```
In [1]: import sys
        import numpy as np
        import os
        # for reading sound files.
        from scipy.io import wavfile
        # For training the neural network.
        # Install pybrain. e.g. pip install pybrain.
                                          import ClassificationDataSet
        from pybrain.datasets
                                          import buildNetwork
        from pybrain.tools.shortcuts
        from pybrain.supervised.trainers import BackpropTrainer
        from pybrain.structure.modules
                                         import SoftmaxLayer
        from pybrain.tools.xml.networkreader import NetworkReader
        from pybrain.tools.xml.networkwriter import NetworkWriter
        from pybrain.structure import FullConnection
        from pybrain.structure import SigmoidLayer
        from pybrain.structure import FeedForwardNetwork
        from pybrain.structure import LinearLayer
        from pybrain.structure import SoftmaxLayer
        from pybrain.structure import TanhLayer
        sq = lambda x: x*x
        # A convenient time step.
        deltaT = 0.2
        lim1 = 5; lim2 = 305
        noisy = 0.05 # We have a very clean speech sample!
        def getData(fil):
          sampFreq, snd = wavfile.read(fil)
          snd = snd/2E15
                                                 # put in range (-1,1)
          duration = snd.shape[0]/sampFreq
          numChunks = int(duration/deltaT)
          sizeChunk = int(len(snd)/numChunks)
          # Frequencies.
          freqs = np.fft.rfftfreq(sizeChunk,deltaT)
          chunksF = []
          for lp in range(0,numChunks):
            chunk = snd[lp*sizeChunk:(lp+1)*sizeChunk] # get a chunk o
        f speech.
            chunksF.append(1E9*np.abs(np.fft.rfft(chunk))) # take the fft,
        conveniently normalized.
          mu = np.mean(chunksF)
          newMean = 0.
          ctr = 0
          for i in range(0,numChunks):
```

```
for j in range(lim1,lim2):
      if abs(chunksF[i][j]) > noisy*mu: # ignore silent por
tions.
       newMean += chunksF[i][j]
       ctr += 1
 # Delete portions of the sound file when the user is not speaking
 mu = newMean/ctr
 zeros = []
 for lp in range(0,numChunks):
   if np.mean(chunksF[lp]) < noisy*mu: zeros.append(lp)</pre>
 data = []
 ctr = 0
  for i in range(0, numChunks):
   if i in zeros: continue
                                   # silent part.
   tmp = []
   for j in range(lim1,lim2): tmp.append(chunksF[i][j])
   data.append(tmp)
 return data
def writeData(data,outFile):
 # Write the speech waveform to file, for plotting purposes.
 l = len(data[0])
 f = open(outFile, "w")
 for i in range(0,len(data)):
   for j in range(0,1):
      f.write(str(i) + " " + str(j) + " " + str(data[i][j]) + "\n")
   f.write("\n")
  f.close()
def getSamples(folders):
 # Each sound file is divided into a number of samples.
 samples = []
  for folder in folders:
   tmp = [v for v in os.listdir(folder) if v <> ".DS Store"]
   samples.append(tmp)
 # Now samples[0] contains all the files for speaker 1,
 # samples[1] contains all the files for speaker 2, and so on.
 speechSpeakers = []
 ctr = 0
  for sample in samples:
   speechSpeaker = []
   for file in sample:
      lsts = getData(folders[ctr]+file)
      for lst in lsts: speechSpeaker.append(lst)
    speechSpeakers.append(speechSpeaker)
```

```
ctr += 1

return speechSpeakers

def XY(speech,numFeatures,N):

tX = np.mat(np.zeros((N,numFeatures)))

tY = []
ctr = 0

for i in range(0,len(speech)):
    for j in range(0,len(speech[i])):
    for k in range(0,numFeatures):
        tX[ctr,k] = speech[i][j][k]
    ctr += 1
    tY.append(i)

return tX,tY
```

```
In [2]:
        # Data comes from the LibriSpeech ASR corpus
        # http://www.openslr.org/12/
        # File: train-clean-100.tar.gz [6.3G](training set of 100 hours "cl
        ean" speech)
        # I have converted flac files to wav files.
        # 5 male speakers and 5 female speakers.
        # Female speakers: 1,3,5,7,9
        # Male speakers: 2,4,6,8,10
        #
        trainingFoldersList = [
                                    "samples/train/speaker1/",
                                     "samples/train/speaker2/"
                                     "samples/train/speaker3/",
                                     "samples/train/speaker4/",
                                     "samples/train/speaker5/",
                                    "samples/train/speaker6/",
                                     "samples/train/speaker7/",
                                     "samples/train/speaker8/"
                                     "samples/train/speaker9/",
                                     "samples/train/speaker10/"
                                 1
        tstFoldersList = [
                              "samples/test/speaker1/",
                              "samples/test/speaker2/'
                              "samples/test/speaker3/"
                              "samples/test/speaker4/",
                              "samples/test/speaker5/",
                              "samples/test/speaker6/",
                              "samples/test/speaker7/"
                              "samples/test/speaker8/",
                              "samples/test/speaker9/",
                              "samples/test/speaker10/"
        numSpeakers = 10
        speech = getSamples(trainingFoldersList)
```

```
In [3]: N1 = 0; numFeatures = lim2-lim1
        for idx in range(0,len(speech)):
          N1 += len(speech[idx])
        trainX, trainY = XY(speech, numFeatures, N1)
        minArray = []; maxArray = []
        for i in range(0,numFeatures):
          minArray.append(np.min(trainX[:,i]))
          maxArray.append(np.max(trainX[:,i]))
        f = open("min max values.dat", "w")
        for i in range(0,numFeatures):
          f.write(str(maxArray[i]) + " " + str(minArray[i]) + "\n")
        f.close()
        # Put training data in range (0,1)
        for i in range(0,N1):
          for j in range(0, numFeatures):
            trainX[i,j] = (trainX[i,j]-minArray[j])/(maxArray[j]-minArray[j
        ])
        speech = getSamples(tstFoldersList)
        N2 = 0;
        for idx in range(0,len(speech)):
          N2 += len(speech[idx])
        tstX,tstY = XY(speech,numFeatures,N2)
        for i in range(0,N2):
          for j in range(0, numFeatures):
            tstX[i,j] = (tstX[i,j]-minArray[j])/(maxArray[j]-minArray[j])
        trainLabels = np.mat(np.zeros((len(trainY),numSpeakers)))
        tstLabels = np.mat(np.zeros((len(tstY),numSpeakers)))
        for i in range(0,len(trainY)):
          for j in range(0,numSpeakers):
            if trainY[i] == j: trainLabels[i,j] = 1
            else: trainLabels[i,j] = 0
        for i in range(0,len(tstY)):
          for j in range(0,numSpeakers):
            if tstY[i] == j: tstLabels[i,j] = 1
            else: tstLabels[i,j] = 0
```

```
In [4]: # Test the classifier.
  def tstClassifier(nn,tstX,tstY,ss):

    # Index corresponding to start of speaker 'n'
    limits = [0]
    for lp in range(1,numSpeakers):
        for i in range(0,len(tstY)):
            if tstY[i] == lp: break
```

```
limits.append(i)
  limits.append(len(tstY))
 # Confusion matrix: Speaker 'm' predicted as speaker 'n'.
 confusion = np.mat(np.zeros((10,10)))
 numS = \{\}
 answers = []
 yes = 0; total = 0
 step = ss
  for speaker in range(0, numSpeakers):
    for lp in range(limits[speaker], limits[speaker+1]-step, step):
      pred = []
      for x in range(lp,lp+step):
        A = [tstX[x,v]  for v  in range(0,numFeatures)]
        # Make a prediction.
        res = list(nn.activate(A))
        pred.append(res.index(np.max(res)))
      # There's a probability for every speaker. Let's pick the lar
gest.
      ctr = {}
      for v in pred:
        if v not in ctr: ctr[v] = 1
        else: ctr[v] += 1
      kys = ctr.keys()
      m = 0; best = 0
      for ky in kys:
        if ctr[ky] > m:
          m = ctr[ky]
          best = ky
      #Overall accuracy.
      if best == tstY[lp]: yes += 1
      # Confusion matrix. Speaker 'i' predicted to be speaker 'j'.
      for r in range(0,10):
        if tstY[lp] == r:
          if r not in numS: numS[r] = 1
          else: numS[r] += 1
        for c in range(0,10):
          if tstY[lp] == r and best == c:
            confusion[r,c] += 1
      total += 1
    answers.append(yes*1./total)
 return answers, confusion, numS
```

```
In [5]: #
# Now, we train the neural network.
#
numClasses = len(speech)
```

```
numHiddenNodes = 10  # Nodes per hidden layer.
numHiddenLayers = 5
                      # 5 layers.
possibleOutputs = 1
                      # Only 1 kind of output, i.e. each sample i
s one speaker.
numTrainingEpochs = 10 # Train 10 epochs.
trndata = ClassificationDataSet(numFeatures, possibleOutputs, nb cl
asses=numClasses)
ctr = 0
for row in range(0,N1):
  tempListOut = []; tempListIn = []
  tempListOut.append(int(trainY[row]))
  for i in range(0, numFeatures):
    tempListIn.append(trainX[row,i])
  trndata.addSample(tempListIn,tempListOut)
trndata._convertToOneOfMany()
#first = True
first = False
outfolder = "networks/"
nnFile = "nn49.xml" # Most recent.
st = 50
# For the first time, we need to create the neural network.
# After that, we just need to train.
if first:
  nn = FeedForwardNetwork()
  inputLayer = LinearLayer(numFeatures)
  nn.addInputModule(inputLayer)
 hiddenLayers = []
  for x in range(numHiddenLayers):
    hiddenLayer = SigmoidLayer(numHiddenNodes)
    nn.addModule(hiddenLayer)
    hiddenLayers.append(hiddenLayer)
  outputLayer = SoftmaxLayer(len(speech))
  nn.addOutputModule(outputLayer)
  inputConnection = FullConnection(inputLayer, hiddenLayers[0])
  outputConnection = FullConnection(hiddenLayers[numHiddenLayers-1]
,outputLayer)
  for x in range(numHiddenLayers-1):
    connect = FullConnection(hiddenLayers[x], hiddenLayers[x-1])
    nn.addConnection(connect)
  nn.addConnection(inputConnection)
  nn.addConnection(outputConnection)
  nn.sortModules()
else: # We already have a network!
 nn = NetworkReader.readFrom(outfolder+nnFile)
trainer = BackpropTrainer(nn, dataset=trndata, momentum=0., verbose
```

```
=True, weightdecay=0.)
for i in range(numTrainingEpochs):
    trainer.trainOnDataset(dataset=trndata)
    if (i+1)%5 == 0:
        A,c,n = tstClassifier(nn,tstX,tstY,1)
        print np.mean(A)
        # Save the network, to save time.
        NetworkWriter.writeToFile(nn, outfolder+"nn" + str(st)+".xml")
        st += 1
        print i+1
```

```
Total error: 0.0143656603969
Total error: 0.0143977004382
Total error: 0.0143211886429
Total error: 0.0143559193331
Total error: 0.0142860859515
0.722311963042
5
Total error: 0.0143589772517
Total error: 0.014311991756
Total error: 0.0143279073167
Total error: 0.0143395705399
Total error: 0.0143196161645
0.756046610387
10
```

```
In [6]: # Performance on the test set, as a function of training epoch.
    # We stop when we see signs of overfitting.
    # On 0.2 seconds of data.
    X = []; Y = []
    for q in range(0,50):
        net = outfolder+"nn"+str(q)+".xml"
        nn = NetworkReader.readFrom(net)
        for st in range(1,2):
        A,c,n = tstClassifier(nn,tstX,tstY,st)
        print net,st*0.2,np.mean(A)
        X.append((q+1)*5)
        Y.append(np.mean(A))
```

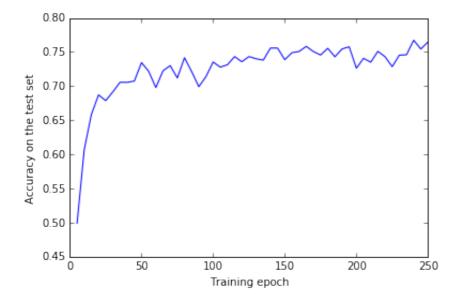
networks/nn0.xml 0.2 0.499117365279 networks/nn1.xml 0.2 0.606735008497 networks/nn2.xml 0.2 0.658612074048 networks/nn3.xml 0.2 0.687217270725 networks/nn4.xml 0.2 0.678839085583 networks/nn5.xml 0.2 0.691663503931 networks/nn6.xml 0.2 0.70569523695 networks/nn7.xml 0.2 0.705575432812 networks/nn8.xml 0.2 0.707493912766 networks/nn9.xml 0.2 0.734532046763 networks/nn10.xml 0.2 0.721658134573 networks/nn11.xml 0.2 0.69792158567 networks/nn12.xml 0.2 0.722454719856 networks/nn13.xml 0.2 0.730107163484 networks/nn14.xml 0.2 0.71199923055 networks/nn15.xml 0.2 0.741639031398 networks/nn16.xml 0.2 0.721392526894 networks/nn17.xml 0.2 0.699211075745 networks/nn18.xml 0.2 0.714028139647 networks/nn19.xml 0.2 0.735380289029 networks/nn20.xml 0.2 0.727852097769 networks/nn21.xml 0.2 0.731408256126 networks/nn22.xml 0.2 0.74333246088 networks/nn23.xml 0.2 0.735741602678 networks/nn24.xml 0.2 0.74336675822 networks/nn25.xml 0.2 0.740176638619 networks/nn26.xml 0.2 0.738092425488 networks/nn27.xml 0.2 0.755807847721 networks/nn28.xml 0.2 0.755728904487 networks/nn29.xml 0.2 0.738764799336 networks/nn30.xml 0.2 0.74898643378 networks/nn31.xml 0.2 0.750807955146 networks/nn32.xml 0.2 0.758286110826 networks/nn33.xml 0.2 0.750738614695 networks/nn34.xml 0.2 0.745580158554 networks/nn35.xml 0.2 0.755489276652 networks/nn36.xml 0.2 0.743030160332 networks/nn37.xml 0.2 0.75452486153 networks/nn38.xml 0.2 0.757615434961 networks/nn39.xml 0.2 0.726292842459 networks/nn40.xml 0.2 0.740774548985 networks/nn41.xml 0.2 0.735083574425 networks/nn42.xml 0.2 0.751037253371 networks/nn43.xml 0.2 0.743174669379 networks/nn44.xml 0.2 0.728544240451 networks/nn45.xml 0.2 0.745473067619 networks/nn46.xml 0.2 0.746040045286 networks/nn47.xml 0.2 0.767353797432 networks/nn48.xml 0.2 0.754406474536 networks/nn49.xml 0.2 0.764805305461

In [7]: # Plot performance with training epoch, on 0.2 seconds of data.
%matplotlib inline
import matplotlib.pyplot as plt

plt.plot(X,Y)
plt.xlabel("Training epoch")
plt.ylabel("Accuracy on the test set")
plt.show()

/usr/local/lib/python2.7/site-packages/matplotlib/font\_manager.py: 273: UserWarning: Matplotlib is building the font cache using fc-l ist. This may take a moment.

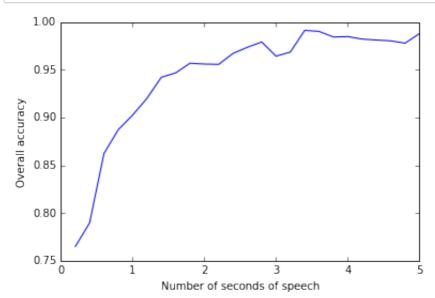
warnings.warn('Matplotlib is building the font cache using fc-li
st. This may take a moment.')



```
In [8]: # Performance on the test set, with 't' seconds of data,
    # Accuracy for t = 0.2,0.4,0.6,0.8,1.0,... seconds.
net = "nn49.xml"
nn = NetworkReader.readFrom(outfolder+net)
X = []; Y = []
for st in range(1,26):
    A,c,n = tstClassifier(nn,tstX,tstY,st)
    print st*0.2,np.mean(A)
    X.append(st*0.2)
    Y.append(np.mean(A))
```

0.2 0.764805305461 0.4 0.789697890647 0.6 0.862283498213 0.8 0.887237110684 1.0 0.902481170611 1.2 0.920037707295 1.4 0.942073795735 1.6 0.946633467992 1.8 0.956715834479 2.0 0.955903690434 2.2 0.955550675976 2.4 0.967076487337 2.6 0.973441010539 2.8 0.979029237841 3.0 0.964161177113 3.2 0.968436822912 3.4 0.991153253831 3.6 0.990071200278 3.8 0.984173735962 4.0 0.984659795058 4.2 0.982023390668 4.4 0.981033772819 4.6 0.980199788274 4.8 0.977751442503 5.0 0.987732329963

```
In [9]: plt.plot(X,Y)
    plt.xlabel("Number of seconds of speech")
    plt.ylabel("Overall accuracy")
    plt.show()
```



```
In [10]: net = "nn49.xml"
    nn = NetworkReader.readFrom(outfolder+net)
    for st in [1,5,10]:
        A,c,n = tstClassifier(nn,tstX,tstY,st)
        # Columns are predictions.
        # First row: Speaker 0: Number of samples predicted as Speaker 0, 1,2,...
        print "With " + str(st*0.2) + " second(s) of speech"
        print c
        print "\n"[:-1]
```

```
With 0.2 second(s) of speech
               2.
                                                                          4.
] ]
   515.
                        3.
                                0.
                                        1.
                                                0.
                                                         0.
                                                                 2.
                                                                                  0.]
       2.
            332.
                      17.
                              66.
                                       35.
                                               18.
                                                         5.
                                                                60.
                                                                          1.
                                                                                22.]
       2.
              23.
                     312.
                                       54.
                                                                        43.
                                6.
                                                 4.
                                                        61.
                                                                 4.
                                                                                  2.1
 [
              35.
       2.
                             374.
                                       33.
                                                 4.
                                                         3.
                                                                 9.
                                                                          1.
                       7.
                                                                                24.]
 [
               9.
                                9.
 [
       6.
                      26.
                                     486.
                                                 1.
                                                       34.
                                                                13.
                                                                        19.
                                                                                  1.]
       0.
              26.
                       3.
                                1.
                                        1.
                                              365.
                                                         4.
                                                                17.
                                                                          0.
 [
                                                                                29.]
       0.
              14.
                      50.
                                0.
                                       59.
                                                0.
                                                      378.
                                                                 2.
                                                                       100.
                                                                                  7.]
              24.
                        4.
                              78.
                                       23.
                                               20.
                                                               300.
       1.
                                                         0.
                                                                          0.
                                                                                48.]
 [
     16.
               3.
                      32.
                                1.
                                       47.
                                                0.
                                                         7.
                                                                 5.
                                                                       489.
 [
                                                                                  1.]
              37.
                      11.
                              29.
                                       15.
                                               18.
                                                        32.
                                                                14.
 [
       0.
                                                                          8.
                                                                               384.]]
With 1.0 second(s) of speech
    105.
               0.
                        0.
                                0.
                                        0.
                                                0.
                                                         0.
                                                                 0.
                                                                          0.
                                                                                  0.]
[[
       0.
              92.
                       0.
                              10.
                                        2.
                                                2.
                                                         0.
                                                                 3.
                                                                          1.
                                                                                  1.]
 [
                                                                         5.
 [
       0.
               4.
                      81.
                                0.
                                        8.
                                                1.
                                                         3.
                                                                 0.
                                                                                  0.]
       0.
               1.
                        0.
                              95.
                                        1.
                                                0.
                                                         0.
                                                                 0.
                                                                          0.
                                                                                  1.]
 [
                                0.
               0.
                        2.
                                     113.
                                                0.
                                                         2.
                                                                          3.
                                                                                  0.]
 [
               1.
                        1.
                                0.
                                        0.
                                               80.
                                                         0.
                                                                 0.
                                                                          0.
                                                                                  7.]
               0.
                                0.
       0.
                        4.
                                        8.
                                                0.
                                                       93.
                                                                 0.
                                                                        17.
                                                                                  0.]
 0.
               2.
                              10.
                                                         0.
                                                                78.
                                                                          0.
                       0.
                                        1.
                                                1.
                                                                                  7.]
 [
       0.
               0.
                        1.
                                0.
                                        1.
                                                0.
                                                         0.
                                                                 0.
                                                                       118.
                                                                                  0.]
 [
                                                                 0.
 [
       0.
               2.
                        0.
                                0.
                                        2.
                                                 1.
                                                         1.
                                                                          0.
                                                                               103.]]
With 2.0 second(s) of speech
                                                             0.
                                               0.
                                                                    0.]
    52.
            0.
                   0.
                          0.
                                 0.
                                        0.
                                                      0.
                                                      0.
                                                             0.
     0.
           49.
                   0.
                          5.
                                 0.
                                        1.
                                               0.
                                                                    0.]
                  46.
                                                             0.
     0.
            0.
                          0.
                                 3.
                                        0.
                                               2.
                                                      0.
                                                                    0.1
 [
                         49.
     0.
            0.
                   0.
                                 0.
                                        0.
                                               0.
                                                      0.
                                                             0.
                                                                    0.]
 [
                                                                    0.]
 [
            0.
                   0.
                          0.
                                59.
                                        0.
                                               0.
                                                      0.
                                                             1.
 [
     0.
            0.
                   0.
                          0.
                                 0.
                                       42.
                                               0.
                                                      0.
                                                             0.
                                                                    2.]
     0.
            0.
                          0.
                                 0.
                                        0.
                                              58.
                                                      0.
                                                             3.
 [
                   0.
                                                                    0.1
            0.
                          2.
                                        1.
                                               0.
                                                     46.
 [
     0.
                   0.
                                 0.
                                                             0.
                                                                    0.]
            0.
                          0.
                                        0.
                                               0.
                                                      0.
     0.
                   0.
                                 0.
                                                            60.
                                                                    0.]
 [
                                                                   54.]]
```

0.

[

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