

Analysis of price-tracking Kandle

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We explore two ideas in this note:

- extracting the deterministic impermanent loss term from the (last price, return) joint distribution of a price-tracking Kandle
- defining a closed-formula (dependent on number of crossings) for the return of a price-tracking Kandle

1 Extracting the deterministic component

Idea: For price-tracking Kandle, as for all variants, there is a strong correlation between the last price and the pay-off (or return). In other words so-called ‘impermanent loss’ dominates the return.

One can ‘neutralise’ the price-deterministic part of the pay-off by adding a correction using the regression line with slope b , intercept c .

Write priceChange for the relative price change $\text{lastPrice}/\text{initialPrice} - 1$:

$$(\text{priceChange}, \text{return}) \longrightarrow (\text{lastPrice}, \text{return} - (b \times \text{priceChange} + c))$$

If $b = 1/2$, $c = 0$ we are subtracting the price-deterministic return of the buy-and-hold where half of the initial cash is invested in the underlying. The exact offsets b , c , to be determined by fitting a regression line to a numerical experiment depend on (r, σ) .

Concretely that can be done eg by shorting a certain amount b of the underlying on a perp market, or by borrowing some of the underlying on a lending platform. What does c represent: the deterministic cost of financing the strategy (a sort of premium)?

For large ratios $r > 1.1$ (aka grid steps), there are < 1 crossings on average and the constant is very nearly $b = \frac{1}{2}$ (ie a complete buy-and-hold). As r decreases, the mean number of crossings increases and the constant b, c seem to decrease. To investigate empirically.

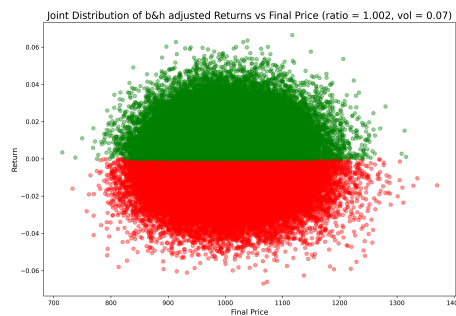


Figure 1: *adjusted return vs exit price (ratio = 1.002, vol = 0.07)*

QS1 - Is there for given σ a value $r^*(\sigma)$ of r that maximises the mean return? [similar to prior work done for plain Kandle]

Todo: Try same experiments with multiple price points (not just 2)

Can we combine sim with reset with a stop loss? Eg, is it useful to stop resetting after k moves, or a certain minimum residual capital (or on a momentum signal)?

2 Analysis of sim reset

Let $\alpha \in [0, 1]$ be the cash ratio (amount of initial capital kept in quote). Set $\alpha' := 1 - \alpha$ to simplify notations.

Proposition 1 *The total gain under a given price run is:*

$$\Gamma(\alpha) = (\alpha + \alpha' r)^{n_u} (\alpha + \alpha' / r)^{n_d} \quad (1)$$

where n_u is the number of up-crossings, n_d is the number of down-crossings along the price trajectories.

Proof: We start with a capital v_B^0 of cash (base), of which we keep αv_B^0 , and buy $\alpha' v_B^0 / p_0$ of base.

Suppose an uc:

$$\begin{pmatrix} \alpha' v_B^0 / p_0 \\ \alpha v_B^0 \end{pmatrix} \xrightarrow{c} \begin{pmatrix} 0 \\ \alpha v_B^0 + (r p_0) \alpha' v_B^0 / p_0 \end{pmatrix}$$

Our new position is worth $v_B^0(\alpha + \alpha' r)$, so a growth rate $\gamma_u = \alpha + \alpha' r$.

Symmetrically for a dc:

$$\begin{pmatrix} \alpha' v_B^0 / p_0 \\ \alpha v_B^0 \end{pmatrix} \xrightarrow{d} \begin{pmatrix} \alpha' v_B^0 / p_0 + \alpha v_B^0 / (p_0 / r) \\ 0 \end{pmatrix}$$

with a new position worth $v_B^0(\alpha' / r + \alpha)$, and we find $\gamma_d = \alpha + \alpha' / r$. \square

NB - Importantly, in the proof above, we assume that on a down-crossing, the price at which one can resell what was just bought is the same price at which we have bought p_0 / r . In reality, on a down-crossing the resell price may be lower: $p_n > p_0 / r \geq p_{n+1}$. There is a loss factor proportional to $p_{n+1} / (p_0 / r) < 1$ and we get $\hat{\gamma}_d = \gamma_d p_{n+1} / (p_0 / r) \leq \gamma_d < 1$ instead. Symmetrically, one could have sold higher on an upcrossing with a “manque à gagner” proportional to $p_{n+1} / (r p_0)$. In the simulations we do mark to market at the real price p_{n+1} . So we get:

$$\begin{aligned} \log \hat{\Gamma} &= N_u \log(\alpha + \alpha' r) + N_d \log(\alpha + \alpha' / r) + \ell(\mathbf{p}, r) \\ \ell(\mathbf{p}, r) &= \sum_{i \downarrow} \log(p_{i+1} / (p_0(i) / r)) \end{aligned}$$

where $p_0(i)$ is the last touched price, ie the mid-point of the strategy right before the i th jump downwards.

This brings a series of remarks:

(1) The smaller the integration step δt , the smaller the loss factor. We should therefore incorporate δt as a bona fide parameter $\ell(\mathbf{p}, r, \delta t)$. In the limit where $\delta t \rightarrow 0^+$, $\log(p_{i+1} / (p_0(i) / r)) \rightarrow 0^-$, which is paradoxical! In that case δt has to be understood not just as an integration step, but as a market reactivity time or a lag vs a driving price (eg a fee, maybe check the LvR paper to get a model of the lag [1])

(2) one could always renege on bids (for a cost q), to mitigate the loss, and just re-sell the remaining inventory [a behaviour which corresponds to $\alpha = 0$?]

(3a) we can also forward the bids (if we do not want to renege) but we are going to pay the same $\log(p_{i+1}/(p_0(i)/r))$ or about where p_{i+1} is now the price on the source dex, or

(3b) re-sell instantly on an other dex (the semi-book is closed for execution) [we may want to generate the volume]

Note that the crossing numbers are random and only depend on vol-and-grid, σ and r .

As $\alpha + \alpha'r > 1$ (barycenter of 1 and $r > 1$), and $\alpha + \alpha'/r < 1$ (barycenter of 1 and $1/r < 1$), it follows that $\Gamma > 1$ ie the strategy is profitable along a given price trajectory as soon as $n_u > n_d$.

In particular if $\alpha = \alpha' = \frac{1}{2}$, we get:

$$\begin{aligned}\log(\Gamma(1/2)) &= n_u \log(\frac{1+r}{2}) + n_d \log(\frac{1+1/r}{2}) \\ &= N(f_u \log(\frac{1+r}{2}) + f_d \log(\frac{1+1/r}{2}))\end{aligned}$$

where in the last line we assume $N = n_u + n_d > 0$ (not always true! especially for large ranges).

The above is > 0 iff:

$$f_d/f_u \leq \frac{\log(1+r) - \log 2}{\log 2 - \log(1+r^{-1})} =: k(r)$$

where $k(r) \geq 1$ is increasing and converges to $\log(1+r)/\log(2) - 1$ for large r . This suggests that larger r s are more likely to generate profits. To make sure we need to study how the f_d, f_u distribution depends on r . It seems unlikely but it could be that larger r are biased downwards, ie f_d/f_u is increasing function of r . [in sims up and down crossings look equally likely]

Todo: study the distribution of f_u as a function of r and σ . [Also look at std]

We can see the effect of changing α .

If $\alpha = 0$ (base only), $\Gamma(0) = r^{n_u - n_d}$. If we go full base, we win iff the process crosses more up than down.

If $\alpha' = 0$ (quote only), $\Gamma(1) = 1$. Indeed if we never buy the underlying, all that can happen is the execution of a bid, the output of which that strategy re-sells immediately at same price (assuming no fees, no slippage), and therefore the initial wealth never changes. Ie (modulo fees and slippage), setting $\alpha = 0$ amounts to not playing.

3 Examples

The cost of financing a price-tracking Kandle (intercept) with $r = 1.005$, $\sigma = 0.06$ is a negative return of $\sim 4.6\%$. See the comparison with the B&H line which has an intercept of zero (Fig. 2). The same cloud of points is represented with the regression line subtracted Fig. 3.

It would be interesting to understand how the parameters of the regression line depend on r and σ .

Fig. 4 gives the distribution of adjusted returns. From the figure it seems one can decompose the return into a random normal component and a deterministic one :

$$ret(r, \sigma) = \mathcal{N}(0, s(r, \sigma)) + (c(r, \sigma) + b(r, \sigma)(\Delta p/p_0)) \quad (2)$$

where $\Delta p/p_0$ is the relative price change at the end of the price sequence.

Todo: explore the coefficients in (Eq 2) above.

4 Random remarks

A small remark: concretely, when running on chain, price-tracking Kandle does not have access to the next price (as the simulation does) only that $p_{n+1} > rp_n$; to correct this one can use multiple offers to improve the upper bound, and/or use an oracle at update time (eg consult another liquidity source).

Euler-Maruyama for large \sqrt{dt} , large σ can lead to negative prices in a GBM; so better use the integral form.

Why is Γ an easier notation for growth of mtm:

$$\begin{aligned} \Gamma &= \prod_i \gamma_i \\ R &= \prod_i (1 + r_i) - 1 \\ r \star s &= r + s + rs \end{aligned}$$

if r, s are small then rs is negligible which explains why people talk with returns.

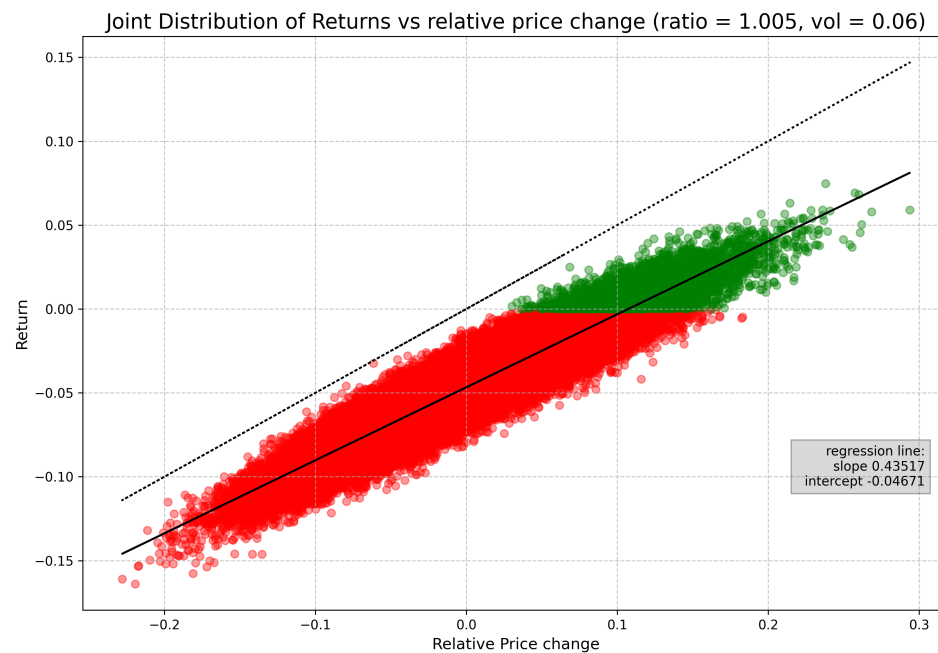


Figure 2: *non-adjusted returns vs relative price change ($r = 1.005$, $\sigma = 0.06$); the dotted line is the buy-and-hold return for half the capital invested in base ($\alpha = 1/2$); the intercept of the regression line is the cost of the strategy per unit of capital provided.*

References

- [1] Jason Milionis, Ciamac C Moallemi, Tim Roughgarden, and Anthony Lee Zhang. Automated market making and loss-versus-rebalancing. [arXiv preprint arXiv:2208.06046](https://arxiv.org/abs/2208.06046), 2022.

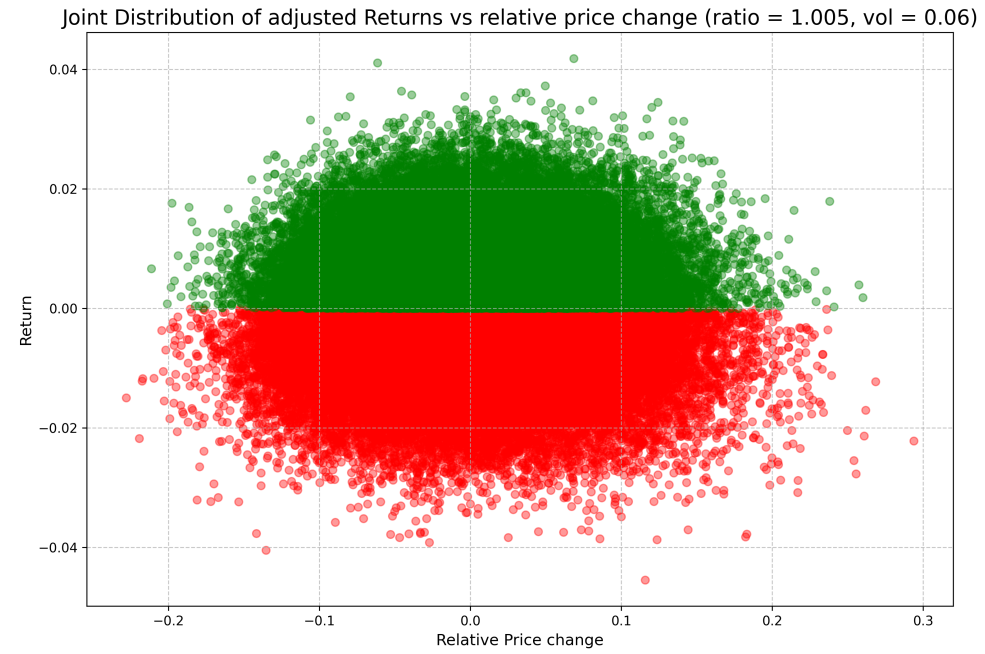


Figure 3: *adjusted returns vs relative price change* ($r = 1.005$, $\sigma = 0.06$).

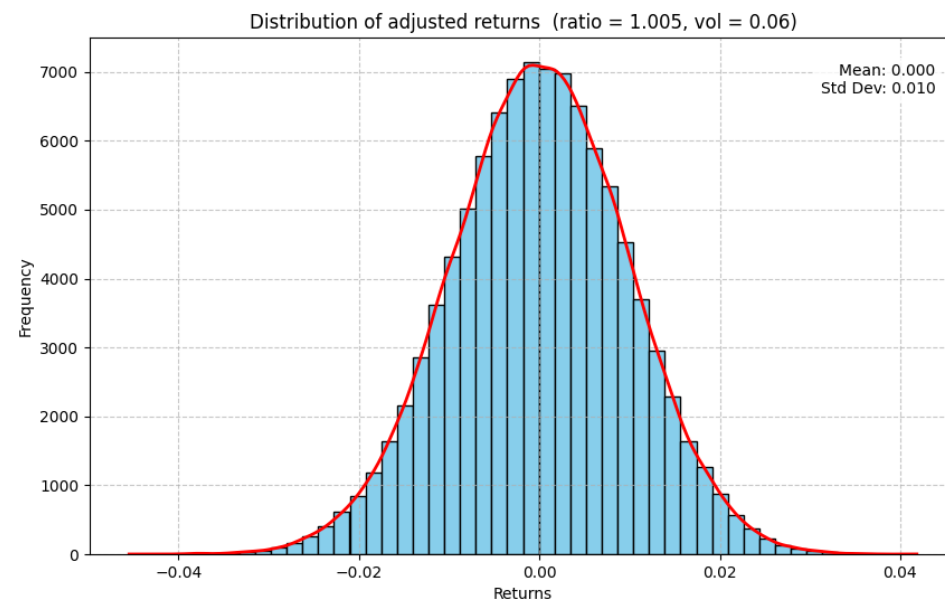


Figure 4: *histograms of adjusted returns* ($r = 1.005$, $\sigma = 0.06$).