sgij_EDA_vertica

August 27, 2019

1 Exploratory data analysis (EDA) on the SGIJ Vertica Dataset

Make a EDA Study for all players activate during a year selected

```
[4]: import math
     from datetime import date
     from datetime import datetime
     import numpy as np
     import pandas as pd
     from pandas.plotting import scatter_matrix
     from pandas.plotting import register_matplotlib_converters
     import matplotlib
     import matplotlib.pyplot as plt
     import vertica python
     from sklearn.cluster import KMeans
     from sklearn import metrics
     from scipy.stats import linregress
     from scipy import stats
     from collections import Counter, OrderedDict
     from IPython.display import display, Markdown
     %matplotlib inline
     register_matplotlib_converters()
```

Connect to Vertica databasef from credentials

NoteNook Year EDA Study

```
[5]: YEAR = 2015
```

Vertica Database connection string

```
'unicode_error': 'strict',
             # SSL is disabled by default
             'ssl': False,
             # using server-side prepared statements is disabled by default
             'use_prepared_statements': False,
             # connection timeout is not enabled by default
             'connection_timeout': 5}
try:
  cnx = vertica_python.connect(**config)
except vertica.connector.Error as err:
  if err.errno == errorcode.ER_ACCESS_DENIED_ERROR:
    print("Something is wrong with your user name or password")
  elif err.errno == errorcode.ER_BAD_DB_ERROR:
    print("Database does not exist")
  else:
    print(err)
```

1.1 Sex player distribution

Get raw Dataset

Plot sex distribution Dataset

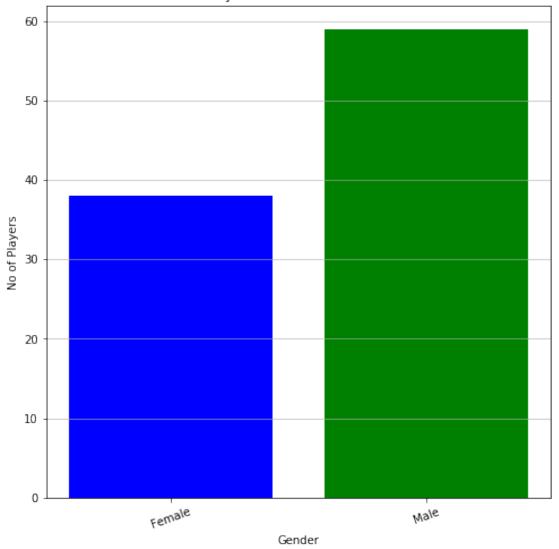
```
[8]: data = np.array(result_sex) # numpy convert to string all values
labels = data[:, 0]
values = data[:, 1].astype(int)

plt.figure(figsize=(8, 8))
plt.bar(labels, values, color=('blue', 'green'))
plt.title('Players Gender Distribution')
plt.xlabel('Gender')
plt.ylabel('No of Players')
plt.xticks(np.arange(labels.size), ('Female', 'Male'), rotation=20)
```

```
plt.grid(axis='y', alpha=0.75)
plt.show()

print('Female Dataset Size: ' + str(values[0]))
print('Male Dataset Size: ' + str(values[1]))
print('-----')
print('Dataset Size: ' + str(sum(values)))
```

Players Gender Distribution



Female Dataset Size: 38
Male Dataset Size: 59

Dataset Size: 97

1.2 Betting Frequency

Get raw dataset for betting frequency, total number of activity days (days on witch a player placed at least one live-action bet))

Plot betting frequency distribution per player

```
[10]: data = np.array(result_betting_frequency) # numpy convert to string all values
    data = data.astype(int)

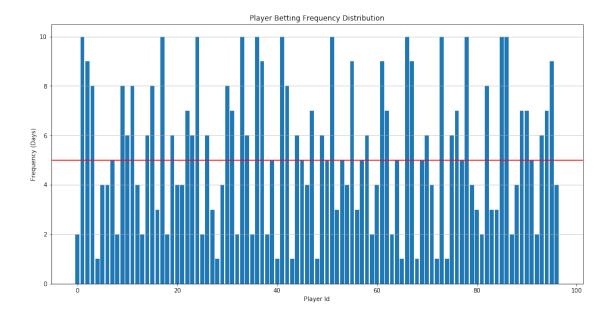
betting_frequency_values = data[:, 0]

mean = round(betting_frequency_values.mean())

#print(betting_frequency_values.size)

plt.figure(figsize=(16, 8))
    plt.bar(np.arange(betting_frequency_values.size), betting_frequency_values)
    plt.title('Player Betting Frequency Distribution')
    plt.xlabel('Player Id')
    plt.ylabel('Frequency (Days)')
    plt.axhline(y=mean, color='r', linestyle='-')
    plt.grid(axis='y', alpha=0.75)
    plt.show()

print('Mean bettings days: ' + str(mean))
```



Mean bettings days: 5.0

1.3 Betting Intensity

Get raw dataset for betting intensity, total number of live-action bets divided by frequency.

Plot betting intensity distribution per player

```
[12]: data = np.array(result_betting_intensity) # numpy convert to string all values
   data = data.astype(int)

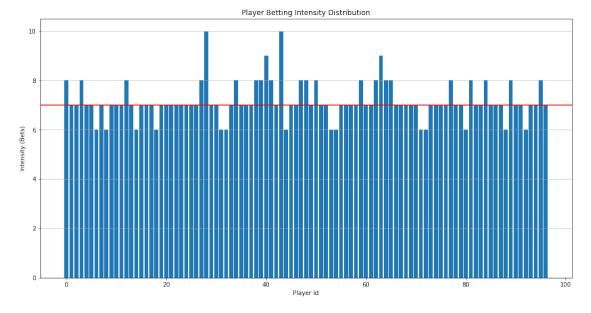
betting_intensity_values = data[:, 0]

mean = round(betting_intensity_values.mean())
```

```
#print(betting_intensity_values.size)

plt.figure(figsize=(16, 8))
plt.bar(np.arange(betting_intensity_values.size), betting_intensity_values)
plt.title('Player Betting Intensity Distribution')
plt.xlabel('Player Id')
plt.ylabel('Intensity (Bets)')
plt.axhline(y=mean, color='r', linestyle='-')
plt.grid(axis='y', alpha=0.75)
plt.show()

print('Mean Betting Intensity : ' + str(mean))
```



Mean Betting Intensity: 7.0

1.4 Betting Intensity Histogram

Get raw dataset for betting Intensity Histogram

```
" GROUP BY pl.operator_id, pl.player_id;") %YEAR

cursor.execute(query)

# return a list of tuples
result_betting_histogram = list(cursor.fetchall())
```

Plot betting betting intesity histogram

```
[14]: data = np.array(result_betting_histogram) # numpy convert to string all values
    data = data.astype(int)

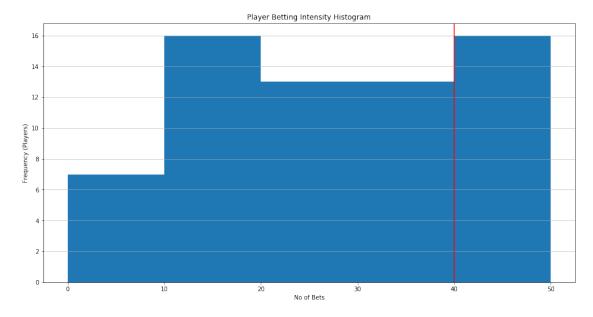
#labels = data[:, 0]

values = data[:, 0]

mean = round(values.mean())

plt.figure(figsize=(16, 8))
    cuenta, cajas, ignorar = plt.hist(values, bins=np.arange(0, 60, 10))
    plt.title('Player Betting Intensity Histogram')
    plt.xlabel('No of Bets')
    plt.ylabel('Frequency (Players)')
    plt.axvline(x=mean, color='r', linestyle='-')
    plt.grid(axis='y', alpha=0.75)
    plt.show()

print('Mean Betting Intensity : ' + str(mean))
```

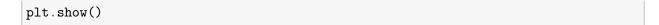


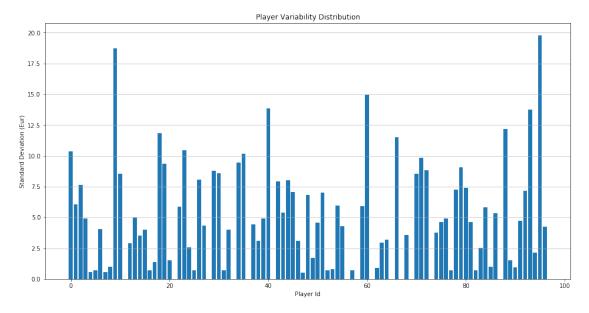
1.5 Betting Variability

Get raw Dataset for Betting Variability. Standard Deviation of wagers

Plot betting betting variability distribution per player

```
[17]: data = np.array(result_betting variability) # numpy convert to string all_
      \rightarrow values()
      data = data.astype(int)
      df = pd.DataFrame({'operator_id': data[:,0],
                          'player_id': data[:,1],
                          'bettings': data[:,2]})
      # calculate the standard deviation grouped by operator and player and replace_
       \rightarrow NaN to zero
      df = df.groupby(['operator_id', 'player_id']).std().fillna(0)
      #print(df.groupby(['operator_id', 'player_id']).head())
      # convert pandas dataframe to numpy vector
      betting_variability_values = df.values.flatten()
      # plot player variance
      plt.figure(figsize=(16, 8))
      plt.bar(np.arange(betting_variability_values.size), betting_variability_values)
      plt.title('Player Variability Distribution')
      plt.xlabel('Player Id')
      plt.ylabel('Standard Deviation (Eur)')
      plt.grid(axis='y', alpha=0.75)
```



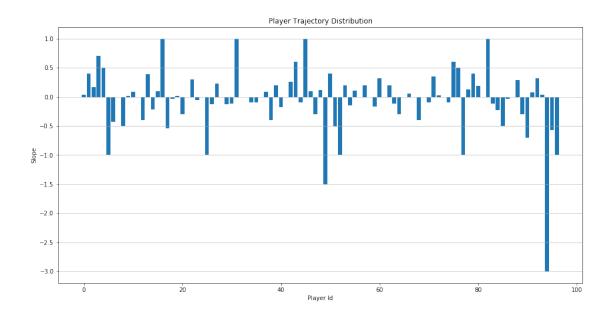


1.6 Betting Trajectory

Get raw Dataset for Betting Trajectory. The tendency to increse or decrease the amount of wagers money

Plot betting trajectory distribution per player

```
'bettings': data[:,2]})
# get bettings grouped by operator and player
result = df.groupby(['operator_id', 'player_id'])['bettings'].apply(list)
# calculate the slope for each bettings collection group by operator and player
betting_trajectory_values = [linregress(np.arange(len(x)), x).slope for x in_
 -result]
# replace NaN to zero
betting_trajectory_values = np.nan_to_num(betting_trajectory_values)
# plot player trayectory
plt.figure(figsize=(16, 8))
plt.bar(np.arange(0, len(betting trajectory_values)), betting_trajectory_values)
plt.title('Player Trajectory Distribution')
plt.xlabel('Player Id')
plt.ylabel('Slope')
plt.grid(axis='y', alpha=0.75)
plt.show()
/home/miguel/.local/lib/python3.5/site-
packages/scipy/stats/_stats_mstats_common.py:116: RuntimeWarning: invalid value
encountered in double_scalars
  slope = r num / ssxm
/home/miguel/.local/lib/python3.5/site-
packages/scipy/stats/_stats_mstats_common.py:126: RuntimeWarning: invalid value
encountered in sqrt
  t = r * np.sqrt(df / ((1.0 - r + TINY)*(1.0 + r + TINY)))
/home/miguel/.local/lib/python3.5/site-
packages/scipy/stats/_stats_mstats_common.py:128: RuntimeWarning: invalid value
encountered in double_scalars
  sterrest = np.sqrt((1 - r**2) * ssym / ssxm / df)
```



1.7 Clustering Betting from predictors (independent characteristics)

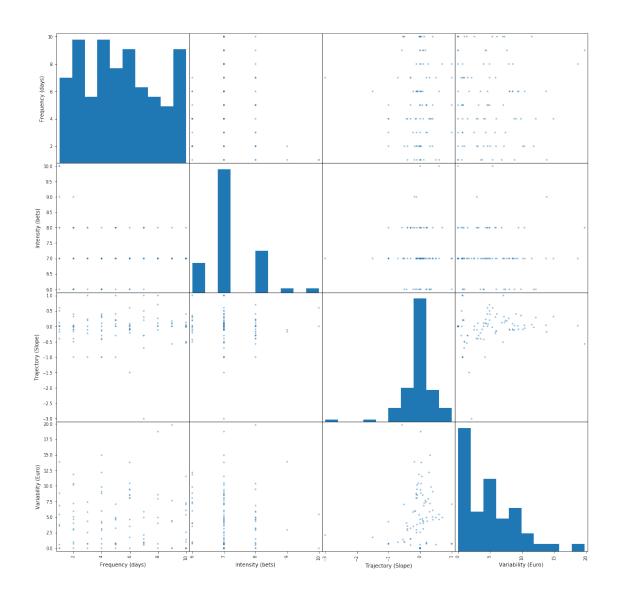
Before clustering to assure comparatibility, we standardized all variables using z transformation

```
[20]: predictors = pd.DataFrame({'Frequency (days)': betting_frequency_values,
                                'Intensity (bets)': betting_intensity_values,
                                'Variability (Euro)': betting_variability_values,
                                'Trajectory (Slope)': betting_trajectory_values})
     X = stats.zscore(predictors)
     # create k-means for the cluster
     model = KMeans(n_clusters=4)
     model.fit(X)
     def ClusterIndicesNumpy(clustNum, labels_array):
         return np.where(labels_array == clustNum)[0]
     labels = Counter(model.labels_)
     centroids = model.cluster_centers_
     columns = ['Cluster 1', 'Cluster 2', 'Cluster 3', 'Cluster 4']
     index =['Frequency(days)', 'Intensity(bets)', 'Variability (Eur)', 'Trajectory
      centroides = centroids.T
     centroides = np.vstack([centroides, [len(ClusterIndicesNumpy(0, model.labels_)),
```

```
Cluster 1 Cluster 2 Cluster 3 Cluster 4
Frequency(days) -1.019761 -0.084759 0.896303 -0.333897
Intensity(bets) 0.929856 -0.564608 -0.140191 -0.332718
Variability (Eur) 0.190396 0.232764 0.284727 -1.888840
Trajectory (Slope) -0.194329 1.204053 -0.531906 -0.711858
N 25.000000 26.000000 35.000000 11.000000
```

Generate Predictors Correlative Scatter MAtrix

```
[21]: scatter_matrix(predictors, figsize=(18, 18))
plt.show()
```



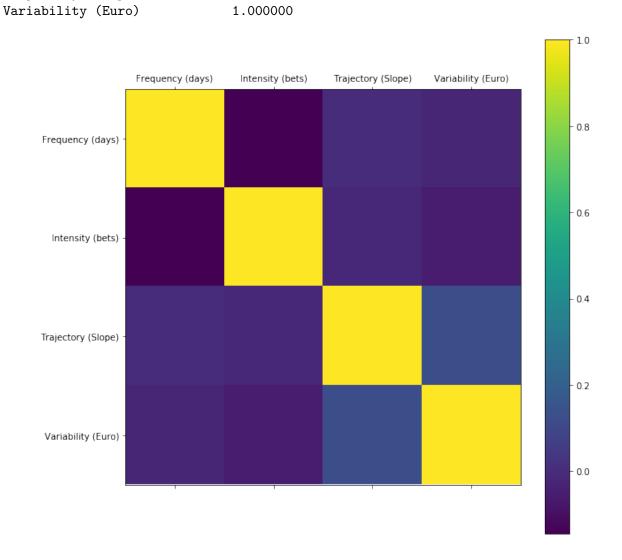
Generate Predictors Correlation Matrix Plot

```
[22]: correlations = predictors.corr()
  #display(Markdown('<font size="2">{}!</font>'.format(correlations)))
  print(correlations)

plt.figure(figsize=(10, 10))
  plt.matshow(correlations, fignum=1)
  plt.xticks(range(len(predictors.columns)), predictors.columns)
  plt.yticks(range(len(predictors.columns)), predictors.columns)
  plt.colorbar()
  plt.show()
```

```
Frequency (days) Intensity (bets) Trajectory (Slope) \ Frequency (days) 1.000000 -0.145690 -0.000841
```

Intensity (bets)	-0.145690	1.000000	-0.019672
Trajectory (Slope)	-0.000841	-0.019672	1.000000
Variability (Euro)	-0.023985	-0.053654	0.131563
	Variability (Euro)		
Frequency (days)	-0.023985		
Intensity (bets)	-0.053654		
Trajectory (Slope)	0.131563		



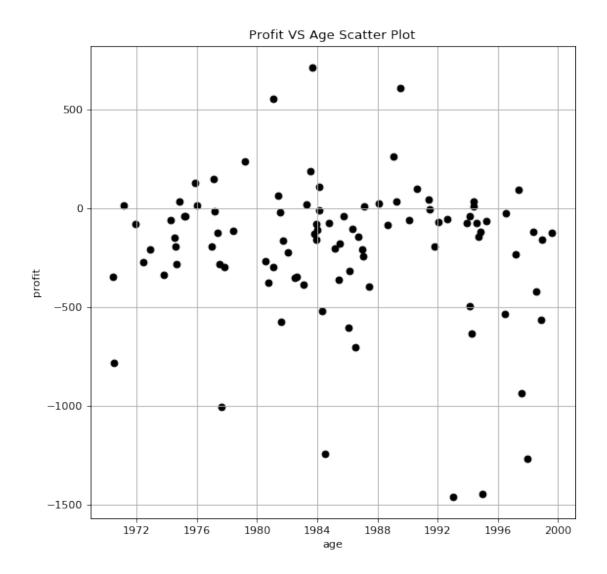
1.8 Clustering Age vs Profit

Study is exist any relation between the age and the profit in the period

Transform date characteristics and create tuples

Plot Age vs Profit Scatter Plot

```
[27]: plt.figure(figsize=(8, 8), dpi=80)
   plt.scatter(dates, profits, color='k')
   plt.title("Profit VS Age Scatter Plot")
   plt.xlabel("age")
   plt.ylabel("profit")
   plt.grid()
   plt.show()
```



Design de k-means with 2 clusters model for dataset

```
[28]: model = KMeans(n_clusters=2).fit(X)
```

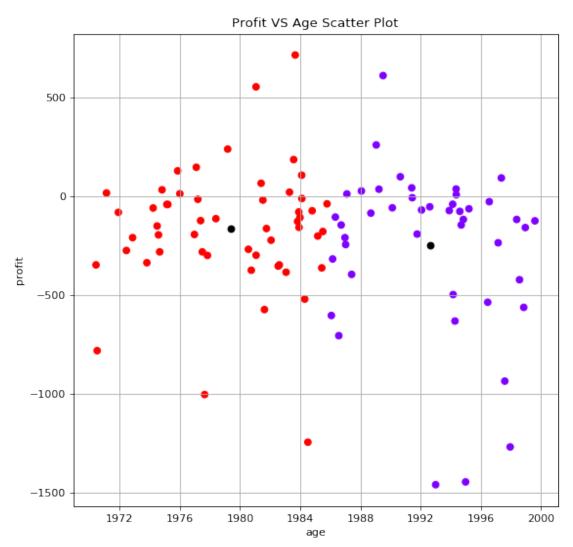
Print k-means centroides

[(datetime.datetime(1992, 8, 9, 11, 24, 33), -247.4772727272727275),

(datetime.datetime(1979, 5, 18, 0, 10, 11), -164.0188679245283)]

Plot the scatter plot and the centroides for tha dataset

```
[30]: plt.figure(figsize=(8, 8), dpi=80)
    plt.scatter(dates, profits, c=model.labels_, cmap='rainbow')
    plt.scatter(centroides_dates ,centroides_profits, color='black')
    plt.title("Profit VS Age Scatter Plot")
    plt.xlabel("age")
    plt.ylabel("profit")
    plt.grid()
    plt.show()
```



```
[]:
```