Partners:

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```
In [ ]:
```

Reading Data; Data Prep.

```
import os
In [444]:
          import pandas as pd
          import numpy as np
          import seaborn as sns
          import json
          import re
  In [2]: os.listdir()
 Out[2]: ['sample.reviews.json',
           'GOOGLE LOCAL.ipynb',
           'reviews.clean.json',
           'users.clean.json',
            '.ipynb checkpoints',
            'places.clean.json']
In [11]:
          def parseData(fname):
              data = []
              with open(fname, 'rb') as f:
                   for l in f.readlines():
                       data.append(eval(1))
               return data
 In [18]: reviews = []
          with open('sample.reviews.json') as f:
```

reviews = json.load(f)

```
In [20]: review_df = pd.DataFrame(reviews)
In [21]: user = parseData('users.clean.json')
In [22]: user_df = pd.DataFrame(user)
In [23]: places = parseData('places.clean.json')
In [24]: places_df = pd.DataFrame(places)
In []:
In []:
```

We notice that more than 90% of the reviews have unique ids, which means that a lot of those users are

inactive users, which is meaningless to analyze, thus we clean the data by ensuring that the each user and place

have more than 3 comments.

```
In [44]:
          # Select Active User
          user count = review df['qPlusUserId'].value counts()
          active user id = user count.where(user count >3).dropna().index
          # Select Active Places
          place count = review df['qPlusPlaceId'].value counts()
          active place id = place count.where(place count >3).dropna().index
          ## Valid User df and valid place df are those with active ids with act
  In [ ]:
          ive reviews
In [26]:
          valid user df = user df[user df['gPlusUserId'].isin(active user id)]
          valid place df = places df[places df['gPlusPlaceId'].isin(active place
In [295]:
          _id)]
          ## select reviews with both user and places being active
```

```
In [27]:
          valid review df = review df[review df['gPlusUserId'].isin(active user
          id) & review df['gPlusPlaceId'].isin(active place id)]
 In [30]:
          valid review df.shape
Out[30]: (5801, 8)
  In [ ]: ## Drop those not inside user df and places df
In [34]: valid review df dropped = valid review df[valid review df['gPlusUserId
          '].isin(user df['qPlusUserId']) & \
                                                    valid review df['qPlusPlaceI
          d'].isin(places df['qPlusPlaceId'])]
In [47]: valid user df.shape
Out[47]: (5929, 6)
In [296]: valid place df.shape
Out[296]: (7298, 8)
In [49]: valid review df dropped.shape
Out[49]: (4461, 8)
```

Task

Task: We focus on predict the rating given a pair of random user and place.

Business Insight: If we can predict how likely a user would like a random place based on user's activities history and the characteristic of the place, we can recommend the places to users by ranking the ratings of places around them.

BaseLine Model

BaseLine Model: we predict the rating by the mean of average rating of the user and the average rating of the place with equal weights.

```
In [50]: from sklearn.utils import shuffle
         df = shuffle(valid review df dropped)
         size = len(df)
         train data = df[:int(size * 0.8)]
         validate data = df[int(size * 0.8): int(size * 0.9)]
         test data = df[int(size * 0.9):]
In [51]: | user rating avg = train data.groupby('gPlusUserId')['rating'].mean().t
         o dict()
         place rating avg = train data.groupby('gPlusPlaceId')['rating'].mean()
         .to dict()
         total avg = train data['rating'].mean()
In [52]: def predict(row):
             hasUser = row['gPlusUserId'] in user rating avg.keys()
             hasPlace = row['gPlusPlaceId'] in place rating avg.keys()
             if hasUser and hasPlace:
                 return (user rating avg[row['gPlusUserId']] + place rating avg
         [row['qPlusPlaceId']]) / 2
             elif hasUser:
                 return user rating avg[row['gPlusUserId']]
             elif hasPlace:
                 return place rating avg[row['gPlusPlaceId']]
             else:
                 return total avg
         predictions = validate data.apply(predict, axis = 1)
         ((validate data['rating'] - predictions) ** 2).mean()
Out[52]: 0.8264152661431119
 In [ ]:
```

Feature Engineering and Feature Augumentation and Data Imputation

User DataFrame

Deal with hours:

Here we get opening hours, closing hours, length of opening and days of opening per week from the original "hours" column by feature augmentation.

```
In [298]:
          open times = []
          close times = []
          num dayss = []
          existence = valid place df['hours'].isna()
          for i in range(len(valid place df)):
              l = valid place df['hours'].iloc[i]
              if existence.iloc[i]:
                   open times.append(np.nan)
                   close times.append(np.nan)
                   num dayss.append(np.nan)
                   continue
              num days = len(1)
              open time = []
              close time = []
              opening time lst = [day[1][0][0] for day in 1]
               for time in opening time 1st:
                   if '24' in time:
                       open time.append(0)
                       close time.append(24)
                   if '-' in time:
                       open time str = time.split('-')[0]
                       close time str = time.split('-')[2]
                       open time num = 0
                       close time num = 0
                       if 'am' in open_time_str:
                           open time num = int(open time str[:-6])
                       elif 'pm' in open time str:
                           open time num = int(open time str[:-6]) + 12
                       else:
                           continue
                       open time.append(open time num)
                       if 'am' in close time str:
                           close time num = int(close time str[:-6])
                       elif 'pm' in close time str:
```

/anaconda3/lib/python3.7/site-packages/numpy/core/fromnumeric.py:292 0: RuntimeWarning: Mean of empty slice.

out=out, **kwargs)

/anaconda3/lib/python3.7/site-packages/numpy/core/_methods.py:85: Ru
ntimeWarning: invalid value encountered in double_scalars
 ret = ret.dtype.type(ret / rcount)

```
In [299]: valid_place_df['open_times'] = open_times
    valid_place_df['close_times'] = close_times
    valid_place_df['num_days'] = num_dayss
```

/anaconda3/lib/python3.7/site-packages/ipykernel_launcher.py:1: Sett
ingWithCopyWarning:

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/indexing.html#indexing-view-versus-copy

"""Entry point for launching an IPython kernel.

/anaconda3/lib/python3.7/site-packages/ipykernel_launcher.py:2: Sett ingWithCopyWarning:

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/indexing.html#indexing-view-versus-copy

/anaconda3/lib/python3.7/site-packages/ipykernel_launcher.py:3: Sett ingWithCopyWarning:

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/indexing.html#indexing-view-versus-copy

This is separate from the ipykernel package so we can avoid doing imports until

deal with price

deal with phone number by checking availablility of phone number

```
In [303]: valid place df['phone available'] = 1 - valid place df['phone'].isna()
          /anaconda3/lib/python3.7/site-packages/ipykernel launcher.py:1: Sett
          ingWithCopyWarning:
          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/panda
          s-docs/stable/indexing.html#indexing-view-versus-copy
            """Entry point for launching an IPython kernel.
In [304]: valid place df['num hours opening'] = valid place df['close times'] -
          valid place df['open times']
          /anaconda3/lib/python3.7/site-packages/ipykernel launcher.py:1: Sett
          ingWithCopyWarning:
          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/panda
          s-docs/stable/indexing.html#indexing-view-versus-copy
            """Entry point for launching an IPython kernel.
```

Data imputation of filling missing values.

```
In [305]:
          val = valid place df['open times'].loc[valid place df['open times'].no
          tna()].mean()
          valid place df['open times'].fillna(val, inplace = True)
In [306]: val = valid place df['close times'].loc[valid place df['close times'].
          notna()].mean()
          valid place df['close times'].fillna(val, inplace = True)
          val = valid place df['num days'].loc[valid place df['num days'].notna(
          ) ] . mean()
          valid place df['num days'].fillna(val, inplace = True)
          val = valid place df['num hours opening'].loc[valid place df['num hour
          s opening'].notna()].mean()
          valid place df['num hours opening'].fillna(val, inplace = True)
In [311]: val = valid place df['price size'].loc[valid place df['price size'] !=
          0].mean()
In [313]: valid place df['price size'].replace(0, val,inplace = True)
          /anaconda3/lib/python3.7/site-packages/pandas/core/generic.py:5890:
          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/panda
          s-docs/stable/indexing.html#indexing-view-versus-copy
            self. update inplace(new data)
          cleaned place df = valid place df
In [324]:
In [325]:
          cleaned place df.set index('gPlusPlaceId', inplace=True)
In [326]:
          final place df = cleaned place df[cleaned place df.columns[7:]]
In [327]: | val = final place df['price size'].loc[final place df['price size'] !=
          0].mean()
```

```
final_place_df['price_size'] = final_place_df['price size'].replace(to
In [328]:
          _replace = 0, value = val)
          /anaconda3/lib/python3.7/site-packages/ipykernel launcher.py:1: Sett
          ingWithCopyWarning:
          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/panda
          s-docs/stable/indexing.html#indexing-view-versus-copy
            """Entry point for launching an IPython kernel.
          valid user df.set index('gPlusUserId', inplace=True)
In [332]:
  In [ ]:
          merged = valid review df dropped.merge(final place df, right index=Tru
In [333]:
          e, left on = 'gPlusPlaceId', how = 'left')
In [334]: user interested cols = ['open times', 'close times', 'num days', 'pric
          e size', 'num hours opening']
In [335]: user avg info = merged.groupby('gPlusUserId')[user interested cols].me
          an()
In [337]: final user df = user avg info.copy()
  In [ ]:
In [361]:
          for cat in list(valid review df dropped.categories.values):
              if cat is not None:
                  for 1 in cat:
                      for s in l.split():
                          if len(re.findall(r'[a-zA-Z]', s)) > 0:
                              words.append(s)
          cat_counts = pd.Series(words).value counts()
          common cat = list(cat counts.where(cat counts > 100).dropna().index)
```

```
In [362]:
          df = valid review df dropped#.dropna(subset = ['categories'])#.loc[val
          id review df dropped['gPlusUserId'] == '110090296112763745586']
          user rating cat = {}
          for i in range(len(df)):
              row = df.iloc[i]
              if row['gPlusUserId'] not in user rating cat.keys():
                  user rating cat[row['gPlusUserId']] = {}
                  for cat in common cat:
                      user_rating_cat[row['gPlusUserId']][cat] = []
                  user rating cat[row['gPlusUserId']]['Other'] = []
              #if row['gPlusUserId'] in user_rating_cat.keys():
              if row['categories'] is not None:
                  for words in row['categories']:
                      for w in words.split():
                           if w in common cat:
                               user rating cat[row['gPlusUserId']][w].append(row[
           'rating'])
                           else:
                               user_rating_cat[row['gPlusUserId']]['Other'].appen
          d(row['rating'])
In [364]:
          place dict = {}
          for i in range(len(df)):
              row = df.iloc[i]
              place = row['gPlusPlaceId']
              if place not in place dict.keys():
                  place_dict[place] = []
              if row['categories'] is not None:
                  for s in row['categories']:
                      for w in s.split():
                           if w in common cat:
                               place_dict[place].append(w)
                           else:
                               place dict[place].append('Other')
          place dict = {s:list(set(place dict[s])) for s in place dict.keys()}
```

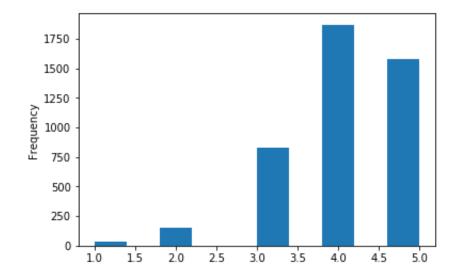
Exploratory Data Analysis

```
In [475]: %matplotlib inline
    rating = valid_review_df_dropped.rating
    print(rating.describe())
    rating.plot(kind='hist')
```

4461.000000 count 4.076440 mean 0.860041 std 1.000000 min 25% 4.000000 50% 4.000000 75% 5.000000 5.000000 max

Name: rating, dtype: float64

Out[475]: <matplotlib.axes._subplots.AxesSubplot at 0x1a21617668>



In [476]: final_place_df.describe()

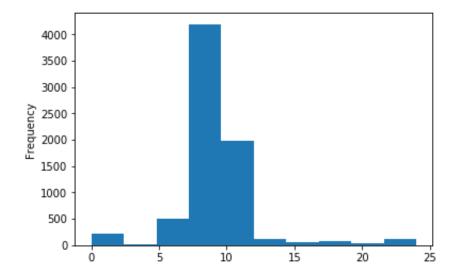
Out[476]:

	open_times	close_times	num_days	price_size	phone_available	num_hours_opening
count	7298.000000	7298.000000	7298.0	7298.000000	7298.000000	7298.000000
mean	9.486074	17.355601	7.0	2.057252	0.950945	7.869528
std	2.955122	4.815404	0.0	0.325923	0.215997	6.730079
min	0.000000	1.000000	7.0	1.000000	0.000000	-23.000000
25%	9.000000	17.355601	7.0	2.057252	1.000000	7.869528
50%	9.486074	17.355601	7.0	2.057252	1.000000	7.869528
75%	10.000000	20.142857	7.0	2.057252	1.000000	11.000000
max	24.000000	24.000000	7.0	3.000000	1.000000	24.000000

In [483]: import matplotlib.pyplot as plt

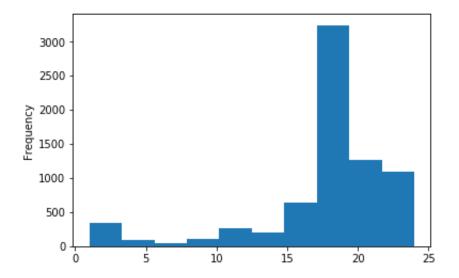
In [487]: final_place_df.open_times.plot(kind='hist')

Out[487]: <matplotlib.axes._subplots.AxesSubplot at 0x1a219faac8>



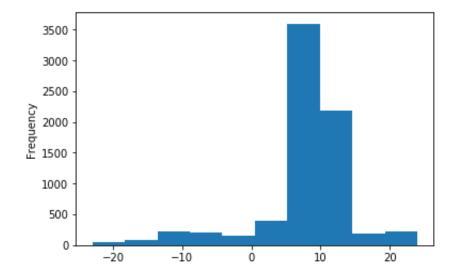
```
In [488]: final_place_df.close_times.plot(kind='hist')
```

Out[488]: <matplotlib.axes._subplots.AxesSubplot at 0x1a21aba518>



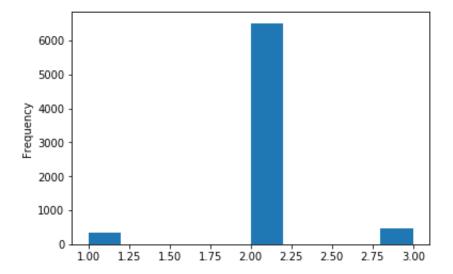
```
In [491]: final_place_df.num_hours_opening.plot(kind='hist')
```

Out[491]: <matplotlib.axes._subplots.AxesSubplot at 0x1a2250cbe0>



```
In [489]: final_place_df.price_size.plot(kind='hist')
```

Out[489]: <matplotlib.axes._subplots.AxesSubplot at 0x1a222f5ef0>



In [492]: final_user_df.describe()

Out[492]:

	open_times	close_times	num_days	price_size	num_hours_opening
count	2287.000000	2287.000000	2287.0	2287.000000	2287.000000
mean	9.953349	16.440953	7.0	2.039904	6.487604
std	3.101200	5.155286	0.0	0.393921	7.084057
min	0.000000	1.000000	7.0	1.000000	-22.714286
25%	9.142857	14.357143	7.0	2.000000	4.000000
50%	9.750000	17.355601	7.0	2.057252	7.869528
75%	11.000000	19.838900	7.0	2.057252	10.183096
max	24.000000	24.000000	7.0	3.000000	24.000000

In []:

Extract Features

```
In [ ]:
In [493]: mean_rate= valid_review_df_dropped.rating.mean()
    avg_place_feat = list(final_place_df.mean().values)
```

```
avg user feat = list(final user df.mean().values)
def feature(row):
   #print(1)
    # constant feature for wx+b
    u = row.iloc[0]
    p = row.iloc[1]
    datum = [1]
    # user data
    #print(4)
    if u in final user df.index:
        u d = list(final user df.loc[u].values)
    else:
        u d = avg user feat
    #print(3)
    #place data
    if p in final place df.index:
        p d = list(final place df.loc[p].values)
    else:
        p d = avg place feat
    # combined data
    cd = []
    cd += [u d[0]-p d[0]]
    cd += [u d[1]-p d[1]]
    cd += [u d[3]-p d[3]]
    cd += [u d[4]-p d[5]]
    cats = []
    if p in place dict:
        cats = place_dict[p]
    total = []
    #print(2)
    if u in user rating cat:
        for c in cats:
            total += user_rating_cat[u][c]
        if len(total)>0:
            avg = np.mean(total)
        else:
            concat = np.concatenate(list(user rating cat[u].values()))
            if len(concat):
                avg = np.mean(concat)
            else:
                avg = mean rate
    else:
        avg = mean rate
     print(datum)
#
     print(u d)
#
     print(cd)
      print(p d)
    datum = datum + u d + p d + cd + [avg]
    return datum
```

```
In [440]: Original_X = valid_review_df_dropped[['gPlusUserId','gPlusPlaceId']]
Y = valid_review_df_dropped.rating
Original_X.head()
```

gPlusPlaceId

Out[440]:

213	110090296112763745586	107762553597304912662
234	102172401918858106420	111280162645644452653
305	113939131914163721697	105539817238070523193
508	109529024903466465111	103566537416444715680
669	104960884941951050236	105361852687606942835

gPlusUserId

```
In [439]: X = Original_X.apply(feature,axis=1)
In []:
```

Train Model and Predict

```
In [451]: from sklearn.model_selection import train_test_split
    from sklearn.svm import SVR

In [456]: X_train, X_test, y_train, y_test = train_test_split(list(X), list(Y),
    test_size=0.1)

In [459]: clf = SVR()
    clf.fit(X_train, y_train)
    predictions = clf.predict(X_test)
    # clf.score(X_test, y_test)
    sum((predictions - y_test)**2) / len(y_test)
```

/anaconda3/lib/python3.7/site-packages/sklearn/svm/base.py:196: Futu reWarning: The default value of gamma will change from 'auto' to 'sc ale' in version 0.22 to account better for unscaled features. Set gamma explicitly to 'auto' or 'scale' to avoid this warning.

"avoid this warning.", FutureWarning)

Out[459]: 0.27126172033829243

The Mean Squared Error here we got from the linear SVR regression is 0.27126, which is a big jump from our baseline model (MSE = 0.826)