

In [52]:

```
import sys
import pandas as pd
import numpy as np
import seaborn as sns
import matplotlib.pyplot as plt
sns.set()
```

In [2]:

```
df=pd.read_csv("G:/研究生学习资料/Illinois Courses/Fall 2019/IE 598 Machine Learning/assignment/HW3/
df.head()
```

Out[2]:

	CUSIP	Ticker	Issue Date	Maturity	1st Call Date	Moodys	S_and_P	Fitch	Bloomberg Composite Rating
0	000324AA1	FLECIN	7/1/2014	7/1/2019	10/23/2017	Nan	Nan	Nan	Nan
1	00080QAB1	RBS	3/15/2004	6/4/2018	Nan	Ba1	BB+	BBB	BB-
2	00081TAD0	ACCO	5/14/2010	3/15/2015	Nan	WR	NR	BB+	NF
3	00081TAH1	ACCO	6/17/2013	4/30/2020	Nan	WR	NR	WD	NF
4	00081TAJ7	ACCO	12/22/2016	12/15/2024	12/15/2019	B1	BB-	BB	BB

5 rows × 37 columns

In [11]:

```
print("Number of Rows of Data = "+str(df.shape[0]))
print("Number of Columns of Data = "+str(df.shape[1]))
print("CSV size: "+str(df.shape))
```

Number of Rows of Data = 2721
 Number of Columns of Data = 37
 CSV size: (2721, 37)

In [31]:

```
print(df.iloc[0,:])  
#Reference: https://zhuanlan.zhihu.com/p/31360526
```

CUSIP	000324AA1
Ticker	FLECIN
Issue Date	7/1/2014
Maturity	7/1/2019
1st Call Date	10/23/2017
Moodys	Nan
S_and_P	Nan
Fitch	Nan
Bloomberg Composite Rating	Nan
Coupon	12
Issued Amount	4.05e+08
Maturity Type	CALLABLE
Coupon Type	PAY-IN-KIND
Maturity At Issue months	60.87
Industry	Real Estate
LiquidityScore	10.8914
Months in JNK	Nan
Months in HYG	Nan
Months in Both	Nan
IN ETF	No
LIQ SCORE	0.108914
n_trades	301
volume_trades	2.64004e+08
total_median_size	1e+06
total_mean_size	877089
n_days_trade	128
days_diff_max	1132
percent_intra_dealer	0.00664452
percent_uncapped	0.292359
bond_type	5
Client Trade Percentage	0.521595
weekly_mean_volume	3.10593e+06
weekly_median_volume	2e+06
weekly_max_volume	1.898e+07
weekly_min_volume	60000
weekly_mean_ntrades	3.54118
weekly_median_ntrades	1

Name: 0, dtype: object

In [32]:

```
print(df.dtypes)
```

CUSIP	object
Ticker	object
Issue Date	object
Maturity	object
1st Call Date	object
Moodys	object
S_and_P	object
Fitch	object
Bloomberg Composite Rating	object
Coupon	float64
Issued Amount	float64
Maturity Type	object
Coupon Type	object
Maturity At Issue months	float64
Industry	object
LiquidityScore	float64
Months in JNK	object
Months in HYG	object
Months in Both	object
IN ETF	object
LIQ SCORE	float64
n_trades	int64
volume_trades	float64
total_median_size	float64
total_mean_size	float64
n_days_trade	int64
days_diff_max	int64
percent_intra_dealer	float64
percent_uncapped	float64
bond_type	int64
Client_Trade_Percentage	float64
weekly_mean_volume	float64
weekly_median_volume	float64
weekly_max_volume	float64
weekly_min_volume	float64
weekly_mean_ntrades	float64
weekly_median_ntrades	int64
dtype:	object

In [34]:

```
# descriptive statistics for the numeric variables
percentiles = np.array([2.5, 25, 50, 75, 97.5])
df.describe()
#https://pandas.pydata.org/pandas-docs/stable/reference/api/pandas.DataFrame.describe.html
```

Out[34]:

	Coupon	Issued Amount	Maturity At Issue months	LiquidityScore	LIQ SCORE	n_trades	volu
count	2721.000000	2.721000e+03	2721.000000	2721.000000	2721.000000	2721.000000	2.7
mean	10.307872	8.299295e+08	113.968997	18.218230	0.182182	2700.696435	7.2
std	63.051382	5.802790e+08	101.893176	7.872071	0.078721	5572.262205	1.0
min	0.000000	3.700000e+08	11.930000	4.388758	0.043888	1.000000	7.0
25%	5.000000	5.000000e+08	65.170000	12.738630	0.127386	116.000000	6.1
50%	6.250000	6.500000e+08	97.370000	16.538471	0.165385	674.000000	3.4
75%	7.750000	1.000000e+09	121.770000	22.120108	0.221201	2467.000000	9.3
max	999.000000	7.364026e+09	1217.570000	54.673908	0.546739	57935.000000	8.9

8 rows × 21 columns



In [56]:

```
# unique categories in each categorical attribute
BCR=df['Bloomberg Composite Rating']
Categories=set(BCR)
sys.stdout.write("Unique Label Values \n")
print(Categories)
sys.stdout.write(" \n")
```

Unique Label Values

```
{'AA-', 'CC+', 'A+', 'BBB+', 'DDD', 'CCC-', 'CC-', 'BBB-', 'CCC', 'A-', 'BB-', 'BBB', 'AAA', 'AA+', 'Nan', 'NR', 'BB', 'B', 'C+', 'DD+', 'B+', 'CC', 'BB+', 'B-', 'CCC+', 'A', 'AA', 'C'}
```

In [57]:

```

sys.stdout.write("\nCounts for Each Value of Categorical Label \n")
catCount={}
for elt in BCR:
    if elt in catCount:
        catCount[elt] += 1
    else:
        catCount[elt]=1
print(catCount)

```

Counts for Each Value of Categorical Label

```

{'Nan': 41, 'BB+': 258, 'NR': 1136, 'BB-': 196, 'BB': 179, 'AA-': 63, 'A': 22, 'B-': 124, 'B+': 150, 'B': 116, 'A+': 26, 'CCC+': 70, 'CCC': 54, 'BBB-': 163, 'BBB': 33, 'CCC-': 16, 'CC': 6, 'CC+': 16, 'AA': 7, 'AAA': 12, 'A-': 8, 'DDD': 2, 'BBB+': 11, 'DD+': 2, 'C': 3, 'C+': 5, 'CC-': 1, 'AA+': 1}

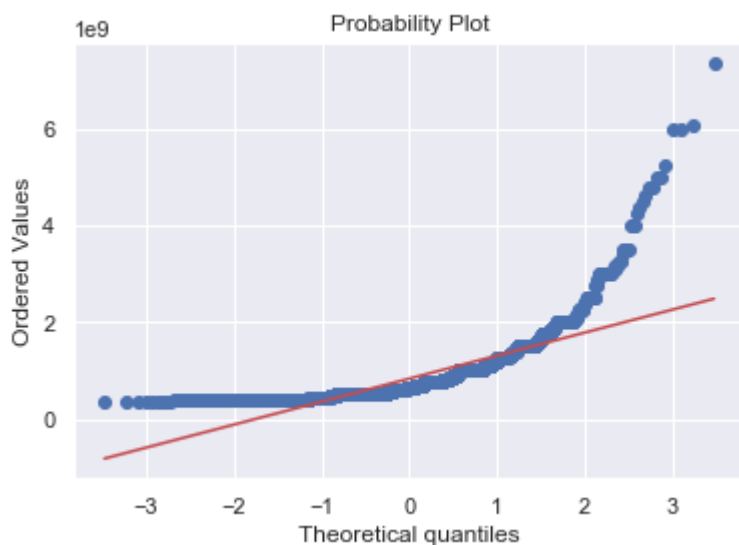
```

In [66]:

```

import scipy.stats as stats
fig=plt.figure()
ax=fig.add_subplot(111)
stats.probplot(df['Issued Amount'], dist="norm",plot=ax)
plt.show()

```



In [67]:

```
from pandas import DataFrame
print(df.head())
print(df.tail())
summary = df.describe()
print(summary)
```

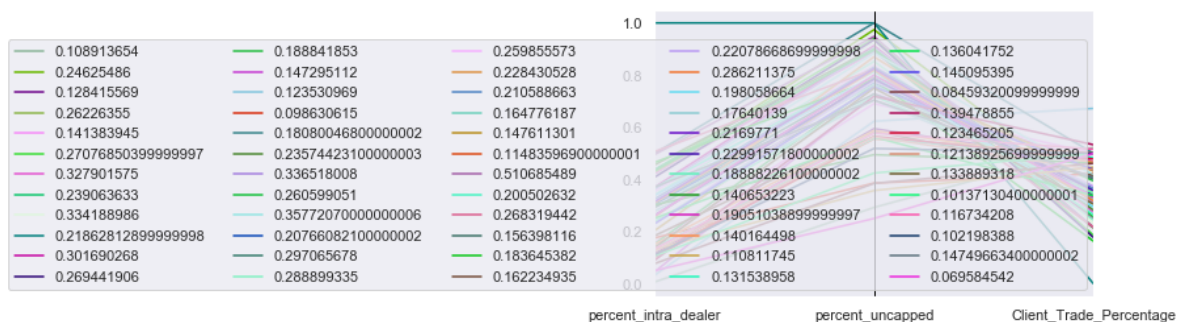
	CUSIP	Ticker	Issue Date	Maturity	1st Call Date	Moody's	S_and_P	\
0	000324AA1	FLECIN	7/1/2014	7/1/2019	10/23/2017	Nan	Nan	
1	00080QAB1	RBS	3/15/2004	6/4/2018	Nan	Ba1	BB+	
2	00081TAD0	ACCO	5/14/2010	3/15/2015	Nan	WR	NR	
3	00081TAH1	ACCO	6/17/2013	4/30/2020	Nan	WR	NR	
4	00081TAJ7	ACCO	12/22/2016	12/15/2024	12/15/2019	B1	BB-	

	Fitch	Bloomberg	Composite Rating	Coupon	...	percent_intra_dealer	\
0	Nan		Nan	12.00	...	0.006645	
1	BBB		BB+	4.65	...	0.425018	
2	BB+		NR	10.63	...	0.115207	
3	WD		NR	6.75	...	0.426332	
4	BB		BB-	5.25	...	0.157216	

	percent_uncapped	bond_type	Client_Trade_Percentage	weekly_mean_volume	\
0	0.292359	5	0.521595	3105926.765	
1	0.974071	2	0.337071	1721696.774	
2	0.594470	5	0.467742	4200313.433	
3	0.892462	3	0.212864	6321559.783	
4	0.600700	5	0.500000	5000000.000	

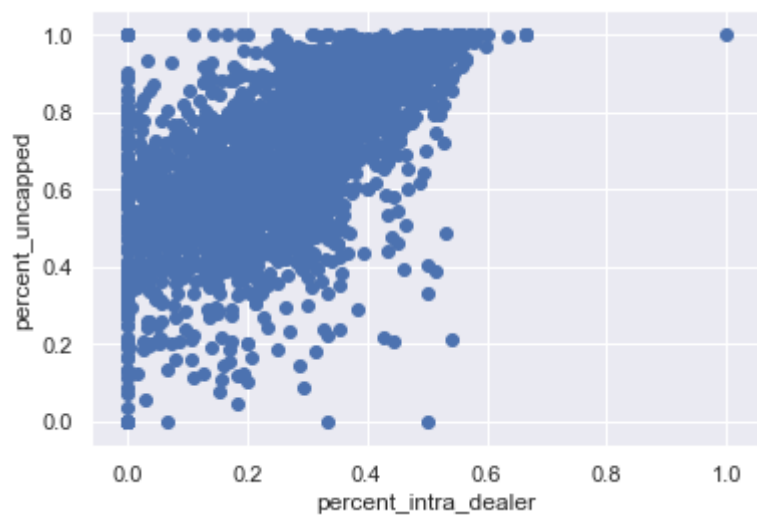
In [96]:

```
from pandas.plotting import parallel_coordinates
tmp = df[['LIQ SCORE', 'percent_intra_dealer', 'percent_uncapped', 'Client_Trade_Percentage']][0:60]
parallel_coordinates(tmp, 'LIQ SCORE')
plt.legend(loc='best', ncol=5)
plt.show()
```



In [98]:

```
plt.scatter(df['percent_intra_dealer'], df['percent_uncapped'])
plt.xlabel("percent_intra_dealer")
plt.ylabel(("percent_uncapped"))
plt.show()
```

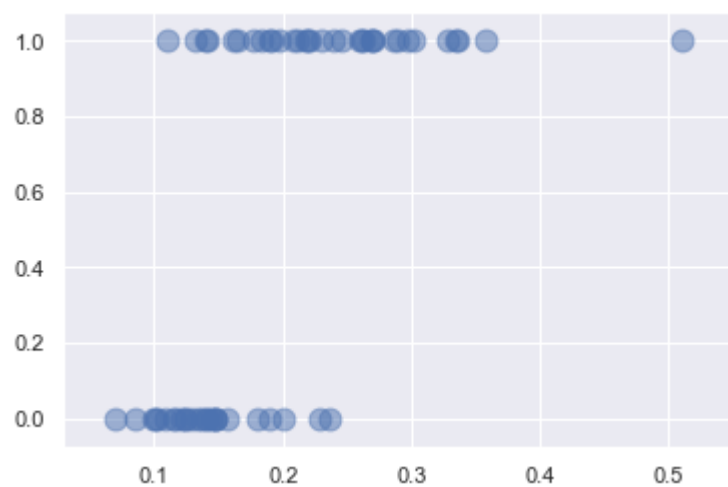


In [117]:

```
target=[]
for i in range(60):
    if df['IN ETF'][i] == "Yes":
        target.append(1.0)
    else:
        target.append(0.0)
tmp=np.array(target)
plt.scatter(df['LIQ SCORE'][:60], target, alpha=0.5, s=120)
```

Out[117]:

<matplotlib.collections.PathCollection at 0x2095ba94128>



In [120]:

```
sys.stdout.write("Correlation between attribute V(n_trades) and W(volume_trades) \n")
tmp=np.corrcoef(df['n_trades'], df['volume_trades'])[0, 1]
print(tmp)
```

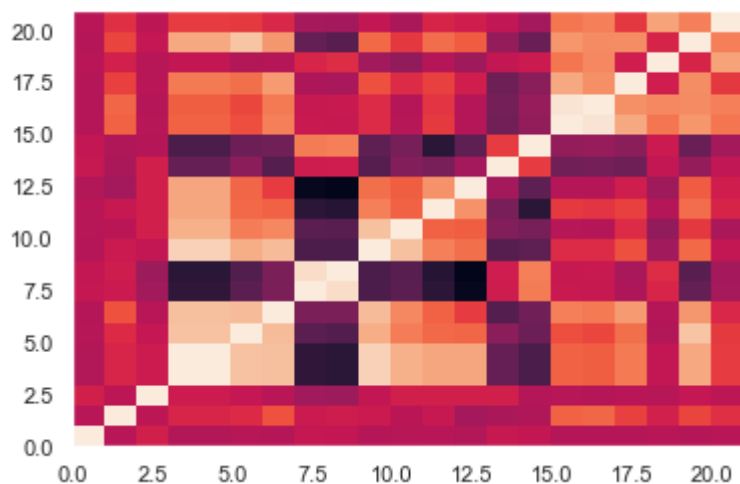
Correlation between attribute V(n_trades) and W(volume_trades)
0.7693223728724927

In [121]:

```
from pandas import DataFrame
corMat = DataFrame(df.corr())
```

In [122]:

```
plt.pcolor(corMat)
plt.show()
```



In [123]:

```
print("My name is {Zihan Chen}")
print("My NetID is: {zihanc7}")
print("I hereby certify that I have read the University policy on Academic Integrity and that I am not in violation.")
```

My name is {Zihan Chen}

My NetID is: {zihanc7}

I hereby certify that I have read the University policy on Academic Integrity and that I am not in violation.

In []: