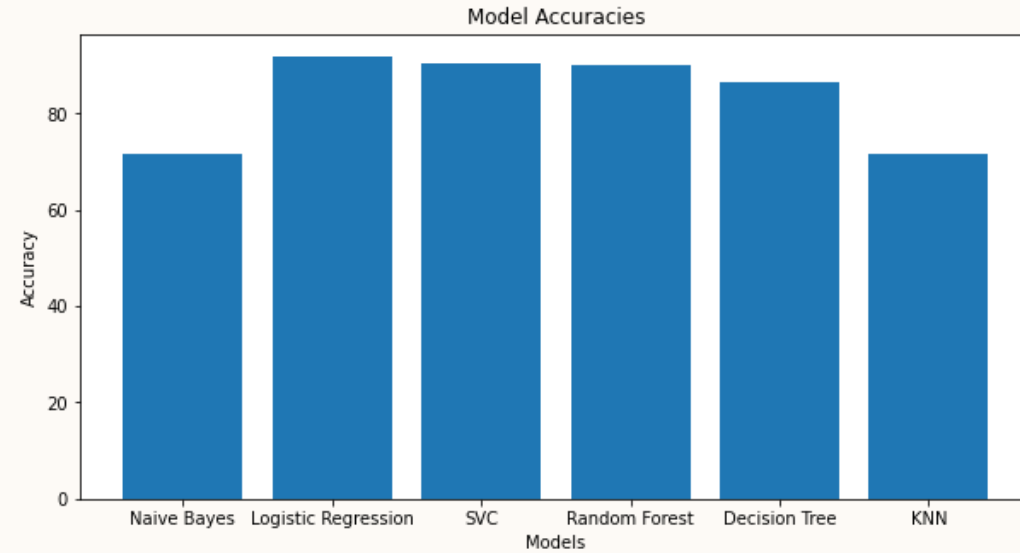


STEPS FOLLOWED

1. Built ML models for the problem statement.
2. Built Recurrent neural network for the same problem statement to prove that I have knowledge on Deep Learning too. (With fine tuning)
3. Used OOPs Concepts.
4. Developed API Server using FastAPI.
5. Integrated Swagger Documentation.
6. Containerized using Docker and deployed it on HuggingFace Spaces. (Bonus).
7. Each and every point mentioned in assignment was completed.

PERFORMANCE (BEFORE FINE TUNING)



RNN (LSTM) - ACCURACY – 90+%

BEST ML MODEL

		precision	recall	f1-score	support
	0	0.95	0.93	0.94	1862
	1	0.75	0.81	0.78	447
accuracy				0.91	2309
macro avg		0.85	0.87	0.86	2309
weighted avg		0.91	0.91	0.91	2309

```
Best Parameters: {'C': 1, 'penalty': 'l2'}
Best Score: 0.9180017257348597
Best Estimator: LogisticRegression(C=1)
Accuracy: 91.85794716327415
Confusion Matrix: [[1801  61]
 [ 127 320]]
```

Classification Report:		precision	recall	f1-score	support
	0	0.93	0.97	0.95	1862
	1	0.84	0.72	0.77	447
accuracy			0.92		2309
macro avg		0.89	0.84	0.86	2309
weighted avg		0.92	0.92	0.92	2309

Fine tuning:

All the models are duly fine tuned using GridSearchCV Algorithm and got the best hyperparameters.

Using of regularization and dropout ensured us models which are not overfitting.

```
#Logistic Regression
parameters = {'C': [0.1, 1, 10, 100, 1000], 'penalty': ['l1', 'l2']}
grid_search = GridSearchCV(LogisticRegression(), parameters, cv=5, n_jobs=-1, verbose=1)
grid_search.fit(sentiment_classifier.X_train, sentiment_classifier.y_train)
print('Best Parameters: ', grid_search.best_params_)
print('Best Score: ', grid_search.best_score_)
print('Best Estimator: ', grid_search.best_estimator_)
best_model = grid_search.best_estimator_
sentiment_classifier.train_model(best_model)
print('Accuracy: ', sentiment_classifier.get_accuracy()*100)
print('Confusion Matrix: ', sentiment_classifier.get_confusion_matrix())
print('Classification Report: ', sentiment_classifier.get_classification_report())
```