

```
In [44]: """
EXERCISE

Techniques used:
- read_csv
- unique()
- .map()
- .value_counts()
- Boolean Indexing
- Data cleaning
- GroupBy
- .plot()
"""
```

```
Out[44]: '\nEXERCISE\n\nTechniques used:\n- read_csv\n- unique()\n- .map()\n- .value_counts()\n- Boolean Indexing\n- Data cleaning\n- GroupBy\n- .plot()\n'
```

```
In [45]: %matplotlib inline

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

```
In [46]: # 2012 election donation dataset
fec = pd.read_csv('data/2012-election.csv', low_memory=False)

# Inspect random row
fec.ix[123]
```

```
Out[46]: cmte_id          C00410118
cand_id          P20002978
cand_nm          Bachmann, Michelle
contbr_nm        RAINEY, PATRICK
contbr_city      WINTER GARDEN
contbr_st        FL
contbr_zip        347771599
contbr_employer  INTL TRADING COMPANY
contbr_occupation BROKER
contb_receipt_amt 250
contb_receipt_dt  20-JUN-11
receipt_desc      NaN
memo_cd           NaN
memo_text         NaN
form_tp          SA17A
file_num         736166
Name: 123, dtype: object
```

```
In [47]: candidates = fec.cand_nm.unique()
candidates
```

```
Out[47]: array(['Bachmann, Michelle', 'Romney, Mitt', 'Obama, Barack',
               'Roemer, Charles E. 'Buddy' III', 'Pawlenty, Timothy',
               'Johnson, Gary Earl', 'Paul, Ron', 'Santorum, Rick', 'Cain, Herman',
               'Gingrich, Newt', 'McCotter, Thaddeus G', 'Huntsman, Jon',
               'Perry, Rick'], dtype=object)
```

```
In [48]: # Map candidates to parties
parties = {'Bachmann, Michelle': 'Republican',
           'Cain, Herman': 'Republican',
           'Gingrich, Newt': 'Republican',
           'Huntsman, Jon': 'Republican',
           'Johnson, Gary Earl': 'Republican',
           'McCotter, Thaddeus G': 'Republican',
           'Obama, Barack': 'Democrat',
           'Paul, Ron': 'Republican',
           'Perry, Rick': 'Republican',
           'Roemer, Charles E. 'Buddy' III': 'Republican',
           'Romney, Mitt': 'Republican',
           'Santorum, Rick': 'Republican'}
```

```
In [49]: # Add party column to dataframe
fec['party'] = fec.cand_nm.map(parties)
```

```
In [50]: # How many individual donations by party
fec['party'].value_counts()
```

```
Out[50]: Democrat      593746
Republican    403430
dtype: int64
```

```
In [51]: # Dataset include refunds (negative contributions)
(fec.contb_receipt_amt > 0).value_counts()
```

```
Out[51]: True      991475
False    10256
dtype: int64
```

```
In [52]: # Simplify dataset to just positive contributions and Barack Obama, and Mitt
data = fec[fec.contb_receipt_amt > 0]
nominees_data = fec[fec.cand_nm.isin(['Obama, Barack', 'Romney, Mitt'])]
```

```
In [53]: # Donation Stats by Occupation
data.contbr_occupation.value_counts()[:10]
```

```
Out[53]: RETIRED                233990
INFORMATION REQUESTED          35107
ATTORNEY                      34286
HOMEMAKER                    29931
PHYSICIAN                    23432
INFORMATION REQUESTED PER BEST EFFORTS 21138
ENGINEER                     14334
TEACHER                      13990
CONSULTANT                   13273
PROFESSOR                    12555
dtype: int64
```

```
In [54]: # Clean data
occupation_map = {
    'INFORMATION REQUESTED PER BEST EFFORTS': 'NOT PROVIDED',
    'INFORMATION REQUESTED': 'NOT PROVIDED',
    'INFORMATION REQUESTED (BEST EFFORTS)': 'NOT PROVIDED',
    'C.E.O.': 'CEO'
}

# If no mapping, pass through original value
f = lambda x: occupation_map.get(x, x)
data.contbr_occupation = data.contbr_occupation.map(f)

employer_map = {
    'INFORMATION REQUESTED PER BEST EFFORTS': 'NOT PROVIDED',
    'INFORMATION REQUESTED': 'NOT PROVIDED',
    'SELF': 'SELF-EMPLOYED',
    'SELF EMPLOYED': 'SELF-EMPLOYED',
}
f = lambda x: employer_map.get(x, x)
data.contbr_employer = data.contbr_employer.map(f)
```

```
In [61]: by_occupation = data.pivot_table('contb_receipt_amt',
                                           index='contbr_occupation',
                                           columns='party', aggfunc='sum')
by_occupation.head()
```

```
Out[61]:
```

party	Democrat	Republican
contbr_occupation		
MIXED-MEDIA ARTIST / STORYTELLER	100	NaN
AREA VICE PRESIDENT	250	NaN
RESEARCH ASSOCIATE	100	NaN
TEACHER	500	NaN
THERAPIST	3900	NaN

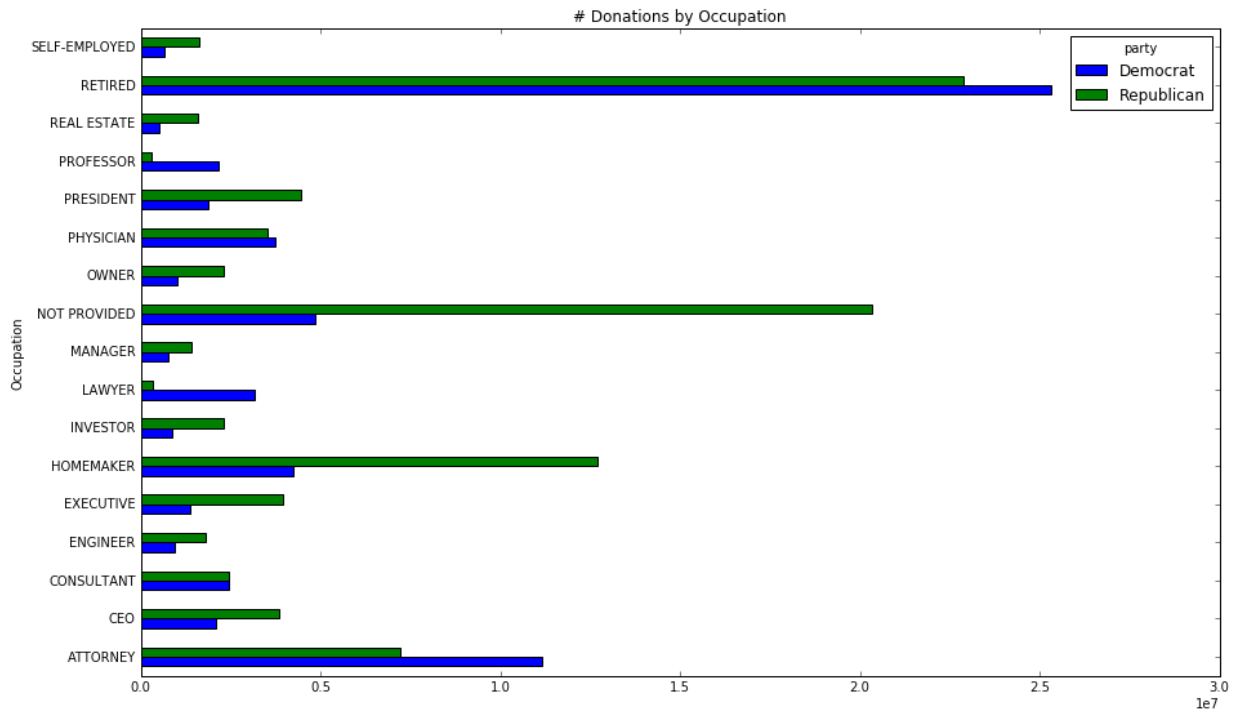
```
In [56]: over_2mm = by_occupation[by_occupation.sum(1) > 2000000]
over_2mm
```

Out[56]:

party	Democrat	Republican
contbr_occupation		
ATTORNEY	11141982.97	7214482.230000
CEO	2074974.79	3862590.520000
CONSULTANT	2459912.71	2434349.400000
ENGINEER	951525.55	1802248.700000
EXECUTIVE	1355161.05	3964650.090000
HOMEMAKER	4248875.80	12715782.760000
INVESTOR	884133.00	2295168.920000
LAWYER	3160478.87	327224.320000
MANAGER	762883.22	1394467.370000
NOT PROVIDED	4866973.96	20337131.830000
OWNER	1001567.36	2294861.920000
PHYSICIAN	3735124.94	3539687.170000
PRESIDENT	1878509.95	4464514.840000
PROFESSOR	2165071.08	294922.730000
REAL ESTATE	528902.09	1585502.250000
RETIRED	25305116.38	22863149.489999
SELF-EMPLOYED	672393.40	1606652.540000

```
In [57]: # We can quickly visualize this with a barchart
ax = over_2mm.plot(kind='barh', figsize=(15,9), title="# Donations by Occupation")
ax.set_ylabel('Occupation')
```

```
Out[57]: <matplotlib.text.Text at 0x104ea6590>
```



```
In [58]: # Get top donor occupations and companies for Obama and Romney
def get_top_amounts(group, key, n=5):
    totals = group.groupby(key)['contb_receipt_amt'].sum()
    return totals.order(ascending=False)[:n]
```

```
In [59]: grouped = nominees_data.groupby('cand_nm')
```

```
# Top 10 donors by occupation
grouped.apply(get_top_amounts, 'contbr_occupation', n=10)
```

```
Out[59]: cand_nm      contbr_occupation      contb_receipt_amt
Obama, Barack  RETIRED                25270507.23
              ATTORNEY                11126932.97
              INFORMATION REQUESTED    4849801.96
              HOMEMAKER               4243394.30
              PHYSICIAN               3732387.44
              LAWYER                  3159391.87
              CONSULTANT              2459812.71
              PROFESSOR               2163571.08
              CEO                     2069784.79
              PRESIDENT               1878009.95
Romney, Mitt   RETIRED                11266949.23
              INFORMATION REQUESTED PER BEST EFFORTS 11173374.84
              HOMEMAKER               8037250.86
              ATTORNEY                5302578.82
              PRESIDENT               2403439.77
              EXECUTIVE               2230653.79
              C.E.O.                  1893931.11
              INVESTOR                1494725.12
              CONSULTANT              1404576.94
              PHYSICIAN               1332996.34
Name: contb_receipt_amt, dtype: float64
```

```
In [60]: # Top 10 donors by employer
grouped.apply(get_top_amounts, 'contbr_employer', n=10)
```

```
Out[60]: cand_nm      contbr_employer      contb_receipt_amt
Obama, Barack  RETIRED                22665902.20
              SELF-EMPLOYED           17038455.96
              NOT EMPLOYED            8584118.70
              INFORMATION REQUESTED    5036178.37
              HOMEMAKER               2599987.04
              SELF                    1076531.20
              SELF EMPLOYED            469290.00
              STUDENT                  318831.45
              VOLUNTEER                257104.00
              MICROSOFT                215585.36
Romney, Mitt   INFORMATION REQUESTED PER BEST EFFORTS 11827237.12
              RETIRED                11264701.35
              HOMEMAKER               8037000.86
              SELF-EMPLOYED           7260882.29
              STUDENT                  488642.82
              CREDIT SUISSE            265650.00
              MORGAN STANLEY           262266.00
              GOLDMAN SACH & CO.        233250.00
              BARCLAYS CAPITAL         162750.00
              H.I.G. CAPITAL           139500.00
Name: contb_receipt_amt, dtype: float64
```

```
In [ ]:
```

