

$$1. (a) f(\vec{x}) = 2x_1^2 + x_1x_2 + x_2^2 - 3x_1 - 5x_2$$

$$4 f(\vec{x}) = 8x_1^2 + 4x_1x_2 + 4x_2^2 - 12x_1 - 20x_2$$

$$\geq 6x_1^2 + 2x_2^2 - 12x_1 - 20x_2$$

$$= 6(x_1-1)^2 + 2(x_2-5)^2 - 56$$

$$\geq 2(x_1-1)^2 + 2(x_2-5)^2 - 56$$

$$\geq x_1^2 + x_2^2 - 108$$

Obviously, when $\|\vec{x}\| \rightarrow \infty$, $f(\vec{x}) \rightarrow +\infty$, so $f(\vec{x})$ is coercive.

(b) $f(\vec{x})$ is continuous on \mathbb{R}^2 . Given the conclusion in (a), we know that $f(\vec{x})$ has global min but does not have global max.

The global mins satisfy

$$\nabla f(\vec{x}) = \vec{0}, \text{ i.e. } \begin{cases} 4x_1 + x_2 - 3 = 0 \\ x_1 + 2x_2 - 5 = 0 \end{cases}$$

$\begin{cases} x_1 = \frac{1}{7} \\ x_2 = \frac{17}{7} \end{cases}$ is the only solution, so it must be the global min

Thus, the global min of $f(\vec{x})$ is $(\frac{1}{7}, \frac{17}{7})$

2. (a) When $\alpha \in \mathbb{R} \rightarrow +\infty$, $f(\alpha \vec{w}_0) \rightarrow 0$, so $\forall \varepsilon > 0$,
 $\exists A > 0$, $\forall \alpha > A$, $|f(\alpha \vec{w}_0)| < \varepsilon$.

Since $f(\vec{w}) > 0$, $f(\vec{w})$ cannot have a global min.

$$\begin{aligned} \text{(b) (i)} \quad f(\vec{w}) &\geq \log(1 + e^{-\eta_i \vec{x}_i^T \vec{w}}) \\ &\geq \log e^{-\eta_i \vec{x}_i^T \vec{w}} \\ &= -\eta_i \vec{x}_i^T \vec{w} \end{aligned}$$

Thus, $f(\vec{w}) \geq h(\vec{w})$

(ii) $h(\vec{w})$ is continuous on S , which is a compact set,
so $h(\vec{w})$ has a global min \vec{w}_0 on S .

If $h(\vec{w}_0) \leq 0$, then $\forall i$, $-\eta_i \vec{x}_i^T \vec{w}_0 \leq 0$, which
contradicts the fact that the dataset is not linearly
separable. Thus, $C \triangleq h(\vec{w}_0) > 0$

(iii) Let $\vec{w}_1 = \frac{\vec{w}}{\|\vec{w}\|}$. Then $\vec{w}_1 \in S$.

$$h(\vec{w}) = h(\|\vec{w}\| \vec{w}_1) = \|\vec{w}\| h(\vec{w}_1) \geq C \|\vec{w}\|.$$

(iv) $f(\vec{w}) \geq C \|\vec{w}\|$, so $f(\vec{w})$ is coercive. We also
know that $f(\vec{w})$ is continuous, so $f(\vec{w})$ has a
global min.

(c) Let $g(x) = \log(1 + e^x)$, $h_i(\vec{w}) = -\eta_i \vec{x}_i^T \vec{w}$.

$$\text{Then } [g(h_i(\vec{w}))]' = g'(h_i(\vec{w})) \cdot h_i'(\vec{w})$$

$$= - \frac{\eta_i e^{-\eta_i \vec{x}_i^T \vec{w}}}{e^{-\eta_i \vec{x}_i^T \vec{w}} + 1} \vec{x}_i^T$$

$$\text{Thus, } \nabla f(\vec{w}) = f'(\vec{w})^T = - \sum_{i=1}^m \frac{\eta_i e^{-\eta_i \vec{x}_i^T \vec{w}}}{e^{-\eta_i \vec{x}_i^T \vec{w}} + 1} \vec{x}_i$$

3. (a) We already know that

$$g(a+t) = g(a) + g'(a)t + \frac{1}{2} g''(a+\theta t) t^2, \quad \theta \in (0,1)$$

$$\text{Let } \hat{d} = \frac{\vec{d}}{\|\vec{d}\|}.$$

Applying the formula to $g(0+t) = f(\vec{x} + t\hat{d})$ will obtain

$$f(\vec{x} + t\hat{d}) = g(0) + g'(0)t + \frac{1}{2} g''(\theta t) t^2$$

Because $g'(t) = f'(\vec{x} + t\hat{d}) \hat{d}$, $g''(t) = \hat{d}^T \nabla^2 f(\vec{x} + t\hat{d}) \hat{d}$,
when $t = \|\vec{d}\|$ we have:

$$f(\vec{x} + \vec{d}) = f(\vec{x}) + \nabla f(\vec{x} + \vec{d}) \vec{d} + \frac{1}{2} \vec{d}^T \nabla^2 f(\vec{x} + \theta \vec{d}) \vec{d},$$

for some $\theta \in (0,1)$

(b) Let $\vec{g}(t) = \nabla f(\vec{x} + t\vec{d})$. Then $\vec{g}'(t) = \nabla^2 f(\vec{x} + t\vec{d}) \cdot \vec{d}$

$$\int_0^1 \vec{g}'(t) dt = \vec{g}(1) - \vec{g}(0) = \nabla f(\vec{x} + \vec{d}) - \nabla f(\vec{x})$$

$$\text{Thus, } \nabla f(\vec{x} + \vec{d}) = \nabla f(\vec{x}) + \int_0^1 \nabla^2 f(\vec{x} + t\vec{d}) \vec{d} dt$$

4. $A = \begin{pmatrix} 6 & 2 & 0 \\ 2 & 5 & -2 \\ 0 & -2 & 4 \end{pmatrix}$

$6 > 0, \begin{vmatrix} 6 & 2 \\ 2 & 5 \end{vmatrix} = 26 > 0, |A| = 80 > 0,$

so A is positive definite

$B = \begin{pmatrix} 1 & 2 & 0 \\ 2 & 2 & 1 \\ 0 & 1 & -3 \end{pmatrix}$

$1 > 0, \begin{vmatrix} 1 & 2 \\ 2 & 2 \end{vmatrix} = -2 < 0$, so B is indefinite.

$C = \begin{pmatrix} 2 & -1 & -1 \\ -1 & 2 & -1 \\ -1 & -1 & 2 \end{pmatrix}$

$2 > 0, \begin{vmatrix} 2 & -1 \\ -1 & 2 \end{vmatrix} = 2 > 0, |C| = 0$

All principal minors are non-negative,
so C is positive semidefinite.