

Stress-Testing a Tactical Allocation Model: Robustness and Improvements

Performance and resilience across different specifications



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Hi there,

Over the past weeks, I've received several thoughtful questions from subscribers about the tactical allocation model I discussed [earlier](#). Many of you asked about robustness: What happens if we change the weighting scheme? What if we adjust the thresholds or modify the defensive set of assets? Others wondered about possible improvements and whether the strategy holds up across variations.

In today's post, I'll take a systematic look at these issues. My main takeaway upfront is that the framework is surprisingly robust. Across weighting choices, signal specification and parameter shifts, the model continues to deliver equity-like returns with far smaller drawdowns than SPY. Still, there are interesting nuances worth highlighting.



Introduction

I won't repeat the full details of the tactical allocation model here. If you'd like a refresher, please see my earlier [post](#) where I outlined the mechanics. Instead, this article focuses on robustness: How the model behaves when we vary key assumptions such as weighting schemes, defensive thresholds, lookback windows, and fallback rules. The goal is to stress-test the framework and see whether its performance holds up across different specifications.

Weighting Schemes

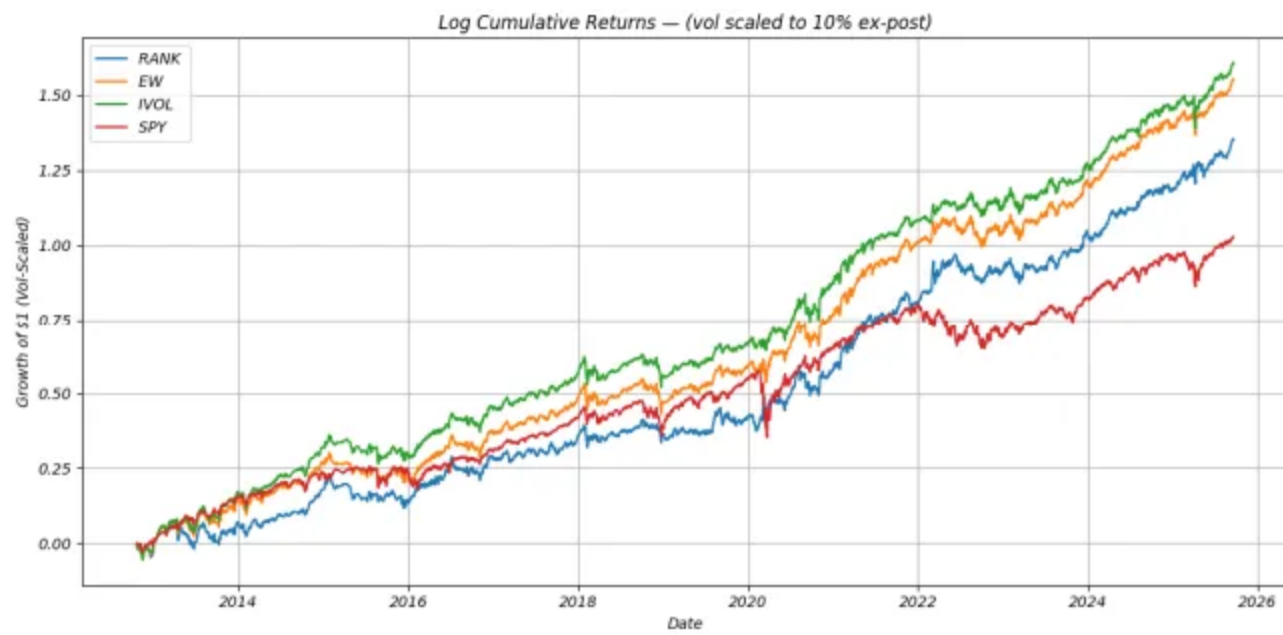
The first dimension to test is *how we combine the defensive assets*. In the base version used rank-based weights as in the original paper, giving more weight to stronger momentum signals. However, signals are noisy, and it's not certain that allocating more to larger but noisy signals is necessarily optimal. Two natural alternatives are worth examining:

- Equal-weighting the defensive set, where each qualifying asset gets the same share.
- Risk-weighting, or inverse-volatility weighting, where more capital is allocated to lower-volatility defensives. I use a rolling 3-month volatility to determine weight

The weighting alternatives only apply to the defensive bucket, whereas the equity exposure, the fallback asset, simply receives the residual weight.

Running the model across rank weights, equal weights, and risk weights yields the following results.

=== Summary Stats (vs SPY) ===						
	RANK	EW	IVOL	SPY		
CAGR %	11.08	12.90	12.87	16.38		
Ann.Vol %	8.96	9.02	8.68	17.90		
Max DD %	-9.40	-10.10	-9.10	-33.72		
Sharpe	1.08	1.26	1.30	0.87		
Corr w/ SPY	0.29	0.65	0.65	1.00		
DD/Vol	-1.05	-1.12	-1.05	-1.88		
=== Annual Returns (%): Strategies vs SPY ===						
	RANK	EW	IVOL	SPY		
Date						
2012	-3.26	-3.13	-2.89	-1.66		
2013	10.16	17.95	19.60	33.92		
2014	10.28	11.98	13.92	15.61		
2015	-3.91	-4.63	-2.44	-0.17		
2016	13.94	13.90	14.32	12.00		
2017	7.18	11.45	10.94	21.71		
2018	0.51	-0.95	-0.57	-4.57		
2019	4.62	10.65	9.28	31.22		
2020	16.52	18.48	21.15	18.33		
2021	22.61	24.01	19.28	28.73		
2022	10.13	3.24	4.14	-18.18		
2023	11.74	16.32	13.02	26.18		
2024	13.61	17.50	17.16	24.89		
2025	16.55	15.87	14.61	13.67		
=== Average Portfolio Weights (full sample) ===						
	TLT	GLD	DBC	UUP	BTAL	SPY
RANK	0.141	0.166	0.151	0.129	0.116	0.297
EW	0.107	0.117	0.104	0.109	0.096	0.469
IVOL	0.091	0.097	0.075	0.191	0.077	0.468



When we compare the three weighting schemes, the differences are subtle but meaningful. Equal-weighting and inverse-volatility weighting delivered higher CAGR ($\approx 12.9\%$) than rank weighting (11%), primarily because they allocate more capital to equities, on average, about 47% versus 30% under rank. That higher equity tilt brings them closer to SPY, with correlations of 0.65 compared to just 0.29 for rank.

Rank weighting, in contrast, produces the lowest equity beta, the smallest max drawdowns, and the most differentiated return stream. Interestingly, equal-weighting and inverse-volatility weighting achieve the highest Sharpe ratios (1.26 and 1.30) compared to 1.08 for rank. In practice, rank weighting may appeal to those explicitly seeking a return stream that has low correlation to equities, while EW and IVOL tend to offer higher equity exposure, higher CAGR, and higher Sharpe ratios.

It's comforting to see that equal-weighting, an even simpler construction than rank-based weights, produces a higher Sharpe ratio. By spreading capital evenly across the defensives, the model achieves a Sharpe ratio of 1.26, outperforming the original rank-based scheme. This suggests that the robustness of the framework does not rely on fine-tuned weightings, as even a naive allocation rule delivers strong risk-adjusted returns.

Excluding Defensive Assets

The main idea of the model is to include defensive assets that are different in nature with low correlation to each other, and which diversify equity risk in different ways. Let's check the performance impact of each specific defensive asset by excluding one asset at a time.

=== Summary Stats (Rank-based): Full vs Excluding One Defensive ===							
	FULL	EXCL_TLT	EXCL_GLD	EXCL_DBC	EXCL_UUP	EXCL_BTAL	SPY
CAGR %	11.08	12.23	11.31	9.68	10.99	11.67	16.38
Ann.Vol %	8.96	9.50	9.04	9.34	10.56	9.77	17.90
Max DD %	-9.40	-9.07	-12.23	-10.15	-14.53	-13.06	-33.72
Sharpe	1.08	1.14	1.10	0.91	0.93	1.05	0.87
Corr w/ SPY	0.29	0.35	0.27	0.16	0.29	0.45	1.00
DD/Vol	-1.05	-0.96	-1.35	-1.09	-1.38	-1.34	-1.88

The exclusion tests confirm that no single defensive asset dominates the strategy, a performance holds up reasonably well even if one is removed. Most exclusions stay close to the full model, though there are nuances:

- Excluding GLD delivers a similar Sharpe as the full model with worse drawdown suggesting gold is a valuable hedge when other assets fail.
- Excluding DBC or UUP weakens results: CAGR falls to 9–10% and Sharpe drops about 0.9. Both provide diversification in specific regimes (inflation shocks for DBC, dollar bull runs for UUP), so their absence seems to hurt resilience.
- Excluding BTAL leads to a bigger drawdown (–13%), confirming its role as an equity hedge during downturns.
- The full mix still looks strongest on balance, offering the best trade-off between CAGR, drawdowns, and low correlation to SPY.

Overall, this exercise reinforces the point that robustness comes from *combining heterogeneous defensives* as each asset contributes differently during the sample period.

Transaction Costs

Some readers also asked about transaction costs. These are investor-specific, driven by trade size, venue, and execution quality. I assume a one-way cost of 5 bps in the base case, but I stress-test higher costs. As costs rise toward 25 bps, the strategy's Sharpe ratio compresses and converges toward SPY's.

=== Summary Stats (Rank-based, varying costs) ===						
	RANK_5bps	RANK_10bps	RANK_15bps	RANK_20bps	RANK_25bps	SPY
CAGR %	11.08	10.59	10.11	9.63	9.16	16.38
Ann.Vol %	8.96	8.95	8.95	8.95	8.95	17.90
Max DD %	-9.40	-9.63	-9.86	-10.08	-10.31	-33.72
Sharpe	1.08	1.03	0.99	0.94	0.89	0.87
Corr w/ SPY	0.29	0.29	0.29	0.29	0.29	1.00
DD/Vol	-1.05	-1.08	-1.10	-1.13	-1.15	-1.88

Changing Defensive Threshold

While it is intuitive to use the risk-free rate as the hurdle for deciding whether an asset's momentum score justifies an allocation, it is worth asking how sensitive the model is to that choice. To test this, I rerun the strategy with six fixed thresholds of 1%, 2%, 3%, 5%, and 7%, and compare them to the benchmark RF-based specification. The results show that performance remains robust: Sharpe ratios and drawdowns are broadly stable across all thresholds, indicating that the model is not overly dependent on the precise choice of hurdle.

=== Summary Stats: RF Benchmark and Absolute Hurdles vs SPY (cost = 5 bps) ===								
	BENCH	0pct	1pct	2pct	3pct	5pct	7pct	SPY
CAGR %	11.08	10.29	10.66	11.11	11.12	12.36	11.77	16.38
Ann.Vol %	8.96	8.87	8.87	8.99	8.97	9.14	9.68	17.90
Max DD %	-9.40	-9.40	-9.31	-9.99	-9.99	-9.07	-12.29	-33.72
Sharpe	1.08	1.01	1.05	1.08	1.09	1.19	1.07	0.87
Corr w/ SPY	0.29	0.27	0.28	0.31	0.34	0.38	0.49	1.00
DD/Vol	-1.05	-1.06	-1.05	-1.11	-1.11	-0.99	-1.27	-1.88

Excluding Lookback Windows

How sensitive is the model to the choice of lookbacks? In its current form, the strategy computes each defensive asset's annualized return over 21, 63, 126, and 252 days a

averages the four. Below, I rerun the model while omitting one window at a time.

The results are reassuring: Performance is robust across all exclusions. Sharpe ratios remain broadly similar from spec to spec, but they are consistently lower than the full specification. That gap suggests the full model benefits from diversifying across horizons, blending short-, medium-, and long-term momentum improves efficiency (higher Sharpe) and reduces drawdowns relative to any single-window subset.

=== Summary Stats: Benchmark vs Excluding One Lookback vs SPY (cost = 5 bps) ===						
	BENCH	EXCL_21	EXCL_63	EXCL_126	EXCL_252	SPY
CAGR %	11.08	10.43	10.09	10.58	10.41	16.38
Ann.Vol %	8.96	8.85	8.73	8.96	9.14	17.90
Max DD %	-9.40	-8.36	-11.83	-11.00	-12.31	-33.72
Sharpe	1.08	1.03	1.01	1.03	1.00	0.87
Corr w/ SPY	0.29	0.29	0.31	0.31	0.31	1.00
DD/Vol	-1.05	-0.94	-1.36	-1.23	-1.35	-1.88

Requiring Fallback Asset to Have Positive Momentum

A thoughtful reader asked: *"What if we only fall back to SPY if it also has positive momentum?"*

In the benchmark setup, the model relies on defensive assets performing well during equity downturns. But one can imagine a scenario, although not yet observed in the backtest, where all defensive assets exhibit weak momentum *and* SPY is entering a drawdown. In that case, the benchmark would still allocate fully to SPY, which could be very costly.

To address this, I rerun the model with an additional filter: SPY must also have a momentum score above the risk-free rate to receive any fallback allocation. If not, the fallback weight instead goes to cash, earning the risk-free rate.

The results show that this modification modestly reduces performance: The Sharpe ratio declines to 0.97 and drawdowns tick higher. However, it adds a layer of protection by ensuring the model never allocates to an asset with poor or negative

momentum. For more conservative investors, this may be a worthwhile trade-off, offering additional peace of mind even at the cost of some performance.

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=== Summary Stats: Benchmark vs SPY-filtered fallback vs SPY (cost = 5 bps) ===
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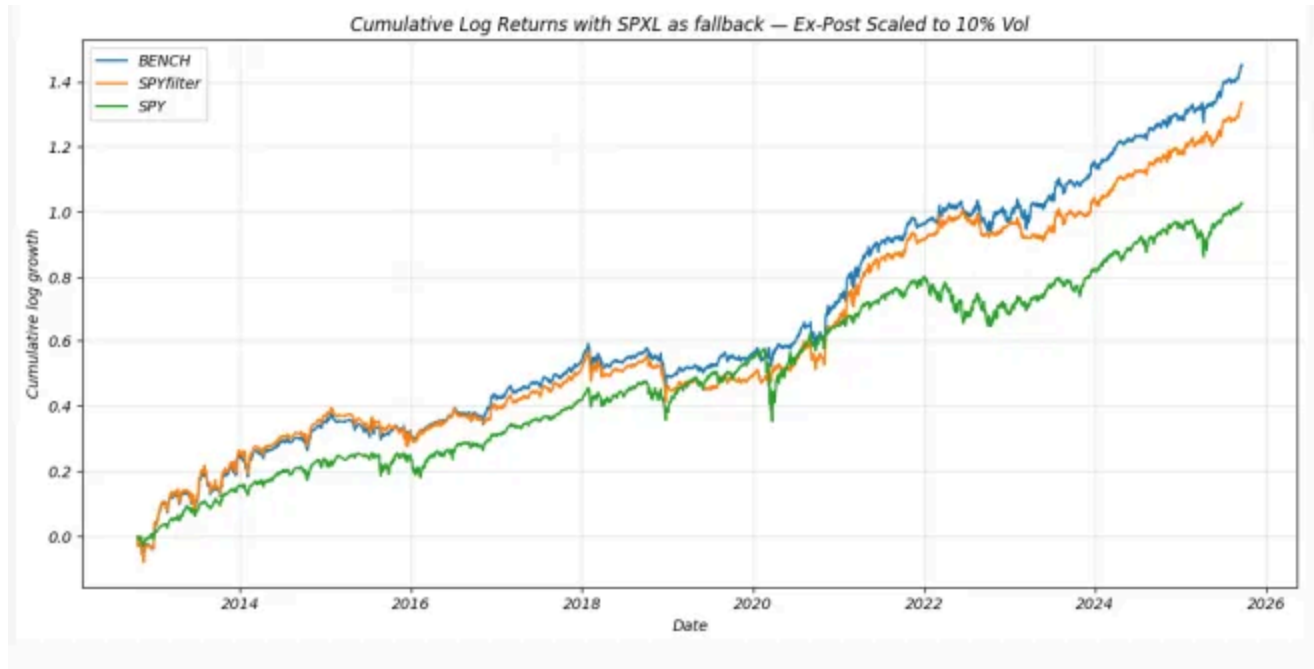
	BENCH	SPYfilter	SPY
CAGR %	11.08	9.99	16.38
Ann.Vol %	8.96	8.94	17.90
Max DD %	-9.40	-10.50	-33.72
Sharpe	1.08	0.97	0.87
Corr w/ SPY	0.29	0.16	1.00
DD/Vol	-1.05	-1.17	-1.88

If an investor is willing to take on some leverage through a triple-levered ETF, SPXL, the model with a momentum filter on SPXL has offered a CAGR of about 20%, a Sharpe of 1.13, and with a 23% drawdown, all beating SPY over time, as shown below under *SPYfilter*.

Fallback Asset: SPXL, with momentum filter

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=== Summary Stats: Benchmark vs SPY-filtered fallback vs SPY (cost = 5 bps) ===
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	BENCH	SPYfilter	SPY
CAGR %	23.46	20.27	16.38
Ann.Vol %	17.33	16.47	17.90
Max DD %	-22.00	-23.09	-33.72
Sharpe	1.23	1.13	0.87
Corr w/ SPY	0.61	0.43	1.00
DD/Vol	-1.27	-1.40	-1.88



Conclusion and Investor Takeaway

Taken together, these robustness checks show that the tactical allocation model is not fragile. Whether I vary the weighting scheme, exclude individual defensives, raise transaction costs, shift thresholds, drop lookback windows, or require positive momentum for the fallback equity sleeve, the broad picture is the same: Equity-like returns with substantially smaller drawdowns. The nuances matter, equal-weighting boosts Sharpe, gold and BTAL play unique hedging roles, and a SPY momentum filter offers extra downside comfort at the cost of a somewhat lower Sharpe and higher drawdowns, but none of these changes overturn the model's core strengths.

Investor takeaway: This is a flexible, resilient framework rather than a parameter-sensitive and data-mined model. Investors can adapt the specification to their own preferences, whether prioritizing higher Sharpe, lower correlation to equities, or stronger downside protection, while still retaining the core benefits of systematic tactical allocation.

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