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The Recursive Future Equation Based On The Ananda-Damayanthi Normalized Similarity Measure. {File Closing Version 2}. ISSN 1751-3030

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Technical Note

Abstract

In this research Technical Note the author have presented a Recursive Future Average Of A Time Series Data Based on Cosine Similarity.

Theory

The Recursive Future Average Of A Time Series Data Based on Cosine Similarity can be given by the following methods:

Method 1:

$$y_{n+1} = \frac{\sum_{i=1}^n (y_i) \{CS(y_i, y_{n+1})\}}{\left\{ \sum_{i=1}^n (\{CS(y_i, y_{n+1})\})^2 \right\}^{1/2}}$$

where $CS(y_i, y_{n+1}) = \left\{ \frac{\text{Smaller of } (y_i, y_{n+1})}{\text{Larger of } (y_i, y_{n+1})} \right\}$

when the Time Series Data is of the kind

$$S = \{y_1, y_2, y_3, \dots, y_{n-1}, y_n\}$$

Method 2:

$$y_{n+1} = \frac{\sum_{i=1}^n (y_i) \{CS(y_i, y_{n+1})\} \{CS(y_i, y_{n+1})\}}{\sum_{i=1}^n \{CS(y_i, y_{n+1})\}}$$

where $CS(y_i, y_{n+1}) = \left\{ \frac{\text{Smaller of } (y_i, y_{n+1})}{\text{Larger of } (y_i, y_{n+1})} \right\}$

when the Time Series Data is of the kind

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$$S = \{y_1, y_2, y_3, \dots, y_{n-1}, y_n\}$$

Deriving motivation from this concept, we further extend this formula using [1] as

$$y_{n+1} = \frac{\left(\sum_{i=1}^n (y_i) \{CS(y_i, y_{n+1})\} + \sum_{i=1}^n (^1 y_i) \{CS(^1 y_i, y_{n+1})\} + \sum_{i=1}^n (^2 y_i) \{CS(^2 y_i, y_{n+1})\} + \dots + \sum_{i=1}^n (^r y_i) \{CS(^r y_i, y_{n+1})\} \right)}{\left(\sum_{i=1}^n \{CS(y_i, y_{n+1})\}^2 + \sum_{i=1}^n \{CS(^1 y_i, y_{n+1})\}^2 + \sum_{i=1}^n \{CS(^2 y_i, y_{n+1})\}^2 + \dots + \sum_{i=1}^n \{CS(^r y_i, y_{n+1})\}^2 \right)^{1/2}}$$

$$\text{where } ^1 y_i = \frac{\left\{ y_i y_{n+1} - \frac{(\text{Smaller of } (y_i, y_{n+1}))^2}{y_i y_{n+1}} \right\}}{y_{n+1}} \text{ and}$$

$$^2 y_i = \frac{\left\{ ^1 y_i y_{n+1} - \frac{(\text{Smaller of } (^1 y_i, y_{n+1}))^2}{^1 y_i y_{n+1}} \right\}}{y_{n+1}}, \dots, \text{i.e., and so on, so forth}$$

$$^k y_i = \frac{\left\{ ^{k-1} y_i y_{n+1} - \frac{(\text{Smaller of } (^{k-1} y_i, y_{n+1}))^2}{^{k-1} y_i y_{n+1}} \right\}}{y_{n+1}}$$

upto

$$^r y_i = \frac{\left\{ ^{r-1} y_i y_{n+1} - \frac{(\text{Smaller of } (^{r-1} y_i, y_{n+1}))^2}{^{r-1} y_i y_{n+1}} \right\}}{y_{n+1}} \text{ such that we can write}$$

$$y_{n+1} = \frac{\left\{ \sum_{i=1}^n \left\{ (y_i) \{CS(y_i, y_{n+1})\} + \sum_{k=1}^r (^k y_i) \{CS(^k y_i, y_{n+1})\} \right\} \right\}}{\left\{ \sum_{i=1}^n \left\{ \{CS(y_i, y_{n+1})\}^2 + \sum_{k=1}^r \{CS(^k y_i, y_{n+1})\}^2 \right\} \right\}^{1/2}}$$

where r is a number such that $^r y_i \rightarrow 0$.

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http://www.philica.com/display_article.php?article_id=1119

References

- 1.Bagadi, R. (2016). Proof Of As To Why The Euclidean Inner Product Is A Good Measure Of Similarity Of Two Vectors. *PHILICA.COM Article number 626*.
http://philica.com/display_article.php?article_id=626
- 2.<http://www.philica.com/advancedsearch.php?author=12897>
- 3.http://www.vixra.org/author/ramesh_chandra_bagadi