CO496 Coursework 1

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1 Differentiation

(a) Expand the completed square form by substituting the variables.

$$Let \ C = \begin{pmatrix} c_{11} & c_{12} \\ c_{21} & c_{22} \end{pmatrix}, \vec{c} = \begin{pmatrix} c_{1} \\ c_{2} \end{pmatrix} \ and \ \vec{x} = \begin{pmatrix} x_{1} \\ x_{2} \end{pmatrix}$$

$$f_{1}(\vec{x}) = (\vec{x} - \vec{c})^{T} C(\vec{x} - \vec{c}) + c_{0}$$

$$= \begin{pmatrix} x_{1} - c_{1} \\ x_{2} - c_{2} \end{pmatrix}^{T} \begin{pmatrix} c_{11} & c_{12} \\ c_{21} & c_{22} \end{pmatrix} \begin{pmatrix} x_{1} - c_{1} \\ x_{2} - c_{2} \end{pmatrix} + c_{0}$$

$$= (x_{1} - c_{1})(c_{11}(x_{1} - c_{1}) + c_{21}(x_{2} - c_{2}))$$

$$+ (x_{2} - c_{2})(c_{12}(x_{1} - c_{1}) + c_{22}(x_{2} - c_{2})) + c_{0}$$

$$= c_{11}(x_{1} - c_{1})^{2} + c_{22}(x_{2} - c_{2})^{2} + (c_{21} + c_{12})(x_{1} - c_{1})(x_{2} - c_{2}) + c_{0}$$

$$= c_{11}(x_{1}^{2} + c_{1}^{2} - 2c_{1}x_{1}) + c_{22}(x_{2}^{2} + c_{2}^{2} - 2c_{2}x_{2})$$

$$+ (c_{12} + c_{21})(x_{1}x_{2} - c_{2}x_{1} - c_{1}x_{2} + c_{1}c_{2}) + c_{0}$$

$$= c_{11}x_{1}^{2} + c_{11}c_{1}^{2} - 2c_{1}c_{11}x_{1} + c_{22}x_{2}^{2} + c_{2}^{2}c_{22} - 2c_{2}c_{22}x_{2}$$

$$+ c_{12}x_{1}x_{2} - c_{2}c_{12}x_{1} - c_{1}c_{12}x_{2} + c_{1}c_{2}c_{21} + c_{0}$$

$$= c_{11}x_{1}^{2} + c_{22}x_{2}^{2} + (c_{12} + c_{21})x_{1}x_{2} + x_{1}(-2c_{1}c_{11} - c_{2}c_{12} - c_{2}c_{21})$$

$$+ x_{2}(-2c_{2}c_{22} - c_{1}c_{12} - c_{1}c_{21}) + c_{1}^{2}c_{11} + c_{2}^{2}c_{22} + c_{1}c_{2}c_{12} + c_{1}c_{2}c_{21} + c_{0}$$

$$(1)$$

Then expand the equation from the original form.

$$f_{1}(\vec{x}) = \vec{x}^{T} \vec{x} + \vec{x}^{T} B \vec{x} - \vec{a}^{T} \vec{x} + \vec{b}^{T} \vec{x}$$

$$= \begin{pmatrix} x_{1} \\ x_{2} \end{pmatrix}^{T} \begin{pmatrix} x_{1} \\ x_{2} \end{pmatrix} + \begin{pmatrix} x_{1} \\ x_{2} \end{pmatrix}^{T} \begin{pmatrix} 3 & -1 \\ -1 & 3 \end{pmatrix} \begin{pmatrix} x_{1} \\ x_{2} \end{pmatrix} - \begin{pmatrix} -1 \\ 0 \end{pmatrix}^{T} \begin{pmatrix} x_{1} \\ x_{2} \end{pmatrix} + \begin{pmatrix} 0 \\ -1 \end{pmatrix}^{T} \begin{pmatrix} x_{1} \\ x_{2} \end{pmatrix}$$

$$= x_{1}^{2} + x_{2}^{2} + x_{1}(3x_{1} - x_{2}) + x_{2}(-x_{1} + 3x_{2}) - x_{1} - x_{2}$$

$$= 4x_{1}^{2} + 4x_{2}^{2} - 2x_{1}x_{2} - x_{1} - x_{2}$$

$$(2)$$

Now, equating the equations (1) and (2) will result in simultaneous equations. Solving them will result in $C = \begin{pmatrix} 4 & -1 \\ -1 & 4 \end{pmatrix}$, $\vec{c} = \frac{1}{6} \begin{pmatrix} 1 \\ 1 \end{pmatrix}$ and $c_0 = -\frac{1}{6}$

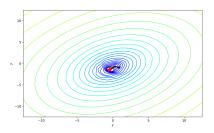


Figure 1: contour of f_2

Figure 2: contour of f_3

(b) Given that the expanded form of $f_1(x)$ is $4x_1^2 + 4x_2^2 - 2x_1x_2 - x_1 - x_2$. If Hessian matrix H of f_1 is positive definite, f_1 has a local minimum. $Hf_1(x_1, x_2) = \begin{pmatrix} \frac{\partial^2 f}{\partial x_1^2} & \frac{\partial^2 f}{\partial x_1 \partial x_2} \\ \frac{\partial^2 f}{\partial x_2 \partial x_1} & \frac{\partial^2 f}{\partial x_2^2} \end{pmatrix} = \begin{pmatrix} 8 & -2 \\ -2 & 8 \end{pmatrix}$

$$Hf_1(x_1, x_2) = \begin{pmatrix} \frac{\partial^2 f}{\partial x_1^2} & \frac{\partial^2 f}{\partial x_1 \partial x_2} \\ \frac{\partial^2 f}{\partial x_2 \partial x_1} & \frac{\partial^2 f}{\partial x_2^2} \end{pmatrix} = \begin{pmatrix} 8 & -2 \\ -2 & 8 \end{pmatrix}$$

If $Hf_1(x_1, x_2)$ is positive definite, the determinant of its characteristic function should give positive eigenvalues only.

$$det(Hf_1(x_1, x_2) - \lambda I) = \begin{vmatrix} 8 - \lambda & -2 \\ -2 & 8 - \lambda \end{vmatrix}$$
$$= (\lambda - 8)^2 - 4$$
$$= \lambda^2 - 16\lambda + 60$$
$$= (\lambda - 6)(\lambda - 10) = 0$$

As such, eigenvalues are 6 or 10 which are both positive. Therefore, $Hf_1(x_1, x_2)$ is positive definite. Thus, the minimum value and its corresponding input can be achieved from the completed square form of $f_1(\vec{x})$.

- ... The minimum value of $f_1 = -\frac{1}{6}$ and the corresponding input $\vec{x} = \frac{1}{6} \begin{pmatrix} 1 \\ 1 \end{pmatrix}$.
- (f) The figure 1 and 2 show the 2d contour plot of f_2 and f_3 respectively. The step size of 0.1 is used to generate the plots.

(g)

- 1) f_2 has only one minima.
- 2) When step size is large, f_2 's gradient, thus the step diverges.
- 3) f_3 has 2 minimas.
- 4) When moderate step size is used, f_3 's step converges to its local minima not the global one.
- 5) When a large step size is used, f_3 's step starts to oscillate within the function and never converges.
- 6) The main difference between f_2 and f_3 is that f_2 diverges far more quickly than f_3 .