## Problem 7.5

Faisal Memon

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## 1 Problem

Calculate the derivative  $\frac{\partial l_i}{\partial f[m{x}_i, m{\phi}]}$  for binary classification loss function

$$l_i = -1(1-y_i)\log[1-sig[f[\boldsymbol{x}_i,\boldsymbol{\phi}]]] - y_i\log[sig[f[\boldsymbol{x}_i,\boldsymbol{\phi}]]]$$
 With  $sig[z] = \frac{1}{1+\exp[-z]}$ 

## 2 Answer

$$\begin{split} l_i &= -1(1-y_i)\log[1-sig[f[\boldsymbol{x}_i,\boldsymbol{\phi}]]] - y_i\log[sig[f[\boldsymbol{x}_i,\boldsymbol{\phi}]]] \\ l_i &= (y_i-1)\log[1-sig[f[\boldsymbol{x}_i,\boldsymbol{\phi}]]] - y_i\log[sig[f[\boldsymbol{x}_i,\boldsymbol{\phi}]]] \\ l_i &= a\cdot b - c\cdot d \end{split}$$
 Where 
$$a &= (y_i-1)$$

$$b = \log[1 - sig[f[\mathbf{x}_i, \boldsymbol{\phi}]]]$$

$$c = y_i$$

$$d = \log[sig[f[\mathbf{x}_i, \boldsymbol{\phi}]]]$$

Using Chain Rule twice

$$\frac{\partial l_i}{\partial f[\boldsymbol{x}_i, \boldsymbol{\phi}]} = a \cdot \frac{\partial b}{\partial f[\boldsymbol{x}_i, \boldsymbol{\phi}]} + b \cdot \frac{\partial a}{\partial f[\boldsymbol{x}_i, \boldsymbol{\phi}]} - c \cdot \frac{\partial d}{\partial f[\boldsymbol{x}_i, \boldsymbol{\phi}]} - d \cdot \frac{\partial c}{\partial f[\boldsymbol{x}_i, \boldsymbol{\phi}]}$$

Using SymPy

Therefore

$$\begin{split} \frac{\partial l_i}{\partial f[\boldsymbol{x}_i, \boldsymbol{\phi}]} &= (y_i - 1) \cdot -\frac{e^{-x}}{\left(1 - \frac{1}{1 + e^{-x}}\right) (1 + e^{-x})^2} + \log[1 - sig[f[\boldsymbol{x}_i, \boldsymbol{\phi}]]] \cdot 0 - y_i \cdot \frac{e^{-x}}{1 + e^{-x}} - \log[sig[f[\boldsymbol{x}_i, \boldsymbol{\phi}]]] \cdot 0 \\ &= (y_i - 1) \cdot -\frac{e^{-x}}{\left(1 - \frac{1}{1 + e^{-x}}\right) (1 + e^{-x})^2} - y_i \cdot \frac{e^{-x}}{1 + e^{-x}} \end{split}$$