

Exercise 2 To show error functionals are convex

$\phi: X \rightarrow \mathbb{R}$  on space  $X$  is convex if the inequality  $\phi(\lambda x + (1-\lambda)y) \leq \lambda \phi(x) + (1-\lambda)\phi(y)$

$x, y \in X$  and  $\lambda \in [0, 1]$

(a)  $X \rightarrow$  linear space  $\|\cdot\|_X$

$$\|\lambda x + (1-\lambda)y\| \leq \|\lambda x\| + \|(1-\lambda)y\| \quad (\text{using } \Delta \text{ inequality})$$

for  $x, y \in X$

$$= \lambda \|x\| + (1-\lambda)\|y\|$$

(by absolutely scalable property of norm)

$\Rightarrow \|\cdot\|_X$  is convex

(b)  $L^2 - H^1$  denoising

$$\phi: H^1(\mathbb{R}^2) \rightarrow \mathbb{R}$$

$$u \mapsto \frac{1}{2} \|u - u_0\|_{L^2}^2 + \frac{\mu}{2} \|\nabla u\|_{L^2}^2$$

given image  $u_0 \in L^2(\mathbb{R}^2)$   
regularization parameters  $\mu > 0$

$\phi(\lambda) := \lambda^2$  is monotone and convex on  $[0, \infty)$

Soln (b):  $\phi(\lambda u + (1-\lambda)v)$

$$= \frac{1}{2} \|(\lambda u + (1-\lambda)v) - u_0\|_2^2$$

$$+ \frac{\mu}{2} \|\nabla(\lambda u + (1-\lambda)v)\|_2^2$$

$$= \frac{1}{2} \|\lambda u + (1-\lambda)v - u_0 - \lambda u_0 + \lambda u_0\|_2^2 + \frac{\mu}{2} \|\nabla(\lambda u) + \nabla((1-\lambda)v)\|_2^2$$

$$= \frac{1}{2} \|\lambda u + (1-\lambda)v - \lambda u_0 - (1-\lambda)u_0\|_2^2 + \frac{\mu}{2} \|\lambda(\nabla u) + (1-\lambda)(\nabla v)\|_2^2$$

$$= \frac{1}{2} \|\lambda(u - u_0) + (1-\lambda)(v - u_0)\|_2^2 + \frac{\mu}{2} \|\lambda(\nabla u) + (1-\lambda)(\nabla v)\|_2^2$$

→ (1)

Now, ~~the~~  $f(x) = \|x\|_2$  is convex (from part (a))

$e(x) = x^2$  is monotone and convex on  $[0, \infty)$

(non decreasing)

range of  $f(x)$

$\phi = \| \cdot \|_2^2$  is a composition of 2 convex functions which is convex too.

Eq. (1) is a summand of 2 convex functions which will again be a convex function.

~~$$\begin{aligned}
 & \phi(\lambda u + (1-\lambda)v) \\
 &= (\lambda u + (1-\lambda)v)^2 \\
 &= \lambda^2 u^2 + (1-\lambda)^2 v^2 \\
 &\quad + 2\lambda(1-\lambda)uv \\
 &\leq e(\lambda u) + e((1-\lambda)v)
 \end{aligned}$$~~