

# Intelligent Data Analysis

## Tutorial 8

### Kernel ERM

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### Goals

A frequently used kernel for learning from sequential data are sequence kernels. The aim of the exercise is to implement such a sequence kernel for various distance measures and apply this kernel to data.

### Problem setting

The *Dynamic Time Warping Kernel* is defined by:

$$k(\mathbf{x}, \mathbf{y}) = \exp(-\lambda D_{DTW}(\mathbf{x}, \mathbf{y}))$$

, where  $D_{DTW}(\mathbf{x}, \mathbf{y})$  is the DTW-distance between two sequences  $\mathbf{x} = \{x_i\}$  with  $i = 1 \dots t_x$  and  $\mathbf{y} = \{y_j\}$  with  $j = 1 \dots t_y$ . The DTW-distance is recursively defined by:

$$\gamma(i, j) = d(x_i, y_j) + \min(\gamma(i-1, j-1), \gamma(i-1, j), \gamma(i, j-1))$$

There are the following different DTW-distances  $d(x_i, y_j)$ :

- Edit distance: binary distance between the data points  $x_i$  and  $y_j$ , that is  $d(x_i, y_j) = [x_i \neq y_j]$ .
- Euclidean DTW distance: (quadratic) euclidean distance between the data points  $x_i$  and  $y_j$ , that is  $d(x_i, y_j) = (x_i - y_j)^2$ .
- Derivative DTW-distance: (quadratic) euclidean distance between the derivatives at the points  $x_i$  and  $y_j$ , that is  $d(x_i, y_j) = (\Delta(x_i) - \Delta(y_j))^2$ , where the derivation can be calculated approximately by

$$\Delta(x_i) = \frac{1}{2} \left( x_i - x_{i-1} + \frac{x_{i+1} - x_{i-1}}{2} \right).$$

### Task 1

Write the following MATLAB-Function

```
function D = dtw(x,y,dst)
```

which computes the DTW-distance between the sequences  $\mathbf{x}$  and  $\mathbf{y}$ . Both sequences are given by vectors. The parameter `dst` controls, which distance is used: `dst=1` stands for the edit distance, `dst=2` stands for the euclidean distance, and all other parameters of `dst` stands for the derivative distance.

## Task 2

Write the following MATLAB-Function

```
function K = dtw_kernel(X,Y,dst,lambda)
```

which computes the DTW-kernel. The matrix  $\mathbf{X}$  is a  $n_X \times m_X$  matrix with  $n_X$  sequences of the length  $m_X$ . The matrix  $\mathbf{Y}$  is a  $n_Y \times m_Y$  matrix with  $n_Y$  sequences of length  $m_Y$ . The parameter `dst` controls, which distance is used and the parameter  $\lambda$  is the kernel parameter. For the matrix  $\mathbf{K}$  the entry  $K_{ij}$  is defined as  $K_{ij} = \exp(-\lambda D_{DTW}(\mathbf{X}_i, \mathbf{Y}_j))$ , where  $\mathbf{X}_i$  is the  $i$ 's row of matrix  $\mathbf{X}$  and  $\mathbf{Y}_j$  is the  $j$ 's row of matrix  $\mathbf{Y}$ .

## Task 3

Analyze the MATLAB functions `alpha = learnKernelRegERM(X,y,lambda,f_krnl,krnl_p)` and `y = classKernelRegERM(alpha,Xtr,f_krnl,krnl_p,X)`, which learns the parameters for a kernel model and classifies data for a previously learned model, respectively.

## Task 4

Use the data set `laser_small.mat` and divide this data set in a training and test set using the function `split_train_test (split=0.7, seed=3, ...)`. Train a model using the instruction `alpha = learnKernelRegERM(train,y_train,1,@dtw_kernel,0.0005)`. Classify the test data and compute the accuracy for the previously trained model.