

Putting the Cure into Strategic Procurement

NLP soft weighted voting ensemble multi-class classifier to predict spend categories using Government of California's 2012 -2015 purchase order data

Dene Stalk

BrainStation Data Science Diploma Capstone Project

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Putting the Cure into Strategic Procurement

DENE STALK

Data Science Diploma

BSc, MBA

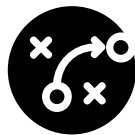
Work Experience

Management Consulting, Internal Strategy, Tourism

Importance of conducting a spend analysis



Identify cost saving opportunities



Facilitate strategic procurement decision making and sourcing

Capstone Overview



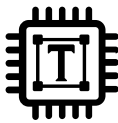
Predict purchase order spend categories



6 weeks



Government of California's 2012-2015 purchase orders



NLP weighted voting ensemble using supplier name, item name and item description

Tools: Python, Pandas, Numpy, Sklearn, Tableau, MySQL

Correctly Classified



**77% of
line items**



**92% of
Total Spend**

Overview of Business Importance, Challenges and Machine Learning's value in procurement

Why a Spend Analysis?



Full spend visibility



Identify cost saving opportunities



Data-driven sourcing

Challenges



Disconnect across regions, languages and business units



Labour intensive



Prone to human error

ML's Procurement Value



Reduces need for manual intervention



Timely strategic decision making



Better inventory and supplier management

Importance of Correctly Classifying Spend Categories

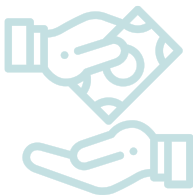
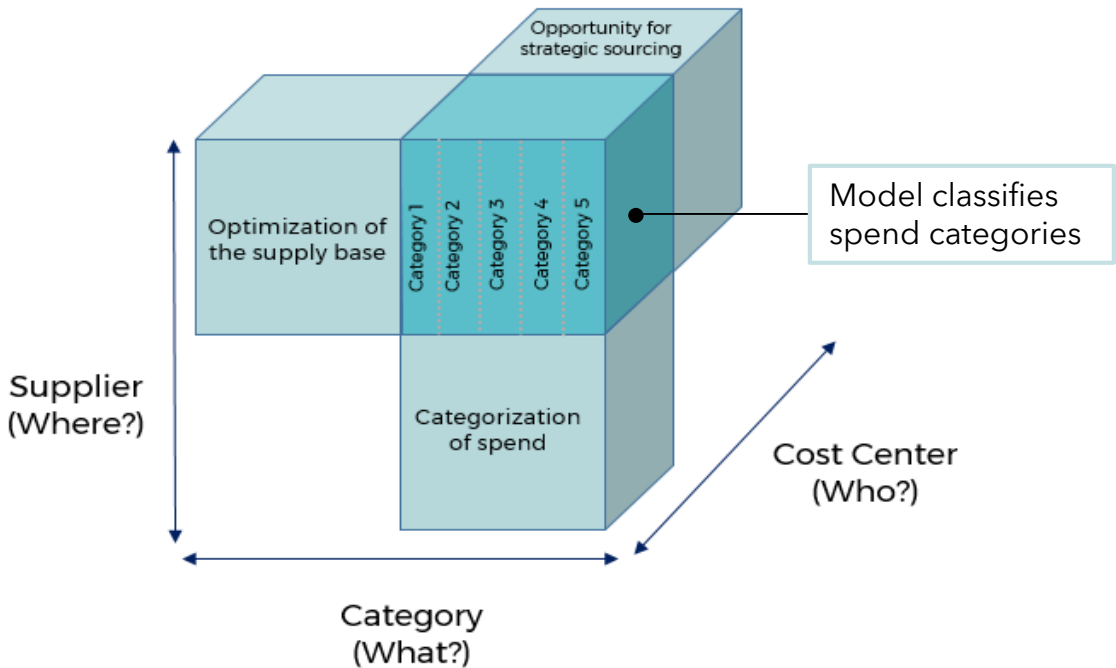


Purchase Order (PO): document issued by a buyer committing to purchasing products or services



Takes an average of 1 year to identify and correct 100k misclassified line items

Increased spend visibility leads to lower Total Cost of Ownership¹



Improved buying power through strategic sourcing



Optimizing inventory management processes

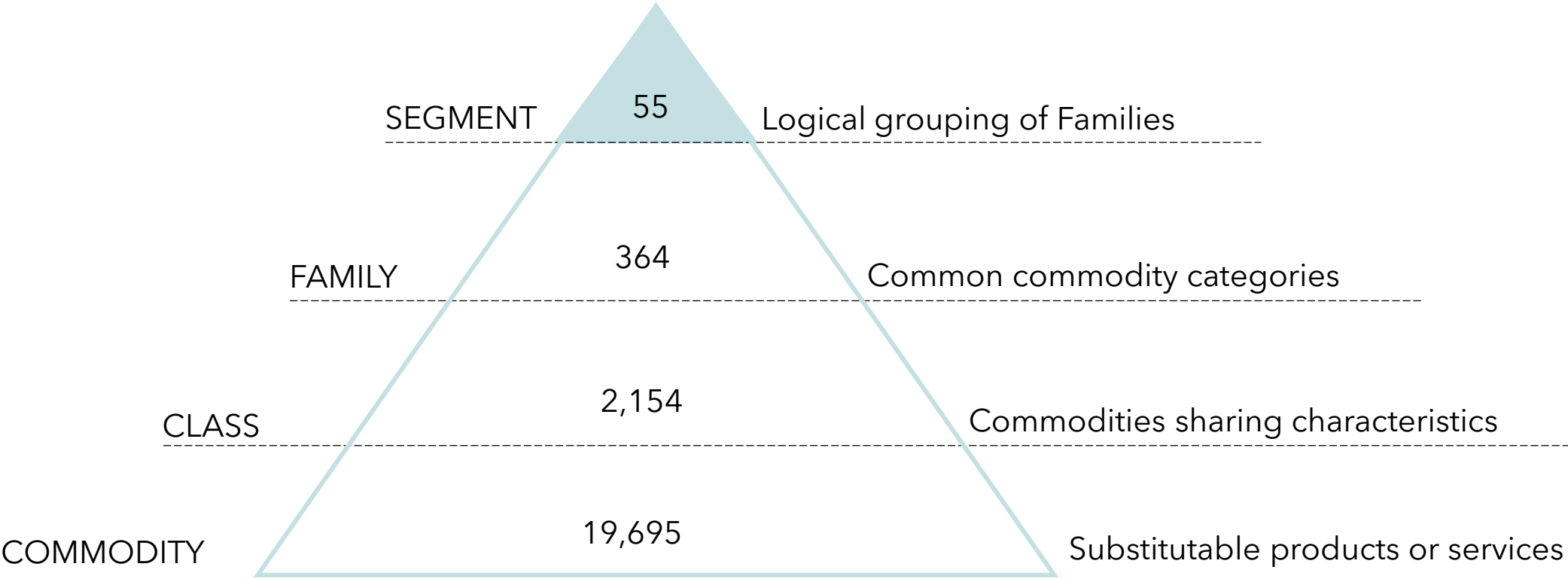


Managing risk exposure to maverick spending²

1 - Total Cost of Ownership is a combination of purchasing and carrying costs
2 - Maverick spending refers to purchases made outside of agreed contracts



UNSPSC : United Nations Standard Products & Services Code



Data Collection and Description



Purchase Order Data 2012-2015

Government of California merged information from various procurement systems, then mapped the UNSPSC taxonomy

Cleaned Dataset Description

3 fiscal years

326k line items

24k suppliers

3.6m words



\$147bn in
Total Spend

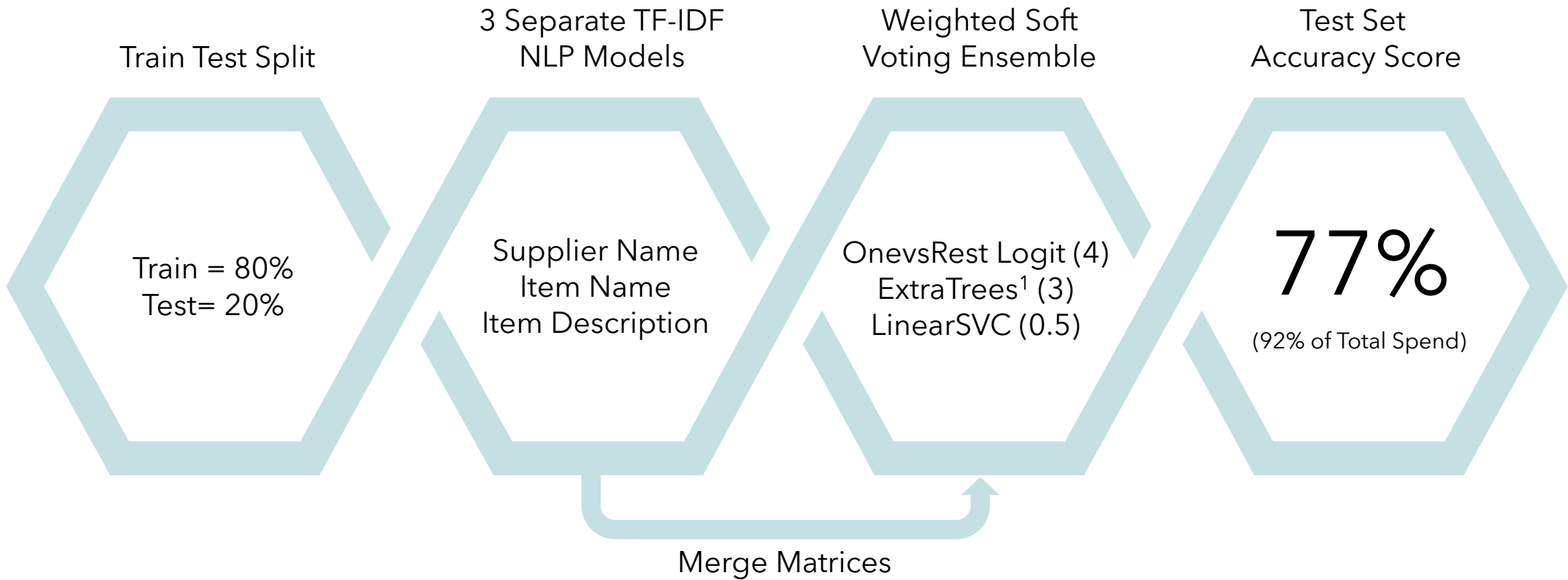
26 Spend categories¹

171k unique item names
208k unique item descriptions

1 - Bottom categories in terms of items lines were placed into either Other Services or Other Goods

Modelling Methodology

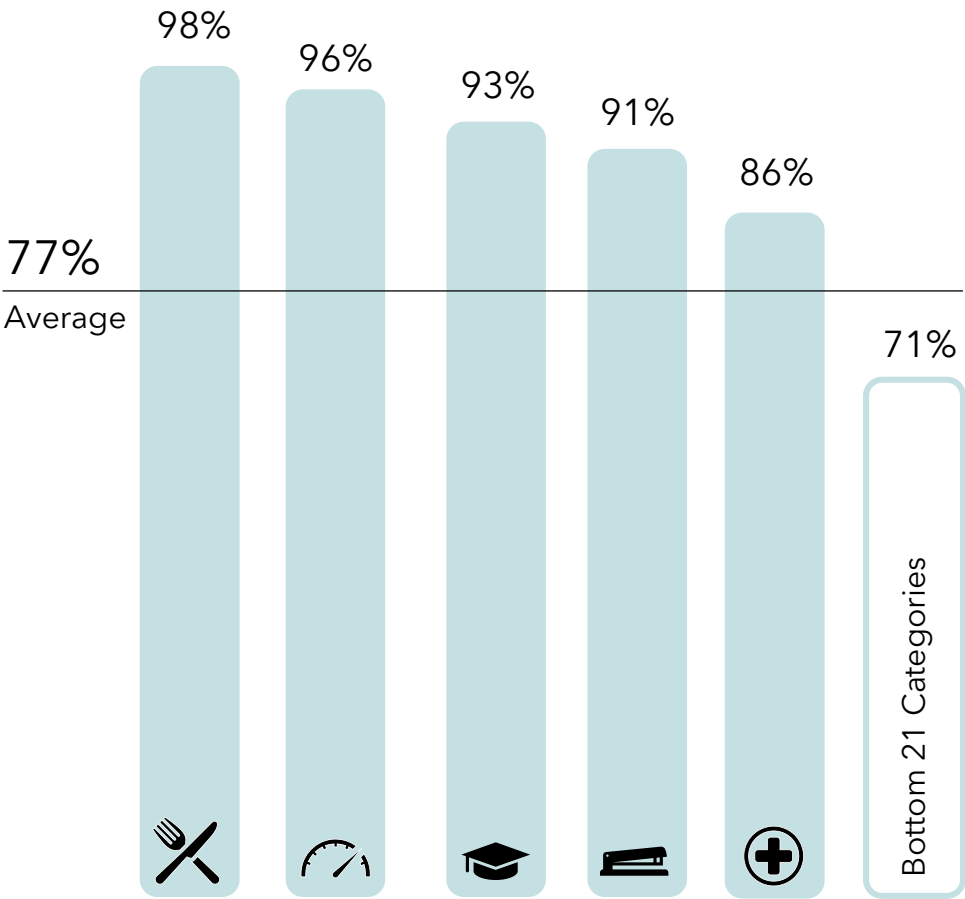
NLP weighted soft voting ensemble predicted 65k line items of unseen data with 77% accuracy



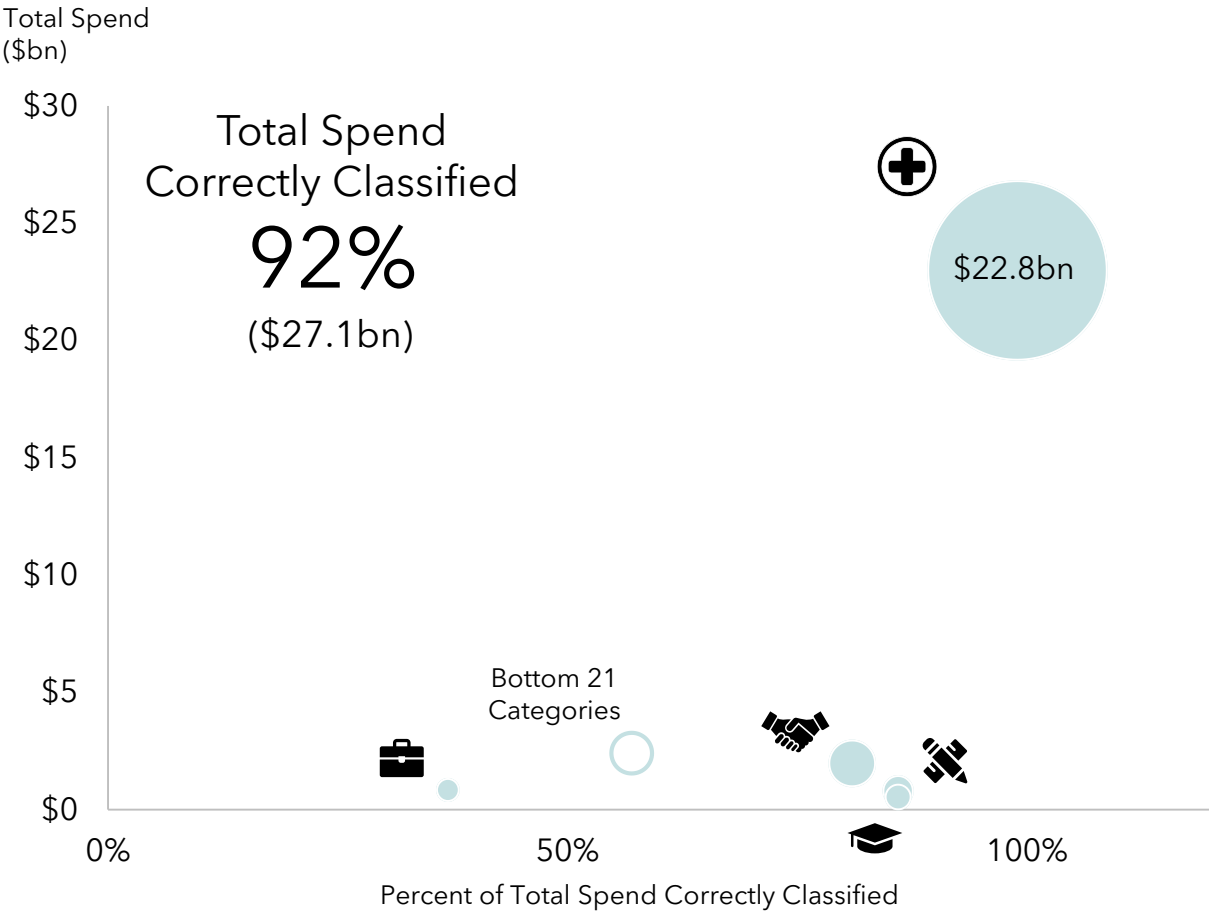
1 - ExtraTrees differs from RandomForest by splitting nodes randomly and does not resample observations

Results

Top 5 Categories by Percent of Line Items Correctly Classified



Top 5 Categories by Total Spend (bubble size indicates Total Spend correctly classified)



- Food & Beverage Products
- Fuel & Lubricants
- Health Services
- Office Supplies
- Education & Training Services
- Professional Services
- Other Services
- Engineering Services

Conclusion

Model Strengths



- Relatively strong predictive value using only 3 free text columns
- Correctly classified the vast majority of spend
- 17 categories' line items correctly classified with over 70% accuracy
- Model is scalable

Model Weaknesses



- Not tested on corporate data
- Health Services over represented in terms of Total Spend
- 3 categories < 50% of line items correctly predicted

Further Analysis



- Meta-model incorporating quantity and unit price
- Deeper analysis on strategic value in relation to supplier management
- Test transferability on new Government of California data and corporate data
- Create an interactive dashboard

Thank you

Dene Stalk

denemstalk@gmail.com
416-573-5784

[linkedin.com/in/denestalk](https://www.linkedin.com/in/denestalk)
github.com/denestalk