
1: Consider the perceptron $x+y-1 > 0$

Consider the training data:

X	Y	Label
0	0	+1
1	1	-1

Give the first 4 updates of the perceptron on this data, taking the order in which the data is given, i.e., loop over the training data in the order given. Give the prediction, state whether it is error or no, and give the updated perceptron

$w_0 = 1, w_1 = 1, w_2 = -1$ in the form $w_0x + w_1y + w_2 > 0$

Data Point(0, 0) \rightarrow label = -1 : Incorrectly Classified, update needed

$w_0 = 0 - 1 = -1, w_1 = 0 - 1 = -1$, line update: $-x - y - 1 > 0$

Data Point (1,1) \rightarrow label = -1: Correctly classified, No line update

Datapoint (0,0) \rightarrow label = -1: Update Needed

$w_0 = 0 - (-1) = 1, w_1 = 0 - (-1) = 1$: line update: $x + y - 1 > 0$

Data point (1, 1) \rightarrow label = + 1: Update Needed

$w_0 = 1 + 1 = 2, w_1 = 1 + 1 = 2$: line update: $2x + 2y - 1 > 0$

Datapoint (0, 0) \rightarrow label = -1: Update Needed

$w_0 = 0 - 2 = -2, w_1 = 0 - 2 = -2$: line update: $-2x - 2y - 1 > 0$

1(Extra Credit): Give the updated perceptrons till convergence

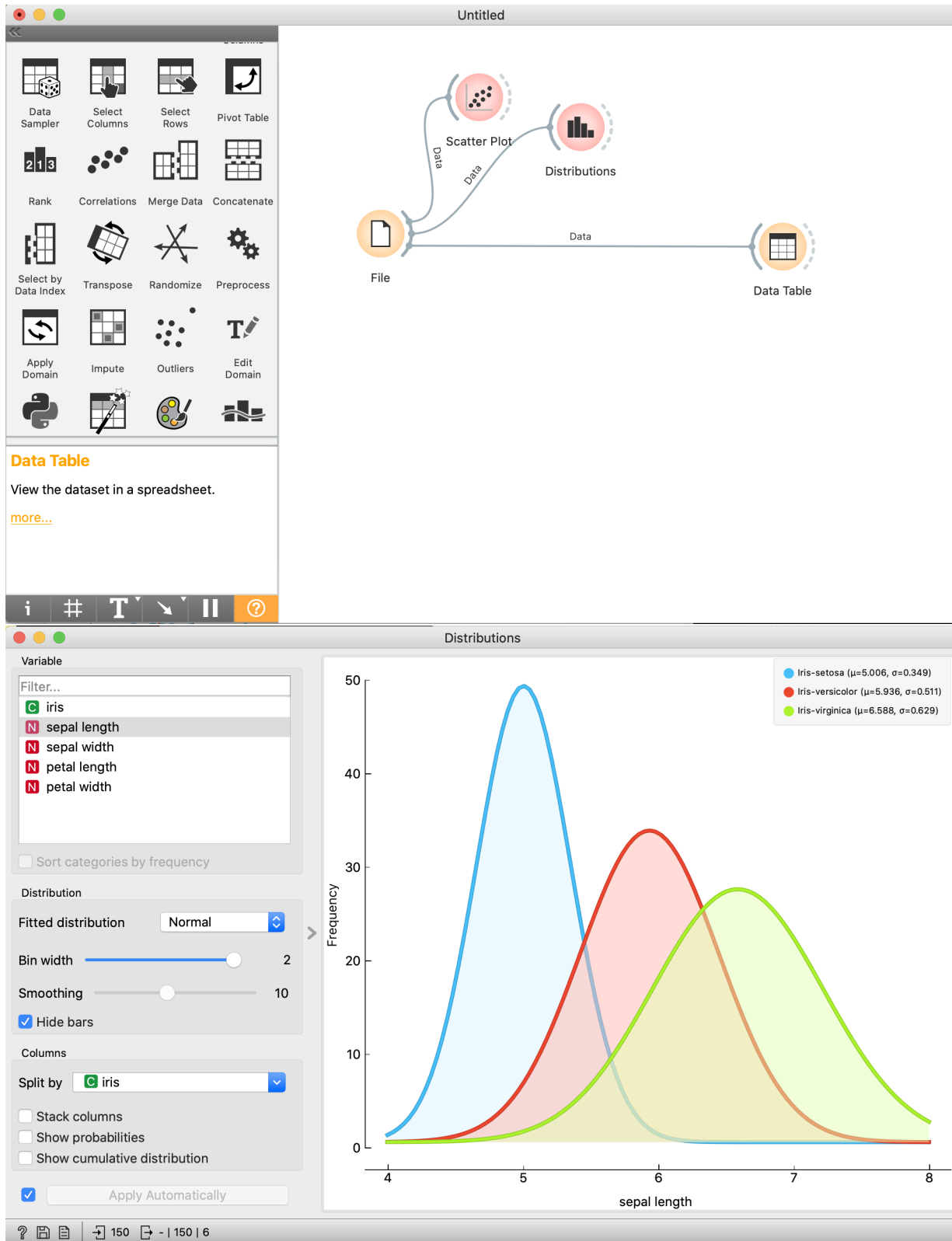
2: Download Orange

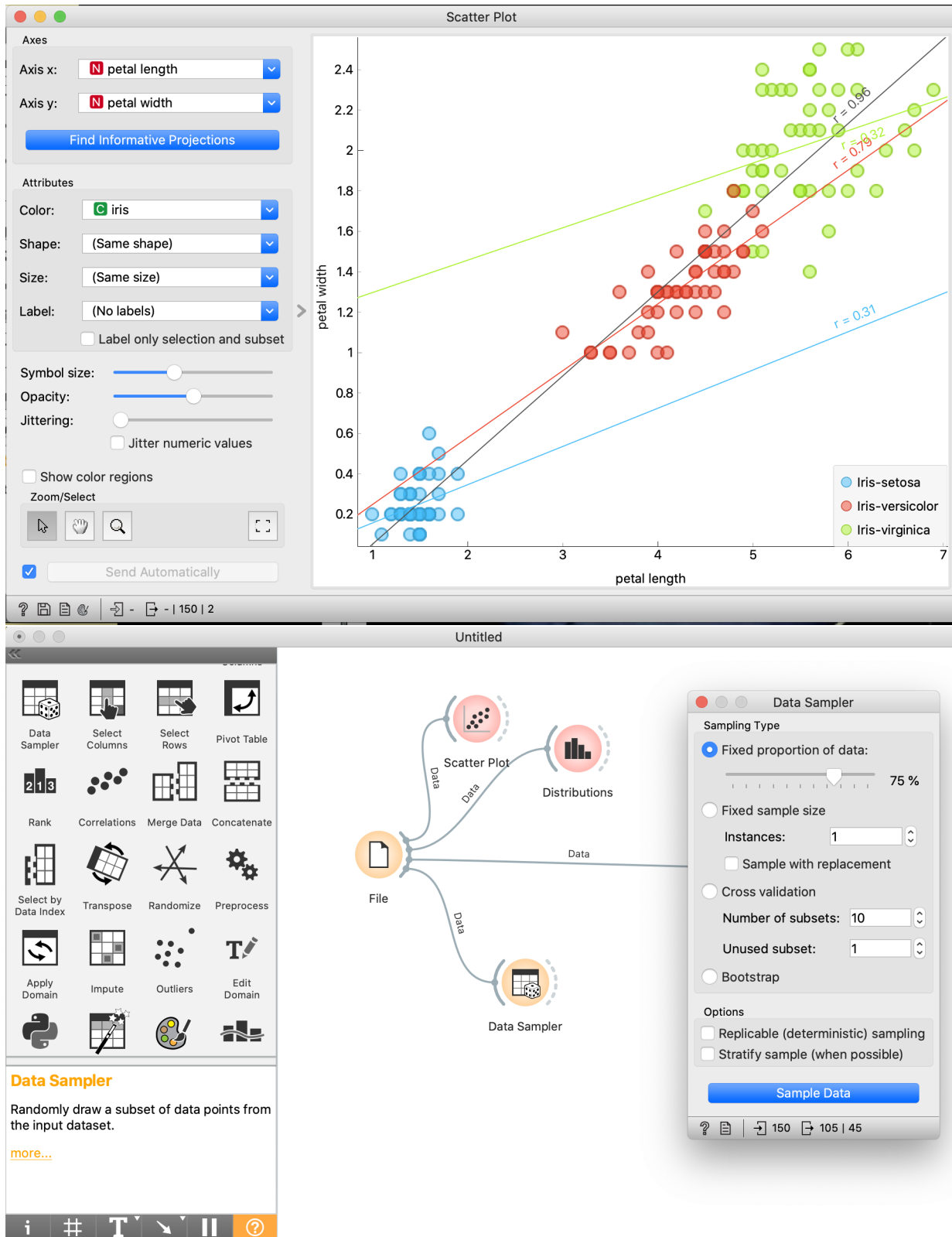
Use iris data set - divide into training and testing

to create models of decision tree, neural network, naive bayesian, KNN

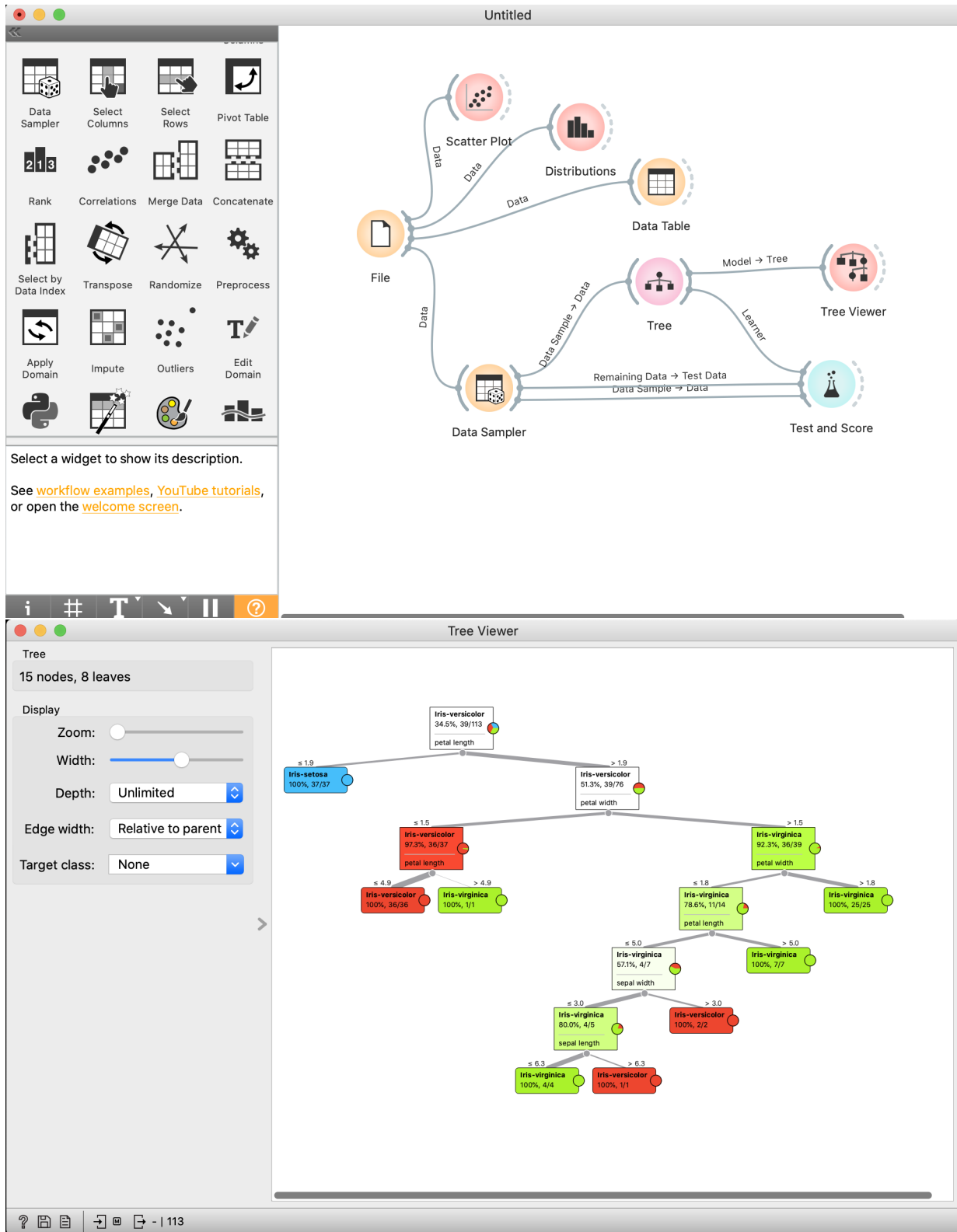
give training and testing errors

give screen shot of the process flow.





Decision Tree



Training Scores:

Test and Score

Sampling

☐ Cross validation

Number of folds: 5

☒ Stratified

☐ Cross validation by feature

☐ Random sampling

Repeat train/test: 10

Training set size: 66 %

☒ Stratified

☐ Leave one out

☒ Test on train data

☐ Test on test data

Target Class

(Average over classes)

Model Comparison

Area under ROC curve

☐ Negligible difference: 0.1

Evaluation Results

Model	AUC	CA	F1	Precision	Recall
Tree	1.000	1.000	1.000	1.000	1.000

Model Comparison by AUC

	Tree
Tree	

Table shows probabilities that the score for the model in the row is higher than that of the model in the column. Small numbers show the probability that the difference is negligible.

Test data is present but unused. Select 'Test on test data' to use it.

Testing Scores:

Test and Score

Sampling

☐ Cross validation

Number of folds: 5

☒ Stratified

☐ Cross validation by feature

☐ Random sampling

Repeat train/test: 10

Training set size: 66 %

☒ Stratified

☐ Leave one out

☐ Test on train data

☒ Test on test data

Target Class

(Average over classes)

Model Comparison

Area under ROC curve

☐ Negligible difference: 0.1

Evaluation Results

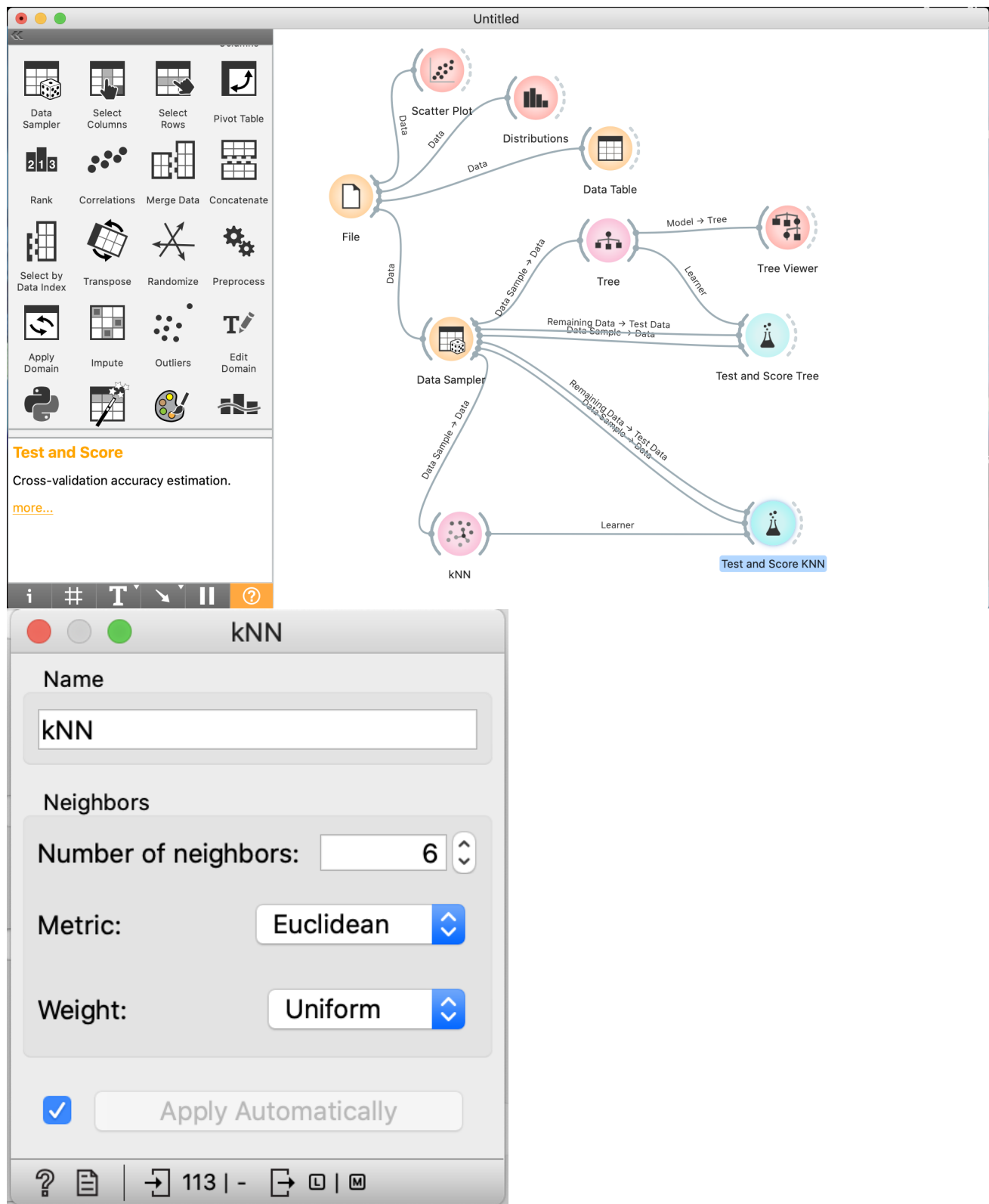
Model	AUC	CA	F1	Precision	Recall
Tree	0.979	0.973	0.973	0.975	0.973

Model Comparison by AUC

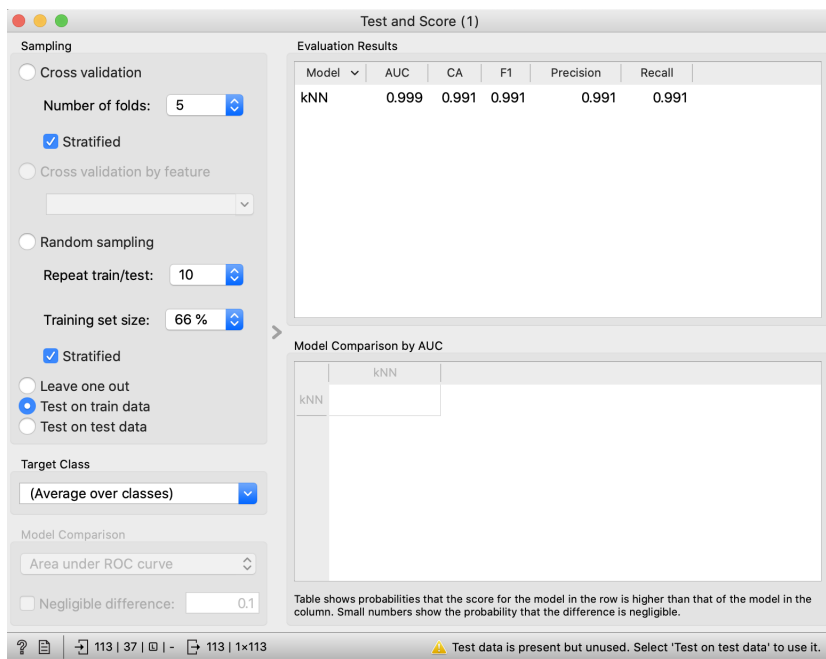
	Tree
Tree	

Table shows probabilities that the score for the model in the row is higher than that of the model in the column. Small numbers show the probability that the difference is negligible.

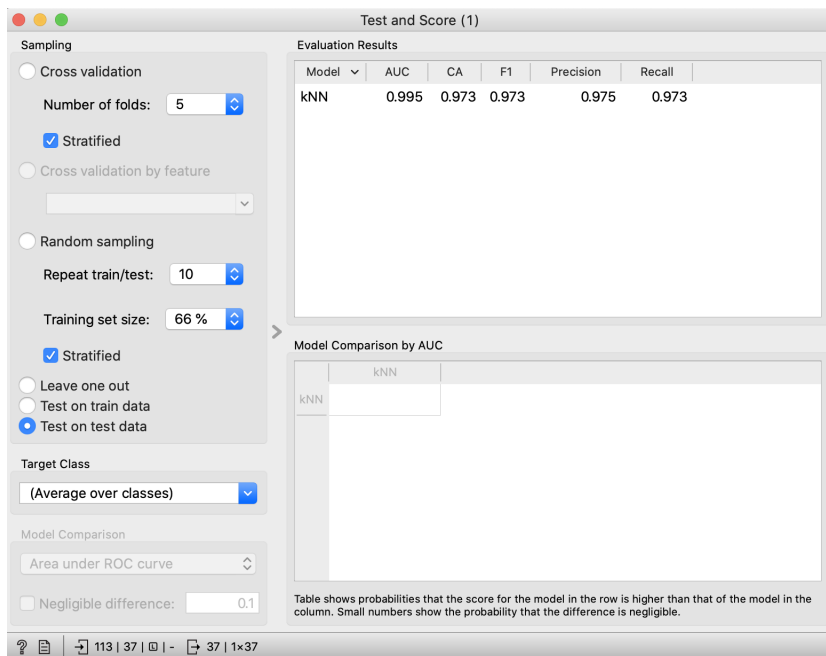
kNN



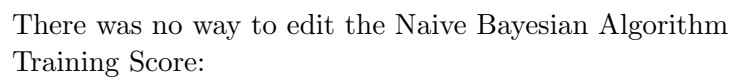
Training Scores:

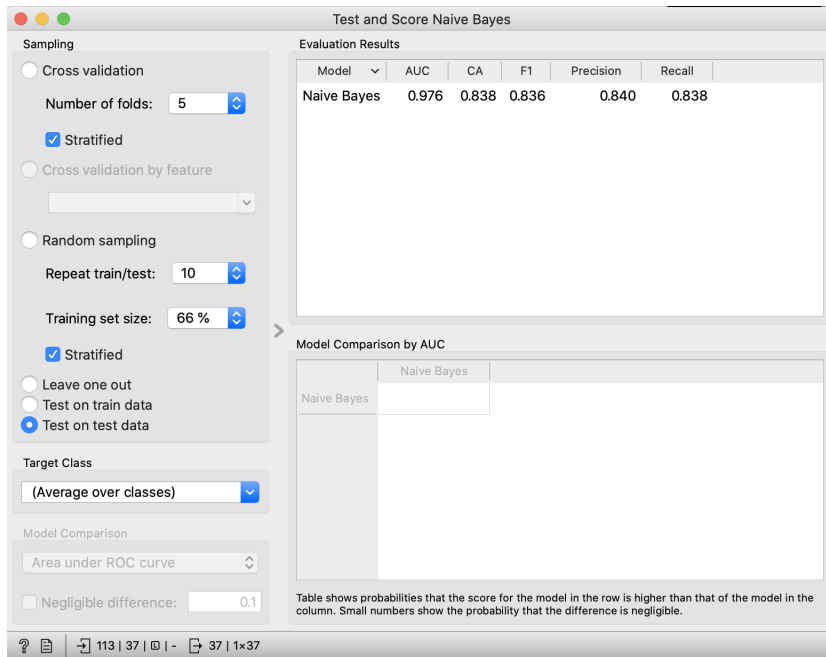


Testing Scores:

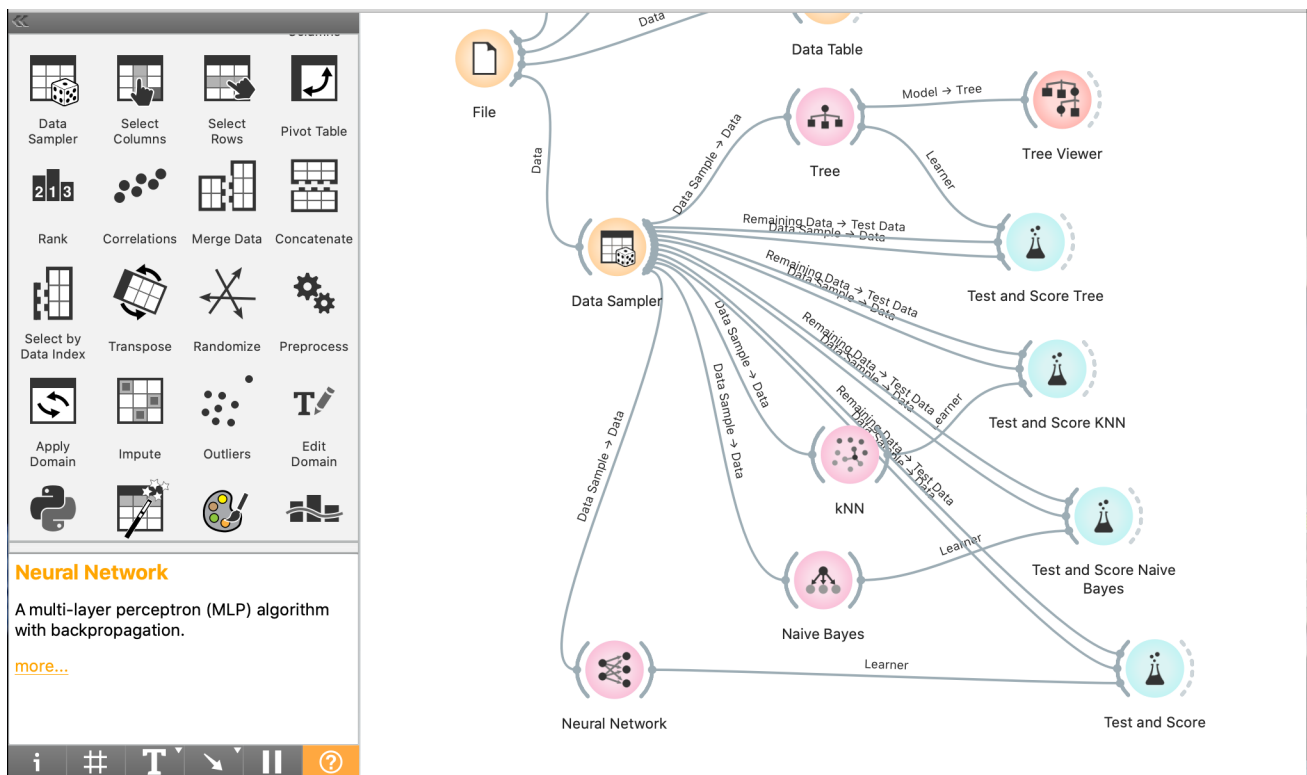


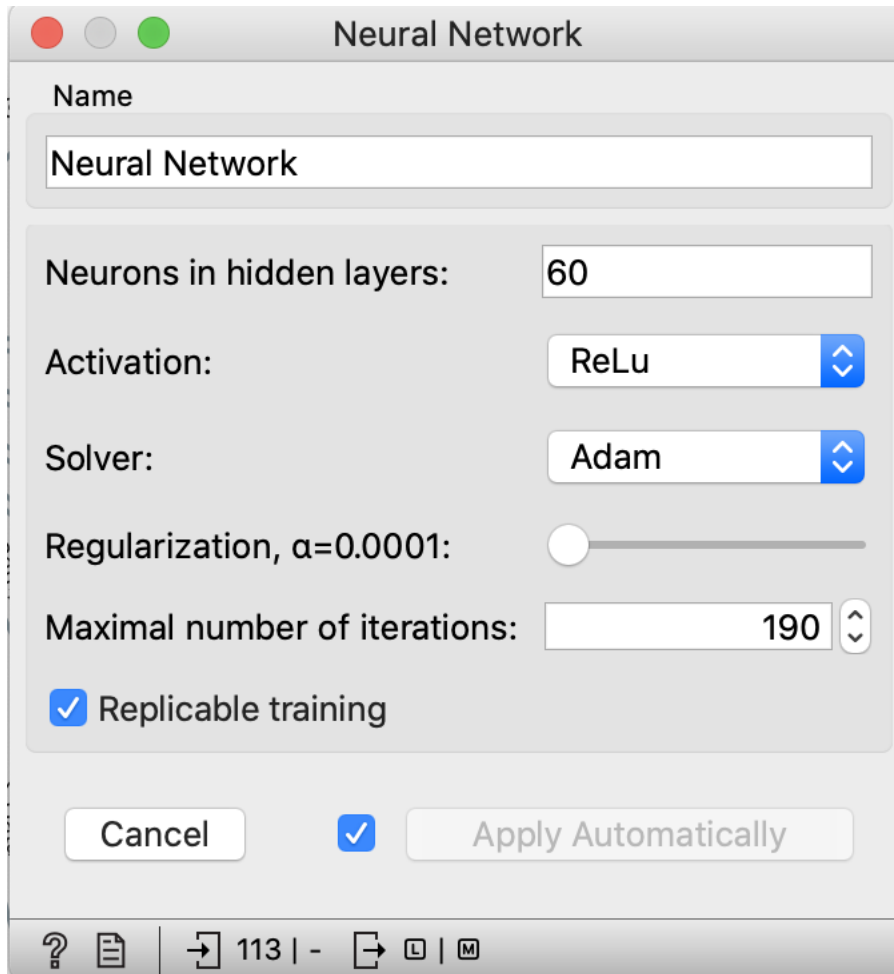
Naive Bayesian





Neural Network





A screenshot of a macOS-style window titled "Neural Network". The window has a title bar with red, yellow, and green window control buttons. Inside, there is a "Name" field containing "Neural Network". Below this, there are several configuration options: "Neurons in hidden layers:" with a text field containing "60"; "Activation:" with a dropdown menu showing "ReLu"; "Solver:" with a dropdown menu showing "Adam"; "Regularization, $\alpha=0.0001$:" with a slider control; "Maximal number of iterations:" with a text field containing "190" and a small up/down arrow button; and a checked checkbox labeled "Replicable training". At the bottom, there are three buttons: "Cancel", a checked checkbox, and "Apply Automatically". The bottom of the window features a status bar with icons for help, document, and a zoom level of "113 | -", along with window management icons.

Neural Network

Name

Neural Network

Neurons in hidden layers: 60

Activation: ReLu

Solver: Adam

Regularization, $\alpha=0.0001$:

Maximal number of iterations: 190

☒ Replicable training

Cancel ☒ Apply Automatically

Training Score:

Test and Score

Sampling

☐ Cross validation

Number of folds: 5

☒ Stratified

☐ Cross validation by feature

☐ Random sampling

Repeat train/test: 10

Training set size: 66 %

☒ Stratified

☐ Leave one out

☒ Test on train data

☐ Test on test data

Target Class

(Average over classes)

Model Comparison

Area under ROC curve

☐ Negligible difference: 0.1

Evaluation Results

Model	AUC	CA	F1	Precision	Recall
Neural Network	0.997	0.965	0.965	0.965	0.965

Model Comparison by AUC

	Neural Network
Neural Network	

Table shows probabilities that the score for the model in the row is higher than that of the model in the column. Small numbers show the probability that the difference is negligible.

Test data is present but unused. Select 'Test on test data' to use it.

Testing Score:

Test and Score

Sampling

☐ Cross validation

Number of folds: 5

☒ Stratified

☐ Cross validation by feature

☐ Random sampling

Repeat train/test: 10

Training set size: 66 %

☒ Stratified

☐ Leave one out

☐ Test on train data

☒ Test on test data

Target Class

(Average over classes)

Model Comparison

Area under ROC curve

☐ Negligible difference: 0.1

Evaluation Results

Model	AUC	CA	F1	Precision	Recall
Neural Network	0.993	0.919	0.919	0.921	0.919

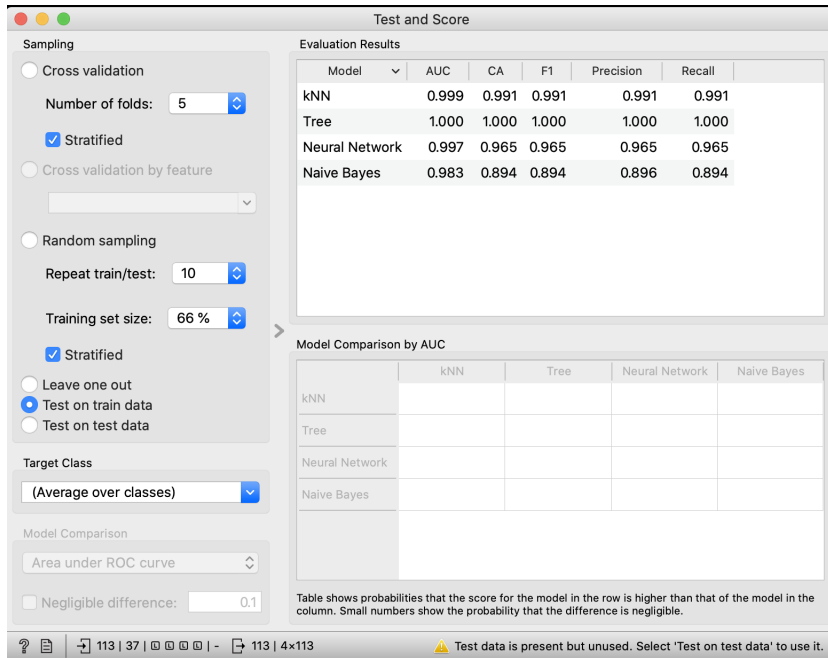
Model Comparison by AUC

	Neural Network
Neural Network	

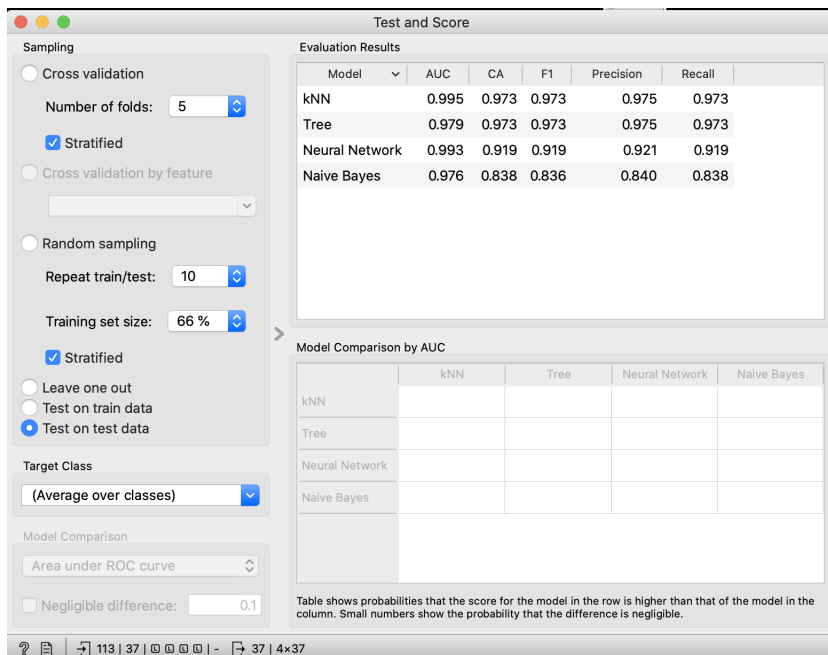
Table shows probabilities that the score for the model in the row is higher than that of the model in the column. Small numbers show the probability that the difference is negligible.

All together scores:

Training:



Testing:



3: Backprop

Consider the example network and initial weights given in

<https://mattmazur.com/2015/03/17/a-step-by-step-backpropagation-example/>
compute the weight updates for W_5 and W_1 with when target outputs are <0.02 ,

0.98>

$$h1_{net} = 0.3775$$

$$out_{h1} = 0.59326992$$

$$out_{o1} = 0.75136507$$

$$out_{o2} = 0.772928465$$

$$Error_{o1} = \frac{1}{2}(0.98 - 0.75136507)^2 = 0.26744$$

$$Error_{o2} = \frac{1}{2}(0.02 - 0.772928465)^2 = 0.02143$$

$$Error_{total} = Error_{o1} + Error_{o2} = 0.26744 + 0.02143 = 0.28887$$

$$\frac{\delta Error_{total}}{\delta w_5} = \frac{\delta Error_{total}}{\delta out_{o1}} * \frac{\delta out_{o1}}{\delta net_{o1}} * \frac{\delta net_{o1}}{\delta w_5}$$

$$\frac{\delta Error_{total}}{\delta out_{o1}} = -(target_{o1} - out_{o1}) = -(0.02 - 0.75136507) = 0.73136507$$

$$\frac{\delta out_{o1}}{\delta net_{o1}} = out_{o1}(1 - out_{o1}) = 0.75136507(1 - 0.75136507) = 0.186815602$$

$$\frac{\delta net_{o1}}{\delta w_5} = out_{h1} = 0.59326992$$

$$\frac{\delta Error_{total}}{\delta w_5} = \frac{\delta Error_{total}}{\delta out_{o1}} * \frac{\delta out_{o1}}{\delta net_{o1}} * \frac{\delta net_{o1}}{\delta w_5} = 0.73136507 * 0.186815602 * 0.59326992 = 0.08105870993$$

$$\text{New value of } w_5 = w_5 - \frac{\delta Error_{total}}{\delta w_5} = 0.4 - 0.08105870993 = 0.31894129007$$

$$\frac{\delta Error_{total}}{\delta w_1} = \frac{\delta Error_{total}}{\delta out_{h1}} * \frac{\delta out_{h1}}{\delta net_{h1}} * \frac{\delta net_{h1}}{\delta w_1}$$

$$\frac{\delta net_{h1}}{\delta w_1} = i_1 = 0.05$$

$$\frac{\delta out_{h1}}{\delta net_{h1}} = out_{h1}(1 - out_{h1}) = 0.59326992(1 - 0.59326992) = 0.231300709$$

$$\frac{\delta Error_{total}}{\delta out_{h1}} = \frac{\delta Error_{o1}}{\delta out_{h1}} + \frac{\delta Error_{o2}}{\delta out_{h1}}$$

$$\frac{\delta Error_{o1}}{\delta out_{h1}} = \frac{\delta Error_{o1}}{\delta net_{o1}} * \frac{\delta net_{o1}}{\delta out_{h1}}$$

$$\frac{\delta Error_{o1}}{\delta out_{o1}} = -(target_{o1} - out_{o1}) = -(0.02 - 0.75136507) = 0.73136507$$

$$\frac{\delta out_{o1}}{\delta net_{o1}} = out_{o1}(1 - out_{o1}) = 0.75136507(1 - 0.75136507) = 0.186815602$$

$$\frac{\delta Error_{o1}}{\delta net_{o1}} = \frac{\delta Error_{o1}}{\delta net_{o1}} * \frac{\delta net_{o1}}{\delta out_{h1}} = 0.73136507 * 0.186815602 = 0.13849562$$

$$\frac{\delta net_{o2}}{\delta out_{h1}} = w_5 = 0.40$$

$$\frac{\delta Error_{o1}}{\delta out_{h1}} = \frac{\delta Error_{o1}}{\delta net_{o1}} * \frac{\delta net_{o1}}{\delta out_{h1}} = 0.13849562 * 0.40 = 0.055399425$$

$$\frac{\delta Error_{o2}}{\delta out_{o2}} = -(target_{o2} - out_{o2}) = -(0.98 - 0.772928465) = -0.20707$$

$$\frac{\delta out_{o2}}{\delta net_{o2}} = out_{o2}(1 - out_{o2}) = 0.772928465(1 - 0.772928465) = 0.17551005299$$

$$\frac{\delta Error_{o2}}{\delta net_{o2}} = \frac{\delta Error_{o2}}{\delta net_{o2}} * \frac{\delta net_{o2}}{\delta out_{h1}} = -0.20707 * 0.17551005299 = -0.03634286667$$

$$\frac{\delta net_{o2}}{\delta out_{h1}} = w_6 = 0.45$$

$$\frac{\delta Error_{o2}}{\delta out_{h1}} = \frac{\delta Error_{o2}}{\delta net_{o2}} * \frac{\delta net_{o2}}{\delta out_{h1}} = -0.03634286667 * 0.45 = -0.01635429$$

$$\frac{\delta Error_{total}}{\delta out_{h1}} = \frac{\delta Error_{o1}}{\delta out_{h1}} + \frac{\delta Error_{o2}}{\delta out_{h1}} = 0.055399425 - 0.01635429 = 0.039045135$$

$$\frac{\delta Error_{total}}{\delta w_1} = \frac{\delta Error_{total}}{\delta out_{h1}} * \frac{\delta out_{h1}}{\delta net_{h1}} * \frac{\delta net_{h1}}{\delta w_1} = 0.039045135 * 0.231300709 * 0.05 = 0.00045155837$$

$$\text{new value of } w_1 = w_1 - \frac{\delta Error_{total}}{\delta w_1} = 0.15 - 0.00045155837 = 0.14954844163$$