CENG 499 - Introduction to Machine Learning

Homework 2

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January 4, 2024

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1 Part 2

1.1 Dataset 1

ID	\mathbf{C}	Kernel	Accuracy
1	0.5	Polynomial	0.68
2	0.5	RBF	1.0
3	2	Polynomial	0.69
4	2	RBF	1.0

Table 1: Hyperparameter configurations

I trained and tested all hyperparameters on the same training and test set. Decision boundaries for each hyperparameter configuration are plotted below.

1.1.1 Decision Boundaries

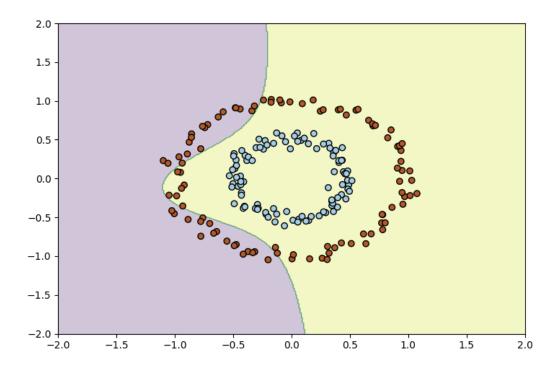


Figure 1: Decision boundary of SVM with C=0.5 and polynomial kernel

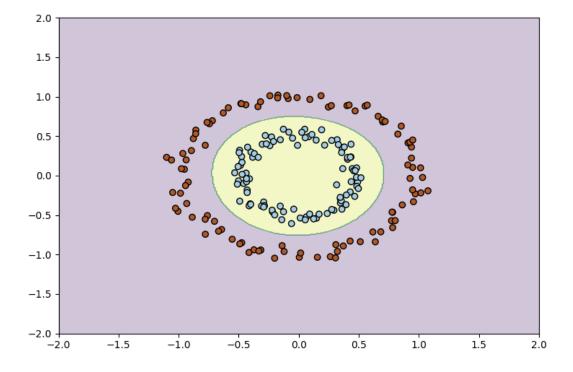


Figure 2: Decision boundary of SVM with C=0.5 and rbf kernel

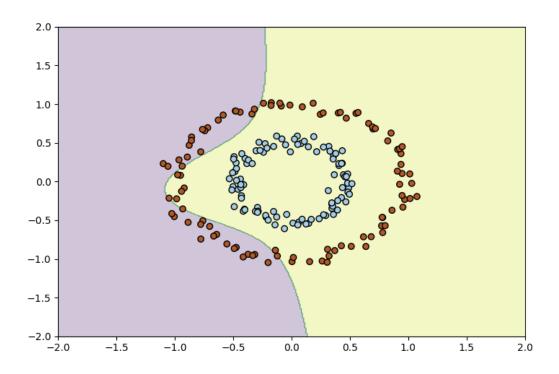


Figure 3: Decision boundary of SVM with C=2 and polynomial kernel

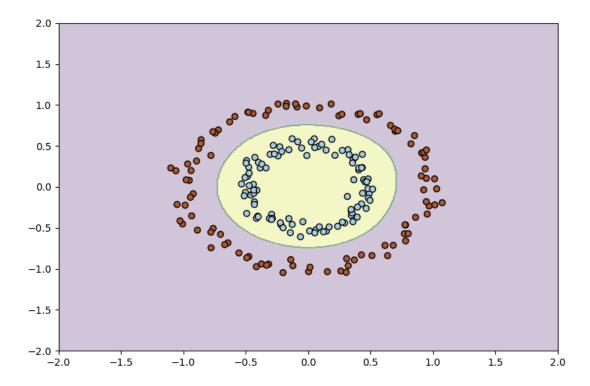


Figure 4: Decision boundary of SVM with C=2 and rbf kernel

1.2 Dataset 2

Hyperparameter configurations on the dataset 1 are kept the same with data set 2.

ID	\mathbf{C}	Kernel
1	0.5	Polynomial
2	0.5	RBF
3	2	Polynomial
4	2	RBF

Table 2: Hyperparameter configurations

I performed cross validation 5 times with 10 splits to find best hyperparameter configurations and plotted accuracy confidence intervals for each configuration.

1.2.1 Confidence Intervals

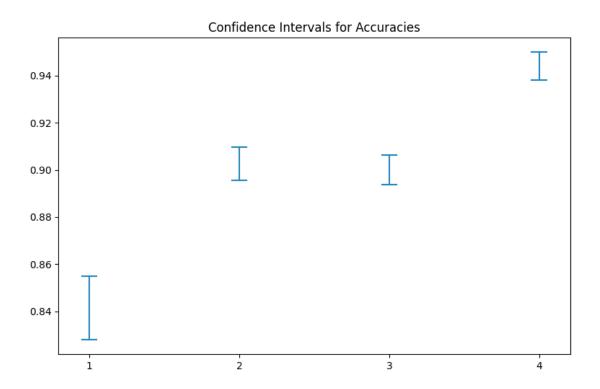


Figure 5: Accuracy confidence intervals

According to the plot, best hyperparameter configuration is the fourth one which is with the C=2 and rbf kernel.

2 Part 3

I applied nested cross validation to find the best performing algorithm for the credit application dataset, among KNN, SVM, decision trees and random forest algorithms.

ID	Algorithm	Parameter
1.1	KNN	metric = cosine, k = 4
1.2	KNN	metric = euclidean, k = 4
2.1	SVM	kernel = linear, C = 1
2.2	SVM	kernel = rbf, C = 1
3.1	Decision Tree	criterion = entropy
3.2	Decision Tree	criterion = gini
4.1	Random Forest	criterion = entropy
4.2	Random Forest	criterion = gini

Table 3: Algorithms and hyperparameter configurations

In the inner cross-validation I picked the best performing hyperparameter for each algorithm. I kept track of how many times a configuration is selected as best for each method. Since random forest is a stochastic model, I tested random forest algorithm 5 times and computed the mean of 5 accuracy and f1 scores in the inner cross validation.

2.1 Hyperparameter Search

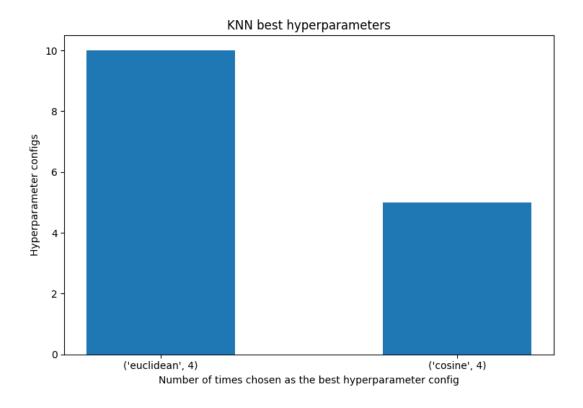


Figure 6: Number of times a hyperparameter is selected as best for KNN

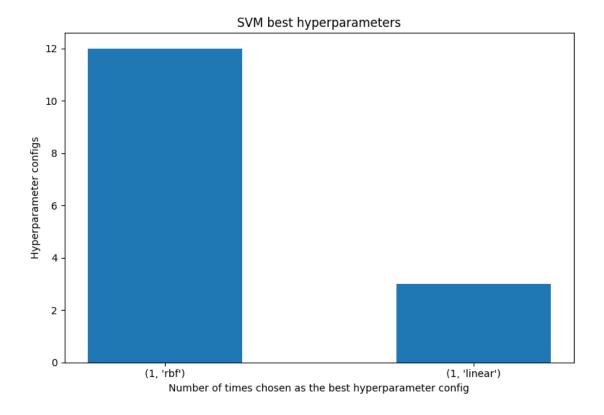


Figure 7: Number of times a hyperparameter is selected as best for SVM

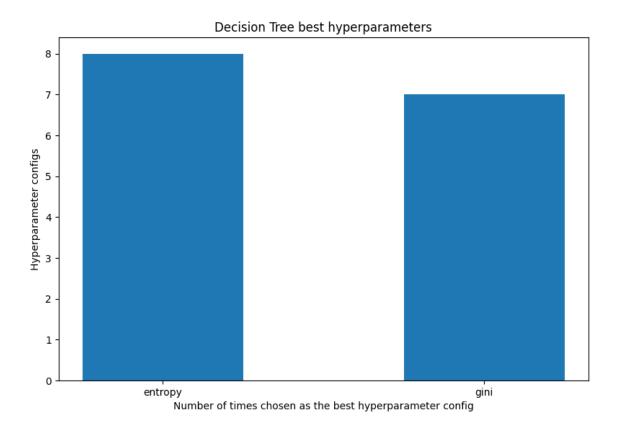


Figure 8: Number of times a hyperparameter is selected as best for decision tree

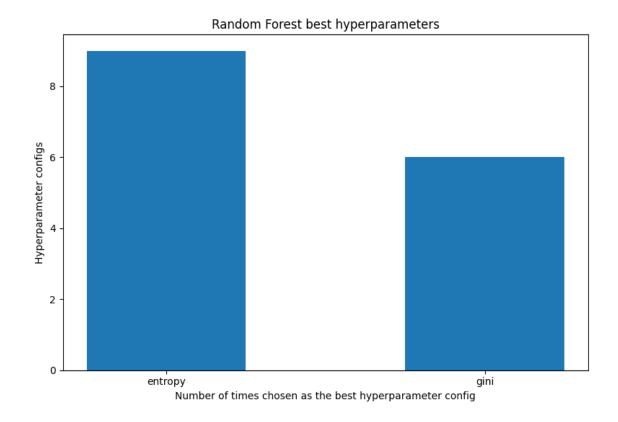


Figure 9: Number of times a hyperparameter is selected as best for random forest

2.2 Final Evaluation

In the outer loop of the nested cross validation, I trained and tested best configurations of methods (which I obtained in the inner cross validation) with the outer training and testing folds. I computed accuracy and f1 scores for each algorithm. As in the inner cross validation, I tested random forest algorithm 5 times and computed mean scores each time.

In the below chart, 1 corresponds to KNN, 2 corresponds to SVM, 3 corresponds to the decision tree and 4 corresponds to the random forest algorithm.

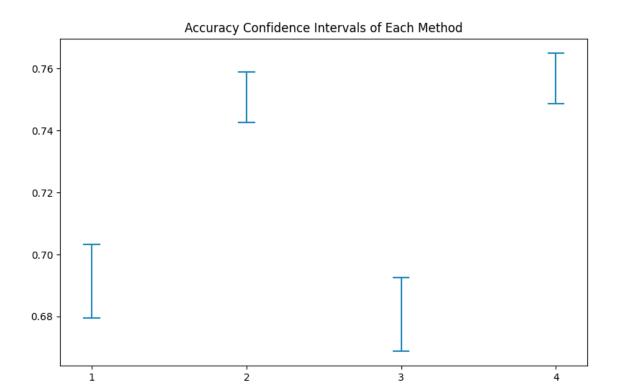


Figure 10: Accuracy confidence intervals for each algorithm

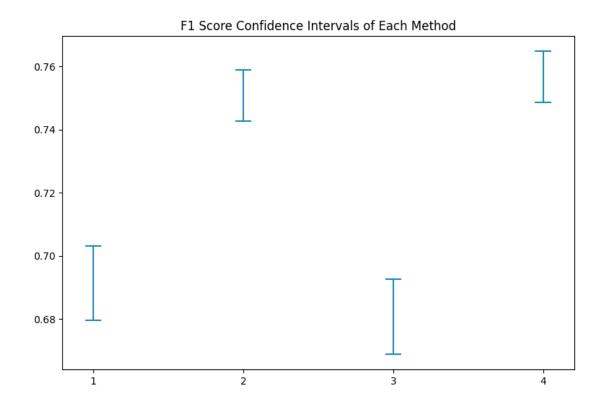


Figure 11: F1 score confidence intervals for each algorithm

According to the confidence interval charts, best method is the random forest algorithm for this dataset.

2.3 Decision Tree Feature Analysis

Feature importances of the decision tree are as follows:

- 1 0.14917455
- 2 0.10145336
- 3 0.04544835
- 4 0.06861055
- 5 0.17243175
- 6 0.03775644
- 7 0.05490751
- 8 0.02717022
- 9 0.02534325
- 10 0.02110837
- 11 0.04255615
- 12 0.04019985
- 13 0.09160992
- 14 0.01699652
- 15 0.02020321
- 16 0.0364606
- 17 0.02917185
- 18 0.0143182
- 19 0.00507937
- 20 0.

Most five important features are the credit amount (0.17), status of existing checking account (0.14), duration in month (0.10), age in years (0.90) and purpose (0.68) respectively.

2.4 SVM Support Vector Analysis

I undid the one hot encoding for both positive and negative support vectors.

Some of the positive class support vectors:

```
 \begin{array}{c} 3.0, \ 12.0, \ 4.0, \ 6.0, \ 0.0, \ 0.0, \ 4.0, \ 0.0, \ 0.0, \ 1.0, \ 0.0, \ 0.0, \ 0.0, \ 0.0, \ 0.0, \ 0.0, \ 0.0, \ 0.0, \ 0.0, \ 0.0, \ 0.0, \ 0.0, \ 0.0, \ 0.0, \ 0.0, \ 0.0, \ 0.0, \ 0.0, \ 0.0, \ 0.0, \ 0.0, \ 0.0, \ 0.0, \ 0.0, \ 0.0, \ 0.0, \ 0.0, \ 0.0, \ 0.0, \ 0.0, \ 0.0, \ 0.0, \ 0.0, \ 0.0, \ 0.0, \ 0.0, \ 0.0, \ 0.0, \ 0.0, \ 0.0, \ 0.0, \ 0.0, \ 0.0, \ 0.0, \ 0.0, \ 0.0, \ 0.0, \ 0.0, \ 0.0, \ 0.0, \ 0.0, \ 0.0, \ 0.0, \ 0.0, \ 0.0, \ 0.0, \ 0.0, \ 0.0, \ 0.0, \ 0.0, \ 0.0, \ 0.0, \ 0.0, \ 0.0, \ 0.0, \ 0.0, \ 0.0, \ 0.0, \ 0.0, \ 0.0, \ 0.0, \ 0.0, \ 0.0, \ 0.0, \ 0.0, \ 0.0, \ 0.0, \ 0.0, \ 0.0, \ 0.0, \ 0.0, \ 0.0, \ 0.0, \ 0.0, \ 0.0, \ 0.0, \ 0.0, \ 0.0, \ 0.0, \ 0.0, \ 0.0, \ 0.0, \ 0.0, \ 0.0, \ 0.0, \ 0.0, \ 0.0, \ 0.0, \ 0.0, \ 0.0, \ 0.0, \ 0.0, \ 0.0, \ 0.0, \ 0.0, \ 0.0, \ 0.0, \ 0.0, \ 0.0, \ 0.0, \ 0.0, \ 0.0, \ 0.0, \ 0.0, \ 0.0, \ 0.0, \ 0.0, \ 0.0, \ 0.0, \ 0.0, \ 0.0, \ 0.0, \ 0.0, \ 0.0, \ 0.0, \ 0.0, \ 0.0, \ 0.0, \ 0.0, \ 0.0, \ 0.0, \ 0.0, \ 0.0, \ 0.0, \ 0.0, \ 0.0, \ 0.0, \ 0.0, \ 0.0, \ 0.0, \ 0.0, \ 0.0, \ 0.0, \ 0.0, \ 0.0, \ 0.0, \ 0.0, \ 0.0, \ 0.0, \ 0.0, \ 0.0, \ 0.0, \ 0.0, \ 0.0, \ 0.0, \ 0.0, \ 0.0, \ 0.0, \ 0.0, \ 0.0, \ 0.0, \ 0.0, \ 0.0, \ 0.0, \ 0.0, \ 0.0, \ 0.0, \ 0.0, \ 0.0, \ 0.0, \ 0.0, \ 0.0, \ 0.0, \ 0.0, \ 0.0, \ 0.0, \ 0.0, \ 0.0, \ 0.0, \ 0.0, \ 0.0, \ 0.0, \ 0.0, \ 0.0, \ 0.0, \ 0.0, \ 0.0, \ 0.0, \ 0.0, \ 0.0, \ 0.0, \ 0.0, \ 0.0, \ 0.0, \ 0.0, \ 0.0, \ 0.0, \ 0.0, \ 0.0, \ 0.0, \ 0.0, \ 0.0, \ 0.0, \ 0.0, \ 0.0, \ 0.0, \ 0.0, \ 0.0, \ 0.0, \ 0.0, \ 0.0, \ 0.0, \ 0.0, \ 0.0, \ 0.0, \ 0.0, \ 0.0, \ 0.0, \ 0.0, \ 0.0, \ 0.0, \ 0.0, \ 0.0, \ 0.0, \ 0.0, \ 0.0, \ 0.0, \ 0.0, \ 0.0, \ 0.0, \ 0.0, \ 0.0, \ 0.0, \ 0.0, \ 0.0, \ 0.0, \ 0.0, \ 0.0, \ 0.0, \ 0.0, \ 0.0, \ 0.0, \ 0.0, \ 0.0, \ 0.0, \ 0.0, \ 0.0, \ 0.0, \ 0.0, \ 0.0, \ 0.0, \ 0.0, \ 0.0, \ 0.0, \ 0.0, \ 0.0, \ 0.0, \ 0.0, \ 0.0, \ 0.0, \ 0.0, \ 0.0, \ 0.0, \ 0.0, \ 0.0, \ 0.0, \ 0.0, \ 0.0, \ 0.0, \ 0.0, \ 0.0, \ 0.0, \ 0.0, \ 0.0, \ 0.0, \ 0.0, \ 0.0, \ 0.0, \ 0.0, \ 0.0, \ 0.0, \
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Some of the negative class support vectors:

```
\begin{array}{c} 1.0,\ 48.0,\ 2.0,\ 3.0,\ 0.0,\ 0.0,\ 3.0,\ 0.0,\ 0.0,\ 1.0,\ 0.0,\ 0.0,\ 0.0,\ 0.0,\ 0.0,\ 0.0,\ 0.0,\ 0.0,\ 0.0,\ 0.0,\ 0.0,\ 0.0,\ 0.0,\ 0.0,\ 0.0,\ 0.0,\ 0.0,\ 0.0,\ 0.0,\ 0.0,\ 0.0,\ 0.0,\ 0.0,\ 0.0,\ 0.0,\ 0.0,\ 0.0,\ 0.0,\ 0.0,\ 0.0,\ 0.0,\ 0.0,\ 0.0,\ 0.0,\ 0.0,\ 0.0,\ 0.0,\ 0.0,\ 0.0,\ 0.0,\ 0.0,\ 0.0,\ 0.0,\ 0.0,\ 0.0,\ 0.0,\ 0.0,\ 0.0,\ 0.0,\ 0.0,\ 0.0,\ 0.0,\ 0.0,\ 0.0,\ 0.0,\ 0.0,\ 0.0,\ 0.0,\ 0.0,\ 0.0,\ 0.0,\ 0.0,\ 0.0,\ 0.0,\ 0.0,\ 0.0,\ 0.0,\ 0.0,\ 0.0,\ 0.0,\ 0.0,\ 0.0,\ 0.0,\ 0.0,\ 0.0,\ 0.0,\ 0.0,\ 0.0,\ 0.0,\ 0.0,\ 0.0,\ 0.0,\ 0.0,\ 0.0,\ 0.0,\ 0.0,\ 0.0,\ 0.0,\ 0.0,\ 0.0,\ 0.0,\ 0.0,\ 0.0,\ 0.0,\ 0.0,\ 0.0,\ 0.0,\ 0.0,\ 0.0,\ 0.0,\ 0.0,\ 0.0,\ 0.0,\ 0.0,\ 0.0,\ 0.0,\ 0.0,\ 0.0,\ 0.0,\ 0.0,\ 0.0,\ 0.0,\ 0.0,\ 0.0,\ 0.0,\ 0.0,\ 0.0,\ 0.0,\ 0.0,\ 0.0,\ 0.0,\ 0.0,\ 0.0,\ 0.0,\ 0.0,\ 0.0,\ 0.0,\ 0.0,\ 0.0,\ 0.0,\ 0.0,\ 0.0,\ 0.0,\ 0.0,\ 0.0,\ 0.0,\ 0.0,\ 0.0,\ 0.0,\ 0.0,\ 0.0,\ 0.0,\ 0.0,\ 0.0,\ 0.0,\ 0.0,\ 0.0,\ 0.0,\ 0.0,\ 0.0,\ 0.0,\ 0.0,\ 0.0,\ 0.0,\ 0.0,\ 0.0,\ 0.0,\ 0.0,\ 0.0,\ 0.0,\ 0.0,\ 0.0,\ 0.0,\ 0.0,\ 0.0,\ 0.0,\ 0.0,\ 0.0,\ 0.0,\ 0.0,\ 0.0,\ 0.0,\ 0.0,\ 0.0,\ 0.0,\ 0.0,\ 0.0,\ 0.0,\ 0.0,\ 0.0,\ 0.0,\ 0.0,\ 0.0,\ 0.0,\ 0.0,\ 0.0,\ 0.0,\ 0.0,\ 0.0,\ 0.0,\ 0.0,\ 0.0,\ 0.0,\ 0.0,\ 0.0,\ 0.0,\ 0.0,\ 0.0,\ 0.0,\ 0.0,\ 0.0,\ 0.0,\ 0.0,\ 0.0,\ 0.0,\ 0.0,\ 0.0,\ 0.0,\ 0.0,\ 0.0,\ 0.0,\ 0.0,\ 0.0,\ 0.0,\ 0.0,\ 0.0,\ 0.0,\ 0.0,\ 0.0,\ 0.0,\ 0.0,\ 0.0,\ 0.0,\ 0.0,\ 0.0,\ 0.0,\ 0.0,\ 0.0,\ 0.0,\ 0.0,\ 0.0,\ 0.0,\ 0.0,\ 0.0,\ 0.0,\ 0.0,\ 0.0,\ 0.0,\ 0.0,\ 0.0,\ 0.0,\ 0.0,\ 0.0,\ 0.0,\ 0.0,\ 0.0,\ 0.0,\ 0.0,\ 0.0,\ 0.0,\ 0.0,\ 0.0,\ 0.0,\ 0.0,\ 0.0,\ 0.0,\ 0.0,\ 0.0,\ 0.0,\ 0.0,\ 0.0,\ 0.0,\ 0.0,\ 0.0,\ 0.0,\ 0.0,\ 0.0,\ 0.0,\ 0.0,\ 0.0,\ 0.0,\ 0.0,\ 0.0,\ 0.0,\ 0.0,\ 0.0,\ 0.0,\ 0.0,\ 0.0,\ 0.0,\ 0.0,\ 0.0,\ 0.0,\ 0.0,\ 0.0,\ 0.0,\ 0.0,\ 0.0,\ 0.0,\ 0.0,\ 0.0,\ 0.0,\ 0.0,\ 0.0,\ 0.0,\ 0.0,\ 0.0,\ 0.0,\ 0.0,\ 0.0,\ 0.0,\ 0.0,\ 0.0,\ 0.0,\ 0.0,\ 0.0,\ 0.0,\ 0.0,\ 0.0,\ 0.0,\ 0.0,\ 0.0,\ 0.0,\ 0.0,\ 0.0,\ 0.0,\ 0.0,\ 0.0,\ 0.0,\ 0.0,\ 0.0,\ 0.0,\ 0.0,\ 0.0,\ 0.0,\ 0.0,\ 0.0,\ 0.0,\ 0.0,\ 0.0,
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For 5th, 6th, 9th, 11th, 12th, 13th, 14th, 17th, 19th attributes values are mostly zero among all support vectors. Also 10th and 20th attribute values consist of mostly 1s. For the 2nd attribute most recurring values are multiples of 6.