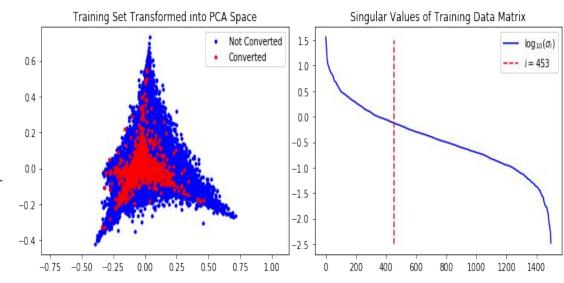


Jared Cook, Elizabeth Herman, Evan North, Katrina Petroske

Background image: https://www.digitalistmag.com/future-of-work/2017/05/04/digital-businesses-need-anomaly-detection-05062375

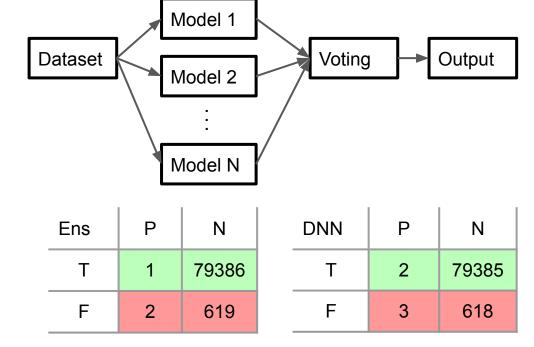
Introduction and Motivation

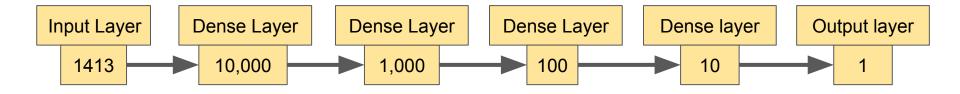
- User shopping traffic from Valassis. Binary classification of "Converted" vs. "Not Converted".
- Conversion is a rare event (~1% of total).
- Objective Question: Can conversion be predicted from a user's interest data?
 - Accurate predictions of conversion leads to more profits.
- Approaches are Dimensionality reduction (PCA) and popular machine learning prediction techniques.
- 99% of variance explained in 453 singular vectors.
 - PCA space provides no clear linear classification.
 - Singular values provide no (useful) clear truncation point





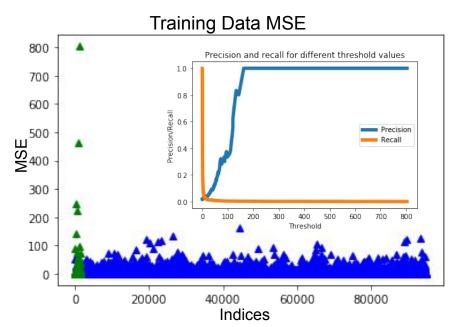
- Ensemble multiple methods
 - Naive Bayes
 - Random Forest
 - Logistic Regression
 - Cross-validation
 - ROC score
- Deep Neural Network (DNN)
 - Nonlinear dimensionality reduction
 - Fully connected, feed forward
 - Motivate autoencoder

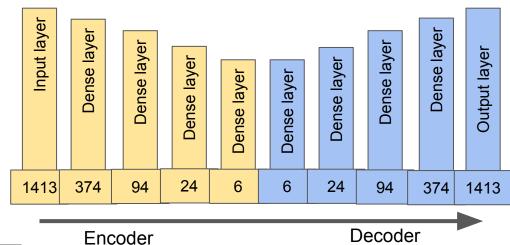




Autoencoder

- Anomaly detection
 - Train on majority group
 - Large MSE → anomaly
 - Hyperparameter: threshold





	ı		ı			I
AE 200	Р	N		AE 100	Р	N
Т	14	77622		Т	31	75591
F	1766	606		F	3797	589
			1			

Conclusions and Future Work

- Naive out of the box machine learning methods do not work well for rare event prediction
- Autoencoder produces best results using the confusion matrix metric
- Future work:
 - Hyperparameter optimization
 - Employ F1, F2, ROC metrics for performance and cross-validation
 - Other anomaly detection methods
 - Boosting methods (Gradient boosting/Adaboost)
 - SVM
 - Further exploration of data for feature selection
 - Non-linear dimension reduction
 - Can we target consumers that are more likely to convert?