### Q1

#### November 4, 2019

### 1 DATASET for Q1

```
[78]: import pandas as pd
     import numpy as np
     import datetime
[79]: #import all datasets
     allergiesCSV = pd.read_csv("../data/allergies.csv")
     observationsCSV = pd.read_csv("../data/observations.csv")
     acsCSV = pd.read_csv("../data/ACS.csv")
     careplansCSV = pd.read_csv("../data/careplans.csv")
     encountersCSV = pd.read_csv("../data/encounters.csv")
     immunizationsCSV = pd.read_csv("../data/immunizations.csv")
     procedutesCSV = pd.read_csv("../data/procedures.csv")
     zipCSV = pd.read_csv("../data/zip_to_zcta_2019.csv")
     conditionsCSV = pd.read_csv("../data/conditions.csv")
     medicationsCSV = pd.read_csv("../data/medications.csv")
     patientsCSV = pd.read_csv("../data/patients.csv")
[80]: | #we only need data from MA, cleaning for necessary data
     massZipCSV = zipCSV[zipCSV.STATE == "MA"]
      →#contains only MA ZCTA
     zipAndIncome = massZipCSV.merge(acsCSV,left_on='ZCTA', right_on='GEO.id2')
     zipAndIncome = zipAndIncome.replace(['2,500-',"***",'-','**'],-1)
      →#necessary for converting to int so it can operated on
[81]: |zipDrop = ['ZIP_CODE', 'PO_NAME', 'STATE', 'ZIP_TYPE', 'zip_join_type', 'GEO.

→id','GEO.id2','GEO.
      →display-label', 'HC01_EST_VC13', 'HC01_MOE_VC13', 'HC02_EST_VC13', 'HC02_MOE_VC13', 'HC03_EST_VC
     zaiMerge = zipAndIncome.copy().drop(zipDrop,axis=1)
     #Values are imported as strings. Converted to ints to run regression
     zaiMerge['HouseholdIncome'] = zipAndIncome['HC01_EST_VC13'].astype(int)
     zaiMerge['HouseholdMOE'] = zipAndIncome['HCO1_MOE_VC13'].astype(int)
      →#zaiMerge is a table with only ZCTA, Household Median Income, & its MOE
     #zaiMerqe['Family'] = zipAndIncome[['HCO2 EST VC13', 'HCO2 MOE VC13']].values.
      →tolist()
```

```
#zaiMerqe['Married'] = zipAndIncome[['HCO3 EST VC13', 'HCO3 MOE VC13']].values.
      →tolist()
     \#zaiMerge['Nonfamily'] = zipAndIncome[['HC04_EST_VC13', 'HC04_MOE_VC13']].values.
      →tolist() #could be interesting for later use
[82]: patientsCSV = patientsCSV.drop(['Unnamed: 0', __
      →'SSN','DRIVERS','PASSPORT','PREFIX','FIRST', 'LAST', 'SUFFIX',

→ 'MAIDEN'],axis=1)
                             #removed data to clean data and make it visually easier.
      → to understand and to ensure privacy
     patientsCSV['HOMEZIP'] = patientsCSV['ADDRESS'].map(lambda x:'0'+str(x)[-7:-3])__
          #lambda fxn grabs zip code. This works since the format an address is _{\sqcup}
      \hookrightarrow fixed
     #maps each cell to respective income through matching zipcodes
     patientsCSV['INCOME'] = patientsCSV['HOMEZIP'].map(lambda x: -1 if x =='0' or x_
      →not in zaiMerge['ZCTA'].unique() else zaiMerge.loc[zaiMerge['ZCTA']==_
      →x]['HouseholdIncome'].values[0])
     patientsCSV['MOE'] = patientsCSV['HOMEZIP'].map(lambda x: -1 if x =='0' or x_
      →not in zaiMerge['ZCTA'].unique() else zaiMerge.loc[zaiMerge['ZCTA']==_
      →x]['HouseholdMOE'].values[0])
     patientsCSV['DEATHDATE'] = patientsCSV['DEATHDATE'].fillna(0)
[83]: emergencyCodes =
      → [50849002,183460006,183452005,183478001,34285007,183495009,32485007,305408004,$05411003]
     emergency = pd.read_csv("../data/encounters.csv")
     emergency = emergency.drop('Unnamed: 0', axis=1)
     #Clean encounters that are too old or not considered an emergency visit
     emergency = emergency[emergency['CODE'].isin(emergencyCodes)]
     emergency = emergency[emergency['DATE'].between('2008', '2017')]
     #last only has most recent encounters for each patient
     last = emergency.drop duplicates("PATIENT", keep='last')
     last['DEATHDATE'] = last['PATIENT'].map(lambda x:__
      →patientsCSV[patientsCSV["ID"]==x]['DEATHDATE'].values[0])
     #Maps date of death for each patient to their encounter
     emergency['DEATHDATE'] = emergency['ID'].map(lambda x: last.
      →loc[last['ID']==x]['DEATHDATE'].values[0] if x in last['ID'].unique() else 0)
     \#I noticed encounters.CSV was missing a lot of reason descriptions that \sqcup
      → conditions. CSV had,
     #so this for loop fills in missing information
     for index,row in emergency.iterrows():
```

```
if isinstance(row['REASONDESCRIPTION'],float):
                                                                              #only_
→applied to encounters that are not filled. will not refill a cell.
       date = row['DATE']
       find = conditionsCSV[conditionsCSV['PATIENT'] == row['PATIENT']]
→has all of rows that pertain to patient and reason narrows it to the
\rightarrow particular encounter
       reason = find[find['START'] == date]['DESCRIPTION']
→#assuming someone doesn't visit the hospital in the same day for a different
\rightarrow reason
       code = find[find['START'] == date]['CODE']
\rightarrowassuming the hospital records all visit and didn't visit double visits if it_\sqcup
\rightarrowoccured
       try:
           emergency['REASONDESCRIPTION'][index] = reason.values[0]
           emergency['REASONCODE'][index] = code.values[0]
       except:
           emergency['REASONDESCRIPTION'][index] = np.nan
           emergency['REASONCODE'][index] = np.nan
```

/usr/local/lib/python3.7/site-packages/ipykernel\_launcher.py:12:
SettingWithCopyWarning:

A value is trying to be set on a copy of a slice from a DataFrame. Try using .loc[row\_indexer,col\_indexer] = value instead

See the caveats in the documentation: http://pandas.pydata.org/pandas-docs/stable/user\_guide/indexing.html#returning-a-view-versus-a-copy if sys.path[0] == '':

/usr/local/lib/python3.7/site-packages/ipykernel\_launcher.py:26: SettingWithCopyWarning:

A value is trying to be set on a copy of a slice from a DataFrame

See the caveats in the documentation: http://pandas.pydata.org/pandas-docs/stable/user\_guide/indexing.html#returning-a-view-versus-a-copy/usr/local/lib/python3.7/site-packages/ipykernel\_launcher.py:27: SettingWithCopyWarning:

A value is trying to be set on a copy of a slice from a DataFrame

See the caveats in the documentation: http://pandas.pydata.org/pandas-docs/stable/user\_guide/indexing.html#returning-a-view-versus-a-copy/usr/local/lib/python3.7/site-packages/ipykernel\_launcher.py:29: SettingWithCopyWarning:

A value is trying to be set on a copy of a slice from a DataFrame

See the caveats in the documentation: http://pandas.pydata.org/pandas-docs/stable/user\_guide/indexing.html#returning-a-view-versus-a-copy/usr/local/lib/python3.7/site-packages/ipykernel\_launcher.py:30: SettingWithCopyWarning:

A value is trying to be set on a copy of a slice from a DataFrame

See the caveats in the documentation: http://pandas.pydata.org/pandas-docs/stable/user\_guide/indexing.html#returning-a-view-versus-a-copy

```
[84]: #Converted to date_time in order to do calculations on the dates
     emergency['DATE'] = pd.to_datetime(emergency['DATE'])
     emergency['DEATHDATE'] = pd.to_datetime(emergency['DEATHDATE'])
     #Finds amount of time between emergency visit and death
     emergency['DEATHDIFFERENCE'] = emergency['DEATHDATE']-emergency['DATE']
     #Dummy variable for all patients who passed away 1 year within visit
     emergency['YearDeath'] = emergency['DEATHDIFFERENCE'].map(lambda x: 1 if_
      →datetime.timedelta(days=0)<x<=datetime.timedelta(days=365) else 0)
     emergency.loc[(emergency['DEATHDIFFERENCE'] < datetime.</pre>
      →timedelta(days=0)), 'DEATHDIFFERENCE'] = np.nan
     \#DEATH is a dummy variable for deaths post discharge, this removes deaths that \sqcup
     →happen on day 0 since that wouldn't be considered post discharge.
     emergency['DEATH'] = emergency['DEATHDIFFERENCE'].map(lambda x: 0 if x ==___
     →datetime.timedelta(days=0) or x!=x else 1)
     #I think this is more indicative of mortality related to post-discharge.
     emergency['DEATH100'] = emergency['DEATHDIFFERENCE'].map(lambda x: 1 if_
      →datetime.timedelta(days=0)<x<=datetime.timedelta(days=100) else 0)
     emergency['DEATH60'] = emergency['DEATHDIFFERENCE'].map(lambda x: 1 if_
      →datetime.timedelta(days=0)<x<=datetime.timedelta(days=60) else 0)
     emergency['DEATH30'] = emergency['DEATHDIFFERENCE'].map(lambda x: 1 if___
      →datetime.timedelta(days=0)<x<=datetime.timedelta(days=30) else 0)
[85]: LOW_LINE = 56763.2 #low income threshole
     #Adding patient information for each encounter. For regression and analysis
     emergency['RACE'] = emergency['PATIENT'].map(lambda x: patientsCSV.
      →loc[patientsCSV['ID']==x]['RACE'].values[0])
     emergency['AGE'] = emergency['PATIENT'].map(lambda x: int(patientsCSV.
      →loc[patientsCSV['ID']==x]['AGE'].values[0]))
     emergency['HOMEZIP'] = emergency['PATIENT'].map(lambda x: patientsCSV.
      →loc[patientsCSV['ID']==x]['HOMEZIP'].values[0])
     emergency['INCOME'] = emergency['PATIENT'].map(lambda x: patientsCSV.
      →loc[patientsCSV['ID']==x]['INCOME'].values[0])
     emergency['MOE'] = emergency['PATIENT'].map(lambda x: patientsCSV.
      →loc[patientsCSV['ID']==x]['MOE'].values[0])
     #Create dummy variables for race for regression
     emergency['BLACK'] = emergency['RACE'].map(lambda x: 1 if x == 'black or_
     →african american' or x=='black' else 0)
     emergency['HISPANIC'] = emergency['RACE'].map(lambda x: 1 if x == 'hispanic'u
      ⇒else 0)
     emergency['ASIAN'] = emergency['RACE'].map(lambda x: 1 if x == 'asian' else 0)
```

### 2 1. Extra Analysis

```
[86]: #Matches patient BMI to patient and encounter

BMI = observationsCSV[observationsCSV['DESCRIPTION']=='Body Mass Index']

BMI["VALUE"] = BMI['VALUE'].astype(float)

patients = BMI['PATIENT'].unique()

emergency['BMI'] = emergency['PATIENT'].map(lambda x:

→BMI[BMI['PATIENT']==x]['VALUE'].values[0] if x in patients else -1)
```

SettingWithCopyWarning:
A value is trying to be set on a copy of a slice from a DataFrame.
Try using .loc[row\_indexer,col\_indexer] = value instead

See the caveats in the documentation: http://pandas.pydata.org/pandas-docs/stable/user\_guide/indexing.html#returning-a-view-versus-a-copy
This is separate from the ipykernel package so we can avoid doing imports until

/usr/local/lib/python3.7/site-packages/ipykernel\_launcher.py:3:

```
[87]: #Homeless dummy variable
homeless = observationsCSV[observationsCSV['DESCRIPTION'] == 'Housing status']
patients = homeless['PATIENT'].unique()
emergency['HOMELESS'] = emergency['PATIENT'].map(lambda x: 1 if x in patients
→else 0)  #Patients with a 'Housing status' description were all
→homeless

[88]: #Variable for Diastolic Blood Pressure
DBP = observationsCSV[observationsCSV['DESCRIPTION'] == 'Diastolic Blood
→Pressure']
```

```
patient = DBP['PATIENT'].unique()
     emergency['DIASTOLICBP'] = emergency['PATIENT'].map(lambda x:__
      →float(DBP[DBP['PATIENT'] == x]['VALUE'].values[0]) if x in patient else -1)
     emergency['BADBP'] = emergency['DIASTOLICBP'].map(lambda x: 1 if x > 80 else 0)
[89]: #Variable for Calcium levels
     Calcium = observationsCSV[observationsCSV['DESCRIPTION']=='Calcium']
     patient = Calcium['PATIENT'].unique()
     emergency['CALCIUM'] = emergency['PATIENT'].map(lambda x:__
      →float(Calcium[Calcium['PATIENT'] ==x]['VALUE'].values[0]) if x in patient
      ⇒else -1)
[90]: #Finds patients who got the flu shot within 6 months of their most recent visiting
     \rightarrow to the hospital.
     emergency['FLU']=None
                                 #1 if patient recieved flu shot w/in 6 months of
      \rightarrow visit, 0 otherwise
     emergency['FLU_DIFF'] = None #difference between time of flu shot and time of_
      →visit, if 0 days then they visited the hospital for the flu shot/happened to⊔
      \rightarrow get it
     flu = immunizationsCSV[immunizationsCSV['CODE'] == 140] #CODE 140 is the code_
      \rightarrow for influenza immunization
     patients = flu['PATIENT'].unique()
     for index, row in emergency.iterrows():
         checked = 0
                         #keeps track of if the inner for loop found a matching
      \rightarrow immunization
         patientID = row['PATIENT']
         if patientID in patients:
                                      #if patient is in the flu dataset, they had anu
      \rightarrow immunization
             date = flu[flu['PATIENT'] == patientID]['DATE'] #list of all dates they_
      →recieved influenza shot
             for d in date:
                 d = datetime.datetime.strptime(d, '%Y-%m-%d')
                 diff = row['DATE'] - d
                 if datetime.timedelta(days=0) <= diff < datetime.
      →timedelta(days=183): #an influenza shot lasts around 6 months/183 days
                     checked = 1
                     emergency['FLU'][index]=1
                      emergency['FLU_DIFF'][index]=diff #can help us determine_
      →cause of visit if there is none (i.e. if =0 days, then the visit was for the
      \rightarrowshot)
             if checked == 0:
                 emergency['FLU'][index]=0
                 emergency['FLU_DIFF'][index]=-1
         else:
```

```
emergency['FLU'][index]=0
emergency['FLU_DIFF'][index]=-1
```

#I found that a lot of the encounter dates in encounters.csv coincided with #dates (found in immunizations.csv) that the patient got a flu shot. They all #coincided with Outpatient Encounters without ReasonDescriptions. So I assume #the reason they went was for the flu shot and not some hidden sickness/  $\rightarrow$  emergency

/usr/local/lib/python3.7/site-packages/ipykernel\_launcher.py:21:
SettingWithCopyWarning:

A value is trying to be set on a copy of a slice from a DataFrame

See the caveats in the documentation: http://pandas.pydata.org/pandas-docs/stable/user\_guide/indexing.html#returning-a-view-versus-a-copy/usr/local/lib/python3.7/site-packages/ipykernel\_launcher.py:22: SettingWithCopyWarning:

A value is trying to be set on a copy of a slice from a DataFrame

See the caveats in the documentation: http://pandas.pydata.org/pandas-docs/stable/user\_guide/indexing.html#returning-a-view-versus-a-copy/usr/local/lib/python3.7/site-packages/ipykernel\_launcher.py:18: SettingWithCopyWarning:

A value is trying to be set on a copy of a slice from a DataFrame

See the caveats in the documentation: http://pandas.pydata.org/pandas-docs/stable/user\_guide/indexing.html#returning-a-view-versus-a-copy/usr/local/lib/python3.7/site-packages/ipykernel\_launcher.py:19: SettingWithCopyWarning:

A value is trying to be set on a copy of a slice from a DataFrame

See the caveats in the documentation: http://pandas.pydata.org/pandas-docs/stable/user\_guide/indexing.html#returning-a-view-versus-a-copy

#### **3 2.** Rates

#### 3.1 2.1 Post Discharge Mortality Rate

```
[24]: seniors = emergency[emergency['SENIOR']==1]
    notSeniors = emergency[emergency['SENIOR']==0]
    minority = emergency[emergency['MINORITY']==1]
    notMinority = emergency[emergency['MINORITY']==0]
    hispanic = emergency[emergency['HISPANIC']==1]
    notHispanic = emergency[emergency['HISPANIC']==0]
    asian = emergency[emergency['ASIAN']==1]
    notAsian = emergency[emergency['ASIAN']==0]
    lowInc = emergency[emergency['LOWINCOME']==1]
```

```
notLowInc = emergency[emergency['LOWINCOME']==0]
#post-discharge 100 day mortalty rate for seniors
seniorMortRate100 = seniors[seniors['DEATH100']==1].shape[0]/seniors.shape[0]
     #3.3%
notMortRate100 = notSeniors[notSeniors['DEATH100'] == 1].shape[0]/notSeniors.
 \rightarrowshape [0]
              #0%
#post-discharge 1 year mortalty rate for seniors
seniorMortRate365 = seniors[seniors['YearDeath']==1].shape[0]/seniors.shape[0]
       #7.1%
notMortRate365 = notSeniors[notSeniors['YearDeath'] == 1].shape[0]/notSeniors.
 →shape[0]
             #0.9%
#post-discharge 1 year mortalty rate for minorities
minorityMortRate = minority[minority['YearDeath']==1].shape[0]/minority.
 \rightarrowshape [0]
                  #2.1%
notMinorityMortRate = notMinority[notMinority['YearDeath']==1].shape[0]/
→notMinority.shape[0] #1.1%
#post-discharge 100 day mortalty rate for hispanic patients
hispanicMortRate100 = hispanic[hispanic['DEATH100']==1].shape[0]/hispanic.
 ⇒shape[0]
                  #0.67%
notHispanicMortRate100 = notHispanic[notHispanic['DEATH100']==1].shape[0]/
 →notHispanic.shape[0] #0.41%
#post-discharge 1 year mortalty rate for hispanic patients
hispanicMortRate = hispanic[hispanic['YearDeath']==1].shape[0]/hispanic.
 \rightarrowshape [0]
                  #2%
notHispanicMortRate = notHispanic[notHispanic['YearDeath']==1].shape[0]/
→notHispanic.shape[0]
                        #2.1%
#post-discharge 100 day mortalty rate for asian patients
asianMortRate100 = asian[asian['DEATH100']==1].shape[0]/asian.shape[0]
                                                                               #1.
 →6%
notAsianMortRate100 = notAsian[notAsian['DEATH100']==1].shape[0]/notAsian.
 →shape[0]
             #0.54%
#post-discharge 1 year mortalty rate for asian patients
asianMortRate = asian[asian['YearDeath']==1].shape[0]/asian.shape[0]
                                                                             #6.
 →5%
notAsianMortRate = notAsian[notAsian['YearDeath']==1].shape[0]/notAsian.
 \rightarrowshape [0]
             #1.7%
#post-discharge 100 day mortalty rate for low income patients
```

```
lowIncMortRate100 = lowInc[lowInc['DEATH100'] == 1] .shape[0]/lowInc.shape[0]

    #.49
notLowIncMortRate100 = notLowInc[notLowInc['DEATH100'] == 1] .shape[0]/notLowInc.
    shape[0] #0.69%

#post-discharge 1 year mortalty rate for low income patients
lowIncMortRate = lowInc[lowInc['YearDeath'] == 1] .shape[0]/lowInc.shape[0]
    #2%
notLowIncMortRate = notLowInc[notLowInc['YearDeath'] == 1] .shape[0]/notLowInc.
    shape[0] #2.1%
```

#### 3.2 2.2 % of Minority, Low Income Patients who are Seniors

#### 3.3 2.3 Characteristics of patients who passed away w/in 1 year of discharge

```
| mortality = emergency[emergency['YearDeath']==1] |
| hispanicMort = mortality[mortality['HISPANIC']==1].shape[0]/mortality.shape[0] |
| #75% of patients who passed away w/in 1 year of discharge were hispanic |
| hispanicRatio = emergency[emergency['HISPANIC']==1].shape[0]/emergency.shape[0] |
| #75.5% |
| seniorMort = mortality[mortality['SENIOR']==1].shape[0]/mortality.shape[0] |
| #65% of patients who passed away w/in 1 year of discharge were seniors |
| seniorRatio = emergency[emergency['SENIOR']==1].shape[0]/emergency.shape[0] |
| #While 65% of YearDeath patients were seniors, only 18.5% of emergency |
| hispanicMort = mortality are seniors |
| incomeRatio = emergency[emergency['LOWINCOME']==1].shape[0]/emergency.shape[0] |
| #41.4% |
| incomeMort = mortality[mortality['LOWINCOME']==1].shape[0]/mortality.shape[0] |
| #40% |
```

```
#6.2% of encounters are Asian, while 20% of mortalities were Asian
asianRatio = emergency[emergency['ASIAN']==1].shape[0]/emergency.shape[0]
                                                                                 ш
asianMort = mortality[mortality['ASIAN']==1].shape[0]/mortality.shape[0]
 →#20%
#8.7% of encounters were black, yet 0% of 1 year discharge mortalities were
\rightarrow b lack
blackRatio = emergency[emergency['BLACK']==1].shape[0]/emergency.shape[0]
blackMort = mortality[mortality['BLACK']==1].shape[0]/mortality.shape[0]
→#0%
#90.4% of encounters were minority and even more (95%) of mortalities were⊔
\rightarrowminority
minorityRatio = emergency[emergency['MINORITY']==1].shape[0]/emergency.shape[0]_u
 →#90.4%
minorityMort = mortality[mortality['MINORITY'] == 1].shape[0]/mortality.shape[0]
 #95%
```

### 3.4 2.4 Characteristics of patients who passed away w/in 30 days of discharge

```
[41]: mortality = emergency[emergency['DEATH30']==1]
     #Similar proportion of mortalities to population of Hispanics
     hispanicMort = mortality[mortality['HISPANIC'] == 1].shape[0]/mortality.shape[0]
      →#75% of patients who passed away w/in 1 year of discharge were hispanic
     hispanicRatio = emergency[emergency['HISPANIC']==1].shape[0]/emergency.shape[0]_u
      →#75.5%
     #While 18.5% of encounters were with seniors, 100% of 30day post discharge
      →mortalities were seniors
     seniorMort = mortality[mortality['SENIOR']==1].shape[0]/mortality.shape[0]
     seniorRatio = emergency[emergency['SENIOR'] ==1].shape[0]/emergency.shape[0]
      →#18.5%
     #More than expected low income encounters passed away within 30 days of \Box
      \rightarrow discharge
     incomeRatio = emergency[emergency['LOWINCOME']==1].shape[0]/emergency.shape[0]
     incomeMort = mortality[mortality['LOWINCOME'] == 1].shape[0]/mortality.shape[0]
      →#50%
     #More asian patient mortalities than expected (6.2% vs 25%)
```

```
asianRatio = emergency[emergency['ASIAN']==1].shape[0]/emergency.shape[0]

$\times #6.2\%$

asianMort = mortality[mortality['ASIAN']==1].shape[0]/mortality.shape[0]

$\times #25\%$

#Less black patient mortalities than expected (8.7\% vs 0\%)

blackRatio = emergency[emergency['BLACK']==1].shape[0]/emergency.shape[0]

$\times #8.7\%$

blackMort = mortality[mortality['BLACK']==1].shape[0]/mortality.shape[0]

$\times #0\%$

#All 30-day post discharge mortalities were minorities

minorityRatio = emergency[emergency['MINORITY']==1].shape[0]/emergency.shape[0]

$\times #90.4\%$

minorityMort = mortality[mortality['MINORITY']==1].shape[0]/mortality.shape[0]

$\times #100\%$
```

#### 3.5 2.5 Characteristics of patients who passed away w/in 60 days of discharge

```
[44]: mortality = emergency[emergency['DEATH60']==1]
     hispanicMort = mortality[mortality['HISPANIC'] == 1].shape[0]/mortality.shape[0]
      →#75% of patients who passed away w/in 1 year of discharge were hispanic
     hispanicRatio = emergency[emergency['HISPANIC'] == 1].shape[0]/emergency.shape[0]_u
      →#75.5%
     #100% of 60day post discharge mortalitis were seniors, they only make up 18.5%
     \rightarrow of the encounters
     seniorMort = mortality[mortality['SENIOR']==1].shape[0]/mortality.shape[0]
      →#100% of patients who passed away w/in 1 year of discharge were seniors
     seniorRatio = emergency[emergency['SENIOR'] ==1].shape[0]/emergency.shape[0]
      →#While 18.5% of YearDeath patients were seniors, only 31.3% of patients are
      ⇒seniors
     incomeRatio = emergency[emergency['LOWINCOME'] == 1].shape[0]/emergency.shape[0] __
      →#41.4%
     incomeMort = mortality[mortality['LOWINCOME'] == 1].shape[0]/mortality.shape[0]
      →#50%
     #More asian patient mortalities than expected
     asianRatio = emergency[emergency['ASIAN']==1].shape[0]/emergency.shape[0]
      →#6.2%
     asianMort = mortality[mortality['ASIAN']==1].shape[0]/mortality.shape[0]
      →#25%
```

```
#Less black patient mortalities than expected
blackRatio = emergency[emergency['BLACK']==1].shape[0]/emergency.shape[0]

$\times #8.7%$
blackMort = mortality[mortality['BLACK']==1].shape[0]/mortality.shape[0]

$\times #0%$

#All 60-day mortalities were minorities
minorityRatio = emergency[emergency['MINORITY']==1].shape[0]/emergency.shape[0]

$\times #90.4%$

minorityMort = mortality[mortality['MINORITY']==1].shape[0]/mortality.shape[0]

$\times #100%$
```

#### 3.6 3. Extra Information (based off of entire dataset (2008-16), not just emergencies)

```
[52]: encounters = pd.read csv("../data/encounters.csv")
     encounters = encounters.drop('Unnamed: 0', axis=1)
     #Clean encounters that are too old
     encounters = encounters[encounters['DATE'].between('2008', '2017')]
     encounters = encounters[encounters['CODE']!=308646001]
     #last only has most recent encounters for each patient
     last = encounters.drop_duplicates("PATIENT", keep='last')
     last['DEATHDATE'] = last['PATIENT'].map(lambda x:__
      →patientsCSV[patientsCSV["ID"] ==x]['DEATHDATE'].values[0])
     #Maps date of death for each patient to their encounter
     encounters['DEATHDATE'] = encounters['ID'].map(lambda x: last.
      →loc[last['ID']==x]['DEATHDATE'].values[0] if x in last['ID'].unique() else 0)
     #I noticed encounters.CSV was missing a lot of reason descriptions that \Box
     \rightarrow conditions. CSV had,
     #so this for loop fills in missing information
     for index.row in encounters.iterrows():
         if isinstance(row['REASONDESCRIPTION'],float):
                                                                                 #only_
      →applied to encounters that are not filled. will not refill a cell.
             date = row['DATE']
             find = conditionsCSV[conditionsCSV['PATIENT'] == row['PATIENT']]
                                                                                 #find
      →has all of rows that pertain to patient and reason narrows it to the
      \rightarrowparticular encounter
             reason = find[find['START'] == date]['DESCRIPTION']
      →#assuming someone doesn't visit the hospital in the same day for a different
             code = find[find['START'] == date]['CODE']
                                                                                 #alson
      →assuming the hospital records all visit and didn't visit double visits if it,
      \rightarrowoccured
```

```
try:
            encounters['REASONDESCRIPTION'][index] = reason.values[0]
            encounters['REASONCODE'][index] = code.values[0]
            encounters['REASONDESCRIPTION'][index] = np.nan
            encounters['REASONCODE'][index] = np.nan
#Converted to date_time in order to do calculations on the dates
encounters['DATE'] = pd.to datetime(encounters['DATE'])
encounters['DEATHDATE'] = pd. to_datetime(encounters['DEATHDATE'])
#Finds amount of time between emergency visit and death
encounters['DEATHDIFFERENCE'] = encounters['DEATHDATE']-encounters['DATE']
#Dummy variable for all patients who passed away 1 year within visit
encounters['YearDeath'] = encounters['DEATHDIFFERENCE'].map(lambda x: 1 if
 datetime.timedelta(days=0)<x<=datetime.timedelta(days=365) else 0)</pre>
encounters.loc[(encounters['DEATHDIFFERENCE'] < datetime.
→timedelta(days=0)), 'DEATHDIFFERENCE'] = np.nan
#DEATH is a dummy variable for deaths post discharge, this removes deaths that
→happen on day 0 since that wouldn't be considered post discharge.
encounters['DEATH'] = encounters['DEATHDIFFERENCE'].map(lambda x: 0 if x ==___
 →datetime.timedelta(days=0) or x!=x else 1)
#I think this is more indicative of mortality related to post-discharge.
encounters['DEATH100'] = encounters['DEATHDIFFERENCE'].map(lambda x: 1 if
 →datetime.timedelta(days=0)<x<=datetime.timedelta(days=100) else 0)
encounters['DEATH60'] = encounters['DEATHDIFFERENCE'].map(lambda x: 1 if___

datetime.timedelta(days=0) < x < = datetime.timedelta(days=60) else 0)
</pre>
encounters['DEATH30'] = encounters['DEATHDIFFERENCE'].map(lambda x: 1 if___
 →datetime.timedelta(days=0)<x<=datetime.timedelta(days=30) else 0)
LOW_LINE = 56763.2 #low income threshole
#Adding patient information for each encounter. For regression and analysis
encounters['RACE'] = encounters['PATIENT'].map(lambda x: patientsCSV.
 →loc[patientsCSV['ID']==x]['RACE'].values[0])
encounters['AGE'] = encounters['PATIENT'].map(lambda x: int(patientsCSV.
 →loc[patientsCSV['ID']==x]['AGE'].values[0]))
encounters['HOMEZIP'] = encounters['PATIENT'].map(lambda x: patientsCSV.
 →loc[patientsCSV['ID']==x]['HOMEZIP'].values[0])
encounters['INCOME'] = encounters['PATIENT'].map(lambda x: patientsCSV.
 →loc[patientsCSV['ID']==x]['INCOME'].values[0])
#Dummy variable for if patient is considered low income. Low income is 80\% of
→state median
encounters['LOWINCOME'] = encounters['INCOME'].map(lambda x: 1 if x < LOW_LINE_
 ⇒else 0)
```

```
#Replacing NA was necessary for the above lambda fxn, however NA is necessary_{\sqcup}
→ for regression since 0 will affect the regression
encounters['INCOME'] = encounters['INCOME'].replace(-1,np.nan)
encounters['SENIOR'] = encounters['AGE'].map(lambda x: 1 if x>=65 else 0)
encounters['MINORITY'] = encounters['RACE'].map(lambda x: 1 if x=='asian' or___
 →x=='hispanic' or x=='black' or x=='black or african american' else 0)
#Matches patient BMI to patient and encounter
BMI = observationsCSV[observationsCSV['DESCRIPTION'] == 'Body Mass Index']
BMI["VALUE"] = BMI['VALUE'].astype(float)
patients = BMI['PATIENT'].unique()
encounters['BMI'] = encounters['PATIENT'].map(lambda x:
→BMI[BMI['PATIENT'] == x]['VALUE'].values[0] if x in patients else -1)
#Homeless dummy variable
homeless = observationsCSV[observationsCSV['DESCRIPTION'] == 'Housing status']
patients = homeless['PATIENT'].unique()
encounters['HOMELESS'] = encounters['PATIENT'].map(lambda x: 1 if x in patients_
 ⇔else 0)
                 \#Patients with a 'Housing status' description were all_\sqcup
homeless
#Variable for Diastolic Blood Pressure
DBP = observationsCSV[observationsCSV['DESCRIPTION'] == 'Diastolic Bloodu
 →Pressure'
patient = DBP['PATIENT'].unique()
encounters['DIASTOLICBP'] = encounters['PATIENT'].map(lambda x:
 →float(DBP[DBP['PATIENT'] == x]['VALUE'].values[0]) if x in patient else -1)
encounters['BADBP'] = encounters['DIASTOLICBP'].map(lambda x: 1 \text{ if } x > 80 \text{ else}_{\sqcup}
 →0)
#Variable for Calcium levels
Calcium = observationsCSV[observationsCSV['DESCRIPTION']=='Calcium']
patient = Calcium['PATIENT'].unique()
encounters['CALCIUM'] = encounters['PATIENT'].map(lambda x:
 →float(Calcium[Calcium['PATIENT'] ==x]['VALUE'].values[0]) if x in patient__
⇒else -1)
#Finds patients who got the flu shot within 6 months of their most recent visit⊔
\rightarrow to the hospital.
encounters['FLU']=None
                             #1 if patient recieved flu shot w/in 6 months of
\rightarrow visit, 0 otherwise
→visit, if 0 days then they visited the hospital for the flu shot/happened to⊔
 \rightarrow get it
```

```
flu = immunizationsCSV[immunizationsCSV['CODE'] == 140] #CODE 140 is the code_
 → for influenza immunization
patients = flu['PATIENT'].unique()
for index, row in encounters.iterrows():
     checked = 0
                     #keeps track of if the inner for loop found a matching_
 \rightarrow immunization
    patientID = row['PATIENT']
    if patientID in patients:
                                  #if patient is in the flu dataset, they had anu
 \rightarrow immunization
         date = flu[flu['PATIENT'] == patientID] ['DATE'] #list of all dates they_
 \rightarrowrecieved influenza shot
        for d in date:
             d = datetime.datetime.strptime(d, '%Y-%m-%d')
             diff = row['DATE'] - d
             if datetime.timedelta(days=0) <= diff < datetime.</pre>
                            #an influenza shot lasts around 6 months/183 days
 →timedelta(days=183):
                 checked = 1
                 encounters['FLU'][index]=1
                 encounters['FLU_DIFF'][index]=diff #can help us determine_
 →cause of visit if there is none (i.e. if =0 days, then the visit was for the
 \rightarrowshot)
         if checked == 0:
             encounters['FLU'][index]=0
             encounters['FLU_DIFF'][index]=-1
    else:
             encounters['FLU'][index]=0
             encounters['FLU_DIFF'][index]=-1
#I found that a lot of the encounter dates in encounters.csv coincided with
#dates (found in immunizations.csv) that the patient got a flu shot. They all
#coincided with Outpatient Encounters without ReasonDescriptions. So I assume
#the reason they went was for the flu shot and not some hidden sickness/
 \rightarrowemergency
/usr/local/lib/python3.7/site-packages/ipykernel_launcher.py:10:
```

```
SettingWithCopyWarning:
A value is trying to be set on a copy of a slice from a DataFrame.
Try using .loc[row_indexer,col_indexer] = value instead

See the caveats in the documentation: http://pandas.pydata.org/pandas-docs/stable/user_guide/indexing.html#returning-a-view-versus-a-copy
# Remove the CWD from sys.path while we load stuff.
/usr/local/lib/python3.7/site-packages/ipykernel_launcher.py:27:
SettingWithCopyWarning:
A value is trying to be set on a copy of a slice from a DataFrame
```

See the caveats in the documentation: http://pandas.pydata.org/pandas-docs/stable/user\_guide/indexing.html#returning-a-view-versus-a-copy/usr/local/lib/python3.7/site-packages/ipykernel\_launcher.py:28: SettingWithCopyWarning:

A value is trying to be set on a copy of a slice from a DataFrame

See the caveats in the documentation: http://pandas.pydata.org/pandas-docs/stable/user\_guide/indexing.html#returning-a-view-versus-a-copy/usr/local/lib/python3.7/site-packages/ipykernel\_launcher.py:24: SettingWithCopyWarning:

A value is trying to be set on a copy of a slice from a DataFrame

See the caveats in the documentation: http://pandas.pydata.org/pandas-docs/stable/user\_guide/indexing.html#returning-a-view-versus-a-copy/usr/local/lib/python3.7/site-packages/ipykernel\_launcher.py:25: SettingWithCopyWarning:

A value is trying to be set on a copy of a slice from a DataFrame

See the caveats in the documentation: http://pandas.pydata.org/pandas-docs/stable/user\_guide/indexing.html#returning-a-view-versus-a-copy/usr/local/lib/python3.7/site-packages/ipykernel\_launcher.py:63: SettingWithCopyWarning:

A value is trying to be set on a copy of a slice from a DataFrame. Try using .loc[row\_indexer,col\_indexer] = value instead

See the caveats in the documentation: http://pandas.pydata.org/pandas-docs/stable/user\_guide/indexing.html#returning-a-view-versus-a-copy/usr/local/lib/python3.7/site-packages/ipykernel\_launcher.py:100: SettingWithCopyWarning:

A value is trying to be set on a copy of a slice from a DataFrame

See the caveats in the documentation: http://pandas.pydata.org/pandas-docs/stable/user\_guide/indexing.html#returning-a-view-versus-a-copy/usr/local/lib/python3.7/site-packages/ipykernel\_launcher.py:101: SettingWithCopyWarning:

A value is trying to be set on a copy of a slice from a DataFrame

See the caveats in the documentation: http://pandas.pydata.org/pandas-docs/stable/user\_guide/indexing.html#returning-a-view-versus-a-copy/usr/local/lib/python3.7/site-packages/ipykernel\_launcher.py:103: SettingWithCopyWarning:

A value is trying to be set on a copy of a slice from a DataFrame

See the caveats in the documentation: http://pandas.pydata.org/pandas-docs/stable/user\_guide/indexing.html#returning-a-view-versus-a-copy/usr/local/lib/python3.7/site-packages/ipykernel\_launcher.py:104: SettingWithCopyWarning:

A value is trying to be set on a copy of a slice from a DataFrame

See the caveats in the documentation: http://pandas.pydata.org/pandas-docs/stable/user\_guide/indexing.html#returning-a-view-versus-a-copy/usr/local/lib/python3.7/site-packages/ipykernel\_launcher.py:106: SettingWithCopyWarning:

A value is trying to be set on a copy of a slice from a DataFrame

See the caveats in the documentation: http://pandas.pydata.org/pandas-docs/stable/user\_guide/indexing.html#returning-a-view-versus-a-copy/usr/local/lib/python3.7/site-packages/ipykernel\_launcher.py:107: SettingWithCopyWarning:

A value is trying to be set on a copy of a slice from a DataFrame

See the caveats in the documentation: http://pandas.pydata.org/pandas-docs/stable/user\_guide/indexing.html#returning-a-view-versus-a-copy

```
[51]: encounters.to_csv("En.csv")
```

46% of senior patients who passed away w/in 30 days of their visit did not visit for care. They visited for a flu shot.

39% of senior patients who passed away w/in 30 days of their visit passed away due to Pneumonia/Viral Illness.

46.15384615384615% of senior patients who passed away within 30 days of their hospital visit visited the hospital for a flu shot 38.46153846153847% of senior patients who passed away within 30 days of visit, visited for Pneumonia/Viral Illness

```
[56]: ID DATE \
669 46b94251-1039-42a9-ae54-3ca75f25e75c 2012-07-07
```

```
1924
       d8ed4921-783c-4e10-bbc9-e8dbd5913c20 2013-04-09
2149
       b297ac62-23e4-4fe7-8f72-8277722eaf52 2009-02-25
3407
       ca457bec-18f4-4ce0-81f9-0800b346428d 2016-07-13
4180
       b8a9dcac-1b61-49ae-9e06-4c793be6e29d 2009-04-16
5761
       508aebb9-5b6c-42fc-990d-3f399b7ba386 2011-11-02
6063
       e7d1b9ab-a5c4-4d0b-ac86-f6582496195e 2014-01-25
7120
       977f504f-1ec5-45c7-aa16-b0010a62745f 2015-02-02
10994
       bb71193a-8aa3-499a-8d18-223e50978588 2014-09-22
11942
       beb50d63-b424-4b33-8315-9c1e4a7ad479 2016-07-02
13148
       2925ea65-0f13-41e7-9e5e-d6165167f831 2015-12-10
13285
       d9a9ddf9-5a40-49b1-9a0d-693bc98e1431 2009-04-26
20345
       736e13f7-4dc5-4e49-8e0e-2e28115fca12 2013-01-05
                                     PATIENT
                                                    CODE
                                                          \
669
       c383f814-9ee4-4ca8-a0ff-ee1369f8d2ee
                                               185349003
1924
       6d66c6a6-8b0d-47c0-a573-b78ec8ceec63
                                               185349003
2149
       481373a7-79df-429d-b3da-e971116c1df1
                                               185349003
3407
       bc6fbe62-116e-424f-943c-bae29fa9f319
                                               185345009
4180
       bc2f58a2-ea29-45e7-93c4-99c03cbfcc4c
                                               185349003
5761
       90e7f959-f0ec-4e9e-bce8-ca0a7d0e6a1f
                                               185349003
6063
       cdbe1927-b954-4f37-9294-ac0ca277d147
                                                34285007
7120
       8b3cf0c5-affd-4cbb-a397-e63484ba4d47
                                                34285007
10994
      d871e7f4-b18d-4e4a-8fbb-f1d2c101efc5
                                               185349003
11942
       e6e63adb-9c2a-4f7a-ade9-3b89ff862a03
                                                34285007
13148
       af7c4d94-3524-4c7b-b8c0-fb774c9e5b6d
                                                34285007
13285
       e3c8d2c8-570d-4c69-a81e-bd88d6c6b0ca
                                               185347001
20345
       2bbaedc8-d765-4950-bfc2-32b16713934c
                                               185349003
                    DESCRIPTION
                                   REASONCODE
                                                         REASONDESCRIPTION
669
           Outpatient Encounter
                                          NaN
                                                                        NaN
1924
                                          NaN
           Outpatient Encounter
                                                                        NaN
2149
           Outpatient Encounter
                                          NaN
                                                                        NaN
          Encounter for symptom
                                               Viral sinusitis (disorder)
3407
                                  444814009.0
4180
           Outpatient Encounter
                                          NaN
                                                                        NaN
5761
           Outpatient Encounter
                                          NaN
                                                                        NaN
6063
             Hospital admission
                                  233604007.0
                                                                 Pneumonia
7120
             Hospital admission
                                  233604007.0
                                                                 Pneumonia
10994
           Outpatient Encounter
                                                                       NaN
                                          NaN
11942
             Hospital admission
                                  233604007.0
                                                                 Pneumonia
13148
             Hospital admission
                                  233604007.0
                                                                 Pneumonia
          Encounter for problem
                                                                        NaN
13285
                                          NaN
20345
      Encounter for 'check-up'
                                   65966004.0
                                                       Fracture of forearm
       DEATHDATE DEATHDIFFERENCE YearDeath
                                                    LOWINCOME
                                                               SENTOR.
                                                                       MTNORTTY
669
                                                            0
                                                                     1
      2012-08-02
                          26 days
                                            1
                                                                               1
                                               . . .
1924
                          22 days
                                                            0
                                                                     1
                                                                               1
      2013-05-01
                                            1
                                               . . .
                                                                     1
2149
      2009-03-18
                          21 days
                                            1
                                               . . .
                                                            1
                                                                               1
```

3407	2016-07	<b>'-</b> 22	9	days		1			1		1	0
4180	2009-05	5-11	25	days		1			0		1	0
5761	2011-11	-12	10	days		1			0		1	1
6063	2014-02	2-06	12	days		1			1		1	1
7120	2015-02	2-15	13	days		1			0		1	1
10994	2014-10	0-01	9	days		1			0		1	1
11942	2016-07	<b>'</b> -10	8	days		1			0		1	1
13148	2015-12	2-21	11	days		1			1		1	1
13285	2009-05	5-23	27	days		1			0		1	1
20345	2013-01	25	20	days		1			0		1	1
	BMI	HOMELESS	DIAST	TOLICBP	BADBP	CA	LCIUM	FLU			FLU_DIFF	•
669	37.40	0		72.0	0		-1.00	1	0	days	00:00:00	
1924	39.06	0		116.0	1		9.45	1	0	days	00:00:00	
2149	30.24	0		111.0	1		10.09	1	0	days	00:00:00	
3407	36.44	0		114.0	1		9.19	0			-1	
4180	35.59	0		111.0	1		9.51	1	0	days	00:00:00	)
5761	31.78	0		75.0	0		9.30	1	0	days	00:00:00	
6063	25.15	0		80.0	0		8.65	1	121	days	00:00:00	
7120	37.09	0		97.0	1		9.27	1	101	days	00:00:00	
10994	25.89	0		71.0	0		-1.00	1	0	days	00:00:00	
11942	35.13	0		111.0	1		9.04	0			-1	
13148	35.71	0		77.0	0		-1.00	0			-1	
13285	41.13	0		82.0	1		10.04	0			-1	
20345	23.84	0		72.0	0		9.09	1	63	days	00:00:00	

[13 rows x 28 columns]

57% of senior patients who passed away w/in 60 days of their hospital visit visited the hospital for a flu shot. An additional 32% passed away after visiting for Pneumonia/Viral Sinusitis.

```
Death60 = seniorPatients[seniorPatients['DEATH60']==1]

print(str(Death60[Death60['FLU_DIFF']==datetime.timedelta(days=0)].shape[0]/

Death60.shape[0] * 100) + "% of senior patients who passed away within 60

days of their hospital visit visited the hospital for a flu shot")

print(str(6/Death60.shape[0]*100) + "% of senior patients passed away after

visiting the hospital for a virus")

Death60
```

57.89473684210527% of senior patients who passed away within 60 days of their hospital visit visited the hospital for a flu shot 31.57894736842105% of senior patients passed away after visiting the hospital for a virus

```
[59]: ID DATE \
669 46b94251-1039-42a9-ae54-3ca75f25e75c 2012-07-07
```

```
1924
       d8ed4921-783c-4e10-bbc9-e8dbd5913c20 2013-04-09
2149
       b297ac62-23e4-4fe7-8f72-8277722eaf52 2009-02-25
3334
       b57f2d3d-e9fc-4d21-be1e-e9b2f84fc714 2014-06-15
3407
       ca457bec-18f4-4ce0-81f9-0800b346428d 2016-07-13
4180
       b8a9dcac-1b61-49ae-9e06-4c793be6e29d 2009-04-16
4530
       12b94e93-09fe-4dcc-98d0-ac67ce07e674 2009-04-03
5761
       508aebb9-5b6c-42fc-990d-3f399b7ba386 2011-11-02
6063
       e7d1b9ab-a5c4-4d0b-ac86-f6582496195e 2014-01-25
6821
       04e0b540-cfaf-4875-895f-8cdb3235a48e 2008-06-20
6851
       4f64dddf-122d-4da0-9b4b-eeaf7e1e0f83 2009-02-11
7120
       977f504f-1ec5-45c7-aa16-b0010a62745f 2015-02-02
9856
       62662cef-3923-43ac-a677-ca370e852dc0 2013-12-15
10994
       bb71193a-8aa3-499a-8d18-223e50978588 2014-09-22
11942
       beb50d63-b424-4b33-8315-9c1e4a7ad479 2016-07-02
12676
       956178d5-c90e-4f0b-8713-263fef438716 2016-03-26
13148
       2925ea65-0f13-41e7-9e5e-d6165167f831 2015-12-10
13285
       d9a9ddf9-5a40-49b1-9a0d-693bc98e1431 2009-04-26
20345
       736e13f7-4dc5-4e49-8e0e-2e28115fca12 2013-01-05
                                     PATIENT
                                                   CODE
669
                                              185349003
       c383f814-9ee4-4ca8-a0ff-ee1369f8d2ee
                                              185349003
1924
       6d66c6a6-8b0d-47c0-a573-b78ec8ceec63
2149
       481373a7-79df-429d-b3da-e971116c1df1
                                              185349003
3334
       18a225df-378e-419a-8aff-bc03ab654103
                                              185349003
3407
       bc6fbe62-116e-424f-943c-bae29fa9f319
                                              185345009
4180
       bc2f58a2-ea29-45e7-93c4-99c03cbfcc4c
                                              185349003
4530
       a77d3e00-6f1a-4bfd-9333-3439be45b15f
                                              185349003
5761
       90e7f959-f0ec-4e9e-bce8-ca0a7d0e6a1f
                                              185349003
6063
       cdbe1927-b954-4f37-9294-ac0ca277d147
                                               34285007
6821
       6ddea441-48d4-4ef1-8213-04ca25c78497
                                              185349003
6851
       6f45458c-8658-449d-a583-effa43cbd8a8
                                              185345009
7120
       8b3cf0c5-affd-4cbb-a397-e63484ba4d47
                                               34285007
9856
       28627be3-81b6-4917-bd4a-d2af28c5ac5a
                                              185349003
10994
       d871e7f4-b18d-4e4a-8fbb-f1d2c101efc5
                                              185349003
11942
       e6e63adb-9c2a-4f7a-ade9-3b89ff862a03
                                               34285007
12676
       a83957db-3e38-4fd4-a4bb-aca5f2090b80
                                              185349003
13148
       af7c4d94-3524-4c7b-b8c0-fb774c9e5b6d
                                               34285007
13285
       e3c8d2c8-570d-4c69-a81e-bd88d6c6b0ca
                                              185347001
20345
       2bbaedc8-d765-4950-bfc2-32b16713934c
                                              185349003
                    DESCRIPTION
                                  REASONCODE
           Outpatient Encounter
669
                                          NaN
1924
           Outpatient Encounter
                                          NaN
2149
           Outpatient Encounter
                                          NaN
3334
           Outpatient Encounter
                                          NaN
3407
          Encounter for symptom
                                 444814009.0
4180
           Outpatient Encounter
                                          NaN
```

4530	Outpatient Encounter NaN						
5761	Outpatient Encounter NaN						
6063	Hospital admission 233604007.0						
6821	Outpatient Encounter NaN						
6851	Encounter for symptom 195662009.0						
7120	Hospital admission 233604007.0						
9856	Outpatient Encounter NaN						
10994	Outpatient Encounter NaN						
11942	Hospital admission 233604007.0						
12676	•						
13148	•						
13285	1						
20345	Encounter for 'check-up' 65966004.0						
	REASONDESCRIPTION DEATHDATE DEATHDIFFERI	ENCE \					
669	NaN 2012-08-02 26 d						
1924		days					
2149		days					
3334	NaN 2014-07-26 41 d	lays					
3407	Viral sinusitis (disorder) 2016-07-22 9 o	days					
4180	NaN 2009-05-11 25 d	days					
4530	NaN 2009-05-15 42 d	days					
5761	NaN 2011-11-12 10 d	days					
6063	Pneumonia 2014-02-06 12 d	days					
6821		days					
6851		days					
7120		days					
9856		days					
10994		days					
11942		days					
12676 13148		days days					
13285		lays lays					
20345	1.01. 2000 00 20	-					
		<i>j</i>					
	YearDeath LOWINCOME SENIOR MINORITY BMI HOMELES	BS \					
669	1 0 1 1 37.40	0					
1924	1 0 1 1 39.06	0					
2149	1 1 1 30.24	0					
3334	1 0 1 0 25.35	0					
3407	1 1 1 0 36.44	0					
4180 4530	1 0 1 0 35.59 1 0 1 1 31.78	0					
4530 5761	1 0 1 1 31.78	0					
6063	1 1 1 1 25.15	0					
6821	1 0 1 1 27.58	0					
6851	1 0 1 1 30.56	0					
		-					

```
7120
                                   0
                                            1
                                                          37.09
                                                                         0
                 1
9856
                                                          39.09
                                                                         0
                                   0
                                            1
                                                       1
                 1
10994
                 1
                    . . .
                                   0
                                            1
                                                       1
                                                          25.89
                                                                         0
                                                          35.13
11942
                 1
                                   0
                                                                         0
                    . . .
12676
                                   0
                                            1
                                                          38.02
                                                                         0
                 1
                                                       1
13148
                 1
                                   1
                                            1
                                                       1
                                                          35.71
                                                                         0
                    . . .
13285
                                   0
                                            1
                                                          41.13
                                                                         0
                 1
                                                       1
                    . . .
                                   0
                                            1
20345
                 1
                                                       1
                                                          23.84
                                                                         0
                    . . .
       DIASTOLICBP BADBP
                             CALCIUM
                                       FLU
                                                       FLU_DIFF
669
               72.0
                               -1.00
                                               0 days 00:00:00
                          0
                                          1
1924
              116.0
                          1
                                9.45
                                         1
                                               0 days 00:00:00
2149
              111.0
                          1
                               10.09
                                               0 days 00:00:00
3334
               87.0
                          1
                                8.55
                                          1
                                               0 days 00:00:00
3407
              114.0
                          1
                                9.19
                                          0
                                                              -1
4180
              111.0
                          1
                                9.51
                                          1
                                               0 days 00:00:00
               86.0
                               -1.00
                                               0 days 00:00:00
4530
                          1
                                         1
5761
                75.0
                          0
                                9.30
                                         1
                                               0 days 00:00:00
               80.0
                                8.65
6063
                          0
                                             121 days 00:00:00
6821
                88.0
                               -1.00
                                               0 days 00:00:00
                          1
               91.0
                                8.60
                                              33 days 00:00:00
6851
                          1
                                         1
7120
               97.0
                          1
                                9.27
                                             101 days 00:00:00
9856
               78.0
                          0
                                9.84
                                               0 days 00:00:00
                                         1
10994
               71.0
                               -1.00
                                               0 days 00:00:00
                          0
                                         1
11942
              111.0
                          1
                                9.04
                                         0
                                                              -1
12676
               85.0
                          1
                               -1.00
                                         1
                                               0 days 00:00:00
                77.0
                               -1.00
13148
13285
               82.0
                          1
                               10.04
                                         0
                                                              -1
20345
               72.0
                          0
                               9.09
                                         1
                                              63 days 00:00:00
```

[19 rows x 28 columns]

#### Finding areas/zipcodes that might benefit most/first from the program

```
[501]: yeardeath = encounters[encounters['DEATH']==1]
  yeardeath = yeardeath.drop_duplicates('PATIENT',keep='last')

locations = yeardeath.groupby('HOMEZIP').count()
  locations.sort_values('PATIENT').tail(10)

emergency_drop = encounters.drop_duplicates('PATIENT', keep='last')

m = emergency_drop[emergency_drop['MINORITY']==1]
  m = m.groupby("HOMEZIP").count()
  top = m.sort_values('PATIENT').tail(10)
```

83.6% of YearDeath==1 seniors were not w/in the healthy BMI range for seniors.

```
[60]: | yearSeniors = seniorPatients[seniorPatients['YearDeath']==1]
     totalYrSeniors = yearSeniors.shape[0]
     under = yearSeniors[yearSeniors['BMI']<25].shape[0]</pre>
     over =yearSeniors[yearSeniors['BMI']>27].shape[0]
     print(str((under+over)/totalYrSeniors *100) +"% of 1 year mortality senior ⊔
      ⇒patients were not within the healthy BMI range for seniors")
     print(str(under/totalYrSeniors*100)+"% of 1 year mortality senior patients were
      →under the healthy BMI range for seniors")
     print(str(over/totalYrSeniors*100)+"% of 1 year mortality senior patients were⊔
      →over the healthy BMI range for seniors")
     yearSeniors = seniorPatients[seniorPatients['YearDeath']==1]
     totalYrSeniors = yearSeniors.shape[0]
     under = yearSeniors[yearSeniors['BMI']<25].shape[0]</pre>
     over =yearSeniors[yearSeniors['BMI']>27].shape[0]
     print(str((under+over)/totalYrSeniors *100) +"% of 1 year mortality senior ⊔
      →patients were not within the healthy BMI range for seniors")
     print(str(under/totalYrSeniors*100)+"% of 1 year mortality senior patients were
      →under the healthy BMI range for seniors")
     print(str(over/totalYrSeniors*100)+"% of 1 year mortality senior patients were
      →over the healthy BMI range for seniors")
```

```
83.56164383561644% of 1 year mortality senior patients were not within the healthy BMI range for seniors
17.80821917808219% of 1 year mortality senior patients were under the healthy BMI range for seniors
65.75342465753424% of 1 year mortality senior patients were over the healthy BMI range for seniors
83.56164383561644% of 1 year mortality senior patients were not within the healthy BMI range for seniors
17.80821917808219% of 1 year mortality senior patients were under the healthy BMI range for seniors
65.75342465753424% of 1 year mortality senior patients were over the healthy BMI range for seniors
```

#### 66.9% of YearDeath==1 seniors did not have normal diastolic blood pressure.

```
[62]: print(str(yearSeniors[yearSeniors['BADBP']==1].shape[0]/totalYrSeniors*100)+"%

of 1 year mortality seniors did not have normal diastolic blood pressure")

print(str(seniorPatients[seniorPatients['BADBP']==1].shape[0]/

ototalSeniors*100)+"% of the entire senior patient population does not have

onormal diastolic blood pressure")
```

```
print(str(yearSeniors[yearSeniors['DIASTOLICBP']>=90].shape[0]/

→totalYrSeniors*100) +"% of 1 year mortality seniors had high diastolic blood

→pressure")

print(str(seniorPatients[seniorPatients['DIASTOLICBP']>=90].shape[0]/

→totalSeniors*100) +"% of the entire senior patient population has high

→diastolic blood pressure")
```

64.38356164383562% of 1 year mortality seniors did not have normal diastolic blood pressure

62.447257383966246% of the entire senior patient population does not have normal diastolic blood pressure

35.61643835616438% of 1 year mortality seniors had high diastolic blood pressure 33.755274261603375% of the entire senior patient population has high diastolic blood pressure

# 11/15 YearDeaths for minority, senior patients were related to curable viruses or injuries sustained from falling.

```
[70]: e = encounters[encounters['YearDeath']==1]
g = e[e['MINORITY']==1]
g = g[g['SENIOR']==1]
h = g.drop_duplicates('PATIENT',keep='last')
g = h.groupby('REASONDESCRIPTION').count()

print("8/14 of known reasons for passing was due to curable viruses and
→sicknesses")
print("3/14 of known reasons were related to injuries from falling")
```

8/14 of known reasons for passing was due to curable viruses and sicknesses 3/14 of known reasons were related to injuries from falling

## The most recent hospital visit for 9/16 senior, minority, patients who passed away w/in 60 days of their visits went to the hospital for a flu shot

```
[74]: e = encounters[encounters['DEATH60']==1]
g = e[e['MINORITY']==1]
g = g[g['SENIOR']==1]
h = g.drop_duplicates('PATIENT',keep='last')
h.groupby('REASONDESCRIPTION').count()
print("9/16 senior, minority, patients who passed away w/in 30 days visited the

→hospital for flu shots")
```

9/16 senior, minority, patients who passed away w/in 30 days visited the hospital for flu shots

# 92.8% of Outpatient Encounters were influenza immunizations/included an influenza immunization

```
[76]: e_drop = encounters.drop_duplicates("PATIENT", keep='last')
outpatient = e_drop[e_drop['CODE']==185349003]
outFluSize = outpatient[outpatient["FLU_DIFF"]==datetime.timedelta(days=0)].

→shape[0]
print(str(outFluSize/outpatient.shape[0]*100)+'% of Outpatient Encounters were/
→included an influenza immunization')
```

92.81767955801105% of Outpatient Encounters were/included an influenza immunization

## Avg. Household Income for a Homeless Patient = 76,309.20. That's 34.4% more than the low income threshold.

The average estimated household income for a homeless patient is \$76309.28 which is 34.43442230177298% more than the low income threshold

[95]:		Unnamed: 0	DATE	PATIENT	\
	202	202	2013-12-16	96b24072-e1fe-49cd-a22a-6dfb92c3994c	
	823	823	2014-07-04	38364c57-80ce-4749-aed6-878cdff95379	
	1465	1465	2007-11-04	abf99602-7cb7-49db-9251-0499968c472f	
	3986	3986	2008-05-20	9943efb3-15d2-4225-ba54-0cb4d4478f6a	
	4256	4256	2015-02-21	41ec5505-df3f-4e0c-9e4c-2cc7e152031e	
	10881	10881	2013-04-16	63b2d7b3-c597-4ede-b477-9b0515799cc8	
	12216	12216	2010-12-29	bc8af009-ec20-409e-813d-fba6a952a2ab	
	16128	16128	2016-11-29	24de5840-c471-4436-93ef-3e5b3e905353	
	16216	16216	2009-07-28	d2e9efc1-431e-4736-8823-e86c16dcb141	
	17213	17213	2011-11-24	af9cd97a-f11a-469a-8781-84e88ac83774	
	20485	20485	2017-02-14	9016a427-917f-4729-ac9c-ef865f17f4d7	
	20737	20737	2010-10-24	186ce071-096a-4eed-9c61-dd773ccf3830	
	22775	22775	2014-07-30	932563cc-7bc0-46b3-8b36-d687202f0705	
	24061	24061	2014-09-05	86082f75-6af5-4df0-a8f7-effd9bd68a58	
	27000	27000	2013-05-13	dbd3794c-5ecc-47bc-8d6a-c8d60fbe9d31	

TOF1

```
27506
            27506
                   2009-08-09
                               4243c357-7791-4eac-a6db-4d18f561c910
29256
            29256
                               6e8b250f-3ce4-442c-96d7-08786c74a966
                   2012-12-27
31687
            31687
                   2017-10-03
                               c95416a1-5c72-4588-9dd4-ff1959ef7687
33982
            33982
                   2013-02-17
                               260e1f37-18b0-4f2b-9a2c-78645485bf1e
                   2008-11-14
35790
            35790
                               0ed686b8-877a-430b-a1a6-5d0be2d6bebc
37416
            37416
                   2012-01-13
                               de5deae8-3379-4bd1-8cf7-e48b5e61212b
48080
            48080
                   2014-04-19
                               fae141ca-fa4e-487d-8562-38a1d4b27ad8
48128
            48128
                   2009-07-22
                               1b74791c-4223-4c9c-81ab-c8427e07484e
48218
            48218
                   2008-03-08
                               16014a34-8dee-4402-89da-edf7884ae808
53557
            53557
                   2008-01-28
                               1db96fae-3f95-4e2d-9e44-49b36d31d734
59809
            59809
                   2012-11-18
                               ab163027-f86b-4571-8fff-2fd20391001b
            62911
                   2016-08-29
                               ecdfa659-f3d3-44d2-870c-a6c6a7fc68e1
62911
65472
            65472
                   2012-07-16
                               94f98532-269c-4a9f-8a34-186d80a16a8e
66999
            66999
                   2008-01-25
                               7a36dd06-0363-4a3e-9d08-47feb4b2c3ce
67676
                   2017-07-14
                               13cda82a-3cfe-4ad7-9c04-5b324f148340
            67676
69415
            69415
                   2017-07-14
                               06c36d78-53eb-45e6-936e-96145a959c14
            69893
                   2016-10-28
                               b5fcea3e-2248-4045-945d-531fa938541a
69893
75427
            75427
                   2015-05-12
                               f9b7b2f0-146d-48d4-ba99-49ba9db0a02f
76058
            76058
                   2014-11-22
                               f7ccb69b-54e3-4969-9d8c-4f75d481256b
                                  ENCOUNTER
                                                 CODE
                                                          DESCRIPTION
       ab703308-e57a-4aa9-b8ef-d0fb07b60ab4
202
                                             71802-3
                                                       Housing status
823
       ddde7e19-206f-455f-99c0-98f73def8965
                                              71802-3
                                                       Housing status
1465
       fa874ebb-3cf7-4140-a168-806c98720029
                                              71802-3
                                                       Housing status
3986
       a1b7f3f5-cbac-4b0a-a371-65f2ec0a4f1f
                                              71802-3
                                                       Housing status
4256
       931ff5a5-7e2b-4f98-8f26-23aa8584b9af
                                              71802-3
                                                       Housing status
10881
       eea684d8-febf-4099-8533-a56663c87960
                                              71802-3
                                                       Housing status
12216
       c2ae4110-443b-4cca-baf4-775ffd4689be
                                              71802-3
                                                       Housing status
16128
       691a9847-3c9e-457b-9ac6-6bf1874898bc
                                              71802-3
                                                       Housing status
16216
       c93595bd-c911-4c10-8255-4466b34b4b78
                                              71802-3
                                                       Housing status
17213
      bbc7bd6d-3897-4f93-a1e0-abfa645ec208
                                              71802-3
                                                       Housing status
20485
       5bb890bb-30f3-4131-9eef-73f346bf0889
                                              71802-3
                                                       Housing status
20737
       558cf428-6f6a-48b4-8bf3-8e39a67d4d08
                                              71802-3
                                                       Housing status
                                                       Housing status
22775
       32521ff1-da2d-4cd5-b35f-580e047caf3f
                                              71802-3
       59ed3e8e-da68-4e6d-974e-0cb62739e1f9
24061
                                              71802-3
                                                       Housing status
27000
      d9168ca9-474f-4e10-a5c2-60993ce81d08
                                              71802-3
                                                       Housing status
27506
      57bd77e0-66e2-43e5-a3be-50247052cd38
                                              71802-3
                                                       Housing status
29256
       ed270417-914c-42de-a9bc-74d06a6919d9
                                              71802-3
                                                       Housing status
31687
       ed98e33d-75d3-4f6f-8fb8-dfc16ee74f36
                                              71802-3
                                                       Housing status
33982
       4f008b65-00aa-4260-9f31-f508d6d10530
                                              71802-3
                                                       Housing status
35790
      0bbf3b0a-f231-428a-b0ff-ed0680215673
                                              71802-3
                                                       Housing status
37416
      b9556eac-4bfb-4591-b275-bbf0cbaa6466
                                              71802-3
                                                       Housing status
48080
       86cf252c-8dca-41aa-b677-c2254038f79c
                                              71802-3
                                                       Housing status
48128
      50cba3d2-4d40-4f2b-8672-1dc2d75c9606
                                              71802-3
                                                       Housing status
       a9841df6-f3d4-4228-bc36-68dc786a7237
48218
                                              71802-3
                                                       Housing status
       22aae102-fd7c-4c4f-8b8e-849ac30e70e8
53557
                                              71802 - 3
                                                       Housing status
59809
      08f2cf51-98cc-407c-a33c-0b750d1228ca
                                              71802-3
                                                       Housing status
```

```
62911
             f2d50f79-c5c3-4859-9945-6ac5b32b806a
                                                   71802-3
                                                             Housing status
      65472
             642d5994-0807-4b70-98b7-8cbb8a27825a
                                                    71802-3
                                                             Housing status
      66999
             ea661a39-ffd4-440a-925b-a7e3d4145516
                                                    71802-3
                                                             Housing status
      67676
             0a6bd323-63d3-4bc9-84aa-bf8c65e87867
                                                    71802-3
                                                             Housing status
             c42832f2-3d9f-46e2-bf63-0df8f81fa556
      69415
                                                   71802-3
                                                            Housing status
      69893
             51229a5e-b92a-4266-9141-ff8217df87c1
                                                   71802-3
                                                            Housing status
             286859b9-7a6d-44d4-a2a4-1a9c07619988
      75427
                                                    71802-3
                                                             Housing status
      76058
            d4f5951e-0047-4613-9d63-8954c49518c5
                                                   71802-3
                                                            Housing status
                           VALUE
                                      UNITS
                                             INCOME
      202
             Patient is homeless
                                  {nominal}
                                             102219
      823
             Patient is homeless
                                 {nominal}
                                              60755
      1465
             Patient is homeless
                                  {nominal}
                                                 -1
      3986
             Patient is homeless
                                  {nominal}
                                             102301
      4256
             Patient is homeless
                                  {nominal}
                                                  -1
      10881
            Patient is homeless
                                  {nominal}
                                              56714
            Patient is homeless
                                  {nominal}
      12216
                                                  -1
            Patient is homeless
      16128
                                  {nominal}
                                              47115
      16216
            Patient is homeless
                                  {nominal}
                                                  -1
            Patient is homeless
      17213
                                  {nominal}
                                             108815
      20485
            Patient is homeless
                                  {nominal}
                                              71065
            Patient is homeless
      20737
                                  {nominal}
                                              70574
      22775
            Patient is homeless {nominal}
                                              49083
      24061
            Patient is homeless {nominal}
                                                 -1
            Patient is homeless
                                  {nominal}
      27000
                                              73839
      27506
            Patient is homeless {nominal}
                                              40876
                                  {nominal}
      29256
            Patient is homeless
                                              51229
            Patient is homeless {nominal}
      31687
                                                 -1
      33982
            Patient is homeless
                                  {nominal}
                                             110043
      35790
            Patient is homeless
                                  {nominal}
                                              88958
      37416
            Patient is homeless
                                  {nominal}
                                              70873
            Patient is homeless
                                  {nominal}
      48080
                                             102090
      48128
            Patient is homeless
                                  {nominal}
                                              56577
      48218
            Patient is homeless
                                  {nominal}
                                              62114
            Patient is homeless
      53557
                                  {nominal}
                                              83283
      59809
            Patient is homeless
                                  {nominal}
                                              91273
      62911 Patient is homeless {nominal}
                                              61527
      65472
            Patient is homeless {nominal}
                                             104600
      66999
            Patient is homeless {nominal}
                                              50044
      67676 Patient is homeless {nominal}
                                             130739
      69415
            Patient is homeless
                                  {nominal}
                                                 -1
      69893
            Patient is homeless
                                  {nominal}
                                              61035
            Patient is homeless
      75427
                                  {nominal}
                                                  -1
      76058
            Patient is homeless
                                 {nominal}
                                                 -1
[508]: emergency['BMIUp'] = emergency['BMI'].map(lambda x: 1 if x > 27 else 0)
      emergency['BMIDown'] = emergency['BMI'].map(lambda x: 1 if x < 25 else 0)</pre>
```

```
emergency['BMIOverall'] = emergency['BMI'].map(lambda x: 1 if x>27 or x<25 else_{\sqcup} _{\to}0)
```

I am a computer science major and have experience coding in Python and Javascript and using Stata. I am currently in Econometrics and becoming better at analyzing data with statistics and equations (vs. how I usually analyze data by questioning it, comparing it to the world around me, sometimes following my gut, asking for outside opinions...)