

# DIFFERENTIAL EQUATIONS IN IMAGE PROCESSING AND COMPUTER VISION

## ASSIGNMENT T4

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## 4.1 Continuous Variational Regularisation

We have

$$E_f(u) = \int_{\Omega} (\Psi_1((u-f)^2) + \alpha\Psi_2(|\nabla u|^2)) \, d\mathbf{x}$$

with

$$\begin{aligned}\Psi_1(s^2) &= \sqrt{s^2 + \varepsilon^2} \\ \Psi_2(s^2) &= \lambda^2 \log(1 + s^2/\lambda^2)\end{aligned}$$

and  $\alpha, \varepsilon, \lambda > 0$ .

It follows:

$$E_f(u) = \int_{\Omega} \left( \sqrt{(u-f)^2 + \varepsilon^2} + \alpha\lambda^2 \log(1 + |\nabla u|^2/\lambda^2) \right) \, d\mathbf{x}$$

We have

$$E_f(u) = \int_{\Omega} F(x_1, x_2, u, u_{x_1}, u_{x_2}) \, d\mathbf{x}$$

which has to satisfy the Euler-Lagrange equation

$$F_u - \partial_{x_1} F_{u_{x_1}} - \partial_{x_2} F_{u_{x_2}} = 0$$

with the boundary condition

$$\mathbf{n}^\top \begin{pmatrix} F_{u_{x_1}} \\ F_{u_{x_2}} \end{pmatrix} = 0.$$

$$\begin{aligned}F &= \sqrt{(u-f)^2 + \varepsilon^2} + \alpha\lambda^2 \log(1 + |\nabla u|^2/\lambda^2) \\ F_u &= \frac{1}{2} \cdot ((u-f)^2 + \varepsilon^2)^{-\frac{1}{2}} \cdot 2(u-f) \\ &= \frac{u-f}{\sqrt{(u-f)^2 + \varepsilon^2}} \\ F_{u_{x_1}} &= \alpha\lambda^2 \frac{1}{1 + |\nabla u|^2/\lambda^2} 2u_{x_1}/\lambda^2 \\ &= 2\alpha \frac{u_{x_1}}{1 + |\nabla u|^2/\lambda^2} \\ F_{u_{x_2}} &= \alpha\lambda^2 \frac{1}{1 + |\nabla u|^2/\lambda^2} 2u_{x_2}/\lambda^2 \\ &= 2\alpha \frac{u_{x_2}}{1 + |\nabla u|^2/\lambda^2}\end{aligned}$$

Just plugging in the results into the Euler-Lagrange equation yields

$$\begin{aligned} \frac{u-f}{\sqrt{(u-f)^2 + \varepsilon^2}} - \partial_{x_1} 2\alpha \frac{u_{x_1}}{1 + |\nabla u|^2/\lambda^2} - \partial_{x_2} 2\alpha \frac{u_{x_2}}{1 + |\nabla u|^2/\lambda^2} &= 0 \\ \frac{u-f}{\sqrt{(u-f)^2 + \varepsilon^2}} - 2\alpha \partial_{x_1} \frac{u_{x_1}}{1 + |\nabla u|^2/\lambda^2} - 2\alpha \partial_{x_2} \frac{u_{x_2}}{1 + |\nabla u|^2/\lambda^2} &= 0 \\ \frac{u-f}{\sqrt{(u-f)^2 + \varepsilon^2}} - 2\alpha \operatorname{div} \left( \frac{\nabla u}{1 + |\nabla u|^2/\lambda^2} \right) &= 0 \end{aligned}$$

with the boundary condition

$$\mathbf{n}^\top \begin{pmatrix} F_{u_{x_1}} \\ F_{u_{x_2}} \end{pmatrix} = 0.$$

This knowledge allows to formulate the gradient descent:

$$\begin{aligned} \partial_t u &= -\gamma \nabla_u E \\ &= -\gamma \cdot \left( \frac{u-f}{\sqrt{(u-f)^2 + \varepsilon^2}} - 2\alpha \operatorname{div} \left( \frac{\nabla u}{1 + |\nabla u|^2/\lambda^2} \right) \right) \end{aligned}$$

## 4.2 Rotation Invariance of Energy Functionals

In both cases it should hold the following:

$$\Psi(u_x, u_y) = \Psi(u_\xi, u_\eta)$$

a.  $\Psi(u_x, u_y) = |u_x| + |u_y|.$

It follows that  $\Psi(u_\xi, u_\eta) = |u_\xi| + |u_\eta|$ . From the hint we can get the following for  $u_\xi$  and  $u_\eta$ :

$$\begin{aligned} u_\xi &= u_x \cdot \cos \vartheta + u_y \cdot \sin \vartheta \\ u_\eta &= -u_x \cdot \sin \vartheta + u_y \cdot \cos \vartheta \end{aligned}$$

$$\Rightarrow \Psi(u_\xi, u_\eta) = |u_x \cdot \cos \vartheta + u_y \cdot \sin \vartheta| + |-u_x \cdot \sin \vartheta + u_y \cdot \cos \vartheta|$$

Let's have a look at the case that  $\vartheta = \frac{\pi}{4}$ . Then we get

$$\Psi(u_\xi, u_\eta) = \frac{\sqrt{2}}{2} (|u_x + u_y| + |u_y - u_x|) \neq |u_x| + |u_y| = \Psi(u_x, u_y)$$

Thus  $\Psi(u_x, u_y) = |u_x| + |u_y|$  is not rotationally invariant.

b.  $\Psi(u_x, u_y) = u_x^2 + u_y^2$ .

It follows  $\Psi(u_\xi, u_\eta) = u_\xi^2 + u_\eta^2$ . Using again the hint, we get for  $u_\xi$  and  $u_\eta$  the same as above.

$$\begin{aligned} \Rightarrow \Psi(u_\xi, u_\eta) &= (u_x \cdot \cos \vartheta + u_y \cdot \sin \vartheta)^2 + (u_y \cdot \cos \vartheta - u_x \sin \vartheta)^2 \\ &= u_x^2 \cos^2 \vartheta + 2u_x u_y \cos \vartheta \sin \vartheta + u_y^2 \sin^2 \vartheta \\ &\quad + u_y^2 \cos^2 \vartheta - 2u_x u_y \cos \vartheta \sin \vartheta + u_x^2 \sin^2 \vartheta \\ &= u_x^2 \underbrace{(\cos^2 \vartheta + \sin^2 \vartheta)}_{=1} + u_y^2 \underbrace{(\cos^2 \vartheta + \sin^2 \vartheta)}_{=1} \\ &= u_x^2 + u_y^2 = \Psi(u_x, u_y) \end{aligned}$$

Thus  $\Psi(u_x, u_y) = u_x^2 + u_y^2$  is rotationally invariant.

### 4.3 Discrete Energy Minimisation

$$E_f(u_1, \dots, u_N) = \sum_{k=1}^N (u_k - f_k)^2 + \alpha \sum_{k=1}^{N-1} (u_{k+1} - u_k)^2 + \beta \sum_{k=2}^{N-1} (u_{k+1} - 2u_k + u_{k-1})^2$$

a. Setting  $\frac{\partial E}{\partial u_i} = 0$  for  $i = 1, \dots, N$  gives

$$\begin{aligned} 0 &= 2(u_1 - f_1) - 2\alpha(u_2 - u_1) + 2\beta(u_3 - 2u_2 + u_1) \\ &= u_1 - f_1 - \alpha(u_2 - u_1) + \beta(u_3 - 2u_2 + u_1) \\ 0 &= 2(u_2 - f_2) - 2\alpha(u_3 - u_2) + 2\alpha(u_2 - u_1) + 2\beta(u_4 - 2u_3 + u_2) \\ &\quad - 4\beta(u_3 - 2u_2 + u_1) \\ &= u_2 - f_2 - \alpha(u_3 - u_2) + \alpha(u_2 - u_1) + \beta(u_4 - 2u_3 + u_2) \\ &\quad - 2\beta(u_3 - 2u_2 + u_1) \\ 0 &= 2(u_i - f_i) - 2\alpha(u_{i+1} - u_i) + 2\alpha(u_i - u_{i-1}) \\ &\quad + 2\beta(u_{i+2} - 2u_{i+1} + u_i) - 4\beta(u_{i+1} - 2u_i + u_{i-1}) \\ &\quad + 2\beta(u_i - 2u_{i-1} + u_{i-2}) \quad (i = 3, \dots, N-2) \\ &= (u_i - f_i) - \alpha(u_{i+1} - u_i) + \alpha(u_i - u_{i-1}) \\ &\quad + \beta(u_{i+2} - 2u_{i+1} + u_i) - 2\beta(u_{i+1} - 2u_i + u_{i-1}) \\ &\quad + \beta(u_i - 2u_{i-1} + u_{i-2}) \quad (i = 3, \dots, N-2) \end{aligned}$$

$$\begin{aligned}
0 &= 2(u_{N-1} - f_{N-1}) - 2\alpha(u_N - u_{N-1}) + 2\alpha(u_{N-1} - u_{N-2}) \\
&\quad - 4\beta(u_N - 2u_{N-1} + u_{N-2}) + 2\beta(u_{N-1} - 2u_{N-2} + u_{N-3}) \\
&= u_{N-1} - f_{N-1} - \alpha(u_N - u_{N-1}) + \alpha(u_{N-1} - u_{N-2}) \\
&\quad - 2\beta(u_N - 2u_{N-1} + u_{N-2}) + \beta(u_{N-1} - 2u_{N-2} + u_{N-3}) \\
0 &= 2(u_N - f_N) + 2\alpha(u_N - u_{N-1}) - 2\beta(u_N - 2u_{N-1} + u_{N-2}) \\
&= u_N - f_N + \alpha(u_N - u_{N-1}) - \beta(u_N - 2u_{N-1} + u_{N-2})
\end{aligned}$$

- b. I would say that the terms where a  $\beta$  stands in front of are the discretisations of the fourth order derivatives. If you combine these terms to one term you get a structure where  $u_{i+2}, u_{i+1}, u_i, u_{i-1}, u_{i-2}$  are involved, so this is not a second order but a fourth order derivative.

#### 4.4 Half-Quadratic Regularisation

$$E_f(u) = \int_{\Omega} \left( (u - f)^2 + \alpha 2\lambda^2 \left( 1 - \exp\left(-\frac{|\nabla u|^2}{2\lambda^2}\right) \right) \right) dx$$

I can rewrite this into

$$E_f(u) = \int_{\Omega} ((u - f)^2 + \alpha \Psi(|\nabla u|^2)) dx$$

with

$$\Psi(|\nabla u|^2) = 2\lambda^2 \left( 1 - \exp\left(-\frac{|\nabla u|^2}{2\lambda^2}\right) \right)$$

$$\begin{aligned}
F &= (u - f)^2 + \alpha \Psi(|\nabla u|^2) \\
F_u &= 2(u - f) \\
F_{u_{x_1}} &= 2\alpha\lambda^2 \left( -\exp\left(-\frac{|\nabla u|^2}{2\lambda^2}\right) \right) \cdot (-2u_{x_1}/2\lambda^2) \\
&= 2\alpha u_{x_1} \exp\left(-\frac{|\nabla u|^2}{2\lambda^2}\right) \\
F_{u_{x_2}} &= 2\alpha\lambda^2 \left( -\exp\left(-\frac{|\nabla u|^2}{2\lambda^2}\right) \right) \cdot (-2u_{x_2}/2\lambda^2) \\
&= 2\alpha u_{x_2} \exp\left(-\frac{|\nabla u|^2}{2\lambda^2}\right)
\end{aligned}$$

Its Euler-Lagrange equation is given by (divided by 2)

$$0 = (u - f) - \alpha \cdot \operatorname{div} \left( \underbrace{\exp \left( -\frac{|\nabla u|^2}{2\lambda^2} \right)}_{\Psi'(|\nabla u|^2)} \nabla u \right)$$

Let's formulate the energy functional for the Half-Quadratic case:

$$E_{HQ}(u, v) := \int_{\Omega} ((u - f)^2 + \alpha(v \cdot |\nabla u|^2 + \eta(v))) \, d\mathbf{x}$$

$E_{HQ}(u, v)$  has the two Euler-Lagrange equations:

$$\begin{aligned} 0 &= (u - f) - \alpha \operatorname{div}(v \nabla u) \\ 0 &= |\nabla u|^2 + \eta'(v) \end{aligned}$$

We know that

$$v := \Psi'(|\nabla u|^2) = \exp \left( -\frac{|\nabla u|^2}{2\lambda^2} \right)$$

To obtain  $|\nabla u|^2$ , let's solve  $v$  for  $|\nabla u|^2$ :

$$\begin{aligned} v &= \exp \left( -\frac{|\nabla u|^2}{2\lambda^2} \right) \\ \log(v) &= -\frac{|\nabla u|^2}{2\lambda^2} \\ |\nabla u|^2 &= -\log(v) \cdot 2\lambda^2 \\ &= -2\lambda^2 \cdot \log(v) \end{aligned}$$

Then we can rewrite the Euler-Lagrange equations

$$\begin{aligned} 0 &= (u - f) - \alpha \operatorname{div}(v \nabla u) \\ 0 &= |\nabla u|^2 \underbrace{- 2\lambda^2 \log(v)}_{=\eta'(v)} \end{aligned}$$

We know that  $\eta'(v) = -2\lambda^2 \log(v)$  and thus

$$\eta(v) = -2\lambda^2 v (\log(v) - 1)$$

This gives

$$E_{HQ}(u, v) := \int_{\Omega} ((u - f)^2 + \alpha(v |\nabla u|^2 - 2\lambda^2 v (\log(v) - 1))) \, d\mathbf{x}$$