

Tactical Analysis

September 15, 2021

1 Import Libraries and Prepare Returns Data

```
[1]: import pandas as pd
import numpy as np
import statsmodels.api as sm
import yfinance as yf
from sklearn import linear_model
import warnings
warnings.filterwarnings("ignore")

[2]: #Load Tactical Index Returns
tactical_index = pd.read_excel('TacticalIndexReturns_20210212.xlsx', skiprows = 1, index_col = 'Date')
tactical_index = tactical_index.iloc[:,[0,1]].replace('*', np.nan) #Replace values null values with nan
tactical_index.index = tactical_index.index
#read in SP500 and FF3 factors data
sp = yf.download('^SP500TR')
sp.columns = [f"SP500 {x}" for x in sp.columns]

ff = pd.read_csv('F-F_Research_Data_Factors.csv', index_col = 'Date', skiprows = 3, parse_dates = True, skipfooter = 99)
ff = (ff / 100) #Divide by 100 to convert from % to decimal
ff_index = ff.index
#read in tactical data (RORO & TCUS)
atacx = yf.download('ATACX')
atacx.columns = [f"ATACX {x}" for x in atacx.columns]
first_atacx = atacx.index[0]

#load and format TCUS Returns
tcus = pd.read_excel('TCUS Returns.xlsx', index_col = 'Month')
tcus.columns = ['TCUS']
tcus.index = tcus.index

#Join DataFrames together on selected data
df = pd.concat([tactical_index, ff, sp, atacx, tcus], axis = 1).fillna(method = 'ffill') #.fillna() to smooth end of month
```

```
df = df.loc[ff_index,['All','Filtered','ATACX Close',
    ↳ 'TCUS','Mkt-RF','SMB','HML','RF','SP500 Close']] # ff_index is the longest
    ↳ index of end of month returns
df['ATACX Returns'] = df['ATACX Close'].pct_change()
df['SP500 Returns'] = df['SP500 Close'].pct_change()
df = df.drop(['ATACX Close', 'SP500 Close'], axis = 1)
df = df.iloc[1:,:]
df.tail(10)
```

[*****100%*****] 1 of 1 completed

[*****100%*****] 1 of 1 completed

```
[2]:
```

	All	Filtered	TCUS	Mkt-RF	SMB	HML	RF	\
Date								
2020-08-31	0.045176	0.038666	0.057602	0.0763	-0.0025	-0.0294	0.0001	
2020-09-30	-0.025827	-0.021815	-0.029419	-0.0363	0.0006	-0.0251	0.0001	
2020-10-31	-0.025827	-0.021815	-0.036605	-0.0210	0.0444	0.0403	0.0001	
2020-11-30	-0.025827	-0.021815	0.178086	0.1247	0.0548	0.0211	0.0001	
2020-12-31	-0.025827	-0.021815	0.097485	0.0463	0.0481	-0.0136	0.0001	
2021-01-31	-0.025827	-0.021815	0.041290	-0.0003	0.0719	0.0285	0.0000	
2021-02-28	-0.025827	-0.021815	0.083357	0.0278	0.0211	0.0708	0.0000	
2021-03-31	-0.025827	-0.021815	0.033834	0.0308	-0.0248	0.0740	0.0000	
2021-04-30	-0.025827	-0.021815	0.039474	0.0493	-0.0309	-0.0074	0.0000	
2021-05-31	-0.025827	-0.021815	0.007127	0.0029	-0.0022	0.0705	0.0000	

	ATACX Returns	SP500 Returns
Date		
2020-08-31	-0.002515	0.071880
2020-09-30	-0.059664	-0.037997
2020-10-31	-0.004915	-0.026593
2020-11-30	0.048271	0.109464
2020-12-31	-0.017134	0.038449
2021-01-31	0.046197	-0.010096
2021-02-28	-0.020412	0.027575
2021-03-31	-0.041676	0.043796
2021-04-30	0.018416	0.053369
2021-05-31	-0.009368	0.006984

Create dummy variables for timing analysis – take value of returns when

```
[3]: df['SP500 Gamma * Dummy'] = np.where(df['SP500 Returns'] > 0, df['SP500
    ↳ Returns'], 0)
df['Mkt-RF Gamma * Dummy'] = np.where(df['Mkt-RF'] > 0, df['Mkt-RF'], 0)
```

2 Regression Order and Intuition

Funds under consideration: * Total tactical index ('All') - an index compiled by Good Harbor Financial of the returns of all funds that are considered 'tactical' by Morningstar. * Filtered tactical index ('Filtered') - The aforementioned index, filtered by β to the market. * ATACX - A mutual fund that rotates around small-caps, large-caps, or emerging markets (risk-on), and Treasuries (risk-off) based on Utilities and Treasuries as risk triggers. (ATAC Funds Website) * TCUS - seeks to "outperform the S&P 500 Index by aligning capital with the US equity market during sustained rallies and positioning defensively in weak equity market conditions." (TCUS Fund Overview)

Order is as follows:

1. Y: Total tactical index X: SP500 Index and SP500 Timing Var
2. Y: Total tactical index X: Mkt-RF Index and Mkt-RF Timing Var
3. Y: Total tactical index X: Mkt-RF Index and Mkt-RF Timing Var, FF3 SMB, HML
4. Y: Filtered tactical index X: SP500 Index and SP500 Timing Var
5. Y: Filtered tactical index X: Mkt-RF Index and Mkt-RF Timing Var
6. Y: Filtered tactical index X: Mkt-RF Index and Mkt-RF Timing Var, FF3 SMB, HML
7. Y: ATACX X: SP500 Index and SP500 Timing Var
8. Y: ATACX X: Mkt-RF Index and Mkt-RF Timing Var
9. Y: ATACX X: Mkt-RF Index and Mkt-RF Timing Var, FF3 SMB, HML
10. Y: TCUS X: SP500 Index and SP500 Timing Var
11. Y: TCUS X: Mkt-RF Index and Mkt-RF Timing Var
12. Y: TCUS X: Mkt-RF Index and Mkt-RF Timing Var, FF3 SMB, HML

Using the methodology proposed by Henriksson and Merton (1981), the regressions below are of the format:

$$R_{fund} = \alpha + \beta(r_{mkt} - rf) + \gamma((r_{mkt} - rf) * D(r_{mkt} > 0))$$

Where: $D = 1, r_{mkt} > 0$ $D = 0, r_{mkt} < 0$

With this multiple regression analysis: α suggests stock-picking ability β refers to market exposure (too high makes you a closet indexer) γ suggests market timing ability

3 Regression Results and Discussion

```
[4]: #1
tactical_index_data = df.loc[tactical_index_index[1:]] #Selecting on this index
↳ prevents unnecessary .ffill()
Y = tactical_index_data['All']
X = tactical_index_data.loc[:,['SP500 Returns','SP500 Gamma * Dummy']]
X = sm.add_constant(X)

model = sm.OLS(Y,X)
results = model.fit()
results.summary()
```

```
[4]: <class 'statsmodels.iolib.summary.Summary'>
"""
```

```

                                OLS Regression Results
=====
Dep. Variable:                  All      R-squared:                  0.757
Model:                            OLS     Adj. R-squared:              0.755
Method:                 Least Squares   F-statistic:                 382.8
Date:                Wed, 15 Sep 2021   Prob (F-statistic):          2.97e-76
Time:                        20:41:07   Log-Likelihood:              663.18
No. Observations:                249     AIC:                        -1320.
Df Residuals:                    246     BIC:                        -1310.
Df Model:                          2
Covariance Type:                nonrobust
=====
=====
                                coef      std err          t      P>|t|      [0.025
0.975]
-----
const                0.0032      0.002       1.865     0.063     -0.000
0.007
SP500 Returns        0.7035      0.044     15.966     0.000     0.617
0.790
SP500 Gamma * Dummy -0.0382      0.077     -0.496     0.620     -0.190
0.114
=====
Omnibus:                 262.508   Durbin-Watson:              2.276
Prob(Omnibus):            0.000   Jarque-Bera (JB):           15923.454
Skew:                     4.042   Prob(JB):                    0.00
Kurtosis:                 41.333   Cond. No.                    79.9
=====
```

Warnings:

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

```
specified.
"""
```

```
[5]: #2
Y = tactical_index_data['All']
X = tactical_index_data.loc[:,['Mkt-RF','Mkt-RF Gamma * Dummy']]
X = sm.add_constant(X)

model = sm.OLS(Y,X)
results = model.fit()
results.summary()
```

```
[5]: <class 'statsmodels.iolib.summary.Summary'>
"""
```

```

                                OLS Regression Results
=====
Dep. Variable:                  All    R-squared:                0.818
Model:                            OLS    Adj. R-squared:            0.817
Method:                 Least Squares    F-statistic:                553.2
Date:                 Wed, 15 Sep 2021    Prob (F-statistic):          9.06e-92
Time:                   20:41:07    Log-Likelihood:             699.34
No. Observations:                  249    AIC:                       -1393.
Df Residuals:                      246    BIC:                       -1382.
Df Model:                           2
Covariance Type:                nonrobust
=====
=====
                                coef    std err          t      P>|t|      [0.025
0.975]
-----
const                0.0027      0.001      1.814      0.071      -0.000
0.006
Mkt-RF                0.6730      0.036     18.529      0.000       0.601
0.744
Mkt-RF Gamma * Dummy  0.0266      0.065      0.406      0.685      -0.102
0.155
=====
Omnibus:                 242.710    Durbin-Watson:              2.122
Prob(Omnibus):            0.000    Jarque-Bera (JB):           10531.422
Skew:                     3.675    Prob(JB):                    0.00
Kurtosis:                 34.001    Cond. No.                    77.9
=====
```

Warnings:

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

```
"""
```

```
[6]: #3
Y = tactical_index_data['All']
X = tactical_index_data.loc[:,['Mkt-RF', 'Mkt-RF Gamma * Dummy', 'SMB', 'HML']]
X = sm.add_constant(X)

model = sm.OLS(Y,X, missing = 'drop')
results = model.fit()
results.summary()
```

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

```
"""
```

```

                        OLS Regression Results
=====
Dep. Variable:          All      R-squared:                0.863
Model:                  OLS      Adj. R-squared:         0.861
Method:                 Least Squares      F-statistic:          384.1
Date:                  Wed, 15 Sep 2021    Prob (F-statistic):    5.29e-104
Time:                  20:41:08           Log-Likelihood:        734.59
No. Observations:      249           AIC:                  -1459.
Df Residuals:          244           BIC:                  -1442.
Df Model:               4
Covariance Type:       nonrobust
=====
=====
                        coef      std err          t      P>|t|      [0.025
0.975]
-----
const                0.0020      0.001       1.520     0.130     -0.001
0.005
Mkt-RF                0.6154      0.032     19.050     0.000      0.552
0.679
Mkt-RF Gamma * Dummy  0.0521      0.057      0.913     0.362     -0.060
0.165
SMB                  0.2359      0.027      8.770     0.000      0.183
0.289
HML                  0.0890      0.025      3.494     0.001      0.039
0.139
=====
Omnibus:              133.356    Durbin-Watson:         2.160
Prob(Omnibus):         0.000    Jarque-Bera (JB):      2076.225
Skew:                  1.719    Prob(JB):              0.00
Kurtosis:              16.722    Cond. No.              78.3
=====
```

Warnings:

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

3.1 Tactical Index Discussion

Regression analysis of the entire tactical index highlights some key points:

- * The index as a whole exhibits high beta to both the S&P500 and the total market (.6-.7). This indicates that a large portion of the index's returns are attributable to market β /closet indexing by individual managers.
- * These coefficients are statistically significant for each regression analysis.
- * When using both the S&P 500 and the total market index, the analysis does not suggest any significant market timing ability. None of the three γ coefficients are significant at any significance level.
- * This suggests that market timing by managers in the total tactical index does not improve their returns.
- * Much of the tactical index returns appear to be explained by exposure to size and value factors, rather than market timing.

```
[7]: #4
Y = tactical_index_data['Filtered']
X = tactical_index_data.loc[:,['SP500 Returns','SP500 Gamma * Dummy']]
X = sm.add_constant(X)

model = sm.OLS(Y,X)
results = model.fit()
results.summary()
```

```
[7]: <class 'statsmodels.iolib.summary.Summary'>
"""
```

```

                        OLS Regression Results
=====
Dep. Variable:          Filtered      R-squared:                0.552
Model:                  OLS          Adj. R-squared:           0.548
Method:                 Least Squares  F-statistic:              151.5
Date:                  Wed, 15 Sep 2021  Prob (F-statistic):       1.31e-43
Time:                  20:41:08        Log-Likelihood:           614.65
No. Observations:      249            AIC:                     -1223.
Df Residuals:          246            BIC:                     -1213.
Df Model:              2
Covariance Type:       nonrobust
=====
=====
                        coef      std err          t      P>|t|      [0.025
0.975]
```

```

-----
-----
const                0.0037      0.002      1.791      0.074      -0.000
0.008
SP500 Returns        0.5348      0.054      9.989      0.000      0.429
0.640
SP500 Gamma * Dummy -0.0228      0.094      -0.243      0.808      -0.207
0.162
=====
Omnibus:                291.465      Durbin-Watson:                2.184
Prob(Omnibus):          0.000      Jarque-Bera (JB):            25503.500
Skew:                   4.687      Prob(JB):                    0.00
Kurtosis:               51.686      Cond. No.                    79.9
=====

```

Warnings:

```

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

```

```

[8]: #5
Y = tactical_index_data['Filtered']
X = tactical_index_data.loc[:,['Mkt-RF', 'Mkt-RF Gamma * Dummy']]
X = sm.add_constant(X)

model = sm.OLS(Y,X)
results = model.fit()
results.summary()

```

```

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

```

```

                        OLS Regression Results
=====
Dep. Variable:          Filtered      R-squared:                0.630
Model:                  OLS           Adj. R-squared:           0.627
Method:                 Least Squares  F-statistic:             209.8
Date:                  Wed, 15 Sep 2021  Prob (F-statistic):      6.86e-54
Time:                  20:41:08        Log-Likelihood:          638.61
No. Observations:      249            AIC:                    -1271.
Df Residuals:          246            BIC:                    -1261.
Df Model:               2
Covariance Type:       nonrobust
=====
=====
                        coef      std err          t      P>|t|      [0.025
0.975]
-----
-----

```



```

-----
const                0.0032      0.002      1.680      0.094      -0.001
0.007
Mkt-RF              0.5243      0.046     11.312      0.000      0.433
0.616
Mkt-RF Gamma * Dummy 0.0307      0.083      0.367      0.714     -0.134
0.195
=====
Omnibus:                285.097   Durbin-Watson:                2.100
Prob(Omnibus):           0.000   Jarque-Bera (JB):            22096.669
Skew:                    4.559   Prob(JB):                     0.00
Kurtosis:                48.240   Cond. No.                     77.9
=====

```

Warnings:

```

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

```

```

[9]: #6
Y = tactical_index_data['Filtered']
X = tactical_index_data.loc[:,['Mkt-RF', 'Mkt-RF Gamma * Dummy', 'SMB', 'HML']]
X = sm.add_constant(X)

model = sm.OLS(Y,X)
results = model.fit()
results.summary()

```

```

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

```

```

                        OLS Regression Results
=====
Dep. Variable:          Filtered   R-squared:                0.713
Model:                  OLS       Adj. R-squared:           0.709
Method:                 Least Squares   F-statistic:             151.8
Date:                  Wed, 15 Sep 2021   Prob (F-statistic):      5.47e-65
Time:                  20:41:08   Log-Likelihood:          670.27
No. Observations:      249   AIC:                     -1331.
Df Residuals:          244   BIC:                     -1313.
Df Model:               4
Covariance Type:       nonrobust
=====
=====
                        coef      std err          t      P>|t|      [0.025
0.975]
-----
-----

```

const	0.0025	0.002	1.465	0.144	-0.001
0.006					
Mkt-RF	0.4563	0.042	10.909	0.000	0.374
0.539					
Mkt-RF Gamma * Dummy	0.0551	0.074	0.746	0.456	-0.090
0.201					
SMB	0.2916	0.035	8.371	0.000	0.223
0.360					
HML	0.0326	0.033	0.989	0.324	-0.032
0.098					

Omnibus:	168.896	Durbin-Watson:	2.113
Prob(Omnibus):	0.000	Jarque-Bera (JB):	3752.009
Skew:	2.271	Prob(JB):	0.00
Kurtosis:	21.466	Cond. No.	78.3

Warnings:

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

3.2 Filtered Tactical Index Discussion

Regression analysis again provides similar findings when filtering tactical managers based on their market β .

* Analysis of the filtered index yields similar results to the entire index. Analysis shows lower market β to the market. This to be expected as managers are specifically selected based on low market β

* Despite this lower market β , our analysis does not suggest any significant market timing abilities by these managers.

* Instead this subset of managers appears to be adding much of their value from exposure to the FF3 size factor.

```
[12]: #CAPM Model - for reference
Y = df.loc[first_atacx:,'ATACX Returns']
X = df.loc[first_atacx:,['SP500 Returns']]
pd.concat([Y, X], axis = 1)
X = sm.add_constant(X)
model = sm.OLS(Y,X, missing = 'drop')
results = model.fit()

results.summary()
```

```
[12]: <class 'statsmodels.iolib.summary.Summary'>
"""
```

```

                                OLS Regression Results
=====
Dep. Variable:          ATACX Returns    R-squared:                0.140
Model:                  OLS              Adj. R-squared:           0.131
Method:                 Least Squares    F-statistic:              16.57
Date:                  Wed, 15 Sep 2021  Prob (F-statistic):      9.28e-05
Time:                  20:46:36          Log-Likelihood:           178.34
No. Observations:      104              AIC:                     -352.7
Df Residuals:          102              BIC:                     -347.4
Df Model:               1
Covariance Type:       nonrobust
=====
=
                                coef      std err          t      P>|t|      [0.025
0.975]
-----
-
const                0.0011      0.005      0.238      0.812      -0.008
0.010
SP500 Returns        0.4595      0.113      4.070      0.000      0.236
0.683
=====
Omnibus:                1.229    Durbin-Watson:           2.023
Prob(Omnibus):          0.541    Jarque-Bera (JB):         1.246
Skew:                   0.252    Prob(JB):                 0.536
Kurtosis:               2.817    Cond. No.                  26.2
=====

```

Warnings:

```

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

```

```

[13]: #7
Y = df.loc[first_atacx:,'ATACX Returns']
X = df.loc[first_atacx:,['SP500 Returns', 'SP500 Gamma * Dummy']]
pd.concat([Y, X], axis = 1)
X = sm.add_constant(X)
model = sm.OLS(Y,X, missing = 'drop')
results = model.fit()
results.summary()

```

```

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

```

```

                                OLS Regression Results
=====
Dep. Variable:          ATACX Returns    R-squared:                0.186

```

```

Model:                                OLS    Adj. R-squared:                0.169
Method:                             Least Squares    F-statistic:                11.51
Date:                               Wed, 15 Sep 2021    Prob (F-statistic):        3.14e-05
Time:                               20:46:37    Log-Likelihood:            181.19
No. Observations:                    104    AIC:                        -356.4
Df Residuals:                        101    BIC:                        -348.4
Df Model:                            2
Covariance Type:                     nonrobust

```

```

=====
=====
                                coef    std err          t      P>|t|      [0.025
0.975]
-----
const                -0.0105      0.007      -1.597      0.113      -0.024
0.003
SP500 Returns         0.0196      0.215       0.091      0.928      -0.407
0.446
SP500 Gamma * Dummy   0.7896      0.331       2.385      0.019       0.133
1.446
=====
Omnibus:              0.146    Durbin-Watson:              2.159
Prob(Omnibus):         0.930    Jarque-Bera (JB):           0.260
Skew:                  -0.081    Prob(JB):                   0.878
Kurtosis:              2.817    Cond. No.                   90.9
=====

```

Warnings:

```

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

```

```

[14]: #8
Y = df.loc[first_atacx:, 'ATACX Returns']
X = df.loc[first_atacx:, ['Mkt-RF', 'Mkt-RF Gamma * Dummy']]
X = sm.add_constant(X)
model = sm.OLS(Y,X, missing = 'drop')
results = model.fit()
results.summary()

```

```

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

```

```

                                OLS Regression Results
=====
Dep. Variable:            ATACX Returns    R-squared:                0.196
Model:                    OLS             Adj. R-squared:            0.180
Method:                    Least Squares   F-statistic:              12.33

```

```

Date:                Wed, 15 Sep 2021    Prob (F-statistic):        1.62e-05
Time:                20:46:38           Log-Likelihood:           181.87
No. Observations:    104                AIC:                     -357.7
Df Residuals:        101                BIC:                     -349.8
Df Model:            2
Covariance Type:     nonrobust

```

```

=====
=====
              coef      std err          t      P>|t|      [0.025
0.975]
-----
-----

```

```

const          -0.0095      0.006      -1.494      0.138      -0.022
0.003
Mkt-RF          0.0654      0.199       0.328      0.743      -0.330
0.461
Mkt-RF Gamma * Dummy  0.7065      0.305       2.315      0.023      0.101
1.312

```

```

=====
Omnibus:                0.155    Durbin-Watson:                2.181
Prob(Omnibus):           0.925    Jarque-Bera (JB):           0.317
Skew:                    -0.062    Prob(JB):                   0.854
Kurtosis:                2.759    Cond. No.                   84.3
=====

```

Warnings:

```

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

```

```

[15]: #9
Y = df.loc[first_atacx:,'ATACX Returns']
X = df.loc[first_atacx:,['Mkt-RF','Mkt-RF Gamma * Dummy', 'SMB','HML']]
X = sm.add_constant(X)

model = sm.OLS(Y,X, missing = 'drop')
results = model.fit()
results.summary()

```

```

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

```

```

              OLS Regression Results
=====
Dep. Variable:    ATACX Returns    R-squared:                0.228
Model:            OLS              Adj. R-squared:            0.196
Method:            Least Squares    F-statistic:              7.295
Date:             Wed, 15 Sep 2021    Prob (F-statistic):       3.43e-05

```

```

Time:                20:46:43    Log-Likelihood:        183.95
No. Observations:    104        AIC:                -357.9
Df Residuals:        99        BIC:                -344.7
Df Model:            4
Covariance Type:     nonrobust

```

```

=====
=====
              coef      std err          t      P>|t|      [0.025
0.975]
-----
const          -0.0084      0.006     -1.327      0.188     -0.021
0.004
Mkt-RF          0.1122      0.210      0.533      0.595     -0.305
0.529
Mkt-RF Gamma * Dummy  0.5960      0.312      1.909      0.059     -0.024
1.216
SMB              0.2311      0.174      1.328      0.187     -0.114
0.576
HML             -0.2357      0.143     -1.651      0.102     -0.519
0.047
=====
Omnibus:                0.821    Durbin-Watson:        2.229
Prob(Omnibus):          0.663    Jarque-Bera (JB):      0.846
Skew:                   -0.021    Prob(JB):              0.655
Kurtosis:               2.560    Cond. No.              88.5
=====

```

Warnings:

```

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

```

3.3 ATACX Strategy Discussion

Results from the ATACX analysis are a bit perplexing. A simple CAPM model shows an SP500 β of .46. This model has a much lower R^2 (.14) value than any of the tactical index models shown prior. Adding the market timing factor provides little improvement to the overall model fit. Alternatively, it suggests that there is market-timing ability in the strategy and this coefficient is significant. At the same time, β_{mkt} drops by roughly .3 and the coefficient loses significance when using both the S&P 500 and the FF3 total market as the index. Results from this model should be taken with a high dose of skepticism because of the relatively poor fit.

```

[ ]: # CAPM Model
      Y = df['TCUS']
      X = df.loc[:, ['SP500 Returns']]
      X = sm.add_constant(X)

```

```
model = sm.OLS(Y,X, missing = 'drop')
results = model.fit()
results.summary()
```

```
[ ]: #10
Y = df['TCUS']
X = df.loc[:,['SP500 Returns', 'SP500 Gamma * Dummy']]
X = sm.add_constant(X)

model = sm.OLS(Y,X, missing = 'drop')
results = model.fit()
results.summary()
```

```
[ ]: #11
Y = df['TCUS']
X = df.loc[:,['Mkt-RF', 'Mkt-RF Gamma * Dummy']]
X = sm.add_constant(X)

model = sm.OLS(Y,X, missing = 'drop')
results = model.fit()
results.summary()
```

```
[ ]: #12
Y = df['TCUS']
X = df.loc[:,['Mkt-RF', 'Mkt-RF Gamma * Dummy', 'SMB', 'HML']]
X = sm.add_constant(X)

model = sm.OLS(Y,X, missing = 'drop')
results = model.fit()
results.summary()
```

3.4 TCUS Discussion

Regression analysis of Tactical Core US provides some interesting findings.

- * When using the S&P 500 as the benchmark, this analysis suggests positive significant market timing ability coupled with low β_{mkt} .

- * All three regressions yield significant coefficients to the market timing factor which suggests that the strategy holds market-timing capability.

- * These results suggest that TCUS exhibits some positive market timing ability. Additionally, it's returns can be attributed to exposure to the FF3 size and value factors as well.

4 Conclusion

In conclusion, our analysis yields a few interesting results:

1. "Tactical" managers as an asset-class can best be classified as low-beta closet indexers. This can be seen by the high R^2 value, but $\beta < 1$.

2. As an asset-class, tactical managers do not provide investors with any significant market-timing capabilities.
3. ATACX shows low β_{mkt} , but high market-timing capabilities.
4. TCUS shows both β_{mkt} and high market-timing capabilities as well
5. These findings may act as evidence that both TCUS and ATACX can attribute their returns to upside market capture.

Further research could focus on how this relationship has changed since quantitative easing has led to a relative decline in market volatility.

[]: