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**Cloud & Virtualization** 

Cluster-Based Network Traffic Prediction Pipeline For Big Data Time Series

WEI CAI

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### INTRODUCTION



#### Rationales

### Proposed clustering-based best-selection forecast pipeline

- Irregularity and more fluctuations
- Big data
- Similarity in traffic patterns across time series







# ARIMA

ARIMA models are designated by Autoregression, Integration and Moving average



### **ARIMA** structure

### **BACKGROUND: LEARNING MODELS**



### XGBoost

### XGBoost models combines weak learners to form a strong model through iterations



XGBoost structure (<u>https://blog.quantinsti.com/xgboost-python/</u>)



# LSTM

XGBoost models combines weak learners to form a strong model through iterations



#### LSTM structure (https://adventuresinmachinelearning.com/keras-lstm-tutorial/)



# Two major gaps

1. Very few research papers have employed a clustering approach into large scale network traffic forecasting workflow 2. Best selection from different models performs better than averaging the results from different models, particularly for network time series with more volatilities



# Raw data with missing polls Imputed clean data





### **Clean data with outliers**

# **Clean data with no outliers**





### **Clustering nodes to 4 clusters**

Cluster Label	Node Counts	
Cluster 0	10,082	
Cluster 1	3,92	
Cluster 2	71	
Cluster 3	1,193	



2020.02 2020.05

weekly timestamp

2020.09 2022.02 2022.05

<sup>507802</sub> <sup>507803</sup> <sup>507307</sup> <sup>507302</sup> <sup>507303</sup></sup>

5620

5615

5610

2018-01





- Grid search was used to automatically discover the optimal order of non-seasonal and seasonable parameters at a cluster level.
- Dataset is split into 70%, 20%, 10% training, test and validation sets, respectively.
- Stationarity of the series are checked.
- SARIMAX models are trained and obtained to make estimation on fresh test data.
- MAPE values for both training dataset and testing data set are calculated.

n sarima

forecasting

Procedures





- Grid search algorithm is used to optimize the parameters at a cluster level.
- Dataset is split into 70%, 20%, 10% training, test and validation sets, respectively.
- XBGoost models are trained and obtained to make estimation on fresh test data.
- MAPE values for both training dataset and testing data set are calculated.

XGBoost forecasting Procedures



- Scale data using MinMaxScaler to speed up the learning process and help model.
- Hyperparameter such as number of layers, layer depths, activation functions, dropout coefficients are repeatedly tuned at a cluster.
- Dataset is split into 70%, 20%, 10% training, test and validation sets, respectively.
- LSTM models are trained and obtained to make estimation on fresh test data.
- MAPE values for both training dataset and testing data set are calculated.

LSTM forecasting Procedures



Table 1 Training MAPE by Models					
Model		Training MAPE		Total	
MAPE Range	<= 5%	>=6% & <= 10%	>=11% & <= 15%	>=16%	
SARIMA	1,360	2,828	2,059	5,295	11,542 nodes
XGBoost	11,333	55	20	134	11,542 nodes
LSTM	3,573	2,979	2,909	2,081	11,542 nodes

Table 2 Test MAPE by Models					
Model		Test MAPE		Total	
MAPE Range	<= 5%	>=6% & <= 10%	>=11% & <= 15%	>=16%	
SARIMA	982	746	1,175	8,639	11,542 nodes
XGBoost	2,244	2,228	2,870	4,200	11,542 nodes
LSTM	2,299	3,996	2,753	2,494	11,542 nodes



Table 3 Node Counts <u>with &lt;= 10% MAPE by Models</u>		
Model	Node counts with good fit (Mape <= 10%)	Node counts with overfit (Mape <= 10%)
SARIMA	956	586
XGBoost	1,195	3,248
LSTM	2,129	2,278

Table 4 Node Counts MAPE by Models		
Model	Node counts with Mape <= 10%)	Node counts with Mape >10%)
SARIMA	101	495
XGBoost	139	457
LSTM	346	250



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