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Information Sciences xxx (2014) xxx-xxx

Contents lists available at ScienceDirect



Information Sciences

journal homepage: www.elsevier.com/locate/ins

## Fuzzy forecasting based on automatic clustering and axiomatic 3 fuzzy set classification

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### ARTICLE INFO

2 B 14 Article history:

15 Received 18 November 2012

16 Received in revised form 28 August 2014

- 17 Accepted 7 September 2014
- 18 Available online xxxx
- 19 Keywords:
- 20 Fuzzy forecasting
- 21 Fuzzy time series
- 22 Axiomatic fuzzy set (AFS) classification
- 23 Automatic clustering
- Trend prediction
- 24 25

## ABSTRACT

In spite of the impressive diversity of models of fuzzy forecasting, there is still a burning need to arrive at models that are both accurate and highly interpretable. This study proposes a new fuzzy forecasting model designed with the use of the two key techniques, namely clustering and axiomatic fuzzy set (AFS) classification. First, clustering algorithm is utilized to generate clustering-based intervals. Second, the fuzzy trend labeled training data set is constructed based on fuzzy logic relationships and fuzzy trends of historical samples. Then, the AFS classification is exploited to yield the semantic interpretation of each fuzzy trend. The main novelty is that the proposed model not only predicts the value but can also capture the trend prevailing in the time series, and obtain its semantic interpretation. The Taiwan Stock Exchange Capitalization Weighted Stock Index (TAIEX), inventory demand, and Spanish electricity prices are used in a series of experiments. The results show that the proposed model has both good interpretability and accuracy.

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#### 43 1. Introduction

44 Investigating the relations within the sequential set of past data and forecasting the future values are the main tasks of the time series. The forecasting problem of time series data plays an important role in various domains, such as finance 45 46 forecasting, population growth, rainfall prediction, and temperature forecasting [30–32,56]. It deals with forecasting future outcomes from a temporally ordered sequence of past observed data points, whose values are usually real numbers. 47 Traditional time series analysis cannot handle the vagueness and uncertainty inherent in certain data due to inaccuracies 48 49 in measurements, incomplete sets of observations, or difficulties in obtaining the measurements [43].

50 Recently, fuzzy logic has been widely recognized as a successful approach for dealing with data uncertainty. For time series, the uncertain values can be modeled as fuzzy variables, resulting in so-called fuzzy time series [43]. Song and Chissom 51 first introduced the theory of fuzzy logic into forecasting time series problems and proposed a new paradigm known as fuzzy 52 53 time series, capable of dealing with vague and incomplete data represented as linguistic values under uncertain circumstances [50–52]. They established a four-step framework to manipulate the forecasting problem: (1) determine and partition 54 55 the universe of discourse into intervals; (2) define fuzzy sets from the universe of discourse and fuzzify the time series; (3) derive fuzzy relationships existing in the fuzzified time series; and (4) forecast and defuzzify the forecasting outputs. 56

http://dx.doi.org/10.1016/j.ins.2014.09.027 0020-0255/© 2014 Elsevier Inc. All rights reserved.

Q1 Please cite this article in press as: W. Wang, X. Liu, Fuzzy forecasting based on automatic clustering and axiomatic fuzzy set classification, Inform. Sci. (2014), http://dx.doi.org/10.1016/j.ins.2014.09.027

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# **ARTICLE IN PRESS**

# **INS 11130** 9 October 2014

W. Wang, X. Liu/Information Sciences xxx (2014) xxx-xxx

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57 Since Song and Chissom's pioneering work, a number of related research works have been reported that follow their 58 framework. In the plethora of currently available models of fuzzy time series, their accuracy has been a holy grail of the over-59 all modeling for a long time. With the emergence of more visible need for interpretable models that are easily comprehended 60 by humans, arose an important need to develop models that are not only accurate but transparent as well.

61 In order to develop a interpretable model, this study exploits axiomatic fuzzy set (AFS) classification which can mimic the 62 human reasoning and in this manner offer a far more transparent and comprehensible way supporting the design of the classifier [35–39]. The proposed classifier consists of AFS fuzzy sets (each class is represented by a fuzzy set) and the degree of 63 64 the new sample belonging to the class is the membership degree of this sample belonging the fuzzy set representing this 65 class. Thus each class character is linguistically interpretable, comprehensible and similar to the classification schemes exer-66 cised by humans.

In this paper, a novel fuzzy forecasting model is presented. The proposed model first utilizes clustering to partition the 67 68 universe of discourse and generate clustering-based intervals. Then, the model applies fuzzy logical relationships and fuzzy trends of historical samples to construct the fuzzy trend labeled training data set and exploits the AFS classification to yield 69 70 its semantic interpretation. Finally, the forecasted data can be obtained by integrating the current state with the fuzzy trend.

- The advantages of the proposed model can be summarized as follows: 71
- When using fuzzy time series for forecasting, it is obvious that the length of intervals in the universe of discourse is 72 73 important due to the fact that it can affect the forecasting accuracy. However, most of the existing fuzzy forecasting models based on fuzzy time series used the static length of intervals, i.e., the same length of intervals 74 [1,3,6,8,20,22,25,27,29,33,40,41,45,46,48,49]. The drawback of the static length of intervals is that the historical data 75 76 are roughly put into the intervals, even if the variance of the historical data is not high. To overcome the drawback, clus-77 tering algorithm is applied to generate intervals.
- Fuzzy sets and fuzzy logic were introduced to capture the uncertainty and human perception. Comparing with the most of 78 existing models which employ "black box" structures without a comprehensive explanation facility for their predictions 79 80 [2,13,19,54,62], the proposed model can yield the semantic interpretation of each fuzzy trend (such as "upward", "unchanged" and "downward"). 81
- The proposed model can predict the fuzzy trend of forecasted data. This trend prediction will help the decision maker to 82 be extra cautious well in advance. 83

To validate the proposed forecasting model, we conduct three real-world experiments in forecasting the stock indices 85 [63], inventory demand [44] and the Spanish electricity prices time series [64]. The experimental results show the proposed 86 model yields higher average forecasting accuracy rates than the existing models [3,6,7,9,16,18,21,23-25,53,57,59-61]. 87

88 The paper is organized as follows. Section 2 overviews works related with fuzzy time series. Section 3 briefly reviews basic concepts of fuzzy time series. Section 4 proposes a novel forecasting model based on the automatic clustering and 89 the AFS classification. Section 5 compares the average forecasting accuracy rates of the proposed model with that of existing 90 91 models, where the historical data of the stock indices, inventory demand and the Spanish electricity prices data sets are used for the experiments. Section 6 concludes the paper. 92

#### 93 2. Related works

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Forecasting using fuzzy time series has been applied to various problem domains, such as the stock prices forecasting 94 95 [6,9,15], inventory demand forecasting [21,23,61], and temperature forecasting [8,41]. Traditional time series models are usually highly dependent on historical data, which can be incomplete, imprecise and ambiguous. These uncertainties are 96 97 likely to be widespread in real-world data and hinder forecasting accuracy, thus limiting the applicability of these models. Unlike traditional time series forecasting approaches, the fuzzy time series is capable of dealing with vague and incomplete 98 99 time series data under uncertain circumstances.

100 Based on the fuzzy time series theory, first forecasting model was introduced by Song and Chissom [50–52], which were used to forecast the time series values based on linguistic values. They presented the fuzzy time series model by means of 101 fuzzy relational equations involving max-min composition operation, and applied the model for forecasting the enrollments 102 in the University of Alabama. The model take a large amount of computation time since max-min operations are used for 103 reasoning. Chen [3] developed first-order fuzzy relationship rules with simplified arithmetic operations rather than the com-104 plicated max-min composition operations to forecast the enrollments of the University of Alabama with better accuracy 105 than the Song and Chissom's model. Later, many studies have focused on improving steps of Song and Chissom's framework 106 using Chen's models. The studies can be divided into three categories, (1) those that enhance the accuracy of model by 107 changing interval length in partition step of the universe of discourse, (2) those that establishment of fuzzy relationships 108 109 with better performance, and (3) those forecasting fuzzy time series by different approaches in forecasting and defuzzication 110 step, separately.

111 In the first category, Huarng [21] investigated the impact of the interval length on forecasting results and proposed two 112 heuristic approaches, i.e., distribution and average based, to determining the interval length. Chen and Hsu [5] used a two-113 phase partitioning method with the statistical distributions of historical data. Huarng and Yu [23] presented ratio-based Download English Version:

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