

Exploratory_Work-Group5

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0.1 Exploratory Data Analysis

5.11 Assignment

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```
[1]: import os
from pyspark.sql import SparkSession
import pyspark.sql.types as typ
import pyspark.sql.functions as F
from pyspark.sql import DataFrame
from functools import reduce
from pyspark.sql.functions import col, asc
from pyspark.sql import SQLContext
from pyspark.sql.types import StructType
from pyspark.sql.types import StructField
from pyspark.sql.types import StringType, DoubleType, DateType
from pyspark.sql import SparkSession
import pandas as pd
import numpy as np
```

```
[2]: spark = SparkSession \
    .builder \
    .getOrCreate()

sc = spark.sparkContext
```

0.1.1 Based on data from : <https://www.fec.gov/data/browse-data/?tab=bulk-data>

This is House/Senate campaign finance data:

```
[3]: #storing all files in dictionaries for easy reference
f_campaigns = {'2019-2020': 'campaign_2020.txt',
               '2017-2018': 'campaign_2018.txt',
               '2015-2016': 'campaign_2016.txt'}
```

```

#no header provided by FEC to upload
header_row = ['CAND_ID',
              'CAND_NAME',
              'CAND_ICI',
              'PTY_CD',
              'CAND_PTY_AFFILIATION',
              'TTL_RECEIPTS',
              'TRANS_FROM_AUTH',
              'TTL_DISB',
              'TRANS_TO_AUTH',
              'COH_BOP',
              'COH_COP',
              'CAND_CONTRIB',
              'CAND_LOANS',
              'OTHER_LOANS',
              'CAND_LOAN_REPAY',
              'OTHER_LOAN_REPAY',
              'DEBTS_OWED_BY',
              'TTL_INDIV_CONTRIB',
              'CAND_OFFICE_ST',
              'CAND_OFFICE_DISTRICT',
              'SPEC_ELECTION',
              'PRIM_ELECTION',
              'RUN_ELECTION',
              'GEN_ELECTION',
              'GEN_ELECTION_PRECENT',
              'OTHER_POL_CMTE_CONTRIB',
              'POL_PTY_CONTRIB',
              'CVG_END_DT',
              'INDIV_REFUNDS',
              'CMTE_REFUNDS']

```

Create dataframes and combine the three files together for analysis:

```

[63]: df_temp20 = sc.textFile(f_campaigns['2019-2020']).map(lambda row: [elem for
    ↪elem in row.split('|')])
df_temp18 = sc.textFile(f_campaigns['2017-2018']).map(lambda row: [elem for
    ↪elem in row.split('|')])
df_temp16 = sc.textFile(f_campaigns['2015-2016']).map(lambda row: [elem for
    ↪elem in row.split('|')])

#default to stringtype for ease of loading, then adjust below:
fields = [*[typ.StructField(h[:], typ.StringType(), True) for h in header_row]]
schema = typ.StructType(fields)

df_20 = spark.createDataFrame(df_temp20, schema)
df_18 = spark.createDataFrame(df_temp18, schema)

```

```
df_16 = spark.createDataFrame(df_temp16, schema)

dfs = [df_20, df_18, df_16]

df = reduce(DataFrame.unionAll, dfs)

#casting necessary numeric values:
df = df.withColumn('TTL_RECEIPTS', df['TTL_RECEIPTS'].cast(DoubleType()))
df = df.withColumn('TTL_INDIV_CONTRIB', df['TTL_INDIV_CONTRIB'].
    ↪cast(DoubleType()))
df = df.withColumn('CAND_CONTRIB', df['CAND_CONTRIB'].cast(DoubleType()))
df = df.withColumn('OTHER_POL_CMTE_CONTRIB', df['OTHER_POL_CMTE_CONTRIB'].
    ↪cast(DoubleType()))
df = df.withColumn('POL_PTY_CONTRIB', df['POL_PTY_CONTRIB'].cast(DoubleType()))
```

Number of records:

```
[51]: df.count()
```

```
[51]: 7149
```

Number of columns:

```
[16]: len(df.columns)
```

```
[16]: 30
```

Statistical summary of response variable:

Our statistical summary will be based on whether a candidate won or lost the relevant political race.

We are still gathering and joining that data to this set.

Statistical summary of potential predictor variables:

Total receipts -

```
[54]: df.select('TTL_RECEIPTS').describe().show()
```

```
+-----+-----+
|summary|      TTL_RECEIPTS|
+-----+-----+
|  count|              7149|
|   mean| 2266553.450893833|
```

stddev	6.004019782547508E7
min	0.0
max	4.824617973E9
+-----+-----+	

Contributions by individuals -

```
[55]: df.select('TTL_INDIV_CONTRIB').describe().show()
```

summary	TTL_INDIV_CONTRIB
+-----+-----+	
count	7149
mean	3570474.139283814
stddev	2.231747534710199E8
min	0.0
max	1.8853982587E10
+-----+-----+	

Contributions by candidates -

```
[57]: df.select('CAND_CONTRIB').describe().show()
```

summary	CAND_CONTRIB
+-----+-----+	
count	7149
mean	621843.8302853543
stddev	3.607460293607057E7
min	0.0
max	2.831281203E9
+-----+-----+	

Contributions from party committees -

```
[64]: df.select('POL_PTY_CONTRIB').describe().show()
```

summary	POL_PTY_CONTRIB
+-----+-----+	
count	7149
mean	1594.4136438662752
stddev	37564.26071346217
min	0.0
max	3100000.0
+-----+-----+	

Contributions from other political committees -

```
[65]: df.select('OTHER_POL_CMTE_CONTRIB').describe().show()
```

```
+-----+-----+
|summary|OTHER_POL_CMTE_CONTRIB|
+-----+-----+
|  count|                7149|
|   mean|    449703.3054748916|
| stddev|  2.2751089972151406E7|
|    min|                 0.0|
|    max|         1.9235003E9|
+-----+-----+
```

Candidate status (C = Challenger, O = Open, I = Incumbent) -

```
[61]: #some data cleaning to do for the blanks
df.groupby('CAND_ICI').count().orderBy('count', ascending = False).show()
```

```
+-----+-----+
|CAND_ICI|count|
+-----+-----+
|      C| 3857|
|      O| 1779|
|      I| 1441|
|      |   72|
+-----+-----+
```

Candidate party affiliation (count) -

```
[41]: df.groupby('CAND_PTY_AFFILIATION').count().orderBy('count', ascending = False).
      ↪show()
```

```
+-----+-----+
|CAND_PTY_AFFILIATION|count|
+-----+-----+
|          DEM| 3227|
|          REP| 3193|
|          IND|  272|
|          LIB|  136|
|          GRE|   55|
|          NPA|   37|
|          DFL|   36|
|          OTH|   35|
|          NNE|   32|
|          UNK|   26|
|          UN |   23|
```

	CON		14	
	W		9	
	NON		7	
	IDP		5	
	NOP		5	
	AMP		3	
	PPY		3	
	SEP		3	
	UNI		3	

+-----+

only showing top 20 rows

Candidate state (count) -

```
[59]: df.groupby('CAND_OFFICE_ST').count().orderBy('count', ascending = False).show()
```

	CAND_OFFICE_ST		count	
--	----------------	--	-------	--

+-----+

	CA		695	
	TX		604	
	OO		513	
	FL		486	
	NY		364	
	PA		263	
	GA		235	
	IL		228	
	NC		224	
	MI		189	
	OH		183	
	VA		171	
	AZ		170	
	NJ		170	
	IN		148	
	TN		147	
	MD		147	
	WA		132	
	MN		129	
	CO		125	

+-----+

only showing top 20 rows

```
[66]: df.select('CAND_NAME',
               'CAND_OFFICE_ST',
               'CAND_PTY_AFFILIATION',
               'CAND_ICI',
```

```
'TTL_RECEIPTS',
'CAND_CONTRIB',
'TTL_INDIV_CONTRIB',
'POL_PTY_CONTRIB',
'OTHER_POL_CMTE_CONTRIB').show(5)
```

```
+-----+-----+-----+-----+-----+
|          CAND_NAME|CAND_OFFICE_ST|CAND_PTY_AFFILIATION|CAND_ICI|TTL_RECEIPTS|
CAND_CONTRIB|TTL_INDIV_CONTRIB|POL_PTY_CONTRIB|OTHER_POL_CMTE_CONTRIB|
+-----+-----+-----+-----+-----+
|      YOUNG, DONALD E|      AK|      REP|      I| 1362383.63|
0.0|      637025.31|      0.0|      584444.63|
|      GALVIN, ALYSE|      AK|      IND|      C| 2266364.63|
3394.63|      2116292.8|      0.0|      109350.0|
|      AVERHART, JAMES|      AL|      DEM|      0| 50126.74|
0.0|      23281.74|      0.0|      0.0|
|      GARDNER, KIANI A|      AL|      DEM|      0| 118661.85|
764.97|      92896.88|      0.0|      19000.0|
|COLLINS, FREDERIC...|      AL|      DEM|      0| 62935.42|
56500.0|      5917.12|      0.0|      0.0|
+-----+-----+-----+-----+-----+
only showing top 5 rows
```

[]: