

Analysis and calibration of mean- reversal strategies on capital markets

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Motivation and Context

Mean Reversion Strategy

Trading strategies based on the assumption that prices of financial instruments return to an equilibrium value (moving average) after temporary deviations

Optimal moving average (MA) length depends on market volatility

Lack of universal formula for predicting optimal MA without extensive testing

Research Question and Hypothesis

Research Question

Can we derive a generalized formula for calculating optimal moving average length based solely on the volatility of a financial instrument?

Hypothesis H_0

Currency pairs with lower volatility exhibit more stable results in mean reversion strategy and allow for more efficient moving average calibration

Methodology

Data & Instruments

Hourly data, 5-10 years

FX pairs, commodities, indices

10+ instruments

Strategy

$MA \pm K \times \sigma$ bands

Long/Short on deviation

Exit on return to MA

Tested Parameters

MA: 30-200 hours

K: 0.5-2.5

Sharpe Ratio, AvgReturn

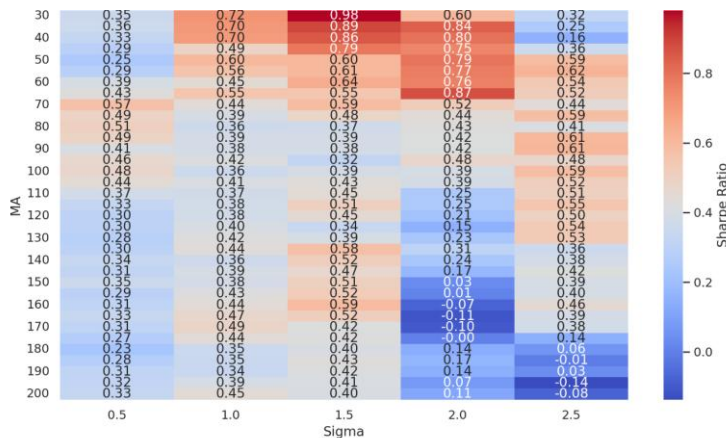
DS, Max Drawdown

Price Series with SMA 20 and 2σ Bands (First 300 days)



Results: Sharpe Ratio by MA and Sigma

Sharpe ratio vs MA vs Sigma



Key Findings

Optimum at MA 35-50

Sigma (K) = 1.5-2.0

Max Sharpe $\approx$ 1.0-1.4

Interpretation

Highest Sharpe ratios achieved with relatively shorter moving averages and medium K values

Directional Success

DS $\approx$ 50-51% – statistically real but not dominant

Color scale: red = higher Sharpe, blue = lower Sharpe

Results by Market Volatility

Low Volatility

Major FX pairs (~5-8% p.a.)

Sharpe: 1.0-1.4 ✓

Max DD: ~15%

DS: ~51%

Annual return: 10-13% p.a.

Medium Volatility

Minor FX pairs (~10-15% p.a.)

Sharpe: 0.8-1.1

Max DD: 20-25%

DS: ~50%

Annual return: single digits
%

Commodities, indices (>30% p.a.)

Sharpe: <0.5 ⚠

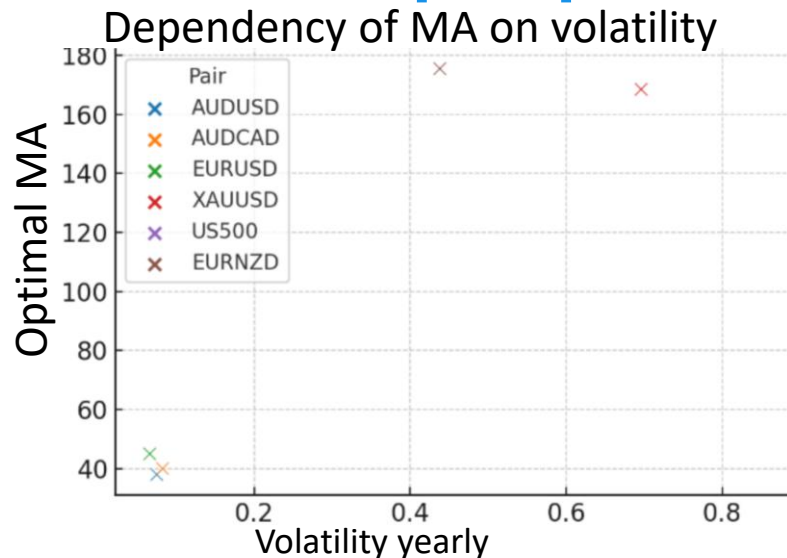
Max DD: >30%

DS: ~50%

Model: 3/ σ fails

Hypothesis confirmed: Less volatile markets exhibit more stable mean reversion results with higher Sharpe ratios and lower drawdowns

Relationship: Optimal MA Length vs Volatility



Empirical Data

Observation: For low volatility, optimal MA is in tens of periods; for high volatility, in hundreds

Currency pair	Length n (days)	Number of trades	Avg profit per trade	Sharpe ratio	Profitable trades(%)	Max. drawdown (%)
EUR/USD	20 days	108	+0,12 %	0,23	69,4 %	-13,8 %
EUR/USD	60 days	84	+0,06 %	0,05	69,0 %	-19,5 %
EUR/USD	120 days	50	+0,17 %	0,07	64,0 %	-21,4 %
USD/JPY	20 days	103	-0,01 %	-0,01	65,0 %	-27,3 %
USD/JPY	60 days	68	+0,73 %	0,57	78,6 %	-12,1 %
USD/JPY	120 days	42	+0,41 %	0,16	71,4 %	-13,7 %

Optimal MA = MA with highest Sharpe ratio for given instrument

Derivation of Predictive Models

Model 1: Simple

$$\text{MA}_{\text{opt}} = 3 / \sigma$$

Application: Low volatility

Range: $\sigma < 0.10$ (~10% p.a.)

Constant K: ≈ 3.0

Example

EURUSD: $\sigma = 0.066$

$\text{MA} = 3 / 0.066 \approx 45$

Model 2: Linear

$$\text{MA}_{\text{opt}} = a \times \sigma - b$$

Where $a = 278.2$; $b = 21.1$

Application: Medium to high volatility

Range: $\sigma \geq 0.10$

R²: ≈ 0.94

Example

EURNZD: $\sigma = 0.438$

$\text{MA} = 278.2 \times 0.438 - 21.1 \approx 100.75$

Comparison of Predictive Models

σ (Volatility)	MA based on $3/\sigma$	MA based on $278.2 \cdot \sigma - 21.1$
0.05 (5 %)	60	-7
0.10 (10 %)	30	6
0.20 (20 %)	15	34
0.50 (50 %)	6	118
1.00 (100 %)	3	257

Recommended Approach

$\sigma < 0.10$: Use Model 1 ($3/\sigma$)

$\sigma \geq 0.10$: Use Model 2 (linear)

Model 1 underestimates MA for high volatility, Model 2 unreliable for $\sigma < 0.05$

Practical Benefit

Eliminates need for extensive testing of hundreds of parameter combinations – just measure volatility

Conclusions

Main Findings

Hypothesis confirmed – low-volatility markets more stable. Derived two predictive models ($3/\sigma$ for $\sigma < 0.10$, linear for $\sigma \geq 0.10$). Mean reversion strategy profitable across markets with proper calibration.

Practical Benefits

Fast MA estimation without testing, eliminates overfitting

Scientific Contribution

Generalized formula, empirical confirmation, $R^2 = 0.94$

Future Research

Multi-factor models, EMA, dynamic ML adaptation

Thank You

Questions?

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