

#### Initiatives on AI/ML based Analytics Government eProcurement System of NIC, GePNIC<sup>©</sup> Government of India

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#### 6<sup>th</sup> Annual Asia Pacific Public Electronic Procurement Network Conference

3<sup>rd</sup> – 5<sup>th</sup> Nov 2020 (Virtual Conference)





#### Introduction

Initiatives to carryout Artificial Intelligence/Machine Learning (AI/ML) based analytics since June 2019.

Few select areas were identified for analysis, as good volume of relevant data is readily available

To start with, a sample set was restricted to 50,000 records for building the model

This model was then applied to data related to tenders published during the period from April 2016 to March 2019 for analysis



#### Focus Areas for AI/ML analytics

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Bidder Segmentation Analysis - based on tender value for which the bidders have participated - to provide focused services to bidders

Analysis on pattern of association - Bidder Participation trend analysis for pre-emptive Alerts

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Forecasting of Volume of Transactions - to envisage and provide Compute and Storage requirements

Invalid Names - Automatic detection of invalid names entered during user registration process





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#### **Bidder Segmentation Analysis**

a) Type of tenders they participate often

- b) Products for which they are bidding
- c) Category of Tenders
- d) Value of tenders

Helps us to segment each bidder to a particular category and this input can be used to for further analysis like send specific Alerts , Predicting participation %

By segmenting each bidder to a particular category, focused services can be planned

**Clustering Method is applied to group the Bidders in these categories** 

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## K-Prototype Clustering

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## Visualisation of Data Clusters

An Elbow shape is formed at the 5th cluster point. So, we choose k=5 as our optimal number of clusters.

The 5 bidder clusters formed can be interpreted as :

- Very low tender value
- Low tender value
- Medium tender value
- High tender value
- Very high value

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# Segmentation of Bidders amongst the various provinces/ implementations





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## Inference drawn from these Segmentation

Inference from the previous graph indicates a clear picture on participation pattern of Bidders in various Provinces - Chandigarh, Delhi, eProcure, J&K, Kerala, Maharashtra, Manipur, Tamil Nadu, UP are given below [ applied For Tender Value]

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#### **APRIORI**



The *Apriori* Algorithm was proposed by Agrawal and Srikant in 1994.



*Apriori* is an association rule mining algorithm used to identify frequent item sets.



#### The three important metrics of apriori are support, confidence and lift.

- **Support** Percentage of a categories that contains the item or item set.
- **Confidence** Percentage of times that category B is bidded, given that category A was bidded
- Lift Measure for the relationship between A and B



#### PRODUCT CATEGORY ASSOCIATION

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antecedents	consequents	support	confidence	lift
<pre>irozenset({'Computer- Data Processing', 'Miscellaneous lorks', 'Computer- H/W'})</pre>	<pre>frozenset({'Miscellaneous', 'Civil Works - Construction Works', 'Electronics Equipment'})</pre>	0.0414707	0.889908	12.2441
<pre>rozenset({'Computer- Data Processing', 'Miscellaneous lorks', 'Miscellaneous'})</pre>	<pre>frozenset({'Civil Works - Construction Works', 'Computer- H/W', 'Electronics Equipment'})</pre>	0.0414707	0.873874	11.7471
<pre>rozenset({'Computer- S/W', 'Miscellaneous Works', 'Civil Works - Construction Works'})</pre>	<pre>frozenset({'Miscellaneous', 'Computer- H/W', 'Electronics Equipment'})</pre>	0.0427533	0.917431	11.7261
<pre>rozenset({'Computer- S/W', 'Civil Works - Construction lorks', 'Food Products'})</pre>	<pre>frozenset({'Miscellaneous', 'Computer- H/W', 'Electronics Equipment'})</pre>	0.0401881	0.903846	11.5524
<pre>rozenset({'Computer- S/W', 'Stationery', 'Civil Works - construction Works'})</pre>	<pre>frozenset({'Miscellaneous', 'Computer- H/W', 'Electronics Equipment'})</pre>	0.0401881	0.903846	11.5524
<pre>rozenset({'Computer- S/W', 'Miscellaneous', 'Food 'roducts'})</pre>	<pre>frozenset({'Civil Works - Construction Works', 'Computer- H/W', 'Electronics Equipment'})</pre>	0.0401881	0.854545	11.4873
<pre>rozenset({'Computer- Data Processing', 'Miscellaneous lorks', 'Civil Works - Construction Works'})</pre>	<pre>frozenset({'Miscellaneous', 'Computer- H/W', 'Electronics Equipment'})</pre>	0.0414707	0.873874	11.1693
<pre>Frozenset({'Computer- S/W', 'Food Products'})</pre>	<pre>frozenset({'Miscellaneous', 'Computer- H/W', 'Electronics Equipment'})</pre>	0.0440359	0.872881	11.1567
<pre>Frozenset({'Computer- S/W', 'Stationery'})</pre>	<pre>frozenset({'Miscellaneous', 'Computer- H/W', 'Electronics Equipment'})</pre>	0.0436084	0.871795	11.1428
rozenset({'Miscellaneous Works', 'Civil Works - onstruction Works', 'Food Products'})	<pre>frozenset({'Miscellaneous', 'Computer- H/W', 'Electronics Equipment'})</pre>	0.0431808	0.87069	11.1287
<pre>rozenset({'Computer- Data Processing', 'Miscellaneous lorks', 'Miscellaneous', 'Computer- H/W'})</pre>	<pre>frozenset({'Civil Works - Construction Works', 'Electronics Equipment'})</pre>	0.0414707	0.92381	11.081
<pre>rozenset({'Medicines', 'Food Products'})</pre>	<pre>frozenset({'Miscellaneous', 'Computer- H/W', 'Electronics Equipment'})</pre>	0.0410432	0.857143	10.9555
<pre>rozenset({'Stationery', 'Food Products'})</pre>	<pre>frozenset({'Miscellaneous', 'Computer- H/W', 'Electronics Equipment'})</pre>	0.0410432	0.857143	10.9555

It can be inferred that the higher the 'lift' value, the more the chance of association. i.e. Those bidders who have bid for tenders for 'Computer Data Processing' would also bid for tenders for 'Computer-HW'.



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### **Forecasting Analytics**





#### A) Forecast using SARIMA model

Based on the three years data, the forecast for the next few years have been. It may be noticed that the years in which the actuals are almost in line with the forecasted value can be observed.







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#### Auto Detection of Invalid Names



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#### Model's Predictions

Type the Name krishna kanth krishna kanth is a Original Name

Type the Name rupam rupam is a Original Name

Type the Name priya priya is a Original Name

Type the Name gem,m gejkjgnjgn gem,m gejkjgnjgn is a Invalid Name

Type the Name beautiful beautiful is a Invalid Name

Type the Name nobody nobody is a Invalid Name



#### Road Ahead

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Various other quantitative measures like number of bidders, number of users, number of corrigendum etc in a particular month can be predicted similar to the number of tender's method.

Similar to the Invalid User Names, Invalid organisation names, product category names etc can be detected.

Various other quantitative measures like number of bidders, number of users, number of corrigendum etc in a particular month can be predicted similar to the number of tender's method.

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- GePNIC Product Details https://gepnic.gov.in
- Dashboard https://eprocure.gov.in/eprocdashboard/







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