

# Data Mining Homework 8

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## 1 Using Weka tool on *diabetes* data

### 1.1 Characterize the TP, FP, TN, FN rates, accuracy, precision and recall obtained from this data.

Output from running weka on the *diabetes.arff* dataset.

```
=== Run information ===
```

```
Scheme:weka.classifiers.trees.J48 -C 0.25 -M 2
```

```
Relation:      pima_diabetes
```

```
Instances:     768
```

```
Attributes:    9
```

```
preg
```

```
plas
```

```
pres
```

```
skin
```

```
insu
```

```
mass
```

```
pedi
```

```
age
```

```
class
```

```
Test mode:10-fold cross-validation
```

```
=== Classifier model (full training set) ===
```

```
J48 pruned tree
```

```
-----
```

```
plas <= 127
```

```
| mass <= 26.4: tested_negative (132.0/3.0)
```

```
| mass > 26.4
```

```
| | age <= 28: tested_negative (180.0/22.0)
```

```
| | age > 28
```

```
| | | plas <= 99: tested_negative (55.0/10.0)
```

```
| | | plas > 99
```

```
| | | | pedi <= 0.561: tested_negative (84.0/34.0)
```

```
| | | | pedi > 0.561
```

```
| | | | | preg <= 6
```

```
| | | | | age <= 30: tested_positive (4.0)
```

```

| | | | | | | | age > 30
| | | | | | | | age <= 34: tested_negative (7.0/1.0)
| | | | | | | | age > 34
| | | | | | | | mass <= 33.1: tested_positive (6.0)
| | | | | | | | mass > 33.1: tested_negative (4.0/1.0)
| | | | | | | | preg > 6: tested_positive (13.0)
plas > 127
| mass <= 29.9
| | plas <= 145: tested_negative (41.0/6.0)
| | plas > 145
| | | age <= 25: tested_negative (4.0)
| | | age > 25
| | | | age <= 61
| | | | mass <= 27.1: tested_positive (12.0/1.0)
| | | | mass > 27.1
| | | | pres <= 82
| | | | | pedi <= 0.396: tested_positive (8.0/1.0)
| | | | | pedi > 0.396: tested_negative (3.0)
| | | | | pres > 82: tested_negative (4.0)
| | | | age > 61: tested_negative (4.0)
| mass > 29.9
| | plas <= 157
| | | pres <= 61: tested_positive (15.0/1.0)
| | | pres > 61
| | | | age <= 30: tested_negative (40.0/13.0)
| | | | age > 30: tested_positive (60.0/17.0)
| | plas > 157: tested_positive (92.0/12.0)

```

Number of Leaves : 20

Size of the tree : 39

Time taken to build model: 0.06 seconds

=== Stratified cross-validation ===

=== Summary ===

Correctly Classified Instances	567	73.8281 %
Incorrectly Classified Instances	201	26.1719 %
Kappa statistic	0.4164	
Mean absolute error	0.3158	
Root mean squared error	0.4463	
Relative absolute error	69.4841 %	
Root relative squared error	93.6293 %	
Total Number of Instances	768	

=== Detailed Accuracy By Class ===

	TP Rate	FP Rate	Precision	Recall	F-Measure	ROC Area	Class
	0.814	0.403	0.79	0.814	0.802	0.751	tested_negative
	0.597	0.186	0.632	0.597	0.614	0.751	tested_positive
Weighted Avg.	0.738	0.327	0.735	0.738	0.736	0.751	

```

a  b  <-- classified as
407 93 | a = tested_negative
108 160 | b = tested_positive

```

Decision Tree Structure:

- Root: **plas**
  - ≤ 127: **mass**
    - ≤ 127: **tested\_negative (1 age '3.0)**
    - > 127: **mass**
      - ≤ 29.9: **plas**
        - ≤ 45: **tested\_negative (1' pl**
        - > 45: **tested\_negative (' age 5.0)**
      - > 29.9: **plas**
        - ≤ 57: **tested\_positive (92.0/12.0)**
        - > 57: **tested\_positive (1 age '1.0)**
    - > 45: **tested\_negative (5' pedi**
    - > 57: **tested\_positive (1 age '1.0)**
  - > 127: **preg**
    - ≤ 1.0: **tested\_negative (8' preg 1.0)**
    - > 1.0: **tested\_positive (60.0/17.0)**

Additional nodes and splits visible in the diagram:

- plas** (≤ 127) → **mass** (≤ 127) → **tested\_negative (1 age '3.0)**
- plas** (≤ 127) → **mass** (≤ 127) → **tested\_negative (1' pl**
- plas** (≤ 127) → **mass** (≤ 127) → **tested\_negative (' age 5.0)**
- plas** (≤ 127) → **mass** (≤ 127) → **tested\_positive (92.0/12.0)**
- plas** (≤ 127) → **mass** (≤ 127) → **tested\_positive (1 age '1.0)**
- plas** (≤ 127) → **mass** (≤ 127) → **tested\_negative (5' pedi**
- plas** (≤ 127) → **mass** (≤ 127) → **tested\_positive (1 age '1.0)**
- plas** (≤ 127) → **preg** (≤ 1.0) → **tested\_negative (8' preg 1.0)**
- plas** (≤ 127) → **preg** (≤ 1.0) → **tested\_positive (60.0/17.0)**
- plas** (≤ 127) → **pres** (≤ 0) → **tested\_positive (' pres 0)**
- plas** (≤ 127) → **age** (≤ 1) → **tested\_positive age '**
- plas** (≤ 127) → **age** (≤ 1) → **tested\_negative (4.0)**
- plas** (≤ 127) → **age** (≤ 1) → **tested\_negative (3.0)**
- plas** (≤ 127) → **age** (≤ 1) → **tested\_negative (4.0/1.0)**

1. **TP rate (sensitivity, recall rate)** defined as  $TPR = TP / (TP + FN)$ . Means there 81.4% would be correctly identified using that model.
2. **FP rate (rate of type I error)** defined as  $FPR = FP / (FP + TN)$ . Increases as the number of false positive increases or true negatives decreases. Related to the false posit
3. **TN rate (specificity)** Defined as  $TNR = TN / (FP + TN)$  measures the proportion of negatives correctly identified. It decreases when false positives decreases.
4. **FN rate (rate of type II error)** means with what probability is the
5. **accuracy** ration between true results (true positives and true negatives) and total events (whole population). Formula:  $accuracy = (TP + TN) / (TP + FP + FN + TN)$
6. **precision** Formula:  $*precision = TP / (TP + FP)$

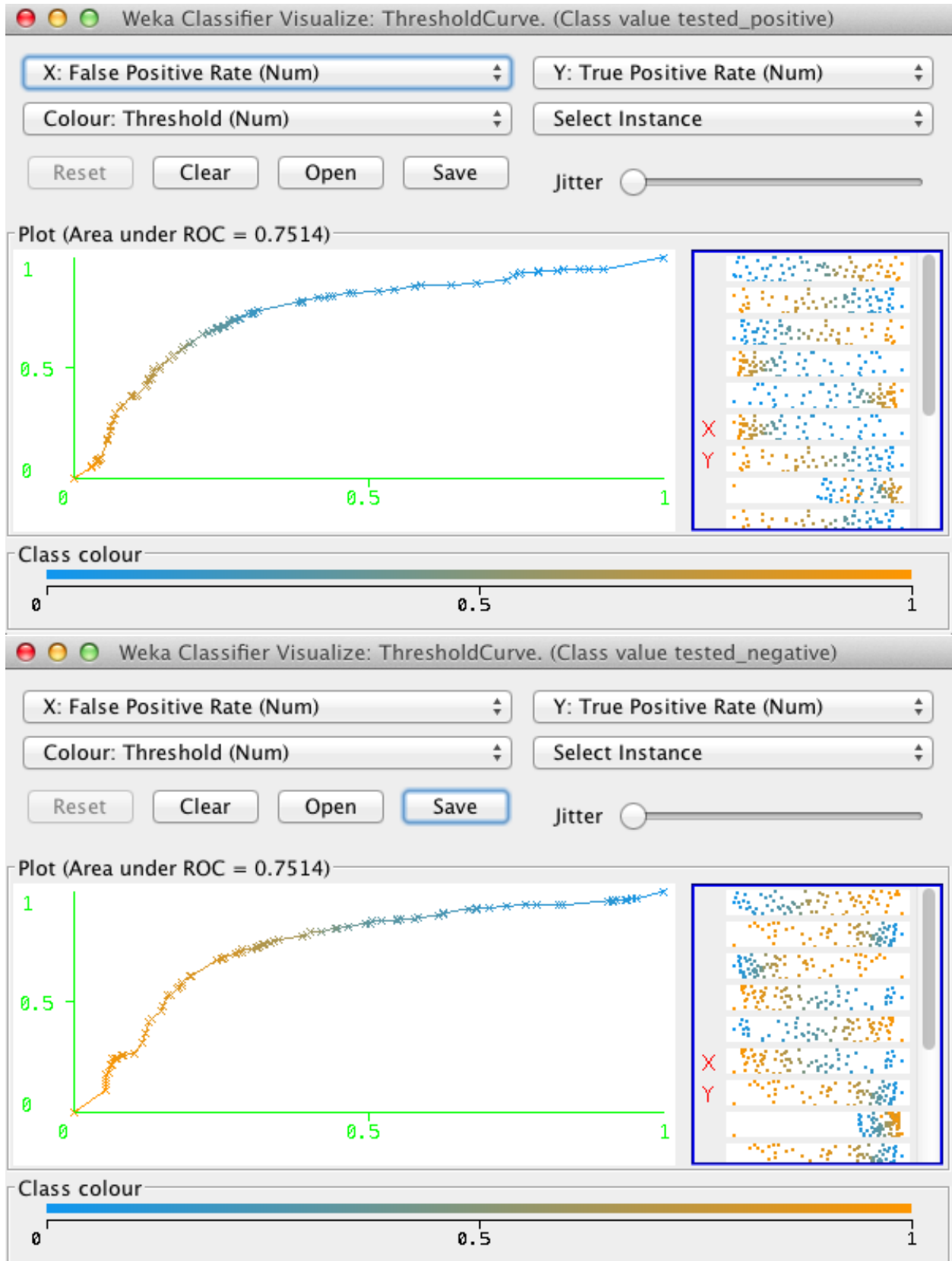
7. **recall (sensitivity)** the fraction of relevant instances returned. Recall of 0.814 means that if there model were to predict diabetes from the population and in there were 268 real cases, the model returned  $268 \times 0.814 = 218$  people (leaving 50 people with possible diabetes un-noticed).

## 1.2 What can be learned from this output?

The model and algorithm that was used to create the model can be rated, based on the results. E.g. we want to achieve a model that predicts with sensitivity of 95% then this model can be dismissed because it doesn't (has 81.4% sensitivity).



## 2 Understanding Receiving Operating Characteristic



The area under the curve or ROC score is 0.7514. It means that the classifier could be made ~25% better.

### 3 Characterizing ROC curves

What you can say about this ROC curve? How this classifier differs from a random guess?

Pick one point on a curve and interpret it using examples and illustrations.

For example, this point represents a classifier that can detect x% of all patients, who have a disease, but y% those who have not, are classified incorrectly. . . .

1. First algorithm makes very few mistakes overall. The number of mistakes increases as goodness the score drops. It can detect 83.5% of the cases correctly.
2. Model A is better because it classifies 84% vs 75.3% (12.7% better).

Compare two ROC curves. Which one is a better model and why?

3. I would always prefer A because it has high precision with both low and high goodness rates.

Compare two ROC curves. When algorithm A would be preferred over algorithm B?

### 4 Calculating confusion matrix and drawing ROC curve

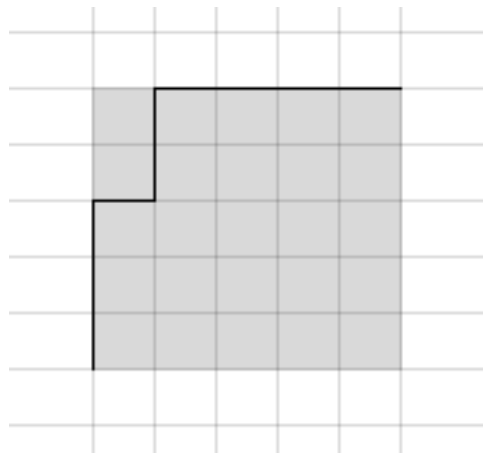


Figure 2: ROC for example data

Confusion matrix:

5 (TP)	1 (FP)
1 (FN)	3 (TN)

1. Precision:  $TP / (TP + FP)$ ,  $5 / (5 + 1) = 0.833(3)$
2. Recall:  $TP / (TP + FN)$ ,  $5 / (5 + 1) = 0.833(3)$

Area under the curve:  $23/25=0.92$

## 5 Analysing example data with Weka

Weka run output:

```
=== Run information ===
```

```
Scheme:weka.classifiers.trees.J48 -C 0.25 -M 2
```

```
Relation:      unbalanced
```

```
Instances:     856
```

```
Attributes:    33
```

```
WBN_GC_L_0.25
```

```
WBN_GC_H_0.25
```

```
WBN_GC_L_0.50
```

```
WBN_GC_H_0.50
```

```
WBN_GC_L_0.75
```

```
WBN_GC_H_0.75
```

```
WBN_GC_L_1.00
```

```
WBN_GC_H_1.00
```

```
WBN_EN_L_0.25
```

```
WBN_EN_H_0.25
```

```
WBN_EN_L_0.50
```

```
WBN_EN_H_0.50
```

```
WBN_EN_L_0.75
```

```
WBN_EN_H_0.75
```

```
WBN_EN_L_1.00
```

```
WBN_EN_H_1.00
```

```
WBN_LP_L_0.25
```

```
WBN_LP_H_0.25
```

```
WBN_LP_L_0.50
```

```
WBN_LP_H_0.50
```

```
WBN_LP_L_0.75
```

```
WBN_LP_H_0.75
```

```
WBN_LP_L_1.00
```

```
WBN_LP_H_1.00
```

```
XLogP
```

```
PSA
```

```
NumRot
```

```
NumHBA
```

```
NumHBD
```

```
MW
```

```
BBB
```

```
BadGroup
```

```
Outcome
```

```
Test mode:10-fold cross-validation
```

```
=== Classifier model (full training set) ===
```

```
J48 pruned tree
```

```
-----
```

```
: Inactive (856.0/12.0)
```

```
Number of Leaves :    1
```

```
Size of the tree :    1
```



Time taken to build model: 0.03 seconds

=== Stratified cross-validation ===

=== Summary ===

Correctly Classified Instances	844	98.5981 %
Incorrectly Classified Instances	12	1.4019 %
Kappa statistic	0	
Mean absolute error	0.0276	
Root mean squared error	0.1176	
Relative absolute error	95.7636 %	
Root relative squared error	99.9943 %	
Total Number of Instances	856	

=== Detailed Accuracy By Class ===

	TP Rate	FP Rate	Precision	Recall	F-Measure	ROC Area	Class
	0	0	0	0	0	0.432	Active
	1	1	0.986	1	0.993	0.432	Inactive
Weighted Avg.	0.986	0.986	0.972	0.986	0.979	0.432	

=== Confusion Matrix ===

```
a  b  <-- classified as
0  12 |  a = Active
0  844 |  b = Inactive
```

## 6 Quiz

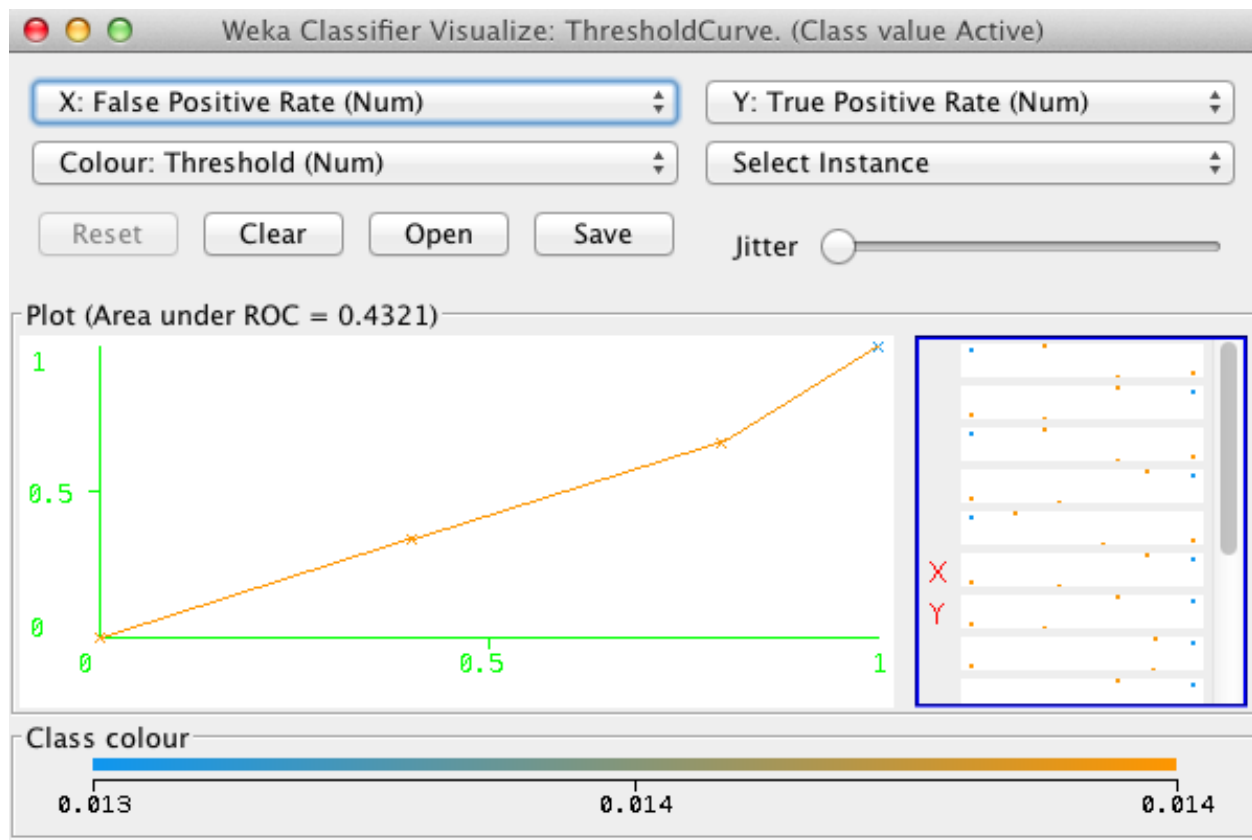
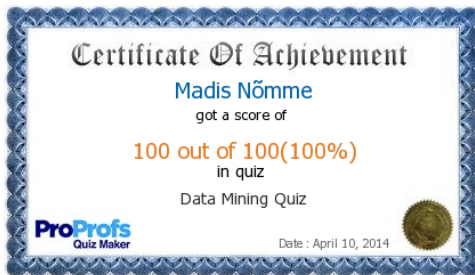


Figure 3: Task 5 ROC curve

Hope you enjoyed the quiz! Your score report is presented below.

#### Your Certificate



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#### Your Score

<b>Name</b>	Madis Nõmme	<b>Score</b>	100 / 100 Points( 100% )
<b>Answered Correctly</b>	10 Questions	<b>Answered Incorrectly</b>	0 Questions
<b>Time taken</b>	4 mins 43 secs		

#### Your Result

Result » Grade A

#### Your Answers

Print Report

Figure 4: Quiz result