Melanoma Skin Cancer Detection

Design of a computer-based system to automatically classify histopathologi

```
in[1]:= nImages = 100; (* number of sample images *)
     nTrainset = 70; (* number of images used for training the ANN *)
    data = Import["/home/gloria/Documents/ETH/Semester2/NN/ImageDatav2.xlsx",
        {"Data", 1, Range[2, nImages + 1], {1, 2, 3}}];
     (* data file has a header in row 1 *)
    target = Import["/home/gloria/Documents/ETH/Semester2/NN/ImageDatav2.xlsx",
        {"Data", 1, Range[2, nImages + 1], 4}];
     (* 4th column corresponds to the ground truth *)
     trainindex = RandomSample[Range[2, nImages], nTrainset];
     (* Randomly split images betweeen training and testing set *)
     testindex = Complement[Range[2, nImages], trainindex];
    Defining the ANN with 1 hidden layer
in[7]:= inNum = 3; hidNum = 3; outNum = 1; (* 3 input units,
    3 hidden units and 1 output unit *)
    wh = Table[Table[Random[Real, {-2.5, 2.5}], {inNum}], {hidNum}];
     (* hidden weights *)
    wo = Table[Table[Random[Real, {-2.5, 2.5}], {hidNum}], {outNum}];
     (* output weights *)
     eta = 0.5; (* learning rate *)
In[11]:=
    sigmoid[x_] := 1 / (1 + Exp[-x]);
    relu[x_] := Map[Max[0, \#] &, x]; (* activation functions *)
    l = nTrainset; (* number images in the training set *)
```

Importing Data - Training & Testing Set

```
In[13]:= (* funtion to train the ANN *)
     TrainEpochs[n_] := Module[{iter = nTrainset}, Table[
          If [k \le l, r = trainindex[[k]], r = trainindex[[Mod[k, l] + 1]]];
          in = data[[r]];
          y = target[[r]];
          outhid = sigmoid[wh.in];
          out = sigmoid[wo.outhid];
          e = y - out;
          outdelta = e out (1 - out);
          hiddelta = outhid (1-outhid) Transpose[wo].outdelta;
          wo += eta e.e Outer[Times, outdelta, outhid];
          wh += eta e.e Outer[Times, hiddelta, in];
          {e.e, wo, wh}, {k, 1, iter}]];
In[14]:= (* function to evaluate the ANN *)
     EvalNetwork[in_] := Module[{outhid}, outhid = sigmoid[wh.in];
        sigmoid[wo.outhid]]
In[15]:= (* function to compute the testing error *)
     testErrorfunc[n_] := Module[{}},
        RootMeanSquare[Table[target[[testindex[[i]]]] - EvalNetwork[data[[testindex[[i]]]]]],
            {i, 1, nImages - nTrainset - 1}] // Flatten] // N]
Training the ANN
In[16]:= nIterations = 15; (* number of iterations *)
     ErrorList = Table[{TrainEpochs[1], testErrorfunc[k]}, {k, 1, nIterations}];
In[18]:= colors = {Green, Red, Orange, Blue, Gray};
ln[19]:= trainplot = Table[ListPlot[ErrorList[[pp, 1, All, 1]] // Flatten,
          PlotJoined → True, Frame → True, FrameLabel → {"Image Number", "SquaredError"},
          PlotRange → All, PlotStyle → colors[[Mod[pp, 5]]]], {pp, 1, nIterations}];
In[20]:= Show[trainplot]
        0.4
                             Image Number
```

```
In[21]:= Performance Evaluation
ln[22]:= TP = 0; FP = 0; TN = 0;
    FN = 0; (* true positives, false positives, true negatives and false negatives *)
    For[i = 1, i ≤ Length[testindex], i++,
      in = data[[testindex[[i]]]];
      y = target[[testindex[[i]]]];
      outhid = sigmoid[wh.in];
      out = Round[sigmoid[wo.outhid]];
      Which[out[[1]] == y == 1, TP += 1, out[[1]] == y == 0, TN += 1,
       out[[1]] # y && out [[1]] == 0, FN += 1, out[[1]] # y && out[[1]] == 1, FP += 1
     ];
NSensitivity = TP / (TP + FN);
    (* Compute Precision, also called Positive Predictive Value (PPV)*)
    NPrecision = TP / (TP + FP);
    (*Compute True Negative Rate (TNR), also called specificity*)
    NSpecificity = TN / (TN + FP);
    (* Compute Accuracy*)
    NAccuracy = (TP + TN) / (TP + TN + FP + FN);
In[28]:= Print["Sensitivity", "=", N[NSensitivity, 3]]
    Print["Precision", "=", N[NPrecision, 3]]
    Print["Specificity", "=", N[NSpecificity, 3]]
    Print["Accuracy", "=", N[NAccuracy, 3]]
    Sensitivity=1.00
    Precision=0.933
    Specificity=0.933
    Accuracy=0.966
```

Visualizing the Classification Task

```
In[32]:= signal = Position[target, _?(# == 1. &)] // Flatten;
     bckg = Position[target, _?(# == 0. &)] // Flatten;
     signaldata = Map[data[[#]] &, signal];
     bckgdata = Map[data[[#]] &, bckg];
     ip = ListPointPlot3D[{signaldata, bckgdata}];
     op =
        ContourPlot3D[EvalNetwork[\{xx, yy, zz\}] = 1 / 2, \{xx, -4, 4\}, \{yy, -4, 4\}, \{zz, -4, 4\}];
     Show [
      op,
       ip]
Out[38]=
        -2
```

Assessing Overfitting - Training & Testing Error

```
testError = Table[ErrorList[[i, 2]], {i, 1, nIterations}];
   ListLinePlot[{trainingError, testError}, Frame → True,
    FrameLabel → {"Training Epoch", "RootMeanSquare Error"},
    PlotLegends → {"Training", "Testing"}]
      0.4
   RootMeanSquare Error
      0.3
                                                 Training
                                                 Testing
      0.1
                               10
                                    12
                                         14
```

Training Epoch