



Classroom Work Assignment C6

**Problem C6.1 (Gradient Domain Methods)**

Some applications such as seamless image cloning benefit from manipulating the gradient of an image rather than its grey values: Discontinuities in the gradient data appear as continuous transitions in the image data. Since the manipulated vector field  $\mathbf{p}$  is not necessarily a gradient vector field anymore, it cannot be integrated exactly.

As a remedy, one searches for a function  $u$  whose gradient approximates  $\mathbf{p}$ . It is found by minimising the energy

$$E(u) = \int_{\Omega} |\nabla u - \mathbf{p}|^2 d\mathbf{x}.$$

- Give an example of a non-integrable vector field i.e. a vector field that is not the gradient of some function.
- Write down the Euler-Lagrange equation of the above energy functional. Also state the corresponding boundary condition.

**Problem C6.2 (Forward-Backward Splitting)**

In the forward backward splitting (FBS) method, the primal variable  $\mathbf{u}$  and the dual variable  $\mathbf{b}$  are obtained by the following iteration:

$$\begin{aligned} \mathbf{b}^{k+1} &= \operatorname{argmax}_{\mathbf{b} \in \mathbb{R}^{2N}} \left\{ -\iota_{C_{\alpha}}(\mathbf{b}) + \langle \mathbf{b}, \mathcal{D}\mathbf{u}^k \rangle - \frac{1}{2\tau} \|\mathbf{b} - \mathbf{b}^k\|_2^2 \right\}, \\ \mathbf{u}^{k+1} &= \operatorname{argmin}_{\mathbf{u} \in \mathbb{R}^N} \left\{ \frac{1}{2} \|\mathbf{u} - \mathbf{f}\|_2^2 + \langle \mathcal{D}^{\top} \mathbf{b}^{k+1}, \mathbf{u} \rangle \right\}. \end{aligned} \quad (1)$$

Show that FBS can be written as

$$\mathbf{b}^{k+1} = \operatorname{argmin}_{\mathbf{b} \in \mathbb{R}^{2N}} \left\{ \iota_{C_{\alpha}}(\mathbf{b}) + \frac{1}{2} \|\mathbf{b} - (\mathbf{b}^k - \tau(\mathcal{D}(\mathcal{D}^{\top} \mathbf{b}^k - \mathbf{f})))\|_2^2 \right\}.$$

You can use the fact that  $\mathbf{u}^{k+1} = \mathbf{f} - \mathcal{D}^{\top} \mathbf{b}^{k+1}$  minimises Equation (1) without proof.

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## Homework Assignment H6

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### Problem H6.1 (Stability of Diffusion-Reaction Discretisations)

2+2+1+3P

Consider the diffusion-reaction equation

$$\frac{\partial u}{\partial t} = \mathbf{div} \left( g(|\nabla u|^2) \nabla u \right) - \frac{u - f}{\alpha} \quad \text{with } \alpha > 0.$$

In this assignment, you can examine stability criteria for a discretisation with the *modified explicit scheme*

$$\frac{\mathbf{u}^{k+1} - \mathbf{u}^k}{\tau} = \mathbf{A}^k(\mathbf{u}^k) \mathbf{u}^k - \frac{1}{\alpha}(\mathbf{u}^{k+1} - \mathbf{f}) \quad (2)$$

and the *fully explicit scheme*

$$\frac{\mathbf{u}^{k+1} - \mathbf{u}^k}{\tau} = \mathbf{A}^k(\mathbf{u}^k) \mathbf{u}^k - \frac{1}{\alpha}(\mathbf{u}^k - \mathbf{f}). \quad (3)$$

- (a) Show that the solution  $\mathbf{u}^{k+1}$  of the modified explicit scheme (2) can be computed as

$$\mathbf{u}^{k+1} = \frac{\alpha \mathbf{v}^{k+1} + \tau \mathbf{f}}{\alpha + \tau}$$

where  $\mathbf{v}^{k+1}$  denotes the solution of the explicit diffusion scheme without reaction term:

$$\frac{\mathbf{v}^{k+1} - \mathbf{u}^k}{\tau} = \mathbf{A}^k(\mathbf{u}^k) \mathbf{u}^k.$$

- (b) Show that the result from (a) implies stability of (2) in terms of the discrete maximum-minimum principle

$$\min_j f_j \leq u_i^{k+1} \leq \max_j f_j$$

for all  $i$  and all  $k \geq 0$  if we use the initialisation  $\mathbf{u}^0 := \mathbf{f}$  and if the explicit scheme without reaction term satisfies

$$\min_j u_j^k \leq v_i^{k+1} \leq \max_j u_j^k$$

for all  $i$  and all  $k \geq 0$ .

- (c) Determine a stability criterion for  $\tau$  for the modified explicit scheme (2), if  $h_1 = h_2 = 1$ ,  $0 < g(s^2) \leq 5$ , and  $\alpha = 10$ .
- (d) For the same parameters as stated in Part (c), determine a stability criterion for  $\tau$  for the fully explicit scheme (3).  
*Hint:* Use Parts (a)–(c) as a guideline.

**Problem H6.2 (Primal-Dual Hybrid Gradient Algorithm)**

6P

In the primal-dual hybrid gradient (PDHG) algorithm, the primal variable  $\mathbf{u}$  and the dual variable  $\mathbf{b}$  are obtained by the following iteration:

$$\begin{aligned}\mathbf{b}^{k+1} &= \operatorname{argmax}_{\mathbf{b} \in \mathbb{R}^{2N}} \left\{ -\iota_{C_\alpha}(\mathbf{b}) + \langle \mathbf{b}, \mathcal{D}\mathbf{u}^k \rangle - \frac{1}{2\tau} \|\mathbf{b} - \mathbf{b}^k\|_2^2 \right\}, \\ \mathbf{u}^{k+1} &= \operatorname{argmin}_{\mathbf{u} \in \mathbb{R}^N} \left\{ \frac{1}{2} \|\mathbf{u} - \mathbf{f}\|_2^2 + \langle \mathcal{D}^\top \mathbf{b}^{k+1}, \mathbf{u} \rangle + \frac{1}{2\sigma} \|\mathbf{u} - \mathbf{u}^k\|_2^2 \right\}.\end{aligned}\quad (4)$$

Show that PDHG can be written as

$$\begin{aligned}\mathbf{b}^{k+1} &= P_{C_\alpha}(\mathbf{b}^k + \tau \mathcal{D}\mathbf{u}^k), \\ \mathbf{u}^{k+1} &= \frac{1}{1 + \frac{1}{\sigma}} \left( \mathbf{f} - \mathcal{D}^\top \mathbf{b}^{k+1} + \frac{1}{\sigma} \mathbf{u}^k \right).\end{aligned}$$

In particular, show that  $\mathbf{u}^{k+1}$  is a unique minimiser of Equation (4).

**Problem H6.3 (Primal-Dual Methods for TV Regularisation)**

1+1+5+3P

- (a) The program `tv-kacanov.c` implements the differentiable  $\varepsilon$ -approximation of TV regularisation by means of the Kačanov method (cf. Lecture 11). The Gauss-Seidel algorithm serves as simple (and fairly slow) iterative solver for the linear systems of equations.

Compile it with `gcc -Wall -O2 -o tv-kacanov tv-kacanov.c -lm` and use it to denoise the image `pruebab1.pgm` with regularisation parameter  $\alpha = 140$  and approximation parameter  $\varepsilon = 0.01$ . Use 50 outer fixed point iterations and 30000 inner Gauss-Seidel iterations to obtain the filtered image `pruebab1-kacanov.pgm`.

- (b) The program `tv-fbs.c` is a forward-backward splitting algorithm for TV regularisation that does not require any  $\varepsilon$ -approximation. Compile it with `gcc -Wall -O2 -o tv-fbs tv-fbs.c -lm` and use it for denoising `pruebab1.pgm` with  $\alpha = 140$ ,  $\tau = 0.2$ , and 100000 iterations. This creates your reference solution `pruebab1-ref.pgm`.

(However, please keep in mind that this cannot be a perfect reference solution, since our discrete models for the primal-dual methods have a directional bias.)

- (c) Supplement the missing code in `tv-fista.c` such that you obtain an implementation of the FISTA algorithm for TV regularisation. Compile it with `gcc -Wall -O2 -o tv-fista tv-fista.c -lm`

- (d) Run both `tv-fbs` and `tv-fista` on `pruebab1.pgm` with  $\alpha = 140$  and  $\tau = 0.2$  (for FBS), but use only 100 iterations. This gives the images `pruebab1-fbs100.pgm` and `pruebab1-fista100.pgm`. Check their accuracy and the accuracy of `pruebab1-kacanov.pgm` w.r.t. the reference solution `pruebab1-ref.pgm` by means of the program `difference.c`. This creates the error images `error-fbs100.pgm`, `error-fista100.pgm`, and `error-kacanov.pgm`. What are your conclusions?

**Submission:** Please create a directory `Ex06_<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 1–2 and answers to the questions in Problem 3
- for Problem 3: the completed file `tv-fista.c` and the seven images `pruebab1-kacanov.pgm`, `pruebab1-ref.pgm`, `pruebab1-fbs100.pgm`, `pruebab1-fista100.pgm`, `error-fbs100.pgm`, `error-fista100.pgm`, and `error-kacanov.pgm`

Compress the directory to a zip file `Ex06_<your_name>.zip`.

Submit the file via CMS.

**Deadline for submission is Friday, December 8, 14:00.**