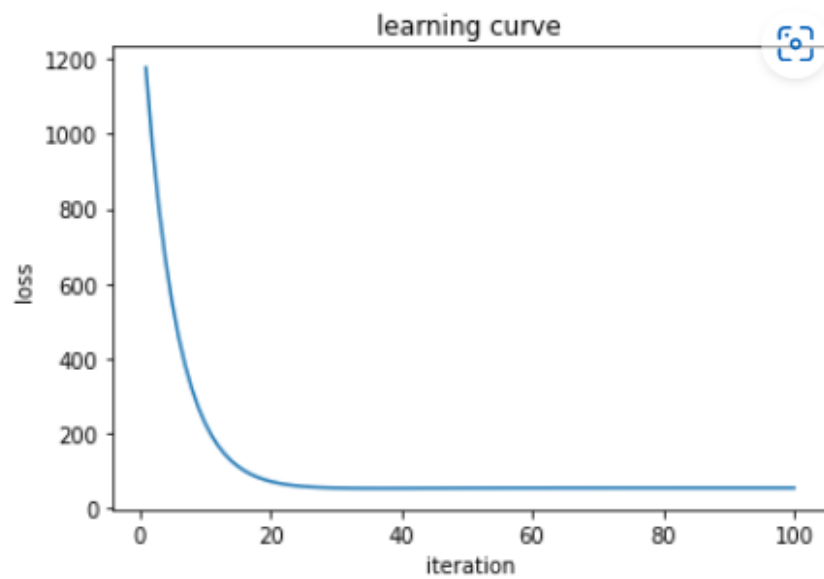


Part. 1

Linear regression model

1.

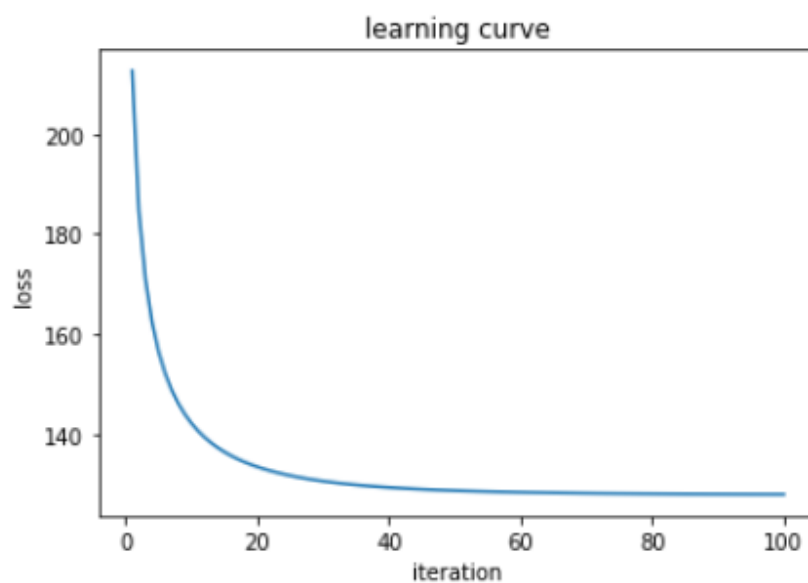


2+3.

Mean Square Error: 55.21442307485028
weights: 52.74050648059757
intercept: -0.33418947537034915

Logistic regression model

1.

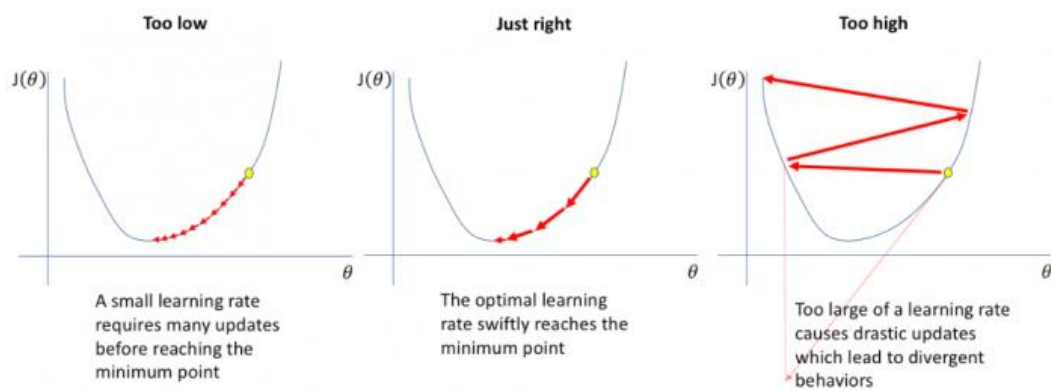


2+3.

```
Cross Entropy Error: 46.75820789486666  
weights: 4.7294759750421145  
intercept: 1.621861070934789
```

Part. 2

1. Gradient Descent 是看過整筆資料過後再 update 一次 weight 和 bias； Mini-Batch Gradient Descent 是會先決定一個 batch size，接著每看完一個 batch 就 update 一次；Stochastic Gradient Descent 則是每看過一個點都會直接 update 一次參數。也就是說，這三者的差異主要在 update 參數的頻率。
2. 會，learning rate 要設剛好才會 converge。首先如果太大的話，update 參數的級距會太大，造成做 training 的時候一直來回震盪(一直 update 過頭)；如果太小，則可能讓 training 卡住(因為每次只 update 一點點)，或是訓練太慢。



3.

$$\begin{aligned}
 3. \quad \sigma(u) + \sigma(-u) &= \frac{1}{1+e^{-u}} + \frac{1}{1+e^u} \\
 &= \left(\frac{1+e^u}{1+e^u} \right) \cdot \frac{1}{1+e^{-u}} + \left(\frac{1+e^{-u}}{1+e^{-u}} \right) \cdot \frac{1}{1+e^u} \\
 &= \frac{2+e^u+e^{-u}}{(1+e^{-u})(1+e^u)} \\
 &= \frac{2+e^u+e^{-u}}{1+e^u+e^u+e^0} = \frac{2+e^u+e^{-u}}{2+e^u+e^{-u}} = 1 \\
 &\Rightarrow \sigma(-u) = 1 - \sigma(u) \\
 \textcircled{2} \quad \text{let } y &= \frac{1}{1+e^{-x}} \quad (y = \sigma(x)) \\
 \frac{1}{y} &= 1+e^{-x} \\
 e^{-x} &= \frac{1}{y} - 1 = \frac{1-y}{y} \\
 \ln e^{-x} &= \ln \frac{1-y}{y} \\
 -x &= \ln \frac{1-y}{y} \\
 x &= -\ln \frac{1-y}{y} = \ln \frac{y}{1-y} \quad \Rightarrow \quad \sigma^{-1}(y) = \ln \frac{y}{1-y}
 \end{aligned}$$

4.

$$\begin{aligned}
 4. \quad a_{nj} &= w_j^T \phi_n \\
 \nabla_{w_j} a_{nj} &= \phi_n \\
 \frac{\partial \mathcal{E}}{\partial y_{nk}} &= -\frac{t_{nk}}{y_{nk}} \quad \frac{\partial y_{nk}}{\partial a_{nj}} = y_{nk}(I_{kj} - y_{nj}) \\
 \frac{\partial \mathcal{E}}{\partial a_{nj}} &= \sum_{k=1}^K \frac{\partial \mathcal{E}}{\partial y_{nk}} \frac{\partial y_{nk}}{\partial a_{nj}} = -\sum_{k=1}^K \frac{t_{nk}}{y_{nk}} y_{nk}(I_{kj} - y_{nj}) \\
 &= -\sum_{k=1}^K t_{nk}(I_{kj} - y_{nj}) = -t_{nj} + \sum_{k=1}^K t_{nk} y_{nj} = y_{nj} - t_{nj} \\
 \nabla_{w_j} \mathcal{E}(w_1, \dots, w_K) &= \sum_{n=1}^N \frac{\partial \mathcal{E}}{\partial a_{nj}} \nabla_{w_j} a_{nj} = \sum_{n=1}^N (y_{nj} - t_{nj}) \phi_n
 \end{aligned}$$