

# Assessing the Efficacy of Kaufman's Adaptive Moving Average on Trading the S&P 500 Index

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2. Data
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5. Conclusion



# Introduction

# Moving Averages

Ordinary:

- Simple Moving Average (SMA)

Adaptive:

- Kaufman's Adaptive Moving Average (KAMA)

Weighted:

- Sinus Weighted Moving Average (SWMA)
- Weighted Moving Average (WMA)
- Exponential Moving Average (EMA)
- Double Exponential Moving Average (DEMA)
- Zero-lag Moving Average (ZMA)

# Simple Moving Average (SMA)

$$SMA_i = \sum_{j=i-n+1}^i w_j X_j, w_j = \frac{1}{n}$$



$n$  : number of periods

$w_j$ : weight factor for day  $j$

$X_j$ : price on day  $j$

# Sinusoidal Weighted Moving Average (SWMA)

$$SWMA_i = \sum_{j=i-n+1}^i w_j X_j, w_j = \frac{\sin\left(\frac{180}{6}i\right)}{\sum_{j=i-n+1}^i \sin\left(\frac{180}{6}i\right)}$$



# Exponential Moving Average (EMA)

$$EMA_i = \sum_{j=i-n+1}^i w_j X_j, w_j = \frac{(1 - \alpha)^{j-i}}{\sum_{k=i-n+1}^i (1 - \alpha)^{k-i}}$$

Or, iteratively,

$$EMA_n(X)_i = \alpha X_i + (1 - \alpha) EMA_n(X)_{i-1}, \alpha = \frac{2}{n+1}.$$

$\alpha$ : smoothing factor

$n$  : number of periods

$X_i$ : price on day  $i$



# Weighted Moving Average (WMA)

$$WMA_i = \sum_{j=i-n+1}^i w_j X_j, w_j = \frac{j-i}{n}$$



$n$  : number of periods

$X_i$ : price on day  $i$



# Double Exponential Moving Average (DEMA)

$$DEMA_n(X) = 2 \times EMA_n(X) - EMA_n(EMA_n(X)),$$



$n$  : number of periods

$X_i$ : price on day  $i$

# Zero-Lag Exponential Moving Average (ZLEMA)

$$\text{ZLEMA} = \text{EMA of } (\text{close} + (\text{close} - \text{close}[n]))$$



$n$  : number of periods

$X_i$ : price on day  $i$

# Kaufman's Adaptive Moving Average (KAMA)

Efficiency ratio:

$$\epsilon = \frac{|X_i - X_{i-n}|}{n \sum_{i=1}^n |X_n - X_{n-1}|}$$

Smoothing constant:

$$c = \left[ \epsilon \left( \frac{f}{f+1} - \frac{s}{s+1} \right) + \frac{s}{s+1} \right]^2$$

KAMA:

$$KAMA_i = KAMA_{i-1} + c(X_i - KAMA_{i-1})$$

$X_i$ : price on day  $i$

$\epsilon$ : efficiency ratio

$c$ : smoothing constant

$KAMA_i$ : KAMA value on day  $i$

$n$ : number of periods for the efficiency ratio

$s$ : number of periods for the slowest EMA

$f$ : number of periods for the fastest EMA

2

30

10

Kaufman's Suggested Values



# Research Questions

1. Can KAMA generate timely trading signals to generate profits in trading S&P 500 Index?
2. What is the optimal set of parameter values of KAMA for generating profitable trades?
3. How to design an objective function for optimizing the parameter values of KAMA that can maximize the risk-adjusted return and prevent overtrading and undertrading?



# Methodology

# Data

## **S&P 500 Index**

Period: 2000-01-03 ~ 2021-11-09 (21 years)

Data: Log of close price

Frequency: Daily



# Assumptions

1. All entry and exit prices are on the next trading day of the signal at the close.
2. Commission fees and taxes are assumed to be 5% of the total trading amount.
3. Limit and stop order options are not allowed at this stage.
4. Risk-free rate is assumed to be zero.



# Trading Strategy

## Signals:

**if**  $KAMA(today) / KAMA(n \text{ days ago}) - 1 > uptrend\_threshold$  **then**

Generate a buy-signal

**else if**  $KAMA(today) / KAMA(n \text{ days ago}) - 1 < downtrend\_threshold$  **then**

Generate a sell-signal





# Trading Strategy

## Enter/exit actions:

**if** *a buy-signal is generated n days ago* **then**

Buy at today's close

**else if** *a sell-signal is generated n days ago* **then**

Sell at today's close



# Training and Test Set

**Train-Test Split = 8 : 2**

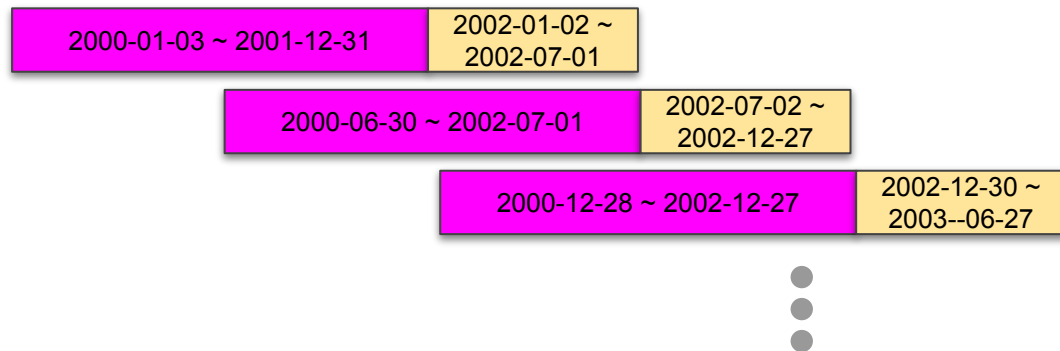
Timeframe	Training (days)	Test (days)	Total (days)	Total (years)
A	200	50	250	~ 1
B	500	125	625	~ 2
C	1000	250	1250	~ 5



# Training and Test Set

## Sliding Window of Horizon

*eg Timeframe B: 500 days (Training) + 125 days (Test)*



# Parameters to be Optimized

```
bounds = [(3, 20),  
          (2, 5),  
          (10, 100),  
          (0, 0.1),  
          (-0.1, 0),  
          (2, 5)]
```

## 1. KAMA

- a. Number of periods for the efficiency ratio ( $n_{delta}$ )
- b. Number of periods for the fastest EMA ( $n_{fast}$ )
- c. Number of periods for the slowest EMA ( $n_{slow}$ )

## 2. Threshold values

- a. Uptrend ( $delta\_lim\_up$ )
- b. Downtrend ( $delta\_lim\_dn$ )
- c. Number of days for trend determination ( $delta\_lim\_len$ )

# Objective Functions

Maximize:

1.  $r$

2.  $r \times w$

3.  $r \times \left( \frac{w + \epsilon}{l + \epsilon} \right)^{\text{sgn}(r)}, \epsilon = 1$

4.  $r \times \left( \frac{w + \epsilon}{l + \epsilon} \right)^{\text{sgn}(r)}, \epsilon = 10^{-6}$

$r$  : return

$w$  : number of winning trades

$l$  : number of losing trades

5.  $r \times (w - l) \times I,$

where  $I = -1$  if  $r < 0$  and  $w < l$ , or otherwise,  $1$ .

6.  $r \times (w^2 - l^2) \times I,$

where  $I = -1$  if  $r < 0$  and  $w < l$ , or otherwise,  $1$ .

# Note:

Directly maximizing Sharpe or Sortino ratios raises errors during training due to zero-trade issue.

# Optimization Method

## Dual Annealing

- `scipy.optimize.dual_annealing` in Python SciPy module
- Max. number of global search iterations (maxiter): 500
- Seed = 1
- Other input values: default  
(`initial_temp=5230.0`, `restart_temp_ratio=2e-05`, `visit=2.62`, `accept=-5.0`, `maxfun=10000000.0`, `no_local_search=False`, `callback=None`, `x0=None`)



# Metrics for Trading Performance Evaluation

1. Sharpe ratio
2. Sortino ratio
3. Return



# Results and Discussion



# Objective Functions

Maximize:

1.  $r$

2.  $r \times w$

3.  $r \times \left( \frac{w + \epsilon}{l + \epsilon} \right)^{\text{sgn}(r)}, \epsilon = 1$

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# Note:

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# Training and Test Set

**Train-Test Split = 8 : 2**

Timeframe	Training (days)	Test (days)	Total (days)	Total (years)
A	200	50	250	~ 1
B	500	125	625	~ 2
C	1000	250	1250	~ 5



# Highlighted Results



Result 1:


Trading Performance  
- Sharpe Ratio

# Objective Functions

Maximize:

1.  $r$

2.  $r \times w$

 3.  $r \times \left( \frac{w + \epsilon}{l + \epsilon} \right)^{\text{sgn}(r)}, \epsilon = 1$

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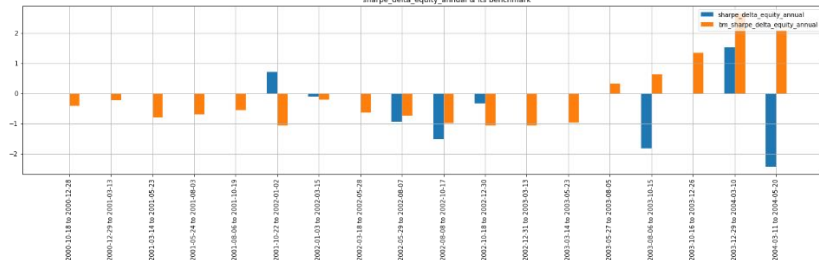
# Note:

Directly maximizing Sharpe or Sortino ratios raises errors during training due to zero-trade issue.

# Sharpe (200+50 days)

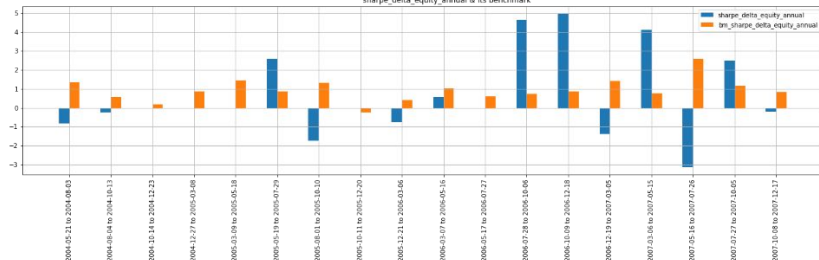
[Window of horizon: 250 days]  
[Objective Function ID: 13]  
[From 2000-10-18 to 2004-05-20]

sharpe\_delta\_equity\_annual & its benchmark



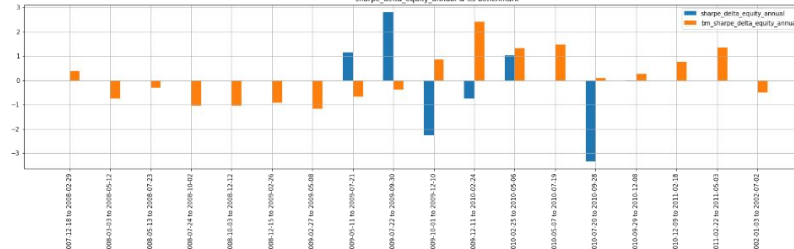
[Window of horizon: 250 days]  
[Objective Function ID: 13]  
[From 2004-05-21 to 2007-12-17]

sharpe\_delta\_equity\_annual & its benchmark



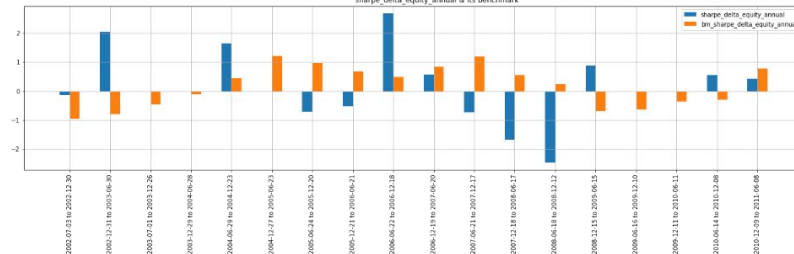
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[From 2007-12-18 to 2002-07-02]

sharpe\_delta\_equity\_annual & its benchmark



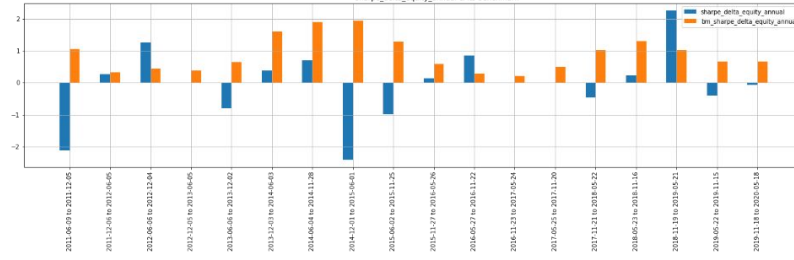
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[From 2002-07-03 to 2011-06-08]

sharpe\_delta\_equity\_annual & its benchmark

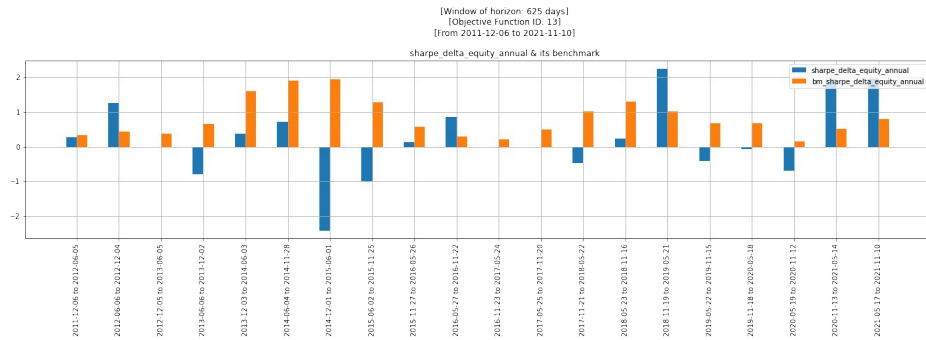
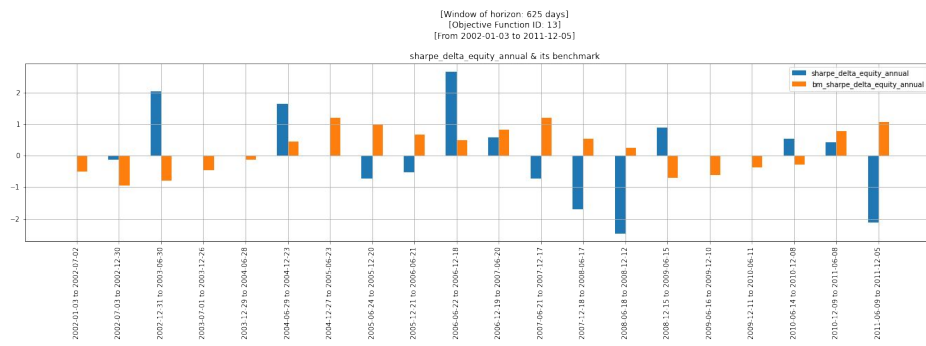


[Window of horizon: 250 days]  
[Objective Function ID: 13]  
[From 2011-06-09 to 2020-05-18]

sharpe\_delta\_equity\_annual & its benchmark

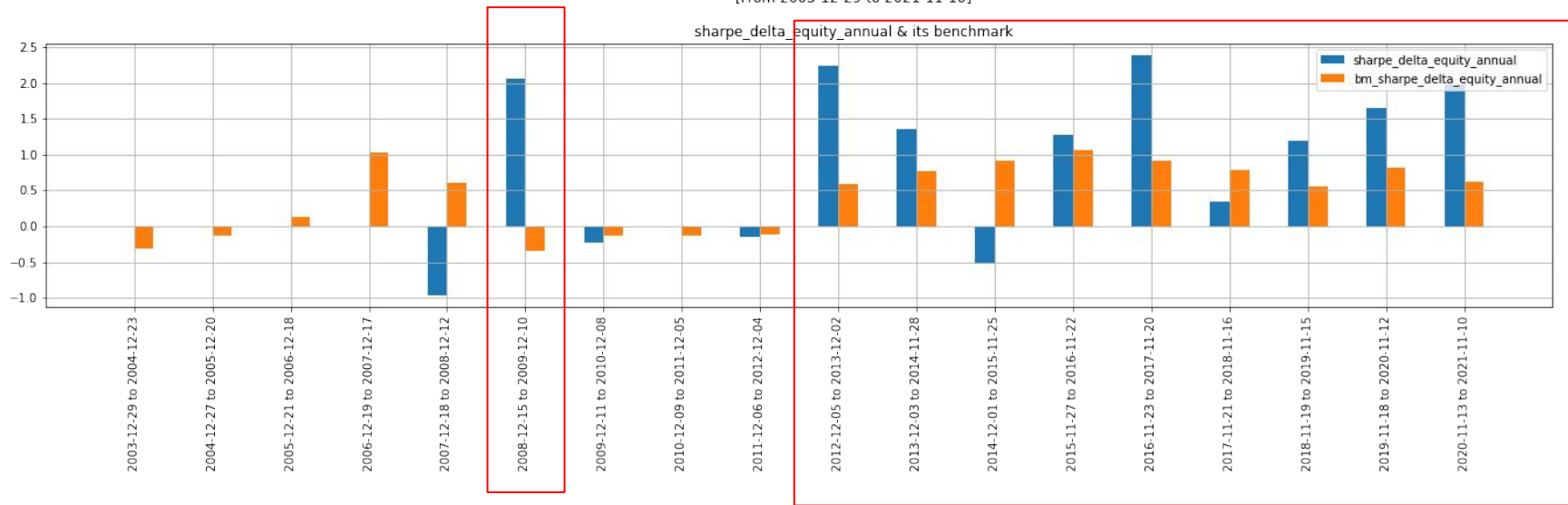


# Sharpe (500+125 days)



# Sharpe (1000+250 days)

[Window of horizon: 1250 days]  
[Objective Function ID: 13]  
[From 2003-12-29 to 2021-11-10]





# Objective Functions

Maximize:

1.  $r$

2.  $r \times w$

3.  $r \times \left( \frac{w + \epsilon}{l + \epsilon} \right)^{\text{sgn}(r)}, \epsilon = 1$

4.  $r \times \left( \frac{w + \epsilon}{l + \epsilon} \right)^{\text{sgn}(r)}, \epsilon = 10^{-6}$

$r$  : return

$w$  : number of winning trades

$l$  : number of losing trades



5.  $r \times (w - l) \times I,$

where  $I = -1$  if  $r < 0$  and  $w < l$ , or otherwise, 1 .

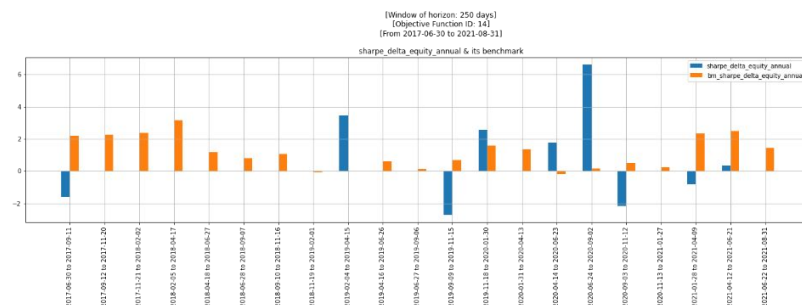
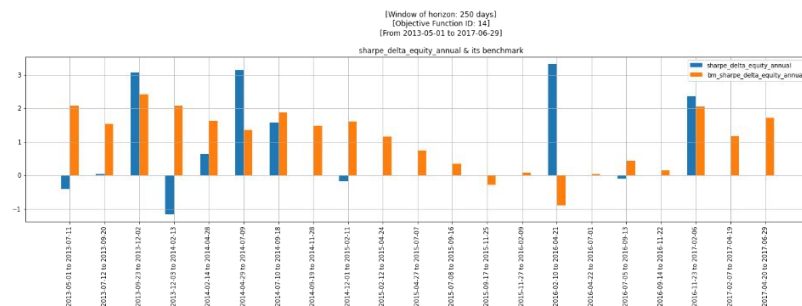
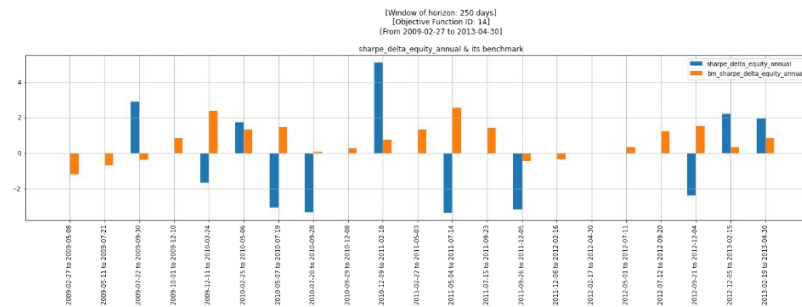
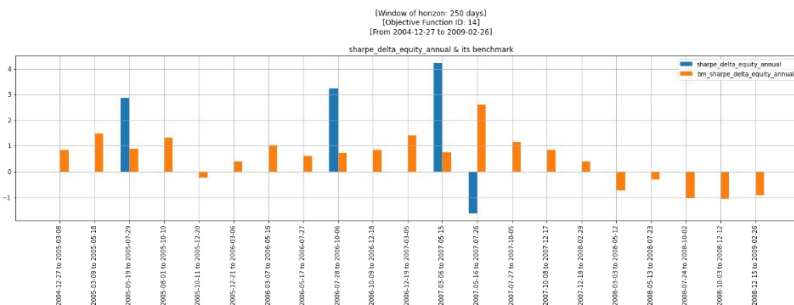
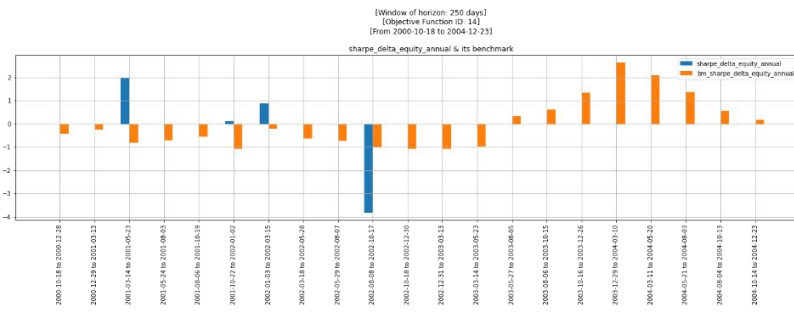
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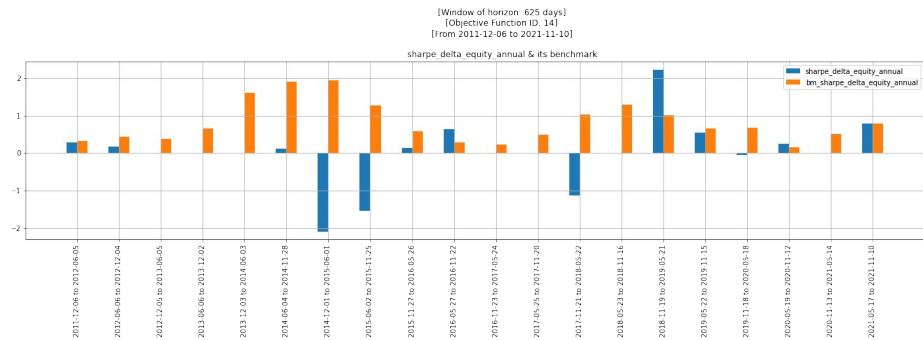
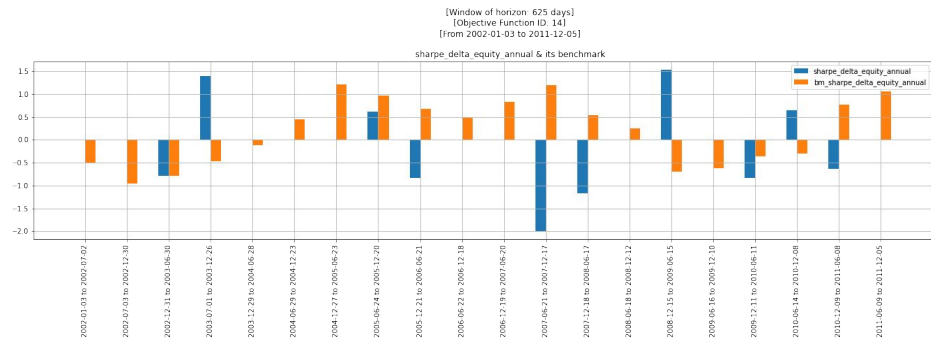
# Note:

Directly maximizing Sharpe or Sortino ratios raises errors during training due to zero-trade issue.

# Sharpe (200+50 days)



# Sharpe (500+125 days)

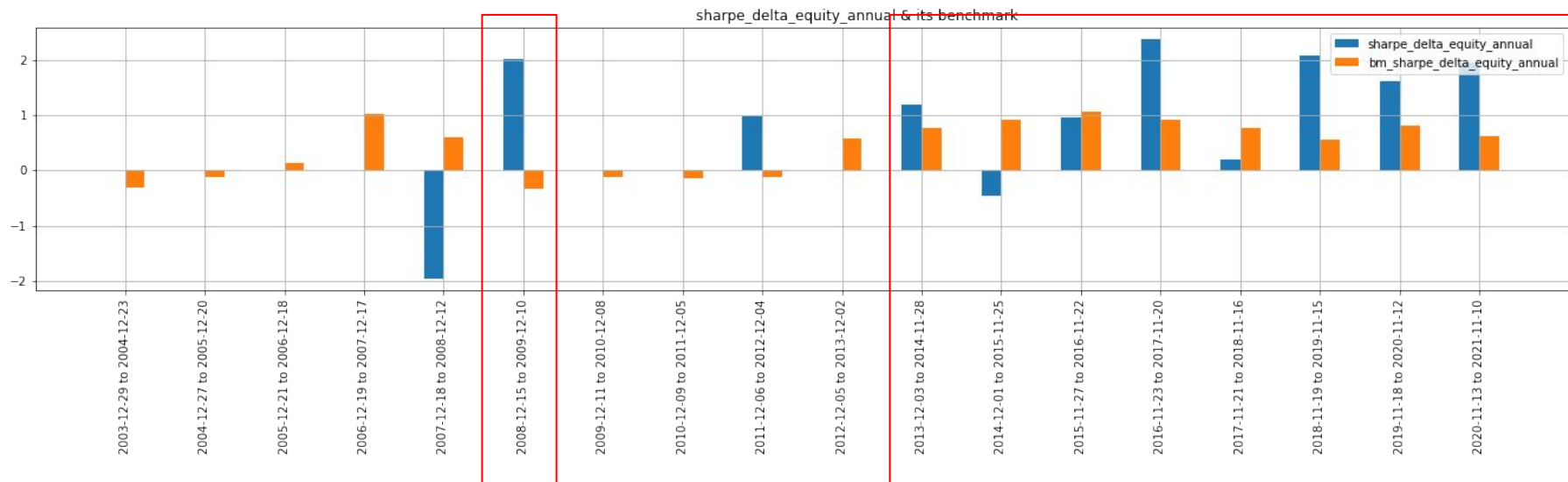


# Sharpe (1000+250 days)

[Window of horizon: 1250 days]

[Objective Function ID: 14]

[From 2003-12-29 to 2021-11-10]





Result 2:

Parameters of KAMA

# Parameters

## 1. KAMA

- Number of periods for the efficiency ratio ( $n\_delta$ )
- Number of periods for the fastest EMA ( $n\_fast$ )
- Number of periods for the slowest EMA ( $n\_slow$ )

	$n\_delta$	$n\_fast$	$n\_slow$	$delta\_lim\_up$	$delta\_lim\_dn$	$delta\_lim\_len$
count	1245.000000	1245.000000	1245.000000	1245.000000	1245.000000	1245.000000
mean	7.849032	3.005112	39.095305	0.008596	-0.029199	3.192675
std	3.850146	0.873958	29.246344	0.011540	0.035967	0.955119
min	3.001896	2.000721	10.010294	0.000002	-0.099981	2.003630
25%	4.401306	2.260350	11.088855	0.000776	-0.076729	2.277016
50%	7.522795	2.566505	29.374803	0.002696	-0.004983	2.924820
75%	10.089374	4.160973	63.722799	0.009125	-0.001579	4.103031
max	19.950902	4.991453	99.583507	0.030233	-0.000007	4.999012

## 2. Threshold values

- Uptrend ( $delta\_lim\_up$ )
- Downtrend ( $delta\_lim\_dn$ )
- Number of days for trend determination ( $delta\_lim\_len$ )

Kaufman's      10      2      30

# Conclusion

# Research Questions

1. Can KAMA generate timely trading signals to generate profits in trading S&P 500 Index?

*Longer time frame (eg 1000+250 days) can potentially achieve that.*





# Research Questions

2. What is the optimal set of parameter values of KAMA for generating profitable trades?

*The optimal set varies with market condition.*

*It can be slightly or significantly different from the set suggested by Kaufman.*



# Research Questions

3. How to design an objective function for optimizing the parameter values of KAMA that can maximize the risk-adjusted return and prevent overtrading and undertrading?

*Sharpe ratio and Sortino ratio may be appropriate for this purpose.*

*Further study on how to implement or modify the objective functions is needed due to zero-trade issue.*

