2018
                           AL
                                    1 2018
                                                   51.8
                                                                 8.2
                                                                              12.9
                                                                                          40.0
                                                                                                                 1.4
             AL 02 2018
                           AL
                                    2 2018
                                                   51.2
                                                                 9.7
                                                                              13.0
                                                                                          38.5
                                                                                                                 1.7
             AL_03_2018
                           AL
                                    3 2018
                                                   51.2
                                                                 10.7
                                                                              13.0
                                                                                          38.1
                                                                                                                2.1
             AL_04_2018
                                    4 2018
                           AL
                                                   51.4
                                                                 8.8
                                                                              11.7
                                                                                          40.7
                                                                                                                 1.3
             AL_05_2018
                           AL
                                    5 2018
                                                   50.8
                                                                 9.3
                                                                              13.2
                                                                                          39.5
                                                                                                                 1.6
             AL_06_2018
                                    6 2018
                                                   52.3
                                                                 8.6
                                                                              12.3
                                                                                          39.1
                                                                                                                 1.3
             AL_07_2018
                           AL
                                    7 2018
                                                   52.3
                                                                12.5
                                                                              14.2
                                                                                          36.3
                                                                                                                 1.6
             AK_01_2018
                           ΑK
                                    1 2018
                                                   47.9
                                                                 9.9
                                                                              16.0
                                                                                          34.5
                                                                                                                 2.7
             AZ_01_2018
                           ΑZ
                                    1 2018
                                                   49.5
                                                                 10.7
                                                                              12.2
                                                                                          37.6
                                                                                                                 2.3
             AZ_02_2018
                           ΑZ
                                    2 2018
                                                   50.5
                                                                 9.5
                                                                              12.8
                                                                                                                 3.0
                                                                                          41.4
             AZ 03 2018
                                    3 2018
                                                   50.0
                                                                13.1
                           ΑZ
                                                                              14.1
                                                                                          31.8
                                                                                                                 2.6
In [21]:
                 demographics data.to csv('Datasets/demographics data 2010 to 2018.csv')
```

```
In [22]:
             demographics data.columns
Out[22]: Index(['state', 'district', 'year', 'female_pct', 'age18_24_pct',
                 'age25_34_pct', 'median_age', 'unmarried_partner_pct',
                'male_living_alone_pct', 'bachelors_deg_or_higher_pct',
                 'past year births to unmarried women pct', 'civilian veteran pct',
                 'live_same_house_past_year_pct',
                'foreign born proportion from LatinAmerica',
                'speak other language at home pct', 'labor force unemployed pct',
                'public transit commuter pct', 'no health insurance pct',
                 'poverty_rate_pct', 'median_housing_value', 'median_household_income',
                'food stamp benefits pct', 'foreign to native born ratio'],
               dtype='object')
In [23]:
             fec = pickle.load(open('Datasets/data_FEC_NATIONALPOLL_2004_2018.p', 'rb'))
In [24]:
          1 fec.index
Out[24]: Index(['AK_01_2004', 'AL_01_2004', 'AL_02_2004', 'AL_03_2004', 'AL_04_2004',
                 'AL 05 2004', 'AL 06 2004', 'AL 07 2004', 'AR 01 2004', 'AR 02 2004',
                'WV_03_2018', 'WI_01_2018', 'WI_02_2018', 'WI 03 2018', 'WI 04 2018',
                'WI 05 2018', 'WI 06 2018', 'WI 07 2018', 'WI 08 2018', 'WY 01 2018'],
               dtype='object', length=3431)
             demographics_data = demographics_data.drop(['district', 'state', 'year'], axis=1)
In [25]:
In [26]:
          1
             # try joining FEC election data with ACS demographic data
           2
           3
             fec demographics data = fec.join(demographics data, how='inner')
In [27]:
          1 set(fec_demographics_data['year'])
Out[27]: {2010, 2012, 2014, 2016, 2018}
In [28]:
             # pickle.dump(fec demographics data, open('Datasets/data FEC NATIONALPOLL DEMOGRA
             # fec demographics data.to csv('Datasets/data FEC NATIONALPOLL DEMOGRAPHICS 2010
             # test = pickle.load(open('Datasets/data FEC NATIONALPOLL DEMOGRAPHICS 2010 2018.
             # fec demographics data.shape, test.shape
In [29]:
          1 fec.shape, demographics_data.shape
Out[29]: ((3431, 22), (2180, 20))
In [30]:
             fec demographics imputed data = fec.join(demographics data, how='outer')
           3
             for index, row in fec demographics imputed data.iterrows():
                  if index not in fec.index:
           5
                      fec demographics imputed data = fec demographics imputed data[fec demographics]
             fec demographics imputed data.shape
Out[30]: (3431, 42)
In [31]:
          1 assert(sorted(fec_demographics_imputed_data.index) == sorted(fec.index))
```

```
In [32]:
              demographics data.columns
Out[32]: Index(['female_pct', 'age18_24_pct', 'age25_34_pct', 'median_age',
                  'unmarried_partner_pct', 'male_living_alone_pct',
                 'bachelors deg or higher pct',
                 'past_year_births_to_unmarried_women_pct', 'civilian_veteran_pct',
                 'live_same_house_past_year_pct',
                 'foreign born proportion from LatinAmerica',
                 'speak other language at home pct', 'labor force unemployed pct',
                 'public transit commuter pct', 'no health insurance pct',
                 'poverty_rate_pct', 'median_housing_value', 'median_household_income',
                 'food stamp benefits pct', 'foreign to native born ratio'],
                dtype='object')
In [33]:
              for index, row in fec_demographics_imputed_data.iterrows():
           1
                   if row['year'] < 2010:</pre>
           2
           3
                       impute_index = index[:-5] + ' 2010'
           4
                       fec_demographics_imputed_data.loc[fec_demographics_imputed_data.index ==
In [34]:
              fec_demographics_imputed_data[fec_demographics_imputed_data.index == 'AK_01_2010
Out[34]:
                     district state
                                          party candidatevotes totalvotes candidate national_poll national_poll_p
                                  vear
                                                                         Don
          AK_01_2010
                        1.0
                             AK 2010.0 republican
                                                     175384
                                                              254335
                                                                                -2.622642
                                                                                               9.824
                                                                        Young
          1 rows × 42 columns
In [35]:
              fec demographics imputed data[fec demographics imputed data.index == 'AK 01 2004
Out[35]:
                     district state
                                          party candidatevotes totalvotes candidate national_poll national_poll_p
                                  vear
                                                                         Don
          AK_01_2004
                        1.0
                             AK 2004.0 republican
                                                     213216
                                                              299996
                                                                                3.680556
                                                                                               -0.989
                                                                        Young
          1 rows × 42 columns
In [36]:
              for col in demographics_data.columns:
                   if fec_demographics_imputed_data[col].isnull().values.any():
                       raise ValueError
           3
In [37]:
              set(fec demographics imputed data['year'].values)
Out[37]: {2004.0, 2006.0, 2008.0, 2010.0, 2012.0, 2014.0, 2016.0, 2018.0}
In [38]:
              pickle.dump(fec_demographics_imputed_data, open('Datasets/data_FEC_NATIONALPOLL_I
              fec demographics imputed data.to csv('Datasets/data FEC NATIONALPOLL DEMOGRAPHICS
              test = pickle.load(open('Datasets/data FEC NATIONALPOLL DEMOGRAPHICSIMPUTED 2004
              fec demographics imputed data.shape, test.shape
Out[38]: ((3431, 42), (3431, 42))
In [39]:
           1 set(test['year'].values)
Out[39]: {2004.0, 2006.0, 2008.0, 2010.0, 2012.0, 2014.0, 2016.0, 2018.0}
In [40]:
              for col in demographics_data.columns:
           2
                   if test[col].isnull().values.any():
           3
                       raise ValueError
```

```
In [ ]: 1 fec[fec.index == 'WI_06_2018']
In [ ]: 1 demographics_data[demographics_data.index == 'WI_06_2018']
In [ ]: 1 fec_demographics_data[fec_demographics_data.index == 'WI_06_2018']
In [ ]: 1 test.columns
In [ ]: 1 fec.shape
In [ ]: 1 demographics_data.shape
In [ ]: 1 demographics_data.columns
In [ ]: 1
```

9 of 9