



Classroom Work Assignment C5

Problem C5.1 (Convex Functionals and Forward Diffusion)

In Lecture 10 we have learned that minimising the 1-D energy functional

$$E(u) = \int_a^b \left((u - f)^2 + \alpha \Psi(u_x^2) \right) dx$$

can be seen as an implicit time discretisation of the diffusion process $\partial_t u = \partial_x (\Psi'(u_x^2) u_x)$.

Assume that the regulariser Ψ is twice differentiable and convex in s , which implies that $\frac{d^2}{ds^2} \Psi(s^2) \geq 0$.

Show that the corresponding diffusion process always performs forward diffusion. Thus, it cannot act contrast-enhancing.

Homework Assignment H5

Problem H5.1 (Parameter Adaptation Under Rescaling) 3+1P

Assume you have a C^1 image on a quadratic domain Ω of size 128×128 , and its grey values are in the interval $[0, 255]$. The spatial grid size is $h_1 = h_2 = 1$. Filtering it with an explicit scheme for EED with parameters $\lambda = 4, \sigma = 1, T = 248$ and $\tau = 0.2$ takes 12.4 seconds on your computer.

- (a) How would you adapt the parameters σ, λ , and T , when your image is rescaled to 256×256 pixels, and its grey value range to $[0, 1]$? The spatial grid size remains 1.
- (b) Estimate the required times for computing the result with the explicit scheme.

Problem H5.2 (Continuous Variational Regularisation) 5+1+3P

Consider the nonconvex 2-D energy functional

$$E(u) = \frac{1}{2} \int_{\Omega} \left((u - f)^2 - 2\alpha\lambda^2 \exp\left(-\frac{|\nabla u|^2}{2\lambda^2}\right) \right) d\mathbf{x}.$$

- (a) State the Euler-Lagrange equation as well as the corresponding boundary condition of the above functional. Which diffusion filter is approximated by the energy functional?
- (b) What can you say about the uniqueness of the minimiser?
- (c) Determine the diffusivity and show that λ is the boundary between forward and backward diffusion.

Problem H5.3 (Energy Functional for Nonlinear Diffusion) 3P

Consider the 2-D nonlinear isotropic diffusion process with rational Perona-Malik diffusivity

$$\partial_t u = \operatorname{div} \left(\frac{1}{1 + |\nabla u|^2 / \lambda^2} \nabla u \right).$$

Compute the energy functional that corresponds to this diffusion process.

Problem H5.4 (Decorrelation Criterion)

1+2+5P

The file `charb_decorr.c` is similar to `iso_non_diff.c` from Assignment H3 and implements isotropic nonlinear diffusion filtering, but with the Charbonnier diffusivity (instead of a Perona-Malik diffusivity) and without Gaussian smoothing. However, this time we want to incorporate and test the decorrelation criterion that frees us from the need to specify the number of iterations.

- (a) Find arguments that support the following claim: For small $\lambda > 0$, the Charbonnier diffusion

$$\partial_t u = \operatorname{div} \left(\frac{1}{\sqrt{1 + |\nabla u|^2 / \lambda^2}} \nabla u \right)$$

at time t approximates TV flow at time λt .

- (b) Supplement the file `charb_decorr.c` with the missing code such that it incorporates the decorrelation criterion from Lecture 9. Compile it with `gcc -Wall -O2 -o charb_decorr charb_decorr.c -lm`
- (c) The images `acros.pgm` and `peppers-noise.pgm` have been degraded by additive Gaussian noise with mean 0. Denoise them with `charb_decorr` with time step size $\tau = 0.2$. Use the correlation coefficient to estimate also the best contrast parameter λ . What are the corresponding λ values? Why is this heuristics justified?

Submission: Please create a directory `Ex05_<your_name>` with the following files (and nothing else):

- a pdf file – which can also be a scanned handwritten solution – that contains
 - the names of all people working together for this assignment
 - the solutions of the theoretical Problems 1–3 and answers to the questions in Problem 4
- for Problem 4: the completed file `charb_decorr.c` and the two resulting images with optimised λ

Compress the directory to a zip file `Ex05_<your_name>.zip`.

Submit the file via CMS.

Deadline for submission is Friday, December 1, 14:00.