# Lab4 Report

# Task 1:

Random Forest:

Experiments:

| n_estimators | accuracy |
|--------------|----------|
| 5            | 77.40%   |
| 10           | 79.40%   |
| 20           | 78.07%   |
| 30           | 77.07%   |

# Best model:

| n_estimators | accuracy | Confusion matrix        |
|--------------|----------|-------------------------|
| 10           | 79.40%   | [221, 17],<br>[ 45, 18] |

# AdaBoost:

# Experiments:

| n_estimators | accuracy |
|--------------|----------|
| 5            | 80.39%   |
| 10           | 82.39%   |
| 20           | 81.73%   |
| 30           | 81.40%   |

#### Best model:

| n_estimators | accuracy | Confusion matrix      |
|--------------|----------|-----------------------|
| 10           | 82.39%   | [230, 8],<br>[45, 18] |

#### Results:

Random forest is an ensemble model using bagging as the ensemble method and decision tree as the individual model. AdaBoost is a boosting ensemble model and works especially well with the decision tree. Boosting model's key is learning from the previous mistakes, so AdaBoost has the better performance on this binary classification problem.

### Task 2:

### 2.1:

Neural Network:

### Best model:

| Hidden layer | nodes | Learning rate | accuracy | Confusion          |
|--------------|-------|---------------|----------|--------------------|
|              |       |               |          | matrix             |
| 1            | 4     | 0.1           | 81.40    | [237 1]<br>[ 55 8] |

## KNN algorithm:

### Best model:

| n_nerghbors accuracy Confusion matrix |
|---------------------------------------|
|---------------------------------------|

| 10 | 81.73% | [230 8]  |
|----|--------|----------|
|    | 01.75% | [ 47 16] |

# Logistic regression:

## Best model:

| solver | Multi_class | accuracy | Confusion matrix    |
|--------|-------------|----------|---------------------|
| lbfgs  | multinomial | 80.73%   | [232 6]<br>[ 52 11] |

# Naïve Bayes:

# Best model:

| Likelihood model | accuracy | Confusion matrix     |
|------------------|----------|----------------------|
| Gaussian         | 80.06%   | [226 12]<br>[ 48 15] |

## Decision Tree:

## Best model:

| accuracy | Confusion matrix     |
|----------|----------------------|
| 71.10%   | [195 43]<br>[ 44 19] |

# 2.2:

Unweighted majority vote:

| accuracy | Confusion matrix    |
|----------|---------------------|
| 82.39%   | [230 8]<br>[ 45 18] |

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# 2.3:

## Weighted majority vote:

| Retio(NN:KNN:LR:NB:DT) | accuracy | Confusion matrix   |
|------------------------|----------|--------------------|
| 81 : 82: 81 : 80 : 71  | 83.39%   | [230 8]<br>[42 21] |

## Results:

The unweighted majority vote ensemble classifier has the similar performance with the best model (n\_nergnbors) in the five models. The weighted majority vote ensemble classifier has the better performance than all the five models.

Task 3:
Unweighted majority vote:

| accuracy | Confusion matrix   |
|----------|--------------------|
| 82.05%   | [230 8]<br>[48 15] |

## Weighted majority vote:

| Retio(NN:KNN:LR:NB:DT:RF:ADA) | accuracy | Confusion matrix    |
|-------------------------------|----------|---------------------|
| 81:82:81:80:71:79:82          | 84.05%   | [229 9]<br>[ 39 24] |

### Results:

The unweighted majority vote ensemble classifier has the similar performance with the best model (AdaBoost) in the seven models. The weighted majority vote ensemble classifier has the better performance than all the seven models and also better than the weighted majority vote ensemble model in task 2.