

# EN.550.661: Nonlinear Optimization I

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## Homework Assignment #1

*Starred exercises require the use of MATLAB.*

**Exercise 1.1:** Compute  $f'(x)$ ,  $\nabla f(x)$ , and  $\nabla^2 f(x)$  for the following functions  $f : \mathbb{R}^n \rightarrow \mathbb{R}$ .

- (a)  $f(x) = \frac{1}{2}x^T H x$ , where  $H \in \mathbb{R}^{n \times n}$  is a constant matrix. What if  $H$  is symmetric?
- (b)  $f(x) = b^T A x - \frac{1}{2}x^T A^T A x$ , where  $A \in \mathbb{R}^{m \times n}$  is a constant matrix and  $b \in \mathbb{R}^m$  is a constant vector.
- (c)  $f(x) = \|x\|_2 = (\sum_{i=1}^n x_i^2)^{1/2}$
- (d)  $f(x) = \|Ax - b\|_2$ , where  $A \in \mathbb{R}^{m \times n}$  is a constant matrix and  $b \in \mathbb{R}^m$  is a constant vector.

**Exercise 1.2:** Let  $\mathcal{S} \subseteq \mathbb{R}^n$ ,  $f : \mathcal{S} \rightarrow \mathbb{R}$ , and suppose that  $x + \alpha s \in \mathcal{S}$  for all  $\alpha \in [0, 1]$ .

- (a) By defining  $\theta(\alpha) = f(x + \alpha s)$  and using the integration-by-parts (Newton) formula

$$\theta(1) - \theta(0) = \int_0^1 \theta'(\alpha) d\alpha,$$

show that

$$|f(x + s) - f(x) - g(x)^T s| \leq \frac{1}{2} \gamma^L(x) \|s\|_2^2$$

whenever  $f \in \mathcal{C}^1$  has a Lipschitz continuous gradient  $g(x)$  with constant  $\gamma^L(x)$  on  $\mathcal{S}$ .

- (b) Justify the integration-by-parts formula

$$\theta(1) - \theta(0) - \theta'(0) = \int_0^1 (1 - \alpha) \theta''(\alpha) d\alpha.$$

Hence, show that

$$|f(x + s) - f(x) - g(x)^T s - \frac{1}{2} s^T H(x) s| \leq \frac{1}{6} \gamma^Q(x) \|s\|_2^3,$$

whenever  $f \in \mathcal{C}^2$  has a Lipschitz continuous Hessian  $H(x)$  with constant  $\gamma^Q(x)$  on  $\mathcal{S}$ .

**Exercise 1.3:** Let  $f_i : \mathbb{R}^n \rightarrow \overline{\mathbb{R}}$  be a *convex* function and  $0 \leq \alpha_i \in \mathbb{R}$  for  $i = 1, \dots, k$ .

- (a) Prove that

$$f(x) = \sum_{i=1}^k \alpha_i f_i(x)$$

is a convex function.

- (b) Prove that

$$f(x) = \max(f_1(x), f_2(x), \dots, f_k(x))$$

is a convex function.

**Exercise 1.4:** Let  $x^*$  be a *local* minimizer of a *convex* function  $f : \mathbb{R}^n \rightarrow \overline{\mathbb{R}}$ . Prove that  $x^*$  is, in fact, a *global* minimizer of  $f$ . Is  $x^*$  a *unique* global minimizer?