

Task 1

- \vec{y} has dimensions of $p \times 1$. Each row represents a probability
- The model will output equal probabilities for all classes

Task 2

- elementwise exponentiation of input
 - sum over all elements, keepdims ensures that resulting sum has same dims as input
 - calculate class probs

$$b) \text{softmax}(\vec{z}) = \frac{e^z}{\sum_{j=1}^n e^{z_j}} ; = \text{softmax}(\vec{z} - \alpha \cdot (1, \dots, 1)^T) = \frac{e^{(z-\alpha)}}{\sum_{j=1}^n e^{(z_j-\alpha)}}$$

- overflow for large numbers
 - underflow " small "
- z.B. subtract max value from all elements

Task 3

- When the model predicts the correct true class with probability 1

$$b) L^{(i)} = - \sum_{j=1}^n y_j^{(i)} \cdot \log(\hat{y}_j^{(i)}) \quad \frac{\partial L^{(i)}}{\partial \hat{y}_j^{(i)}} = \hat{y}_j^{(i)} - y_j^{(i)}$$

$$\frac{\partial L^{(i)}}{\partial \hat{y}_j^{(i)}} = - \frac{y_j^{(i)}}{\hat{y}_j^{(i)}}$$

$$\frac{\partial \hat{y}_j^{(i)}}{\partial \theta_j^{(i)}} = \hat{y}_j^{(i)} (1 - \hat{y}_j^{(i)})$$

$$\frac{\partial \theta_j^{(i)}}{\partial \theta_j^{(i)}} = 1$$

$$\frac{\partial L^{(i)}}{\partial \theta_j^{(i)}} = \frac{\partial L^{(i)}}{\partial \hat{y}_j^{(i)}} \cdot \frac{\partial \hat{y}_j^{(i)}}{\partial \theta_j^{(i)}}$$

$$= - \frac{y_j^{(i)}}{\hat{y}_j^{(i)}} \cdot \frac{\partial \hat{y}_j^{(i)}}{\partial \theta_j^{(i)}}$$

$$= - \frac{y_j^{(i)}}{\hat{y}_j^{(i)}} \cdot \hat{y}_j^{(i)} (1 - \hat{y}_j^{(i)})$$

$$= -x_j^{(i)} (1 - x_j^{(i)})$$

$$= x_j^{(0)} - x_j^{(i)}$$