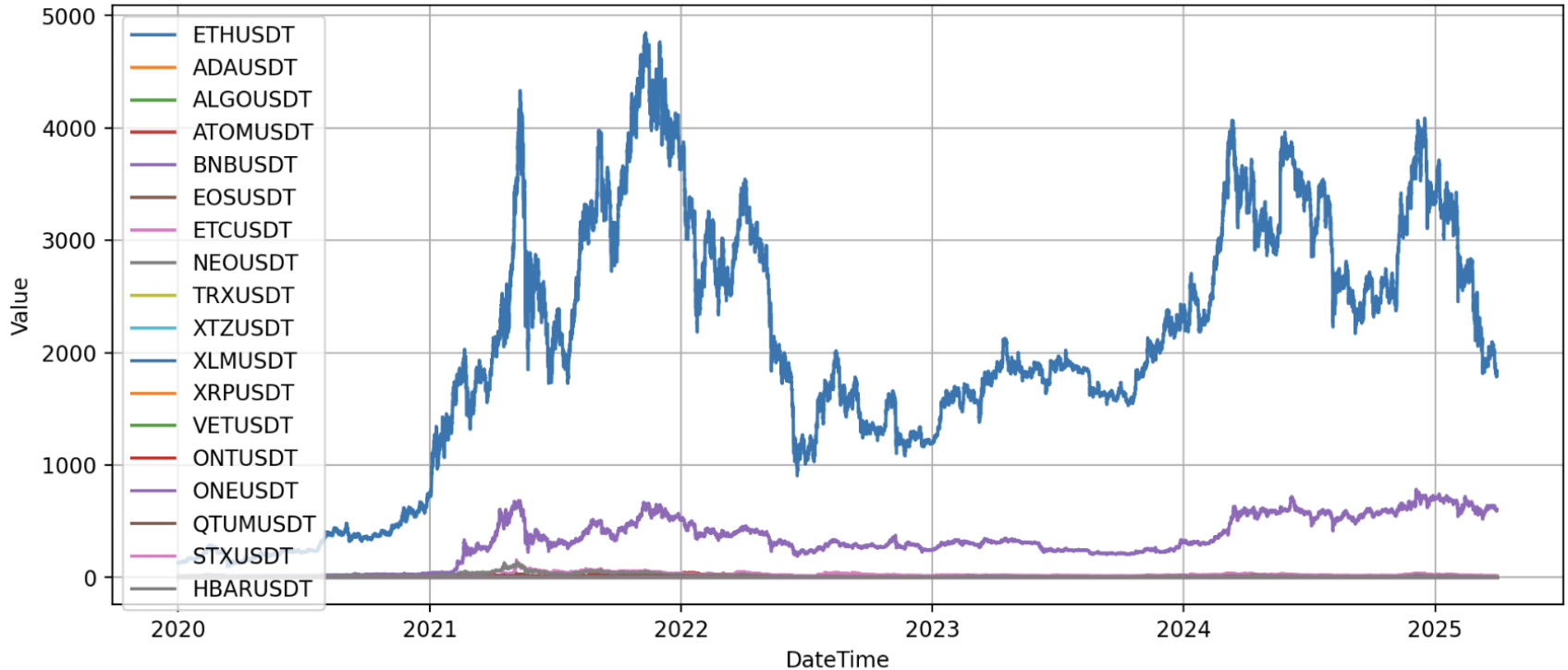


# CRYPTO HEDGE FUND MVP



# Structure

## 1. Hedge Fund Model:

*Vision and components*

## 2. Risk management

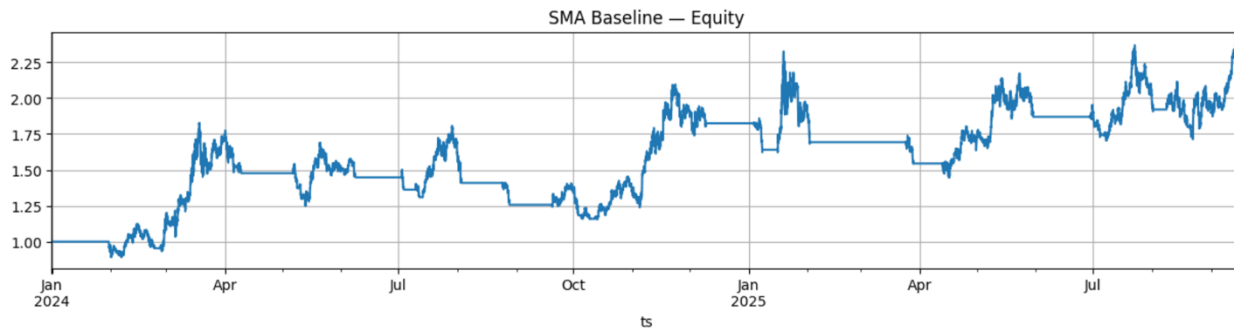
*Main concepts and metrics*

## 3. Portfolio management

*Main concepts and metrics*

## 4. System Architecture

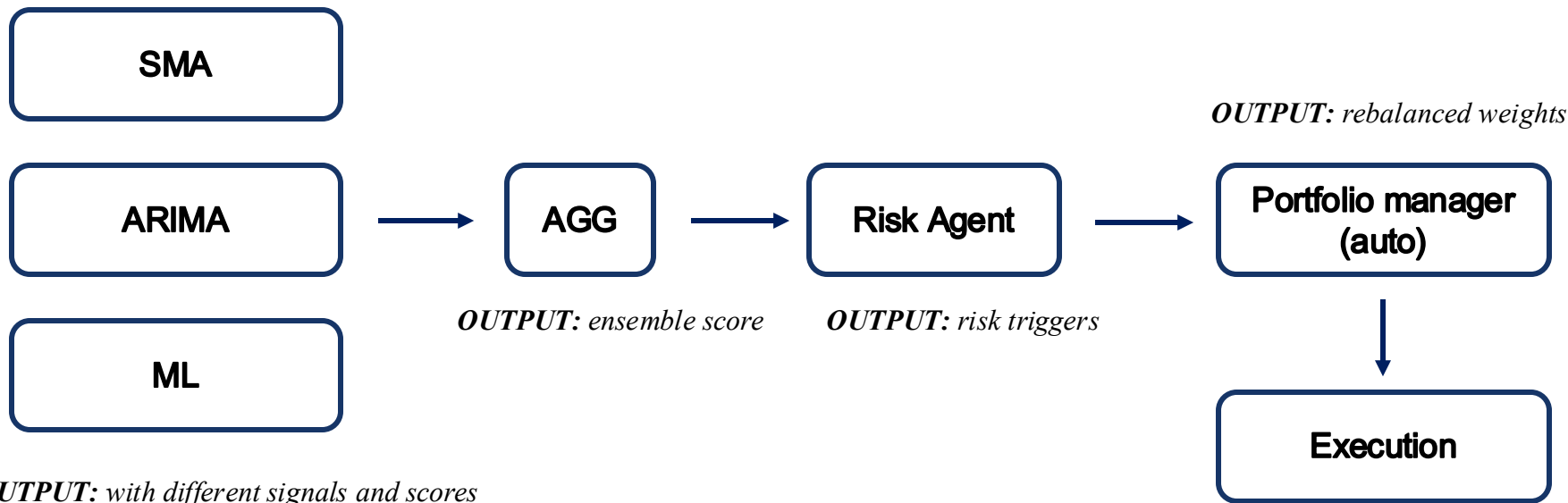
*Architecture example*



# Hedge Fund Model

## *AI Role*

Optimization in decision making, calculate and graduate market risk metric: forecasting models and trading agents



| Model                                      | Pros   | Cons   | Possible Role  |
|--|--|--|--|
| SMA  | Easy to implement and interpretate<br>Captures medium/long-term trends | Lags behind price (slow reaction to regime shifts)                       | Serves as a <b>baseline</b> strategy and benchmark; useful for long-term trend detection   |
| ARIMA                                      | Captures autocorrelation and mean-reversion patterns                   | Struggles with non-linear dynamics and regime shifts                     | Used to model short-term <b>statistical dependencies</b> in returns  |
| GARCH                                      | Captures volatility clustering (common in crypto)                      | Focused only on <b>variance</b> , not mean returns                       | Incorporated into the <b>risk management module</b> : volatility forecasts to reduce leverage or scale down positions  |
| Logistic Regression                        | Simple, fast, and interpretable classification model                   | Linear decision boundary<br>Sensitive to feature scaling and correlation | Serves as a <b>baseline ML classifier</b> for directional forecasts;<br><br>useful for interpretability  |
| Boosting                                   | Handles <b>non-linearities</b> and feature interactions                | Risk of overfitting without careful tuning                               | Provides the core predictive agent for signal generation;  |
| Aggregation (Ensemble / Voting / Weighted) | Diversifies across models → reduces overfitting                        | Requires monitoring of model correlations and weights                    | Acts as the <b>meta-agent</b> in architecture: combines outputs from different models into one unified trading signal, balancing simplicity with robustness. |

# Risk management

## *Role of Risk Management in a Hedge Fund*

**Capital Preservation:** protects the fund against catastrophic losses.

**Stability:** ensures smoother equity curves, less volatility in returns.

**Position Sizing:** adjusts exposure according to market risk levels.

**Regime Adaptation:** reduces leverage or exits during high-volatility or illiquid periods.

**Investor Confidence:** provides transparency and discipline for stakeholders.

### **Core Metrics:**

- Volatility
- Drawdowns
- VaR
- Sharp Ratio

# Risk management

## *AI Role*

*AI-driven agents monitor **volatility**, **liquidity**, and tail risks in real time, dynamically adjusting exposure, enforcing stop signals, and protecting capital during turbulent periods.*

## **Volatility Estimation:**

*Historical volatility*

***GARCH** models for conditional variance*

***ML-based** forecasting*

## **Liquidity Estimation:**

***Bid-ask** spread and *Amihud* liquidity ratio*

***Order book** depth analysis (volume at top levels).*

***Slippage simulation** based on order size vs. market depth.*

# Portfolio management

Popular **theories** of portfolio management:

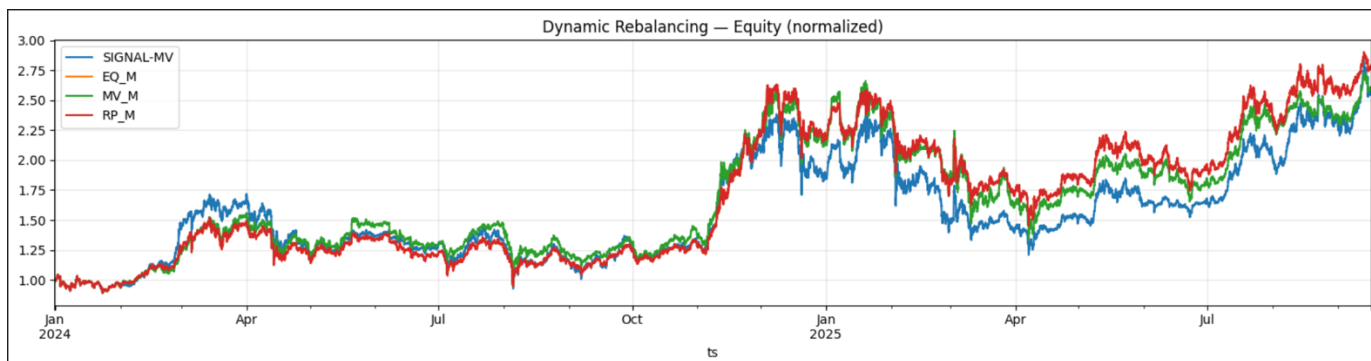
- Modern Portfolio Theory (Markowitz)
- Risk Parity
- Black–Litterman
- Factor Models

## *Optimal portfolio:*

Balanced **return, volatility, drawdown and liquidity constraints.**

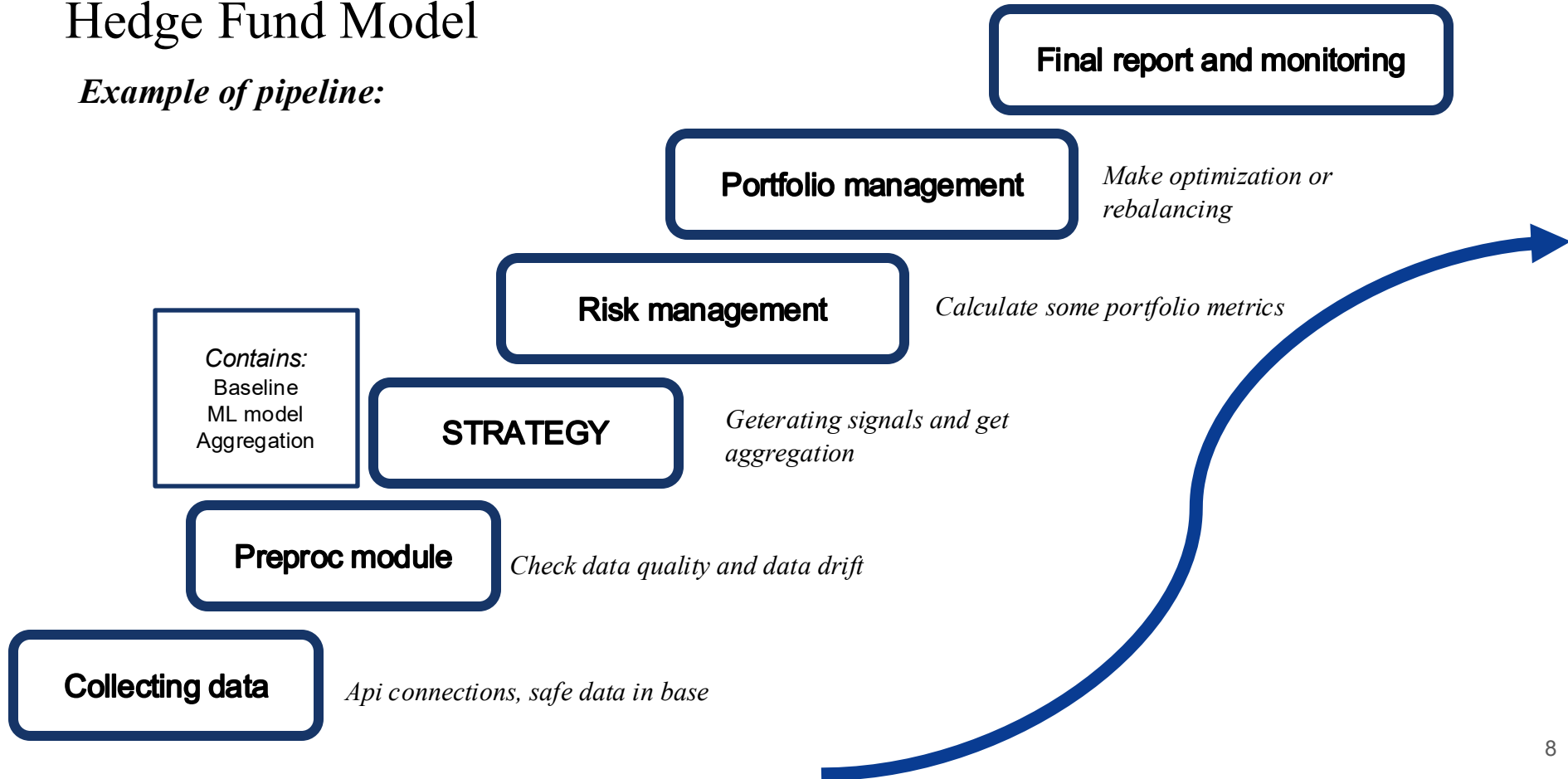
Avoid over-concentration by accounting for **correlations** between assets.

Evaluated using **risk-adjusted performance metrics.**



# Hedge Fund Model

*Example of pipeline:*





# Key Takeaways:

- The AI-based hedge fund model combines **classical theory** with **ML** for adaptive, real-time decision-making.
- **Risk management** is central to portfolio stability, with **AI agents** dynamically adjusting for volatility and liquidity.
- **Portfolio management** leverages theories to optimize risk-adjusted returns.

## Next Steps:

- *Expand* the library of signals using strong, complementary models
- *Enhance* the aggregation process and define clear risk-adjusted performance ranges
- *Fast Scalability*
- *Incorporate bot* for live monitoring and manual adjustments.
- *Set monitoring system*