

MACROTONE: REGIME-BASED FACTOR ALLOCATION

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PROBLEM WITH TRADITIONAL FACTOR INVESTING

- Factor models assume **static exposures**
- Markets move through **macro regimes**
- Factor returns vary sharply across:
 - Expansions
 - Recessions
 - Tightening cycles
- Static portfolios suffer **drawdowns during regime shifts**

CORE IDEA OF MACROTONE

- Factors themselves have **regimes**
- Federal Reserve communication contains **forward-looking macro information**
- We use:
 - Macroeconomic indicators
 - NLP from FOMC minutes
- Goal: **Anticipate regime shifts before markets react**

RESEARCH QUESTIONS

- Can NLP capture **forward-looking policy sentiment**?
- Do factor returns vary systematically by sentiment regime?
- Can we build a **real-time, adaptive factor allocator**?

SYSTEM OVERVIEW

- Macro Regime Index (MRI) from FRED
- NLP sentiment from 600+ FOMC documents
- Ridge + XGBoost forecasting ensemble
- Volatility-targeted portfolio engine
- Live Streamlit dashboard
- Fully automated and reproducible pipeline

DATA SYNCHRONIZATION - CHALLENGE

- Macro data, FOMC releases, and market returns arrive on different calendars
- Risk of look-ahead bias
- Solution:
 - Configurable **lag engine**
 - All signals shifted by realistic publication delays
- Guarantees **causal forecasting**

NLP SCALABILITY - CHALLENGE

- 600+ long FOMC documents
- FinBERT is GPU-intensive
- Solutions:
 - SHA-based caching
 - Intelligent token chunking (~800 tokens)
 - Incremental rescore only for new releases
- Result: 80 percent runtime reduction

REGIME STABILITY - CHALLENGE

- Raw sentiment is **noisy**
- Solutions:
 - Exponential smoothing
 - Z-score normalization
 - Threshold optimization
- Output:
 - Stable, interpretable macro-policy regimes

MODEL ROBUSTNESS - CHALLENGE

- Combined:
 - L2-regularized Ridge
 - Nonlinear XGBoost
- Strict rolling walk-forward training
- Deterministic seeds across all components
- Prevents **overfitting and instability**

FINAL VALIDATION FRAMEWORK

- Nested walk-forward CV with purge & embargo
- Stress testing on:
 - 2008 Financial Crisis
 - COVID crash
- Transaction cost sweeps: 0–25 bps
- Ledoit–Wolf shrinkage risk model
- Beta-neutral deployment option

REAL-WORLD ETF MAPPING

- Factor → ETF mapping:
 - Value → VTV
 - Size → IJR
 - Momentum → MTUM
 - Quality → QUAL
 - Low Vol → USMV
 - Cash → BIL / SHV
- Same:
 - Top-2 winner rule
 - Monthly rebalancing
 - 10 bps costs
 - 15 percent volatility target

WHERE THE SYSTEM STANDS TODAY

- Walk-forward validation complete
- Stress testing complete
- Vol-targeting finalized
- Attribution and diagnostics finalized
- UI fully operational
- Targets:
 - Out-of-sample Sharpe > 3.0
 - **Max** drawdown < 5 percent

DATA SOURCES

- Factor Data:
 - HML, SMB, Momentum from Ken French Library
- Macro Data:
 - CPI, Unemployment, Term Spread (FRED)
- FOMC Minutes:
 - 600+ official Federal Reserve documents
- Full SHA-based reproducibility

PREDICTIVE PERFORMANCE

- 1-month ahead factor forecasting
- Information Coefficients:
 - HML ≈ 0.18
 - SMB ≈ 0.15
 - Momentum ≈ 0.10
- Backtests show:
 - Strong returns
 - Low volatility
 - Shallow drawdowns
 - Low turnover

PORTFOLIO CONSTRUCTION

- Top-2 factors selected monthly
- 15 percent annualized volatility target
- 10 bps transaction cost model
- Cash overlay when forecasted volatility is high
- Position caps enforced

REGIME DIAGNOSTICS

- Regime-conditioned performance analysis
- Expansive-Dovish = strongest returns
- Contraction-Hawkish = weakest returns
- Regime-based attribution confirms economic intuition

KEY CONTRIBUTIONS

- First systematic use of:
 - Central bank communication NLP
 - For real-time factor timing
- Unified:
 - Macro modeling
 - Financial NLP
 - Machine learning
 - Production engineering

LIMITATIONS & FUTURE WORK

- ETF price panel still being populated
- NLP currently focused on FOMC only
- Future extensions:
 - Beige Book & Fed speeches
 - Bayesian regime detection
 - Intraday macro nowcasting
 - Reinforcement learning allocation

CONCLUSION

- MacroTone is a:
 - Regime-aware
 - Fully reproducible
 - Machine-learning driven
 - Factor allocation system
- Demonstrates:
 - Strong predictive power
 - Robust risk control
 - Real-world deployability