

COMP206

Answer to Question 1.1

While $f \in R$

$$F(x) = -2x_1^2 - 2x_1x_2 + 2x_2^2 + x_1 + x_2 + 3$$

So that we can find the derivative as below

$$x'_1 = -4x_1 - 2x_2 + 1$$

$$x'_2 = -2x_1 - 4x_2 + 1$$

$$x''_1 = -4$$

$$x''_2 = -4$$

$$x'_1x'_2 = -2$$

$$x'_1x'_2 = -2$$

(1)

We can find our hession matrix as : $\begin{bmatrix} -4 & -2 \\ -2 & -4 \end{bmatrix}$

$$|D_1| = -4 < 0$$

$$|D_2| = 16 - 4 = 12 > 0$$

Thus, by definition that says $|D_1|$ is less than 0 and $|D_2|$ is bigger than 0

It means it is neither positive nor negative definite.

So, it is not convex.

Also we can find this solution with other test but it will give the same result.

Answer to Question 1.2

$$F(x, y, z) = 2x^2 + xz + y^2 + 2yz + 1/2 z^2$$

$$\frac{\partial^2 F}{\partial x^2} = 4 \quad \frac{\partial^2 F}{\partial x \partial y} = 0 \quad \frac{\partial^2 F}{\partial x \partial z} = 1$$

$$\frac{\partial^2 F}{\partial y \partial x} = 0 \quad \frac{\partial^2 F}{\partial y^2} = 2 \quad \frac{\partial^2 F}{\partial y \partial z} = 2$$

$$\frac{\partial^2 F}{\partial z \partial x} = 1 \quad \frac{\partial^2 F}{\partial z \partial y} = 2 \quad \frac{\partial^2 F}{\partial z^2} = 1$$

(2)

$$|D_1| = 4$$

$$|D_2| = 8$$

$$|D_3| = -10$$

$$\text{Our Hession matrix will be} = \begin{bmatrix} 4 & 0 & 1 \\ 0 & 2 & 2 \\ 1 & 2 & 1 \end{bmatrix}$$

It is neither positive nor negative definite while $|D_1| > 0$, $|D_2| > 0$, but the last one $|D_3| < 0$ thus it is not convex

Answer to Answer To Question 3

We can conclude while f is convex that is :

$$f(\lambda y + (1 - \lambda)x) \leq \lambda f(y) + (1 - \lambda)f(x),$$

$\forall x \in [0, 1], x, y \in \text{domain of } f \text{ function while } \lambda \text{ is between } 0 \text{ and } 1$

$$\text{So we can say that } f(x + \lambda(y - x)) \leq f(x) + \lambda(f(y) - f(x))$$

$$\text{So as } \lambda \text{ goes } 0, \text{ we find : } f(y) - f(x) \geq \frac{f(x + \lambda(y - x)) - f(x)}{\lambda}$$

$$\text{As conclusion we find that : } f(y) - f(x) \geq \nabla f^T(x)(y - x)$$

Thus if f is convex we can say that our proposition is true.

Answer to Answer To Question 4

If we say that f is convex, Then we can include function g as :

$$g(y) = f(y) - \nabla f(x)^T(y - x)$$

g is convex by the definition so that we can write :

$$\nabla g(y) = \nabla f(y) - \nabla f(x)$$

And

$$\nabla^2 g(y) = \nabla^2 f(y)$$

for all $y \in R^n$

$$\text{Thus } \nabla g(x) = 0$$

Thus, it implies $\nabla^2 g(x)$ is positive semi definite

In addition to that I found Taylor Theorem as below

$$f(y) = f(x) + \nabla f(x)^T(y - x) + \frac{1}{2}(y - x)^T \nabla^2 f(x + t(y - x))(y - x)$$

for some t is in 0 and 1. Since f is positive semi definite the quadratic equation will be always non negative Thus We will get

$$f(y) \geq f(x) + \nabla f(x)^T(y - x)$$

It is what we want to prove

(3)