

1) (a)

$$(i) f(x) = e^x - 1, \nabla^2 f(x) = e^x > 0 \text{ for all } x.$$

Also, $f(x)$ is twice continuously differentiable $\Rightarrow f(x)$ is strictly convex and also strictly concave[#]

$$(ii) \nabla^2 f(x) = \begin{bmatrix} 0 & 1 \\ 1 & 0 \end{bmatrix} \Rightarrow \text{neither PSD or NSD.} \Rightarrow \text{neither convex nor concave} \#$$

It's strictly concave[#] because $\{(x_1, x_2) \in \mathbb{R}_+^2 \mid x_1 x_2 \geq \alpha\}$ are convex. It's not strictly convex[#]

$$(iii) \nabla^2 f(x) = \begin{bmatrix} \alpha(\alpha-1)x_1^{\alpha-2}x_2^{1-\alpha} & \alpha(\alpha-1)x_1^{\alpha-1}x_2^{-\alpha} \\ \alpha(\alpha-1)x_1^{\alpha-1}x_2^{-\alpha} & (1-\alpha)(-\alpha)x_1^{\alpha}x_2^{-\alpha-1} \end{bmatrix} = -\alpha(1-\alpha)x_1^{\alpha}x_2^{1-\alpha} \begin{bmatrix} x_1^{-1} \\ -x_2^{-1} \end{bmatrix} \begin{bmatrix} x_1^{-1} \\ -x_2^{-1} \end{bmatrix}^T$$

$$\Rightarrow -\nabla^2 f(x) \succeq 0 \Rightarrow f \text{ is concave.}$$

(b)

$$(i) J(\theta) = \frac{1}{2} (A^T \theta - y)^T (A^T \theta - y) = \frac{1}{2} (\theta^T A A^T \theta - \theta^T A y - y^T A^T \theta + y^T y).$$

$$\nabla_{\theta} J(\theta) = A A^T \theta - A y \# \quad \theta^{(t+1)} = \theta^{(t)} - \alpha (A A^T \theta - A y) \#$$

(ii)

$$\text{To minimize } J \Rightarrow \nabla_{\theta} J(\theta) = 0 \Rightarrow A A^T \theta = A y \Rightarrow \theta = (A^T)^{-1} y \#$$

2)

N = numbers of skip.

W = Win.

We know $P(W|N=0) = \frac{1}{2}$, now we assume k Rose Gold have been skipped.

$$\Rightarrow P(W|N=n) = \frac{50-k}{100-n}.$$

$$P(W|N=n+1) = \frac{50-k}{100-(n+1)} \cdot \frac{100-n-(50-k)}{100-n} + \frac{49-k}{100-(n+1)} \cdot \frac{50-k}{100-n} = \frac{50-k}{100-n}.$$

$$\Rightarrow P(W|N=n) = P(W|N=n+1) = P(W|N=0) = \frac{1}{2} \#$$

\therefore The probability of correct prediction is always $\frac{1}{2}$ regardless of N .