

Grace Chang

Multidimensional Optimizer Math

Note, this is under the function

$$y = -b\mathbf{X} + \frac{1}{2}\mathbf{X}'\mathbf{A}\mathbf{X}$$

Gradient

eg)

$$b = \begin{bmatrix} 5 \\ -10 \end{bmatrix} \text{ and } A1 = \begin{bmatrix} 2 & 1 \\ 1 & 2 \end{bmatrix} \text{ and } A2 = \begin{bmatrix} 1 & -2 \\ -2 & 20 \end{bmatrix}. \text{ } x \text{ is just a } 1 \times 2 \text{ matrix of whatever I want.}$$

Then

$$\begin{aligned} f(x_1, x_2) &= -\vec{b} \cdot \vec{x} + \frac{1}{2}\vec{x}^\top A \vec{x} \\ &= -b_1x_1 - b_2x_2 + \frac{1}{2} \begin{bmatrix} x_1 & x_2 \end{bmatrix} A \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} \\ &= -b_1x_1 - b_2x_2 + \frac{1}{2} \begin{bmatrix} x_1 & x_2 \end{bmatrix} \begin{bmatrix} a_{11} & a_{12} \\ a_{21} & a_{22} \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} \\ &= -b_1x_1 - b_2x_2 + \frac{1}{2} \begin{bmatrix} x_1 & x_2 \end{bmatrix} \begin{bmatrix} a_{11}x_1 + a_{12}x_2 \\ a_{21}x_1 + a_{22}x_2 \end{bmatrix} \\ &= -b_1x_1 - b_2x_2 + \frac{1}{2} [a_{11}x_1^2 + a_{12}x_1x_2 + a_{21}x_1x_2 + a_{22}x_2^2] \end{aligned}$$

Thus we get the gradient

$$\nabla f(x_1, x_2) = \left(-b_1 + \frac{1}{2}(2a_{11}x_1 + a_{12}x_2 + a_{21}x_2), -b_2 + \frac{1}{2}(a_{12}x_1 + a_{21}x_1 + 2a_{22}x_2) \right)$$

Hessian Calculations

$$\begin{aligned} H_{ij} &= \begin{bmatrix} \frac{\partial^2 f}{\partial x_1^2} & \frac{\partial^2 f}{\partial x_1 \partial x_2} \\ \frac{\partial^2 f}{\partial x_2 \partial x_1} & \frac{\partial^2 f}{\partial x_2^2} \end{bmatrix} \\ &= \begin{bmatrix} \frac{1}{2}(2a_{11}) & \frac{1}{2}(a_{12} + a_{21}) \\ \frac{1}{2}(a_{12} + a_{21}) & \frac{1}{2}(2a_{22}) \end{bmatrix} \\ &= \begin{bmatrix} a_{11} & \frac{1}{2}(a_{12} + a_{21}) \\ \frac{1}{2}(a_{12} + a_{21}) & a_{22} \end{bmatrix} \end{aligned}$$

Because

$$\begin{bmatrix} a & b \\ c & d \end{bmatrix}^{-1} = \frac{1}{ad - bc} \begin{bmatrix} d & -b \\ -c & a \end{bmatrix}$$

Thus

$$H^{-1} = \frac{1}{a_{11}a_{22} - \frac{1}{4}(a_{12} + a_{21})^2} \begin{bmatrix} a_{22} & -\frac{1}{2}(a_{12} + a_{21}) \\ -\frac{1}{2}(a_{12} + a_{21}) & a_{11} \end{bmatrix}$$

Nonlinear Least Squares Optimizer

Using this exponential decay function:

$$f(x, a, k) = a \cdot \exp(-k \cdot x)$$

Thus

$$\begin{aligned} \frac{\partial f(x(i))}{\partial a} &= \exp(-k \cdot x) \\ \frac{\partial f(x(i))}{\partial k} &= -ax \exp(-k \cdot x) \end{aligned}$$

Thus the gradient components are:

$$\begin{aligned} G(1) &= \text{sum} \{ [a \exp(-k \cdot x_i) - y_i] [\exp(-k \cdot x)] \} \\ G(2) &= \text{sum} \{ [a \exp(-k \cdot x_i) - y_i] [-ax \exp(-k \cdot x)] \} \end{aligned}$$

For the Hessian:

$$\begin{aligned} H(1,1) &= \text{sum} \{ [\exp(-k \cdot x)]^2 \} \\ H(2,2) &= \text{sum} \{ (-a \cdot x_i \cdot \exp(-k \cdot x_i))^2 \} \\ H(1,2) = H(2,1) &= \text{sum} \{ \exp(-k \cdot x_i) (-a \cdot x_i \exp(-k \cdot x_i)) \} \end{aligned}$$

Next State

$$\begin{bmatrix} a_{i+1} \\ k_{i+1} \end{bmatrix} = \begin{bmatrix} a_i \\ k_i \end{bmatrix} - [H + \lambda \cdot \text{Diag}H]^{-1} G$$