

**Practical Lab**  
**Variational Methods and Inverse Problems in Imaging**  
Summer term 2014  
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**Depth from stereo – Problem sheet 4**

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At first we will consider the ROF model, i.e. we aim for minimizing  $E[u] = F[u] + G[\Lambda u]$ , where  $F[v] = \frac{\alpha}{2} \int_{\Omega} (v - g)^2 dx$ ,  $\Lambda v = \nabla v$ ,  $G[q] = \int_{\Omega} |q| dx$ . We will however consider the three dimensional case, i.e.  $\Omega \subset \mathbb{R}^3$ , as we will later embed a lifted model for stereo reconstruction into this framework. Details will be presented in the lecture. In the following you will find a rough sketch on how to tackle the ROF model.

To find a minimizer for the ROF model we will employ a primal dual method, i.e. we will alternately minimize a primal and a dual energy. In the lecture you have already seen the conjugate. The problem is that  $E^*$  in our case is not easy to compute. However one can find a so called pre dual  $D$  that fulfills  $E = D^*$ . It is given by:

$$D[q] = \frac{1}{2\alpha} \|\operatorname{div} q + \alpha g\|^2 - \frac{\alpha}{2} \|g\|^2 + I_{K_1}[q], \quad \text{where } I_{K_1}[q] = \begin{cases} 0 & \text{if } |q| \leq 1 \\ \infty & \text{else} \end{cases}$$

Minimizers  $u$  of  $E$  and  $p$  of  $D$  will form a saddle point of the functional

$$\mathcal{L}[v, q] = \frac{\alpha}{2} \|v - g\|^2 - (v, \operatorname{div} q) - I_{K_1}[q].$$

This saddle point problem can be tackled using an efficient gradient descent / ascent approach which alternately solves the following problems

$$\left( -\partial_{\tau} p_h^{n+1} + \nabla_h \tilde{u}_h^{n+1}, q_h - p_h^{n+1} \right)_h \leq 0 \quad \forall q_h \in \mathcal{V}_h^d \quad (\text{p})$$

$$\partial_{\tau} u_h^{n+1} = -\alpha(u_h^{n+1} - g) + \operatorname{div} p_h^{n+1}, \quad u_h \in \mathcal{V}_h \quad (\text{u})$$

where  $\partial_{\tau} a^{n+1} = \frac{1}{\tau}(a^{n+1} - a^n)$  and  $\tilde{u}_h^{n+1} = u_h^n + \tau \partial_{\tau} u_h^n = 2u_h^n - u_h^{n-1}$ .

We will solve these problems on three dimensional simplicial grids and use piecewise affine finite elements as the ansatz space  $\mathcal{V}_h$ .

Problem (p) can be solved point wise explicitly, i.e. for every node  $z \in \mathcal{N}_h$  the following update formula can be used.

$$p_h^{n+1} = \operatorname{Proj}_{\mathcal{K}} \left( p_h^n + \tau \nabla_h \tilde{u}_h^{n+1} \right), \quad \text{where } \operatorname{Proj}_{\mathcal{K}}(p) = \frac{p}{\max\{1, |p|\}}$$

Problem (u) will be solved in the usual finite element setting:

$$\left( (1 + \alpha) u_h^{n+1}, v_h \right)_h = (u_h^n, v_h)_h - \tau \left( p_h^{n+1}, \nabla_h v_h \right)_h + \alpha (g, v_h)_h$$

The scalar product  $(\cdot, \cdot)_h$  represents lumped masses, i.e.

$$(q_h, p_h)_h = \int_{\Omega} \mathcal{I}_h[q_h \cdot p_h] \, dx = \sum_{z \in \mathcal{N}_h} \left( \int_{\Omega} \phi_z \, dx \right) p_h(z) \cdot q_h(z)$$

This approach will lead to a diagonal mass matrix on the left hand side whose inverse can be computed and applied efficiently.

**Tasks:**

- Go through the sketched algorithm and think about which kind of operators you will need to implement to tackle problems (u) and (p).
- Use your projection framework to implement  $\text{Proj}_{\mathcal{K}}$ .
- Implement the finite element scheme to solve problem (u).