

Team
Members:

- Aayush Shangani
- Darsh Shetty
- Pavan Antala
- Sheroz Shaikh

DS595: Natural Language Processing



Team Contribution





Aayush

ML + TF IDF + BERT

Darsh

ML + Word2VEC + DL BERT FINE TUNE

Pavan

ML POS + DL BERT FINE TUNE LORA + RNN

Sheroz

DL LSTM GRU BI-LSTM





Motivation





Assuming:

We are working for the multinational company similar to Deloitte, Mckinsey etc.



Client









Pizza

NLP Engineer/Data Scientist





Dataset Overview

We are working for the multinational company similar to Deloitte, Mckinsey etc.









Users

Ranking

Business

Reviews







User-Checkins





Problem Domain





How data filtered:



Big Data Set







What we finalized:









Little Caesars

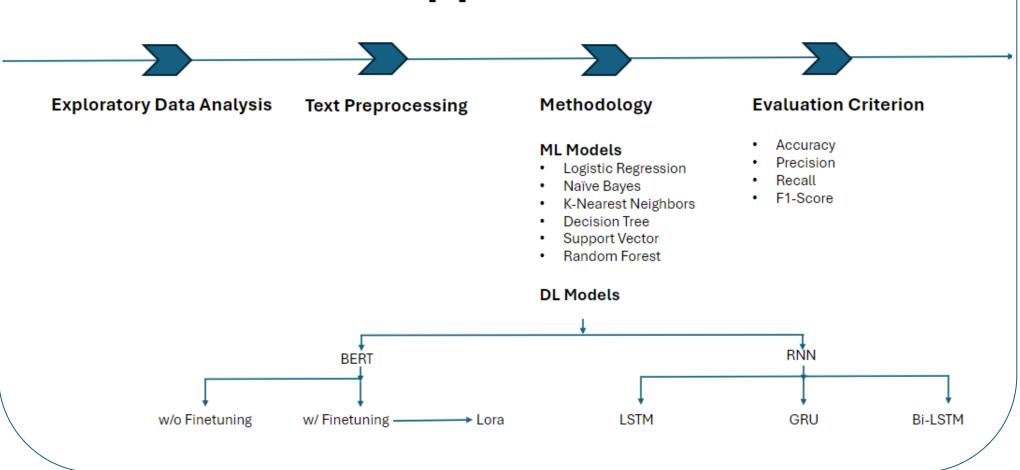












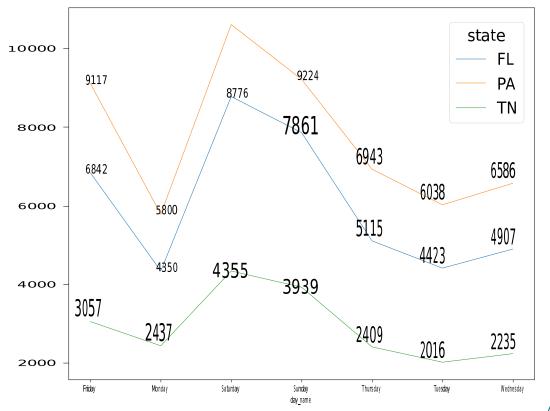






Line Plot of day_name by state









Features

ML Models

DL Models

TF-IDF



Logistic-Regression



SVM



BERT



Part of Speech



Naïve Bayes



KNN



RNN



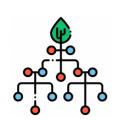
Word2Vec



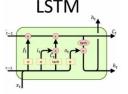
Decision Tree



Random Forest

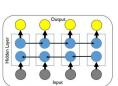


LSTM



GRU

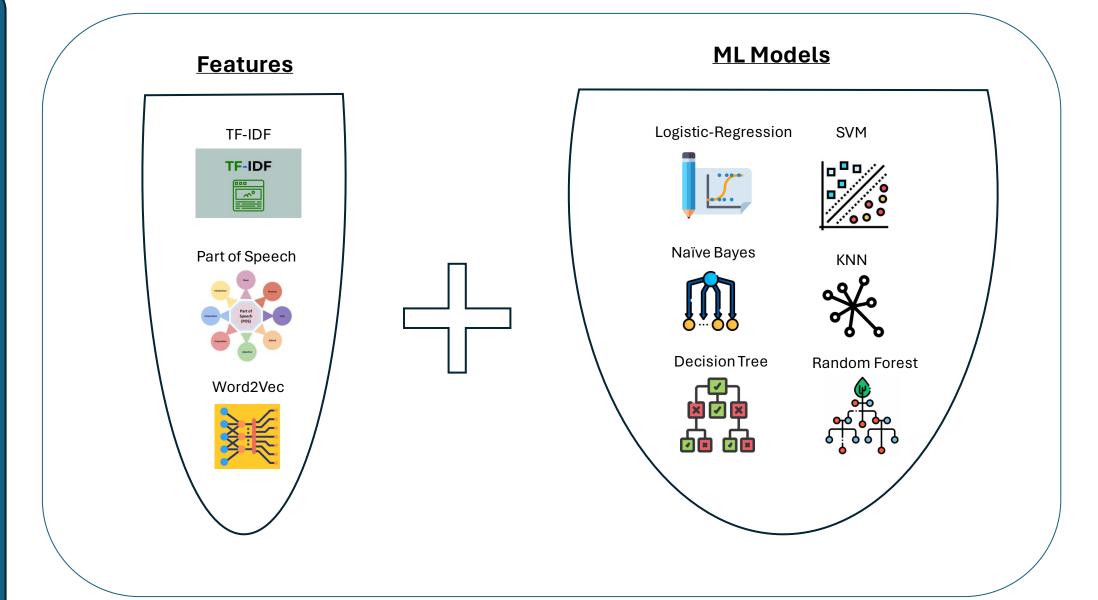
Bi-LSTM















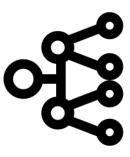


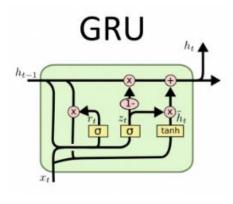
DL Models

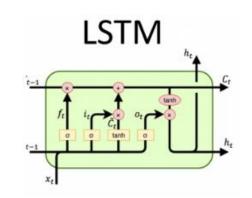
BERT

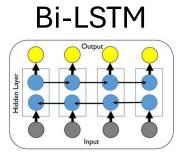


RNN











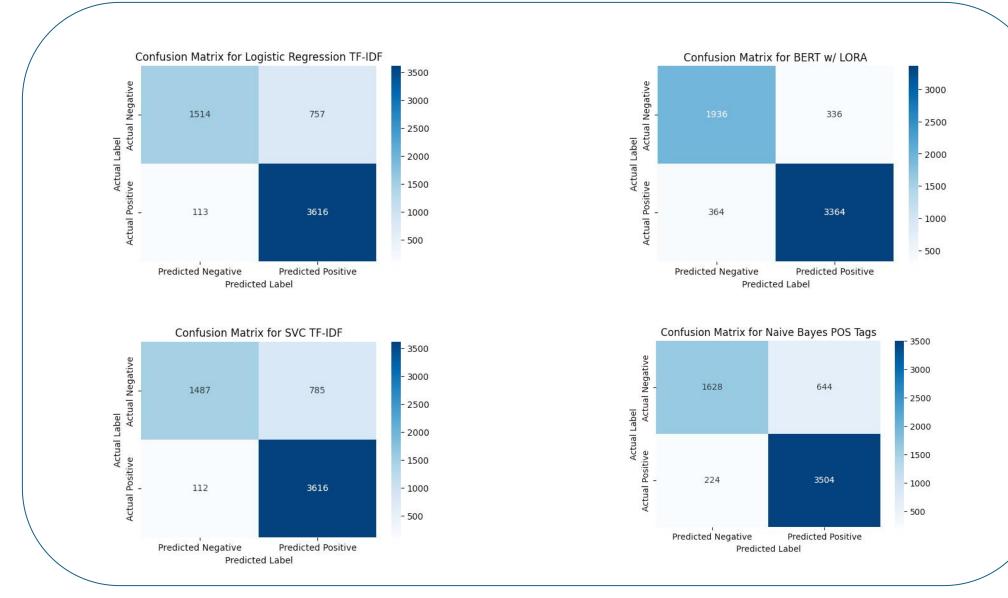




ML Model	∨ Feature ∨	Accuracy	Precision ~	Recall ~	F1-Score ~
Logistic Regression	TF-IDF	85.51	82.69	96.99	89.27
Naive Bayes	Part of Speech Tag	85.51	84.46	93.98	88.97
Support Vector	TF-IDF	85.05	82.17	96.99	88.97
Support Vector	Part of Speech Tag	84.58	85.71	90.23	87.91
K-Nearest Neighbors	TF-IDF	82.71	82.88	90.98	86.74
Random Forest	TF-IDF	80.84	78.75	94.74	86.01
Logistic Regression	Part of Speech Tag	81.78	85.07	85.17	85.39
Random Forest	Part of Speech Tag	78.97	76.50	95.49	84.95
Naive Bayes	TF-IDF	78.04	74.16	99.25	84.89
Decision Tree	Word2Vec	76.34	85.51	83.91	84.70
Random Forest	Word2Vec	73.36	74.68	86.47	80.14
K-Nearest Neighbors	Part of Speech Tag	64.95	64.22	98.50	77.74
Support Vector	Word2Vec	62.15	62.15	100.00	76.66
Decision Tree	Part of Speech Tag	71.03	78.86	72.93	75.78
Decision Tree	TF-IDF	69.16	74.81	75.94	75.37
Logistic Regression	Word2Vec	61.68	63.63	94.93	74.38
K-Nearest Neighbors	Word2Vec	64.02	67.50	81.20	73.72
Naive Bayes	Word2Vec	57.48	75.00	47.37	58.06







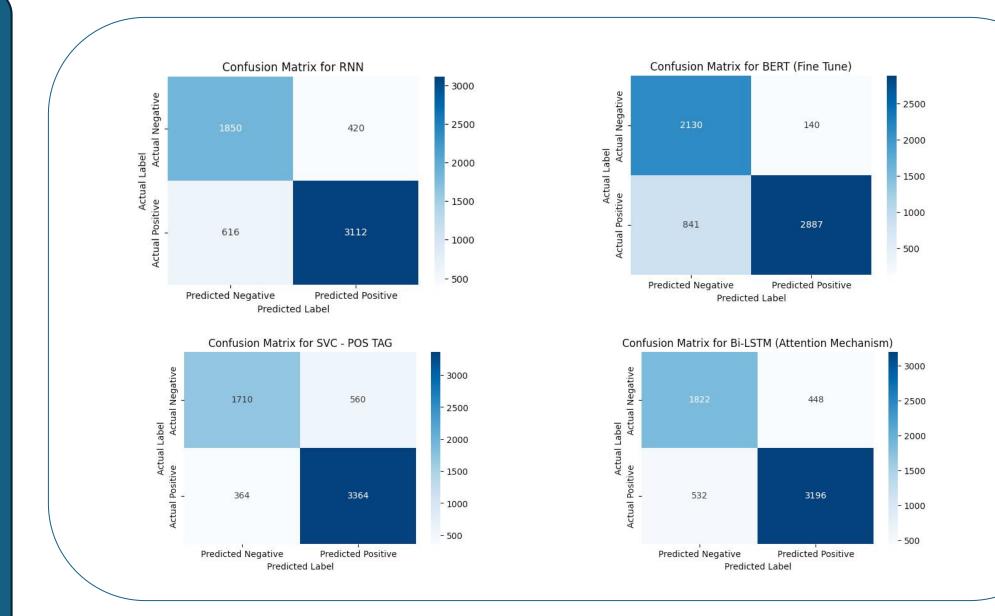






DL Model	Feature	✓ Accuracy ✓	Precision ~	Recall ~	F1-Score ~
BERT	Lora	88.32	90.91	90.23	90.57
Bi-LSTM	Attention Mechanism	83.64	87.69	85.71	86.69
RNN	-	82.71	88.10	83.46	85.71
BERT	Fine Tune	83.64	95.37	77.44	85.48
BERT	w/o Fine Tune	82.06	90.33	79.36	84.49
LSTM	-	83.18	89.09	80.33	84.48
GRU	-	74.77	78.33	77.05	77.69











$\overline{\mathsf{v}}$ ML + DL Model $\overline{}$ v Precision $\overline{}$ Feature Accuracy Recall F1-Score BERT 88.32 90.91 90.23 90.57 Lora TF-IDF 82.69 96.99 Logistic Regression 85.51 89.27 Part of Speech Tag 85.51 84.46 93.98 88.97 Naive Bayes 85.05 96.99 82.17 88.97 Support Vector TF-IDF Part of Speech Tag 84.58 85.71 90.23 87.91 Support Vector K-Nearest Neighbors TF-IDF 82.71 90.98 82.88 86.74 Bi-LSTM 83.64 87.69 85.71 86.69 Attention Mechanism 80.84 94.74 Random Forest TF-IDF 78.75 86.01 RNN 82.71 88.10 83.46 85.71 77.44 BERT Fine Tune 83.64 95.37 85.48 Logistic Regression Part of Speech Tag 81.78 85.07 85.17 85.39 Random Forest Part of Speech Tag 78.97 76.50 95.49 84.95 TF-IDF 78.04 74.16 99.25 Naive Bayes 84.89 Decision Tree Word2Vec 76.34 85.51 83.91 84.70 BERT w/o Fine Tune 82.06 90.33 79.36 84.49 LSTM 83.18 89.09 80.33 84.48 Word2Vec Random Forest 73.36 74.68 86.47 80.14 K-Nearest Neighbors Part of Speech Tag 64.95 64.22 98.50 77.74 GRU 74.77 78.33 77.05 77.69 62.15 Support Vector Word2Vec 62.15 100.00 76.66 Part of Speech Tag 71.03 78.86 72.93 75.78 Decision Tree 74.81 75.94 Decision Tree TF-IDF 69.16 75.37 Logistic Regression Word2Vec 61.68 63.63 94.93 74.38 K-Nearest Neighbors 64.02 67.50 81.20 Word2Vec 73.72 Word2Vec 47.37 Naive Bayes 57.48 75.00 58.06







Conclusion





- A combined model that incorporates an attention mechanism could yield improved results.
- A model that merges attention mechanisms with NLP techniques may surpass other approaches in performance.





