

Evaluating a Survivorship–Bias–Free 12–1 Month Momentum Strategy in the S&P 500 (2005–2024)

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Abstract

We re-examine the canonical 12–1 month cross-sectional momentum strategy in the S&P 500 from 2005 to 2024 using a *survivorship–bias–free* constituent history, turnover-based transaction costs, and multi-factor regressions with Newey–West errors. Each month we go long the top decile and short the bottom decile of the 12–1 momentum signal, equal-weight both legs, and rebalance monthly. The strategy **does not** generate positive abnormal returns: the net annualized return is -2.07% with a Sharpe ratio of -0.34 (2% annual risk-free) and a maximum drawdown of -61% . Average one-way turnover is 27% per month; at 10 bps per side the implied average monthly cost is 0.027%. Factor regressions show that alpha is negative and statistically significant once the momentum factor (UMD) is included; R^2 rises from ≈ 0.28 (FF5) to ≈ 0.82 (FF5 + UMD). These findings suggest that naïve large-cap momentum primarily reflects exposure to the well-documented momentum factor and does not deliver residual alpha once realistic frictions are accounted for.

Keywords: momentum; cross-sectional; turnover; transaction costs; factor models; survivorship bias.

JEL: G11; G12; C58.

1 Introduction

Momentum—the tendency for recent winners to continue outperforming losers—is one of the most robust empirical regularities in asset pricing (Jegadeesh and Titman, 1993; Carhart, 1997; Asness et al., 2019). Early studies document large abnormal returns from simple cross-sectional strategies, but subsequent work emphasizes their fragility. Momentum portfolios have experienced sharp crashes (Daniel and Moskowitz, 2016), while trading frictions erode profitability in practice (Novy-Marx and Velikov, 2016). Even commercial momentum indices have often underperformed broad benchmarks.

Research question. *Can a naïve 12–1 month momentum rule, applied to the S&P 500 with realistic frictions, still generate statistically significant abnormal returns?*

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Contribution. This paper provides a transparent replication using freely available data and reproducible code. Specifically:

- (i) We eliminate survivorship bias by reconstructing dated index membership and masking returns outside inclusion months.
- (ii) We apply a turnover-based transaction cost model calibrated at 10 basis points per side, and report sensitivity to alternative cost assumptions.
- (iii) We assess abnormal performance using multi-factor regressions (CAPM, FF3, FF5, and FF5 + UMD) with Newey–West robust errors.

Preview of results. The strategy underperforms: net returns are negative, drawdowns severe, and alphas vanish once momentum exposure is controlled for. Our findings reinforce the interpretation of momentum as a priced risk factor rather than a persistent source of residual alpha in U.S. large caps.

2 Data

Universe and sample. The universe comprises all firms that were S&P 500 constituents at any point between January 2005 and December 2024 ($T=252$ months). Historical additions and deletions are scraped from Wikipedia and merged with Yahoo Finance adjusted-close prices. Returns are computed only when a ticker is an active constituent, removing survivorship bias.

Price source and cleaning. End-of-month adjusted closes are downloaded via `yfinance` (`interval="1mo"`). Nonpositive prices are excluded. Monthly returns are winsorized cross-sectionally at the 1st and 99th percentiles when at least 50 names are present, after filtering out obvious glitches ($|\log(1 + r)| > 1.5$).

Factors and benchmarks. Monthly Fama–French factors (Mkt-RF, SMB, HML, RMW, CMA), the risk-free rate, and the momentum factor (UMD) are obtained from the Ken French data library. The U.S. market benchmark is defined as $\text{Mkt-RF} + R_f$.

Descriptive statistics. Table 1 reports descriptive statistics for the strategy and the U.S. market. The strategy has negative mean returns, high kurtosis, and deeper drawdowns.

Table 1: Summary statistics (monthly, decimals). Sample: 2005–2024, $N=252$ months.

Series	N	Mean (m)	Vol (m)	Sharpe	Ann Ret	Ann Vol	Skew	Kurtosis	Max DD
Strategy (net)	252	-0.0017	0.0341	-0.34	-0.0207	0.1183	-2.283	16.967	-0.613
US Market	252	0.0094	0.0439	0.61	0.1184	0.1519	-0.553	1.403	-0.503

3 Methodology

Signal and portfolio formation. For each month t , compute the cumulative return from $t-13$ to $t-1$. Rank stocks into percentiles; go long the top 10% and short the bottom 10%, equal-weighted within legs. Portfolios are rebalanced monthly.

Transaction costs. Let w_t denote portfolio weights at t . Dollar turnover is $\frac{1}{2} \sum_i |w_{i,t+1}^{\text{pre}} - w_{i,t}|$. Net return is

$$r_{t+1}^{\text{net}} = w_t^\top r_{t+1} - c \cdot \text{turnover}_{t+1},$$

with $c=10$ bps. Average turnover is 27%, implying costs of 0.027% per month.

Performance metrics. We report annualized mean, volatility, Sharpe (2% RF), and maximum drawdown. Risk-adjusted alphas are estimated with OLS and Newey–West standard errors (6 lags).

4 Results

4.1 Performance with turnover costs

At 10 bps per side, the strategy yields a net annualized return of -2.07% , volatility of 11.83%, Sharpe of -0.34 , and a maximum drawdown of -61% . Performance deteriorates monotonically as costs increase (Table 2).

Table 2: Transaction-cost sensitivity. Average monthly turnover $\approx 27\%$.

Cost (bps)	Ann Ret	Vol	Sharpe	Max DD	Avg Turnover
5	-1.91%	11.83%	-0.33	-60.20%	27.0%
10	-2.07%	11.83%	-0.34	-61.28%	27.0%
15	-2.23%	11.83%	-0.36	-62.33%	27.0%
25	-2.55%	11.83%	-0.39	-64.34%	27.0%

4.2 Factor regressions

Table 3 reports regressions of excess returns on CAPM, FF3, FF5, and FF5 + UMD. Alphas are insignificant under CAPM/FF5, but turn negative and significant once UMD is included, with R^2 rising to 0.82.

5 Conclusion

Using dated membership, turnover-based costs, and factor regressions with robust errors, we find that a naïve 12–1 month momentum strategy in the S&P 500 fails to deliver positive abnormal

Table 3: Factor regressions (net returns). HAC SEs, 6 lags. Alphas annualized.

Model	Alpha (ann)	t(Alpha)	R^2	n
CAPM	-0.72%	-0.32	0.147	252
FF3	-1.65%	-0.70	0.269	252
FF5	-1.56%	-0.65	0.279	252
FF5 + UMD	-4.03%	-3.23	0.823	252

returns during 2005–2024. Returns are largely explained by momentum exposure, and realistic frictions render the implementation unprofitable.

Limitations and future work.

- (i) Our membership reconstruction relies on Wikipedia; more authoritative CRSP membership files may refine accuracy.
- (ii) Turnover costs are modeled uniformly at 10 bps; liquidity-adjusted costs may vary across stocks.
- (iii) Factor regressions use standard Fama–French sets; exploring alternative factors or structural models may yield further insight.

Reproducibility note. All code, data, and scripts to replicate this analysis are available at <https://github.com/dshan12/Momentum-Research.git>. Tables and figures in this paper are auto-generated.

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