

# Carrie\_Little\_AAI\_500\_FinalProject\_FamaFrench

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## 1 Risk Parity - Opportunity dataset

Carrie Little

### 1.0.1 Import Necessary Libraries

```
[1]: # Carrie Little - AAI5000 Final Project Code
#
# Import All Necessary Libraries
import numpy as np
import pandas as pd
import matplotlib.pyplot as plt
import seaborn as sns
import scipy.stats as stats
import cvxpy as cp
```

### 1.0.2 Models

Fama-French Factor Model

```
[2]: # http://mba.tuck.dartmouth.edu/pages/faculty/ken.french/data_library.html
fama_french_factors = pd.read_csv('F-F_Research_Data_Factors.csv',
    ↪ parse_dates=['Date']) # Load Dataset as Dataframe
fama_french_factors.head()
    ↪ # Display 1st 5 in Dataframe
```

```
[2]:
```

	Date	Mkt-RF	SMB	HML	RF
0	2014-11-30	2.55	-2.06	-3.10	0.0
1	2014-12-31	-0.06	2.49	2.27	0.0
2	2015-01-31	-3.11	-0.56	-3.59	0.0
3	2015-02-28	6.13	0.63	-1.86	0.0
4	2015-03-31	-1.12	3.04	-0.38	0.0

```
[3]: df = pd.read_csv('Opportunity_Set.csv', parse_dates=['Date']) # Load
    ↪ Dataset as Dataframe
df.head()
```

	Date	Vanguard LifeStrategy Income Fund (VASIX)	\
0	2014-11-30	0.0094	
1	2014-12-31	-0.0005	
2	2015-01-31	0.0141	
3	2015-02-28	0.0033	
4	2015-03-31	0.0018	

	Vanguard Total World Stock ETF (VT) \
0	0.0126
1	-0.0199
2	-0.0163
3	0.0595
4	-0.0121

	PIMCO 25+ Year Zero Coupon US Trs ETF (ZROZ) \
0	0.0414
1	0.0612
2	0.1600
3	-0.1007
4	0.0117

	AQR Diversified Arbitrage I (ADAIX)	iShares Gold Trust (IAU)
0	-0.0066	-0.0053
1	-0.0117	0.0133
2	-0.0059	0.0865
3	0.0069	-0.0579
4	0.0000	-0.0222

	Bitcoin Market Price USD (~BTC) \
0	0.0969
1	-0.1777
2	-0.2677
3	0.1062
4	-0.0150

	AQR Risk-Balanced Commodities Strategy I (ARCIX) \
0	-0.0726
1	-0.0412
2	-0.0287
3	0.0044
4	-0.0573

	AQR Long-Short Equity I (QLEIX)	AQR Style Premia Alternative I (QSPIX)	\
0	0.0248	0.0412	
1	0.0140	0.0002	
2	0.0156	-0.0112	
3	0.0236	-0.0390	

4	-0.0027	0.0256
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	AQR Equity Market Neutral I (QMNIX)	AQR Macro Opportunities I (QGMIX) \
0	0.0257	0.0154
1	0.0195	0.0039
2	0.0290	-0.0070
3	-0.0078	0.0091
4	0.0049	0.0261

	AGF U.S. Market Neutral Anti-Beta (BTAL) \
0	0.0235
1	0.0294
2	0.0320
3	-0.0568
4	0.0000

	AQR Managed Futures Strategy HV I (QMHIX) \
0	0.1159
1	0.0461
2	0.0721
3	-0.0108
4	0.0655

	Invesco DB US Dollar Bullish (UUP)	ProShares VIX Mid-Term Futures (VIXM)
0	0.0165	-0.0298
1	0.0213	0.0553
2	0.0484	0.0762
3	0.0028	-0.1145
4	0.0278	0.0033

```
[4]: import pandas as pd
import statsmodels.api as sm

asset_return = df['Vanguard LifeStrategy Income Fund (VASIX)'] # Replace with
↳ asset name

# Step 3: Prepare the factors and align with asset returns by date
factor_returns = fama_french_factors[['Mkt-RF', 'SMB', 'HML']] # Ensure both
↳ asset returns and factors are aligned by date
factor_returns_RF = fama_french_factors['RF'] # Risk-free rate
excess_asset_returns = asset_return - factor_returns_RF # Calculate excess
↳ returns over risk-free rate

# Add a constant (alpha) to the model
X = sm.add_constant(factor_returns[['Mkt-RF', 'SMB', 'HML']])
y = excess_asset_returns
```

```
# Fit the model
model = sm.OLS(y, X).fit()

# Display the summary of the regression
print(model.summary())
```

#### OLS Regression Results

```
=====
Dep. Variable:          y      R-squared:          0.025
Model:                  OLS    Adj. R-squared:      -0.001
Method:                 Least Squares  F-statistic:    0.9667
Date:                   Thu, 03 Oct 2024  Prob (F-statistic): 0.411
Time:                   05:43:47  Log-Likelihood:  57.434
No. Observations:      118      AIC:           -106.9
Df Residuals:          114      BIC:           -95.78
Df Model:               3
Covariance Type:       nonrobust
=====
```

	coef	std err	t	P> t	[0.025	0.975]
const	-0.1267	0.014	-8.839	0.000	-0.155	-0.098
Mkt-RF	0.0016	0.003	0.518	0.605	-0.005	0.008
SMB	0.0066	0.005	1.260	0.210	-0.004	0.017
HML	0.0020	0.004	0.532	0.596	-0.005	0.009

```
=====
Omnibus:                18.265  Durbin-Watson:          0.069
Prob(Omnibus):           0.000  Jarque-Bera (JB):        22.950
Skew:                    -1.080  Prob(JB):                1.04e-05
Kurtosis:                 2.930  Cond. No.                 4.91
=====
```

Notes:

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

```
[5]: # Merge the two datasets on the 'Date' column
merged_df = pd.merge(df, fama_french_factors, on='Date')

# Set the index to 'Date' for easier handling
merged_df.set_index('Date', inplace=True)

# Prepare the assets for regression (list of asset columns)
asset_columns = [
    'Vanguard LifeStrategy Income Fund (VASIX)',
    'Vanguard Total World Stock ETF (VT)',
    'PIMCO 25+ Year Zero Coupon US Trs ETF (ZROZ)',
    'AQR Diversified Arbitrage I (ADAIX)',
```

```

    'iShares Gold Trust (IAU)',
    'Bitcoin Market Price USD (^BTC)'
]

# We need to calculate excess returns for the assets (subtracting the risk-free
↪rate RF)
for asset in asset_columns:
    merged_df[asset] = merged_df[asset] - (merged_df['RF'] / 100) # Converting
↪RF to percentage

# Prepare the factor data for the regression (convert from percentage to
↪decimal)
factor_columns = ['Mkt-RF', 'SMB', 'HML']
merged_df[factor_columns] = merged_df[factor_columns] / 100

# Perform Fama-French 3-factor regression for each asset
import statsmodels.api as sm

betas = pd.DataFrame()

for asset in asset_columns:
    X = sm.add_constant(merged_df[factor_columns]) # Add constant (alpha)
    y = merged_df[asset] # Asset's excess returns
    model = sm.OLS(y, X).fit() # Fit OLS regression

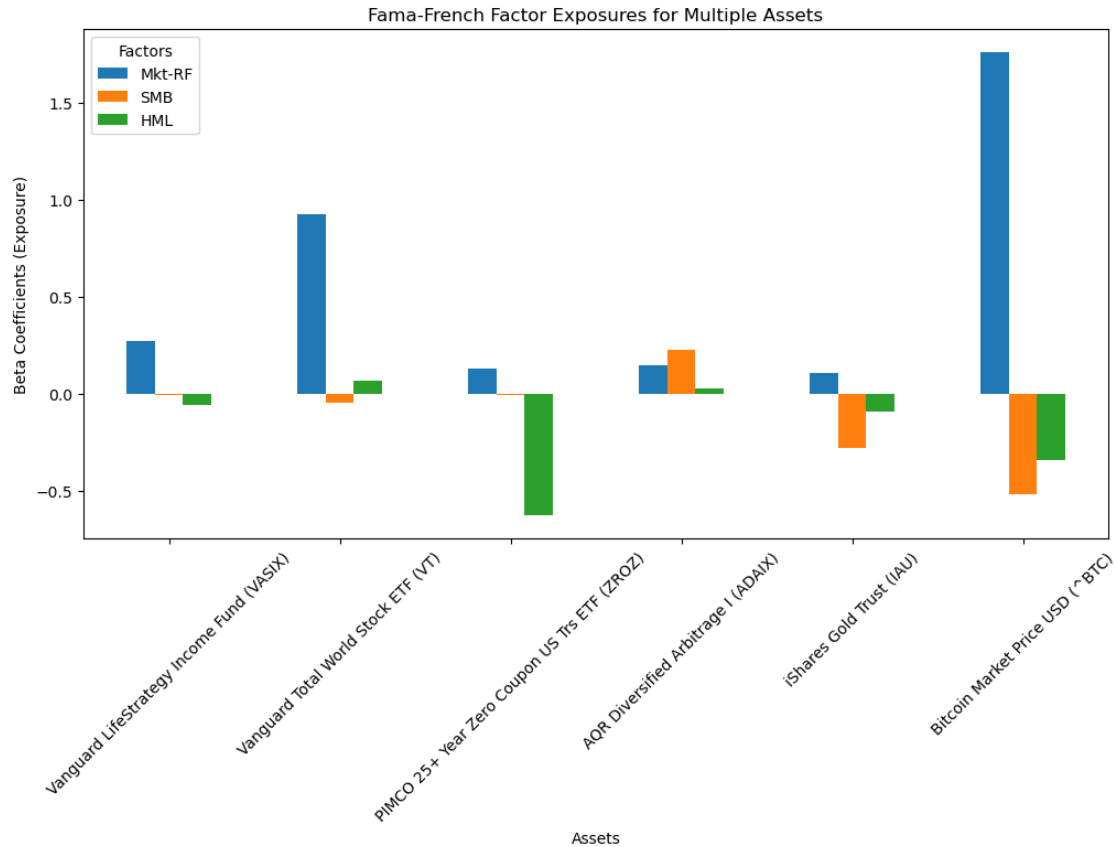
    # Collect the coefficients (betas) for each asset
    betas[asset] = model.params[1:] # Ignore the intercept (alpha), take only
↪factor betas

# Rename the index for factors
betas.index = factor_columns
betas

# Now we can visualize the beta exposures for all assets
betas.T.plot(kind='bar', figsize=(12, 6))
plt.title('Fama-French Factor Exposures for Multiple Assets')
plt.xlabel('Assets')
plt.ylabel('Beta Coefficients (Exposure)')
plt.xticks(rotation=45)
plt.legend(title='Factors')

# Display the plot
plt.show()

```



[6]: *# Let's extend the analysis to include all assets in the opportunity dataset.*

```
# Extract all asset columns (ignoring the factor columns)
all_asset_columns = merged_df.columns.difference(factor_columns + ['RF']).
↳ tolist()

# Initialize an empty DataFrame to store the betas for all assets
all_betas = pd.DataFrame()

# Perform Fama-French 3-factor regression for each asset
for asset in all_asset_columns:
    X = sm.add_constant(merged_df[factor_columns]) # Add constant (alpha)
    y = merged_df[asset] # Asset's excess returns
    model = sm.OLS(y, X).fit() # Fit OLS regression

    # Collect the coefficients (betas) for each asset
    all_betas[asset] = model.params[1:] # Ignore the intercept (alpha), take
↳ only factor betas

# Rename the index for factors
```

```
all_betas.index = factor_columns
all_betas
```

```
[6]:      AGF U.S. Market Neutral Anti-Beta (BTAL) \
Mkt-RF      -0.496720
SMB          -0.610488
HML          -0.195573

      AQR Diversified Arbitrage I (ADAIX) \
Mkt-RF      0.143773
SMB         0.226056
HML         0.028488

      AQR Equity Market Neutral I (QMNIX)  AQR Long-Short Equity I (QLEIX) \
Mkt-RF      -0.079728                    0.342733
SMB         -0.325849                    -0.377814
HML         0.390752                     0.442353

      AQR Macro Opportunities I (QGMIX) \
Mkt-RF      -0.043512
SMB         0.074967
HML         0.078516

      AQR Managed Futures Strategy HV I (QMHIX) \
Mkt-RF      -0.293050
SMB         -0.161875
HML         -0.028837

      AQR Risk-Balanced Commodities Strategy I (ARCIX) \
Mkt-RF      0.436298
SMB         -0.198439
HML         0.385313

      AQR Style Premia Alternative I (QSPIX) \
Mkt-RF      0.028831
SMB         -0.406157
HML         0.536513

      Bitcoin Market Price USD (^BTC)  Invesco DB US Dollar Bullish (UUP) \
Mkt-RF      1.756685                    -0.172204
SMB         -0.516295                    0.070058
HML         -0.343898                    -0.023386

      PIMCO 25+ Year Zero Coupon US Trs ETF (ZROZ) \
Mkt-RF      0.127749
SMB         -0.009423
```

HML	-0.628553
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	ProShares VIX Mid-Term Futures (VIXM) \
Mkt-RF	-1.375098
SMB	0.084818
HML	-0.419645

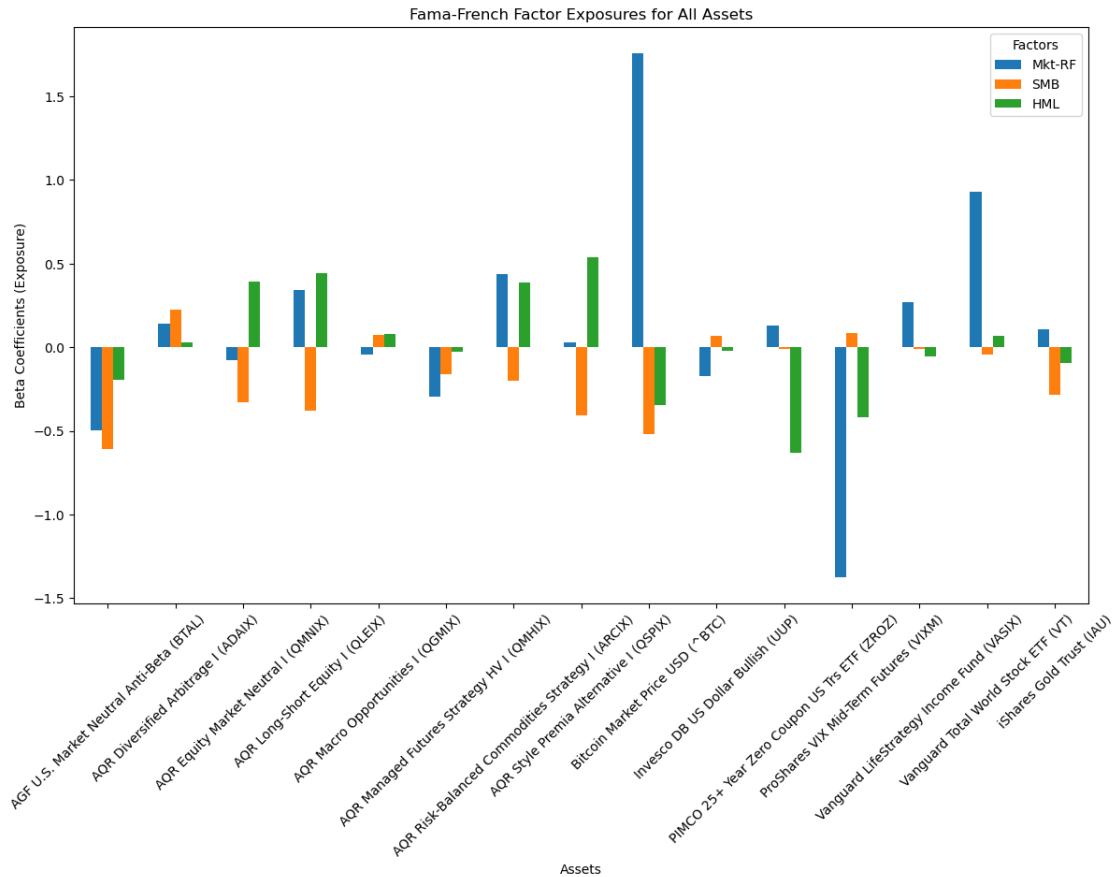
	Vanguard LifeStrategy Income Fund (VASIX) \
Mkt-RF	0.270746
SMB	-0.009690
HML	-0.055494

	Vanguard Total World Stock ETF (VT)	iShares Gold Trust (IAU)
Mkt-RF	0.926904	0.105052
SMB	-0.045200	-0.281462
HML	0.069572	-0.094277

```
[7]: # Visualize the beta exposures for all assets
all_betas.T.plot(kind='bar', figsize=(14, 8))
plt.title('Fama-French Factor Exposures for All Assets')
plt.xlabel('Assets')
plt.ylabel('Beta Coefficients (Exposure)')
plt.xticks(rotation=45)
plt.legend(title='Factors')

# Display the plot
plt.show()
```





Betas ( 1, 2, 3 ) : These represent the asset's sensitivity to each factor.

A high 1 (market) means the asset is strongly correlated with market movements. A positive 2 (SMB) means the asset has small-cap exposure. A positive 3 (HML) means the asset has exposure to value stocks.

[ ]:

[ ]:

## References

Agresti, Alan, and Maria Kateri. Foundations of Statistics for Data Scientists: With R and Python. CRC Press, Taylor & Francis Group, 2022.

Agresti, Alan, and Maria Kateri. (2022) Appendix B2. Chapter 2: Python for Probability Distributions. In Foundations of Statistics for Data Scientists: With R and Python (p. 385-389). CRC Press, Taylor & Francis Group, 2022.

ChatGPT, (2024) GPT-4o version, OpenAI. [Large language model]. <https://chatgpt.com/>

Opportunity Dataset - need link/website info

Fama French Factors , Kenneth French's website. [http://mba.tuck.dartmouth.edu/pages/faculty/ken.french/data\\_](http://mba.tuck.dartmouth.edu/pages/faculty/ken.french/data_)

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