

# **236330 - Introduction to Optimization: Homework #1**

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**Analytical Differentiation:****Task 1:**

Find the Gradient and Hessian of:

$$f_1(x) = \varphi(Ax)$$

Where,

$$\begin{aligned} f_1 : \mathbb{R}^n &\rightarrow \mathbb{R} \\ x &\in \mathbb{R}^n \\ A &\in \mathbb{R}^{m \times n} \\ \varphi : \mathbb{R}^m &\rightarrow \mathbb{R} \end{aligned}$$

Given  $\nabla\varphi$  and  $\nabla^2\varphi$ .

**Solution:**

We start by finding the gradient:

Denoting  $u = Ax$ . Hence  $du = dAx$ .

From the external definition of gradient, we know that,

$$d\varphi = \langle \nabla\varphi, du \rangle = \nabla\varphi^T du = \nabla\varphi^T A dx$$

Since  $df = \nabla f^T dx$  and  $d\varphi = df$  (because  $f(x) = \varphi(Ax)$ ) we get that:

$$g = \nabla f = A^T \nabla\varphi(Ax)$$

Now for the Hessian,

$$dg = d\nabla f = dA^T \nabla\varphi(u) = A^T d\nabla\varphi(u)$$

From the external definition of Hessian we know that  $d\nabla\varphi(u) = H du$ .

$$A^T d\nabla\varphi(u) = A^T H(u) du = A^T H(Ax) A dx$$

Since  $H = \nabla^2\varphi$ , it yields that:

$$H(x) = A^T \nabla^2\varphi(Ax) A$$

**Task 2:**

Develop the Gradient and Hessian of:

$$f_2(x) = h(\varphi(x))$$

Where,

$$\begin{aligned} f_2 : \mathbb{R}^n &\rightarrow \mathbb{R} \\ x &\in \mathbb{R}^n \\ \varphi : \mathbb{R}^m &\rightarrow \mathbb{R} \\ h : \mathbb{R} &\rightarrow \mathbb{R} \end{aligned}$$

Given  $\nabla\varphi$ ,  $\nabla^2\varphi$ ,  $h'$ ,  $h''$ .

**Solution:**

We start by finding the gradient:

Denoting  $u = \varphi(x)$ . From the external definition of the gradient, we know that:

$$du = \langle \nabla\varphi(x), dx \rangle$$

$$dh = h'(u) du$$

Combining these two together yields:

$$dh = h'(u) \langle \nabla\varphi(x), dx \rangle = \langle h'(u) \nabla\varphi(x), dx \rangle = \langle h'(\varphi(x)) \nabla\varphi(x), dx \rangle$$

Since  $df = \nabla f^T dx$  we get that:

$$g = \nabla f = (h'(\varphi(x)) \nabla\varphi(x))^T \stackrel{(1)}{=} h'(\varphi(x)) \nabla\varphi(x)$$

(1) - Since  $h'(\varphi(x))$  is a scalar and  $\nabla\varphi(x) = \nabla\varphi^T(x)$ .

Now for the Hessian,

$$\begin{aligned}
dg &= d\nabla f = d(h'(u) \nabla \varphi(x)) \\
&= dh'(u) \nabla \varphi(x) + h'(u) d\nabla \varphi(x) \\
&= h''(u) du \nabla \varphi(x) + h'(u) \nabla^2 \varphi(x) dx \\
&\stackrel{(1)}{=} h''(u) \nabla \varphi(x) du + h'(u) \nabla^2 \varphi(x) dx \\
&= h''(u) \cdot \nabla \varphi(x) \cdot \langle \nabla \varphi(x), dx \rangle + h'(u) \nabla^2 \varphi(x) dx \\
&= h''(u) \cdot \nabla \varphi(x) \cdot \nabla \varphi^T(x) dx + h'(u) \nabla^2 \varphi(x) dx \\
&= (h''(u) \cdot \nabla \varphi(x) \cdot \nabla \varphi^T(x) + h'(u) \nabla^2 \varphi(x)) \cdot dx
\end{aligned}$$

(1) - Since  $du = \langle \nabla \varphi(x), dx \rangle$  is a scalar.

From the external definition of Hessian we know that  $dg = Hdx$ , hence:

$$H = h''(u) \cdot \nabla \varphi(x) \cdot \nabla \varphi^T(x) + h'(u) \nabla^2 \varphi(x)$$