# 183.605 Machine Learning for Visual Computing Assignment 2

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# 1 Assignment 2

## 1.1 The dual optimization problem

#### Tasks:

• Generate a suitable training set of linearly separable data

The training set is generated in generateTrainingData.m. The function generateTrainingData(N, xRange, yRange, linear) takes the number of sample points, the domains (defined by a lower and an upper bound) from which the x and y coordinates are sampled and a flag which indicates if the resulting data should be linearly separable. A set of random 2D coordinates is created using the MATLAB function rand. Linear separability is achieved by labelling the according to the condition  $x_i + y_i > \bar{x} + \bar{y}$ , where  $\bar{x}$  and  $\bar{y}$  denote the domain centers.

• Plot the input vectors in  $\mathbb{R}^2$  and visualize corresponding target values (e.g. by using color).

The resulting training data is displayed in Figure 1. Sample points with class label 1 are marked red, points with label -1 are merked green.

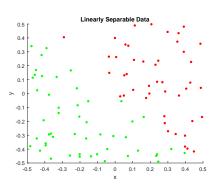


Figure 1: Linearly separable data with color-coded class labels (1=red, -1=green) for N=100.

• Visualize the support vectors and plot the decision boundary.

Figure 2 shows the support vectors defined by trainSVM.

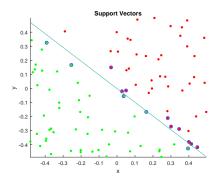


Figure 2: Training data with support vectors marked by blue circles and decision boundary plotted in blue.

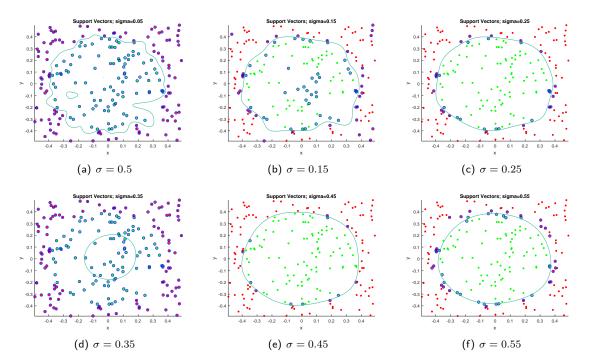


Figure 3: Training data with support vectors marked by blue circles and decision boundary plotted in blue, for different values of the RBF parameter  $\sigma$ .

### 1.2 The kernel trick

#### Tasks:

• Try different values for  $\sigma$  (the RBF parameter).

The radial basis function kernel is defined by  $K(x,y) = exp(-\frac{\|x-y\|^2}{\sigma^2})$ . The corresponding support vectors are shown in Figure 3 for different values of the RBF parameter  $\sigma$ . In the lecture notes about RBF-networks, we discussed a selection of  $\sigma = 2 * avgdist$ , where avgdist denotes the average distance of the centers. Having distances between the data points of approximately 10 units (or slightly more),  $\sigma = 25$  is then selected.

• Generate a non-linearly separable training set, plot the data, visualize the support vectors and plot the decision boundary.

See also Figure ??fig:sv