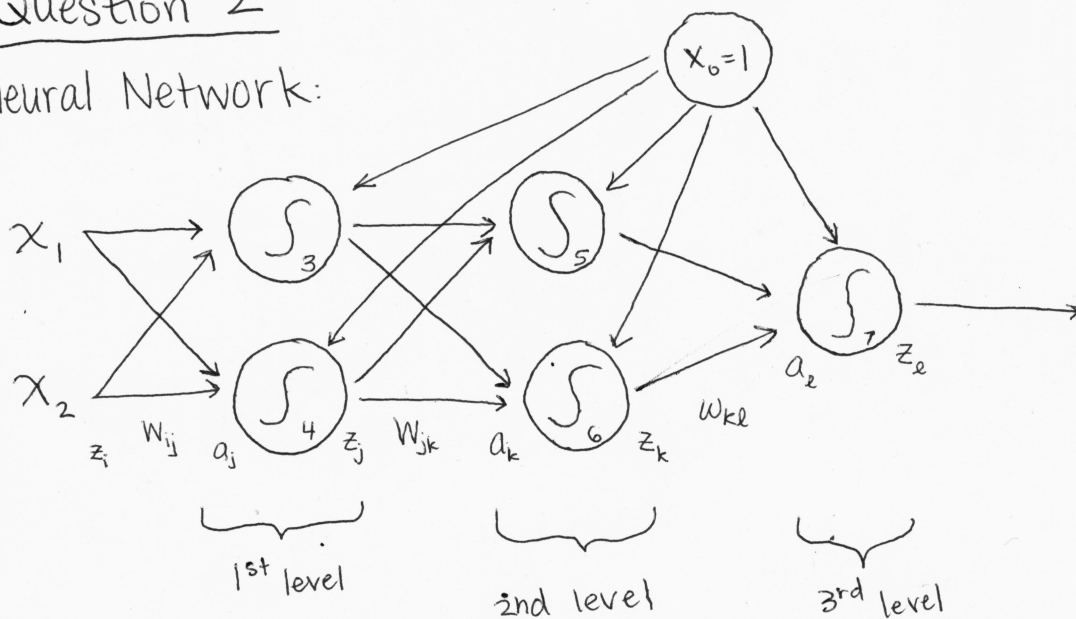


Question 2

Neural Network:



[gradient descent w/ logistic loss]

Logistic Loss function:

$$L(y, f(\vec{x}; \theta)) = (y-1) \log(1-f(\vec{x}; \theta)) - y \log(f(\vec{x}; \theta))$$

where $f(\vec{x}; \theta) = \frac{1}{1 + \exp(-\theta \vec{x})}$

where $f(x^n; \theta) = \frac{1}{1 + \exp(-\theta x^n)}$

$$\begin{aligned} \frac{\partial R}{\partial w_{kl}} &= \frac{1}{N} \sum_n \left[\frac{\partial L^n}{\partial a_l^n} \right] \left(\frac{\partial a_l^n}{\partial w_{kl}} \right) = \frac{1}{N} \sum_n \left[\frac{\partial (y-1) \log(1-f(a_l^n; \theta)) - y \log(f(a_l^n; \theta))}{\partial a_l^n} \right] (z_k^n) \\ &= \frac{1}{N} \sum_n \left[\frac{(y-1) \cdot f'(a_l^n; \theta)}{1-f(a_l^n; \theta)} - \frac{y \cdot f'(a_l^n; \theta)}{f(a_l^n; \theta)} \right] (z_k^n) \quad \text{where } f'(a_l^n; \theta) = \frac{\theta \exp(\theta a_l^n)}{\exp(\theta a_l^n) + 1}^2 \\ &\quad \text{define as } \delta_l^n \quad \& \quad f(a_l^n; \theta) = z_l^n \\ &= \frac{\sum_n \delta_l^n z_k^n}{N} \quad (\text{output layer derivative}) \end{aligned}$$

Cost function: $= \frac{1}{N} \sum_n (y-1) \log(1-f(\sum_k w_{kl} f(\sum_j w_{jk} f(\sum_i w_{ij} x_i^n)))) - y \log(f(\sum_k w_{kl} f(\sum_j w_{jk} f(\sum_i w_{ij} x_i^n))))$

$$\begin{aligned} \frac{\partial R}{\partial w_{jk}} &= \frac{1}{N} \sum_n \left[\frac{\partial L^n}{\partial a_k^n} \right] \left(\frac{\partial a_k^n}{\partial w_{jk}} \right) = \frac{1}{N} \sum_n \left[\sum_l \frac{\partial L^n}{\partial a_l^n} \frac{\partial a_l^n}{\partial a_k^n} \right] \left(\frac{\partial a_k^n}{\partial w_{jk}} \right) = \frac{1}{N} \sum_n \left[\sum_l \delta_l^n \frac{\partial a_l^n}{\partial a_k^n} \right] (z_j^n) \\ &= \frac{1}{N} \sum_n \left[\sum_l \delta_l^n w_{kl} f'(a_k^n; \theta) \right] z_j^n \quad (\text{where } f' \text{ as before}) \\ &\quad \text{define as } \delta_k^n \\ &= \frac{1}{N} \sum_n \delta_k^n z_j^n \quad (\text{middle layer derivative}) \end{aligned}$$