Variational Methods for Computer Vision

Lecture: M. Möller Exercises: J. Geiping Winter Semester 17/18 Visual Scene Analysis Institute for Computer Science University of Siegen

## Weekly Exercises 4

Room: H-C 6336 Friday, 17.11.2017, 14:15-15:45

Submission deadline: Tuesday, 14.11.2017, 14:15 in Room H-C 6336 Programming: Email your solution to jonas.geiping@uni-siegen.de

## Theory

Exercise 1 (2 Points). Let

$$W = \begin{pmatrix} A & B \\ B^T & B^T A^{-1} B \end{pmatrix}$$

for  $W \in \mathbb{R}^{N \times N}$ ,  $A \in \mathbb{R}^{r \times r}$ , and  $B \in \mathbb{R}^{r \times N - r}$  symmetric matrices. Prove that  $\operatorname{rank}(W) \leq r$  by showing the identity

$$W = \bar{U} \Sigma \bar{U}^T$$
, for  $\bar{U} = \begin{pmatrix} U \\ B^T U \Sigma^{-1} \end{pmatrix} \in \mathbb{R}^{N \times r}$ 

where  $A = U\Sigma U^T$  is the eigendecomposition of A, and deduce from this identity, that the rank is r or less.

It can be shown that this W with  $rank(W) \leq r$  has a (reduced) eigendecomposition  $W = VDV^T$  for some  $V \in \mathbb{R}^{N \times r}$  and  $D \in \mathbb{R}^{r \times r}$  being diagonal <sup>1</sup>, but you can skip this step.

**Exercise 2** (4 points). Let  $W \in \mathbb{R}^{N \times N}$  be a symmetric rank r matrix with eigendecomposition  $W = VDV^T$ ,  $D_{ii} \leq 1 \ \forall i \in \{1, ..., n\}$ . Show that the solution to

$$\hat{u} = \underset{u}{\operatorname{arg\,min}} \frac{1}{2} \|u - f\|^2 + \frac{\alpha}{2} \langle u, (I - W)u \rangle$$

is  $\hat{u} = VXV^T f$ , where  $X \in \mathbb{R}^{r \times r}$  is a diagonal matrix with entries  $X_{ii} = \frac{1}{1 + \alpha - \alpha D_{ii}}$ . Start with the optimality condition and insert what you know about W. How would you implement this formula (pseudocode) without using a full matrix-vector product in dimensions  $n \times n$ .

<sup>&</sup>lt;sup>1</sup>See Fowlkes et al. Spectral Grouping Using the Nyström method, (2004)

## **Understanding Code**

**Exercise 3** (4 points). Download the functions "IntegralNLM.m", "IntegralImage.m", and "ImShift.m" from supplementary material for sheet 4. Answer the following questions:

- What does "ImShift.m" do?
- What does "IntegralImage.m" do?
- State a mathematical formula for the (i, j)-th entry of the variable ssd in IntegralNLM.m.
- Based on ssd, state a formula for PatchDist.

## **Programming**

Exercise 4 (6 points). Load the image cityscape provided in the supplementary material for sheet 4 and create a noisy version by adding Gaussian noise (help randn). Use the downloaded functions from the previous exercise to generate a similarity matrix W of the noisy image.

- $\bullet$  Perform nonlocal means denoising with the similarity matrix W
- Compute a diagonal matrix D with

$$d_i = \sum_j W_{i,j}$$

on the diagonal. Perform nonlocal regularization with

$$R(u) = \frac{\alpha}{2} \langle u, Lu \rangle$$

for L = D - W and for  $L = I - D^{-1/2}WD^{-1/2}$ . Tune the values of  $\alpha$ .

• Check the *PSNR* <sup>2</sup>, (psnr(image,reference\_image) in Matlab) of your output versus the reference image (in our case the original image). Compare this value to the PSNR value for your result from the previous exercise (Huber-TV denoising).

Which denoising algorithm yields the best results?

<sup>&</sup>lt;sup>2</sup>PSNR, the peak signal to noise ratio measures the distance of two images in a logarithmic scale, if the images are very similar, then their PSNR will be high.