

② Given,

$b \in \mathbb{R}^n$  is a vector

$A \in \mathbb{R}^{n \times n}$  is a matrix

$$f: \mathbb{R} \rightarrow \mathbb{R} \quad f(\lambda) = b^T x + \lambda^T A x$$

a) The gradient is

$$\nabla f(\lambda) = \frac{\partial f(\lambda)}{\partial \lambda}$$

$$= \frac{\partial (b^T x + \lambda^T A x)}{\partial \lambda}$$

$$\Rightarrow \nabla f(\lambda) = b + 2A(\lambda)$$

assuming  $A$  is symmetric

$$\boxed{\nabla f(\lambda) = b + 2A\lambda}$$

~~The~~ The Hessian is

$$H(f(\lambda)) = \frac{\partial^2 f(\lambda)}{\partial \lambda^2}$$

$$= \frac{\partial^2 (b^T x + \lambda^T A x)}{\partial \lambda^2}$$

$$= \frac{\partial (b + 2A\lambda)}{\partial \lambda}$$

$$\boxed{H(f(\lambda)) = 2A}$$

(assuming  $A$  is symmetric)

(b) The first order Taylor expansion series is

$$f_1(\lambda) = f(\lambda_0) + \nabla f(\lambda_0)^T (\lambda - \lambda_0)$$

$$= 0 + (b^T + 0)^T (\lambda - 0)$$

$$= 0 + b^T \lambda$$

$$\boxed{f_1(\lambda) = b^T \lambda}$$

For second order Taylor series expansion,

$$f_2(\lambda) = f(\lambda_0) + \nabla f(\lambda_0)^T (\lambda - \lambda_0) + \frac{1}{2} (\lambda - \lambda_0)^T \nabla^2 f(\lambda_0) (\lambda - \lambda_0)$$

$$= 0 + b^T + \frac{1}{2} (1)^T 2A(\lambda)$$

$$= b^T + \frac{1}{2} 2A \lambda$$

$$\boxed{f_2(\lambda) = b^T + \lambda^T A \lambda}$$

→ All Taylor approximations are approximate and not accurate completely because all Hessian, gradient and Taylor expansions are done locally and they are accurate at that local point but not everywhere. But, they are accurate in this problem.

③ given,

$A \in \mathbb{R}^{n \times n}$  is a square matrix.

(a) The necessary and sufficient conditions for  $A$  to be positive definite is all eigen values are positive.

$$\text{i.e. } \lambda_i > 0$$

(b) The necessary and sufficient conditions for  $A$  to have a full rank is all eigen values are non zero.

$$\text{i.e. } \lambda_i \neq 0$$

(c) If  $y \in \mathbb{R}^n$ ,  $y \neq 0 \Rightarrow A^T y = 0$

then  $Ax = b$  will have a solution only if  $y^T b = 0$

See or (2) Question (2)

④ FOR Stinger diet.

Let,

define  $x_i$  = quantity of food type 'i' to be purchased

$c_i$  = unit price of food type

So the problem would be

$$\text{minimize: } \sum_{i=1}^N c_i x_i$$

The constraints considered are to ensure nutrition requirements  $r_j$

$$\sum_{i=1}^N a_{ij} x_i \geq b_j \quad \text{where } j = 1, 2, 3, \dots, M$$

where,

$a_{ij}$  = quantity of nutrition type 'j' in food type 'i'

$b_j$  = required quantity of nutrition type 'j' for a month