Project overview:

We are a Private Equity firm looking to create an insurance company. Our goal is to find the largest market opportunities by state and area type based on diversity of services and quantity of providers. We are also looking to find a reasonable network of providers to cover in a few specialties of interest for each state.

- 1. We investigated the diversity of medicare-accepted services available to populations living in different area types (i.e. metropolitan, rural...)
- 2. We also found the three most diverse area types to explore where the greatest opportunities for expansion would be
- 3. We investigated the quantity of medicare-accepted services available to populations living in different area types
- 4. We investigated the proportion of male versus female doctors who take medicare, and compared these proportions to the total number of male and female doctors in the US
 - a. Total number of female doctors versus male doctors who take medicare
 - b. Proportion of female doctors who take medicare out of the total number of female practicing physicians versus proportion of male doctors who take medicare out of the total number of male practicing physicians
 - i. According to Becker Hospital Review, there are 366,759 (36%) practicing female physicians and 652,017 (64%) practicing male physicians, for a total of 1,018,776 practicing physicians in the US total
- 5. We investigated the proportion of individuals versus corporations who take medicare, so as to further understand our target market
- 6. We investigated the diversity of medicare-accepted services available in each state, and found the top 10 most diverse states in regards to number of unique service types, so as to better understand the opportunities for the highest impact states when making a referral network
- 7. We investigated the quantity of medicare-accepted services available in each state to further understand our target market with the greatest opportunities
- 8. Another question that we wanted to answer: We were looking to create a referral network in each state for internal medicine with cardiology doctors to map out coverage.
 - We made a google storage bucket and got the link to this file, in order to access it in mapper_init() to pair up the dataset
 - This code is in the mapper_pairs.py document

To provide a report to the PE firm, we visualized our findings with barplots and pie charts. Additionally, we stored our data in GoogleCloud to work with large amounts of data efficiently. Analogously to Lab3, we ran all over our map reduce code through Google Cloud. Additionally, we used a Google Storage Bucket to store our dataset.

Extraneous files:

- We also attempted to create a referral network based on zip codes, but unfortunately, every library includes too many "NaN"s (does not have information about many zip codes) to obtain the latitude and longitude for each zip code
 - o However, the code (which otherwise works) is included in the git repository

- 1. Unique kinds of providers by area type (area_providers.py)
 - We are investigating the diversity of services available to populations living in different area types

```
Analyzing number of unique provider types available in an area
  def mapper(self, , line):
      line cols = COMMA MATCHER.split(line)
      if len(line cols) > 13 and "flow" in line cols[14]:
          provider type = line cols[16]
          provider type = ""
          yield (area, provider type)
  def combiner(self, area, provider type):
      types = set(provider type)
       for unq type in types:
          yield (area, unq type)
  def reducer(self, area, provider type):
       types = set(provider type)
      yield area, len(types)
  MRAreaProviders.run()
```

Output:

```
\"Secondary flow 30% to <50% to a larger urbanized area of 50,000 and greater\"" 59
"\"Metropolitan area low commuting: primary flow 10% to <30% to a urbanized area of 50,000 and greater\"" 23
"\"Micropolitan area core: primary flow within an urban cluster of 10,000 to 49,999\"" 69
"\"Metropolitan area core: primary flow within an urbanized area of 50,000 and greater\"" 80
"\"Metropolitan area high commuting: primary flow 30% or more to a urbanized area of 50,000 and greater\"" 58
"\"Small town high commuting: primary flow 30% or more to a urban cluster of 2,500 to 9,999\"" 20
"\"Small town low commuting: primary flow 10% to <30% to a urban cluster of 2,500 to 9,999\"" 14
```

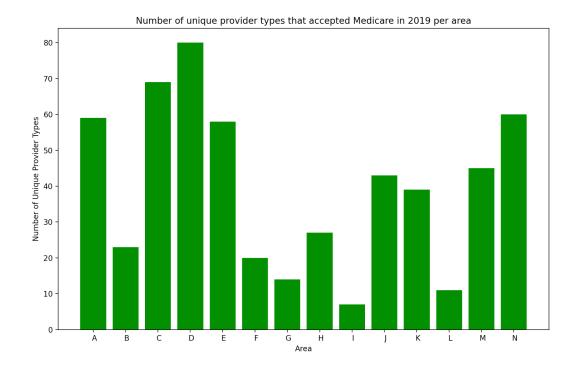
```
"\"Secondary flow 30% to <50% to a urban cluster of 10,000 to 49,999\"" 27
"\"Secondary flow 30% to <50% to a urban cluster of 2,500 to 9,999\"" 7
"\"Secondary flow 30% to <50% to a urbanized area of 50,000 and greater\"" 43
"\"Micropolitan high commuting: primary flow 30% or more to a urban cluster of 10,000 to 49,999\""
                                                                                                  39
"\"Micropolitan low commuting: primary flow 10% to <30% to a urban cluster of 10,000 to 49,999\""
"\"Rural areas: primary flow to a tract outside a urbanized area of 50,000 and greater or UC\"" 45
"\"Small town core: primary flow within an urban cluster of 2,500 to 9,999\"" 60
Processed Output (area providers.csv):
"Secondary flow 30% to <50% to a larger urbanized area of 50000 and greater", A, 59
"Metropolitan area low commuting: primary flow 10% to <30% to a urbanized area of 50000 and greater", B, 23
"Micropolitan area core: primary flow within an urban cluster of 10000 to 49999", C, 69
"Metropolitan area core: primary flow within an urbanized area of 50000 and greater", D, 80
"Metropolitan area high commuting: primary flow 30% or more to a urbanized area of 50000 and greater", E, 58
"Small town high commuting: primary flow 30% or more to a urban cluster of 2500 to 9999", F, 20
"Small town low commuting: primary flow 10% to <30% to a urban cluster of 2500 to 9999", G, 14
"Secondary flow 30% to <50% to a urban cluster of 10000 to 49999", H, 27
"Secondary flow 30% to <50% to a urban cluster of 2500 to 9999", I, 7
"Secondary flow 30% to <50% to a urbanized area of 50000 and greater", J, 43
"Micropolitan high commuting: primary flow 30% or more to a urban cluster of 10000 to 49999", K, 39
"Micropolitan low commuting: primary flow 10% to <30% to a urban cluster of 10000 to 49999", L, 11
"Rural areas: primary flow to a tract outside a urbanized area of 50000 and greater or UC", M, 45
```

Barchart code for area_providers.py: (barchart_area_providers.py)

"Small town core: primary flow within an urban cluster of 2500 to 9999", N, 60

```
import matplotlib.pyplot as plt
import pandas as pd

data = pd.read_csv('area_providers.csv')
    df = pd.DataFrame(data)
    X = list(df.iloc[:, 1])
Y = list(df.iloc[:, 2])
    plt.bar(X, Y, color='g')
plt.title("Number of unique provider types that accepted Medicare in 2019 per area")
plt.xlabel("Area")
plt.ylabel("Number of Unique Provider Types")
plt.show()
```



Key

- A: Secondary flow 30% to <50% to a larger urbanized area of 50000 and greater (59)
- B: Metropolitan area low commuting: primary flow 10% to <30% to a urbanized area of 50000 and greater (23)
- C: Micropolitan area core: primary flow within an urban cluster of 10000 to 49999 (69)
- D: Metropolitan area core: primary flow within an urbanized area of 50000 and greater (80)
- E: Metropolitan area high commuting: primary flow 30% or more to a urbanized area of 50000 and greater (58)
- F: Small town high commuting: primary flow 30% or more to a urban cluster of 2500 to 9999 (20)
- G: Small town low commuting: primary flow 10% to <30% to a urban cluster of 2500 to 9999 (14)
- H: Secondary flow 30% to <50% to a urban cluster of 10000 to 49999 (27)
- I: Secondary flow 30% to <50% to a urban cluster of 2500 to 9999 (7)
- J: Secondary flow 30% to <50% to a urbanized area of 50000 and greater (43)
- K: Micropolitan high commuting: primary flow 30% or more to a urban cluster of 10000 to 49999 (39)
- L: Micropolitan low commuting: primary flow 10% to <30% to a urban cluster of 10000 to 49999 (11)
- M: Rural areas: primary flow to a tract outside a urbanized area of 50000 and greater or UC (45)
- N: Small town core: primary flow within an urban cluster of 2500 to 9999 (60)

TOP 10 MOST DIVERSE AREAS BY SERVICE TYPE:

```
from mrjob.job import MRJob
import re
import heapq
class MRAreaProviders(MRJob):
  def mapper(self, , line):
     provider type = ""
     if area:
        yield (area, provider type)
  def combiner(self, area, provider_type):
     types = set(provider_type)
     for unq type in types:
        yield (area, unq type)
  def reducer(self, area, provider type):
     types = set(provider_type)
     num_types = len(types)
        self.h.append((num types, area))
            heapq.heapify(self.h)
         if num types > min count:
             heapq.heapreplace(self.h, (num_types, area))
```

```
def reducer_final(self):
    self.h.sort(reverse=True)
    for num_types, area in self.h:
        yield area, num_types

if __name__ == '__main__':
    MRAreaProviders.run()
```

Output:

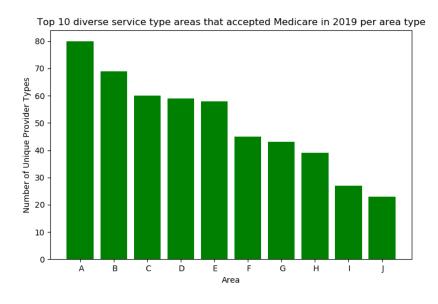
,,

Metropolitan area core: primary flow within an urbanized area of 50000 and greater, A, 80 Micropolitan area core: primary flow within an urban cluster of 10000 to 49999, B, 69 Small town core: primary flow within an urban cluster of 2500 to 9999, C, 60 Secondary flow 30% to <50% to a larger urbanized area of 50000 and greater, D, 59 Metropolitan area high commuting: primary flow 30% or more to a urbanized area of 50,000 and greater, E, 58

Rural areas: primary flow to a tract outside a urbanized area of 50000 and greater or UC, F, 45 Secondary flow 30% to <50% to a urbanized area of 50000 and greater, G, 43 Micropolitan high commuting: primary flow 30% or more to a urban cluster of 10000 to 49999, H, 39

Secondary flow 30% to <50% to a urban cluster of 10000 to 49999, I, 27 Metropolitan area low commuting: primary flow 10% to <30% to a urbanized area of 50000 and greater, J, 23

Barchart:



2. Number of providers by area type (area_any_providers.py)

a. We are investigating the quantity of services available to populations living in different area types

```
Analyzing number of providers available in an area
from mrjob.job import MRJob
class MRAreaAnyProviders(MRJob):
  def mapper(self, _, line):
      if area:
         yield (area, 1)
  def combiner(self, area, counts):
      yield area, sum(counts)
  def reducer(self, area, counts):
  MRAreaAnyProviders.run()
```

Output:

"\"Metropolitan area core: primary flow within an urbanized area of 50,000 and greater\"" 221750

3131

658

65

[&]quot;\"Metropolitan area high commuting: primary flow 30% or more to a urbanized area of 50,000 and greaten\"" 4241

[&]quot;\"Metropolitan area low commuting: primary flow 10% to <30% to a urbanized area of 50,000 and greater\""

[&]quot;\"Micropolitan area core: primary flow within an urban cluster of 10,000 to 49,999\"" 13786

[&]quot;\"Micropolitan high commuting: primary flow 30% or more to a urban cluster of 10,000 to 49,999\""

[&]quot;\"Micropolitan low commuting: primary flow 10% to <30% to a urban cluster of 10,000 to 49,999\""

[&]quot;\"Rural areas: primary flow to a tract outside a urbanized area of 50,000 and greater or UC\"" 1389

[&]quot;\"Secondary flow 30% to <50% to a larger urbanized area of 50,000 and greater\""

[&]quot;\"Secondary flow 30% to <50% to a urban cluster of 10,000 to 49,999\"" 164

```
"\"Secondary flow 30% to <50% to a urban cluster of 2,500 to 9,999\"" 21
"\"Secondary flow 30% to <50% to a urbanized area of 50,000 and greater\"" 1075
"\"Small town core: primary flow within an urban cluster of 2,500 to 9,999\"" 4426
"\"Small town high commuting: primary flow 30% or more to a urban cluster of 2,500 to 9,999\"" 190
"\"Small town low commuting: primary flow 10% to <30% to a urban cluster of 2,500 to 9,999\"" 65
```

Processed Output (area_any_providers.csv):

,,

Metropolitan area core: primary flow within an urbanized area of 50000 and greater, A, 221750 Metropolitan area high commuting: primary flow 30% or more to a urbanized area of 50000 and greater, B, 4241 Metropolitan area low commuting: primary flow 10% to <30% to a urbanized area of 50000 and greater, C, 215 Micropolitan area core: primary flow within an urban cluster of 10000 to 49999, D, 13786 Micropolitan high commuting: primary flow 30% or more to a urban cluster of 10000 to 49999, E, 658 Micropolitan low commuting: primary flow 10% to <30% to a urban cluster of 10000 to 49999, F, 65 Rural areas: primary flow to a tract outside a urbanized area of 50000 and greater or UC, G, 1389 Secondary flow 30% to <50% to a larger urbanized area of 50000 and greater, H, 3131 Secondary flow 30% to <50% to a urban cluster of 10000 to 49999, I, 164 Secondary flow 30% to <50% to a urban cluster of 2500 to 9999, J, 21 Secondary flow 30% to <50% to a urbanized area of 50000 and greater, K, 1075 Small town core: primary flow within an urban cluster of 2500 to 9999, L, 4426 Small town high commuting: primary flow 30% or more to a urban cluster of 2500 to 9999, N, 190 Small town low commuting: primary flow 10% to <30% to a urban cluster of 2500 to 9999, N, 65

Barchart code for area_any_providers.py: (barchart_area_any_providers.py)

```
Makes a barchart for the number of providers per area type.

'''
import matplotlib.pyplot as plt
import pandas as pd

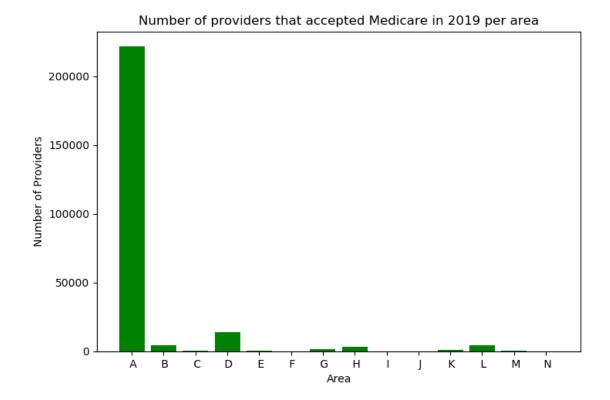
def bar_plot(log_scale):
    data = pd.read_csv('area_any_providers.csv')

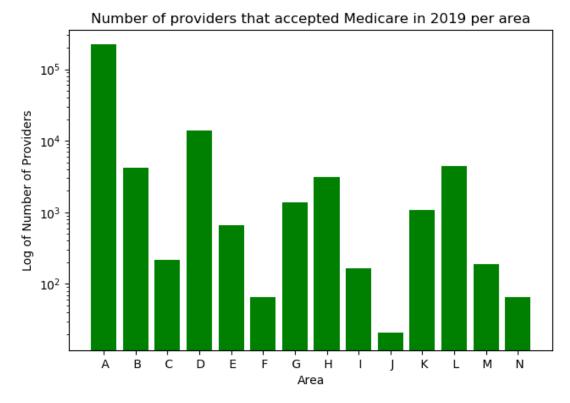
    df = pd.DataFrame(data)
    X = list(df.iloc[:, 1])
    Y = list(df.iloc[:, 2])

if log_scale:
    plt.yscale("log")

plt.bar(X, Y, color='g')

plt.title("Number of providers that accepted Medicare in 2019 per area")
plt.xlabel("Area")
if log_scale:
    plt.ylabel("Log of Number of Providers")
else:
    plt.ylabel("Number of Providers")
```





Key

- A: Metropolitan area core: primary flow within an urbanized area of 50000 and greater (221750)
- B: Metropolitan area high commuting: primary flow 30% or more to a urbanized area of 50000 and greater (4241)
- C: Metropolitan area low commuting: primary flow 10% to <30% to a urbanized area of 50000 and greater (215)
- D: Micropolitan area core: primary flow within an urban cluster of 10000 to 49999 (13786)
- E: Micropolitan high commuting: primary flow 30% or more to a urban cluster of 10000 to 49999 (658)
- F: Micropolitan low commuting: primary flow 10% to <30% to a urban cluster of 10000 to 49999 (65)
- G: Rural areas: primary flow to a tract outside a urbanized area of 50000 and greater or UC (1389)
- H: Secondary flow 30% to <50% to a larger urbanized area of 50000 and greater (3131)
- I: Secondary flow 30% to <50% to a urban cluster of 10000 to 49999 (164)
- J: Secondary flow 30% to <50% to a urban cluster of 2500 to 9999 (21)
- K: Secondary flow 30% to <50% to a urbanized area of 50000 and greater (1075)
- L: Small town core: primary flow within an urban cluster of 2500 to 9999 (4426)
- M: Small town high commuting: primary flow 30% or more to a urban cluster of 2500 to 9999 (190)
- N: Small town low commuting: primary flow 10% to <30% to a urban cluster of 2500 to 9999 (65)

- 3. Proportion of men/women providers who take Medicare (gender_take_medicare.py)
 - a. We are investigating if men or women are more likely to take Medicare, comparing the national averages of men/women who practice

Output:

"F" 76082 "M" 175632

Processed Output (gender_take_medicare.csv):

, F, 76082 M, 175632

Pie chart of gender_take_medicare.py (pichart_gender_take_medicare.py):

```
Visualizing the proportion of gender that take Medicare
```

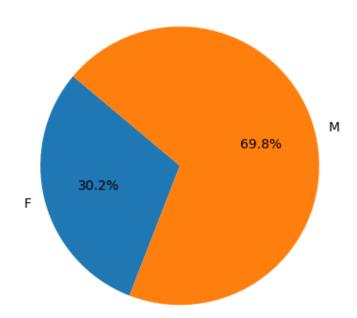
```
import matplotlib.pyplot as plt
import pandas as pd

df = pd.read_csv('gender_take_medicare.csv')

x = list(df.iloc[:, 0])
y = list(df.iloc[:, 1])

colors = ["#1f77b4", "#ff7f0e"]
explode = (0, 0)
plt.pie(y, labels=x, explode=explode, colors=colors,
autopct='%1.1f%%', shadow=True, startangle=140)
plt.title("Proportion of genders of individuals who took Medicare in 2019")
plt.show()
```

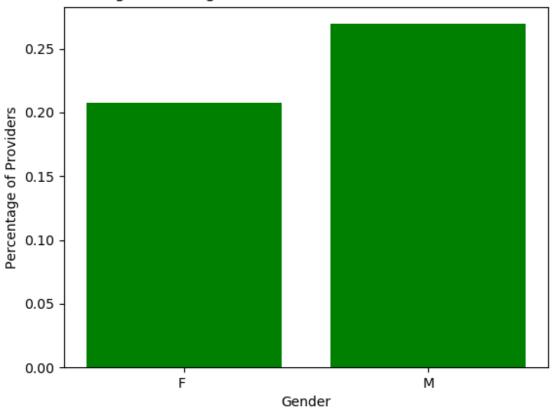
Proportion of genders of providers who took Medicare in 2019



Barplot: Percentage of each gender who took Medicare versus total in the US

 According to Becker Hospital Review, there are 366,759 (36%) practicing female physicians and 652,017 (64%) practicing male physicians, for a total of 1,018,776 practicing physicians in the US total

Percentage of each gender who took Medicare versus total in US



4. Proportion of individuals vs. corporations who take Medicare (corporation_vs_individual.py)

```
Analyzing number of corporations vs. individuals who take Medicare
from mrjob.job import MRJob
import re
  def mapper(self, , line):
      COMMA MATCHER = re.compile(r", (?=(?:[^\"']*[\"'][^\"']*[\"'])*[^\"']*$)")
  def combiner(self, type, counts):
      yield type, sum(counts)
  def reducer(self, type, counts):
      yield type, sum(counts)
```

Output:

```
"corporation" 1129
"individual" 251714
```

<u>Processed Output</u> (corporation_vs_individual.csv):

corporation, 1129 individual, 251714

Pie chart for corporation_vs_individual.py (pichart_corporation_vs_individual.py)

```
Visualizing the proportion of gender that take Medicare
```

```
import matplotlib.pyplot as plt
import pandas as pd

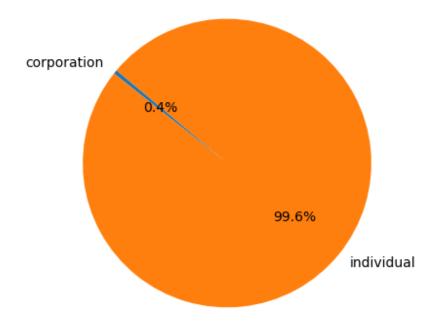
df = pd.read_csv('corporation_vs_individual.csv')

x = list(df.iloc[:, 0])
y = list(df.iloc[:, 1])

colors = ["#1f77b4", "#ff7f0e"]
explode = (0, 0)
plt.pie(y, labels=x, explode=explode, colors=colors,
autopct='%1.1f%%', shadow=True, startangle=140)
plt.title("Proportion of individuals vs corporations that took Medicare in 2019")

plt.show()
```

Proportion of individuals vs corporations that took Medicare in 2019



5. Number of unique provider types in state that take Medicare (state_take_medicare.py)

```
Analyzing number of unique types of providers in states who take Medicare
from mrjob.job import MRJob
import re
class MRStateProviders(MRJob):
  def mapper(self, _, line):
      COMMA_MATCHER = re.compile(r",(?=(?:[^\"']*[\"'][^\"']*[\"'])*[^\"']*$)")
      line cols = COMMA MATCHER.split(line)
          provider_type = line_cols[16]
          provider_type = ""
          yield (state, provider_type)
  def combiner(self, state, provider type):
      types = set(provider_type)
      for unq type in types:
          yield (state, unq_type)
  def reducer(self, state, provider type):
      types = set(provider_type)
      yield state, len(types)
  MRStateProviders.run()
```

Output:

"AE" 6 "AK" 42 "AL" 64 "AP" 5

- "AR" 59
- "AZ" 64
- "CA" 69
- "CO" 62
- "CT" 59
- "DC" 50
- "DE" 52
- "FL" 69
- "GA" 63
- "GU" 17
- "HI" 53
- "IA" 56
- "ID" 55
- "IL" 70
- "IN" 62
- "KS" 57
- "KY" 58
- "LA" 58
- "MA" 64
- "MD" 62
- "ME" 54
- "MI" 61
- "MN" 63
- "MO" 61
- "MP" 5
- "MS" 51
- "MT" 54
- "NC" 68
- "ND" 51
- "NE" 56
- "NH" 57
- "NJ" 64
- "NM" 55
- "NV" 53
- "NY" 73
- "OH" 65
- "OK" 55
- "OR" 61
- "PA" 70
- "PR" 52
- "RI" 52
- "SC" 60
- "SD" 51
- "TN" 62

```
"TX"
      69
"UT"
      58
"VA"
      64
"VI"
     17
"VT"
      48
"WA" 65
"WI"
      64
"WV"
       51
"WY"
       39
<u>Processed Output</u> (state_take_medicare.csv):
AE, 6
AK, 42
AL, 64
AP, 5
AR, 59
AZ, 64
CA, 69
CO, 62
CT, 59
DC, 50
DE, 52
FL, 69
GA, 63
GU, 17
HI, 53
IA, 56
ID, 55
IL, 70
IN, 62
KS, 57
KY, 58
LA, 58
MA, 64
MD, 62
ME, 54
MI, 61
MN, 63
MO, 61
MP, 5
MS, 51
MT, 54
```

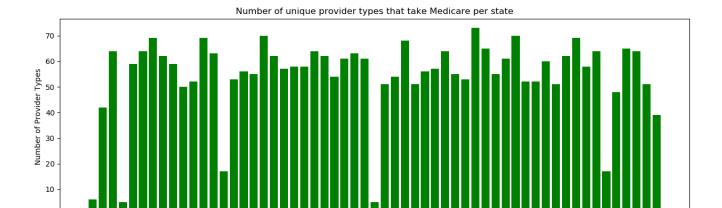
NC, 68

ND, 51 NE, 56 NH, 57 NJ, 64 NM, 55 NV, 53 NY, 73 OH, 65 OK, 55 OR, 61 PA, 70 PR, 52 RI, 52 SC, 60 SD, 51 TN, 62 TX, 69 UT, 58 VA, 64 VI, 17 VT, 48 WA, 65 WI, 64 WV, 51 WY, 39

Barchart of unique provider types in state that take Medicare (barchart_state_take_medicare.py)

```
import matplotlib.pyplot as plt
import pandas as pd

data = pd.read_csv('state_take_medicare.csv')
    df = pd.DataFrame(data)
    X = list(df.iloc[:, 0])
    Y = list(df.iloc[:, 1])
    plt.bar(X, Y, color='g')
    plt.title("Number of unique provider types that take Medicare per state")
    plt.xlabel("States")
    plt.ylabel("Number of Provider Types")
    plt.show()
```



TOP 10 STATES (top_ten_states.py):

```
from mrjob.job import MRJob
import re
import heapq
class MRStateProviders heap(MRJob):
   def mapper(self, _, line):
       COMMA MATCHER =
re.compile(r",(?=(?:[^\"']*[\"'][^\"']*[\"'])*[^\"']*$)")
       line cols = COMMA MATCHER.split(line)
       if len(line cols[10]) == 2 and line cols[10] != "AA" and
line cols[10] != "ZZ" and \
       line cols[10] != "XX" and line cols[10] != "US" and not
line cols[10].isnumeric():
           provider type = line cols[16]
           state = ""
          provider type = ""
       if state:
           yield (state, provider type)
   def combiner(self, state, provider_type):
```

```
types = set(provider type)
      for unq type in types:
          yield (state, unq type)
  def reducer init(self):
  def reducer(self, state, provider type):
      types = set(provider type)
      num types = len(types)
      if len(self.h) < 10:
          self.h.append((num types, state))
          if len(self.h) == 10:
              heapq.heapify(self.h)
          if num types > min count:
              heapq.heapreplace(self.h, (num types, state))
  def reducer final(self):
      self.h.sort(reverse=True)
      for num types, state in self.h:
          yield state, num types
if name == ' main ':
MRStateProviders heap.run()
```

Processed Output (top_ten_states.csv):

NY, 73 PA, 70 IL, 70 TX, 69 FL, 69 CA, 69

NC, 68

WA, 65

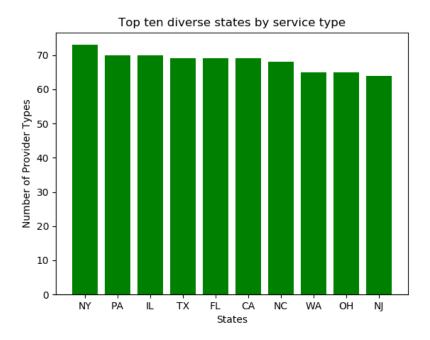
OH, 65

NJ, 64

Barchart (barchart_top_ten_states.py):

```
import matplotlib.pyplot as plt
import pandas as pd
data = pd.read_csv('top_ten_states.csv')
```

```
df = pd.DataFrame(data)
X = list(df.iloc[:, 0])
Y = list(df.iloc[:, 1])
plt.bar(X, Y, color='g')
plt.title("Top ten diverse states by service type")
plt.xlabel("States")
plt.ylabel("Number of Provider Types")
plt.show()
```



6. Number of providers in a state (state_any_take_medicare.py)

```
Analyzing number of providers in states who take Medicare
1 1 1
from mrjob.job import MRJob
import re
class MRStateAnyTakeMedicare(MRJob):
  def mapper(self, , line):
     line cols = COMMA MATCHER.split(line)
and line cols[10] != "XX" and line cols[10] != "US" and not line cols[10].isnumeric():
  def combiner(self, state, counts):
  def reducer(self, state, counts):
  MRStateAnyTakeMedicare.run()
```

Output:

```
"AE"
      6
"AK"
      554
"AL"
      3330
"AP"
      6
"AR"
      1994
"AZ"
      4605
"CA"
      20397
"CO"
      4248
"CT"
      3801
"DC"
      1116
"DE"
      827
"FL"
       16703
"GA"
      6736
```

```
"GU"
       51
"HI"
       764
"IA"
       2033
"ID"
       1111
"IL"
       9307
"IN"
       5542
"KS"
       2067
"KY"
       3284
"LA"
       3637
"MA"
       8365
"MD"
       4603
"ME"
       1347
"MI"
       7884
"MN"
       5618
"MO"
       5291
"MP"
       6
"MS"
       1602
"MT"
       933
"NC"
       8870
"ND"
       911
"NE"
       1349
"NH"
       1531
"NJ"
       7422
"NM"
       1537
"NV"
       1824
"NY"
       19370
"OH"
       9756
"OK"
       2312
"OR"
       3624
"PA"
       12984
"PR"
       1494
"RI"
       1268
"SC"
       4065
"SD"
       819
"TN"
       5560
"TX"
       15963
"UT"
       1971
"VA"
       6390
"VI"
       43
"VT"
       610
"WA"
       6388
"WI"
       6045
"WV"
       1488
"WY"
       327
<u>Processed Output</u> (state_any_take_medicare.csv):
AE, 6
```

AK, 554

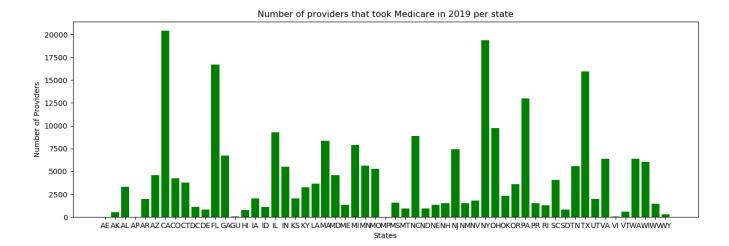
- AL, 3330
- AP, 6
- AR, 1994
- AZ, 4605
- CA, 20397
- CO, 4248
- CT, 3801
- DC, 1116
- DE, 827
- FL, 16703
- GA, 6736
- GU, 51
- HI, 764
- IA, 2033
- ID, 1111
- IL, 9307
- IN, 5542
- KS, 2067
- KY, 3284
- LA, 3637
- MA, 8365
- MD, 4603
- ME, 1347
- MI, 7884
- MN, 5618
- MO, 5291
- MP, 6
- MS, 1602
- MT, 933
- NC, 8870
- ND, 911
- NE, 1349
- NH, 1531
- NJ, 7422
- NM, 1537
- NV, 1824
- NY, 19370
- OH, 9756
- OK, 2312
- OR, 3624
- PA, 12984
- PR, 1494
- RI, 1268
- SC, 4065
- SD, 819
- TN, 5560
- TX, 15963
- UT, 1971
- VA, 6390

VI, 43 VT, 610 WA, 6388 WI, 6045 WV, 1488 WY, 327

Barchart for state_any_take_medicare.py (barchart_state_any_take_medicare.py)

```
import matplotlib.pyplot as plt
import pandas as pd

data = pd.read_csv('state_any_take_medicare.csv')
    df = pd.DataFrame(data)
    X = list(df.iloc[:, 0])
Y = list(df.iloc[:, 1])
    plt.bar(X, Y, color='g')
plt.title("Number of providers that took Medicare in 2019 per state")
plt.xlabel("States")
plt.ylabel("Number of Providers")
    plt.show()
```



Another question that we wanted to answer: We were looking to create a referral network in each state for internal medicine with cardiology doctors to map out coverage.

- We were not able to get this code to fully run, but after many edits, this should work
- We made a google storage bucket (as explained up top) and got the link to this file

File name: mapper_pairs.py

```
from mrjob.job import MRJob
from mrjob.step import MRStep
Trying to count the number of internal medicine + cardiology pairs
class build top pairs(MRJob):
  def mapper_init(self):
      gcs file = gcs.open('gs://big-data-project storage-bucket/med data.csv')
  def mapper(self, _, line):
      row1 = line.strip().split(",")
          row2 = row.strip().split(",")
          state2 = row2[10]
  def combiner(self, state, counts):
```

https://stackoverflow.com/questions/48279061/gcs-read-a-text-file-from-google-cloud-storage-directly-into-python

Another question that we aimed to solve was also finding the average age of mothers and fathers at the time of a baby's birth so as to gage our target coverage audience. Here was the dataset we used: (avg_mom_age.py, avg_dad_age.py)

https://console.cloud.google.com/bigquery?p=bigquery-public-data&d=samples&t=natality&page=table&ga=2.86294480.999355211.1653932284-1312598613.1652540771&project=lab2-351619&ws=!1m5!1m4!4m3!1sbigquery-public-data!2ssamples!3snatality

• The code is shown below

```
Average age of mothers who gave birth

from mrjob.job import MRJob
import re

class MRAvgMotherAge(MRJob):

def mapper(self, _, line):
    COMMA_MATCHER = re.compile(r",(?=(?:[^\"]*[\""]*[\""]*[\""])*[^\""]*$)")
    line_cols = COMMA_MATCHER.split(line)
    mother_age = line_cols[14]
    yield None, mother_age

def reducer(self, _, mother_ages):
    sum = 0
    count = 0
    for age in mother_ages:
        sum += age
        count += 1
        yield None, sum/count

if __name__ == '__main__':
    MRAvgMotherAge.run()
```

```
Average age of fathers at time of birth
```

```
from mrjob.job import MRJob
import re

class MRAvgFatherAge(MRJob):

    def mapper(self, _, line):
        COMMA_MATCHER = re.compile(r",(?=(?:[^\"]*[\""][^\""]*[\""]]*[\""]]*[\""]]*[\""]]*[\""]]*[\""]]*[\""]]*[\""]]*[\""]]*[\""]]*[\""]]*[\""]]*[\""]]*[\""]]*[\""]]*[\""]]*[\""]]*[\""]]*[\""]]*[\""]]*[\""]]*[\""]]*[\""]]*[\""]]*[\""]]*[\""]]*[\""]]*[\""]]*[\""]]*[\""]]*[\""]]*[\""]]*[\""]]*[\""]]*[\""]]*[\""]]*[\""]]*[\""]]*[\""]]*[\""]]*[\""]]*[\""]]*[\""]]*[\""]]*[\""]]*[\""]]*[\""]]*[\""]]*[\""]]*[\""]]*[\""]]*[\""]]*[\""]]*[\""]]*[\""]]*[\""]]*[\""]]*[\""]]*[\""]]*[\""]]*[\""]]*[\""]]*[\""]]*[\""]]*[\""]]*[\""]]*[\""]]*[\""]]*[\""]]*[\""]]*[\""]]*[\""]]*[\""]]*[\""]]*[\""]]*[\""]]*[\""]]*[\""]]*[\""]]*[\""]]*[\""]]*[\""]]*[\""]]*[\""]]*[\""]]*[\""]]*[\""]]*[\""]]*[\""]]*[\""]]*[\""]]*[\""]]*[\""]]*[\""]]*[\""]]*[\""]]*[\""]]*[\""]]*[\""]]*[\""]]*[\""]]*[\""]]*[\""]]*[\""]]*[\""]]*[\""]]*[\""]]*[\""]]*[\""]]*[\""]]*[\""]]*[\""]]*[\""]]*[\""]]*[\""]]*[\""]]*[\""]]*[\""]]*[\""]]*[\""]]*[\""]]*[\""]]*[\""]]*[\""]]*[\""]]*[\""]]*[\""]]*[\""]]*[\""]]*[\""]]*[\""]]*[\""]]*[\""]]*[\""]]*[\""]]*[\""]]*[\""]]*[\""]]*[\""]]*[\""]]*[\""]]*[\""]]*[\""]]*[\""]]*[\""]]*[\""]]*[\""]]*[\""]]*[\""]]*[\""]]*[\""]]*[\""]]*[\""]]*[\""]]*[\""]]*[\""]]*[\""]]*[\""]]*[\""]]*[\""]]*[\""]]*[\""]]*[\""]]*[\""]]*[\""]]*[\""]]*[\""]]*[\""]]*[\""]]*[\""]]*[\""]]*[\""]]*[\""]]*[\""]]*[\""]]*[\""]]*[\""]]*[\""]]*[\""]]*[\""]]*[\""]]*[\""]]*[\""]]*[\""]]*[\""]]*[\""]]*[\""]]*[\""]]*[\""]]*[\""]]*[\""]]*[\""]]*[\""]]*[\""]]*[\""]]*[\""]]*[\""]]*[\""]]*[\""]]*[\""]]*[\""]]*[\""]]*[\""]]*[\""]]*[\""]]*[\""]]*[\""]]*[\""]]*[\""]]*[\""]]*[\""]]*[\""]]*[\""]]*[\""]]*[\""]]*[\""]]*[\""]]*[\""]]*[\""]]*[\""]]*[\""]]*[\""]]*[\""]]*[\""]]*[\""]]*[\""]]*[\""]]*[\""]]*[\""]]*[\""]]*[\""]]*[\""]]*[\""]]*[\""]]*[\""]]*[\""]]*[\""]]*[\""]]*[\""]]*[\""]]*[\""]]*[\""]]*[\""]]*[\""]]*[\""]]*[\""]]*[\""]]*[\""]]*[\""]]*[\""]]*[\""]]*[\""]]*[\""]]*[\""]]*[\""]]*[\""]]*[\""]]*[\""]]*[\""]]*[\""]]*[\""]]*[\""]]*[\""]]*[\""]]*[\""]]*[\""]]*[\""]]*[\""]]*[\"
```

Extraneous GoogleCloud terminal code from lab 3: scp -i ~/.ssh/google-cloud-cs123 med_data.csv bkatsnelson@34.68.254.233:~/

CODE FOR ZIP CODES:

```
from mrjob.job import MRJob
from mrjob.step import MRStep
import heapq
import csv

class build_top_pairs(MRJob):
    def mapper_init():
        with open('med_data.csv') as dataset:
            self.data = dataset.read_lines()

    def mapper(self, _, line):
        row1 = line.strip().split(",")
```

```
state1 = row1[10]
           for row in self.data:
               row2 = row.strip().split(",")
               state2 = row2[10]
and state1 == state2:
                   yield (row1 + row2), (state1, 1)
  def combiner(self, rows, state, counts):
      yield rows, state, sum(counts)
  def reducer init(self):
  def reducer(self, rows, state, counts):
     row1 = rows[0:29]
     row2 = rows[29:]
     yield rows, state, sum(counts)
              self.h.append((dist, (row1, row2)))
              if len(self.h) == 3:
                  heapq.heapify(self.h)
              if dist > min count:
                  heapq.heapreplace(self.h, (dist, (row1, row2)))
  def reducer final(self):
     self.h.sort(reverse = True):
     for dist, (row1, row2) in self.h:
          state = row1[10]
          last name1 = row1[2]
          profession1 = row1[16]
          zip code1 = row1[12]
          last name2 = row2[2]
          profession2 = row2[16]
          zip code2 = row2[12]
```

```
yield (("SEPARATE FIELDS").join([state, first name1, last name1,
profession1, zip code1, \setminus
          first name2, last name2, profession2, zip code2, distance])),
None
if __name__ == "__main__":
 build top pairs.run()
## MAP TO CSV IMPLEMENTATION:
import csv
## pre_csv = task_p.py < filename.csv > output.csv
fieldnames = ["state", "first name1", "last name1", "profession1",
"zip code1", \
"distance"]
with open("final output.csv", "w") as f:
 writer = csv.DictWriter(f, fieldnames=fieldnames)
 writer.writeheader()
 with open(filename) as f old:
          row = line.split("\t")
          row keep = row[0]
          fields = row keep.split("SEPARATE FIELDS")
          row to add = {"state": fields[0],
              "first name1": fields[1],
              "last name1": fields[2],
              "profession1": fields[3],
              "zip code1": fields[4],
              "first name2": fields[5],
              "last name2": fields[6],
              "profession2": fields[7],
              "zip code2": fields[8],
              "distance": fields[9]}
          writer.writerow(row to add)
search = SearchEngine(simple zipcode=False)
data = []
with open(filename) as f:
```

```
# Don't want to iterate twice through the same person
used ids = set()
 def mapper(self, _, line):
              row2 = row.strip().split(",")
Medicine") \
list
                  heapq.heapify(self.h)
```

```
heapq.heapreplace(self.h, (dist, (row1, row2)))
          distance = dist*(-1)
profession1, zip code1, \setminus
None
if name == " main ":
## MAP TO CSV IMPLEMENTATION:
import csv
## pre csv = task p.py < filename.csv > output.csv
fieldnames = ["state", "first name1", "last name1", "profession1",
"zip code1", \
with open("final output.csv", "w") as f:
 writer.writeheader()
```

EXTRANEOUS CODE:

```
from mrjob.job import MRJob
from mrjob.step import MRStep
import heapq
import csv
class build_top_pairs(MRJob):
  def mapper init(self):
     gcs_file = gcs.open(filename)
     contents = gcs_file.read()
     gcs_file.close()
  def mapper(self, _, line):
     gcs_file = gcs.open(filename)
     contents = gcs_file.read()
     row1 = line.strip().split(",")
     state1 = row1[10]
     # loop through dataset for every iteration of mapper
     for row in self.data:
       row2 = row.strip().split(",")
       state2 = row2[10]
       if ("Internal Medicine" in row1) and ("Cardiology" in row2) and state1 == state2:
          yield state1, 1
```

```
gcs_file.close()
  def combiner(self, state, counts):
     yield state, sum(counts)
  def reducer(self, state, counts):
     yield state, sum(counts)
if __name__ == "__main__":
 build_top_pairs.run()
from mrjob.job import MRJob
from mrjob.step import MRStep
import heapq
import csv
class build top pairs(MRJob):
  def mapper_init(self):
     with open('med data.csv') as dataset:
       self.data = dataset.read_lines()
  def mapper(self, _, line):
     row1 = line.strip().split(",")
     state1 = row1[10]
     # loop through dataset for every iteration of mapper
     for row in self.data:
       row2 = row.strip().split(",")
       state2 = row2[10]
       if ("Internal Medicine" in row1) and ("Cardiology" in row2) and state1 == state2:
          yield state1, 1
  def combiner(self, state, counts):
     yield state, sum(counts)
  def reducer(self, state, counts):
     yield state, sum(counts)
```

```
if __name__ == "__main__":
   build_top_pairs.run()
```

```
reader=csv.reader(open('med data.csv', 'r'), delimiter=',')
writer=csv.writer(open('med data paired.csv', 'w'), delimiter=',')
for row1 in reader:
   row1 string = ",".join(row1)
  COMMA MATCHER1 =
re.compile(r",(?=(?:[^\"']*[\"'][^\"']*[\"'])*[^\"']*$)")
  line1 = COMMA MATCHER1.split(row1 string)
  state1 = line1[10]
  for row2 in reader:
       row2 string = ",".join(row2)
      COMMA MATCHER2 =
re.compile(r",(?=(?:[^\"']*[\"'][^\"']*[\"'])*[^\"']*$)")
       line2 = COMMA MATCHER2.split(row2 string)
      if len(line2) > 11:
          state2 = line2[10]
          state2 = "None"
state1 == state2:
          writer.writerow(row2)
```

```
import logging
import os
import cloudstorage as gcs
import webapp2
from google.appengine.api import app identity
def get(self):
  self.response.headers['Content-Type'] = 'text/plain'
  self.response.write('Demo GCS Application running from Version: '
  self.response.write('Using bucket name: ' + bucket name + '\n\n')
def create file(self, filename):
   """Create a file.
    filename: filename.
  self.response.write('Creating file %s\n' % filename)
  write retry params = gcs.RetryParams(backoff factor=1.1)
  gcs_file = gcs.open(filename,
                     content type='text/plain',
                     options={'x-goog-meta-foo': 'foo',
                     retry params=write retry params)
  gcs file.write('abcde\n')
  gcs file.write('f'*1024*4 + '\n')
  gcs file.close()
  self.tmp filenames to clean up.append(filename)
```

```
def read_file(self, filename):
    self.response.write('Reading the full file contents:\n')

    gcs_file = gcs.open(filename)
    contents = gcs_file.read()
    gcs_file.close()
    self.response.write(contents)
```