Image Processing and Computer Vision (IPCV)



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Example Solutions for Classroom Assignment 9 (C9)

Problem 1 (Convexity of a Discrete Energy Function)

We want to analyse the convexity of

$$E(\mathbf{u}) := \frac{1}{2} \sum_{k=1}^{N} (u_k - f_k)^2 + \frac{\alpha}{2} \sum_{k=1}^{N-1} (u_{k+1} - u_k)^2$$

with $\boldsymbol{u}, \boldsymbol{f} \in \mathbb{R}^N$ and $\alpha > 0$. We recall the definition of convexity:

A function $g: \mathbb{R}^n \to \mathbb{R}$ is called convex if for any $\boldsymbol{x}, \boldsymbol{y} \in \mathbb{R}^n$ and $t \in [0, 1]$ the following inequality holds

$$g(t\boldsymbol{x} + (1-t)\boldsymbol{y}) \leqslant tg(\boldsymbol{x}) + (1-t)g(\boldsymbol{y})$$

If for any $x, y \in \mathbb{R}^n$ with $x \neq y$ and $t \in (0, 1)$ we even have

$$g(t\boldsymbol{x} + (1-t)\boldsymbol{y}) < tg(\boldsymbol{x}) + (1-t)g(\boldsymbol{y})$$

then the function is called strictly convex. For functions from \mathbb{R} to \mathbb{R} the above definition of strict convexity is equivalent to requiring that the second derivative is strictly positive everywhere. Therefore, linear operations are clearly convex but not strictly convex. Furthermore, $f(x) = x^2$ is a strictly convex function because for $x \neq y$ and 0 < t < 1 we have

$$(tx + (1-t)y)^{2} < tx^{2} + (1-t)y^{2}$$

$$\Leftrightarrow t^{2}x^{2} + 2t(1-t)xy + (1-t)^{2}y^{2} < tx^{2} + (1-t)y^{2}$$

$$\Leftrightarrow 0 < t(1-t)x^{2} + t(1-t)y^{2} - 2t(1-t)xy$$

$$\Leftrightarrow 0 < t(1-t)(x-y)^{2}$$

where the last inequality is always fulfilled under the above assumptions. From this it follows that the squared norm in \mathbb{R}^n is also strictly convex since

$$||t\boldsymbol{x} + (1-t)\boldsymbol{y}||^2 = \sum_{i=1}^n (tx_i + (1-t)y_i)^2$$

$$< \sum_{i=1}^n tx_i^2 + (1-t)y_i^2$$

$$= t\left(\sum_{i=1}^n x_i^2\right) + (1-t)\left(\sum_{i=1}^n y_i^2\right)$$

$$= t||\boldsymbol{x}||^2 + (1-t)||\boldsymbol{y}||^2$$

Using these results it is now possible to show that the above energy is strictly convex. First let us now rewrite the energy in a more comfortable form.

$$E(\boldsymbol{u}) = \frac{1}{2} \|\boldsymbol{u} - \boldsymbol{f}\|^2 + \frac{\alpha}{2} \|D\boldsymbol{u}\|^2$$

where $D \in \mathbb{R}^{(N-1)\times N}$ is a matrix with the following structure

$$D = \begin{pmatrix} -1 & 1 & 0 & 0 & 0 & \dots & 0 \\ 0 & -1 & 1 & 0 & 0 & \dots & 0 \\ & & & \dots & & & \\ 0 & & \dots & 0 & -1 & 1 & 0 \\ 0 & & \dots & & 0 & -1 & 1 \end{pmatrix}$$

Now let $u, v \in \mathbb{R}^n$, $u \neq v$ and $0 < \beta < 1$, then we have

$$E(\beta \boldsymbol{u} + (1 - \beta)\boldsymbol{v}) = \frac{1}{2} \|\beta \boldsymbol{u} + (1 - \beta)\boldsymbol{v} - \boldsymbol{f}\|^{2} + \frac{\alpha}{2} \|D(\beta \boldsymbol{u} + (1 - \beta)\boldsymbol{v})\|^{2}$$

$$= \frac{1}{2} \|\beta (\boldsymbol{u} - \boldsymbol{f}) + (1 - \beta) (\boldsymbol{v} - \boldsymbol{f})\|^{2}$$

$$+ \frac{\alpha}{2} \|\beta D\boldsymbol{u} + (1 - \beta)D\boldsymbol{v}\|^{2}$$

$$< \frac{1}{2} \beta \|\boldsymbol{u} - \boldsymbol{f}\|^{2} + \frac{1}{2} (1 - \beta) \|\boldsymbol{v} - \boldsymbol{f}\|^{2}$$

$$+ \frac{\alpha}{2} \|\beta D\boldsymbol{u} + (1 - \beta)D\boldsymbol{v}\|^{2}$$

$$\leq \frac{1}{2} \beta \|\boldsymbol{u} - \boldsymbol{f}\|^{2} + \frac{1}{2} (1 - \beta) \|\boldsymbol{v} - \boldsymbol{f}\|^{2}$$

$$+ \frac{\alpha}{2} (\beta \|D\boldsymbol{u}\|^{2} + (1 - \beta) \|D\boldsymbol{v}\|^{2})$$

$$= \beta E(\boldsymbol{u}) + (1 - \beta) E(\boldsymbol{v})$$

Thus, the energy is strictly convex. Note that the first inequality is always strict because if $\boldsymbol{u} \neq \boldsymbol{v}$ then we also have $\boldsymbol{u} - \boldsymbol{f} \neq \boldsymbol{v} - \boldsymbol{f}$ but the second inequality must not necessarily be strict. If both \boldsymbol{u} and \boldsymbol{v} are in the nullspace of D, then $D\boldsymbol{u} = D\boldsymbol{v} = 0$ and we cannot have a strict inequality. Concerning possible minimisers, we observe that the energy is bounded below and coercive (e.g. $\|\boldsymbol{v}\| \to \infty$ implies $E(\boldsymbol{v}) \to \infty$ as well.) Thus, there exists a closed and bounded level set which contains the minimiser. Finally, because of the strict convexity of the energy, we can even state that this minimum is unique.

Remark: The coercivity of E is really necessary because functions like the exponential function are also strictly convex and bounded below but they fail to

have a minimum. They possess an infimum to which one can come arbitrarily close, but which one cannot attain. Here, the coercivity assures that there exists a compact level set that contains the infimum. Since E is a continuous function, it must attain its minimum in this compact level set (Extreme value theorem or Theorem of Weierstrass/Bolzano).

Problem 2 (Euler-Lagrange Equations)

Given is the 3-D energy functional

$$E(u) = \int_{\Omega} \left(\frac{(u-f)^2}{2} + \alpha \sqrt{\epsilon + |\nabla u|^2} \right) dx dy dz$$

The Euler-Lagrange equation in 3-D looks as follows:

$$F_{u} - \frac{\partial}{\partial x} F_{u_{x}} - \frac{\partial}{\partial y} F_{u_{y}} - \frac{\partial}{\partial z} F_{u_{z}} = 0$$

where we have in this case:

$$F(x, y, u, u_x, u_y, u_z) = \frac{(u - f)^2}{2} + \alpha \sqrt{\epsilon + |\nabla u|^2}$$
$$= \frac{(u - f)^2}{2} + \alpha \sqrt{\epsilon + u_x^2 + u_y^2 + u_z^2}$$

Let us compute the derivatives in the Euler-Lagrange equation:

$$F_{u_x} = u - f$$

$$F_{u_x} = \alpha \frac{2u_x}{2\sqrt{\epsilon + u_x^2 + u_y^2 + u_y^2}} = \alpha \frac{u_x}{\sqrt{\epsilon + u_x^2 + u_y^2 + u_z^2}}$$

$$F_{u_y} = \alpha \frac{2u_y}{2\sqrt{\epsilon + u_x^2 + u_y^2 + u_z^2}} = \alpha \frac{u_y}{\sqrt{\epsilon + u_x^2 + u_y^2 + u_z^2}}$$

$$F_{u_z} = \alpha \frac{2u_z}{2\sqrt{\epsilon + u_x^2 + u_y^2 + u_z^2}} = \alpha \frac{u_z}{\sqrt{\epsilon + u_x^2 + u_y^2 + u_z^2}}$$

$$\frac{\partial}{\partial x} F_{u_x} = \alpha \frac{u_{xx} \sqrt{\epsilon + u_x^2 + u_y^2 + u_z^2} - u_x \frac{u_x u_{xx} + u_y u_{yx} + u_z u_{zx}}{\sqrt{\epsilon + u_x^2 + u_y^2 + u_z^2}}}{(\epsilon + u_x^2 + u_y^2 + u_z^2)}$$

$$= \alpha \frac{u_{xx} (\epsilon + u_x^2 + u_y^2 + u_z^2) - u_x^2 u_{xx} - u_x u_y u_{yx} - u_x u_z u_{zx}}{(\epsilon + u_x^2 + u_y^2 + u_z^2)^{\frac{3}{2}}}$$

$$= \alpha \frac{u_{xx}(\epsilon + u_y^2 + u_z^2) - u_x u_y u_{xy} - u_x u_z u_{xz}}{(\epsilon + u_x^2 + u_y^2 + u_z^2)^{\frac{3}{2}}}$$

$$\frac{\partial}{\partial y} F_{u_y} = \alpha \frac{u_{yy}(\epsilon + u_x^2 + u_z^2) - u_y u_x u_{xy} - u_y u_z u_{zy}}{(\epsilon + u_x^2 + u_y^2 + u_z^2)^{\frac{3}{2}}}$$

$$\frac{\partial}{\partial z} F_{u_z} = \alpha \frac{u_{zz}(\epsilon + u_x^2 + u_y^2) - u_z u_x u_{xz} - u_z u_y u_{yz}}{(\epsilon + u_x^2 + u_y^2 + u_z^2)^{\frac{3}{2}}}$$

The resulting Euler-Lagrange equation is:

$$u - f - \alpha \frac{u_{xx}(\epsilon + u_y^2 + u_z^2) - u_x u_y u_{xy} - u_x u_z u_{xz}}{(\epsilon + u_x^2 + u_y^2 + u_z^2)^{\frac{3}{2}}}$$

$$- \alpha \frac{u_{yy}(\epsilon + u_x^2 + u_z^2) - u_y u_x u_{xy} - u_y u_z u_{zy}}{(\epsilon + u_x^2 + u_y^2 + u_z^2)^{\frac{3}{2}}}$$

$$- \alpha \frac{u_{zz}(\epsilon + u_x^2 + u_y^2) - u_z u_x u_{xz} - u_z u_y u_{yz}}{(\epsilon + u_x^2 + u_y^2) - u_z u_x u_{xz} - u_z u_y u_{yz}}$$

$$= 0$$

$$\Leftrightarrow 2(u-f) - \alpha \frac{u_{xx}(\epsilon + u_y^2 + u_z^2) + u_{yy}(\epsilon + u_x^2 + u_z^2) + u_{zz}(\epsilon + u_x^2 + u_y^2)}{(\epsilon + u_x^2 + u_y^2 + u_z^2)^{\frac{3}{2}}}$$

$$- 2 \frac{u_x u_y u_{xy} + u_x u_z u_{xz} + u_z u_y u_{yz}}{(\epsilon + u_x^2 + u_y^2 + u_z^2)^{\frac{3}{2}}} = 0$$