

# Fama-French Factor Model

## Factor Definitions

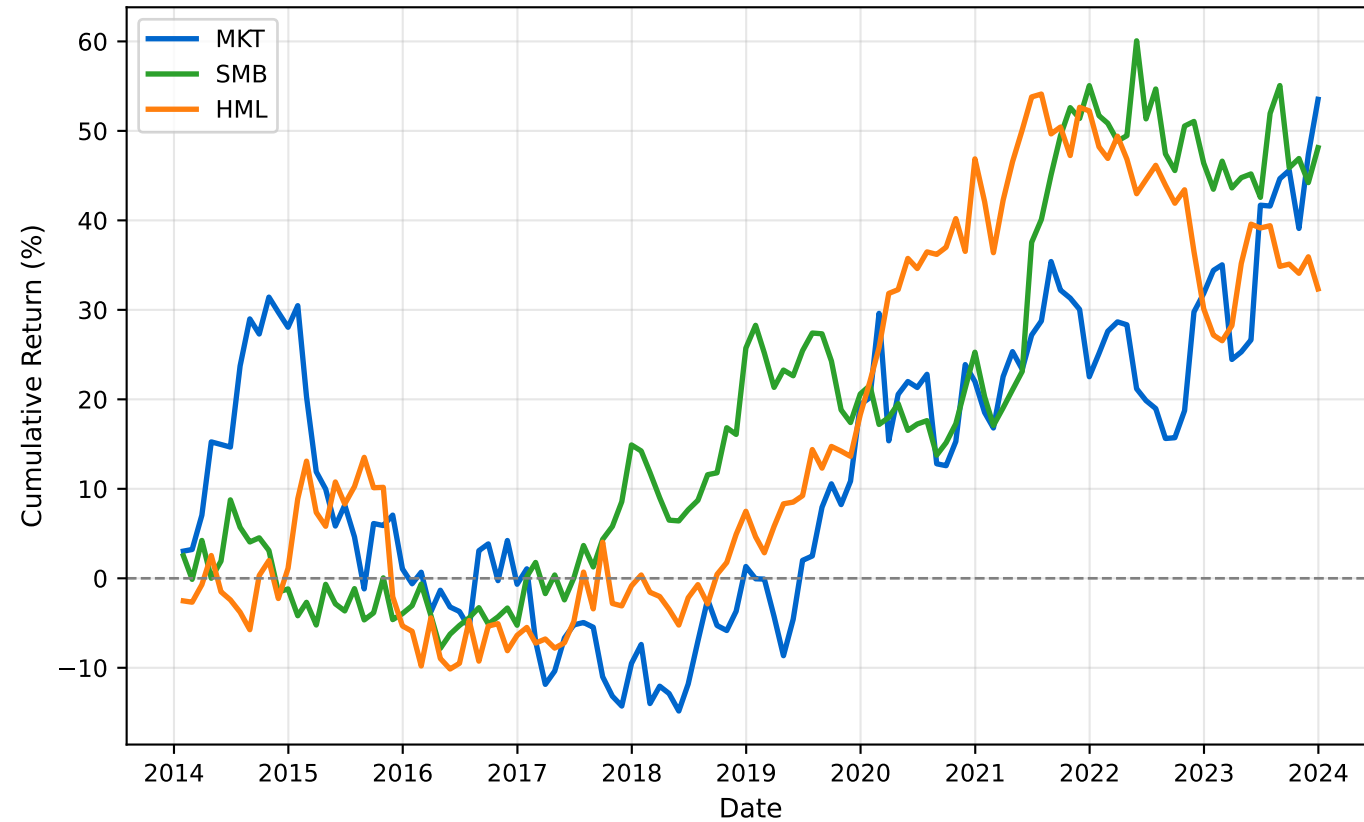
### FAMA-FRENCH THREE-FACTOR MODEL (1993)

1. MKT (Market)
  - Market return minus risk-free rate
  - $R_m - R_f$
  - Captures broad market exposure
2. SMB (Small Minus Big)
  - Small cap return minus large cap return
  - Size effect: small stocks outperform
  - Based on market capitalization
3. HML (High Minus Low)
  - Value stocks minus growth stocks
  - Value effect: high B/M outperform
  - Based on Book-to-Market ratio

### FAMA-FRENCH FIVE-FACTOR MODEL (2015)

4. RMW (Robust Minus Weak)
  - Profitable firms minus unprofitable
  - Profitability effect
5. CMA (Conservative Minus Aggressive)
  - Low investment minus high investment
  - Investment effect

## Cumulative Factor Returns (10 Years)



## Factor Statistics

### FACTOR STATISTICS (Monthly)

Factor	Mean	Std	Sharpe (Ann.)
MKT	0.44%	4.15%	0.37
SMB	0.37%	3.02%	0.43
HML	0.28%	3.17%	0.31

### CORRELATION MATRIX

	MKT	SMB	HML
MKT	1.00	0.10	-0.11
SMB	0.10	1.00	0.13
HML	-0.11	0.13	1.00

Key: Low correlations = good diversification

## Python Implementation

Loading Fama-French Data with pandas

```
import pandas as pd
import numpy as np
from sklearn.linear_model import LinearRegression

# Load Fama-French factors (from Ken French website)
# Format: Date, Mkt-RF, SMB, HML, RF (in %)
ff_data = pd.read_csv('F-F_Research_Data_Factors.csv',
                      skiprows=3)
ff_data['Date'] = pd.to_datetime(ff_data['Date'],
                                format='%Y%m')

# Load stock returns
stock_returns = pd.read_csv('stock_returns.csv')

# Merge and prepare for regression
merged = pd.merge(stock_returns, ff_data, on='Date')
merged['excess_return'] = merged['return'] - merged['RF']

# Fit FF3 model
X = merged[['Mkt-RF', 'SMB', 'HML']]
y = merged['excess_return']

model = LinearRegression()
model.fit(X, y)

print(f"Alpha: {model.intercept_:.4f}")
print(f"MKT beta: {model.coef_[0]:.3f}")
print(f"SMB beta: {model.coef_[1]:.3f}")
print(f"HML beta: {model.coef_[2]:.3f}")
```