

# How the Scalar function works

the main problem of the data generated by the model is the scale of the data itself. As we can see in the following picture the series seems realistic if we just look at the trend but the magnitude of the data is totally wrong. To avoid this I've created an algorithm to force the series to the real data range.

## Introduction

The scalar aims to scale the time series without changing the trend. The idea is that when we have a peak in the time series of the return of a stock the role of the peak depends by the distance of the peak itself from the second highest peak in the series. For example let's take the following series:

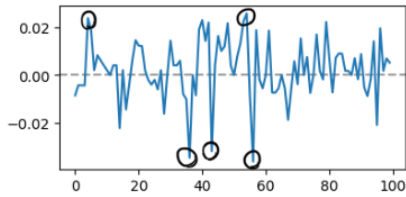


Figure 1

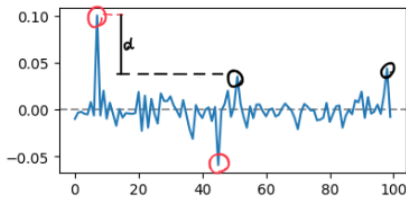


Figure 2

As we can see in (Figure 1) the 2 highest peaks are not very far from each-other and that makes them lower in terms of magnitude compared to the peak we

can see in the (Figure 2). This because a peak can have 2 different meanings in a time series:

- a strong movement in the price but still not that unusual
- a shock, due to news or particular events.

is very unlikely to have two big shocks in the same window. Following this logic when the 1<sup>st</sup> and 2<sup>nd</sup> peaks in a series are close they are probably just strong movements in the price. On the other side when the 1<sup>st</sup> peak in terms of magnitude is way bigger than the 2<sup>nd</sup> one probably It's a shock so these types of shocks will have bigger values.

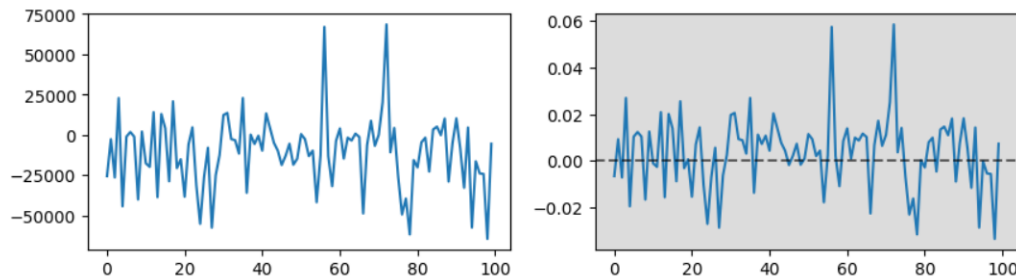
## Scalar Logic

Based on the above reasoning the Scalar aims assign to "lonely" peaks in the generated series the value that usually shocks have in the real data and assign to close peaks value that are typical of strong market movements.

## Scalar Implementation

To do it the scalar compute the quantiles of the max values distribution and same thing for the distribution of the difference between the 1<sup>st</sup> and 2<sup>nd</sup> peaks. Then the algorithm take the sample and see where the distance between the 1<sup>st</sup> and 2<sup>nd</sup> peaks are among the quantiles of the differences for real data. Then pick a random value from a uniform distribution where the lower and upper boundaries are the corresponding quantiles in the max distribution. Same thing for the negative peaks.

## Example:



In the above example we can see on the left the series generated by our model and on the right the series after we apply the Scalar. As we can see the scalar preserves the pattern generated by our model and just rescale the data so we have realistic data.