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Q1: Distinguish between Classification and Regression with the help of relevant scenarios.

Answer:

Classification and Regression are two major prediction problems which are usually dealt in Data mining. Predictive modeling is the technique of developing a model or function using the historic data to predict the new data. The significant difference between Classification and Regression is that classification maps the input data object to some discrete labels. On the other hand, regression maps the input data object to the continuous real values.

Let's take an **example in classification**, suppose we want to predict the possibility of the rain in some regions on the basis of some parameters. Then there would be two labels rain and no rain under which different regions can be classified.

Let's take the similar **example in regression** also, where we are finding the possibility of rain in some particular regions with the help of some parameters. In this case, there is a probability associated with the rain. Here we are not classifying the regions within rain and no rain labels instead we are classifying them with their associated probability.

Comparison Chart

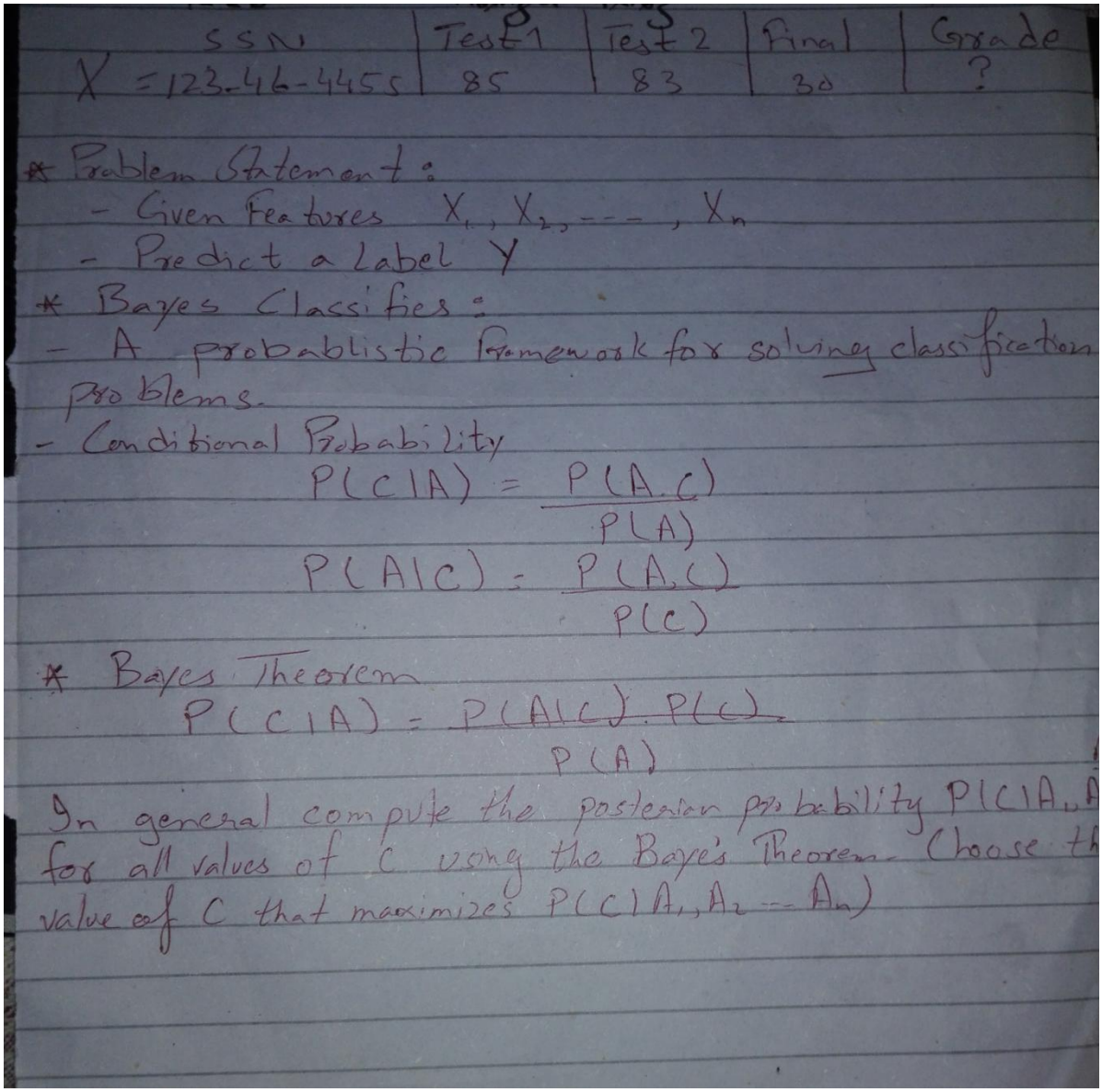
COMPARING BASE	CLASSIFICATION	REGRESSION
Basic	The discovery of model or functions where the mapping of objects is done into predefined classes.	A devised model in which the mapping of objects is done into values.
Predicts	Discrete values	Continuous values
Algorithms used	Decision tree, logistic regression, etc.	Regression tree (Random forest), Linear regression, etc.
Nature of data	Unordered	Ordered
Calculation method	Measuring accuracy	Measurement of root mean square error

Q2: Perform Naïve Bayes or Decision tree classification for new instance where (SSN = 123-46-4455, Test1= 85, Test2= 31 and Final= 30) Find Grade.

"SSN"	"Test1"	"Test2"	"Final"	"Grade"
"123-45-6789"	100	83	49	"D"
"123-12-1234"	96	97	48	"D"
"567-89-0123"	60	40	44	"C"

"087-65-4321"	36	45	47	"B-"
"456-78-9012"	88	77	45	"A-"
"234-56-7890"	80	90	46	"C-"
"345-67-8901"	-1	4	43	"F"
"632-79-9939"	30	40	50	"B+"

Answer:



~~$P(c) = \frac{Nom}{Total\ Sample}$~~

* Class

$P(A-) = 1$		$P(A-) = 1/8$	$\Rightarrow \infty$
$P(B+) = 1$		$P(B+) = 1/8$	$\Rightarrow \infty$
$P(B-) = 1$		$P(B-) = 1/8$	$\Rightarrow \infty$
$P(C-) = 1$		$P(C-) = 1/8$	$\Rightarrow \infty$
$P(C) = 1$		$P(C) = 1/8$	$\Rightarrow \infty$
$P(D) = 2$		$P(D) = 2/8$	$\Rightarrow$
$P(F) = 1$		$P(F) = 1/8$	$\Rightarrow \infty$

* For Discreate Attributes

$$P(A_i | C_k) = |A_{ik}| / N_{kc}$$

where, $|A_{ik}|$ is the number of instances having attributes A_i and belongs to Class c_k

~~e.g $P(Status = married | No) = 4/7$~~ } ~~google example~~

~~$P(Refund = Yes | Yes) = 0/3 = 0$~~ }

① $P(X | Class = A-)$

$= P(Test1 = 85 | class = A-)$ *

$P(Test2 = 83 | (class = A-)$ *

$P(Final = 30 | (class = A-)$

* For Non-Discreate values we have to use Normal Distribu

$$P(A_i | C_i) = \frac{1}{\sqrt{2\pi \sigma_{ij}^2}} e^{\left(\frac{A_i - \mu_{ij}}{2\sigma_{ij}^2}\right)}$$

where μ_{ij} = mean

σ^2_{ij} = Varience

e.g $X = (5, 8, 7, 6, 9)$ → mean

$$\sigma^2 = \frac{\Sigma (X_i - \bar{X})^2}{n-1}$$ → no of sample

where $\bar{X} = \frac{5+8+7+6+9}{5} = 7$

$s^2 = \frac{(5-7)^2 + (8-7)^2 + (7-7)^2 + (6-7)^2 + (9-7)^2}{5-1}$ → P(

$s^2 = 2.5$

(1)

→ P (Test 1 = 85 | Class = A-)

$= \frac{1}{\sqrt{2\pi\sigma_{ij}^2}} \times e\left(\frac{A_i - \mu_{ij}}{2\sigma_{ij}^2}\right)$ —— (A)

$A_i = 85$ — (1)

$\mu_{ij} = \frac{88}{1} = 88$ — (2)

$\sigma_{ij}^2 = \frac{88-88}{1-1} = 0$ — (3)

Put in (A)

$= \frac{1}{\sqrt{2\pi(0)}} \times e\left(\frac{85-88}{2(0)}\right)$

$= \frac{1}{0} \times e\left(\frac{-3}{0}\right)$

$= \infty$

→ P (Test 2 = 83 | Class A-)

$A_i = 83$ — (1)

$\mu_{ij} = \frac{77}{1} = 77$ — (2)

$\sigma_{ij}^2 = \frac{77-77}{1-1} = 0$ — (3)

Put in (A) ⇒ $\frac{1}{\sqrt{2\pi(0)^2}} \times e\left(\frac{83-77}{2(0)}\right)$

$= \frac{1}{0} \times e\left(\frac{6}{0}\right)$

$= \infty$

$)^2 + (9-7)^2$ → P(Final Test1 = 30 | class = A-)

$A_i = 30$

$\mu_{ij} = \frac{45}{1} = 45$

$\sigma_{ij}^2 = \frac{45-45}{1-1} = 0$

PA in (A)

$= \frac{1}{\sqrt{2\pi(0)}} \propto e\left(\frac{30-45}{2(0)}\right)$

$= \infty$

(2) → $P(Test1 = 85 \mid Class = B+)$

$= \frac{1}{2\pi\sigma_{ij}^2} \times e\left(\frac{A_i - \mu_{ij}}{2\sigma_{ij}^2}\right)$ ——— (A)

$A_i = 85$

$\mu_{ij} = 30/1 = 30$

$\sigma_{ij}^2 = 30 - 30/1-1 = 0$

PA in A get ∞

⇒ $P(Test2 = 83 \mid Class\ B+)$

$= \frac{1}{2\pi\sigma_{ij}^2} \times e\left(\frac{A_i - \mu_{ij}}{2\sigma_{ij}^2}\right)$

$A_i = 83$

$\mu_{ij} = 40$

$\sigma_{ij}^2 = = 0$

Put in A get ∞

⇒ $P(Final = 80 \mid Class\ B+)$

$A_i = 80$

$\mu_{ij} = 80$

$\sigma_{ij}^2 = 0$

PA in A ∞

(5) → $P(\text{Test 1} = 85 \mid \text{Class} = D)$

$$= \frac{1}{\sqrt{2\pi\sigma_{ij}^2}} e\left(\frac{A_i - \mu_{ij}}{2\sigma_{ij}^2}\right)$$

$A_i = 85$

$\mu_{ij} = \frac{100+96}{2} = \frac{196}{2} = 98$

$\sigma_{ij}^2 = \frac{(100-98)^2 + (96-98)^2}{2-1} = \frac{(2)^2 + (-2)^2}{1} = \frac{4+4}{1} = 8$

Put in (A)

$$= \frac{1}{\sqrt{2\pi(8)}} e\left(\frac{85-98}{2(8)}\right)$$

$$= \frac{1}{\sqrt{50}} e\left(\frac{-13}{16}\right)$$

$$= \frac{1}{7} e(-0.8)$$

$$= 0.1 \times 0.4$$

$$= 0.4$$

$P(\text{Test 2} = 83 \mid \text{Class } D)$

$$= \frac{1}{\sqrt{2\pi\sigma_{ij}^2}} e\left(\frac{A_i - \mu_{ij}}{2\sigma_{ij}^2}\right)$$

$A_i = 83$

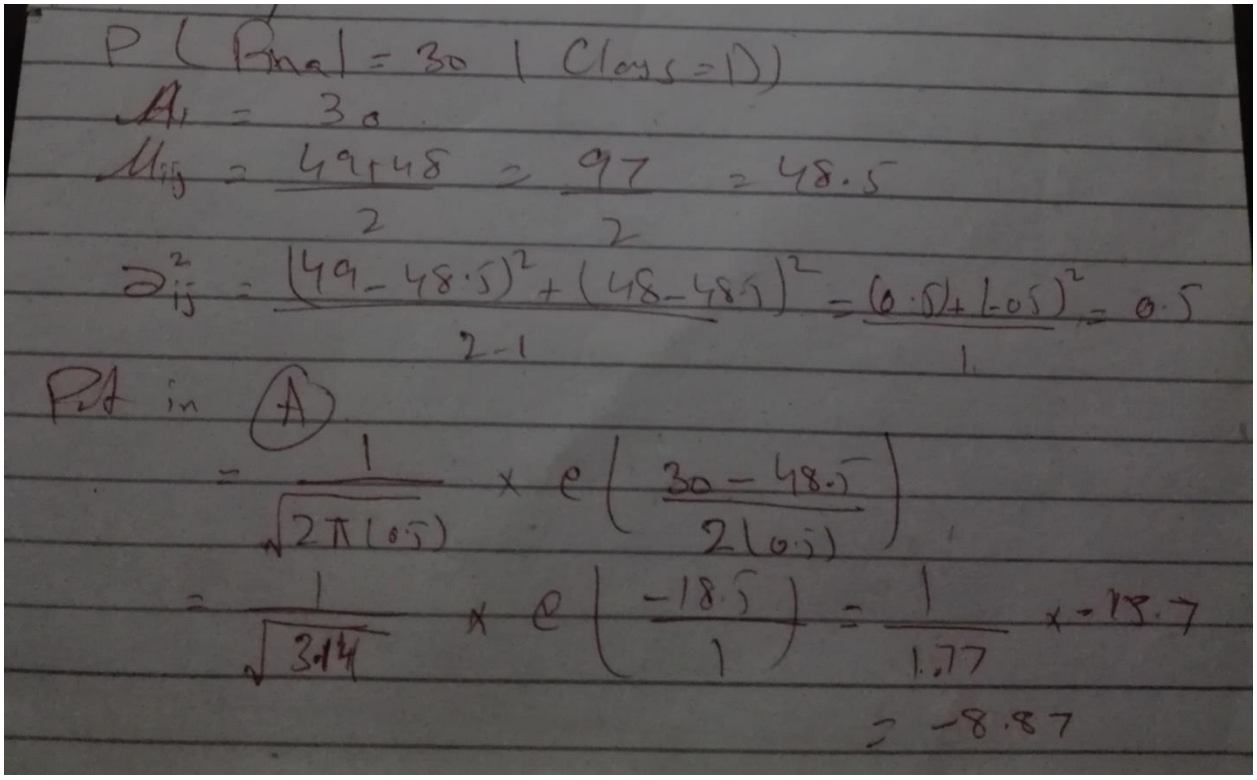
$\mu_{ij} = \frac{83+97}{2} = \frac{180}{2} = 90$

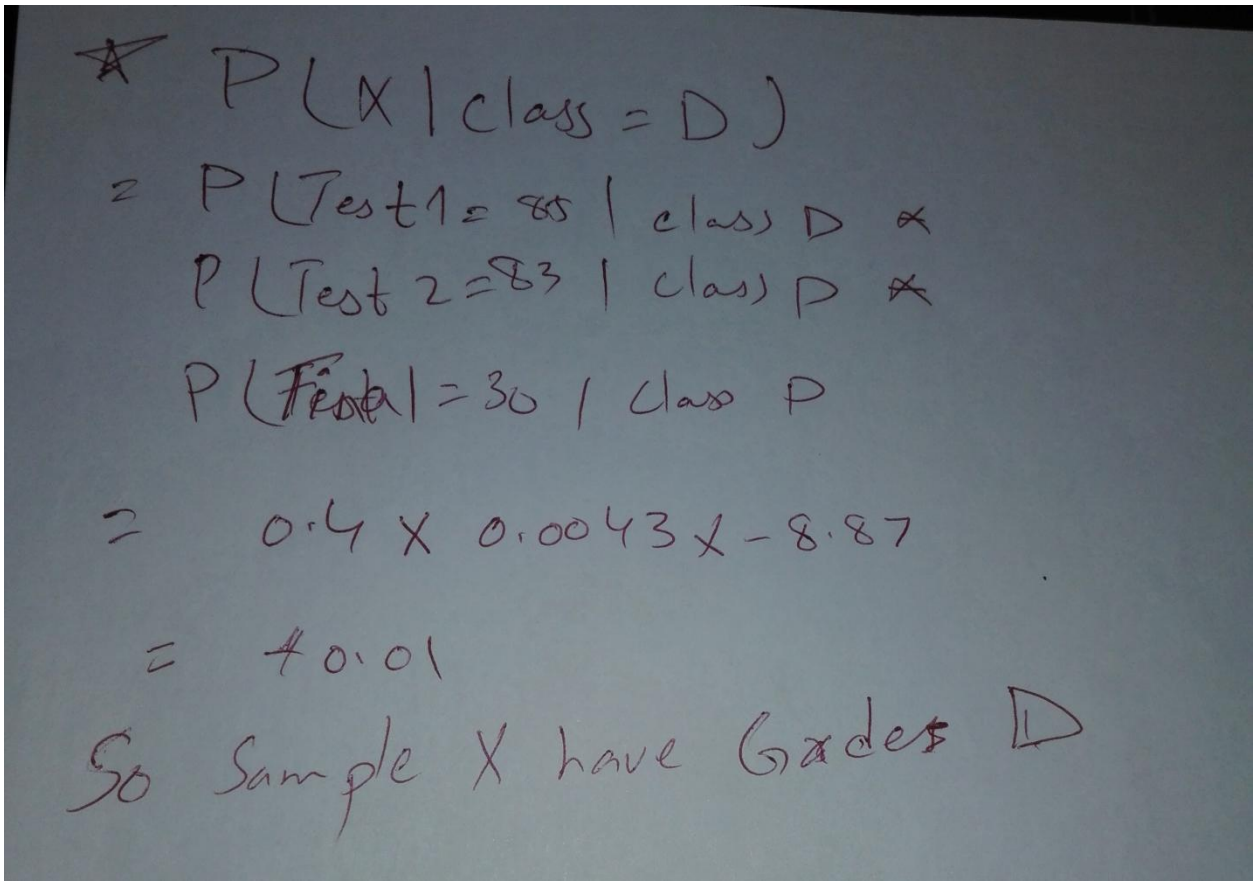
$\sigma_{ij}^2 = \frac{(83-90)^2 + (97-90)^2}{2-1} = \frac{49+49}{1} = 98$

Put in (A) $= \frac{1}{2\pi(98)} e\left(\frac{83-90}{2(98)}\right)$

$$= \frac{1}{615} e\left(\frac{-7}{196}\right)$$

$$= 0.0016 \times e(-0.03) = 0.0043$$





Q3: Find a Dataset related to any field and perform several classification techniques (Naïve Bayes, Decision tree, SVM, or any) to predict a class of a new

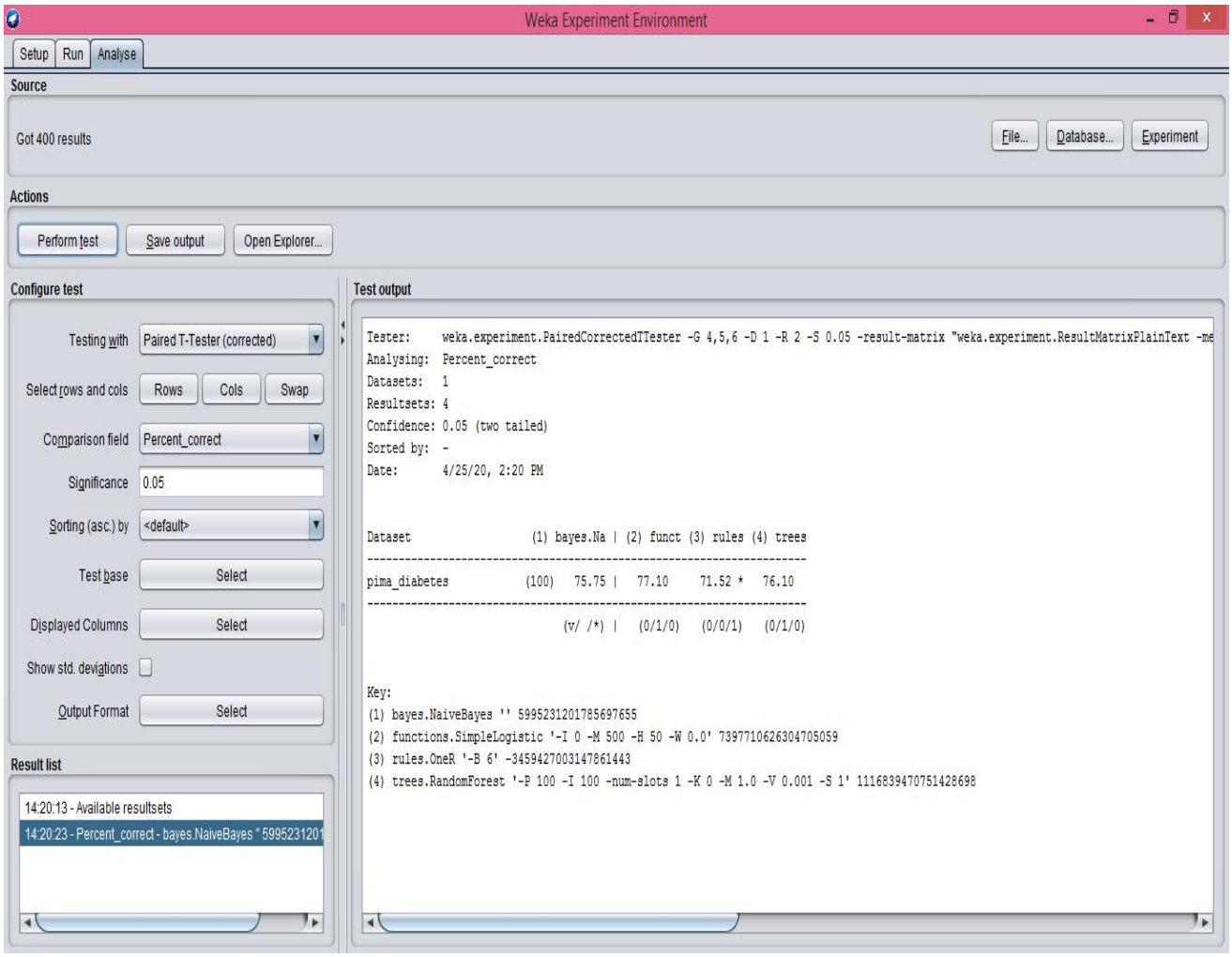
instance using WEKA. Compare the results (Accuracy, Precision, Recall, MARE, MMRE) of classification algorithms in a Table.

Take snapshots of the all the steps you perform for the classification.

Answer:

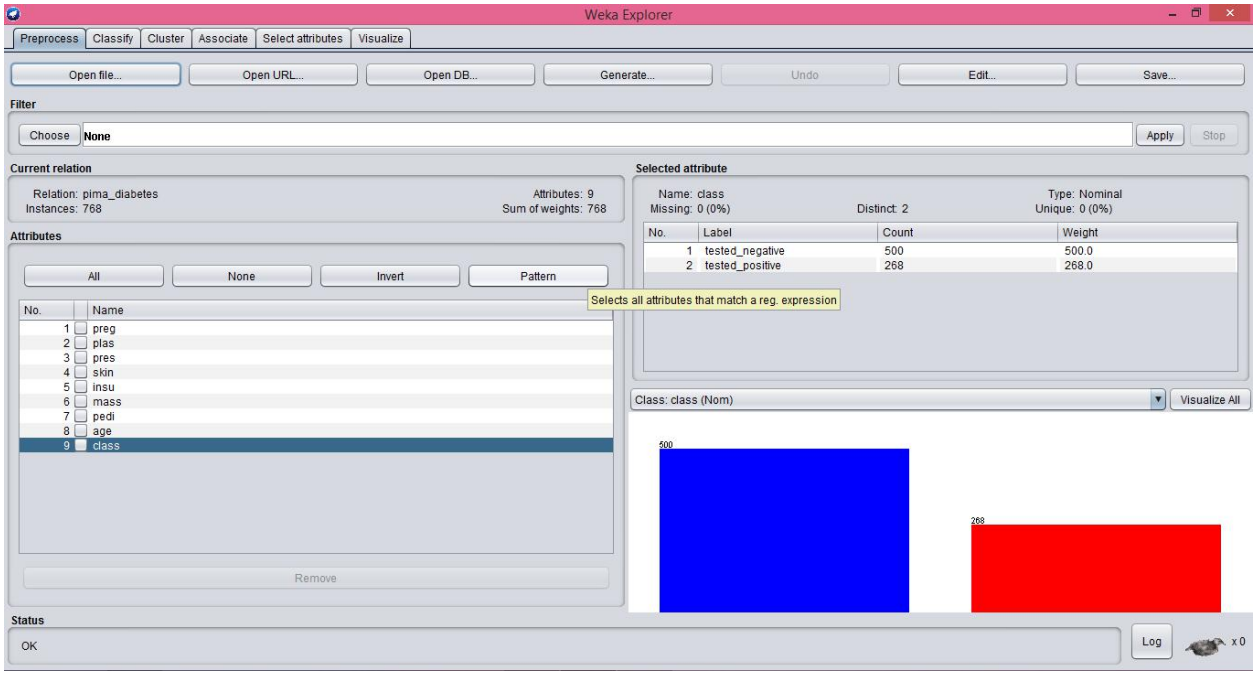
Dataset used: Weka default dataset named diabetes

Classification techniques applied: Simple Logistic, Naïve Bayesian, Random Forest and OneR

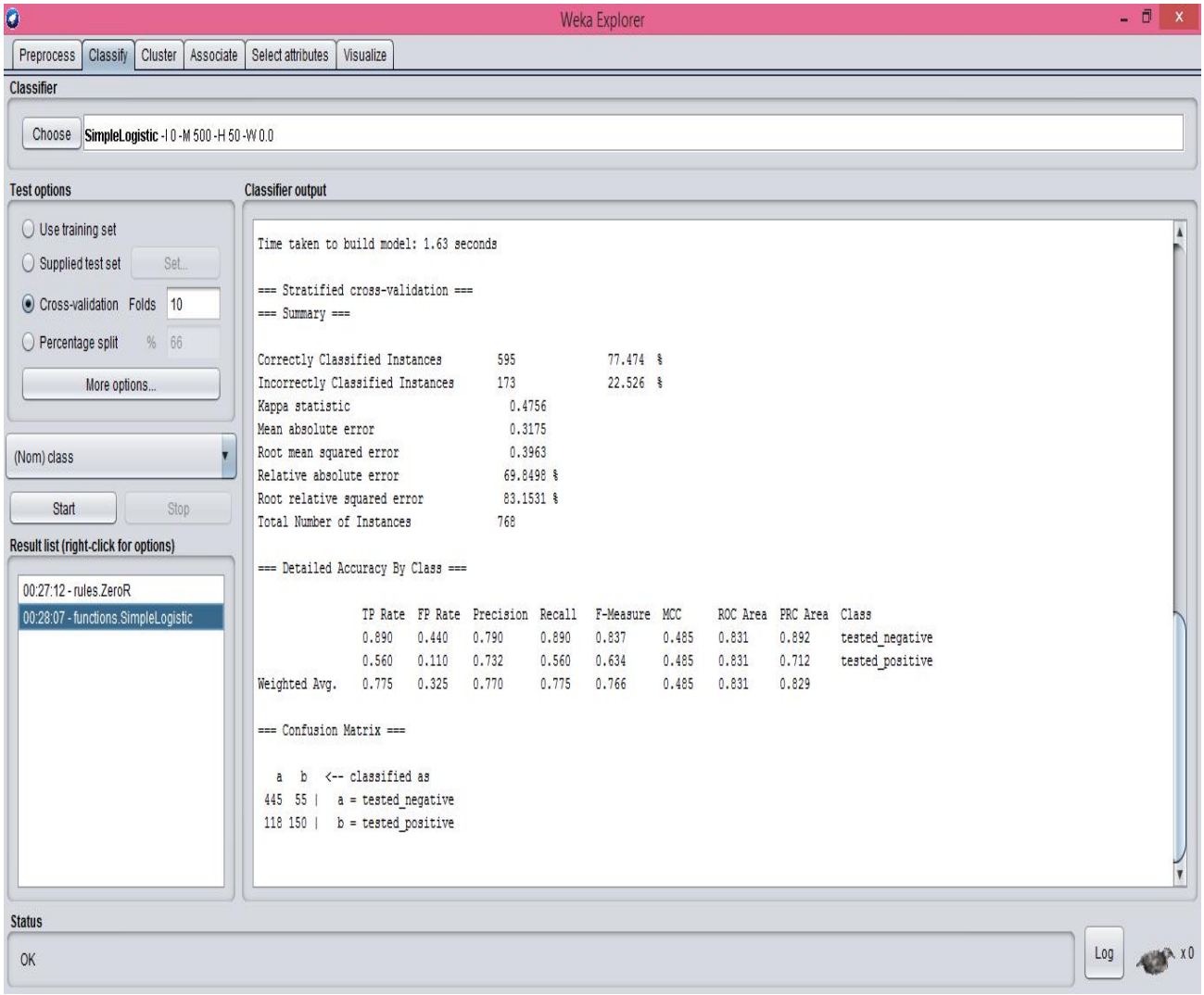


- OneR (71%) is significantly worse than Random Forest (76%)
- OneR (71%) is significantly worse than Naïve Bayesian (75%)
- OneR (71%) is significantly worse than Simple Logistic (77%)
- Shows that Simple Logistic (77%) performs significantly better for the particular data set.

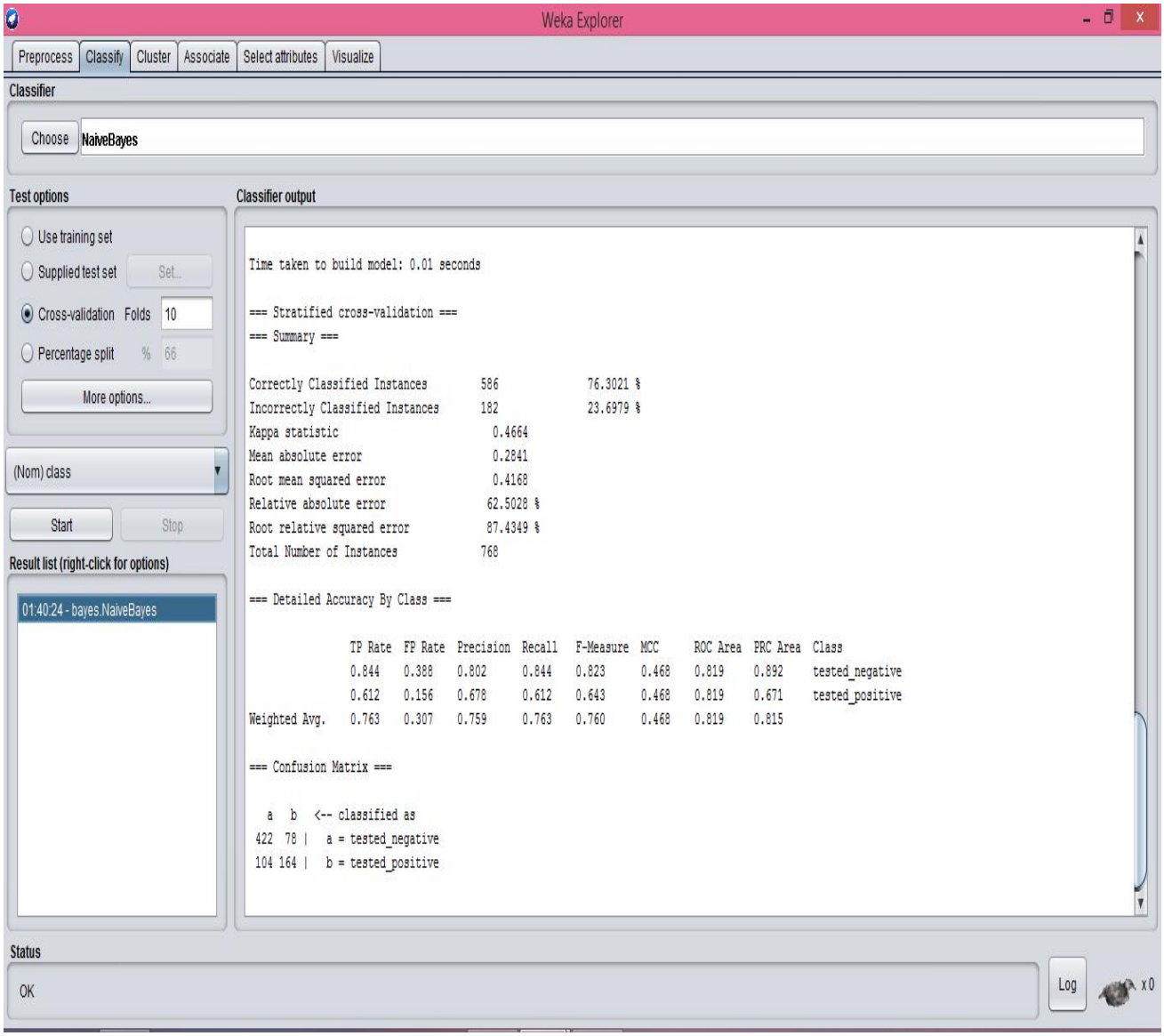
1. Below is the snap shot from WEKA explorer taking the diabetes data set showing its class with 2 attributes **1)** Tested Positive **2)** Tested Negative



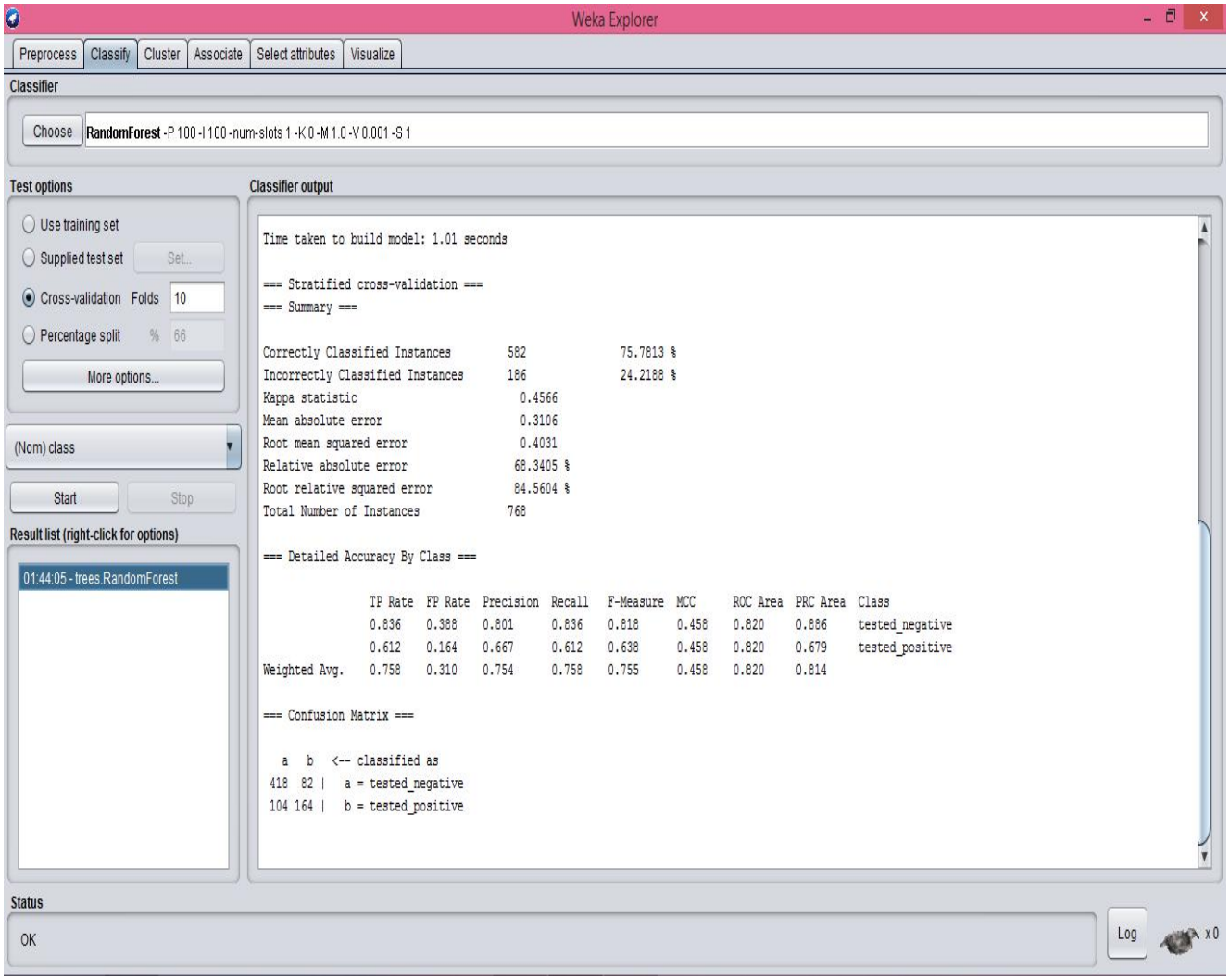
2. Below is the snap shot from Weka explorer by applying classifier Simple Logistic:



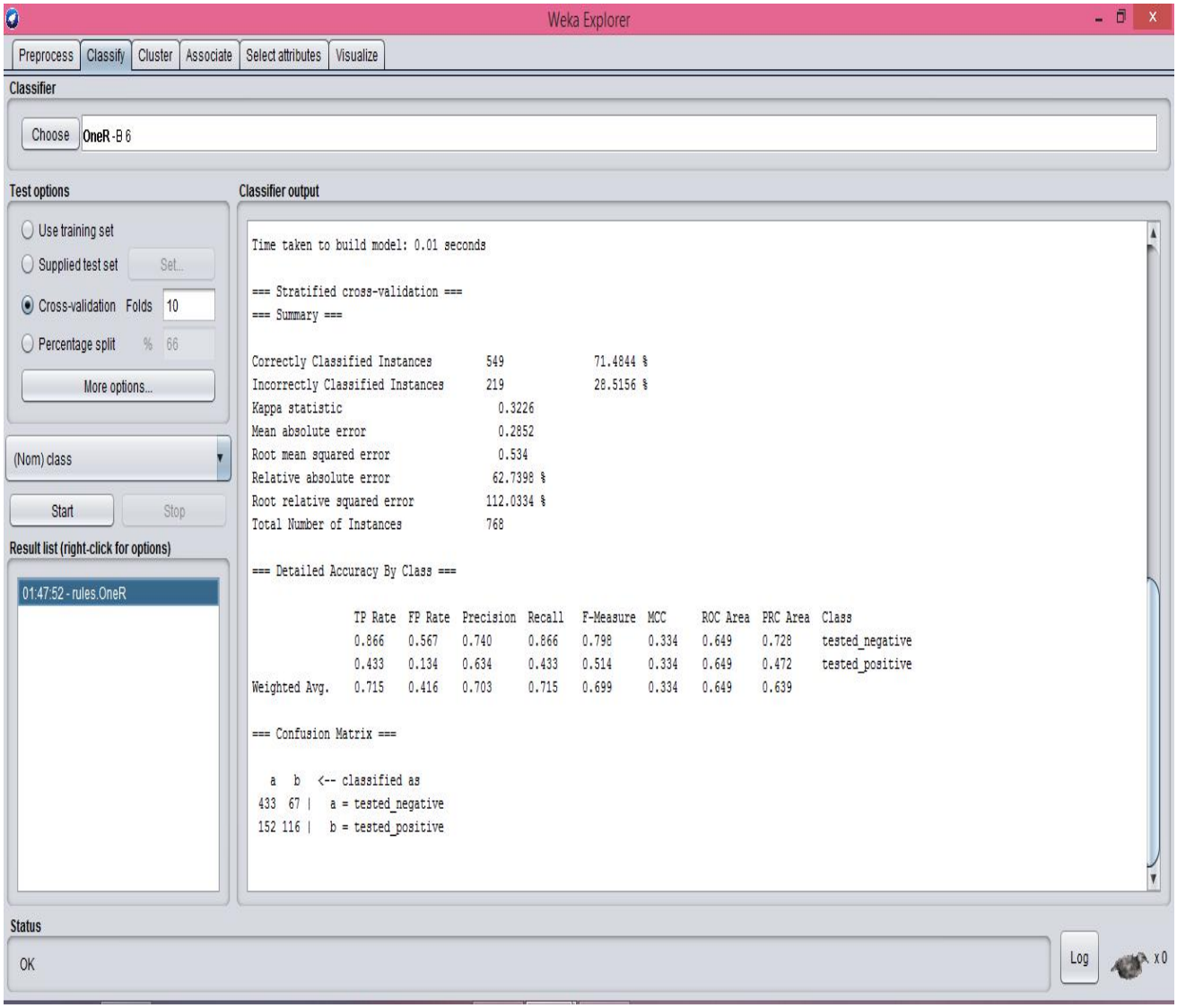
3. Below is the snap shot from Weka explorer by applying classifier Naïve Bayesian:



4. Below is the snap shot from Weka explorer by applying classifier Random Forest



5. Below is the snap shot from Weka explorer by applying classifier OneR



COMPARISON TABLE

Applied algorithms	Precision	Recall	F measure	MARE
Naive Baysian	0.759	0.763	0.760	62.5028%
Random Forest	0.754	0.758	0.755	68.3406%
Simple Logistic	0.770	0.775	0.766	69.84%
OneR	0.703	0.715	0.699	62.7298%