

Asset Pricing Model Comparison and Market Interpretation

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- Modern finance relies on models that link expected returns to risk exposures.
- I estimate two competing frameworks:
 - Capital Asset Pricing Model (CAPM)
 - Fama–French Five–Factor Model (FF5)
- The goal is to understand:
 - Which model better explains returns?
 - How do exposures differ across sectors?
 - What do alphas imply for risk–adjusted performance?

- Five sector-diversified assets:
 - Apple (AAPL) – Technology
 - JPMorgan (JPM) – Financials
 - ExxonMobil (XOM) – Energy
 - Walmart (WMT) – Consumer Retail
 - XLV – Healthcare ETF
- Daily returns: 2019–2024.
- Factor data downloaded from the Kenneth French database.
- Returns converted to excess returns (asset minus risk-free rate).

- For each asset, I estimate both:

$$R_{it} - R_{ft} = \alpha + \beta_{MKT}(MKT - RF)_t + \epsilon_t$$

$$R_{it} - R_{ft} = \alpha + \beta_{MKT}(MKT - RF)_t + \beta_{SMB}SMB_t + \beta_{HML}HML_t \\ + \beta_{RMW}RMW_t + \beta_{CMA}CMA_t + \epsilon_t$$

- I compare:
 - Alphas (risk-adjusted performance)
 - Betas (factor sensitivities)
 - R^2 (model fit)
- Python implementation ensures reproducibility.

Example Python Structure

```
data = load_prices_and_ff5(tickers, start, end)
summary_df = run_factor_regressions(data, tickers)
summary_df.to_csv("factor_regression_summary.csv")
make_factor_visuals(summary_df)
```

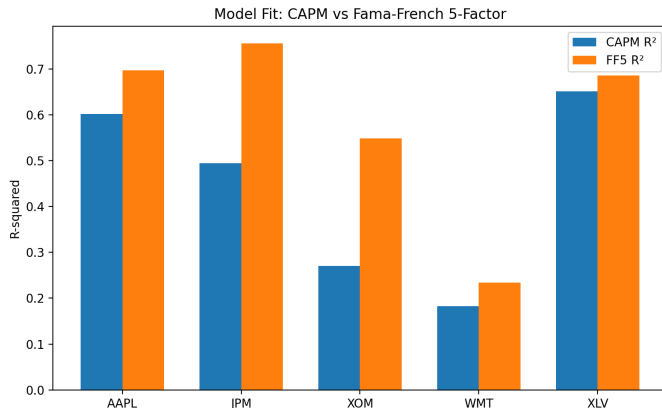
Note: Full code is available in the project script and uses pandas, statsmodels, yfinance, and matplotlib.

Regression Output Summary

Ticker	CAPM β_M	CAPM R^2	FF5 β_M	FF5 R^2
AAPL	1.16	0.60	1.26	0.70
JPM	1.04	0.49	1.05	0.76
XOM	0.81	0.27	0.89	0.55
WMT	0.44	0.18	0.53	0.23
XLV	0.69	0.65	0.75	0.69

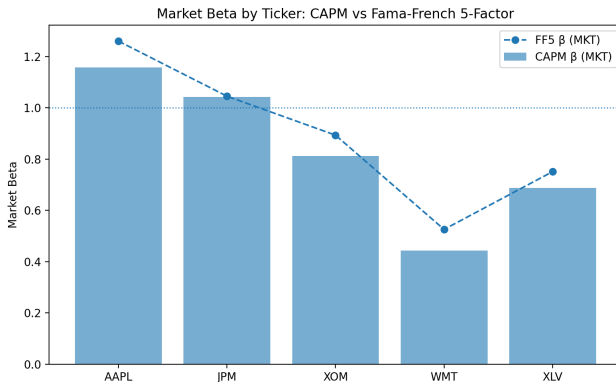
Interpretation: Across every asset, adding Fama–French factors increases R^2 , meaning multi-factor exposure explains returns better than market exposure alone.

CAPM vs FF5 Fit Visual



- FF5 materially improves explanatory power, especially for JPM and XOM.
- This suggests the single-beta CAPM omits important sources of priced risk.

Market Betas: CAPM vs FF5



- AAPL behaves as a high-beta growth stock ($\beta > 1$).
- WMT and XLV load less on market risk (betas well below 1).
- FF5 market betas are slightly higher because other factors absorb noise, revealing the core market sensitivity more clearly.

FF5 Factor Exposures and Significance

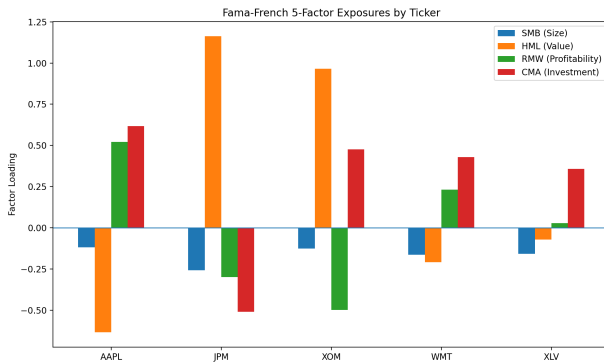
Ticker	MKT	SMB	HML	RMW	CMA
AAPL	1.26***	-0.12**	-0.63***	0.52***	0.62***
JPM	1.05***	-0.26***	1.16***	-0.30***	-0.51***
XOM	0.89***	-0.13*	0.96***	-0.50***	0.48***
WMT	0.53***	-0.16***	-0.21***	0.23***	0.43***
XLV	0.75***	-0.16***	-0.07**	0.03 (ns)	0.36***

*** $p < 0.01$, ** $p < 0.05$, * $p < 0.10$, (ns) not significant

Insights:

- The value factor (HML) is systematically important across assets.
- Profitability (RMW) is significant in tech and retail, but not for the healthcare ETF.
- Investment (CMA) plays a strong role for all assets, reflecting how capital intensity and reinvestment shape returns.

FF5 Factor Loading Visual



- AAPL and WMT load heavily on profitability and investment.
- JPM and XOM load strongly on value, consistent with “value stock” intuition.
- XLV’s profitability sensitivity (RMW) is near zero, suggesting healthcare returns are less driven by relative earnings strength.

Interpreting Alpha Estimates

- CAPM alphas are small and mixed in sign across assets.
- FF5 alphas are also close to zero once I control for all five factors.
- Based on the regression output, these alphas are not strongly statistically significant.
- **Interpretation:** Once risk exposures are measured properly, excess risk-adjusted performance is close to zero, which is consistent with (approximate) market efficiency.

- **AAPL:** Growth exposure, significant profitability and investment betas, negative value loading.
- **JPM:** Deep value exposure; negative size beta consistent with large-cap banking.
- **XOM:** Cyclical value stock with strong HML and investment exposure.
- **WMT:** Stable retailer whose returns are shaped by profitability and investment discipline.
- **XLV:** Defensive healthcare exposure with moderate market beta and limited profitability loading.

Binomial Option Pricing: Pfizer (PFE)

Setup:

- Underlying: Pfizer (PFE).
- Valuation date: Nov 26, 2025; expiry: Nov 26, 2026 ($T \approx 1$ year).
- Dividend yield $q = 3\%$ per year; tree steps $N = 500$.
- Volatility estimated from recent PFE returns (20-day window used below).

Prices (20-day vol, $\sigma \approx 28.31\%$, $r \approx 3.75\%$):

Strike	Euro Call	Amer Call	Euro Put	Amer Put
25	3.22	3.23	2.34	2.37
26	2.77	2.78	2.86	2.90
27	2.37	2.37	3.42	3.47

Interpreting the Binomial Results

- **American vs European:**

- American calls are slightly more valuable than European calls (about \$0.01).
- American puts are more valuable than European puts (about \$0.03–\$0.05).

- **Economic intuition:**

- Dividends plus the right to exercise early justify a premium for American options.
- Early exercise is especially relevant for in-the-money puts, which is where we see the largest differences.

- **Connection to the rest of the project:**

- Factor models explain expected stock returns; the binomial model shows how those stock dynamics translate into derivative prices.

Final Takeaways

- The Fama–French 5–factor model outperforms CAPM in explaining returns for all five assets.
- Factor exposures align with sector characteristics (growth vs value, cyclical vs defensive).
- Alphas are small and mostly insignificant, consistent with limited opportunities for persistent abnormal returns.
- The binomial model for PFE confirms theoretical predictions about American vs European option pricing under dividends.

Thank you!