$$|H-\chi_{1}| = -\chi \left(12\chi^{2}+2^{-\chi}\right) - 1 = 0 = 7 \chi^{2} - \left(12\chi^{2}+2\right)\chi - 1 = 0$$

$$\chi = \frac{12\chi^{2}+2}{\sqrt{(2\chi^{2}+2)^{2}-4\chi(-4)}} = \frac{12\chi^{2}+2}{\sqrt{(2\chi^{2}+2)^{2}+4}}$$

$$\frac{2}{\sqrt{(2\chi^{2}+2)^{2}+4}} = \frac{12\chi^{2}+2}{\sqrt{(2\chi^{2}+2)^{2}-4\chi(-4)}} = \frac{12\chi^{2}+2}{\sqrt{(2\chi^{2}+2)^{2}+4}}$$

$$\frac{2}{\sqrt{(2\chi^{2}+2)^{2}+4}} = \frac{12\chi^{2}+2}{\sqrt{(2\chi^{2}+2)^{2}-4\chi(-4)}} = \frac{12\chi^{2}+2}{\sqrt{(2\chi^{2}+2)^{2}+4}}$$

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$$\frac{12\chi^{2}+2}{\sqrt{(2\chi^{2}+2)^{2}+4}} = \frac{12\chi^{2}+2}{\sqrt{(2\chi^{2}+2)^{2}+4}} = \frac{12\chi^{2$$

CONTINUED ---

 $\frac{\partial^2 f}{\partial x^2} = 12x^2 + 2 \qquad ; \qquad \frac{\partial^2 f}{\partial y^2} = 1 = \frac{\partial^2 f}{\partial y^3 x}$ 

 $|H^{-}| |12x^{2+2}| | |1 - \lambda I = [12x^{2} + 2 - \lambda 1]$ 

 $2)a)f(x,y) = x^{4} + xy + x^{2}$ 

2f = 4713 + 4+272; 2f = \$1200

... the matrix is not PSD

consider the point 
$$(1, -8)$$
At this pt, the Hessian is  $[12x1^2+2] = [14]$ 

Hence, the Hessian is not PSD at (1, -8)

This matrix is not PSD if 
$$\frac{1}{2} V \in \mathbb{R}^2 | V^T H V \approx 0$$
  
consider  $V = \begin{bmatrix} 1 \\ 1 \end{bmatrix}$ 

This matrix is not PSD if 
$$\frac{1}{2}$$
  $V \in \mathbb{R}^2$   $\left[ V^T + V \approx 0 \right]$   
consider  $V = \begin{bmatrix} 1 \\ -8 \end{bmatrix}$   
 $V^T + V = \begin{bmatrix} 1 \\ -8 \end{bmatrix} \begin{bmatrix} 14 \\ 10 \end{bmatrix} \begin{bmatrix} 1 \\ -8 \end{bmatrix} = \begin{bmatrix} 1 \\ -8 \end{bmatrix} \begin{bmatrix} 6 \\ 1 \end{bmatrix} = 6 - 8 = -2 < 0$ 

Proof Prove Hessian 
$$\left(\frac{1}{2n}\left(x^{T}w-y\right)^{T}\left(x^{T}w-y\right)\right)$$
 is PSD

From HWL  $\nabla x\left(Ax+b\right)^{T}\left(Ax+b\right) = A^{T}(Ax+b)$ 

$$\therefore \nabla x\left(Ax+b\right)^{T}\left(Ax+b\right) = A^{T}(Ax+b)$$

$$= \frac{1}{2n}\left(x^{T}\right)^{T}\left(x^{T}w-y\right)$$

$$= \frac{1}{2n}\left(x^{T}w-y\right)$$

$$= \frac{1}{2n}\left(x^{T}w-xy\right)$$

$$= \frac{1}{2n}\left(x^{T}w-xy\right)$$
Now we must show that  $1 \times x^{T}$  is PSD

We can ignore the  $\frac{1}{2n}$  as it is simply a tre scalar

If  $x \times x^{T}$  is PSD then

 $V^{T}(XX^{T})V \geqslant 0 \quad \forall V \in \mathbb{R}^{2n} \text{ it } XX^{T} \text{ in an } N \times n \text{ Cohr matrix}$   $= (V^{T}X)(X^{T}V) = (X^{T}V)^{T}(X^{T}V) \geqslant 0$ 

XTV is an nx 1 vector

:, (x TV) is 1 x n

. In  $K^{T}$ ,  $(X^{T}V)^{T}(X^{T}V)$  is the sum of squared of the elements of  $X^{T}V$  which are always >, 0 assuming  $K^{T}X^{T}V \in IR^{n}$ 

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Hence froved

$$\sigma(x) = \frac{1}{1 + e^{-x}}$$

(a) 
$$\sigma(-\pi) = 1 - \sigma(\pi)$$

Soln:

$$\sigma(-\pi) = \frac{1}{1 + e^{\pi}} = \frac{1}{1 + e^{\pi}}$$

$$= \frac{1+e^{x}-e^{x}}{1+e^{x}} \qquad (+-e^{x} \text{ in the numerator})$$

$$= \frac{1+e^{\chi}}{1+e^{\chi}} - \frac{e^{\chi}}{1+e^{\chi}}$$

$$= 1 - \frac{1}{e^{-\varkappa(1+e^{\varkappa})}}$$

$$=$$
  $1-\frac{1}{e^{-\varkappa_{+}}e^{\varkappa_{-}\varkappa_{-}}}$ 

$$= 1 - \frac{1}{1 + e^{-\chi}} = \frac{1 - o(\chi)}{1 - o(\chi)}$$

$$fhus, \quad \sigma(-n) = 1 - \sigma(n)$$

Hence proved.

(b) 
$$\sigma'(n) = \frac{\partial \sigma}{\partial x}(n) = \sigma(n)(1-\sigma(n))$$
.  
801)  $\sigma(n) = \frac{1}{1+e^{-n}}$  (let  $1+e^{-n} = f(n)$ )  $-(D)$ 

$$\frac{\partial \sigma}{\partial x} = \frac{\partial}{\partial x} \left[\frac{1}{1+e^{-n}}\right]$$

$$= -\frac{1}{1+e^{-n}} \cdot \int_{-1}^{1} f(x) dx$$

$$= -\frac{1}{1+e^{-n}} \cdot (-e^{-n}) - unng(D)$$

$$= \frac{+e^{-n}+1}{(1+e^{-n})^2}$$

$$= \frac{1}{1+e^{-n}} - \frac{1}{(1+e^{-n})^2} = \frac{1}{1+e^{-n}} \left(1 - \frac{1}{1+e^{-n}}\right)$$

Fince 
$$\sigma(n)$$
 has  $n$  as the only variable,  $\sigma(n) = \frac{\partial \sigma(n)}{\partial n}$ .

Thus,  $\sigma'(n) = \frac{\partial (\sigma(n))}{\partial n} = \sigma(n)(1 - \sigma(n))$ .

Hence proved.

= o(n) (1-o(n))

## **Question 3**

Reported Fmse on test and validation set

Optimal hyperparameters after testing on the validation set: [50, 0.001, 500, 0.01]

```
Mini_batch_size = 50
Learning_rate = 0.001
Epochs = 500
Alpha = 0.01
```

Optimal weights list after training - .txt file included in the submission