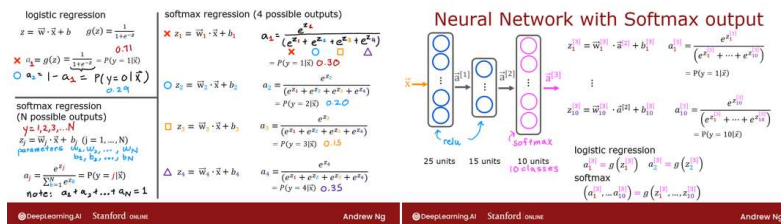


## Optional Lab - Softmax Function

In this lab, we will explore the softmax function. This function is used in both Softmax Regression and in Neural Networks when solving Multiclass Classification problems.

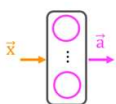


```
In [1]: import numpy as np
import matplotlib.pyplot as plt
plt.style.use('./deeplearning.mplstyle')
import tensorflow as tf
from tensorflow.keras.models import Sequential
from tensorflow.keras.layers import Dense
from IPython.display import display, Markdown, Latex
from sklearn.datasets import make_blobs
%matplotlib widget
from matplotlib.widgets import Slider
from lab_utils_common import dlc
from lab_utils_softmax import plt_softmax
import logging
logging.getLogger("tensorflow").setLevel(logging.ERROR)
tf.autograph.set_verbosity(0)
```

**Note:** Normally, in this course, the notebooks use the convention of starting counts with 0 and ending with N-1,  $\sum_{i=0}^{N-1}$ , while lectures start with 1 and end with N,  $\sum_{i=1}^N$ . This is because code will typically start iteration with 0 while in lecture, counting 1 to N leads to cleaner, more succinct equations. This notebook has more equations than is typical for a lab and thus will break with the convention and will count 1 to N.

## Softmax Function

In both softmax regression and neural networks with Softmax outputs,  $N$  outputs are generated and one output is selected as the predicted category. In both cases a vector  $\mathbf{z}$  is generated by a linear function which is applied to a softmax function. The softmax function converts  $\mathbf{z}$  into a probability distribution as described below. After applying softmax, each output will be between 0 and 1 and the outputs will add to 1, so that they can be interpreted as probabilities. The larger inputs will correspond to larger output probabilities.



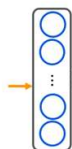
$$\begin{aligned} z_1 &= \vec{w}_1 \cdot \vec{x} + b_1 \\ &\vdots \\ z_N &= \vec{w}_N \cdot \vec{x} + b_N \end{aligned}$$

## Softmax Function

$$a_1 = \frac{e^{z_1}}{(e^{z_1} + \dots + e^{z_N})} = P(y = 1|\vec{x})$$

$$a_N = \frac{e^{z_N}}{(e^{z_1} + \dots + e^{z_N})} = P(y = N|\vec{x})$$

## Neural Network with Softmax Output



$$\begin{aligned} z_1^{[3]} &= \bar{w}_1^{[3]} \cdot \vec{a}^{[2]} + b_1^{[3]} \\ &\vdots \\ z_N^{[3]} &= \bar{w}_N^{[3]} \cdot \vec{a}^{[2]} + b_N^{[3]} \end{aligned}$$

### Softmax Function

$$a_1^{[3]} = \frac{e^{z_1^{[3]}}}{(e^{z_1^{[3]}} + \dots + e^{z_N^{[3]}})} = P(y = 1|\vec{x})$$

$$a_N^{[3]} = \frac{e^{z_N^{[3]}}}{(e^{z_1^{[3]}} + \dots + e^{z_N^{[3]}})} = P(y = N|\vec{x})$$

The softmax function can be written:

$$a_j = \frac{e^{z_j}}{\sum_{k=1}^N e^{z_k}} \quad (1)$$

The output **a** is a vector of length N, so for softmax regression, you could also write:

$$\mathbf{a}(x) = \begin{bmatrix} P(y = 1 | \mathbf{x}; \mathbf{w}, b) \\ \vdots \\ P(y = N | \mathbf{x}; \mathbf{w}, b) \end{bmatrix} = \frac{1}{\sum_{k=1}^N e^{z_k}} \begin{bmatrix} e^{z_1} \\ \vdots \\ e^{z_N} \end{bmatrix} \quad (2)$$