

```
## Load libraries
import numpy as np
import sys
import matplotlib.pyplot as plt
import matplotlib.cm as cm
plt.style.use('dark_background')
%matplotlib inline
```

```
pip install tensorflow
```

```
Collecting tensorflow
  Obtaining dependency information for tensorflow from https://files.pythonhosted.org/packages/8c/ca/c3f014272924e690c34fd2698dcd877ab2d881fe4196cc461a98d2b4693/tensorflow-2.15.0-cp39-cp39-win\_amd64.whl.metadata
  Downloading tensorflow-2.15.0-cp39-cp39-win_amd64.whl.metadata (3.6 kB)
Collecting tensorflow-intel==2.15.0 (from tensorflow)
  Obtaining dependency information for tensorflow-intel==2.15.0 from https://files.pythonhosted.org/packages/1b/86/cda55138387a291fb684691d159742e6196daad571f01de60f809e266d40/tensorflow\_intel-2.15.0-cp39-cp39-win\_amd64.whl.metadata
  Downloading tensorflow_intel-2.15.0-cp39-cp39-win_amd64.whl.metadata (5.1 kB)
Collecting absl-py==1.0.0 (from tensorflow-intel==2.15.0->tensorflow)
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  Downloading absl_py-2.0.0-py3-none-any.whl.metadata (2.3 kB)
Collecting astunparse==1.6.0 (from tensorflow-intel==2.15.0->tensorflow)
  Downloading astunparse-1.6.3-py2.py3-none-any.whl (12 kB)
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  Downloading flatbuffers-23.5.26-py2.py3-none-any.whl.metadata (850 bytes)
Collecting gast==0.5.0,!=0.5.1,!=0.5.2,>=0.2.1 (from tensorflow-intel==2.15.0->tensorflow)
  Downloading gast-0.5.4-py3-none-any.whl (19 kB)
Collecting google-pasta==0.1.1 (from tensorflow-intel==2.15.0->tensorflow)
  Downloading google_pasta-0.2.0-py3-none-any.whl (57 kB)
----- 0.0/57.5 kB ? eta -:--:--
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  Downloading h5py-3.10.0-cp39-cp39-win_amd64.whl.metadata (2.5 kB)
Collecting libclang==13.0.0 (from tensorflow-intel==2.15.0->tensorflow)
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  Downloading ml_dtypes-0.2.0-cp39-cp39-win_amd64.whl.metadata (20 kB)
Requirement already satisfied: numpy<2.0.0,>=1.23.5 in c:\users\saravi\anaconda3\envs\aimlsem1\lib\site-packages (from tensorflow-intel==2.15.0->tensorflow) (1.23.5)
Collecting opt-einsum==2.3.2 (from tensorflow-intel==2.15.0->tensorflow)
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Requirement already satisfied: packaging in c:\users\saravi\anaconda3\envs\aimlsem1\lib\site-packages (from tensorflow-intel==2.15.0->tensorflow) (23.1)
Requirement already satisfied: protobuf!=4.21.0,!=4.21.1,!=4.21.2,!=4.21.3,!=4.21.4,!=4.21.5,<5.0.0dev,>=3.20.3 in c:\users\saravi\anaconda3\envs\aimlsem1\lib\site-packages (from tensorflow-intel==2.15.0->tensorflow) (4.24.0)
Requirement already satisfied: setuptools in c:\users\saravi\anaconda3\envs\aimlsem1\lib\site-packages (from tensorflow-intel==2.15.0->tensorflow) (68.0.0)
Requirement already satisfied: six>=1.12.0 in c:\users\saravi\anaconda3\envs\aimlsem1\lib\site-packages (from tensorflow-intel==2.15.0->tensorflow) (1.16.0)
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  Downloading termcolor-2.4.0-py3-none-any.whl.metadata (6.1 kB)
Requirement already satisfied: typing-extensions>=3.6.6 in c:\users\saravi\anaconda3\envs\aimlsem1\lib\site-packages (from tensorflow-intel==2.15.0->tensorflow) (4.7.1)
Collecting wrapt<1.15,>=1.11.0 (from tensorflow-intel==2.15.0->tensorflow)
  Downloading wrapt-1.14.1-cp39-cp39-win_amd64.whl (35 kB)
Collecting tensorflow-io-gcs-filesystem==0.23.1 (from tensorflow-intel==2.15.0->tensorflow)
  Downloading tensorflow_io_gcs_filesystem-0.31.0-cp39-cp39-win_amd64.whl (1.5 MB)
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```

```
import tensorflow as tf
```

```
WARNING:tensorflow:From c:\Users\SARAVI\anaconda3\envs\aimlsem1\lib\site-packages\keras\src\losses.py:2976: The name tf.losses.sparse_softmax_cross_entropy is deprecated. Please use tf.compat.v1.losses.sparse_softmax_cross_entropy instead.
```

```
tf.__version__
```

```
'2.15.0'
```

```
# Generate artificial data with 5 samples, 4 features per sample
# and 3 output classes
num_samples = 5 # number of samples
num_features = 4 # number of features (a.k.a. dimensionality)
num_labels = 3 # number of output labels
# Data matrix (each column = single sample)
X = np.random.choice(np.arange(3, 10), size = (num_features, num_samples), replace = True)
# Class labels
y = np.random.choice([0, 1, 2], size = num_samples, replace = True)
print(X)
print('-----')
print(y)
print('-----')
# One-hot encode class labels
y = tf.keras.utils.to_categorical(y)
print(y)

[[6 4 7 7 7]
 [4 3 7 8 5]
 [5 8 4 6 5]
 [9 6 9 5 8]]
-----
[1 2 0 2 1]
-----
[[0. 1. 0.]
 [0. 0. 1.]
 [1. 0. 0.]
 [0. 0. 1.]
 [0. 1. 0.]]
```

A generic layer class with forward and backward methods

```
class Layer:
    def __init__(self):
        self.input = None
        self.output = None

    def forward(self, input):
        pass

    def backward(self, output_gradient, learning_rate):
        pass
```

The softmax classifier steps for a generic sample **x** with (one-hot encoded) true label **y** (3 possible categories) using a randomly initialized weights matrix (with bias abosrbed as its last last column):

- 1. Calculate raw scores vector for a generic sample **x** (bias feature added):

$$\mathbf{z} = \mathbf{W}\mathbf{x}.$$

- 2. Calculate softmax probabilities (that is, softmax-activate the raw scores)

$$\mathbf{a} = \text{softmax}(\mathbf{z}) \Rightarrow \begin{bmatrix} a_0 \\ a_1 \\ a_2 \end{bmatrix} = \text{softmax} \left(\begin{bmatrix} z_0 \\ z_1 \\ z_2 \end{bmatrix} \right) = \begin{bmatrix} \frac{e^{z_0}}{e^{z_0} + e^{z_1} + e^{z_2}} \\ \frac{e^{z_1}}{e^{z_0} + e^{z_1} + e^{z_2}} \\ \frac{e^{z_2}}{e^{z_0} + e^{z_1} + e^{z_2}} \end{bmatrix}$$

- 3. Softmax loss for this sample is (where output label *y* is not yet one-hot encoded)

$$\begin{aligned} L &= -\log([a]_y) \\ &= -\log([\text{softmax}(\mathbf{z})]_y) \\ &= -\log([\text{softmax}(\mathbf{W}\mathbf{x})]_y). \end{aligned}$$

- 4. Predicted probability vector that the sample belongs to each one of the output categories is given a new name

$$\hat{\mathbf{y}} = \mathbf{a}.$$

- 5. One-hot encoding the output label

$$\overbrace{y \rightarrow \mathbf{y}}$$

e.g. 2 $\rightarrow \begin{bmatrix} 0 \\ 0 \\ 1 \end{bmatrix}$

results in the following representation for the softmax loss for the sample which is also referred to as the categorical crossentropy (CCE) loss:

$$L = L(\mathbf{y}, \hat{\mathbf{y}}) = \sum_{k=0}^2 -y_k \log(\hat{y}_k).$$

6. Calculate the gradient of the loss for the sample w.r.t. weights by following the computation graph from top to bottom (that is, backward):

$$\begin{array}{c}
 L \\
 \downarrow \\
 \hat{\mathbf{y}} = \mathbf{a} \\
 \downarrow \\
 \mathbf{z} \\
 \downarrow \\
 \mathbf{W}
 \end{array}
 \Rightarrow \nabla_{\mathbf{W}}(L) = \nabla_{\mathbf{W}}(\mathbf{z}) \times \nabla_{\mathbf{z}}(\mathbf{a}) \times \nabla_{\mathbf{a}}(L)$$

$$= \underbrace{\nabla_{\mathbf{W}}(\mathbf{z})}_{\text{first term}} \times \underbrace{\nabla_{\mathbf{z}}(\mathbf{a})}_{\text{second to last term}} \times \underbrace{\nabla_{\hat{\mathbf{y}}}(L)}_{\text{last term}}.$$

7. Now focus on the last term $\nabla_{\hat{\mathbf{y}}}(L)$:

$$\nabla_{\hat{\mathbf{y}}}(L) = \begin{bmatrix} \nabla_{\hat{y}_0}(L) \\ \nabla_{\hat{y}_1}(L) \\ \nabla_{\hat{y}_2}(L) \end{bmatrix} = \begin{bmatrix} -y_0/\hat{y}_0 \\ -y_1/\hat{y}_1 \\ -y_2/\hat{y}_2 \end{bmatrix}$$

8. Now focus on the second to last term $\nabla_{\mathbf{z}}(\mathbf{a})$:

$$\begin{aligned}
 \nabla_{\mathbf{z}}(\mathbf{a}) &= \nabla_{\mathbf{z}} \left(\begin{bmatrix} a_0 \\ a_1 \\ a_2 \end{bmatrix} \right) \\
 &= \begin{bmatrix} \nabla_{z_0}(a_0) & \nabla_{z_0}(a_1) & \nabla_{z_0}(a_2) \\ \nabla_{z_1}(a_0) & \nabla_{z_1}(a_1) & \nabla_{z_1}(a_2) \\ \nabla_{z_2}(a_0) & \nabla_{z_2}(a_1) & \nabla_{z_2}(a_2) \end{bmatrix} \\
 &= \begin{bmatrix} a_0(1-a_0) & -a_1a_0 & -a_2a_0 \\ -a_0a_1 & a_1(1-a_1) & -a_1a_2 \\ -a_0a_2 & -a_1a_2 & a_2(1-a_2) \end{bmatrix}.
 \end{aligned}$$

9. On Monday, we will focus on the first term to complete the gradient calculation using the computation graph.

```

## Softmax activation class
class Softmax(Layer):
    def forward(self, input):
        self.output = np.array(tf.nn.softmax(input))

    def backward(self, output_gradient, learning_rate):
        return(np.dot((np.identity(np.size(self.output))-self.output.T) * self.output, output_gradient))

## Define the loss function and its gradient
def cce(y, yhat):
    return(-np.sum(y*np.log(yhat)))

def cce_gradient(y, yhat):
    return(-y/yhat)

# TensorFlow in-built function for categorical crossentropy loss
#cce = tf.keras.losses.CategoricalCrossentropy()

# Step-1: add the bias feature to all the samples
X_bias = np.vstack((X, np.ones(X.shape[1])))
X_bias

array([[6., 4., 7., 7., 7.],
       [4., 3., 7., 8., 5.],
       [5., 8., 4., 6., 5.],
       [9., 6., 9., 5., 8.],
       [1., 1., 1., 1., 1.]])

# Step-2: initialize the entries of the weights matrix randomly
W = np.random.rand(num_labels, num_features + 1)

# Step-3: create softmax layer object softmax
softmax = Softmax()

```

```
# Step-4: run over each sample
for i in range(X.shape[1]):
    z = np.dot(W, X_bias[:, i])
    softmax.forward(z)
    cce_loss = cce(y[i, :], softmax.output)
    print(f"Sample {i + 1}, Cross-Entropy Loss: {cce_loss} and Gradient: {cce_gradient(y[i, :], softmax.output)}")

Sample 1, Cross-Entropy Loss: 7.033108378769083 and Gradient: [ -0.          -1133.54863819  -0.          ]
Sample 2, Cross-Entropy Loss: 10.28275450582269 and Gradient: [ -0.          -0.          -29224.26118621]
Sample 3, Cross-Entropy Loss: 0.0028763987446593026 and Gradient: [-1.00288054 -0.          -0.          ]
Sample 4, Cross-Entropy Loss: 10.3623275346211 and Gradient: [ -0.          -0.          -31644.74992358]
Sample 5, Cross-Entropy Loss: 5.549089202804743 and Gradient: [ -0.          -257.00337132  -0.          ]

# (d) Print gradient
gradient = cce_gradient(y[i, :], softmax.output)
print("Gradient:", gradient)

Gradient: [ -0.          -257.00337132  -0.          ]
```