## Untitled

January 12, 2020

# 1 Multifactor CAPM Analysis

```
[124]: #Importing Libraries
      import pandas as pd
      import numpy as np
      import matplotlib.pyplot as plt
      import statsmodels.api as sm
[125]: #Importing the dataset
      df = pd.read_csv("Multifactor CAPM.csv",parse_dates=True)
[126]: df.head()
[126]:
              Date
                       SMB %
                                 HML %
                                           WML % VIX %
                                                             Rm %
                                                                       Rf % \
      0 01-Jan-15 0.634136 1.568746 0.241641
                                                  -0.63 0.349101
                                                                  0.031936
      1 02-Jan-15 0.229000 0.281989 0.504802
                                                 -8.19 0.905457
                                                                   0.031936
      2 05-Jan-15 0.700766 -0.323167 0.609444
                                                   2.59 -0.103746
                                                                   0.031936
      3 06-Jan-15 0.377347 -1.698323 -0.268163 23.09 -2.778106
                                                                  0.031936
      4 07-Jan-15 0.306515 -0.611127 1.035972
                                                   4.13 -0.033870 0.031936
          Rm-Rf % MSCI World (Rgm %) Rgm-Rf % Info Edge Returns %
      0 0.317165
                                 0.00 -0.031936
                                                            0.506121
      1 0.873521
                                -0.29 -0.321936
                                                            0.052699
      2 -0.135682
                                -1.96 -1.991936
                                                           -1.234857
      3 -2.810042
                                -1.01 -1.041936
                                                           -1.114008
      4 -0.065806
                                 0.47 0.438064
                                                            1.000719
```

#### 1.0.1 Single Factor CAPM

```
[127]: x1 = df['Rm-Rf %']
y1 = df['Info Edge Returns %']

[128]: x1 = sm.add_constant(x1)
```

```
[129]: model1 = sm.OLS(endog = y1, exog =x1).fit()
model1.summary()
```

# [129]: <class 'statsmodels.iolib.summary.Summary'>

#### OLS Regression Results

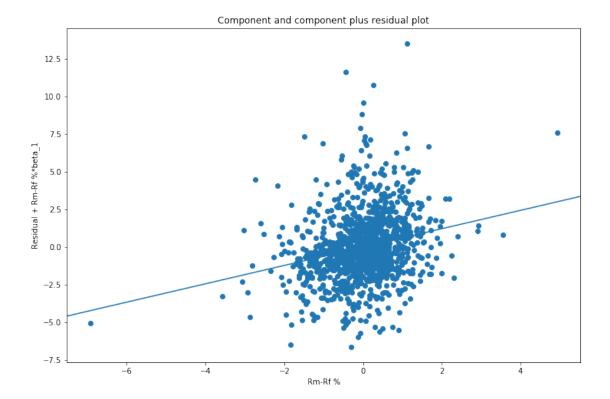
Dep. Variable:	Info Edge Returns %	R-squared:	0.054
Model:	OLS	Adj. R-squared:	0.053
Method:	Least Squares	F-statistic:	70.03
Date:	Sun, 12 Jan 2020	Prob (F-statistic):	1.57e-16
Time:	16:56:01	Log-Likelihood:	-2711.7
No. Observations:	1234	AIC:	5427.
Df Residuals:	1232	BIC:	5438.

Df Model: 1
Covariance Type: nonrobust

=========	=======			========	========	========
	coef	std err	t	P> t	[0.025	0.975]
const Rm-Rf %	0.1070 0.6091	0.062 0.073	1.724 8.368	0.085 0.000	-0.015 0.466	0.229 0.752
Omnibus: Prob(Omnibus) Skew: Kurtosis:	:	219.64 0.00 0.89 6.10	00 Jarqu 02 Prob(	•		1.993 659.992 4.84e-144 1.17

#### Warnings:

11 11 11



#### 1.0.2 Multifactor CAPM

```
[131]: x2 = df[['Rm-Rf %','Rgm-Rf %','SMB %','HML %', 'WML %','VIX %']]
    y2 = df['Info Edge Returns %']

[132]: x2 = sm.add_constant(x2)

[133]: model2 = sm.OLS(endog = y2, exog = x2).fit()
    model2.summary()
```

[133]: <class 'statsmodels.iolib.summary.Summary'>

#### OLS Regression Results

Dep. Variable:	Info Edge Returns %	R-squared:	0.059		
Model:	OLS	Adj. R-squared:	0.054		
Method:	Least Squares	F-statistic:	12.76		
Date:	Sun, 12 Jan 2020	Prob (F-statistic):	5.13e-14		
Time:	16:56:05	Log-Likelihood:	-2708.5		
No. Observations:	1234	AIC:	5431.		
Df Residuals:	1227	BIC:	5467.		
Df Model:	6				

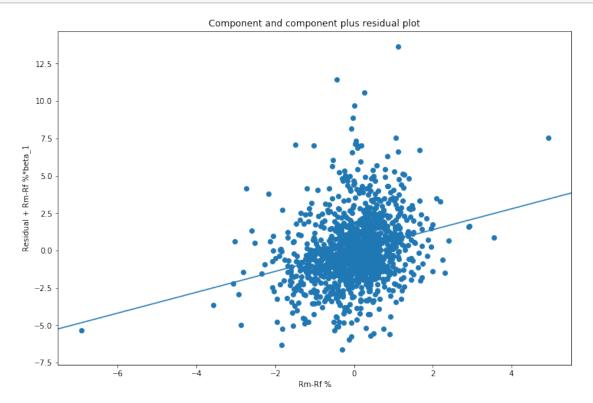
=========			
Covariance 1	l'ype:	nonrobust	

========	========	========	=======			========
	coef	std err	t	P> t	[0.025	0.975]
const Rm-Rf %	0.1081	0.063 0.105	1.725 6.641	0.085	-0.015 0.492	0.231
Rgm-Rf %	-0.0819	0.093	-0.878	0.380	-0.265	0.101
SMB %	0.0804	0.097	0.826	0.409	-0.111	0.271
HML %	-0.1540	0.076	-2.027	0.043	-0.303	-0.005
WML %	-0.0838	0.084	-0.994	0.321	-0.249	0.082
VIX %	-0.0074	0.016	-0.480	0.632	-0.038	0.023
========	=======					
Omnibus:		221	.835 Durb	oin-Watson:		1.990
Prob(Omnibu	s):	0	.000 Jarq	ue-Bera (JB	):	672.584
Skew:		0	.898 Prob	(JB):		8.92e-147
Kurtosis:		6	.140 Cond	l. No.		9.70
========	=======		=======			

## Warnings:

[1] Standard Errors assume that the covariance matrix of the errors is correctly specified.

11 11 11



[]: