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# Artificial Intelligence

## Training a Neural Network

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*This code is used to build, train and test a neural network.*

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## Load dataset

```
CellDNA = csvread('C:\tmp\CellDNA.csv');
```

## Define Y

```
Y = CellDNA(:, [14]);
```

## Convert Y to 0 and 1

Convert the Y target to binary values, where Y = 0 means bacterium is not interesting to study

```
Y(Y ~= 0) = 1;  
Y(Y == 0) = 0;
```

## Convert Y to Dummy datatype

Y should be a dummy variable, but dummyvar function cannot accept a vector of 0s and 1s. Instead, all values must be positive.

```
Y = Y + 1;  
Y = dummyvar(Y);
```

```
% Then transpose. NN requires Y be transposed
Y_t = Y.';
```

## Define and Standardize X

```
% Define X
XAll = CellDNA(:,1:13);

% Standardize X
XAll = zscore(XAll);

% Then transpose. NN requires Y be transposed
X_t = XAll.';
```

## Define NN Architecture

We're going with a neuron structure of 2 hidden layers, with the below named neuron count in each

```
NN = patternnet([15 5])

% Define the two hidden layers using function poslin
% poslin = Positive linear transfer function = ReLu
% <https://www.mathworks.com/help/deeplearning/ref/poslin.html poslin>
NN.layers{1}.transferFcn = 'poslin';
NN.layers{2}.transferFcn = 'poslin';

NN =

    Neural Network

        name: 'Pattern Recognition Neural Network'
    userdata: (your custom info)

    dimensions:

        numInputs: 1
        numLayers: 3
        numOutputs: 1
        numInputDelays: 0
        numLayerDelays: 0
        numFeedbackDelays: 0
        numWeightElements: 95
        sampleTime: 1

    connections:

        biasConnect: [1; 1; 1]
        inputConnect: [1; 0; 0]
        layerConnect: [0 0 0; 1 0 0; 0 1 0]
        outputConnect: [0 0 1]
```

*subobjects:*

```
    input: Equivalent to inputs{1}
    output: Equivalent to outputs{3}

    inputs: {1x1 cell array of 1 input}
    layers: {3x1 cell array of 3 layers}
    outputs: {1x3 cell array of 1 output}
    biases: {3x1 cell array of 3 biases}
    inputWeights: {3x1 cell array of 1 weight}
    layerWeights: {3x3 cell array of 2 weights}
```

*functions:*

```
    adaptFcn: 'adaptwb'
    adaptParam: (none)
    derivFcn: 'defaultderiv'
    divideFcn: 'dividerand'
    divideParam: .trainRatio, .valRatio, .testRatio
    divideMode: 'sample'
    initFcn: 'initlay'
    performFcn: 'crossentropy'
    performParam: .regularization, .normalization
    plotFcns: {'plotperform', plottrainstate, ploterrhist,
              plotconfusion, plotroc}
    plotParams: {1x5 cell array of 5 params}
    trainFcn: 'trainscg'
    trainParam: .showWindow, .showCommandLine, .show, .epochs,
               .time, .goal, .min_grad, .max_fail, .sigma,
               .lambda
```

*weight and bias values:*

```
    IW: {3x1 cell} containing 1 input weight matrix
    LW: {3x3 cell} containing 2 layer weight matrices
    b: {3x1 cell} containing 3 bias vectors
```

*methods:*

```
    adapt: Learn while in continuous use
    configure: Configure inputs & outputs
    gensim: Generate Simulink model
    init: Initialize weights & biases
    perform: Calculate performance
    sim: Evaluate network outputs given inputs
    train: Train network with examples
    view: View diagram
    unconfigure: Unconfigure inputs & outputs
```

## Train Model

Train a NN on X data using Y target This is a classification model, so use cross-entropy as loss function.

```
NN = train(NN, X_t, Y_t);
```

## Predict Model

With the model trained, we now predict

```
[Y_hat] = NN(X_t);
```

## Accuracy, Precision, Recall, FScore

First compute the confusion matrix, cfs

```
[conf, cfs] = confusion(Y_t, Y_hat)

% Class 0 (not flagged for study)
TP0 = cfs(1,1);
TN0 = cfs(2,2);
FP0 = cfs(1,2);
FN0 = cfs(2,1);
precision0 = TP0/(TP0 + FP0)
recall0 = TP0/(TP0 + FN0)
fscore0 = 2*((precision0 * recall0)/(precision0 + recall0));

% Class 1 (flagged for study)
TP1 = cfs(2:2,2);
TN1 = cfs(2,2);
FP1 = cfs(1:1,2);
FN1 = cfs(2:2,1);
precision1 = TP1/(TP1 + FP1)
recall1 = TP1/(TP1 + FN1)
fscore1 = 2*((precision1 * recall1)/(precision1 + recall1));

% Calculate accuracy
accuracy = sum(diag(cfs))/sum(cfs(:))

conf =

    0.1027

cfs =

    972    45
     80   120

precision0 =

    0.9558
```

```
recall0 =
```

```
    0.9240
```

```
precision1 =
```

```
    0.7273
```

```
recall1 =
```

```
    0.6000
```

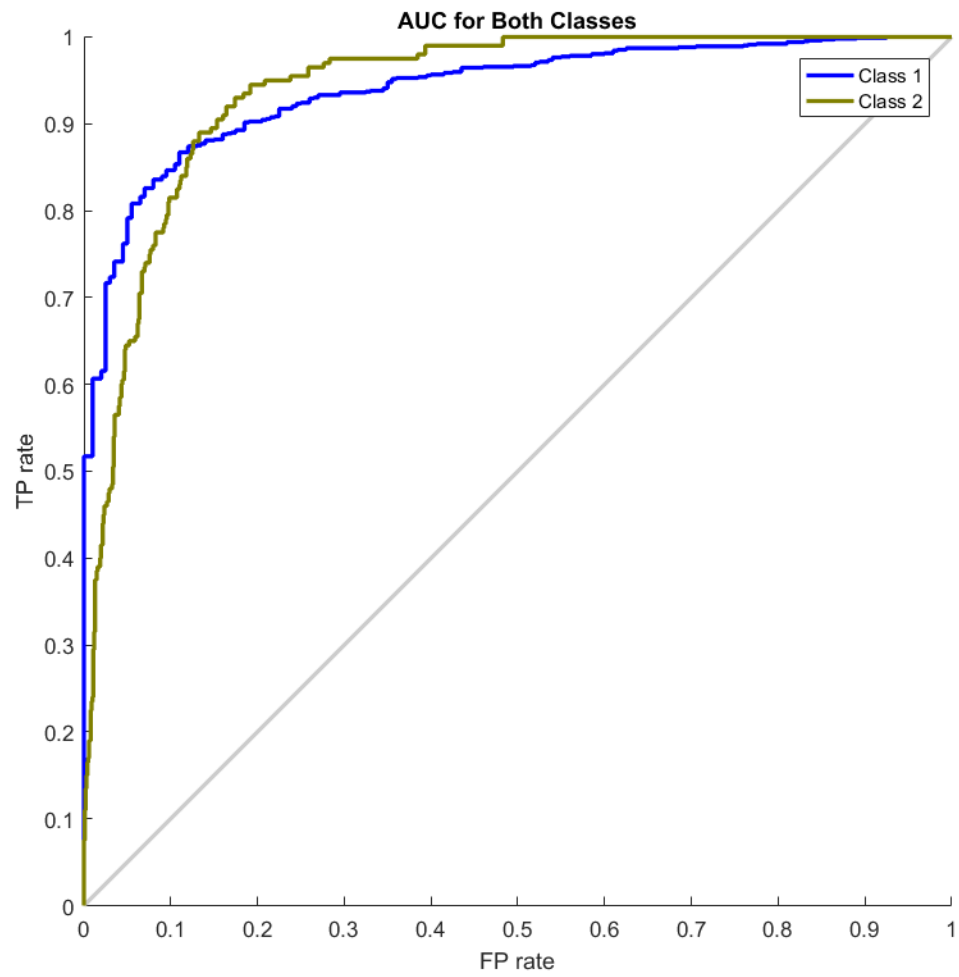
```
accuracy =
```

```
    0.8973
```

## Plot ROC curve

Place both lines on plot

```
figure(1)
plotroc(Y_t, Y_hat)
xlabel('FP rate'), ylabel('TP rate')
title('AUC for Both Classes')
```



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