DL project1

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1 Introduction

This project evaluates the performance of classification and regression algorithms on a game of Tic Tac Toe. The project lists statistical accuracy and confusion matrices for several classifiers and regressors like linear regression, linear SVM, k-nearest neighbors, and multilayer perceptron using the provided datasets. The datasets have been synthetically generated. The AI learns from these games to produce reasonable moves for one of the players in the game.

2 Program Execution

- 1. To train the models python3 buildModels.py
- 2. To play the game python3 tictactoe.py

3 Implementation

We split each of the 3 datasets as 80% train and 20% test sets. Then we used 10-fold cross validation to the training set to determine the optimal parameters for the model and used the model on the test set to report the test accuracies.

4 Classifiers

4.1 Cross-validation

4.1.1 Linear SVM

Table 1 and 2 shows the different parameter combinations with the mean 10-fold train and validation accuracies of tictac_final (final board classification) and tictac_single (Intermediate boards optimal play) datasets respectively.

Parameters	Train Accuracy($\%$)	\mid Validation Accuracy(%)
C: 10, gamma: 1	97.8	98
C: 10, gamma: 0.1	97.4	98
C: 10, gamma: 0.5	97.7	98
C: 12, gamma: 1	97.1	98
C: 12, gamma: 0.1	97.8	98
C: 12, gamma: 0.5	97.9	98

Table 1: Accuracy scores of SVM classifier on tictac_final (final board classification) dataset. All the accuracies are reported based on the mean score of 10 fold cross validation

4.1.2 k-nearest neighbors

For the k-nearest neighbors, we take a range for the values of k from 1 to 100 incremented by two steps. Then we plotted the mean validation accuracies of the 10 fold cross validation vs k values to get the optimal k. Figure 1 shows the plots for $tictac_final$ (final board classification) and $tictac_single$ (Intermediate boards optimal play) datasets respectively (from left to right).

4.1.3 Multilayer perceptron

We used 10-fold cross validation to find out the best possible combination of hyper-parameters for Multilayer perceptron (MLP) classifier. Table 3 and 4 shows the different parameter combinations with the mean 10-fold validation accuracies of *tictac_final* (final board classification) and *tictac_single* (Intermediate boards optimal play) datasets respectively.

Parameters	Train Accuracy(%)	Validation Accuracy(%)
C: 10, gamma: 1	39.5	39.02
C: 10, gamma: 0.1	39.5	39.02
C: 10, gamma: 0.5	39.5	39.028
C: 12, gamma: 1	39.5	39.02
C: 12, gamma: 0.1	39.5	39.02
C: 12, gamma: 0.5	39.5	39.028

Table 2: Accuracy scores of SVM classifier on *tictac_single* (Intermediate boards optimal play) dataset. All the accuracies are reported based on the mean score of 10 fold cross validation

Parameters	Validation Accuracy(%)
activation: relu, hidden_layer_sizes: (100, 100) activation: relu, hidden_layer_sizes: (120, 120)	99.08 99.2
activation: relu, hidden_layer_sizes: (150, 150) activation: relu, hidden_layer_sizes: (200, 200)	97.1 97.9

Table 3: Validation accuracy scores of MLP classifier on tictac_final (final board classification) dataset. All the accuracies are reported based on the mean score of 10 fold cross validation

Parameters	Validation Accuracy(%)
activation: relu, hidden_layer_sizes: (100, 100)	92.9
activation: relu, hidden_layer_sizes: (120, 120)	93.1
activation: relu, hidden_layer_sizes: (150, 150)	93.1
activation: relu, hidden_layer_sizes: $(200, 200)$	93.2

Table 4: Validation accuracy scores of MLP classifier on *tictac_single* (Intermediate boards optimal play) dataset. All the accuracies are reported based on the mean score of 10 fold cross validation

4.2 Performance on the test set

4.2.1 On the final board classification dataset

Table 5 shows the test accuracies for all the 3 classifiers on tictac_final (final board classification) dataset. Here, all the three algorithms perform really good, with K Nearest Neighbors getting the best of them all. The reason for KNN doing the best can be attributed to the smaller size of the dataset and the fact that the data is very well defined. Also, since no complicated classification is involved in this case, KNN gets to perform the best of them all.

Algorithms	Optimal Parameters	Test Accuracy(%)
Linear SVM	C – 10, gamma - 1	98.95
K Nearest Neighbors	Optimal K value - 1	100
Multi-Layer Perceptron	Activation function - relu	99.4
	Hidden Layer configuration: (120, 120)	

Table 5: Test accuracy scores of all 3 classifiers on tictac_final (final board classification) dataset

4.2.2 Intermediate boards optimal play (single label)

Table 5 shows the test accuracies for all the 3 classifiers on *tictac_single* (final board classification) dataset. In this scenario, Multi-layer Perceptron gives the best accuracy amongst all the algorithm. The dataset is substantially large and with a hidden layer configuration of (200, 200) the MLP algorithms learns the best out of all the algorithms. Linear SVM performs the worst because the data is not linearly separable. We have also tried the RBF Kernel which gives an accuracy of 93%, and this is because RBF is a non-linear (Gaussian) kernel.

4.3 Classifiers trained on 10% of the data

Table 7 and 8 show the performance of the classifiers when trained on $1/10^{th}$ of the whole dataset.

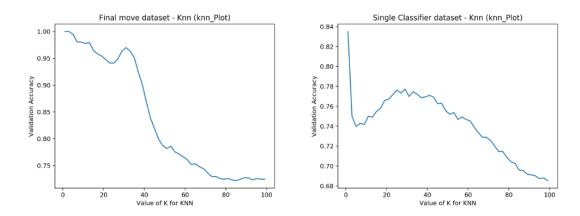


Figure 1: (Left to Right) K values vs validation accuracy plot of the knn classifiers of tictac_final and tictac_single datasets

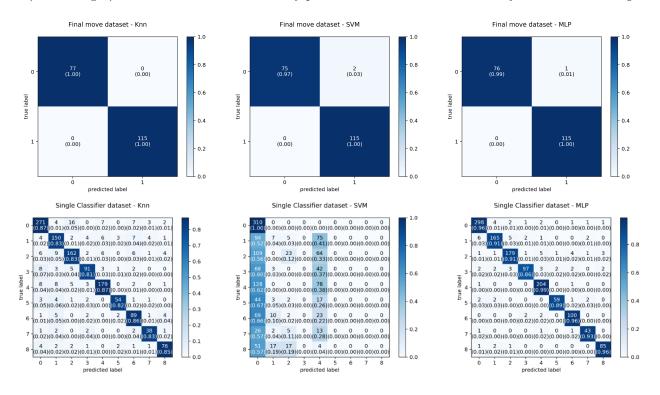


Figure 2: Confusion Matrices of the tictac_final and tictac_single datasets

Algorithms	Optimal Parameters	$\mid \text{Test Accuracy}(\%)$
Linear SVM	C – 10, gamma - 1	84.6
K Nearest Neighbors	Optimal K value - 1	31.8
Multi-Layer Perceptron	Activation function - relu	93.82
	Hidden Layer configuration: (120, 120)	

Table 6: Test accuracy scores of all 3 classifiers on $tictac_single$ (Intermediate boards optimal play) dataset

Algorithms	Optimal Parameters	Test Accuracy(%)
Linear SVM	C – 10, gamma - 1	85
K Nearest Neighbors	Optimal K value - 1	100
Multi-Layer Perceptron	Activation function - relu	100
	Hidden Layer configuration: (120, 120)	

Table 7: Test accuracy scores of all 3 classifiers on $1/10^{th}$ of $tictac_final$ (final board classification) dataset

4.4 Confusion Matrix

Figure 2 and 3 show the confusion matrices for the tictac_final and tictac_single datasets trained on the entire datasets and on the $1/10^{th}$ of the datasets respectively.

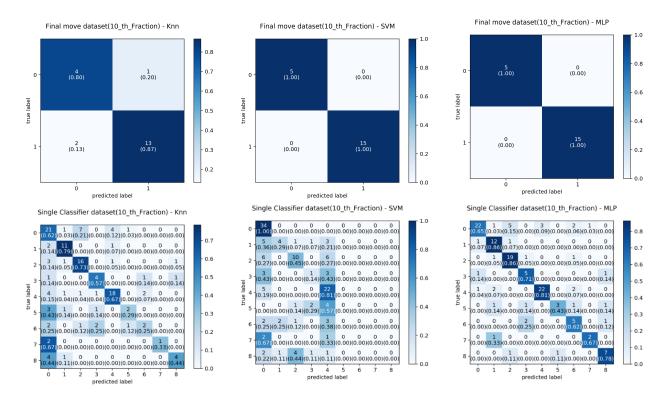


Figure 3: Confusion Matrices of $1/10^{th}$ the $tictac_final(1^{st} \text{ row})$ and $tictac_single(2^{nd} \text{ row})$ datasets for knn

Algorithms	Optimal Parameters	Test Accuracy(%)
Linear SVM	C – 10, gamma - 1	60
K Nearest Neighbors	Optimal K value - 1	54.19
Multi-Layer Perceptron	Activation function - relu	74.04
	Hidden Layer configuration: (120, 120)	

Table 8: Test accuracy scores of all 3 classifiers on $1/10^{th}$ of $tictac_single$ (Intermediate boards optimal play) dataset

5 Regressors

5.1 Cross validation

Table 9 shows the different parameters and their values used for the 10-fold cross validation of the MLP-Regressor. For knn-Regressor, we used a range for the values of k from 3 to 100 with a increment of 1 step.

Parameters	Values
max_iter	1000, 500
hidden_layer_sizes	(200, 100, 100, 50), (200, 100, 50)
activation	tanh, relu
solver	adam, lbfgs, sgd
alpha	0.0001
learning_rate	constant

Table 9: Grid search parameters for MLP regressors for the $tictac_multi$ dataset. All the accuracies are reported based on the mean score of 10 fold cross validation

5.2 Performance on the test set

In the regression problem, the Multi-Layer Perceptron algorithm performs the best out of all the three. The reason for this can be attributed to the availability of large dataset and the depth of the network. Also, the activation function that gets selected is tanh and this fits our dataset perfectly since our data ranges from [-1, 1]. The KNN algorithms performs well, but not the best as KNN being a lazy learning model with local approximation having commonly the distance function and k value as its hyperparameters.

Linear Regression on other hand, predicts a continuous value ranging from -ve infinity to infinity. The algorithm tries to fit a straight line in this problem which demands a classification. Also, Linear Regression is highly sensitive to imbalance data and such can be seen from a comparison between the R2 and RMSE errors between the two.

Table 10 shows the performance of the regressor models on the held-out test set.

Algorithms	Optimal Parameters	Test Accuracy(%)
Linear Regression	NA	72.62
K Nearest Neighbors	Optimal K value - 1	100
Multi-Layer Perceptron	Activation function - tanh	97.61
	Hidden Layer configuration: (200, 100, 100, 50)	
	alpha: 0.0001, solver: lbfgs, max_iter: 1000, learning_rate: constant	

Table 10: Test accuracy scores of all 3 classifiers on $1/10^{th}$ of $tictac_final$ (final board classification) dataset