MACHINE LEARNING FROM DATA

Lab Session 2 – Feature selection using PCA and MDA

1	Goa	al	
		tructions	
		oduction and previous study	
		Phoneme dataset	
		Characteristics of the dataset	
	4.2.	Classification using all the features or a manually selected subset	4
		iture selection on the Phoneme dataset	
		Dimensionality reduction using PCA and MDA	

1. Goal

The goal of this session is to

- Use and compare two methods for dimensionality reduction: multiple discriminant analysis (MDA or Fisher discriminant analysis) and principal component analysis (PCA)
- Test the two methods using synthetic and real datasets

2. Instructions

- Download and uncompress the file Mlearn_Lab2.zip
- Answer the questions in the document Mlearn Lab2 report surnames.doc
- Write the new code in a Colab Notebook Mlearn_lab2_4_Phoneme_PCA_MDA_surnames.ipynb.

3. Introduction and previous study

Read the slides corresponding to lecture 3: Feature selection by dimensionality reduction.

Read and run the code in the Colab Notebook **Mlearn_lab2_1_IntroPCA.ipynb** to understand the use of PCA for dimensionality reduction.

4. Feature selection on a synthetic Gaussian dataset

Now we will use Colab Notebook Mlearn_lab2_2_Synthetic_PCA_MDA.ipynb

This script generates training and test sets with 3-dimensional vectors (d=3) corresponding to three classes (c=3). The classes follow Gaussian distributions, with mean and covariance matrices specified in the code.

A linear and a quadratic classifier are used to classify the data, and error probabilities are computed. Next, the classification is repeated using two features and then using only one feature. The goal is to compare the performance of the classifiers as we reduce the number of features using PCA or MDA.

Run the code.

Note that different values of the seed parameter will produce different eigenvectors for the class covariance matrices.

Use SNR = 10dB.

Q1: Complete the table with the training and test errors for the linear (LC) and the quadratic (QC) classifiers when using three, two and one feature, and SNR=10dB. In this case PCA is used for feature selection. Discuss the results. Analyse the scatter plots in two dimensions and in one dimension.

Repeat the analysis for MDA, using the same seed to generate exactly the same dataset. Q2: Complete the table with the training and test errors for the linear (LC) and the quadratic (QC) classifiers when using three, two and one feature, and SNR=10dB. In this case MDA is used for feature selection. Discuss the results. Analyse the scatter plots in two dimensions and in one dimension.

Repeat the previous analysis using a different SNR value, SNR= 0 dB:

Q3: Use PCA for feature selection. Complete the table with the training and test errors for the linear (LC) and the quadratic (QC) classifiers when using three, two and one feature, and SNR= 0 dB. Discuss the results. Analyse the scatter plots in two dimensions and in one dimension.

Q4: Use MDA for feature selection. Complete the table with the training and test errors for the linear (LC) and the quadratic (QC) classifiers when using three, two and one feature, and SNR= 0 dB. Discuss the results. Analyse the scatter plots in two dimensions and in one dimension.

5. The Phoneme dataset

5.1. Characteristics of the dataset

The Phoneme dataset can be found here: https://web.stanford.edu/~hastie/ElemStatLearn/data.html
This is the information provided by the authors:

These data arose from a collaboration between Andreas Buja, Werner Stuetzle and Martin Maechler, and we used as an illustration in the paper on Penalized Discriminant Analysis by Hastie, Buja and Tibshirani (1995), referenced in the text.

The data were extracted from the TIMIT database (TIMIT Acoustic-Phonetic Continuous Speech Corpus, NTIS, US Dept of Commerce) which is a widely used resource for research in speech recognition. A dataset was formed by selecting five phonemes for classification based on digitized speech from this database. The phonemes are transcribed as follows: "sh" as in "she", "dcl" as in "dark", "iy" as the vowel in "she", "aa" as the vowel in "dark", and "ao" as the first vowel in "water". From continuous speech of 50 male speakers, 4509 speech frames of 32 msec duration were selected, approximately 2 examples of each phoneme from each speaker. Each speech frame is represented by 512 samples at a 16kHz sampling rate, and each frame represents one of the above five phonemes. The breakdown of the 4509 speech frames into phoneme frequencies is as follows:

```
aa ao dcl iy sh
695 1022 757 1163 872
```

From each speech frame, we computed a log-periodogram, which is one of several widely used methods for casting speech data in a form suitable for speech recognition. Thus, the data used in what follows consist of 4509 log-periodograms of length 256, with known class (phoneme) memberships.

The data contain 256 columns labelled "x.1" - "x.256", a response column labelled "g", and a column labelled "speaker" identifying the different speakers.

Open Colab Notebook Mlearn_lab2_3_Phoneme.ipynb

Read the notebook and identify code sections for:

- Reading the dataset and selecting a subset of 64 features for each observation (from the original vector of size 256)
- Removing the sample mean for each observation
- Plotting the vectors (using all the 256 features, up to 8kHz)
- Dividing the dataset into training (70%) and test (30%) subsets. Note that the partition is random and it keeps the same proportion of training/test observations for each class
- Classifying the data using linear and quadratic classifiers
- Computing error probabilities and confusion matrices
- When using 2 features, showing a scatter plot and decision boundaries between classes.

5.2. Classification using all the features or a manually selected subset

Run the code in Mlearn_lab2_3_Phoneme.ipynb

Answer the following questions:

Q5. Include the plots of the phoneme spectra.

Q6. Include the error probabilities for the training and test sets obtained with the linear classifier (LC) and the quadratic classifier (QC), using all the features. Discuss the results.

Q7. Include the confusion matrices for the test set obtained with the linear classifier (LC) and the quadratic classifier (QC), using all the features. Discuss the results.

Suppose that due to computational issues we can only use two features per observation. Considering the plots in Q1, manually select two features that seem to be the most discriminative.

Use the variable V coor

Q8. Which features would you choose? Show the error probabilities for the training and test sets obtained with the linear and the quadratic classifier. Compare with the previous case (using all features) and discuss the results.

Q9. Include the scatter plot and decision boundaries. Discuss the results.

6. Feature selection on the Phoneme dataset

6.1. Dimensionality reduction using PCA and MDA

Previously we have reduced the dimensionality of the Phoneme dataset by observing the data and manually selecting features. In this section we will reduce dimensionality following a principal component analysis criterion.

Edit a new Colab Notebook named Mlearn_lab2_4_Phoneme_PCA_MDA_surname.ipynb

Write the code to:

Read the complete dataset (256-dimensional vectors). Perform feature selection using PCA, and then apply a linear and a quadratic classifier. Compute the training and test classification error for the two classifiers

Lab 2: Feature selection – PCA and MDA Machine Learning from Data

when using a number of features nfeat ranging from 1 to 256.

Show the four error curves (training/test errors for LC and QC as a function of the dimension nfeat) in the same figure, using different colours for the curves.

Note: use the training dataset to find the projection matrix and to train the classifier. The test dataset should only be used for evaluating the error probabilities.

Q10. Show the error curves for the linear and the quadratic classifier on the training and on the test set.

Q11. Discuss which dimension is the most adequate for the linear classifier and which is the best one for the quadratic classifier. Remember that it is important not to overfit on the training data (the test error should not be much larger than the training error).

Now, repeat the analysis but using MDA instead of PCA for dimensionality reduction. Write the code to use MDA for a number of features d' ranging from 1 to dmax.

Q12. Which is the maximum number of features *dmax*? Show the error curves for the linear and the quadratic classifier on the training and on the test set.

Q13. Compare results and discuss the use of PCA and MDA for the Phoneme dataset using between 1 and dmax features