

1 [Q1]

Last week, we talked about the gradient and the hessian. Gradient is the direction which the function descends fastest.

2 [Q2]

done

3 [Q3]

3.1 (a)

Assume $f(\mathbf{x}), f(\mathbf{y})$ are convex, then we have,

$$f(\mathbf{x}) \geq f(\mathbf{y}) + \langle \mathbf{x} - \mathbf{y}, \nabla f(\mathbf{y}) \rangle$$

$$f(\mathbf{y}) \geq f(\mathbf{x}) + \langle \mathbf{y} - \mathbf{x}, \nabla f(\mathbf{x}) \rangle$$

Combine the 2 inequalities together and we can get,

$$\begin{aligned} f(\mathbf{x}) + f(\mathbf{y}) &\geq f(\mathbf{y}) + f(\mathbf{x}) + \langle \mathbf{x} - \mathbf{y}, \nabla f(\mathbf{y}) \rangle + \langle \mathbf{y} - \mathbf{x}, \nabla f(\mathbf{x}) \rangle \\ &\quad - \langle \mathbf{x} - \mathbf{y}, \nabla f(\mathbf{y}) \rangle - \langle \mathbf{y} - \mathbf{x}, \nabla f(\mathbf{x}) \rangle \geq 0 \\ &\quad \langle \mathbf{y} - \mathbf{x}, \nabla f(\mathbf{y}) \rangle + \langle \mathbf{y} - \mathbf{x}, -\nabla f(\mathbf{x}) \rangle \geq 0 \\ &\quad \langle \mathbf{y} - \mathbf{x}, \nabla f(\mathbf{y}) - \nabla f(\mathbf{x}) \rangle \geq 0 \end{aligned}$$

3.2 (b)

3.3 (c)