

cs-236330-optimization-hw1

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1 Task 1

1.1 Full Derivation of the Gradient of $f_1(x) = \varphi(Ax)$

Let: $Ax = u$

Therefore, the differential: $du = Adx$

And then, by the definition of external differential: $df_1 = \langle \nabla \varphi, du \rangle$

by the definition of euclidean Inner Product:

$$\langle \nabla \varphi, du \rangle = \nabla \varphi^T du = \nabla \varphi^T Adx = (A^T \nabla \varphi(Ax))^T dx$$

And to conclude:

$$\nabla f_1(x) = A^T \nabla \varphi(Ax) \quad (1)$$

1.2 Full Derivation of the Hessian of $f_1(x) = \varphi(Ax)$

Let: $g = \nabla f_1(x) = \nabla f_1(x) = A^T \nabla \varphi(Ax)$

By the external definition of the Hessian: $dg = Hdx$

Therefore:

$$d(A^T \nabla \varphi(Ax)) = A^T d(\nabla \varphi(u)) = A^T \langle \nabla \varphi, du \rangle = A^T \nabla^2 \varphi du = A^T \nabla^2 \varphi Adx \quad (2)$$

And then we conclude:

$$H = A^T \nabla^2 \varphi(Ax) A \quad (3)$$

2 Task 2

2.1 Full Derivation of the Gradient of $f_2(x) = h(\varphi(x))$

$$df_2(x) = d(h'(\varphi(x))) = \langle h'(\varphi(x)), d\varphi(x) \rangle = h'(\varphi(x)) d\varphi(x) (*)$$

$$dd\varphi(x) = \langle \nabla \varphi(x), dx \rangle = \nabla \varphi(x)^T dx$$

$$(*) df_2(x) = h'(\varphi(x)) \nabla \varphi(x)^T dx \Rightarrow \nabla f_2(x) = h'(\varphi(x)) \nabla \varphi(x)$$

2.2 Full Derivation of the Hessian of $f_2(x) = h(\varphi(x))$

$$\begin{aligned}
\text{Let: } g(x) &= \nabla f_2(x) = h'(\varphi(x)) \nabla \varphi(x) \\
dg &= d(h'(\varphi(x)) \nabla \varphi(x)) = \\
dh'(\varphi(x)) \nabla f_2(x) &+ h'(\varphi(x)) d\nabla f_2(x) = \\
\langle h''(\varphi(x)), d\varphi(x) \rangle \nabla \varphi(x) &+ h'(\varphi(x)) \langle \nabla^2 \varphi(x), dx \rangle = \\
h''(\varphi(x)) \nabla \varphi(x)^T dx \nabla \varphi(x) &+ h'(\varphi(x)) \nabla^2 \varphi(x) dx = \\
(h''(\varphi(x)) \nabla \varphi(x)^T \nabla \varphi(x) &+ h'(\varphi(x)) \nabla^2 \varphi(x)) dx \text{ And Therefore:}
\end{aligned}$$

$$H = h''(\varphi(x)) \nabla \varphi(x)^T \nabla \varphi(x) + h'(\varphi(x)) \nabla^2 \varphi(x) \quad (4)$$

3 Task 3

3.1 Full Derivation of the Gradient of:

$$\varphi \left(\begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix} \right) = \sin(x_1 x_2 x_3) \quad (5)$$

$$d\varphi \left(\begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix} \right) = \langle \nabla \sin(x_1 x_2 x_3), dx \rangle = \nabla \sin(x_1 x_2 x_3)^T dx$$

$$\nabla \sin(x_1 x_2 x_3) = \left(\begin{bmatrix} x_2 x_3 \\ x_1 x_3 \\ x_1 x_2 \end{bmatrix} \right) \cos(x_1 x_2 x_3) \Rightarrow$$

$$\nabla \varphi \left(\begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix} \right) = \left(\begin{bmatrix} x_2 x_3 \\ x_1 x_3 \\ x_1 x_2 \end{bmatrix} \right) \cos(x_1 x_2 x_3)$$

3.2 Full Derivation of the Hessian

$$g = \nabla \varphi \left(\begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix} \right)$$

$$dg = H dx$$

$$dg = \nabla^2 \varphi \left(\begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix} \right)^T dx$$

$$H = \nabla^2 \varphi \left(\begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix} \right)^T =$$

$$\begin{bmatrix} 0 & x_3 & x_2 \\ x_3 & 0 & x_1 \\ x_2 & x_1 & 0 \end{bmatrix} \cos(x_1 x_2 x_3) - \begin{bmatrix} x_2 & x_3 \\ x_1 & x_3 \\ x_1 & x_2 \end{bmatrix} \begin{bmatrix} x_2 & x_1 & x_1 \\ x_3 & x_3 & x_2 \end{bmatrix} \sin(x_1 x_2 x_3)$$

3.3 Full Derivation of the First & Second Derivatives of:

$$h(x) = \exp(x) \tag{6}$$

$$dh(x) = h'(x), dx \rangle = h'(x)dx$$

$$h'(x) = \exp(x)$$

$$dh''(x) = \exp(x)$$

3.4 Analytical Calculations:

Run the functions `run_f1` & `run_f2` in the file `hw1.py`

Each function calls `f1/f2` accordingly and receive a class called `f1/f2_par` which contains its parameters.

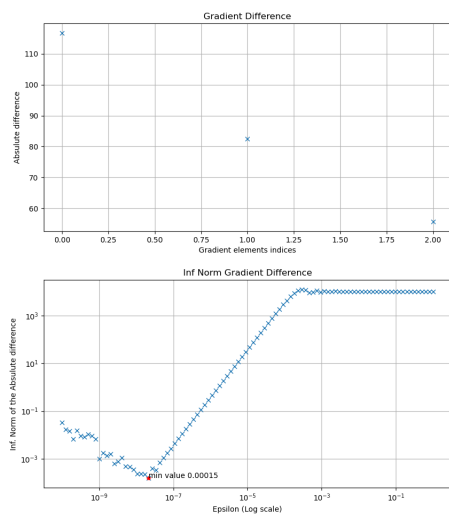
Inside `f1/f2_par` there are pointers to functions that do the calculations (called `phi`, `gradient_phi` and so on)

4 Task 4 - Numerical Calculations:

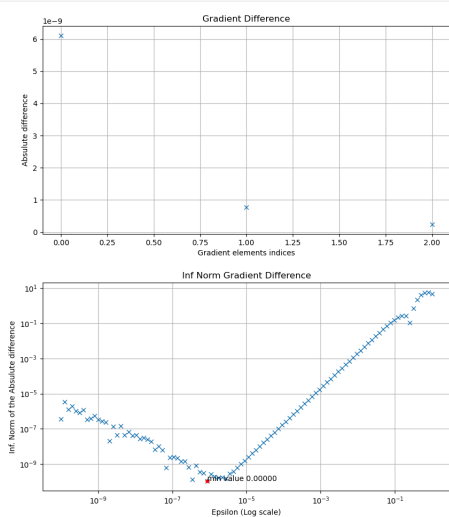
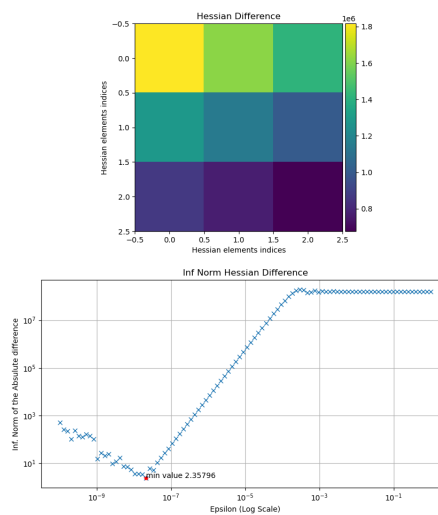
Run the function `run_numdiff` in the file `hw1.py`

Code structure is very similar to the code in section 3.

5 Task 5 - Comparisons



f1



f2

