**Student Intervention judgement**

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**1. Classification or Regression**

In this problem, we’d like to know whether student pass the final exam or not. Therefore, I use classification method. Classification can be applied to the problem of which the answer belongs to particular class like “agree or disagree”, “A team or B team or C team” and “pass the exam or not”.

**2. Evaluating Model Performance**

Total number of students: 395

Number of features: 30

Number of students who passed: 265

Number of students who failed: 130

Graduation rate of the class: 67.09%

**3. Preparing the data**

I used *train\_test\_split* from *sklearn* for random data assignment for training and test dataset. I adopted *label\_binarize* from *sklearn* to binarize the target column from “yes and no” to “0 and 1”.

**4. Training and Evaluating Models**

(1) Decision Tree

Select Reason: The structure is simple and easy to implement.

Application: Buyer behavior analysis.

Strengths: The result tree can be visualized and easy to interpret it. The condition can be explained by boolean logic, therefore the model is easy to understand.

Weaknesses: The result can be changed easily by a little change of the data. It is difficult to find the optimal parameters because this is one of the NP-complete problems.

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| --- | --- | --- | --- |
| **Decision Tree** | Training data size | | |
| 100 | 200 | 300 |
| Training time | 0.001 | 0.001 | 0.003 |
| Prediction time | 0.000 | 0.000 | 0.001 |
| F1 for training data | 1.000 | 1.000 | 1.000 |
| F1 for test data | 0.729 | 0.708 | 0.695 |

(2) Random Forest

Select Reason: High prediction performance.The structure is simple and calculation speed is high.

Application: This method used for image recognition on medical fields.

Strengths: High complexity model can be used in low variance because random forest uses ensemble methods.

Weaknesses: The model tends to be complicated because this model uses various decision trees and the number of the parameters can be large. In small data, this model doesn’t work well because this model picks training data and variables randomly.

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| **Random Forest** | Training data size | | |
| 100 | 200 | 300 |
| Training time | 0.031 | 0.025 | 0.025 |
| Prediction time | 0.001 | 0.002 | 0.001 |
| F1 for training data | 0.992 | 1.000 | 0.993 |
| F1 for test data | 0.782 | 0.797 | 0.761 |

(3) Support Vector Machine

Select Reason: It is known that this method shows high predictability on various problems.

Application: This method often used for text processing like detecting spam mail.

Strengths: High predictability because of high generalization ability. By Kernel trick, this method works when the dimensions is larger than the number of samples.

Weaknesses: Training time is long on big data size.

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| **SVM** | Training data size | | |
| 100 | 200 | 300 |
| Training time | 0.024 | 0.004 | 0.027 |
| Prediction time | 0.002 | 0.003 | 0.006 |
| F1 for training data | 0.859 | 0.869 | 0.869 |
| F1 for test data | 0.784 | 0.776 | 0.759 |

**5. Choosing the Best Model**

I chose Support Vector Machine as the best model. Decision Tree is simple but that shows low F1 score and it doesn’t have plenty tuning potential. Random Forest shows good F1 score but it tends to take more training time than SVM.

SVM is a process for finding the of a border. The border predicts a data belongs which group like Yes or No, Team A or Team B. The border is the most far from the two groups nearest data. SVM can write the border not only in straight line but also other shapes by a trick, called kernel trick.

The final F1 score applied GridSearchCV is as followings:

Best parameter choice: {'kernel': 'rbf', 'C': 100, 'gamma': 0.0001}

Best F1 score: 0.824511363636