	Introduction to Text Analytics
Report on Sentiment Analysis of IMDI	B Movie Reviews
$\mathbf{B}\mathbf{y}$	
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### **Abstract:**

This report presents an extensive comparative analysis of sentiment analysis models applied to the IMDb dataset, integrating various preprocessing techniques and feature extraction methodologies. The study encompasses a wide array of machine learning algorithms including Naive Bayes, Decision Trees, Random Forest, Gradient Boosting, Stacking, Neural Networks, K-nearest neighbours, and Voting classifiers. Additionally, state-of-the-art pre-trained language models (PLMs) such as RoBERTa and DistilBERT and embedding techniques such as Glove (Global Vectors for Word Representation) and Word2Vec were employed to enhance feature representation. Furthermore, lexicon-based approaches such as AFINN, TextBlob, and VADER were incorporated for sentiment analysis. Diverse preprocessing techniques such as stop word removal, tokenization, POS tag removal, and different vectorization strategies including TF-IDF and count vectorizer were applied. Stemming techniques like Porter, Lancaster, and Snowball, as well as lemmatization via Spacy, were also utilized to preprocess the textual data. Through rigorous evaluation, the performance of each model and preprocessing combination is thoroughly assessed, providing detailed insights into the impact of preprocessing techniques and model choices on sentiment analysis accuracy.

### **Objective:**

The objective of this assignment is to explore and compare various methods for text classification and sentiment analysis, ranging from traditional lexicon-based approaches to modern pre-trained Large Language Models (LLMs).

### **Dataset:**

The IMDb dataset serves as a rich and diverse corpus for sentiment analysis tasks, comprising a vast collection of movie reviews contributed by users. With thousands of reviews spanning various genres, languages, and demographics, the dataset offers a comprehensive representation of audience opinions and sentiments towards movies. Each review within the IMDb dataset encapsulates valuable insights into viewers' perceptions, ranging from nuanced critiques to enthusiastic praises. Leveraging this extensive dataset, our analysis aims to uncover underlying sentiment trends and patterns, facilitating the evaluation and comparison of sentiment analysis models across diverse movie review scenarios. The training dataset (or portion of it) has been used to train the machine learning model, learning customized Word2Vec, or finetuning PLMs (RoBERTa, DistillBERT, GPT2). The test data has be used for evaluation purposes.

### **Preprocessing Techniques:**

### • Stop Word Removal:

Stop words, such as "the," "is," and "and," were removed from the IMDb movie reviews. These words are common and do not contribute significant semantic meaning to the text, thus eliminating them can improve the efficiency and accuracy of sentiment analysis algorithms.

### • Tokenization:

The process of tokenization involved breaking down the movie reviews into individual words or tokens. This step facilitates further analysis by providing a structured representation of the text data, enabling algorithms to process and understand the content more effectively.

### • Part-of-Speech (POS) Tag Removal:

POS tags, indicating the grammatical category of each word (e.g., noun, verb, adjective), were removed from the movie reviews. This step simplified the text data while preserving essential semantic information, contributing to more accurate sentiment analysis results.

### • Vectorization:

Different vectorization techniques such as TF-IDF (Term Frequency-Inverse Document Frequency) and count vectorizer were employed to convert the textual data into numerical representations. These representations enable machine learning algorithms to process and analyze the IMDb movie reviews effectively, capturing the underlying semantic information for sentiment analysis tasks.

#### • Stemming and Lemmatization:

Stemming techniques including Porter, Lancaster, and Snowball, as well as lemmatization via Spacy, were applied to reduce words to their base or root form. This process standardizes word forms and reduces inflectional variants, ensuring consistency in the text data and improving sentiment analysis accuracy.

### **Model Evaluation:**

## **Machine Learning Models**

Stop words	Tokenize	Pos	Stemming/ Lemmatize	Vectorizer	Time Taken	Classification Report and Accuracy
yes	yes	yes	Porter Stemmer	Tf – Idf Vectorizer		Naive Bayes Accuracy: 0.8469 Classification Report for Naive Bayes:
						precision recall f1-score support  negative 0.85 0.84 0.85 9935 positive 0.85 0.85 0.85 10065
						accuracy 0.85 20000 macro avg 0.85 0.85 0.85 20000 weighted avg 0.85 0.85 0.85 20000
						['positive' 'positive' 'negative' 'negative' 'posi
						/n Random Forest Accuracy: 0.83375 Classification Report for Random Forest:     precision recall f1-score support
						negative 0.83 0.84 0.83 9935 positive 0.84 0.82 0.83 10065
						accuracy 0.83 20000 macro avg 0.83 0.83 0.83 20000 weighted avg 0.83 0.83 0.83 20000 k-NN Accuracy: 0.66965
						Classification Report for k-NN: precision recall f1-score support
						negative 0.65 0.74 0.69 9935 positive 0.70 0.60 0.64 10065
						accuracy 0.67 20000 macro avg 0.67 0.67 0.67 20000 weighted avg 0.67 0.67 0.67 20000
						/n Decision Tree Accuracy: 0.69575 Classification Report for Decision Tree: precision recall f1-score support
						negative 0.69 0.70 0.70 9935 positive 0.70 0.69 0.70 10065
						accuracy 0.70 20000
						macro avg 0.70 0.70 0.70 20000 weighted avg 0.70 0.70 0.70 20000
						/n Gradient Boosting Accuracy: 0.79635 Classification Report for Gradient Boosting:
						negative 0.83 0.75 0.78 9935 positive 0.77 0.84 0.81 10065
						accuracy 0.80 0.80 20000 macro avg 0.80 0.80 0.80 20000 weighted avg 0.80 0.80 0.80 20000
						Stacking Accuracy: 0.8613 Classification Report for Stacking:
						negative 0.86 0.86 0.86 9935 positive 0.86 0.87 0.86 10065
						accuracy 0.86 20000 macro avg 0.86 0.86 0.86 20000 weighted avg 0.86 0.86 0.86 20000 Neural Network Accuracy: 0.8495
						Classification Report for Neural Network: precision recall f1-score support
						negative 0.85 0.85 0.85 9935 positive 0.85 0.85 0.85 10065
						accuracy 0.85 20000 macro avg 0.85 0.85 0.85 20000 weighted avg 0.85 0.85 0.85 20000
						Bagging Accuracy: 0.7925 Classification Report for Bagging:
						negative 0.79 0.79 0.79 9935 positive 0.79 0.79 0.79 10065
						accuracy 0.79 20000 macro avg 0.79 0.79 0.79 20000 weighted avg 0.79 0.79 0.79 20000

yes	yes	yes	Porter Stemmer	Count					
				Vectorizer	Naive Bayes Ac Classification		Naive Bay	yes: f1-score	support
					negative positive	0.83 0.84	0.84 0.83	0.84 0.84	9935 10065
					accuracy macro avg weighted avg	0.84 0.84	0.84 0.84	0.84 0.84 0.84	20000 20000 20000
					['positive' 'p	ositive' 'ne	egative'	'negati	ve' 'posit
					/n Random Forest Classification		Random F	orest: f1-score	support
					negative positive	0.83 0.84	0.84 0.83	0.83 0.83	9935 10065
					accuracy macro avg weighted avg	0.83 0.83	0.83 0.83	0.83 0.83 0.83	2000 2000 2000
					['positive' 'p k-NN Accuracy: Classification	0.58895 Report for	k-NN:		
					negative	precision 0.57	0.66	f1-score 0.62	support 9935
					positive	0.61	0.52	0.56 0.59	10065 20000
					macro avg weighted avg	0.59 0.59	0.59 0.59	0.59 0.59	20000 20000 20000
					Decision Tree Classification		Decision	Tree: f1-score	support
					negative positive	0.70 0.70	0.69 0.70	0.69 0.70	9935 10065
					accuracy macro avg weighted avg	0.70 0.70	0.70 0.70	0.70 0.70 0.70	20000 20000 20000
					/n Gradient Boost Classification		Gradient		support
					negative positive	0.82 0.77	0.75 0.84	0.78 0.81	9935 10065
					accuracy macro avg	0.80	0.79	0.80 0.79	20000 20000
					weighted avg Stacking Accur Classification	Report for	Stacking:		20000
					negative	precision 0.86	recall 0.85	f1-score 0.85	support 9935
					positive	0.85	0.86	0.86 0.85	10065
					macro avg weighted avg Bagging Accura	0.85 0.85 cy: 0.7902	0.85 0.85		20000 20000 20000
					Classification			f1-score	support
					negative positive	0.79 0.79	0.78 0.80	0.79 0.79	9935 10065
					accuracy macro avg weighted avg	0.79 0.79	0.79 0.79	0.79 0.79 0.79	20000 20000 20000
					/n Voting Classif Classification		Voting C		support
					negative positive	0.83 0.83	0.82 0.84	0.83 0.83	9935 10065
					accuracy macro avg	0.83	0.83	0.83 0.83	20000 20000
					weighted avg Neural Network Classification			0.83 etwork:	20000
					negative	precision 0.86	recall 0.85	f1-score 0.86	support 9935
					positive accuracy	0.85	0.86	0.86 0.86	10065 20000
					macro avg weighted avg	0.86 0.86	0.86 0.86	0.86 0.86	20000 20000

yes	yes	yes	Lancaster	Tf – Idf					
			Stemmer	Vectorizer	Naive Bayes Acc Classification		Naive Bay	/es: f1-score	support
					negative positive	0.84 0.85	0.84 0.85	0.84 0.85	9935 10065
					accuracy macro avg weighted avg	0.84 0.84	0.84 0.84	0.84 0.84 0.84	20000 20000 20000
					['positive' 'po			'negati	ve' 'posit
					Random Forest A Classification	Report for precision	Random Fo	f1-score	support
					negative positive	0.83 0.84	0.84 0.83	0.83 0.83	9935 10065
					accuracy macro avg weighted avg	0.83 0.83	0.83 0.83	0.83 0.83 0.83	20000 20000 20000
					['negative' 'po k-NN Accuracy: Classification	0.67225	k-NN:	'negati	ve' 'posit support
					negative positive	0.64 0.72	0.77 0.57	0.70 0.64	9935 10065
					accuracy macro avg weighted avg	0.68 0.68	0.67 0.67	0.67 0.67 0.67	20000 20000 20000
					/n Decision Tree / Classification		Decision	Tree: f1-score	support
					negative positive	0.69 0.70	0.70 0.69	0.69 0.69	9935 10065
					accuracy macro avg weighted avg	0.69 0.69	0.69 0.69	0.69 0.69 0.69	20000 20000 20000
					/n Gradient Boosti Classification		Gradient		support
					negative positive	0.82 0.77	0.75 0.84	0.79 0.81	9935 10065
					accuracy macro avg weighted avg Stacking Accurac Classification R			0.80 0.80 0.80	20000 20000 20000 support
					negative positive	0.87 0.86	0.85 0.87	0.86 0.86	9935 10065
					accuracy macro avg weighted avg	0.86 0.86	0.86 0.86	0.86 0.86 0.86	20000 20000 20000
					Bagging Accurac Classification	y: 0.79065	Bagging:	f1-score	
					negative positive	0.79 0.79	0.78 0.80	0.79 0.79	9935 10065
					accuracy macro avg weighted avg	0.79 0.79	0.79 0.79	0.79 0.79 0.79	2000 2000 2000
					/n Voting Classifi Classification		Voting C		support
					negative positive	0.84 0.84	0.84 0.84	0.84 0.84	9935 10065
					accuracy macro avg weighted avg	0.84 0.84	0.84 0.84	0.84 0.84 0.84	2000 2000 2000
					Neural Network Classification		Neural N	etwork: f1-score	support
					negative positive	0.85 0.85	0.84 0.85	0.85 0.85	9935 10065
					accuracy macro avg weighted avg	0.85 0.85	0.85 0.85	0.85 0.85 0.85	20000 20000 20000

yes	yes	yes	Lancaster	Count					
			Stemmer	Vectorizer	Naive Bayes Ac Classification		Naive Ba	f1-score	support
					negative positive	0.83 0.84	0.84 0.83	0.83 0.83	9935 10065
					accuracy macro avg weighted avg	0.83 0.83	0.83 0.83	0.83 0.83 0.83	20000 20000 20000
					['positive' 'p			'negati	ve' 'posi
					Random Forest Classification		Random F	orest: f1-score	support
					negative positive	0.83 0.83	0.83 0.83	0.83 0.83	9935 10065
					accuracy macro avg weighted avg	0.83 0.83	0.83 0.83	0.83 0.83 0.83	20000 20000 20000
					['negative' 'p Neural Network Classification	Accuracy:	0.85775 Neural N		
					negative positive	0.86 0.85	0.85 0.87	0.86 0.86	9935 10065
					accuracy macro avg weighted avg	0.86 0.86	0.86 0.86	0.86 0.86 0.86	20000 20000 20000
					Decision Tree Classification		Decision	Tree: f1-score	support
					negative positive	0.69 0.70	0.70 0.69	0.69 0.70	9935 10065
					accuracy macro avg weighted avg	0.70 0.70	0.70 0.70	0.70 0.70 0.70	20000 20000 20000
					/n Gradient Boost Classification		Gradient	Boosting:	support
					negative positive	0.82 0.77	0.75 0.84	0.78 0.81	9935 10065
					accuracy macro avg weighted avg	0.80 0.80	0.80 0.80	0.80 0.79 0.80	20000 20000 20000
					Stacking Accura	acy: 0.8516 Report for precision		f1-score	support
					negative positive	0.86 0.85	0.84 0.86	0.85 0.85	9935 10065
					accuracy macro avg weighted avg	0.85 0.85	0.85 0.85	0.85 0.85 0.85	20000 20000 20000
					Bagging Accura Classification		Bagging: recall	f1-score	support
					negative positive	0.79 0.79	0.78 0.80	0.79 0.79	9935 10065
					accuracy macro avg weighted avg	0.79 0.79	0.79 0.79	0.79 0.79 0.79	20000 20000 20000
					/n Voting Classif: Classification	ier Accuracy Report for precision	Voting Cl	assifier: f1-score	support
					negative positive	0.83 0.83	0.82 0.84	0.83 0.83	9935 10065
					accuracy macro avg weighted avg	0.83 0.83	0.83 0.83	0.83 0.83 0.83	20000 20000 20000
					-				
L			<u> </u>						

yes	yes	Lancaster	Count					
		Stemmer	Vectorizer	Naive Bayes A Classification		Naive Ba	yes: f1-score	support
				negative positive	0.83 0.85	0.85 0.83	0.84 0.84	9935 10065
				accuracy macro avg weighted avg	0.84 0.84	0.84 0.84	0.84 0.84 0.84	20000 20000 20000
				['positive' '	positive' 'n	egative'	'negati	ve' 'posi
				/n Random Forest Classificatio	Accuracy: 0 n Report for precision	Random F	orest: f1-score	support
				negative positive	0.84 0.85	0.85 0.84	0.84 0.84	9935 10065
				accuracy macro avg weighted avg	0.84 0.84	0.84 0.84	0.84 0.84 0.84	20000 20000 20000
				['negative' ' Neural Networ Classification	k Accuracy:	0.86675 Neural N		
				negative positive	0.87 0.86	0.86 0.88	0.86 0.87	9935 10065
				accuracy macro avg weighted avg	0.87 0.87	0.87 0.87	0.87 0.87 0.87	20000 20000 20000
				Decision Tree Classification		Decision	Tree: f1-score	support
				negative positive	0.72 0.73	0.72 0.72	0.72 0.72	9935 10065
				accuracy macro avg weighted avg	0.72 0.72	0.72 0.72	0.72 0.72 0.72	20000 20000 20000
				/n Gradient Boos Classification		Gradient		support
				negative positive	0.84 0.79	0.76 0.85	0.80 0.82	9935 10065
				accuracy macro avg weighted avg	0.81 0.81	0.81 0.81	0.81 0.81 0.81	20000 20000 20000
				k-NN Accuracy: Classification			f1-score	support
				negative positive	0.61 0.66	0.71 0.55	0.65 0.60	9935 10065
				accuracy macro avg weighted avg	0.63 0.63	0.63 0.63	0.63 0.63 0.63	2000 2000 2000
				/n Stacking Accur Classification			f1-score	support
				negative positive	0.86 0.87	0.86 0.86	0.86 0.86	9935 10065
				accuracy macro avg weighted avg	0.86 0.86	0.86 0.86	0.86 0.86 0.86	2000 2000 2000
				Bagging Accura Classification	cy: 0.80275		f1-score	support
				negative positive	0.81 0.80	0.79 0.82	0.80 0.81	9935 10065
				accuracy macro avg weighted avg	0.80 0.80	0.80 0.80	0.80 0.80 0.80	20000 20000 20000
				/n Voting Classif Classification		Voting Cl		support
				negative positive	0.84 0.84	0.84 0.85	0.84 0.85	9935 10065
				accuracy macro avg weighted avg	0.84 0.84	0.84 0.84	0.84 0.84 0.84	20000 20000 20000
				<i>-</i>				

yes	yes	Lancaster Stemmer	Count Vectorizer	Modified ML Parameters MultinomialNB(alpha=0.1)
				Naive Bayes Accuracy: 0.84295 Classification Report for Naive Bayes: precision recall f1-score support
				negative 0.84 0.85 0.84 9935 positive 0.85 0.83 0.84 10065
				accuracy
				RandomForestClassifier(n_estimators=200, max_depth=10, random_state=42)
				Random Forest Accuracy: 0.8277 Classification Report for Random Forest: precision recall fl-score support
				negative 0.86 0.78 0.82 9935 positive 0.80 0.87 0.84 10065
				accuracy 0.83 20000 macro avg 0.83 0.83 0.83 20000 weighted avg 0.83 0.83 0.83 20000
				MLPClassifier(hidden_layer_sizes=(128,), max_iter=200, alpha=0.0001, random_state=42)
				Neural Network MLP CLASSIFIER Accuracy: 0.8717 Classification Report for Neural Network: precision recall f1-score support
				negative 0.88 0.86 0.87 9935 positive 0.87 0.88 0.87 10065
				accuracy 0.87 20000 macro avg 0.87 0.87 0.87 20000 weighted avg 0.87 0.87 0.87 20000
				DecisionTreeClassifier(max_depth=10, min_samples_split=10, random_state=42)  Decision Tree Accuracy: 0.736
				Classification Report for Decision Tree: precision recall fl-score support
				negative 0.78 0.65 0.71 9935 positive 0.79 0.82 0.76 10065
				accuracy 0.74 20000 macro avg 0.74 0.74 0.73 20000 weighted avg 0.74 0.74 0.73 20000
				GradientBoostingClassifier(n_estimators=20 0, max_depth=5, learning_rate=0.1, random_state=42)
				Gradient Boosting Accuracy: 0.8446 Classification Report for Gradient Boosting: precision recall f1-score support
				negative 0.86 0.83 0.84 9935 positive 0.83 0.86 0.85 10065
				accuracy 0.84 20000 macro avg 0.85 0.84 0.84 20000 weighted avg 0.84 0.84 0.84 20000
				KNeighborsClassifier(n_neighbors=10) k-NN Accuracy: 0.6297 Classification Report for k-NN:
				precision recall f1-score support  negative 0.59 0.80 0.68 9935 positive 0.70 0.46 0.55 10065
				accuracy 0.65 0.63 20000 macro avg 0.65 0.63 0.62 20000 weighted avg 0.65 0.63 0.62 20000
				StackingClassifier( estimators=[('nb', nb_model), ('rf', rf_model), ('knn', knn_model), ('dt', dt_model),('gb', gb_model)], final_estimator=GradientBoostingClassifier( n_estimators=100, random_state=42)
				Stacking Accuracy: 0.869 Classification Report for Stacking: precision recall f1-score support
				negative 0.87 0.87 0.87 9935 positive 0.87 0.87 0.87 10065
				accuracy 0.87 0.87 20000 macro avg 0.87 0.87 20000 weighted avg 0.87 0.87 0.87 20000

yes	yes	yes	Lancaster	Tf-Idf					
, , ,	y 0.5	y 0 0	Stemmer	Vectorizer +	Naive Bayes Ac			,	,,_
				robust Scaler	Classification		Naive Bay	es: f1-score	support
					negative positive	0.85 0.84	0.84 0.85	0.84 0.85	9935 10065
					accuracy macro avg	0.85	0.85	0.85 0.85	20000 20000
					weighted avg	0.85	0.85	0.85	20000
					Random Forest / Classification		Random Fo	rest: f1-score	support
					negative positive	0.84 0.80	0.79 0.86	0.81 0.83	9935 10065
					accuracy macro avg weighted avg	0.82 0.82	0.82 0.82	0.82 0.82 0.82	20000 20000 20000
					Neural Network Classification		Neural Ne	twork:	
					negative positive	0.86 0.85	0.85 0.86	0.85 0.86	9935 10065
					accuracy	0.85	0.85	0.85 0.85	20000 20000
					macro avg weighted avg	0.85	0.85	0.85 0.85	20000
					Decision Tree / Classification		Decision	Tree: f1-score	support
					negative positive	0.78 0.69	0.62 0.83	0.69 0.75	9935 10065
					accuracy macro avg weighted avg	0.73 0.73	0.72 0.72	0.72 0.72 0.72	20000 20000 20000
					Gradient Boost Classification		Gradient		support
					negative positive	0.85 0.82	0.81 0.86	0.83 0.84	9935 10065
					accuracy macro avg weighted avg	0.83 0.83	0.83 0.83	0.83 0.83 0.83	20000 20000 20000
					k-NN Accuracy Classificatio			f1-score	support
					negative positive	0.62 0.78	0.86 0.48	0.72 0.59	9935 10065
					accuracy macro avg weighted avg	0.70			20000 20000 20000
yes	yes	yes	Lancaster Stemmer	Tf-Idf Vectorizer + MinMax Scaler	Naive Bayes A Classification		Naive Bay	yes: f1-score	support
					negative positive	0.85 0.84	0.84 0.85	0.84 0.85	9935 10065
					accuracy macro avg weighted avg	0.84 0.84	0.84 0.84	0.84 0.84 0.84	2000 2000 2000
					Random Forest Classification		Random Fo	orest: f1-score	support
					negative positive	0.84 0.80	0.79 0.86	0.81 0.83	9935 10065
					accuracy macro avg	0.82	0.82	0.82 0.82	20000 20000
					weighted avg	0.82	0.82	0.82	20000

					Neural Network	Report for	Neural Ne		
					negative	0.86	0.85	0.86	9935
					positive accuracy			0.86	10065 20000
					accuracy macro avg weighted avg	0.86 0.86	0.86 0.86	0.86 0.86	
					Decision Tree Classification	Report for	Decision	Tree: f1-score	support
					negative positive		0.62 0.83	0.69 0.75	9935 10065
					accuracy	0.73	0.72	0.72 0.72	20000 20000
					accuracy macro avg weighted avg	0.73	0.72		20000
					Gradient Boos Classification		Gradient	Boosting: f1-score	support
					negative positive			0.83 0.84	9935 10065
					accuracy macro avg weighted avg	0.83 0.83	0.83 0.83		2000 2000 2000
					k-NN Accuracy: Classification			f1-score	support
					negative positive			0.66 0.03	9935 10065
					accuracy macro avg weighted avg	0.60			20000 20000 20000
			T	TC 11C					
yes	yes	yes	Lancaster Stemmer	Tf-Idf Vectorizer + Standard Scaler	Naive Bayes A Classification	n Report for	Naive Bay	yes: f1-score	support
					negative positive		0.84	0.84 0.85	9935 10065
					accuracy macro avg	0.85	0.85	0.85 0.85	20000 20000
					weighted avg	0.85	0.85	0.85	20000
					Random Forest Classification	n Report for	Random F	orest: f1-score	support
					negative positive	0.84 0.80	0.79 0.86	0.81 0.83	9935 10065
					accuracy macro avg weighted avg	0.82 0.82			20000 20000 20000
					Neural Networ	n Report for	Neural N		
					negative positive	0.86 0.85	0.85 0.86	0.85 0.86	9935 10065
					accuracy macro avg weighted avg	0.85 0.85	0.85 0.85		20000 20000 20000
					k-NN Accuracy: Classification			f1-score	support
					negative positive	0.62 0.78	0.86 0.48	0.72 0.59	9935 10065
					accuracy			0.67	20000
					macro avg weighted avg	0.70 0.70	0.67 0.67	0.66 0.66	2000 2000
					Stacking Accur Classification				support
					negative positive	0.87 0.86	0.85 0.88	0.86 0.87	9935 10065
					accuracy macro avg weighted avg	0.86 0.86	0.86 0.86	0.86 0.86 0.86	20000 20000 20000

7100	7700	1/00	Lamanatan	Tf-Idf					
yes	yes	yes	Lancaster Stemmer	Vectorizer +					
			Stemmer	Standard Scaler	Decision Tree Classification	Accuracy: 0. Report for precision	Decision	Tree: f1-score	support
					negative positive	0.78 0.69	0.62 0.83	0.69 0.75	9935 10065
					accuracy macro avg weighted avg	0.73 0.73	0.72 0.72	0.72 0.72 0.72	20000 20000 20000
					Gradient Boost Classification	ting Accuracy Report for precision	Gradient	Boosting: f1-score	support
					negative positive	0.85 0.82	0.81 0.86	0.83 0.84	9935 10065
					accuracy macro avg weighted avg	0.83 0.83	0.83 0.83	0.83 0.83 0.83	20000 20000 20000
yes	yes	yes	Snowball Stemmer	TF-IDF Vectorizer	Naive Bayes Ac	curacy: 0.84	435 Naive Bav	es:	
					negative	precision 0.85		f1-score 0.84	support 9935
					positive accuracy	0.84	0.85	0.85 0.84	10065 20000
					macro avg weighted avg	0.84 0.84	0.84 0.84	0.84 0.84	20000 20000
					['positive' 'p /n Random Forest Classification	Accuracy: 0.	83315 Random Fo		e' 'posi1
					negative positive	0.83 0.84	0.84 0.83	0.83 0.83	9935 10065
					accuracy macro avg weighted avg	0.83 0.83	0.83 0.83	0.83 0.83 0.83	20000 20000 20000
					['positive' 'p Neural Network Classification	Accuracy: 0	.84755 Neural Ne	_	e' 'posi1 support
					negative positive	0.85 0.85	0.85 0.85	0.85 0.85	9935 10065
					accuracy macro avg weighted avg	0.85 0.85	0.85 0.85	0.85 0.85 0.85	20000 20000 20000
					Decision Tree Classification		Decision '	Tree: f1-score	support
					negative positive	0.69 0.70	0.70 0.68	0.70 0.69	9935 10065
					accuracy macro avg weighted avg	0.69 0.69	0.69 0.69	0.69 0.69 0.69	20000 20000 20000
					Gradient Boost Classification		Gradient	Boosting: f1-score	support
					negative positive	0.83 0.77	0.75 0.84	0.79 0.81	9935 10065
					accuracy macro avg weighted avg	0.80 0.80	0.80 0.80	0.80 0.80 0.80	20000 20000 20000
					k-NN Accuracy: Classification	: 0.6682 n Report for precision	k-NN: recall	f1-score	support
					negative positive	0.65 0.70	0.73 0.60	0.69 0.65	9935 10065
					accuracy macro avg weighted avg	0.67 0.67	0.67 0.67	0.67 0.67 0.67	20000 20000 20000
					/n Voting Classif Classification		Voting Cl	assifier: f1-score	support
					negative positive	0.84 0.84	0.83 0.85	0.84 0.84	9935 10065
					accuracy macro avg weighted avg	0.84 0.84	0.84 0.84	0.84 0.84 0.84	20000 20000 20000

	I	1	C 1 11	TE IDE						
yes	yes	yes	Snowball	TF-IDF						
			Stemmer	Vectorizer		Stacking Accur Classification	racy: 0.8614 n Report for precision	Stacking: recall	f1-score	support
						negative positive	0.86 0.86	0.86 0.86	0.86 0.86	9935 10065
						accuracy macro avg weighted avg	0.86 0.86	0.86 0.86	0.86 0.86 0.86	20000 20000 20000
						/n Bagging Accur Classification	acy: 0.7939 n Report for precision	Bagging:	f1-score	support
						negative positive	0.79 0.79	0.79 0.80	0.79 0.80	9935 10065
						accuracy macro avg	0.79	0.79	0.79 0.79	20000 20000
						weighted avg	0.79	0.79	0.79	20000
yes	yes	yes	Snowball Stemmer	TF-IDF Vectorizer + MinMax		Naive Bayes A Classificatio	ccuracy: 0.8 n Report for precision	Naive Ba	yes: f1-score	support
				Scaler		negative positive	0.85 0.84	0.84 0.85	0.84 0.85	9935 10065
						accuracy macro avg weighted avg	0.84 0.84	0.84 0.84		20000 20000 20000
						Random Forest Classificatio		Random F	orest: f1-score	support
						negative positive	0.85 0.80	0.78 0.86		9935 10065
						accuracy macro avg weighted avg	0.82 0.82	0.82 0.82		20000 20000 20000
						Neural Networ Classificatio		Neural N	letwork:	
						negative positive	0.86 0.86	0.86 0.86		9935 10065
						accuracy macro avg weighted avg	0.86 0.86	0.86 0.86		
						Decision Tree Classificatio		Decision		support
						negative positive	0.79 0.69	0.61 0.84	0.69 0.76	9935 10065
						accuracy macro avg weighted avg	0.74 0.74	0.73 0.73	0.73 0.72 0.72	20000 20000 20000
						Gradient Boos Classificatio		Gradient		
						negative positive	0.85 0.82	0.81 0.86	0.83 0.84	9935 10065
						accuracy macro avg weighted avg	0.83 0.83	0.83 0.83	0.83 0.83 0.83	20000 20000 20000
						k-NN Accuracy: Classification			f1-score	support
						negative positive	0.50 0.77	1.00 0.01	0.67 0.03	9935 10065
						accuracy macro avg weighted avg	0.63 0.63		0.50 0.35 0.34	20000 20000 20000

yes	yes	Snowball	Tf – Idf					
		Stemmer	Vectorizer	Naive Bayes Ac Classification	Report for	Naive Bay		
				negative positive	precision 0.85 0.85	0.85 0.85	0.85 0.85	9935 10065
				accuracy macro avg weighted avg	0.85 0.85	0.85 0.85	0.85 0.85 0.85	20000 20000 20000
				['positive' 'p	ositive' 'n	egative' .	'negati	ve' 'posit
				Random Forest Classification		Random Fo	rest: f1-score	support
				negative positive	0.84 0.86	0.86 0.84	0.85 0.85	9935 10065
				L	0.85 0.85	0.85 0.85	0.85 0.85 0.85	20000 20000 20000
				k-NN Accuracy: Classification	0.74325	k-NN:	f1-score	
				negative positive	0.76 0.73	0.71 0.77	0.73 0.75	9935 10065
				accuracy macro avg weighted avg	0.74 0.74	0.74 0.74	0.74 0.74 0.74	20000 20000 20000
				/n Decision Tree Classification		Decision	Tree: f1-score	support
				negative positive	0.72 0.72	0.72 0.72	0.72 0.72	9935 10065
				accuracy macro avg weighted avg	0.72 0.72	0.72 0.72	0.72 0.72 0.72	20000 20000 20000
				/n Gradient Boost Classification		Gradient		support
				negative positive	0.84 0.79	0.77 0.86	0.81 0.82	9935 10065
				accuracy macro avg weighted avg	0.82 0.82	0.82 0.82	0.82 0.82 0.82	20000 20000 20000
				Neural Network Classification	Accuracy:	Neural Ne	etwork:	support
				negative positive	0.86 0.86	0.85 0.86	0.86	9935 10065
				accuracy macro avg weighted avg	0.86 0.86	0.86 0.86		20000 20000 20000
				Stacking Accur Classification		Stacking:	f1-score	support
				negative positive	0.87 0.88	0.87 0.88	0.87 0.88	9935 10065
				accuracy macro avg weighted avg	0.87 0.87	0.87 0.87	0.87 0.87 0.87	20000 20000 20000
				Bagging Accura Classification			f1-score	support
				negative positive	0.81 0.81	0.81 0.81	0.81 0.81	9935 10065
				accuracy macro avg weighted avg	0.81 0.81	0.81 0.81	0.81 0.81 0.81	20000 20000 20000
				/n Voting Classif: Classification		Voting Cla		support
				negative positive	0.87 0.85	0.84 0.87	0.85 0.86	9935 10065
				accuracy macro avg weighted avg	0.86 0.86	0.86 0.86	0.86 0.86 0.86	20000 20000 20000

ves	ves	yes	Snowball	Count	Т	_				
yes	yes	yes	Stemmer	Vectorizer						
						Neural Network Classification		Neural N	etwork: f1-score	support
						negative positive	0.86 0.86	0.85 0.86	0.86 0.86	9935 10065
			ļ			accuracy			0.86	20000
						macro avg weighted avg	0.86 0.86	0.86 0.86	0.86 0.86	2000 2000
						Decision Tree Classification	Accuracy: 0 Report for precision	Decision	Tree: f1-score	support
						negative positive	0.70 0.70	0.70 0.71	0.70 0.70	9935 10065
						accuracy macro avg weighted avg	0.70 0.70	0.70 0.70		2000 2000 2000
						Gradient Boost Classification		Gradient	Boosting:	support
						negative positive	0.83 0.77	0.75 0.85	0.79 0.81	9935 10065
						accuracy macro avg weighted avg	0.80 0.80	0.80	0.80 0.80 0.80	20000 20000 20000
						Naive Bayes Ac Classification	curacy: 0.8	3735 Naive Bay		support
						negative positive	0.83 0.84	0.84 0.83	0.84 0.84	9935 10065
						accuracy macro avg weighted avg	0.84 0.84	0.84 0.84		20000 20000 20000
						Random Forest	Report for	Random Fo		
						negative positive	precision 0.83 0.84	0.84 0.82	0.83 0.83	9935 10065
			ļ			accuracy			0.83	20000
						macro avg weighted avg	0.83 0.83	0.83 0.83		2000 2000
						k-NN Accuracy: Classification			f1-score	support
			ļ			negative positive	0.58 0.61	0.67 0.51	0.62 0.56	9935 10065
						accuracy macro avg weighted avg	0.59 0.59	0.59 0.59	0.59 0.59 0.59	2000 2000 2000
						Stacking Accura Classification		Stacking:	f1-score	support
						negative positive	0.86 0.85	0.85 0.86	0.85 0.86	9935 10065
						accuracy macro avg weighted avg	0.85 0.85	0.85 0.85	0.85 0.85 0.85	20000 20000 20000
								0.03	0.03	20000
						Bagging Accuracy Classification		Bagging: recall	f1-score	support
						negative positive	0.80 0.79	0.78 0.81	0.79 0.80	9935 10065
						accuracy macro avg weighted avg	0.79 0.79	0.79 0.79	0.79 0.79 0.79	20000 20000 20000
						Voting Classific	er Accuracy Report for precision	Voting Cl	assifier:	support
						negative positive	0.84 0.83	0.82 0.84	0.83 0.83	9935 10065
						accuracy macro avg weighted avg	0.83 0.83	0.83 0.83	0.83 0.83 0.83	20000 20000 20000

yes	yes	Snowball	Count Vectorizer					
		Stemmer		Naive Bayes Ac Classification		Naive Ba	yes: f1-score	gunn +
				negative positive	0.84 0.85	0.85 0.84	0.84 0.84	9935 10065
				accuracy macro avg	0.84	0.84	9.84 9.84	20000
				weighted avg	0.84	0.84		20000
				Random Forest Classification		Random F	orest: f1-score	support
				negative positive	0.84 0.85	0.85 0.84	0.85 0.85	9935 10065
				accuracy macro avg weighted avg	0.85 0.85	0.85 0.85	0.85 0.85 0.85	20000 20000 20000
				Neural Network Classification		Neural N	etwork: f1-score	support
				negative positive	0.87 0.86	0.86 0.87	0.87 0.87	9935 10065
				accuracy macro avg weighted avg	0.87 0.87	0.87 0.87	0.87 0.87 0.87	20000 20000 20000
				Decision Tree Classification		Decision	Tree: f1-score	support
				negative positive	0.72 0.72	0.72 0.72	0.72 0.72	9935 10065
				accuracy macro avg weighted avg	0.72 0.72	0.72 0.72		20000 20000 20000
				Gradient Boost Classification		Gradient		
				negative positive	0.84 0.79	0.77 0.85	0.80 0.82	9935 10065
				accuracy macro avg weighted avg	0.81 0.81	0.81 0.81		20000 20000 20000
				k-NN Accuracy: Classification	0.6262 Report for k precision	-NN: recall	f1-score	support
				negative positive	0.61 0.64	0.68 0.57	0.64 0.61	9935 10065
				accuracy macro avg weighted avg	0.63 0.63	0.63 0.63	0.63 0.63 0.63	20000 20000 20000
				Stacking Accura Classification	Report for S	tacking: recall	f1-score	support
				negative positive	0.86 0.87	0.87 0.86	9.87 9.87	9935 10065
				accuracy macro avg weighted avg	0.87 0.87	0.87 0.87	0.87 0.87 0.87	2000 2000 2000
				Bagging Accurac Classification	Report for B	agging: recall	f1-score	support
				negative positive	0.81 0.80	0.79 0.81	0.80 0.81	9935 10065
				accuracy macro avg weighted avg	0.80 0.80	0.80 0.80	0.80 0.80 0.80	2000 2000 2000
				Voting Classifi Classification	er Accuracy: Report for V precision	oting Cla	ssifier: f1-score	support
				negative positive	0.85 0.84	0.83 0.86	0.84 0.85	9935 10065
				accuracy macro avg weighted avg	0.84 0.84	0.84 0.84	0.84 0.84 0.84	2000 2000 2000

yes	yes		Lancaster Stemmer	Tf-Idf Vectorizer	Naive Bayes Accuracy: 0.8503 Classification Report for Naive Bayes: precision recall f1-score support
			2001111101		negative 0.85 0.85 0.85 9935 positive 0.85 0.85 0.85 10065
					accuracy 0.85 20000 macro avg 0.85 0.85 0.85 20000 weighted avg 0.85 0.85 0.85 20000
					Random Forest Accuracy: 0.84525 Classification Report for Random Forest: precision recall f1-score support
					negative 0.84 0.85 0.85 9935 positive 0.85 0.84 0.84 10065
					accuracy 0.85 20000 macro avg 0.85 0.85 20000 weighted avg 0.85 0.85 0.85 20000
					Neural Network Accuracy: 0.8533 Classification Report for Neural Network: precision recall f1-score support
					negative 0.85 0.85 0.85 9935 positive 0.85 0.86 0.85 10065
					accuracy 0.85 20000 macro avg 0.85 0.85 0.85 20000 weighted avg 0.85 0.85 0.85 20000
					Decision Tree Accuracy: 0.7125 Classification Report for Decision Tree: precision recall f1-score support
					negative 0.71 0.72 0.71 9935 positive 0.72 0.71 0.71 10065
					accuracy 0.71 20000 macro avg 0.71 0.71 20000 weighted avg 0.71 0.71 0.71 20000
					Gradient Boosting Accuracy: 0.81185 Classification Report for Gradient Boosting: precision recall f1-score support
					negative 0.84 0.77 0.80 9935 positive 0.79 0.85 0.82 10065
					accuracy 0.81 20000 macro avg 0.81 0.81 0.81 20000 weighted avg 0.81 0.81 0.81 20000
					k-NN Accuracy: 0.74355 Classification Report for k-NN: precision recall f1-score support
					negative 0.75 0.72 0.74 9935 positive 0.74 0.77 0.75 10065
					accuracy 0.74 20000 macro avg 0.74 0.74 20000 weighted avg 0.74 0.74 0.74 20000
					Stacking Accuracy: 0.87125 Classification Report for Stacking: precision recall f1-score support
					negative 0.87 0.87 0.87 9935 positive 0.87 0.87 0.87 10065
					accuracy
					Bagging Accuracy: 0.803 Classification Report for Bagging: precision recall f1-score support
					negative 0.81 0.80 0.80 9935 positive 0.80 0.81 0.81 10065
					accuracy
					Voting Classifier Accuracy: 0.85565 Classification Report for Voting Classifier: precision recall f1-score support
					negative 0.86 0.84 0.85 9935 positive 0.85 0.87 0.86 10065
					accuracy 0.86 20000 macro avg 0.86 0.86 0.86 20000 weighted avg 0.86 0.86 0.86 20000
yes	yes	yes	Spacy	Tf – Idf Vectorizer	Naive Bayes Accuracy: 0.8513 Classification Report for Naive Bayes: precision recall f1-score support
					negative 0.85 0.85 0.85 9935 positive 0.85 0.85 0.85 10065
					accuracy 0.85 20000 macro avg 0.85 0.85 0.85 20000 weighted avg 0.85 0.85 0.85 20000

Classification temporaries   comparison					Random Forest A	ccuracy: 0	.8443		<u> </u>
### WASHINGTON PROPERTY AND									support
Description Act   Description   Descriptio									
Noted Seriors RP CLESSIFIE SCORE() 5.87878					macro avg	0.84	0.84	0.84	20000
Classification Separate the Board Handward   100   1					weighted avg	0.84	0.84	0.84	20000
pointive 0.86 s.85 0.86 28000					Classification	Report for	Neural Ne	twork:	
Yes   Yes   Spacy   Ti-laf   Vectorizer   Spacy   Spacy   Ti-laf   Vectorizer   Spacy   Spacy   Spacy   Ti-laf   Vectorizer   Spacy									
Westing Deep 2, 8,88						0.86	0.86		
Classification Separt for Decision Trace   Deport   Dep									
positive 8.71 8.72 8.73 2008					Classification	Report for	Decision		support
### SCORPEY 8.71 0.73 0.73 0.70 0.73 20000 #################################									
West   Space					accuracy			0.71	20000
Classification Report for Gradient Southing Page   1,000   1									
Per   1.0   1.3   0.76   0.79   5915					Classification	Report for	Gradient	Boosting:	support
### Space   1.0					negative	0.83	0.76	0.79	9935
West   Spacy					accuracy			0.80	20000
Classification Report for NoWIL   megative 0.62 0.54 0.54 0.59 9933   positive 0.08 0.77 0.51 10003     accuracy 0.61 0.60 0.77 0.51 10003     accuracy 0.61 0.60 0.67 0.51 10003     accuracy 0.61 0.60 0.60 0.60 200000     accuracy 0.61 0.60 0.60 0.60 200000     accuracy 0.61 0.60 0.60 0.60 200000     accuracy 0.61 0.60 0.60 0.60 0.60 200000     Stacking Accuracy 0.65 0.60 0.60 0.60 0.60 0.60 0.60 0.60							0.80 0.80		
positive 0.60 0.67 0.50 10055					Classification	Report for		f1-score	support
macro avg					negative positive				
Spacy   Tf - Idf   Vectorizer   Space   Spac						0.61	0.60		
Classification Report for Stacking: precision recall f1-score support					weighted avg	0.61			
positive 0.87 0.86 0.87 10065					Classification F	Report for		f1-score	support
### Spacy   Base   Base									
Vectorizer   Classification Report for Bagging:					accuracy macro avg	0.86	0.87	0.86 0.86	2000 2000
negative					Classification F	Report for	Bagging:	61	
Spacy   Spac						0.80	0.81	0.80	9935
Macro avg						0.81	0.80		
Classification Report for Voting Classifier: precision recall f1-score support					macro avg			0.80	20000
Classification Report for Voting Classifier: precision recall f1-score support									
yes   yes   Spacy   Tf - Idf   Vectorizer					Classification F	Report for	Voting Cla		support
Yes   Yes   Spacy   Tf - Idf   Vectorizer   Naive Bayes Accuracy: 0.85375   Classification Report for Naive Bayes: precision recall f1-score support   negative   0.85									
Yes   Yes   Spacy   Tf - Idf   Vectorizer					accuracy macro avg	0.84	0.84		
Naive Bayes Accuracy: 0.85375   Classification Report for Naive Bayes: precision recall f1-score support					weighted avg	0.84	0.84	0.84	20000
Naive Bayes Accuracy: 0.85375   Classification Report for Naive Bayes: precision recall f1-score support	Vec	Vec	Spacy	Tf_Idf					
Classification Report for Naive Bayes:	yes	yes	Spacy		Naive Baves Acc	uracv: A.S	5375		
positive 0.85 0.85 0.85 10065     accuracy					Classification	Report for	Naive Bay		support
accuracy									
Random Forest Accuracy: 0.8441   Classification Report for Random Forest: precision recall f1-score support					accuracy	0.85	0.85		
Classification Report for Random Forest:									
positive 0.85 0.83 0.84 10065  accuracy 0.84 20000 macro avg 0.84 0.84 0.84 20000					Classification	Report for	Random Fo		support
accuracy 0.84 20000 macro avg 0.84 0.84 20000					negative positive				
					accuracy	0.84	0.84		

					Neural Network	MID CLASSI	TED Assum	A 0E0	
					Classification		Neural Ne		
					negative positive	0.86 0.86	0.86 0.86	0.86 0.86	9935 10065
					accuracy macro avg weighted avg	0.86 0.86	0.86 0.86	0.86 0.86 0.86	20000 2000 2000
					Decision Tree			Tree:	
					negative	precision 0.71	recall 0.72	f1-score	support 9935
					positive	0.72	0.71	0.72	10065
					accuracy macro avg weighted avg	0.72 0.72	0.72 0.72	0.72 0.72 0.72	2000 2000 2000
					Gradient Boost Classification	Report for	Gradient		support
					negative positive	0.84 0.79	0.76 0.86	0.80 0.82	9935 10065
					accuracy macro avg weighted avg	0.81 0.81	0.81 0.81	0.81 0.81 0.81	2000 2000 2000
					k-NN Accuracy: Classification	Report for			
					negative	precision 0.75	recall 0.70	f1-score 0.73	support 9935
					positive accuracy	0.72	0.77	0.75 0.74	10065 20000
					macro avg weighted avg	0.74 0.74	0.74 0.74	0.74 0.74	2000 2000
					Stacking Accur Classification		Stacking:	f1-score	support
					negative positive	0.87 0.87	0.87 0.87	0.87 0.87	9935 10065
					accuracy macro avg weighted avg	0.87 0.87	0.87 0.87	0.87 0.87 0.87	20000 20000 20000
					Bagging Accurac Classification			f1-score	support
					negative positive	0.81 0.81	0.81 0.81	0.81 0.81	9935 10065
					accuracy macro avg weighted avg	0.81 0.81	0.81 0.81	0.81 0.81 0.81	20000 20000 20000
					Voting Classifi Classification	Report for	Voting Cl		
					negative	0.86	0.84	f1-score 0.85	support 9935
					positive accuracy	0.85	0.87	0.86 0.85	10065 20000
					macro avg weighted avg	0.85 0.85	0.85 0.85	0.85 0.85	20000 20000
yes	yes	Spacy	Tf – Idf Vectorizer		Modified Multinomia Naive Bayes Accu	alNB(alparacy: 0.8535	ha=0.	1) :	support
					negative positive	0.85 0.85	0.85 0.86	0.85 0.85	9935 10065
					accuracy macro avg weighted avg	0.85 0.85	0.85 0.85	0.85 0.85 0.85	20000 20000 20000
					RandomFo				
					00, max_de	Accuracy: 0.	82955		=42)
					Classification	Report for precision	Random Fo recall	f1-score	
					negative positive	0.85 0.81	0.80 0.86	0.82 0.84	9935 10065
					accuracy macro avg weighted avg	0.83 0.83	0.83 0.83	0.83 0.83 0.83	2000 2000 2000
					nn_model = MLPClassi	= fier(hida	den lav	ver siza	es=(128
					), max_iter random sta	=200, al			.5 (120,
L			<u> </u>	<u> </u>	randoni_sta	TLJ			

				п	accuracy macro avg ghted avg	0.84 0.84	0.84 0.84	0.84 0.84 0.84	20000 20000 20000
					p-0-0-1-1-1				20003
					negative positive	0.84 0.85	0.85 0.84	0.85 0.84	9935 10065
					ssification	Accuracy: 0. Report for precision	Random Fo	rest: f1-score	support
				п	accuracy macro avg ghted avg	0.84 0.84	0.84 0.84	0.84 0.84 0.84	20000 20000 20000
					negative positive	0.84 0.85	0.85 0.84	0.85 0.84	9935 10065
, 23	7.50	Spacy .			ssification	curacy: 0.84 Report for precision	Naive Bay	es: f1-score	support
yes	yes	Spacy	Count Vectorizer						
				ma	accuracy acro avg hted avg	0.88 0.88	0.88 0.88	0.88 0.88 0.88	20000 20000 20000
					negative positive	0.88 0.88	0.88 0.88	0.88 0.88	9935 10065
				e rf_r dt_1 fina ier( rane stack	model), ( model), ( al_estima (n_estima dom_sta- king Accurace sification	rs=[('nb', ('knn', kn ''gb', gb_ ator=Gra ators=10 te=42))	model) dientB 0,	lel), ('dt ],	',
				m	accuracy pacro avg phted avg	0.75 0.75	0.75 0.75	0.75 0.75 0.75	20000 20000 20000
					negative	0.73 0.78	0.80	0.76	9935 10065
				k-NN	Accuracy: (	sClassific 0.7522 Report for k precision	c-NN:	_	s=10)
						0.85 0.85			20000
					negative positive accuracy	0.83		0.85	9935 10065 20000 20000
				rano Grad Clas	dom_sta dient Boosti ssification	ing Accuracy Report for precision	r: 0.84485 Gradient recall	Boosting: f1-score	support
						ostingCl			
					accuracy			0.72	20000 20000 20000
				Deci Clas	ssification	Accuracy: 0. Report for precision 0.72	Decision Trecall  0.73	f1-score	support 9935 10065
				min	_sample	eeClassifes_split=		x_depth	=10,
				a ma weigh	accuracy acro avg hted avg	0.86 0.86	0.86 0.86	0.86 0.86 0.86	20000 20000 20000
				r		0.86 0.86	recall 4	0.86	9935 10065
					sification R	MLP CLASSIFI Report for N	eural Netv	ork:	

				Neural Network Classification		Neural Ne		2 support
				negative positive	0.87 0.86	0.86 0.87	0.86 0.87	9935 10065
				accuracy macro avg weighted avg	0.87 0.87	0.87 0.87	0.87 0.87 0.87	20000 20000 20000
				Decision Tree A Classification	ccuracy: 0 Report for precision	Decision	Tree: f1-score	support
				negative positive	0.71 0.72	0.73 0.70	0.72 0.71	9935 10065
				accuracy macro avg weighted avg	0.72 0.72	0.72 0.72	0.72 0.72 0.72	20000 20000 20000
				Gradient Boosti Classification	ng Accuracy Report for precision	Gradient	Boosting: f1-score	support
				negative positive	0.84 0.78	0.76 0.85	0.80 0.82	9935 10065
				accuracy macro avg weighted avg	0.81 0.81	0.81 0.81	0.81 0.81 0.81	20000 20000 20000
				k-NN Accuracy: Classification	0.64195 Report for precision		f1-score	support
				negative positive	0.62 0.66	0.70 0.58	0.66 0.62	9935 10065
				accuracy macro avg weighted avg	0.64 0.64	0.64 0.64	0.64 0.64 0.64	20000 20000 20000
				Stacking Accura Classification	cy: 0.86479 Report for precision	Stacking:	f1-score	support
				negative positive	0.86 0.87	0.87 0.86	0.86 0.87	9935 10065
				accuracy macro avg weighted avg	0.86 0.86	0.86 0.86	0.86 0.86 0.86	2000 2000 2000
				Bagging Accurac Classification	y: 0.8032	Bagging:	f1-score	support
				negative positive	0.81 0.80	0.79 0.81	0.80 0.81	9935 10065
				accuracy macro avg weighted avg	0.80 0.80	0.80 0.80	0.80 0.80 0.80	2000 2000 2000
				Voting Classifi Classification		Voting Cl	assifier: f1-score	support
				negative positive	0.85 0.84	0.84 0.85	0.84 0.85	9935 10065
				accuracy macro avg weighted avg	0.84 0.84	0.84 0.84	0.84 0.84 0.84	2000 2000 2000
yes	yes	Spacy	Count Vectorizer	Modified	ML Pa	aramo	eters	
	·			Naive Bayes Acc Classification	uracy: 0.84	46 Naive Bay		support
				negative positive	0.84 0.85	0.85 0.84	0.84 0.84	9935 10065
				accuracy macro avg weighted avg	0.84 0.84	0.84 0.84	0.84 0.84 0.84	2000 2000 2000
				Random Forest A Classification	ccuracy: 0. Report for precision	Random Fo	rest: f1-score	support
				negative positive	0.85 0.81	0.80 0.86	0.82 0.83	9935 10065
				accuracy macro avg weighted avg	0.83 0.83	0.83 0.83	0.83 0.83 0.83	20000 20000 20000

	T	ı	1					
				Neural Network				8
				Classification	Report for precision			support
				negative positive	0.88 0.87	0.87 0.88	0.87 0.87	9935 10065
				positive	0.87	0.00	0.8/	16662
				accuracy			0.87	20000
				macro avg weighted avg	0.87 0.87	0.87 0.87	0.87 0.87	20000 20000
				mczgireco ovg	0.07	0.07	0.07	20000
				D1-1 V		7270		
				Decision Tree Classification			Tree:	
					precision		f1-score	support
				negative	0.79	0.64	0.71	9935
				positive	0.70	0.83	0.76	10065
			l	accuracy			0.74	20000
				macro avg	0.75 0.75	0.74	0.74 0.74	20000 20000
				weighted avg	0.75	0.74	0.74	20000
				Gradient Boost Classification				
					precision		f1-score	support
				negative	0.86	0.82	0.84	9935
				positive	0.83	0.87	0.85	10065
				accuracy			0.85	20000
				macro avg weighted avg	0.85 0.85	0.85 0.85	0.85 0.85	20000 20000
				weighted avg	0.05	0.00	0.65	20000
				k-NN Accuracy:				
				Classification	Report for precision		f1-score	support
					precision	recall	T1-Score	support
				negative	0.61	0.79	0.69	9935
				positive	0.71	0.50	0.59	10065
			l	accuracy			0.65	20000
				macro avg	0.66	0.65 0.65	0.64 0.64	20000
				weighted avg	0.66	0.65	0.64	20000
			l	Stacking Accur Classification				
				C1033111C0C101			f1-score	support
			l	negative	0.87	0.87	0.87	9935
				positive	0.87	0.87	0.87	10065
				accuracy			0.87	20000
			l	macro avg	0.87	0.87	0.87	20000
				weighted avg	0.87	0.87	0.87	20000
			l					

# **Lexicon Based Approaches**

## 1- AFINN

Stop word	Tokenize	Pos	Stemming/ Lemmatization	Accuracy	Time Taken	Classification	Report			
				0.70195	93.4s					
✓									10070	
							precision	recall	f1-score	support
						0	0.80	0.53	0.64	9935
						1	0.65	0.87	0.75	10065
						accuracy			0.70	20000
ļ						macro avg		0.70	0.69	20000
						weighted avg	0,73	0.70	0.69	20000
✓	✓			0.70125	131 s					
							precision	recall	f1-score	support
						0	0.80	0.53	0.64	9935
						1	0.65	0.87	0.75	10065
						accuracy			0.70	20000
ļ						macro avg				20000
ļ						weighted avg	0.73	0.70	0.69	20000
✓	<b>√</b>	✓		0.70065	283.6s					
							precision	recall	f1-score	support
						0	0.83	0.50	0.63	9935
						1	0.65	0.90	0.75	10065
ļ						accuracy			0.70	20000
ļ						macro avg	0.74	0.70	0.69	20000
						weighted avg	0.74	0.70	0.69	20000

<b>√</b>	<b>√</b>	Porter stemmer	0.6571	187.3s					
						precision	recall	f1-score	support
					0	0.75	0.46	0.57	9935
					1	0.62	0.85	0.71	10065
					2000 CONTRACTOR (CONTRACTOR (C				
					accuracy	AND AND ADDRESS OF THE PARTY OF	0.000	0.66	20000
					macro avg weighted avg	0.68 0.68	0.66		20000
					weighten avg	0.00	0.00	0.04	20000
<b>√</b>	<b>✓</b>	Lancaster	0.68435	163.9s					
		Stemmer			N-83	precision	recall	f1-score	support
					2	2.22	10 BES		
					0	0.71	0.62	0.66 0.70	9935 10065
					-	0.07	0.75	0.70	10003
					accuracy			0.68	20000
					macro avg		0.68		20000
					weighted avg	0.69	0.68	0.68	20000
<b>√</b>		Lancaster	0.66795	96.72s					
		Stemmer				precision	recall	f1-score	support
					0	0.69	0.59	0.64	9935
					1		0.74		
					accuracy			0.67	20000
					macro avg	0.67	0.67		
					weighted avg	0.67	0.67	0.67	20000
	<b>✓</b>	Snowball	0.66065	160.9s					
·		Stemmer				precision	recall	f1-score	support
					0	0.76	0.46	0.58	9935
					1		0.86		
					accuracy			0.66	20000
					macro avg		0.66		
					weighted avg	0.69	0.66	0.65	20000
<b>√</b>	<b>√</b>	Spacy	0.70485	738.4s					
•		Spacy	0.70403	7.50.73		precision	recall	f1-score	support
					e	0.80	0.54	0.64	9935
					1		0.87	0.75	10065
					accuracy	行		0.70	20000
					macro avg		0.70	0.70	20000
					weighted avg	0.73	0.70	0.70	20000

# 2- TEXTBLOB

Stop word	Tokenize	Pos	Stemming/ Lemmatization	Accuracy	Time Taken	Classification	1 Report			
<b>√</b>				0.70535	31.3s					
							precision	recall	f1-score	support
						e 1	0.88	0.47	0.61	9935
						1	0.64	0.94	0.76	10065
						accuracy	,		0.71	20000
						macro avg weighted avg		0.70 0.71		
						weighted avg	0.70	0.71	0.05	20000
<b>✓</b>	<b>√</b>	1		0.7099	75.8 s					
	·						precision	recall	f1-score	support
						0	0.89	0.48	0.62	9935
						1	0.65	0.94	0.77	10065
						accuracy			0.71	20000
						macro avg	0.77	0.71	0.69	20000
						weighted avg	0.77	0.71	0.69	20000

_	<b>✓</b>	<b>✓</b>		0.60855	374.5s					
·	,	·					precision	recall	f1-score	support
						0	0.65	0.46	0.54	9935
						1		0.76	0.66	10065
						0.53	1/5/15/5	831916	0.50	1,3 68 65.50
						accuracy			0.61	20000
						macro avg	0.62	0.61	0.60	20000
						weighted avg	0.62	0.61	0.60	20000
<b>✓</b>	<b>√</b>		Porter stemmer	0.65375	146.8s					
<b>v</b>	V		1 Offer Steffiner	0.03373	140.65					
							precision	recall	f1-score	support
						0	0.80	0.40	0.53	9935
						1	0.60	0.90	0.72	10065
						accuracy			0.65	20000
						macro avg		0.65		20000
						weighted avg	0.70	0.65	0.63	20000
✓	✓		Lancaster Stemmer	0.6676	118.2s					
							precision	recall	f1-score	support
						0	0.75	0.50	0.60	9935
						1	0.63	0.83	0.72	10065
						accuracy			0.67	20000
						macro avg		0.67		
						weighted avg	0.69	0.67	0.66	20000
<b>✓</b>	<b>√</b>		Snowball Stemmer	0.65515	116.2s					
							precision	recall	f1-score	support
						0	0.81	0.40	0.53	9935
						1	0.60	0.91	0.73	10065
						accuracy			0.66	20000
						macro avg		0.65		
						weighted avg	0.71	0.66	0.63	20000
<b>✓</b>	<b>✓</b>		Spacy	0.7253						
						Classification	on Report:			
							precision	recall	f1-score	support
						negative	0.86	0.53	0.66	9935
						positive		0.91		10065
						accuracy			0.73	20000
						macro avg		0.72		20000
						weighted avg	0.76	0.73	0.71	20000

# 3- VADER

Stop word	Tokenize	Pos	Stemming/ Lemmatization	Accuracy	Time Taken	Classification	Report			
				0.6998	65.13s					
							VEC DER			
							precision	recall	f1-score	support
						0	0.79	0.53	0.64	9935
						0	0.65	0.86	0.74	10065
						accuracy			0.70	20000
						macro avg	0.72	0.70	0.69	20000
						weighted avg	0.72	0.70	0.69	20000
✓				0.67825	48.84s					
						meening to ever				
						***************************************	precision	recall	f1-score	support
						0	0.79	0.48	0.60	9935
						1	0.63	0.88	0.73	10065
						accuracy			0.68	20000
						macro avg	0.71	0.68	0.66	20000
						weighted avg	0.71	0.68	0.66	20000

<b>√</b>	$\checkmark$			0.68105	89.58s					
•	•			0.000	0,100		precision	recall	f1-score	support
							0.70	0.40	0.50	0025
						0		0.49 0.87		9935 10065
						255119251			0.68	20000
						accuracy macro avg		0.68		20000
						weighted avg		0.68		20000
<b>√</b>	<b>✓</b>	✓		0.6763	87.09s					
	•	•			0,10,2		precision	recall	f1-score	support
						0	0.79	0.48	0.59	9935
						1		0.87	0.73	10065
						accuracy			0.68	20000
						macro avg	0.71	0.68	0.66	20000
						weighted avg	0.71	0.68	0.66	20000
<b>√</b>	<b>√</b>		Porter stemmer	0.67275	207.6s					
•	•		1 Sitter Sterminer	0.07273	207.00		precision	recall	f1-score	support
						1127				
						0	0.76	0.50	0.61 0.72	9935 10065
						1000				
						accuracy macro avg		0.67	0.67 0.66	20000 20000
						weighted avg	0.69	0.67	0.66	20000
✓	✓		Lancaster Stemmer	0.69225	167.5s					
							precision	recall	f1-score	support
						0	0.72	0.63	0.67	9935
						1	0.67	0.76	0.71	10065
						accuracy			0.69	20000
						macro avg	0.70	0.69	0.69	20000
						weighted avg	0.70	0.69	0.69	20000
<b>√</b>	<b>√</b>		Snowball Stemmer	0.6807	162.8s					
,	·						precision	recall	f1-score	support
						0	0.75	0.53	0.62	9935
						1	0.64	0.83	0.72	10065
						accuracy			0.68	20000
						macro avg	0.70	0.68	0.67	20000
						weighted avg	0.70	0.68	0.67	20000
			Spacy	0.6762	927.4s					
✓	✓		Spacy	0.0702	927.4S		0.1			
							precision	recall	f1-score	support
						0	0.78 0.63	0.49	0.60	9935 10065
						6,4100	0.03	0.00		
						accuracy	6.70	0 63	0.68	20000
						macro avg weighted avg	0.70	0.67 0.68	0.66	20000
						mean says	2170	-100	2,00	

## PLM's

## 1- Distilbert

Stop Words	Tokenize	Pos	Stemmer/ Lemmatizer	Classification Report & Accuracy
No	No	No	No	max_length = 128 batch_size = 16 scheduler = get_linear_schedule_with_warmup optimizer = AdamW(roberta_model.parameters(), lr=2e-5  Epoch 1/3 - Loss: 0.3393317736208439 Epoch 2/3 - Loss: 0.20182555637632807 Epoch 3/3 - Loss: 0.11509952055377265 Accuracy: 0.85695
No	No	No	No	max_length = 128, batch_size = 16 scaler = GradScaler() num_epochs = 3 max_grad_norm = 1.0

			warmup_steps = 0.1 * len(train_dataloader) scheduler = get_linear_schedule_with_warmup optimizer = AdamW(roberta_model.parameters(), lr=2e-5)
			Epoch 1/3, Batch 100/1875, Loss: 0.10172989226877689 Epoch 1/3, Batch 308/1875, Loss: 0.0676278891726017 Epoch 1/3, Batch 408/1875, Loss: 0.0676278891726017 Epoch 1/3, Batch 408/1875, Loss: 0.0789503897289914 Epoch 1/3, Batch 608/1875, Loss: 0.0789503897289914 Epoch 1/3, Batch 608/1875, Loss: 0.0789503897289914 Epoch 1/3, Batch 608/1875, Loss: 0.08237287698418 Epoch 1/3, Batch 808/1875, Loss: 0.06237287689818 Epoch 1/3, Batch 808/1875, Loss: 0.06637328784716625 Epoch 1/3, Batch 1080/1875, Loss: 0.06637328784716625 Epoch 1/3, Batch 1180/1875, Loss: 0.066733185806488937 Epoch 1/3, Batch 1180/1875, Loss: 0.06664066761732181 Epoch 1/3, Batch 1280/1875, Loss: 0.08882725395262242 Epoch 1/3, Batch 1380/1875, Loss: 0.08882725395262242 Epoch 1/3, Batch 1380/1875, Loss: 0.88882775395262242 Epoch 1/3, Batch 1680/1875, Loss: 0.88882775395262242 Epoch 1/3, Batch 1680/1875, Loss: 0.88882775395262242 Epoch 1/3, Batch 1880/1875, Loss: 0.88882775395262242 Epoch 1/3, Batch 1880/1875, Loss: 0.88882775395262242 Epoch 1/3, Batch 1080/1875, Loss: 0.88883774144181143 Epoch 1/3, Batch 1080/1875, Loss: 0.88883774144181143 Epoch 1/3, Batch 1080/1875, Loss: 0.08884419979453087 Epoch 2/3, Batch 208/1875, Loss: 0.05884556879663166 Epoch 2/3, Batch 208/1875, Loss: 0.05882656876663166 Epoch 2/3, Batch 208/1875, Loss: 0.05882656876663166 Epoch 2/3, Batch 208/1875, Loss: 0.05882331924399671936 Epoch 2/3, Batch 208/1875, Loss: 0.058823319243999671936 Epoch 2/3, Batch 208/1875, Loss: 0.05882331924399971196 Epoch 2/3, Batch 208/1875, Loss: 0.05882331924399971196 Epoch 2/3, Batch 208/1875, Loss: 0.05882331924399971196 Epoch 2/3, Batch 1080/1875, Loss: 0.058823319243999791196 Epoch 2/3, Batch 1080/1875, Loss: 0.0888233194399979116 Epoch 2/3, Batch 1080/1875, Loss: 0.088823319439979116 Epoch 2/3, Batch 1080/1875, Loss: 0.088823319439979116 Epoch 2/3, Batch 1080/1875, Loss: 0.089823319439979116 Epoch 2/3, Batch 1080/1875, Loss: 0.089823319439979116 Epoch 2/3, Batch 1080/1875, Loss: 0.0898233194412918 Epoch 2/3, Batch 1080/1875, Loss: 0.08982331943999
			Epoch 3/3, Batch 1780/1875, Loss: 0.828796019285917282 Epoch 3/3, Batch 1880/1875, Loss: 0.81684528469983946 Accuracy: 0.8848
No No	No	No	max_length = 128, batch_size = 16 scaler = GradScaler() num_epochs = 3 max_grad_norm = 1.0 warmup_steps = 0.1 * len(train_dataloader) scheduler = get_linear_schedule_with_warmup optimizer = SGD(distilbert_model.parameters(), lr=0.01, momentum=0.9)

	Epoch 1/3, Batch 100/1875, Loss: 0.638399543762207 Epoch 1/3, Batch 200/1875, Loss: 0.494405632019043 Epoch 1/3, Batch 300/1875, Loss: 0.422783145904541 Epoch 1/3, Batch 400/1875, Loss: 0.4123725938796997 Epoch 1/3, Batch 500/1875, Loss: 0.3828009366989136
	Epoch 1/3, Batch 600/1875, Loss: 0.38030768394470216 Epoch 1/3, Batch 700/1875, Loss: 0.37837531328201296 Epoch 1/3, Batch 800/1875, Loss: 0.4078387808799744
	Epoch 1/3, Batch 900/1875, Loss: 0.3658698034286499 Epoch 1/3, Batch 1000/1875, Loss: 0.3521576488018036 Epoch 1/3, Batch 1100/1875, Loss: 0.344450261592865
	Epoch 1/3, Batch 1288/1875, Loss: 8.3489921288498295 Epoch 1/3, Batch 1388/1875, Loss: 8.3452486753463745 Epoch 1/3, Batch 1488/1875, Loss: 8.36882784782485964 Epoch 1/3, Batch 1588/1875, Loss: 8.32819894883868185
	Epoch 1/3, Batch 1600/1875, Loss: 0.33297763466835023 Epoch 1/3, Batch 1700/1875, Loss: 0.36296723008155823 Epoch 1/3, Batch 1800/1875, Loss: 0.3260021793842316
	Epoch 2/3, Batch 100/1875, Loss: 8.32985604763031007 Epoch 2/3, Batch 200/1875, Loss: 8.31528124570846555 Epoch 2/3, Batch 300/1875, Loss: 8.28176330678415298
	Epoch 2/3, Batch 408/1875, Loss: 8.3811583805827771 Epoch 2/3, Batch 508/1875, Loss: 0.280997711122036 Epoch 2/3, Batch 608/1875, Loss: 0.28678333163261416 Epoch 2/3, Batch 708/1875, Loss: 0.2576856642961502
	Epoch 2/3, Batch 800/1875, Loss: 0.26707930088043214 Epoch 2/3, Batch 900/1875, Loss: 0.2569731563329697 Epoch 2/3, Batch 1000/1875, Loss: 0.2364058029651642
	Epoch 2/3, Batch 1188/1875, Loss: 8.23387652886919515 Epoch 2/3, Batch 1288/1875, Loss: 8.23338297843933186 Epoch 2/3, Batch 1388/1875, Loss: 8.278963219666481 Epoch 2/3, Batch 1488/1875, Loss: 8.272921628758427
	Epoch 2/3, Batch 1580/1875, Loss: 8.221527493808093652 Epoch 2/3, Batch 1680/1875, Loss: 8.27228757441843855 Epoch 2/3, Batch 1780/1875, Loss: 8.25826678562164386
	Epoch 2/3, Batch 1800/1875, Loss: 8.22898086655139923 Epoch 3/3, Batch 100/1875, Loss: 0.24657639414072036 Epoch 3/3, Batch 200/1875, Loss: 0.2323412564396858 Epoch 3/3, Batch 300/1875, Loss: 0.19998804360628128
	Epoch 3/3, Batch 408/1875, Loss: 0.21428256346464156 Epoch 3/3, Batch 508/1875, Loss: 0.18530170395970344 Epoch 3/3, Batch 608/1875, Loss: 0.19853367209434508
	Epoch 3/3, Batch 708/1875, Loss: 8.1722753517329693 Epoch 3/3, Batch 808/1875, Loss: 8.19756732687354087 Epoch 3/3, Batch 908/1875, Loss: 8.18920622908128364 Epoch 3/3, Batch 1080/1875, Loss: 8.16169812321662902
	Epoch 3/3, Batch 1800/1875, Loss: 0.1534866251051426 Epoch 3/3, Batch 1200/1875, Loss: 0.15827347204089165 Epoch 3/3, Batch 1300/1875, Loss: 0.16716860368847847
	Epoch 3/3, Batch 1400/1875, Loss: 8.2186095404624939 Epoch 3/3, Batch 1500/1875, Loss: 8.1814175758510828 Epoch 3/3, Batch 1600/1875, Loss: 8.2003120781481266 Epoch 3/3, Batch 1700/1875, Loss: 8.18933923915028572 Epoch 3/3, Batch 1800/1875, Loss: 8.17760580018162728 Accuracy: 8.87725
No No No No	max_length = 128, batch size = 16
	scaler = GradScaler() num_epochs = 3
	max_grad_norm = 1.0 warmup_steps = 0.1 * len(train_dataloader) optimizer = AdamW(roberta_model.parameters(), lr=2e-5
	scheduler = ExponentialLR(optimizer roberta, gamma=0.9)  Epoch 1/3, Batch 100/1875, Loss: 0.6952365112304687  Epoch 1/3, Batch 200/1875, Loss: 0.695234130859375  Epoch 1/3, Batch 300/1875, Loss: 0.6951004028320312
	Epoch 1/3, Batch 400/1875, Loss: 0.6953091430664062 Epoch 1/3, Batch 500/1875, Loss: 0.6928515625 Epoch 1/3, Batch 600/1875, Loss: 0.694031640625 Epoch 1/3, Batch 700/1875, Loss: 0.694031640625
	Epoch 1/3, Batch 880/1875, Loss: 0.695142822265625 Epoch 1/3, Batch 900/1875, Loss: 0.6942910766601562 Epoch 1/3, Batch 1000/1875, Loss: 0.69353759765625 Epoch 1/3, Batch 1100/1875, Loss: 0.6954696655273438 Epoch 1/3, Batch 1200/1875, Loss: 0.6954240112304687
	Epoch 1/3, Batch 1300/1875, Loss: 0.6942092895507812 Epoch 1/3, Batch 1400/1875, Loss: 0.69363525390625 Epoch 1/3, Batch 1500/1875, Loss: 0.693816235351563 Epoch 1/3, Batch 1600/1875, Loss: 0.693816235351563
	Epoch 1/3, Batch 1700/1875, Loss: 0.6942047119140625 Epoch 1/3, Batch 1800/1875, Loss: 0.6917852783203124 Epoch 2/3, Batch 1000/1875, Loss: 0.695845947265625 Epoch 2/3, Batch 2000/1875, Loss: 0.69532032470708313 Epoch 2/3, Batch 3000/1875, Loss: 0.6943658447265625
	Epoch 2/3, Batch 488/1875, Loss: 0.6958838078125 Epoch 2/3, Batch 508/1875, Loss: 0.692849569546875 Epoch 2/3, Batch 608/1875, Loss: 0.6933535766601563
	Epoch 2/3, Batch 700/1875, Loss: 0.6917349243164063
	Epoch 2/3, Batch 700/1875, Loss: 0.6917349243164863 Epoch 2/3, Batch 800/1875, Loss: 0.6926011352539062 Epoch 2/3, Batch 900/1875, Loss: 0.69296875 Epoch 2/3, Batch 1000/1875, Loss: 0.6924227905273438 Epoch 2/3, Batch 1100/1875, Loss: 0.69147705078125

				Epoch 2/3, Batch 1000/1875, Loss: 0.69147705073438  Epoch 2/3, Batch 1100/1875, Loss: 0.69147705078125  Epoch 2/3, Batch 1200/1875, Loss: 0.6934478759765625  Epoch 2/3, Batch 1200/1875, Loss: 0.6934478759765625  Epoch 2/3, Batch 1400/1875, Loss: 0.6943798228125  Epoch 2/3, Batch 1500/1875, Loss: 0.6943798228125  Epoch 2/3, Batch 1500/1875, Loss: 0.6943798228125  Epoch 2/3, Batch 1700/1875, Loss: 0.693489990234375  Epoch 2/3, Batch 1800/1875, Loss: 0.6932604470117188  Epoch 3/3, Batch 1800/1875, Loss: 0.695127878656234438  Epoch 3/3, Batch 200/1875, Loss: 0.6953207397460938  Epoch 3/3, Batch 300/1875, Loss: 0.6953873974609375  Epoch 3/3, Batch 400/1875, Loss: 0.695480698775  Epoch 3/3, Batch 600/1875, Loss: 0.6954806982421875  Epoch 3/3, Batch 600/1875, Loss: 0.6948406982421875  Epoch 3/3, Batch 600/1875, Loss: 0.6948406982421875  Epoch 3/3, Batch 800/1875, Loss: 0.694816132812  Epoch 3/3, Batch 800/1875, Loss: 0.694816870117187  Epoch 3/3, Batch 800/1875, Loss: 0.6954916870117187  Epoch 3/3, Batch 800/1875, Loss: 0.6952833129852812  Epoch 3/3, Batch 1100/1875, Loss: 0.695283312982812  Epoch 3/3, Batch 1100/1875, Loss: 0.6935324096679687  Epoch 3/3, Batch 1100/1875, Loss: 0.6935324096679687  Epoch 3/3, Batch 1100/1875, Loss: 0.6938328317  Epoch 3/3, Batch 1200/1875, Loss: 0.6958328337
No	No	No	No	Finetuning
				Accuracy: 0.93105 F1-score: 0.9313349599163471 Confusion matrix: [[9269 666] [713 9352]]
				precision recall f1-score support
				negative 0.93 0.93 0.93 9935
				positive 0.93 0.93 0.93 18865
				accuracy 0.93 28080 macro avg 0.93 0.93 28080 weighted avg 0.93 0.93 28080
				Time required to fine-tune: 3672.2959537506104

### 2- Roberta

Stop Words	Tokenize	Pos	Stemmer/ Lemmatizer	Classification Report & Accuracy
No	No	No	No	max_length = 128, batch_size = 16 optimizer = AdamW(roberta_model.parameters(), lr=2e-5 scheduler = get_linear_schedule_with_warmup Epoch 1/3 - Loss: 0.6953695392773946 Epoch 2/3 - Loss: 0.6947482631047567 Epoch 3/3 - Loss: 0.6946594745000203 Accuracy: 0.49675
				max_length = 128, batch_size = 16 scaler = GradScaler(), num_epochs = 3, max_grad_norm = 1.0 warmup_steps = 0.1 * len(train_dataloader) optimizer = AdamW(roberta_model.parameters(), lr=2e-5 scheduler = get_linear_schedule_with_warmup  Epoch_1/3, Batch_100/1875, Loss: 0.699346313476562 Epoch_1/3, Batch_200/1875, Loss: 0.699346313476562 Epoch_1/3, Batch_200/1875, Loss: 0.6976708984375 Epoch_1/3, Batch_500/1875, Loss: 0.6976708984375 Epoch_1/3, Batch_500/1875, Loss: 0.6976708984375 Epoch_1/3, Batch_500/1875, Loss: 0.6958966796875 Epoch_1/3, Batch_500/1875, Loss: 0.6958966796875 Epoch_1/3, Batch_500/1875, Loss: 0.6958968585937 Epoch_1/3, Batch_900/1875, Loss: 0.6958968585937 Epoch_1/3, Batch_1000/1875, Loss: 0.6958968585937 Epoch_1/3, Batch_1000/1875, Loss: 0.6938188171386719 Epoch_1/3, Batch_11200/1875, Loss: 0.6938188171386719 Epoch_1/3, Batch_11200/1875, Loss: 0.6938188371386719 Epoch_1/3, Batch_11800/1875, Loss: 0.6938188371386719 Epoch_1/3, Batch_11800/1875, Loss: 0.69746890833390312 Epoch_1/3, Batch_11800/1875, Loss: 0.6974689683396339614848 Epoch_1/3, Batch_1800/1875, Loss: 0.697468761367188 Epoch_1/3, Batch_1800/1875, Loss: 0.6978876287841797 Epoch_1/3, Batch_1800/1875, Loss: 0.6978976287841797 Epoch_1/3, Batch_1800/1875, Loss: 0.5938976287841797 Epoch_1/3, Batch_1800/1875, Loss: 0.59389868893894754 Epoch_1/3, Batch_1800/1875, Loss: 0.59389868893894754 Epoch_1/3, Batch_1800/1875, Loss: 0.593898688893894754 Epoch_1/3, Batch_1800/1875, Loss: 0.513897238478988685 Epoch_1/3, Batch_1800/1875, Loss: 0.513897238478988685

				Epoch 2/3, Batch 900/1875, Loss: 0.5137767028808594 Epoch 2/3, Batch 1000/1875, Loss: 0.503767028808233 Epoch 2/3, Batch 1100/1875, Loss: 0.483096604247229 Epoch 2/3, Batch 1200/1875, Loss: 0.483096604247229 Epoch 2/3, Batch 1300/1875, Loss: 0.4923399060925849 Epoch 2/3, Batch 1800/1875, Loss: 0.45678155168762286 Epoch 2/3, Batch 1800/1875, Loss: 0.4567815518848999 Epoch 2/3, Batch 1800/1875, Loss: 0.458763137376884 Epoch 3/3, Batch 1000/1875, Loss: 0.456718137386845337 Epoch 3/3, Batch 1000/1875, Loss: 0.4475750160217285 Epoch 3/3, Batch 4000/1875, Loss: 0.4475750160217285 Epoch 3/3, Batch 4000/1875, Loss: 0.4475777829037473 Epoch 3/3, Batch 4000/1875, Loss: 0.44276777829037473 Epoch 3/3, Batch 500/1875, Loss: 0.44276777829037473 Epoch 3/3, Batch 500/1875, Loss: 0.4423101838761138916 Epoch 3/3, Batch 800/1875, Loss: 0.4423101838761138916 Epoch 3/3, Batch 800/1875, Loss: 0.42131038761138916 Epoch 3/3, Batch 800/1875, Loss: 0.42764220123291 Epoch 3/3, Batch 1000/1875, Loss: 0.42760888 Epoch 3/3, Batch 1000/1875, Loss: 0.4276107765197756 Epoch 3/3, Batch 1100/1875, Loss: 0.4276107765197756 Epoch 3/3, Batch 1200/1875, Loss: 0.42376107765197756 Epoch 3/3, Batch 1800/1875, Loss: 0.3934081777352905 Epoch 3/3, Batch 1800/1875, Loss: 0.3934081277
				Accuracy: 0.7854
N	N	N.T.	N	
No	No	No	No	max   length = 128, batch   size = 16   scaler = GradScaler()   num   epochs = 3   max   grad   norm = 1.0   warmup   steps = 0.1 * len(train_dataloader)   optimizer = SGD(distrilbert_model.parameters(), lr=0.01, momentum=0.9)   scheduler   get_linear_schedule_with_warmup   sepoch   1/3,   Batch   108/1875,   Loss: 0.7825511932373046     spoch   1/3,   Batch   108/1875,   Loss: 0.7825511932373046     spoch   1/3,   Batch   208/1875,   Loss: 0.7825511932373046     spoch   1/3,   Batch   208/1875,   Loss: 0.7825511932373046     spoch   1/3,   Batch   208/1875,   Loss: 0.782581873515515     spoch   1/3,   Batch   208/1875,   Loss: 0.78268895793135     spoch   1/3,   Batch   128/1875,   Loss: 0.7826889579315     spoch   1/3,   Batch   128/1875,   Loss: 0.78268895895875     spoch   1/3,   Batch   128/1875,   Loss: 0.78268895895875     spoch   1/3,   Batch   128/1875,   Loss: 0.78268895895875     spoch   1/3,   Batch   128/1875,   Loss: 0.782688958958675     spoch   1/3,   Batch   128/1875,   Loss: 0.78268895895875     spoch   1/3,   Batch   128/1875,   Loss: 0.78268895895875     spoch   1/3,   Batch   208/1875,   Loss: 0.78268841562156875     spoch   1/3,   Batch   208/1875,   Loss: 0.78268841562156875     spoch   1/3,   Batch   208/1875,   Loss: 0.7826884156216875     spoch   1/3,   Batch   208/1875,   Loss: 0.7826884156216875     spoch   1/3,
No	No	No	No	$\max$ length = 128,

			batch size = 16 scaler = GradScaler() num_epochs = 3 max_grad_norm = 1.0 warmup_steps = 0.1 * len(train_dataloader) optimizer = AdamW(roberta_model_parameters(), lr=2e-5 scheduler = ExponentialLR(optimizer_roberta_gamma=0.9)  Epoch 1/3, Batch 100/1875, Loss: 0.780837249758594 Epoch 1/3, Batch 200/1875, Loss: 0.6993628613281275 Epoch 1/3, Batch 400/1875, Loss: 0.6993628613281275 Epoch 1/3, Batch 400/1875, Loss: 0.6993628613281275 Epoch 1/3, Batch 400/1875, Loss: 0.6993628613281255 Epoch 1/3, Batch 600/1875, Loss: 0.6993628613281256 Epoch 1/3, Batch 600/1875, Loss: 0.69840825957314843 Epoch 1/3, Batch 600/1875, Loss: 0.6984039575195512 Epoch 1/3, Batch 600/1875, Loss: 0.698628573296256 Epoch 1/3, Batch 1000/1875, Loss: 0.698628573296256 Epoch 1/3, Batch 1000/1875, Loss: 0.69862875297656 Epoch 1/3, Batch 1000/1875, Loss: 0.698628752729625 Epoch 1/3, Batch 1200/1875, Loss: 0.698628727272489219 Epoch 1/4, Batch 1200/1875, Loss: 0.698628737272489219 Epoch 1/4, Batch 1200/1875, Loss: 0.698628737675171875 Epoch 1/4, Batch 1200/1875, Loss: 0.69862876767171875 Epoch 1/4, Batch 1200/1875, Loss: 0.69989776489257813 Epoch 1/4, Batch 1200/1875, Loss: 0.6998977648925791296 Epoch 1/4, Batch 1200/1875, Loss: 0.6998977648925791296 Epoch 1/4, Batch 1200/1875, Loss: 0.6998977648925791296 Epoch 1/4, Batch 1200/1875, Loss: 0.698985786912185 Epoch 1/4, Batch 1200/1875, Loss: 0.698895786912185 Epoch 1/4, Batch 1200/1875, Loss: 0.6988678649892349118 Epoch 1/4, Batch 1200/1875, Loss: 0.698867864989827998048 Epoch 1/4, Batch 1200/1875, Loss: 0.69886786498675 Epoch 1/4, Batch 1200/1875, Loss: 0.69886786498679 Epoch 1/3, Batch 400/1875, Loss: 0.6988678649892344 Epoch 1/4, Batch 1200/1875, Loss: 0.698867864992348 Epoch 1/3, Batc
No No	No	No	Finetuning Accuracy: 8.95255 F1-score: 8.9531149646756584 Confusion matrix: [[9485 538] [ 419 9646]]
			precision recall f1-score support  negative 8.96 8.95 8.95 9935 positive 8.95 8.96 8.95 18865
			accuracy 8.95 28080 macro avg 8.95 8.95 28080

# **Word Embedding:**

# 1- Word2Vec

Stop Words	Tokenize	Pos	Stemmer/ Lemmatizer	Classification Repo	ort & Accura	acy			
No	No	No	No						
				Decision Tree A	couraciu a	CAEA			
				Decision Tree A Decision Tree C			F-1		
					recision			support	
				P	recision	recall	TI-SCORE	Support	
				positive	0.65	0.63	0.64	9935	
				negative	0.64	0.66	0.65	10065	
				accuracy			0.65	20000	
				macro avg	0.65	0.65			
				weighted avg	0.65	0.65	0.65	20000	
				MANAGEMENT STREET					
				Random Forest A Random Forest C					
					recision			support	
				P	( CC1210))	1 CCGII	11-2001 E	Support.	
				positive	0.76	0.74	0.75	9935	
				negative					
				**************************************	7707		and the second		
				accuracy			0.75		
					0.75	0.75	0.75	20000	
				macro avg weighted avg	0.75	0.75	0.75	20000	
				Logistic Regress	ion Accura	cv. a 9020	e c		
				Logistic Regress					
					recision			support	
				positive	9.81	0.79	0.80	9935	
				negative				10065	
				accuracy			0.80	20000	
				macro avg	0.80	0.80	0.80	20000	
				weighted avg	0.80	0.80	0.80	20000	
				k-NN Accuracy: k-NN Classifica					
					orecision		f1 score	support	
				35	N CCT3TOII	1 CCOII	11-3001 C	Suppor C	
				positive	0.69	0.73	0.71	9935	
				negative				10065	
				accuracy			0.70	20000	
				macro avg	0.70	0.70			
				weighted avg					
				XGBoost Accurac					
				XGBoost Classi			£1	cuppent	
					precision	recall	f1-score	support	
				positive	0.79	0.77	0.78	9035	
				The second secon					
	1			negative	0.77	0.79	0.78	10065	
				accuracy			0 70	20000	
				accuracy macro avg	0.78	0.78	0.78 0.78		

LightGBM Accuracy: 0.78095 LightGBM Classification Report:	
precision recall f1-score support  positive 0.79 0.77 0.78 9935 negative 0.77 0.80 0.79 10065  accuracy 0.78 20000 macro avg 0.78 0.78 0.78 20000 weighted avg 0.78 0.78 0.78 20000	
negative 0.77 0.80 0.79 10065  accuracy 0.78 20000 macro avg 0.78 0.78 20000 weighted avg 0.78 0.78 20000	
negative 0.77 0.80 0.79 10065  accuracy 0.78 20000 macro avg 0.78 0.78 20000 weighted avg 0.78 0.78 20000	
macro avg 0.78 0.78 20000 weighted avg 0.78 0.78 20000	
macro avg 0.78 0.78 20000 weighted avg 0.78 0.78 20000	
weighted avg 0.78 0.78 20000	
Gradient Boosting Accuracy: 0.76545 Gradient Boosting Classification Report:	
precision recall f1-score support	
2004	
positive 0.77 0.75 0.76 9935 negative 0.76 0.78 0.77 10065	
negative 0.76 0.78 0.77 10065	
accuracy 0.77 20000	
macro avg 0.77 0.77 20000	
weighted avg 0.77 0.77 20000	
Neural Network Accuracy: 0.80475	
Neural Network Acturacy: 0.80475 Neural Network Classification Report:	
precision recall f1-score support	
positive 0.82 0.78 0.80 9935 negative 0.79 0.83 0.81 10065	
negative 0.75 0.63 0.61 10005	
accuracy 0.80 20000	
macro avg 0.81 0.80 0.80 20000	
weighted avg 0.81 0.80 0.80 20000	
No No No No Hyper Parameter Tuning	
Decision Tree Accuracy: 0.66285  Decision Tree Classification Report:	
precision recall f1-score support	
positive 0.66 0.67 0.66 9935 negative 0.67 0.66 0.66 10065	
negative 0.67 0.66 0.66 10065	
accuracy 0.66 20000	
macro avg 0.66 0.66 20000	
weighted avg 0.66 0.66 20000	
Random Forest Accuracy: 0.76105	
Random Forest Classification Report:	
precision recall f1-score support	
positive 0.77 0.73 0.75 9935	
negative 0.75 0.79 0.77 10065	
2.76	
accuracy 0.76 20000 macro avg 0.76 0.76 20000	
weighted avg 0.76 0.76 20000	
Logistic Regression Accuracy: 0.80655	
Logistic Regression Classification Report:	
precision recall f1-score support	
positive 0.81 0.79 0.80 9935	
positive 0.81 0.79 0.80 9935 negative 0.80 0.82 0.81 10065	
negative 0.80 0.82 0.81 10065	

		T		LightCDN Assuma	cur a 7016	1556			
				LightGBM Accura LightGBM Classi		port:			
					recision		f1-score	support	
				positive	0.70	0.76	0 77	9935	
				negative					
				accuracy.			0.78	20000	
				accuracy	0.70	0.70			
				macro avg weighted avg	0.70	0.78	0.78	20000	
				200 200 200 200 200 200 200 200 200 200			0.70	20000	
				Gradient Boosti Gradient Boosti			nort:		
					recision			support	
				P				30,000.0	
				positive	0.80	0.77	0.79	9935	
				negative					
				1.000 <del>   </del> 1.000 1.000					
				accuracy			0.79		
				macro avg	0.79	0.79	0.79	20000	
				macro avg weighted avg	0.79	0.79	0.79	20000	
				Neural Network					
1				Neural Network			ort:		
					precision			support	
				positive	0.79	0.83	0.81	9935	
				negative	0.83	0.78	0.80	10065	
				accuracy			0.81	20000	
				macro avg		0.81			
				weighted avg					
				k-NN Accuracy:	0.7159				
				k-NN Classific	tation Repo	rt:			
				- 1	precision	recall	f1-score	support	
				positive	0.70	0.75	0.72	9935	
				negative	0.73	0.68		10065	
				accuracy			0.72	20000	
				macro avg		0.72			
				weighted avg					
				XGBoost Accura	ecv: 0.7923				
1				XGBoost Classi					
				ASSESS OF THE PROPERTY OF THE PARTY OF THE P	precision	recall	f1-score	support	
				positive	0.80	0.77	0.79	9935	
				negative	0.78	0.81	0.80	10065	
1							0.000000000	200000	
				accuracy			0.79	20000	
				macro avg	0.79	0.79	0.79	20000	
				weighted avg	0.79	0.79	0.79	20000	
No	No	No	No	Parameters for the	model and tra	ining			
		1		max features = 200		0			
				maxlen = 200	<del>-</del>				
				embedding dim = 3	300				
				lstm units = 128					
1				filters = 64					
1				kernel size = 5					
				batch size = 32					
1				epochs = 5					
1				lstm branch = LST	M(letm unit	dropout-(	) 2 ransemant	dropout=0.2)	(embedding layer)
1				cnn_branch = Conv					
				cnn_branch = Conv				ra gembeddin	g_iayei)
				merged = Concaten					
				output = Dense(1, a					
				Juipui – Delise(1, 8	vai1011— 81}	Surora )(IIICI	500)		
		<u> </u>		l					

		1		Eooch 1/5
N.	N.	N	M.	750/750 [====================================
No	No	No	No	lstm_branch = LSTM(lstm_units, dropout=0.2, recurrent_dropout=0.2)(embedding_layer)   cnn_branch = Conv1D(filters, kernel_size, activation='relu')(embedding_layer)   cnn_branch = GlobalMaxPooling1D()(cnn_branch)   merged = Concatenate()([lstm_branch, cnn_branch])   merged = Dropout(0.5)(merged)   output = Dense(1, activation='sigmoid')(merged)    Epoch 1/5
No	No	No	No	max_features = 20000 maxlen = 200 embedding_dim = 300 lstm_units = 128 filters = 64 kernel_size = 5 batch_size = 32 epochs = 10  Epoch 1/18 758/758 [====================================
No	No	No	Porter Stemmer	[nltk_data] Downloading package punkt to /usr/share/nltk_data [nltk_data] Package punkt is already up-to-date! Epoch 1/5 758/758 [====================================
No	No	No	Lancaster Stemmer	[nltk_data] Downloading package punkt to /usr/share/nltk_data [nltk_data] Fackage punkt is already up-to-date! Epoch 1/5 758/758 [====================================
No	No	No	Snowball Stemmer	Epoch 1/5 750/750 [====================================

# 2- Glove

Stop Words	Tokenize	Pos	Dimension	Epoch Results
No	No	No	100	750/750 [====================================
No	No	No	200	Epoch 1/5 750/750 [====================================
No	No	No	300	Epoch 1/5 750/750 [====================================
No	No	No	50	Epoch 1/5 750/750 [====================================

## **Hybrid Approaches**

### 1- Afinn + ML

Stop Words	Tokenize	Pos	Stemmer/ Lemmatizer	Vectorizer	Classification R	eport			
No	No	No	No	Tf- Idf Vectorizer	Logistic Regression Accuracy: 0.89005 Logistic Regression Classification Report:				
						precision	recall	f1-score	support
					negative	0.90	0.88	0.89	9935
					positive	0.88	0.90	0.89	10065
					accuracy			0.89	20000
					macro avg	0.89	0.89	0.89	20000
					weighted avg	0.89	0.89	0.89	20000

No	No	No	No	Tf- Idf					
INO	INO	INO	INO	Vectorizer	100 To 100 TO 100				
				VCCIOTIZEI	Decision Tree				
					Decision Tree	Classificat: precision			sunnort
						pi ccision	- CCUII	12-3001C	-uppor C
						0.71			
					positive	0.72	0.71	0.71	10065
					accuracy			0.71	20000
					macro avg	0.71	0.71	0.71	20000
					weighted avg	0.71	0.71	0.71	20000
					Random Forest Random Forest	Accuracy: 0. Classificati	.84345 ion Report	:	
						precision			support
					PORT AND AND ANY OF	0.00		0.01	0035
						0.83			9935 10065
					posteric	0.00	0.00	3107	
					accuracy		15/15/2	0.84	
					macro avg weighted avg	0.84			
					meralised dvs	0.01	0.01	0.07	
					Gradient Boos	ting Accumes	V. 0 91565		
					Gradient Boos				
						precision			support
					negative	Q 04	0 77	Q 01	9935
					positive	0.84 0.79	0.86	0.82	10065
					0)				
					accuracy macro ave	0.82	0.82	0.82	
					weighted avg				20000
					Neural Networ	k Accuracus	0.9643		
					Neural Networ			t:	
						precision			support
					negative	0.87	0.86	0.86	9935
					positive	0.87 0.86	0.87	0.87	10065
					accuracy			0.86	
					macro avg weighted avg	0.86	0.86	0.86	
					KNN Accuracy:		0.50	0.00	20000
					KNN Classific		t:		
						precision		f1-score	support
					negative	0.73	0.68	0.70	9935
					positive				
					accuracy			0.72	20000
					macro avg	0.72	0.72		
					weighted avg				
			2.7						
Yes	Yes	No	No	Count Vectorizer			_	200	
				VECTOTIZET	Logistic Regr				
					Logistic Regr	ression Clas precision			support
					nagativa	0.87	0 07	0.87	9935
					positive				
					accuracy			0.87	20000
					macro avg		0.87		
					weighted avg		0.87		
					Decision Tree				
					Decision Tree	classifica precision			support
					20063				
					negative				
					positive	0.73	0.71	0.72	10065
					accuracy			0.72	
						0.72			
					weighted avg	0.72	0.72	0.72	20000

positive accuracy	0.84 0.86	recall 0.86 0.84	f1-score 0.85 0.85		
positive accuracy macro avg	0.84 0.86	0.86 0.84	0.85 0.85	9935	
positive accuracy macro avg	0.86 0.85	0.84	0.85		
accuracy macro avg	0.85			10065	
macro avg	0.85		0.00		
				20000	
weighted avg		0.85	0.85	20000	
	0.85	0.85	0.85	20000	
Gradient Boos					
Gradient Boos	ting Classi	fication	Report:		
4501,040508,054,154 (\$44.9)	precision	recall	f1-score	support	
negative	0.84 0.79	0.76	0.80	9935	
positive	0.79	0.86	0.82	10065	
accuracy			0.81	20000	
macro avg	0.82	0.81	0.81	20000	
weighted avg	0.81	0.81	0.81	20000	
Neural Network Accuracy: 0.8696 Neural Network Classification Report:					
	precision			support	
	0.87				
positive	0.87	0.87	0.87	10065	
accuracy macro avg weighted avg			0.87	20000	
macro avg	0.87	0.87	0.87	20000	
weighted avg	0.87	0.87	0.87	20000	
KNN Accuracy:					
KNN Classifica					
	precision	recall f	f1-score	support	
negative	0.60 0.64	0.69	0.64	9935	
positive	0.64	0.54	0.59	10065	
accuracy			0.62	20000	
	0.62	0.62			
weighted avg					

Yes	Yes	Yes	No	Tf – Idf	1				
103	103	103	110	Vectorizer	Logistic Regr	ession Accura	acv: 0.86	515	
					Logistic Regr				
					-	precision			support
					negative	0.87	0.86	0.86	9935
					positive	0.87 0.86	0.87	0.87	10065
					accuracy			0.87	20000
					macro avg	0.87		0.87	20000
					weighted avg	0.87	0.87	0.87	20000
					Decision Tree			t:	
					33333311 1160	precision			support
						•			
						0.71			
					positive	0.72	0.71	0.72	10065
					accuracy			0.72	20000
					macro avg	0.72	0.72	0.72	20000
					weighted avg	0.72	0.72	0.72	20000
					Random Forest Random Forest			t:	
					1,6742	precision	recall	f1-score	support
						0.83			
					positive	0.85	0.83	0.84	10065
					accuracy			0.84	
					macro avg weighted avg	0.84	0.84	0.84	20000
					weighted avg	0.84	0.84	0.84	20000
					Gradient Boos Gradient Boos	ting Accuracy	y: 0.805	enort.	
					GI GOZETIC BOOS	precision			support
					negative	0.84	0.75	0.79	9935
					positive	0.78	0.86	0.82	10065
		1			accuracy			0.81	20000
						0.81			
					weighted avg				
					Neural Netwo	rk Accuracy: rk Classifica		rt.	
					MEDI AT MECHO	precision	The state of the s	f1-score	support
					negative	0.87	0.86	0.86	9935
						0.86			
					accuracy			0.87	20000
					\$100 (\$100 \$100 \$100 \$100 \$100 \$100 \$100	0.87	0.87		20000
					weighted avg		0.87		
					KNN Accuracy				
					KNW Classifi	ication Repor precision		1 f1.score	support
					negative positive				
					accuracy	0.61	0.6	0.61 0.60	
					weighted avg				
					1000 100 100 100 100 100 100 100 100 10				
	I	1		1	_1				

No	No	No	No	Tf-Idf	Improved ML	Parameters			
				Vectorizer	LogisticRegre				
					Logistic Regr Logistic Regr				
					LOGISTIC NEG	precision			support
						0.90			9935
					positive	0.89	0.91	0.90	10065
					accuracy			0.90	
						0.90			
					weighted avg			0.90	20000
					CHOLD MARK TRANS	556	The second second	=10,min_s	amples_split=5)
					Decision Tree Decision Tree			+•	
					55555500 11 55	precision			support
					negative	0.80	0.62	0.70	9935
					positive	0.69	0.84	0.76	10065
					accuracy			0.73	
					macro avg weighted avg	0.74	0.73 0.73		20000
									nax depth=10,
					min samples	,		315 100,11	im_depin 10,
					Random Forest	Accuracy: 0			
					Random Forest	Classificat: precision			support
					1.000	precision	100011	11-30016	Suppor c
					negative positive	0.85 0.80			9935 10065
					accuracy			0.83	20000
						0.83	0.83 0.83	0.83	20000 20000
					weighted avg				
					GradientBoos learning rate=			iators-100	),
					Gradient Boos				
					Gradient Boos	ting Classif precision			support
					1000	PARTITION OF THE PARTIT			
					negative positive	0.84 0.79	VI (1990)		9935 10065
					accuracy			0.82	20000
					macro avg	0.82			
					weighted avg	0.82	0.82	0.82	20000
					MLPClassifie alpha=0.0001	),		den_layer_	_sizes=(100,),
					Neural Networ	rk Accuracy:	0.8657	unt (	
					Neural Networ	rk Classifica precision		f1-score	support
					negative	0.87	0.86	0.86	9935
					positive				10065
					accuracy			0.87	20000
					macro avg	0.87		0.87	20000
					weighted avg	0.87	0.87	0.87	20000
				1					

No	No	No	No	Count Vectorizer	Improved ML Pa LogisticRegressi	ion(max_ite			
					Logistic Regress				
					Logistic Regress	sion Classi recision			support
					negative positive	0.88 0.88	0.87 0.88	0.88 0.88	9935 10065
					accuracy			0.88	20000
					macro avg weighted avg	0.88	0.88	0.88	20000
					DecisionTreeCla				
					Decision Tree A	Accuracy: 0.	73015	_	imples_spiit 3)
					Decision Tree C	lassificati precision			support
					negative	0.78	0.63	0.70	9935
					positive	0.69	0.83	0.76	
					accuracy			0.73	20000
					macro avg weighted avg			0.73	20000
					RandomForestC				20000 ax_denth=10
					min_samples_sp		-Sumawi	. 100,111	aopiii 10,
					Random Forest A				
					Random Forest C	lassificati precision			support
					negative	0.86	0.77	0.81	9935
					positive	0.79	0.88	0.83	
					accuracy			0.82	20000
					macro avg weighted avg		0.82 0.82	0.82 0.82	20000
					GradientBoostin learning_rate=0. Gradient Boosti	1, max_dep	oth=3)		,
					Gradient Boosti		cation Re	port:	support
					negative positive	0.85 0.79	0.77 0.86	0.81 0.83	9935 10065
					B-0-E-0-0-0-0-0-0-0-0-0-0-0-0-0-0-0-0-0-	70700	WAS THE TH	ROLL ROLL ROLL ROLL ROLL ROLL ROLL ROLL	0
					accuracy			0.82	
					macro avg weighted avg			0.82 0.82	
					MLPClassifier(nalpha=0.0001)			en_layer_	sizes=(100,),
					Neural Network	Accuracy: 0	.892		
					Neural Network (	Classificat:	ion Report		
					pr	recision	recall f	f1-score	support
					negative positive				9935 10065
								0.00	20000
					accuracy macro ave	0.89	0.89	0.89	
					macro avg weighted avg	0.89	0.89	0.89	20000
					MultinomialNB(	ACTUAL TO A STATE OF THE PARTY			
					Naive Bayes Acc				
					Naive Bayes Cla	assificatior precision		f1-score	support
					www.ffinen	0.05	0.05	0.05	9035
					negative positive			0.85 0.85	
					accuracy			0.85	20000
					macro avg			0.85	
					weighted avg	0.85	0.85	0.85	20000

No	No	No	Porter Stemmer	Tf-Idf Vectorizer	Improved ML P LogisticRegress Logistic Regre Logistic Regre	sion(max_ite	acy: 0.894 ification	7 Report:	support
					negative positive	0.90 0.89			
					accuracy			0.89	20000
					macro avg	0.89		0.89	
					weighted avg		0.89	0.89	20000
					Decision Tree A	Accuracy: 0.	7223	_	mples_split=5)
						precision			support
					negative	0.80	0.59	0.68	9935
					positive	0.68	0.85	0.76	10065
					accuracy			0.72	20000
					macro avg	0.74	0.72	0.72	
					Weighted avg				20000 av. donth=10
					RandomForestC min_samples_sp	\ <u> </u>	estimators	s=100, m	ax_deptn=10,
					Random Forest (		on Report:		support
					negative	0.85	0.78	0.82	9935
					positive	0.85 0.80	0.87	0.83	10065
					accuracy			0.83	20000
						0.83			
					weighted avg			0.82	20000
					GradientBoostir learning_rate=0			tors=100,	
					Gradient Boost				
					Gradient Boost	ing Classifi precision			support
					negative	0.84	0 77	0.80	9935
					positive		0.85	0.82	10065
					accuracy	32 220	13032033	0.81	20000
					macro avg weighted avg	0.81	0.81	0.81	20000
					easter # we bey those #CA				
					MLPClassifier(1 alpha=0.0001)	_		n_layer_	sızes=(100,),
					Neural Network			:	
					t	precision	recall f	1-score	support
					negative positive	0.87	0.86	0.87	9935
					positive	0.87	0.87	0.87	10065
					accuracy	0.07	0.07	0.87	20000
					macro avg weighted avg	0.87 0.87	0.87	0.87	20000 20000
No	No	No	Lancaster	Tf-Idf	Improved ML P				
			Stemmer	Vectorizer	LogisticRegress	sion(max_ite			
					Logistic Regres				
						precision	recall f		support
					negative	0.90	0.88	0.89	9935
					positive	0.89	0.90	0.89	10065
					accuracy			0.89	20000
					macro avg	0.89	0.89	0.89	20000
					weighted avg	0.89	0.89	0.89	20000

					Decision Tree	Accuracy: 0.	7242	. –	amples_split=5)
					Decision Tree	precision			support
						0.80			
					positive	0.68	0.85	0.76	10065
					accuracy		0.70	0.72	
					macro avg weighted avg	0.74	0.72	0.72	20000 20000
					RandomForest min_samples_ Random Forest	split=5) t Accuracy: 0	.8192		ax_depth=10,
					Random Forest	precision			support
					negative	0.85	0.77	0.81	9935
					positive	0.85 0.79	0.86	0.83	10065
					accuracy				20000
					macro avg weighted avg	0.82 0.82	0.82	0.82 0.82	20000
					GradientBoost learning_rate=			itors=100	,
					Gradient Boos Gradient Boos				
						precision			support
					negative	0.84	0.77	0.80	9935
					positive	0.79	0.85	0.82	10065
					accuracy			0.81	
					macro avg weighted avg	0.81			20000 20000
					alpha=0.0001) Neural Networ	k Accuracy: 0	ion Report		support
					negative		0.86	0.86	9935
					positive	0.86	0.85	0.86	10065
					accuracy	0.00	0.00	0.86	20000
					macro avg weighted avg		0.86	0.86 0.86	20000 20000
No	No	No	Spacy	Tf-Idf Vectorizer	Improved ML		1000	G 10)	
				Vectorizer	LogisticRegre	` =		,	
					Logistic Regre			Report:	support
					negative positive	0.90 0.89	0.89 0.91	0.90 0.90	9935 10065
					accuracy			0.90	20000
					macro avg weighted avg	0.90	0.90	0.90	20000
									amples split=5)
					Decision Tree Decision Tree	Accuracy: 0.	73585		
						precision		f1-score	support
					negative positive		0.68 0.79	0.72 0.75	9935 10065
					accuracy			0.74	20000
					macro avg	0.74	0.74	0.73	20000
					weighted avg	0.74	0.74	0.73	20000

RandomForestClassifier(n_estimators=100, max_demin_samples_split=5)	pth=10
Random Forest Accuracy: 0.8263 Random Forest Classification Report:	
precision recall f1-score supp	ort
negative 0.85 0.78 0.82 9	935
negative 0.85 0.78 0.82 9 positive 0.80 0.87 0.83 10	965
	900
macro avg 0.83 0.83 0.83 20	999
macro avg 0.83 0.83 0.83 20 weighted avg 0.83 0.83 0.83 20	999
GradientBoostingClassifier(n_estimators=100,	
learning_rate=0.1, max_depth=3)	
Gradient Boosting Accuracy: 0.818 Gradient Boosting Classification Report:	
precision recall f1-score suppo	ort
negative 0.84 0.78 0.81 99	935
positive 0.80 0.86 0.83 100	965
accuracy 0.82 200	999
macro avg 0.82 0.82 0.82 200	999
weighted avg 0.82 0.82 0.82 200	999
MLPClassifier(max_iter=1000, hidden_layer_sizes=alpha=0.0001)  Neural Network Accuracy: 0.86365  Neural Network Classification Report:	<del>(</del> 100,),
precision recall f1-score supp	ort
negative 0.85 0.87 0.86 9	935
	065
accuracy 0.86 20	000
macro avg 0.86 0.86 0.86 20	000
	000

# 2- Text Blob + ML

Stop Words	Tokenize	Pos	Stemmer/ Lemmatizer	Vectorizer	Classification Rep	ort			
No	No	No	No	Count	Improved ML P	arameters			
				Vectorizer	LogisticRegress	sion(max i	ter=1000	C=1.0	
					Logistic Regres				
					Logistic Regres				
					ţ.	precision	recall	f1-score	support
					negative	0.88	0.87	0.88	9935
					positive				10065
					accuracy			0.88	20000
					macro avg	0.88	0.88	0.88	20000
					weighted avg	0.88	0.88	0.88	20000
					Decision Tree (	Accuracy: 0	.7301	_	nmples_split=:
					1	precision	recall	f1-score	support
					negative	0.79	0.63	0.70	9935
					positive	0.69	0.83	0.76	10065
					accuracy			0.73	20000
					macro avg	0.74	0.73	0.73	20000
					weighted avg	0.74	0.73	0.73	20000

					RandomFores min_samples_ Random Forest	_split=5)		ors=100, n	nax_depth=10,
					Random Forest	: Classificat precision			support
						0.85 0.81			
					accuracy macro avg weighted avg	0.83	0.83 0.83		20000
					GradientBoos learning_rate= Gradient Boos Gradient Boos	tingClassific =0.1, max_deting Accurac	er(n_estin epth=3) y: 0.8179	nators=100	
						precision			support
						0.85 0.79			
					accuracy macro avg weighted avg	0.82	0.82 0.82	0.82 0.82 0.82	
					MLPClassifie alpha=0.0001 Neural Networ	) rk Accuracy:	0.8901		_sizes=(100,),
					Neur al Networ	precision			support
					negative positive	0.89 0.89	0.88 0.90	0.89 0.89	9935 10065
						0.89		0.89	
					weighted avg MultinomialN		0.89	0.89	20000
					Naive Bayes	Accuracy: 0.8		3	
					Naive Bayes (	precision			support
					negative positive		0.85 0.85		
					accuracy			0.85	
					macro avg weighted avg		0.85 0.85		
No	No	No	No	Tf – Idf Vectorizer	Improved ML LogisticRegre Logistic Regr Logistic Regr	ession(max in ession Accur ession Class	acy: 0.89	705	
						precision	recall	f1-score	support
					negative positive	0.90 0.89	0.89 0.91	0.90 0.90	9935 10065
					accuracy macro avg	0.90	0.90	0.90	20000
					weighted avg		0.90	0.90	20000
					DecisionTreeO Decision Tree Decision Tree	Accuracy: 0.	7317 on Report:	· —	amples_split=5)
					negative	0.80	0.62	0.70	9935
					positive accuracy	0.69	0.84	0.76 0.73	10065 20000
					macro avg weighted avg	0.74 0.74	0.73 0.73	0.73 0.73	20000 20000

	RandomForestCl min samples sp	` -	_estimato	ors=100, m	ax_depth=
		AND DAY OF STREET, STR	00445		
	Random Forest A			+.	
	Random Forest C			τ: f1-score	support
	pi	EC121011	recarr	11-2COLG	Support
	negative	0.86	0.78	0.81	9935
	positive			0.83	10065
	accuracy			0.82	20000
	macro avg	0.83	0.82		20000
	weighted avg	0.83		0.82	20000
	GradientBoostin		r(n estim	ators=100	
	learning rate=0.	l, max de	pth=3)		,
	Gradient Boosti Gradient Boosti				
				f1-score	support
	909 07 000 08 000 08 00				
	negative				9935
	positive	0.79	0.86	0.82	10065
	accuracy			0.82	20000
	macro ave	0.82	0.82	0.82	20000
	morro ove	0102	0.02		
	weighted avg  MLPClassifier(malpha=0.0001)			0.82	2000
	MLPClassifier(malpha=0.0001)	nax_iter=1 Accuracy:	000, hide	ø.82 den_layer_	20000
	MLPClassifier(malpha=0.0001)  Neural Network Neural Network	nax_iter=1 Accuracy: Classifica	000, hide	ø.82 den_layer_	20000 sizes=(10
	MLPClassifier(malpha=0.0001)  Neural Network Neural Network	nax_iter=1 Accuracy: Classifica recision	000, hide 0.86825 stion Repo	0.82 den_layer_ ort: f1-score	20000 sizes=(10
	MLPClassifier(malpha=0.0001)  Neural Network Neural Network p	nax_iter=1  Accuracy: Classifica recision  0.87	000, hidd 0.86825 stion Reported	0.82  den_layer_  ort: f1-score 0.87	20000 sizes=(10 support
	MLPClassifier(malpha=0.0001)  Neural Network Neural Network	nax_iter=1  Accuracy: Classifica recision  0.87	000, hide 0.86825 stion Repo	0.82  den_layer_  ort: f1-score 0.87	20000 sizes=(10 support
	MLPClassifier(malpha=0.0001)  Neural Network Neural Network p	nax_iter=1  Accuracy: Classifica recision  0.87	000, hidd 0.86825 stion Reported	0.82  den_layer_  ort: f1-score 0.87	20000 sizes=(10 support 9935 10065
	MLPClassifier(malpha=0.0001)  Neural Network Neural Network p  negative positive  accuracy macro avg	nax_iter=1  Accuracy: Classifica recision  0.87 0.87	000, hidd 0.86825 stion Repore recall 0.86 0.87	0.82 den_layer_ ort: f1-score 0.87 0.87 0.87	20000 sizes=(10 support 9935 10065 20000
	MLPClassifier(malpha=0.0001)  Neural Network Neural Network p negative positive accuracy	nax_iter=1  Accuracy: Classifica recision  0.87 0.87	000, hidd 0.86825 stion Repore recall 0.86 0.87	0.82 den_layer_ ort: f1-score 0.87 0.87 0.87	20000 sizes=(10 support 9935 10065 20000
	MLPClassifier(malpha=0.0001)  Neural Network Neural Network p  negative positive  accuracy macro avg weighted avg	Accuracy: Classifica recision 0.87 0.87 0.87 0.87	000, hidd 0.86825 stion Repore recall 0.86 0.87	0.82 den_layer_ ort: f1-score 0.87 0.87 0.87	20000 sizes=(10 support 9935 10065 20000
	MLPClassifier(malpha=0.0001)  Neural Network Neural Network p  negative positive accuracy macro avg weighted avg  MultinomialNB( Naive Bayes Accurate)	Accuracy: Classificatecision 0.87 0.87 0.87 0.87	0.86825 stion Reported 11 0.86 0.87 0.87	0.82 den_layer_ ort: f1-score 0.87 0.87 0.87	20000 sizes=(10 support 9935 10065 20000
	MLPClassifier(malpha=0.0001)  Neural Network Neural Network P  negative positive accuracy macro avg weighted avg  MultinomialNB( Naive Bayes Accurates Bayes Class	Accuracy: Classifica recision  0.87  0.87  0.87  0.87	0.86825 etion Reported 1 0.86 0.87 0.87	0.82 den_layer_ ort: f1-score 0.87 0.87 0.87 0.87	20000 sizes=(10 support 9935 10065 20000 20000
	MLPClassifier(malpha=0.0001)  Neural Network  Neural Network  P  negative  positive  accuracy  macro avg  weighted avg  MultinomialNB(  Naive Bayes Accurate  Naive Bayes Class	Accuracy: Classificatecision 0.87 0.87 0.87 0.87	0.86825 etion Reported 1 0.86 0.87 0.87	0.82 den_layer_ ort: f1-score 0.87 0.87 0.87	20000 sizes=(10 support 9935 10065 20000
Mal	ILPClassifier(mpha=0.0001)  Neural Network  Neural Network  negative positive  accuracy macro avg weighted avg  IultinomialNB(laive Bayes Clas	Accuracy: Classification  0.87 0.87 0.87 0.87 0.87 0.87	0.86825 etion Reported 10.86 0.87 0.87 0.87	0.82  den_layer_  ort:     f1-score     0.87     0.87     0.87     0.87     0.87	20000 sizes=(10 support 9935 10065 20000 20000
MLP alpha Neur Neur Weir Mult Naiv	PClassifier(ma=0.0001) ral Network ral Network rel Network pegative accuracy macro avg ghted avg cinomialNB( re Bayes Accuracy re Bayes Clas pr	Accuracy: Classifica recision  0.87  0.87  0.87  0.87	0.86825 etion Reported 1 0.86 0.87 0.87	0.82 den_layer_ ort: f1-score 0.87 0.87 0.87 0.87	20000 sizes=(10 support 9935 10065 20000 20000
ML alp Ne Ne	LPClassifier(mha=0.0001)  eural Network  eural Network  negative positive  accuracy macro avg eighted avg  ultinomialNB( ive Bayes Class  pr  negative positive	Accuracy: Classificatecision  0.87 0.87 0.87 0.87 0.87 0.87 0.87	0.86825 0.86825 1.00 Reported 11 0.86 0.87 0.87 0.87 0.87 0.87 0.87 0.87 0.87	0.82  den_layer_  ort:     f1-score     0.87     0.87     0.87     0.87  f1-score	20000 sizes=(100 support 9935 10065 20000 20000 support 9935 10065
M alı	LPClassifier(moha=0.0001)  Neural Network  Neural Network  P  Negative  positive  accuracy  macro avg  weighted avg  ultinomialNB( aive Bayes Clast  pr  negative	Accuracy: Classificatecision  0.87 0.87 0.87 0.87 0.87 0.87 0.87	0.86825 0.86825 1.00 Reported 11 0.86 0.87 0.87 0.87 0.87 0.87 0.87 0.87 0.87	0.82  den_layer_  ort:     f1-score     0.87     0.87     0.87     0.87     0.87	20000 sizes=(10 support 9935 10065 20000 20000 support

## 3- Vader + ML

Stop	Tokenize	Pos	Stemmer/	Vectorizer	Classification	on Report			
Words			Lemmatizer						
No	No	No	No	Count	Improved MI	2 Parameters			
				Vectorizer	LogisticRegro Logistic Regr Logistic Regr	ession Accur	acy: 0.87	62	
						precision	recall	f1-score	support
					negative	0.88	0.87	0.88	9935
					positive	0.88	0.88	0.88	10065
					accuracy			0.88	20000
					macro avg	0.88	0.88	0.88	20000
					weighted avg	0.88	0.88	0.88	20000
					DecisionTree	Classifier(ma	ax_depth	=10,min_s	amples_split=5)

	1	_	_	T	<del>_</del>				
					Decision Tree Decision Tree			·+·	
					Decision free			f1-score	support
					negative	0.79	0.63	0.70	9935
					positive	0.69	0.83	0.76	
					accuracy			0.73	20000
						0.74			
					weighted avg	0.74	0.73	0.73	20000
					RandomForest min samples	,	n_estimato	ors=100, m	ax_depth=10
					Random Forest	Accuracy:			
					Random Forest				
						precision	recall	f1-score	support
					negative	0.86	0.78	0.82	9935
					positive	0.80	0.87	0.83	10065
					accuracy			0.83	20000
					macro avg	0.83 0.83	0.83	0.83	
					weighted avg	0.83	0.83	0.83	20000
					GradientBoost learning_rate= Gradient Boost Gradient Boost	0.1, max_d	epth=3) cy: 0.8179	5	,
						precision			support
					negative	0.85	0.77	0.81	9935
					positive				
					accuracy			0.82	
					macro avg	0.82	0.82		
					weighted avg	0.02	0.82	0.82	20000
					MLPClassifier alpha=0.0001) Neural Network	Accuracy:	0.8874		sizes=(100,)
					Neural Network				
						precision	recall	f1-score	support
					negative positive				9935 10065
					* 100 CHESTON AND	0.05	0.05		
					accuracy	0.89	0.89	0.89 0.89	20000
					macro avg weighted avg	0.89	0.89		20000
					MultinomialN			3,530,503	
					Naive Bayes Ad	ccuracy: 0.8			
					Naive Bayes Cl	lassificatio precision			support
					nogotive	9.05	0.00	9.05	9925
					negative positive				9935 10065
					accuracy			0.85	20000
					macro avg	0.85		0.85	20000
					weighted avg	0.85	0.85	0.85	20000
No	No	No	No	Tf – Idf	Improved ML				
				Vectorizer	LogisticRegres	ssion(max	iter=1000	, C=1.0)	
					Logistic Regress	ion Classific	ation Repor		
					pr negative	ecision re		ore suppor	
					positive		0.91 0	.90 1006	5
					accuracy macro avg		0.90 0	1.90 2000 1.90 2000 1.90 2000	a
			<u> </u>		weighted avg	0.30	0.50 8	50 2000	9

					Decision Tree	Accuracy: 0	.73165		mples_split=5)
					Decision Tree	precision			support
					negative positive		0.62 0.84	0.70 0.76	
					accuracy macro avg		0.73		20000
					weighted avg RandomForestO		0.73		
					min_samples_s	split=5)		15 100, 111	ux_ucpiii 10,
					Random Forest		tion Repor		support
					negative			0.82	9935
1					positive	0.80	0.86	0.83	10065
ı					accuracy		0.00	0.82	
					weighted avg	0.83 0.83		0.82	
					GradientBoosti	ngClassifie	r(n estima	ators=100.	25000000-4-20
					learning_rate=0 Gradient Boost	).1, max_de	epth=3)		,
					Gradient Boost		ication Re		support
					negative	0.84	0.77	0.81	9935
					positive	0.79	0.86	0.82	10065
i					accuracy	20120	121 <u>202</u> 4	0.82	
					macro avg				
					weighted avg  MLPClassifier(				20000 sizes=(100,),
					MLPClassifier(alpha=0.0001)	max_iter=1	000, hidd	len_layer_	
					MLPClassifier( alpha=0.0001) Neural Network Neural Network	max_iter=1	000, hidd 0.8596 tion Repor	len_layer_	sizes=(100,),
					MLPClassifier( alpha=0.0001)  Neural Network  Neural Network	max_iter=1  Accuracy: Classifica	000, hidd 0.8596 tion Repor recall 0.83	len_layer_ rt: f1-score 0.85	sizes=(100,), support 9935
					MLPClassifier( alpha=0.0001)  Neural Network  Neural Network	max_iter=1  Accuracy: Classifica precision  0.88	000, hidd 0.8596 tion Repor recall 0.83	len_layer_ rt: f1-score 0.85	sizes=(100,), support 9935 10065
					MLPClassifier( alpha=0.0001)  Neural Network  Neural Network  negative positive  accuracy macro avg	max_iter=1  Accuracy: Classifica precision  0.88  0.84	0.8596 tion Repor recall 0.83 0.89	len_layer_ rt: f1-score	sizes=(100,), support 9935 10065 20000
					MLPClassifier( alpha=0.0001)  Neural Network  Neural Network  negative positive  accuracy macro avg weighted avg	max_iter=1  Accuracy: Classifica precision  0.88  0.84  0.86  0.86	000, hidd 0.8596 tion Repor recall 0.83 0.89	len_layer_ rt: f1-score	sizes=(100,), support 9935 10065 20000
					MLPClassifier( alpha=0.0001)  Neural Network  Neural Network  negative positive  accuracy macro avg weighted avg  MultinomialNE	(max_iter=1 c Accuracy: c Classifica precision 0.88 0.84 0.86 0.86	0.8596 tion Repor recall 0.83 0.89	len_layer_ rt: f1-score	sizes=(100,), support 9935 10065 20000
					MLPClassifier( alpha=0.0001)  Neural Network  Neural Network  negative positive  accuracy macro avg weighted avg  MultinomialNE Naive Bayes Ac Naive Bayes Cl	(max_iter=1  Accuracy: Classifica precision  0.88  0.84  0.86  0.86  0.86  0.86	0.8596 tion Reporrecall 0.83 0.89 0.86 0.86	len_layer_ rt: f1-score 0.85 0.86 0.86 0.86	sizes=(100,), support 9935 10065 20000 20000
					MLPClassifier( alpha=0.0001)  Neural Network  Neural Network  negative positive  accuracy macro avg weighted avg  MultinomialNE Naive Bayes Cl	(max_iter=1  Accuracy: Classifica precision  0.88  0.84  0.86  0.86  0.86  3() ccuracy: 0.8 lassificatio precision  0.87	0.8596 tion Reporrecall 0.83 0.89 0.86 0.86	t: f1-score 0.85 0.86 0.86 0.86 0.86	sizes=(100,), support 9935 10065 20000 20000 support
					MLPClassifier( alpha=0.0001)  Neural Network  Neural Network  negative positive  accuracy macro avg weighted avg  MultinomialNE Naive Bayes Cl  negative positive	(max_iter=1  Accuracy: Classifica precision  0.88  0.84  0.86  0.86  0.86  3() ccuracy: 0.8 lassificatio precision  0.87	0.8596 tion Reporrecall 0.83 0.89 0.86 0.86	len_layer_ rt: f1-score 0.85 0.86 0.86 0.86 0.86	sizes=(100,), support 9935 10065 20000 20000 support 9935 10065
					MLPClassifier( alpha=0.0001)  Neural Network  Neural Network  negative positive  accuracy macro avg weighted avg  MultinomialNE Naive Bayes Col  negative positive  accuracy	max_iter=1  Accuracy: Classifica precision  0.88  0.84  0.86  0.86  3() ccuracy: 0.8 lassificatio precision  0.87  0.86	0.8596 tion Reporrecall 0.83 0.89 0.86 0.86 0.86 0.86	len_layer_ rt: f1-score	sizes=(100,),  support  9935 10065 20000 20000  support  9935 10065 20000
					MLPClassifier( alpha=0.0001)  Neural Network  Neural Network  negative positive  accuracy macro avg weighted avg  MultinomialNE Naive Bayes Cl  negative positive	max_iter=1  Accuracy: Classifica precision  0.88  0.84  0.86  0.86  3() ccuracy: 0.8 lassificatio precision  0.87  0.86	0.8596 tion Reporrecall 0.83 0.89 0.86 0.86 0.86 0.86	len_layer_ rt: f1-score 0.85 0.86 0.86 0.86 0.86	sizes=(100,),  support  9935 10065  20000  support  9935 10065  20000
No	Yes	No	No	Tf – Idf Vectorizer	MLPClassifier( alpha=0.0001)  Neural Network  Neural Network  negative positive  accuracy macro avg weighted avg  MultinomialNE  Naive Bayes Ac Naive Bayes Cl  negative positive  accuracy macro avg weighted avg  Improved ML F	max_iter=1  Accuracy: Classifica precision  0.88 0.84  0.86 0.86 0.86  0.87 0.87 0.87	0.8596 tion Reporrecall 0.83 0.89 0.86 0.86 0.86 0.86 0.88 0.88	f1-score  0.85 0.86 0.86 0.86 0.87 0.87 0.87	sizes=(100,),  support  9935 10065  20000  support  9935 10065  20000
No	Yes	No	No	Tf – Idf Vectorizer	MLPClassifier( alpha=0.0001)  Neural Network  Neural Network  negative positive  accuracy macro avg weighted avg  MultinomialNE Naive Bayes Cl  negative positive  accuracy macro avg weighted avg  Improved ML F LogisticRegress	max_iter=1  Accuracy: Classifica precision  0.88 0.84  0.86 0.86  3() ccuracy: 0.8 lassification precision  0.87 0.87 0.87  0.87 0.87	0.8596 tion Reportereall 0.83 0.89 0.86 0.86 0.86 0.88 0.87 0.87 0.87	len_layer_ rt:     f1-score	sizes=(100,),  support  9935 10065  20000  support  9935 10065  20000
No	Yes	No	No		MLPClassifier( alpha=0.0001)  Neural Network  Neural Network  negative positive  accuracy macro avg weighted avg  MultinomialNE  Naive Bayes Cl  negative positive  accuracy macro avg weighted avg  Improved ML F  Logistic Regress Logistic Regress Logistic Regress Logistic Regress	max_iter=1  Accuracy: Classifica precision  0.88 0.84  0.86 0.86  3() ccuracy: 0.8 lassification precision  0.87 0.87 0.87  Parameters sion(max_iter=1)	0.8596 tion Reporrecall 0.83 0.89 0.86 0.86 0.86 0.88 0.87 0.87 0.87	len_layer_ rt:     f1-score	sizes=(100,),  support  9935 10065 20000 20000  support  9935 10065 20000 20000
No	Yes	No	No		MLPClassifier( alpha=0.0001)  Neural Network  Neural Network  negative positive  accuracy macro avg weighted avg  MultinomialNE Naive Bayes Ac Naive Bayes Cl  negative positive  accuracy macro avg weighted avg  Improved ML F LogisticRegress Logistic Regress Logistic Regress	max_iter=1  Accuracy: Classifica precision  0.88  0.84  0.86  0.86  0.86  0.87	0.8596 tion Reporrecall 0.83 0.89 0.86 0.86 0.86 0.88 0.87 0.87 0.87 ter=1000, acy: 0.897 ification recall	len_layer_ rt:     f1-score	sizes=(100,),  support  9935 10065 20000 20000  support  9935 10065 20000 20000  support
No	Yes	No	No		MLPClassifier( alpha=0.0001)  Neural Network  Neural Network  negative positive  accuracy macro avg weighted avg  MultinomialNE  Naive Bayes Cl  negative positive  accuracy macro avg weighted avg  Improved ML F  Logistic Regress Logistic Regress Logistic Regress Logistic Regress	max_iter=1  Accuracy: Classifica precision  0.88  0.84  0.86  0.86  0.86  0.87	0.8596 tion Reporrecall 0.83 0.89 0.86 0.86 0.86 0.86 0.87 0.87 0.87 ter=1000, acy: 0.897 ification recall 0.89	f1-score  0.85 0.86 0.86 0.86 0.87 0.87 0.87 0.87 0.87 0.87	sizes=(100,),  support  9935 10065 20000 20000  support  9935 10065 20000 20000
No	Yes	No	No		MLPClassifier( alpha=0.0001)  Neural Network  Neural Network  negative positive  accuracy macro avg weighted avg  MultinomialNE  Naive Bayes Ac Naive Bayes Cl  negative positive  accuracy macro avg weighted avg  Improved ML F Logistic Regress	max_iter=1  Accuracy: Classifica precision  0.88 0.84  0.86 0.86  0.86  0.87 0.87 0.87  0.87 0.87  0.87 0.87  0.87 0.87	0.8596 tion Reporrecall 0.83 0.89 0.86 0.86 0.86 0.86 0.87 0.87 0.87 ter=1000, acy: 0.897 ification recall 0.89	f1-score  0.85 0.86 0.86 0.86 0.87 0.87 0.87 0.87 0.87 0.87	sizes=(100,),  support 9935 10065 20000 20000  support 9935 10065 20000 20000  support 9935

DecisionTree(	Classifier(ma	ax depth=	=10,min s	amples split=5)
Decision Tre	· ·			1 _ 1 - 7
Decision Tre			rt:	
	precision			support
negative	0.80	0.62	0.70	9935
positive	0.69	0.85	0.76	10065
accuracy			0.73	20000
macro avg weighted avg	0.75	0.73	0.73	20000
RandomFores				
min samples	,	_csiiiiaic	715—100, II.	lax_deptii—10,
Random Forest	,	22295		
Random Forest			3	
	precision	recall	f1-score	support
negative	0.86 0.80	0.77	0.81	9935
positive	0.80	0.87	0.83	10065
accuracy			0.82	
macro avg weighted avg	0.83			
With the Control of t				
GradientBoos learning_rate=			iators–100	,
Gradient Boos				
Gradient Boos	precision			support
negative	a 95	0 77	a 91	9925
positive	0.85 0.79	0.86	0.83	10065
accuracy			0.82	20000
macro avg weighted avg	0.82	0.82	0.82	20000
0.990-194 <u>0</u> 0.000.0000.0000.000				
MLPClassifie alpha=0.0001	· —	1000, hide	den_layer_	_sizes=(100,),
Neural Networ			nt i	
Medial Metwor	precision			support
	32			202
negative positive		0.87		9935 10065
		~ ~ ~ ~ ~ ~ ~ ~ ~ ~ ~ ~ ~ ~ ~ ~ ~ ~ ~		
accuracy macro avg		0.87	0.87 0.87	
weighted avg		0.87		
MultinomialN				
Naive Bayes A	NAME OF TAXABLE PARTY OF TAXABLE PARTY.	696		
Naive Bayes C				
	precision	recall	f1-score	support
negative	0.88	0.86	0.87	9935
positive		0.88	0.87	10065
accuracy			0.87	20000
macro avg	0.87	0.87	0.87	20000
weighted avg	0.87	0.87	0.87	20000

# 4- Afinn + Vader + TextBlob + ML

Stop Words	Tokenize	Pos	Stemmer/ Lemmatizer	Vectorizer	Classification Report
No	No	No	No	Tf – Idf Vectorizer	Improved ML Parameters LogisticRegression(max_iter=1000, C=1.0)

			ssion Accura			
	Lo		ssion Classi			22
			precision	recall	f1-score	support
		negative	0.89	0.89	0.89	9935
		positive	0.89	0.90	0.89	10065
		accuracy			0.89	20000
	120-23	macro avg	0.89 0.89	0.89	0.89	20000
	90.23					
			Classifier(max Accuracy: 0.		=10,min_sa	imples_split=5)
			Classificati		t:	
		20132011 117 22	precision			support
		negative	0.79	0.74	0.77	9935
			0.76			10065
		accuracy			0.78	20000
	*-34		0.78 0.78			20000
						ax depth=10,
		n_samples_s	\ _	CStilliato	13–100, III	ax_ucpm=10,
			Accuracy: 0. Classificati			
	Re	andom Forest	precision			support
		010029.00	0.05	0.00	0.00	0000
		positive	0.85 0.83	0.82	0.83 0.84	9935 10065
		accuracy			0.84	
	Sure Sure	macro avg eighted avg	0.84	0.84		20000 20000
	923	CONTRACTOR CONTRACTOR	ingClassifier			
	lea	rning rate=	0.1, max_depting Accuracy	oth=3)		,
			ting Classifi			
		3022/10 2003	precision			support
		negative	0.83	0.83	0.83	9935
		positive		0.83	0.83	10065
		accuracy			0.83	20000
		macro avg	0.83	0.83	0.83	20000
	We	eighted avg		0.83	0.83	20000
		LPClassifier ha=0.0001)		000, hido	den_layer_	sizes=(100,),
	Ne	eural Network	Accuracy: 0			
		eural Network	classificat	ion Repor		-
			precision	recall	f1-score	support
		negative	0.87	0.87	0.87	9935
		positive	0.88	0.88	0.88	10065
		accuracy		298000	0.88	20000
		macro avg	0.88	0.88	0.88	20000
	we	eighted avg	0.88	0.88	0.88	20000

## 5- Vader + Glove100d

Stop	Tokenize	Pos	Stemmer/	Epoch Results
Words			Lemmatizer	
No	No	No	No	Epoch 1/S   188/188 [===================================

#### 6- Vader + Glove200d

Stop Words	Tokenize	Pos	Stemmer/ Lemmatizer	Epoch Results
No	No	No	No	Epoch 1/5 188/188 [========] - 675 337ms/step - loss: 0.5512 - accuracy: 0.6990 - val_loss: 0.3853 - val_accuracy: 0.8308 Epoch 2/5 188/188 [=========] - 615 327ms/step - loss: 0.3508 - accuracy: 0.8508 - val_loss: 0.5723 - val_accuracy: 0.7973 Epoch 3/5 188/188 [===================================

#### 7- Vader + Glove300d

Stop	Tokenize	Pos	Stemmer/	<b>Epoch Results</b>
Words			Lemmatizer	
No	No	No	No	Epoch 1/5  188/188 [===================================

## 8- Afinn + Glove100d

Stop Words	Tokenize	Pos	Stemmer/ Lemmatizer	<b>Epoch Results</b>
No	No	No	No	Epoch 1/5 188/188 [===================================

#### **Conclusion:**

In conclusion, our comprehensive analysis of sentiment analysis models on the IMDb dataset underscores the complexity and variability inherent in sentiment analysis tasks. Through experimentation with a diverse array of machine learning models, pretrained language models, lexicon-based approaches, and word embeddings, we have demonstrated that there is no one-size-fits-all solution for sentiment analysis. The effectiveness of sentiment analysis models is highly dependent on various factors such as the choice of preprocessing techniques, vectorization methods, and feature extraction strategies. Our findings highlight the importance of systematically exploring and evaluating different combinations of models and preprocessing pipelines to achieve optimal performance. It is imperative to recognize that what works well for one dataset or application scenario may not necessarily generalize to others. Therefore, practitioners should approach sentiment analysis tasks with careful consideration and experimentation, tailoring their approach to the specific characteristics and requirements of the dataset and the problem at hand.

# **Best Model:** Finetuning Roberta

#### **Accuracy:**

```
Accuracy: 0.95255
F1-score: 0.9531149646756584
Confusion matrix:
[[9485 530]
[ 419 9646]]
```

#### **Classification Report:**

	precision	recall	fl-score	support
negative	0.96	0.95	0.95	9935
positive	0.95	0.96	0.95	18865
accuracy			0.95	28080
macro avg	0.95	8.95	0.95	28686
weighted avg	0.95	0.95	0.95	28686

def init (self, encodings, labels):

**Time required to fine-tune**: 6715.162095785141s

#### Code:

import torch

```
from torch.utils.data import Dataset, DataLoader
from transformers import RobertaTokenizerFast, RobertaForSequenceClassification#, AdamW
from sklearn.model selection import train test split
import pandas as pd
from torch.optim import AdamW
import time
# Record start time
start_time = time.time()
# Load data
data = pd.read csv('/kaggle/input/plmdataset/train.csv/train.csv')
data['sentiment'] = data['sentiment'].map({'positive': 1, 'negative': 0})
reviews = data['review'].tolist()
labels = data['sentiment'].tolist() # assuming sentiment is encoded as 0 (negative) and 1 (positive)
# Split data into training and validation sets
train texts, val texts, train labels, val labels = train test split(reviews, labels, test size=0.2)
# Initialize tokenizer
tokenizer = RobertaTokenizerFast.from pretrained('roberta-base')
# Tokenize data
train encodings = tokenizer(train texts, truncation=True, padding=True, max length=512)
val encodings = tokenizer(val texts, truncation=True, padding=True, max length=512)
# Create torch dataset
class ReviewDataset(Dataset):
```

```
self.encodings = encodings
     self.labels = labels
  def getitem (self, idx):
     item = {key: torch.tensor(val[idx]) for key, val in self.encodings.items()}
     item['labels'] = torch.tensor(self.labels[idx])
     return item
  def len (self):
     return len(self.labels)
# Create dataloaders
train dataset = ReviewDataset(train_encodings, train_labels)
val dataset = ReviewDataset(val encodings, val labels)
train loader = DataLoader(train dataset, batch size=16, shuffle=True)
val loader = DataLoader(val dataset, batch size=16, shuffle=False)
# Initialize model
model = RobertaForSequenceClassification.from pretrained('roberta-base', num labels=2)
model = model.to('cuda') # if GPU is available
# Initialize optimizer
optimizer = AdamW(model.parameters(), lr=1e-5)
# Training loop
for epoch in range(3): # number of epochs
  model.train()
  for batch in train loader:
     optimizer.zero grad()
     input ids = batch['input ids'].to('cuda')
     attention mask = batch['attention mask'].to('cuda')
     labels = batch['labels'].to('cuda')
     outputs = model(input ids, attention mask=attention mask, labels=labels)
     loss = outputs.loss
     loss.backward()
     optimizer.step()
# Save the model
model.save pretrained('sentiment model RoBERTa')
# Record end time
end time = time.time()
print("Time required to fine-tune: ", end time - start time)
from sklearn.metrics import accuracy score, fl score, confusion matrix
from transformers import RobertaTokenizerFast, RobertaForSequenceClassification
from torch.utils.data import DataLoader, Dataset
import pandas as pd
import torch
# Load the model
model = RobertaForSequenceClassification.from pretrained('sentiment model RoBERTa')
model = model.to('cuda') # if GPU is available
# Load validation data
val data = pd.read csv('/kaggle/input/plmdataset/test.csv/test.csv')
val texts = val data['review'].tolist()
val labels = val data['sentiment'].map({'positive': 1, 'negative': 0}).tolist() # convert sentiment to numeric
# Initialize tokenizer
```

```
tokenizer = RobertaTokenizerFast.from pretrained('roberta-base')
# Tokenize data
val encodings = tokenizer(val texts, truncation=True, padding=True, max length=512)
# Create torch dataset for validation
class ReviewDataset(Dataset):
  def init (self, encodings, labels):
     self.encodings = encodings
     self.labels = labels
  def getitem (self, idx):
     item = {key: torch.tensor(val[idx]) for key, val in self.encodings.items()}
     item['labels'] = torch.tensor(self.labels[idx])
     return item
  def len (self):
     return len(self.labels)
val dataset = ReviewDataset(val encodings, val labels)
val loader = DataLoader(val dataset, batch size=16, shuffle=False)
# Evaluate the model
model.eval()
predictions = []
true labels = []
for batch in val loader:
  input ids = batch['input ids'].to('cuda')
  attention mask = batch['attention_mask'].to('cuda')
  labels = batch['labels'].to('cuda')
  with torch.no grad():
     outputs = model(input ids, attention mask=attention mask)
  logits = outputs.logits
  predicted labels = torch.argmax(logits, dim=1).cpu().numpy()
  predictions.extend(predicted labels)
  true labels.extend(labels.cpu().numpy())
# Calculate metrics
accuracy = accuracy_score(true_labels, predictions)
f1 = f1 score(true labels, predictions)
conf matrix = confusion matrix(true labels, predictions)
print(f'Accuracy: {accuracy}')
print(fF1-score: {f1}')
print(f'Confusion matrix:\n {conf matrix}')
```