TECHNISCHE UNIVERSITÄT MÜNCHEN

Fakultät für Elektrotechnik und Informationstechnik Lehrstuhl für Datenverarbeitung PD Dr. Martin Kleinsteuber

Information Retrieval in High Dimensional Data Lab #2, 26.10.2017

Statistical Decision Making

Task 1. Consider the two-dimensional, discrete random variable $X = [X_1 \ X_2]^{\top}$ subjected to the joint probability density p_X as described in the following table.

$$\begin{array}{c|c|c|c|c} p_X(X_1, X_2) & X_2 = 0 & X_2 = 1 \\ \hline X_1 = 0 & 0.4 & 0.3 \\ X_1 = 1 & 0.2 & 0.1 \\ \hline \end{array}$$

- a) Compute the marginal probability densities p_{X1}, p_{X2} and the conditional probability $P(X_2 = 0 | X_1 = 0)$ as well as the expected value $\mathbb{E}[X]$ and the covariance matrix $\mathbb{E}[(X \mathbb{E}[X])(X \mathbb{E}[X])^{\top}]$.
- b) Write a PYTHON function toyrnd that expects the positive integer parameter n as its input and returns a matrix X of size (2,n), containing n samples drawn independently from the distribution p_X , as its output.
- c) Verify your results in a) by generating 10000 samples with toyrnd and computing the respective empirical values¹.
- Task 2. The MNIST training set consists of handwritten digits from 0 to 9, stored as PNG files of size 28 × 28 and indexed by label. Download the provided ZIP file from Moodle and make yourself familiar with the directory structure.
 - a) Grayscale images are typically described as matrices of uint8 values. For numerical calculations, it is more sensible to work with floating point numbers. Load two (abitrary) images from the database and convert them to matrices I1 and I2 of float64 values in the interval [0, 1].
 - b) The matrix equivalent of the euclidean norm $\|\cdot\|_2$ is the *Frobenius* norm. For any matrix $\mathbf{A} \in \mathbb{R}^{m \times n}$, it is defined as

$$\|\mathbf{A}\|_F = \sqrt{\operatorname{tr}(\mathbf{A}^\top \mathbf{A})},\tag{1}$$

Unless stated otherwise, we are working with the *biased* estimator $\frac{1}{n}\sum_{i=1}^{n} \left(\mathbf{x}_{i} - \left(\frac{1}{n}\sum_{j=1}^{n} \mathbf{x}_{j}\right)\right) \left(\mathbf{x}_{i} - \left(\frac{1}{n}\sum_{j=1}^{n} \mathbf{x}_{j}\right)\right)^{\top}$ of the covariance

where tr denotes the trace of a matrix. Compute the distance $\|\mathbf{I}_1 - \mathbf{I}_2\|_F$ between the images I1 and I2 by using three different procedures in PYTHON:

- Running the numpy.linalg.norm function with the 'fro' parameter
- Directly applying formula (1)
- Computing the euclidean norm between the vectorized images
- c) In the following, we want to solve a simple classification problem by applying k-Nearest Neighbours. To this end, choose two digit classes, e.g. 0 and 1, and load n_train = 500 images from each class to the workspace. Convert them according to subtask a) and store them in vectorized form in the matrix X_train of size (784, 2*n_train). Provide an indicator vector Y_train of length 2*n_train that assigns the respective digit class label to each column of X_train.

From each of the two classes, choose another set of $n_{test=10}$ images and create the according matrices X_{test} and Y_{test} . Now, for each sample in the test set, determine the k=20 training samples with the smallest Frobenius distance to it and store their indices in the $(2*n_{test}, k)$ matrix NN. Generate a vector Y_{kNN} containing the respective estimated class labels by performing a majority vote on NN. Compare the result with Y_{test} .

Helpful Numpy functions

Required packages: numpy (np), scipy

```
scipy.misc.imread(path)
                            import image from path as uint8-array
np.array(x)
                            turn x into a numpy array
                            norm of an array x, different norms possible
np.lingalg.norm(x)
                            computes matrix multiplication arrays x and y
np.dot(x, y)
np.sqrt(x)
                            computes square root of x
                            computes matrix trace of x
np.trace(x)
np.sum(x, axis)
                            sums entries of array over axis x
                            returns indices required to sort array x by size
np.argsort(x)
np.zeros(shape)
                            generates array of all zeros of a given shape
np.ones(shape)
                            generates array of all ones of a given shape
np.random.rand(shape)
                            generate array of random numbers
np.reshape(x,shape)
                            reshape array x to a given shape
np.ravel(x, 'F')
                            returns a flattened array, stack columns
np.expand_dims(x, axis)
                            add dimension to array
np.concatenate((x,y))
                            concatenates two arrays
                            vertically stack two arrays
np.vstack((x,y))
```