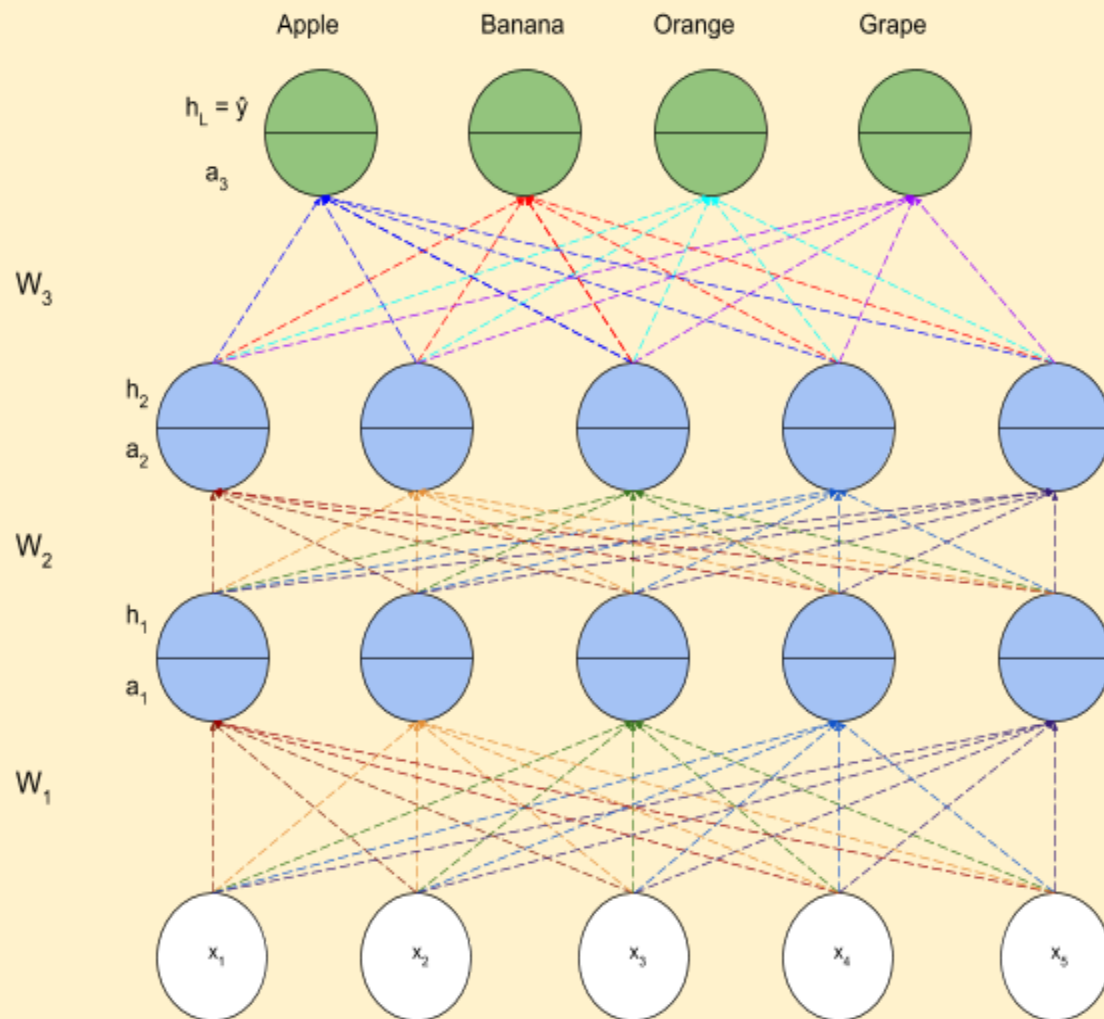


### Output Layer of a Multi-Class Classification Problem

Deciding the output layer

1. The Output Activation function is chosen depending on the task at hand (can be a softmax, linear etc)
2. Consider the following multi-class classification problem



3. At the last layer, we compute  $a_3 = W_3 h_2 + b_3$
4. We need to apply a function to  $\hat{y}_i = O(a_{3i})$  such that the 4 outputs form a probability distribution.
5. Output activation function has to be chosen such that the output is probability.
6. Let's assume  $a_3 = [3 \ 4 \ 10 \ 3]$ 
  - a. Take each entry and divide by the sum of all entries to get a probability distribution
  - b.  $\hat{y}_1 = \frac{3}{(3+4+10+3)} = 0.15$
  - c.  $\hat{y}_2 = \frac{4}{(3+4+10+3)} = 0.20$
  - d.  $\hat{y}_3 = \frac{10}{(3+4+10+3)} = 0.50$
  - e.  $\hat{y}_4 = \frac{3}{(3+4+10+3)} = 0.15$
  - f. However, **this will not work** if any of the entries are negative

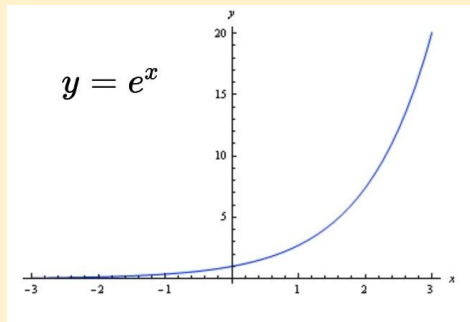
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7. So we consider the softmax function

8.  $\text{softmax}(z_i) = \frac{e^{z_i}}{\sum_{j=1}^k e^{z_j}}$  for  $i = 1 \dots k$

9. Note: the output of  $e^x$  is always positive, irrespective of the input



10. This property is important to counter the negative-value shortcoming that we observed in the previous example

11. Now, let us illustrate the softmax function at the last layer of our Neural Network

12.  $a = [a_1 \ a_2 \ a_3 \ a_4]$

13.  $\text{softmax}(a) = \left[ \frac{e^{a_1}}{\sum_{j=1}^k e^{a_j}} \ \frac{e^{a_2}}{\sum_{j=1}^k e^{a_j}} \ \frac{e^{a_3}}{\sum_{j=1}^k e^{a_j}} \ \frac{e^{a_4}}{\sum_{j=1}^k e^{a_j}} \right]$

14. Raising the numerators to  $e^x$  ensures that they are all positive

15. The denominator is just the sum of all the values raised to  $e^x$

16.  $\text{softmax}(a_i)$  is the  $i^{\text{th}}$  element of the softmax output

17. So for our multi-class fruit classifier, the equations are as follows

Layer	Pre-activation	Activation/Output
Hidden Layer 1	$a_1 = W_1 * x + b_1$	$h_1 = g(a_1)$
Hidden Layer 2	$a_2 = W_2 * h_1 + b_2$	$h_2 = g(a_2)$
Output Layer	$a_3 = W_3 * h_2 + b_3$	$\hat{y} = \text{softmax}(a_3)$