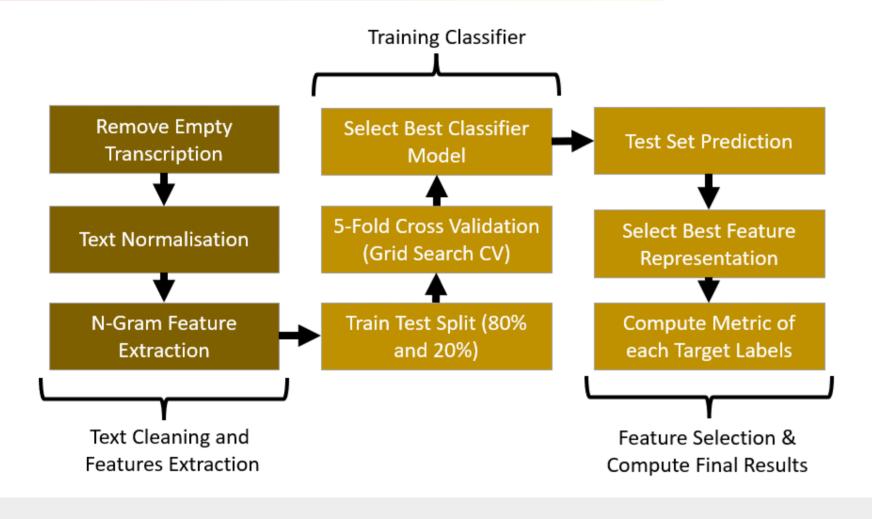
Text Classification of Medical Transcript

01 Problem

- 1. Healthcare workers spend **a lot of time** to identify key issues of each transcript
- 2. Transcripts with complex cases that lead to information **overload and delays**

03 Solution Overview



05 N-Gram Feature Extraction

5 N-gram Feature vectors are extracted

Feature Vectors	Lengths of array dimension	
	Rows	Columns
Unigram	298	7038
Unigram + Bigram	298	56270
Bigram	298	49232
Bigram + Trigram	298	114215
Trigram	298	64983

07 Evaluation Criteria

- 1. Metric Score: Macro F1
- 2. Train-test split 80% and 20%.
- 3. 5-fold cross validation on training set to find optimised hyperparameters.
- 4. Using trained model with best parameter to make a unseen prediction.

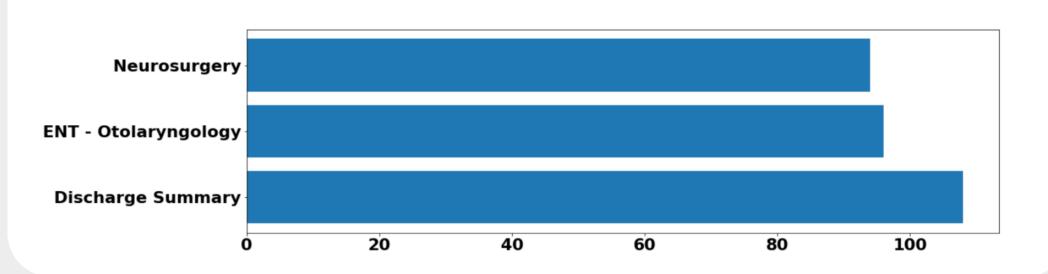
09 Results and Findings

Test Prediction Metric Score using RFC

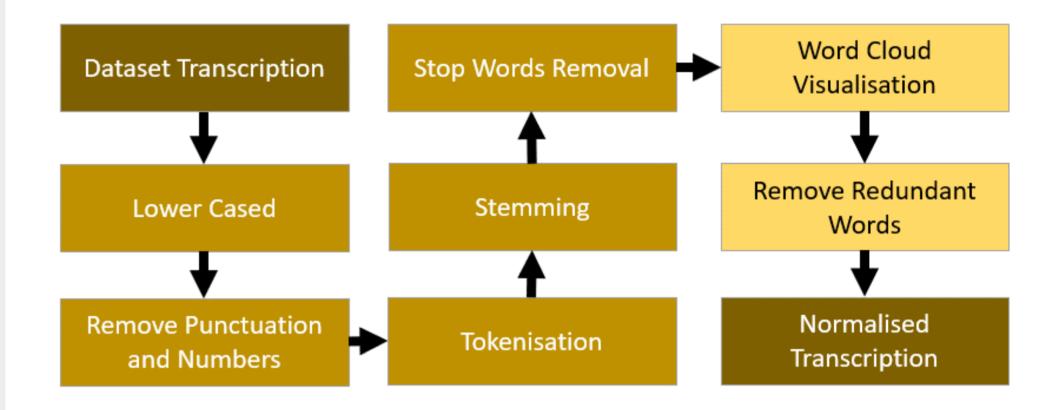
Features	macro F1
Unigram	0.9336
Unigram + Bigram	0.8526
Bigram	0.8372
Bigram + Trigram	0.8499
Trigram	0.5773

02 Dataset

Medical transcript dataset is used, which is sourced from Kaggle and MT samples. It has 300 observations with 3 label classes



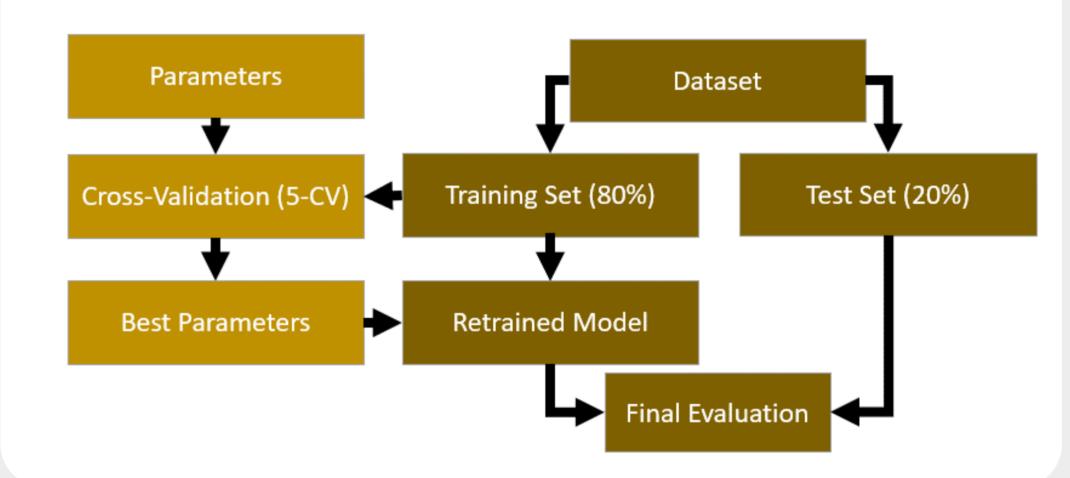
04 Text Normalisation



-42.9% Less words after text cleaning

Deployed text normalisation tools in website http://www.morris-lee.com/nlp

06 Experiment Setup



08 Text Classifiers

- 1. K-Nearest Neighbour (n_neighbors = 7)
- 2. Decision Tree Classifier (max_depth = None, min_samples_split = 2)
- 3. Random Forest Classifier (RFC) (max_depth = 35, n_estimators = 16)

10 Key Findings

- 1. Unigram is the best feature vector in classifying the medical transcript, obtaining best result, **0.93 macro F1.**
- 2. This may indicates that the adjacent word ordering less important in the classification task **in this context**.