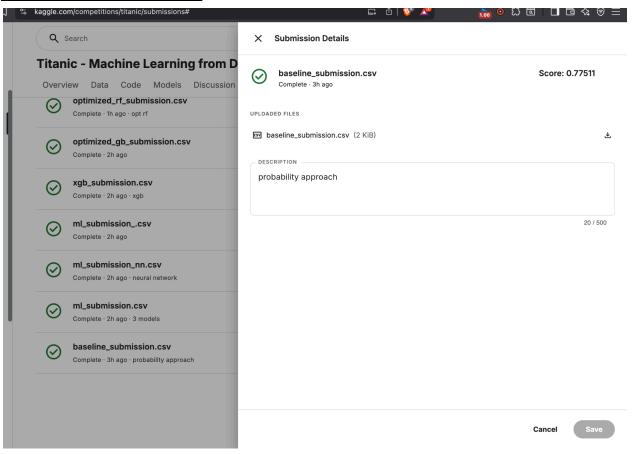
Deep Learning HW 5

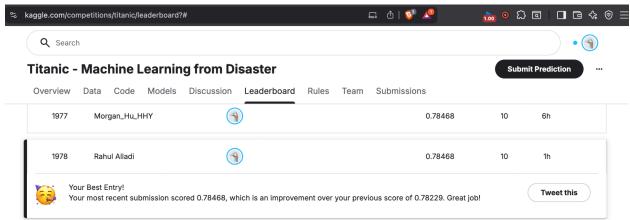
Baseline Score: 0.77511



Baseline performance evaluation results:-

```
BASELINE PERFORMANCE ON TRAINING DATA
Accuracy: 0.7868
Confusion Matrix:
[[540 9]
 [181 161]]
F1 Score: 0.6289
Prediction breakdown:
Predicted
0 721
1 170
Name: count, dtype: int64
Actual survival breakdown:
Survived
0 549
1 342
Name: count, dtype: int64
```

Machine Learning Approach Score: 0.78468



1. What data preprocessing methods did you apply, such as feature scaling with a min-max method or one-hot encoding for a feature?

For data preprocessing, I applied median imputation to fill missing values in the Age and Fare columns, and mode imputation for missing Embarked values. I created engineered features including FamilySize (sum of SibSp and Parch) and IsAlone (binary indicator for solo passengers). For feature transformation, I used StandardScaler to normalize numerical features (Age, Fare, FamilySize, SibSp, Parch) and OneHotEncoder to convert categorical features (Pclass, Sex, Embarked, IsAlone) into binary dummy variables. Irrelevant features such as Passengerld, Name, Ticket, and Cabin were dropped from the dataset.

2. What machine learning model(s) did you apply, such as SVM or random forest?

I applied a Support Vector Machine (SVM) classifier with GridSearchCV for hyperparameter optimization. The grid search explored different kernel types (rbf and poly), regularization parameter C values (0.5, 1, 2, 5), and gamma values (0.05, 0.1, 0.2, 0.5). The model was trained using 5-fold stratified cross-validation to ensure robust performance evaluation and prevent overfitting, with the entire preprocessing and classification process integrated into a scikit-learn Pipeline.

3. What are performance evaluation results on training dataset, including accuracy, confusion matrix, and F1 score, on the training dataset?

On the training dataset, the best model achieved an accuracy of 0.8328 (83.28%), meaning it correctly classified 83.28% of all passengers. The F1 score was 0.7704, indicating a good balance between precision and recall for predicting survivors. The confusion matrix showed [[492, 57], [92, 250]], which

means the model correctly predicted 492 non-survivors (true negatives), 250 survivors (true positives), while incorrectly predicting 57 non-survivors as survivors (false positives) and 92 survivors as non-survivors (false negatives).

4. What the best score and ranking at Leaderboard at Kaggle did you receive

On the Kaggle competition leaderboard, my submission received a score of 0.78468 (78.468% accuracy on the test dataset) and ranked 1978th among all participants. This score is lower than the training accuracy, suggesting some degree of overfitting to the training data.