



## General Information:

Lecture (3 SWS): Mon 08.15 – 09:45 (H16) and Tue 08.15 – 09:45 (H16)  
Exercises (1 SWS): Wed 12.15 – 13.15 (00.151-113) and Thu 12.30 – 13.30 (00.151-113)  
Certificate: Oral exam at the end of the semester  
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## Mean Shift Algorithm

**Exercise 1** The mean shift algorithm can be used to determine a local maximum (or a saddle point) of a probability density function  $p(\mathbf{x})$ . In particular, it is feasible to determine the *mode* of the density. For a mathematical formulation of the mean shift algorithm, the Parzen window approach is used to model  $p(\mathbf{x})$ . The mean shift iterations are equivalent to a gradient ascent for  $p(\mathbf{x})$ .

- Outline the main steps of the mean shift algorithm.
- Derive the mean shift vector for the following kernels:
  - Epanechnikov kernel
  - Gaussian kernel

**Exercise 2** Let  $\mathcal{S} = \{\mathbf{x}_1, \dots, \mathbf{x}_n\}$  be a set of  $N = 8$  samples defined as:

$$\mathcal{S} = \left\{ \begin{pmatrix} 0.1 \\ 0.1 \end{pmatrix}, \begin{pmatrix} 0.1 \\ 0.2 \end{pmatrix}, \begin{pmatrix} 0.2 \\ 0.25 \end{pmatrix}, \begin{pmatrix} 0.3 \\ 0.2 \end{pmatrix}, \begin{pmatrix} 0.5 \\ 0.7 \end{pmatrix}, \begin{pmatrix} 0.7 \\ 0.8 \end{pmatrix}, \begin{pmatrix} 0.8 \\ 0.9 \end{pmatrix}, \begin{pmatrix} 0.9 \\ 0.8 \end{pmatrix} \right\}$$

- Draw the samples in the 2-dimensional feature space.
- Perform one mean shift iteration and draw the corresponding mean shift vectors using the following starting points:
  - $\mathbf{x}^0 = (0 \ 0)^\top$
  - $\mathbf{x}^0 = (1 \ 1)^\top$

Use the Epanechnikov kernel with kernel width  $\lambda = 0.25$ .

- Sketch the mean shift vectors if the mean shift iterations are performed until convergence.
- Explain how the mean shift algorithm can be used for an automatic clustering. How do you determine the number of clusters? Compare mean shift clustering to hard- and soft-clustering.

**Exercise 3** **Matlab exercise:** In terms of image processing, the mean shift algorithm can be employed for edge-preserving smoothing. This filtering technique can be used to denoise images. The key idea of mean shift filtering is to represent each pixel of an image by a feature vector  $\mathbf{x}$  and to define a joint probability density function  $p(\mathbf{x})$  for the image. Mean shift iterations are performed to find a local



Figure 1: Noisy (left) and denoised image (right) using mean shift filtering.

maximum of  $p(\mathbf{x})$  next to a given pixel. For the sake of simplicity, we consider 2-dimensional, intensity (gray value) images. For details of mean shift for edge-preserving smoothing please refer to

Comaniciu, D. and Meer, P. *Mean shift: a robust approach toward feature space analysis*. IEEE Transactions on Pattern Analysis and Machine Intelligence (2002), Volume 24, Issue: 5, pp. 603 - 619

- (a) Define a feature vector  $\mathbf{x}_i$  to model the  $i$ -th pixel for a given input image. Explain how the feature vector can be extended to handle color images represented in the RGB color space.
- (b) Explain how the mean shift algorithm can be employed to denoise  $\mathbf{x}_i$ . In particular, describe which parameters are required and explain the influence of the parameters to the outcome of mean shift.
- (c) Implement the edge-preserving smoothing using the mean shift algorithm. Without loss of generalization, we use the Epanechnikov kernel for the mean shift iterations.
- (d) Test your algorithm using synthetic image data:
  - Load the example *Cameraman* image.
  - Apply your mean shift algorithm to smooth the noisy image.
  - The width of the Epanechnikov kernel can be selected empirically by visual inspection of the denoised image.
  - Compare the input and the denoised image qualitatively.