

## Aim: Classification using Deep Neural Network

```
In [1]: import numpy as np
import pandas as pd
```

```
In [2]: from keras.datasets import imdb
(X_train, y_train), (X_test, y_test) = imdb.load_data(num_words=10000)
data = np.concatenate((X_train, X_test), axis=0)
label = np.concatenate((y_train, y_test), axis=0)
```

```
In [3]: print("Review is ",X_train[5])
print("Review is ",y_train[5])
```

```
Review is [1, 778, 128, 74, 12, 630, 163, 15, 4, 1766, 7982, 1051, 2, 32, 8
5, 156, 45, 40, 148, 139, 121, 664, 665, 10, 10, 1361, 173, 4, 749, 2, 16, 3
804, 8, 4, 226, 65, 12, 43, 127, 24, 2, 10, 10]
Review is 0
```

```
In [4]: vocab=imdb.get_word_index()
print(vocab)
```

```
{'fawn': 34701, 'tsukino': 52006, 'nunnery': 52007, 'sonja': 16816, 'van
i': 63951, 'woods': 1408, 'spiders': 16115, 'hanging': 2345, 'woody': 228
9, 'trawling': 52008, "hold's": 52009, 'comically': 11307, 'localized': 4
0830, 'disobeying': 30568, "royale": 52010, "harpo's": 40831, 'canet': 5
2011, 'aileen': 19313, 'acurately': 52012, "diplomat's": 52013, 'rickma
n': 25242, 'arranged': 6746, 'rumbustious': 52014, 'familiarness': 52015,
"spider'": 52016, 'hahahah': 68804, "wood'": 52017, 'transvestism': 4083
3, "hangin'": 34702, 'bringing': 2338, 'seamier': 40834, 'wooded': 34703,
'bravora': 52018, 'grueling': 16817, 'wooden': 1636, 'wednesday': 16818,
"'prix": 52019, 'altagracia': 34704, 'circuitry': 52020, 'crotch': 11585,
'busybody': 57766, "tart'n'tangy": 52021, 'burgade': 14129, 'thrace': 520
23, "tom's": 11038, 'snuggles': 52025, 'francesco': 29114, 'complainers':
52027, 'templarios': 52125, '272': 40835, '273': 52028, 'zaniacs': 52130,
'275': 34706, 'consenting': 27631, 'snuggled': 40836, 'inanimate': 15492,
'uality': 52030, 'bronte': 11926, 'errors': 4010, 'dialogs': 3230, "yomad
a's": 52031, "madman's": 34707, 'dialoge': 30585, 'usetnet': 52033, 'video
drome': 40837, "kid'": 26338, 'pawed': 52034, "'girlfriend'": 30569, "'pl
easeure": 52035, "'reloaded'": 52036, "kazakos'": 40839, 'rocque': 52037,
'mailings': 52038, 'brainwashed': 11927, 'mcanally': 16819, "tom'": 5203
...}
```

```
In [5]: X_train.shape
```

```
Out[5]: (25000,)
```

```
In [6]: X_test.shape
```

```
Out[6]: (25000,)
```

```
In [7]: y_train
```

```
Out[7]: array([1, 0, 0, ..., 0, 1, 0], dtype=int64)
```

```
In [8]: y_test
```

```
Out[8]: array([0, 1, 1, ..., 0, 0, 0], dtype=int64)
```

```
In [9]: def vectorize(sequences, dimension = 10000):  
    # Create an all-zero matrix of shape (len(sequences), dimension)  
    results = np.zeros((len(sequences), dimension))  
    for i, sequence in enumerate(sequences):  
        results[i, sequence] = 1  
    return results
```

```
In [10]: test_x = data[:10000]  
test_y = label[:10000]  
train_x = data[10000:]  
train_y = label[10000:]
```

```
In [11]: test_y
```

```
Out[11]: array([1, 0, 0, ..., 1, 0, 0], dtype=int64)
```

```
In [12]: print("Label:", label[0])
```

```
Label: 1
```

```
In [13]: print(data[0])
```

```
[1, 14, 22, 16, 43, 530, 973, 1622, 1385, 65, 458, 4468, 66, 3941, 4, 173, 3  
6, 256, 5, 25, 100, 43, 838, 112, 50, 670, 2, 9, 35, 480, 284, 5, 150, 4, 17  
2, 112, 167, 2, 336, 385, 39, 4, 172, 4536, 1111, 17, 546, 38, 13, 447, 4, 1  
92, 50, 16, 6, 147, 2025, 19, 14, 22, 4, 1920, 4613, 469, 4, 22, 71, 87, 12,  
16, 43, 530, 38, 76, 15, 13, 1247, 4, 22, 17, 515, 17, 12, 16, 626, 18, 2,  
5, 62, 386, 12, 8, 316, 8, 106, 5, 4, 2223, 5244, 16, 480, 66, 3785, 33, 4,  
130, 12, 16, 38, 619, 5, 25, 124, 51, 36, 135, 48, 25, 1415, 33, 6, 22, 12,  
215, 28, 77, 52, 5, 14, 407, 16, 82, 2, 8, 4, 107, 117, 5952, 15, 256, 4, 2,  
7, 3766, 5, 723, 36, 71, 43, 530, 476, 26, 400, 317, 46, 7, 4, 2, 1029, 13,  
104, 88, 4, 381, 15, 297, 98, 32, 2071, 56, 26, 141, 6, 194, 7486, 18, 4, 22  
6, 22, 21, 134, 476, 26, 480, 5, 144, 30, 5535, 18, 51, 36, 28, 224, 92, 25,  
104, 4, 226, 65, 16, 38, 1334, 88, 12, 16, 283, 5, 16, 4472, 113, 103, 32, 1  
5, 16, 5345, 19, 178, 32]
```

```
In [14]: index = imdb.get_word_index()
reverse_index = dict([(value, key) for (key, value) in index.items()])
decoded = " ".join( [reverse_index.get(i - 3, "#") for i in data[0]] )
print(decoded)
```

```
# this film was just brilliant casting location scenery story direction everyone's really suited the part they played and you could just imagine being there robert # is an amazing actor and now the same being director # father came from the same scottish island as myself so i loved the fact there was a real connection with this film the witty remarks throughout the film were great it was just brilliant so much that i bought the film as soon as it was released for # and would recommend it to everyone to watch and the fly fishing was amazing really cried at the end it was so sad and you know what they say if you cry at a film it must have been good and this definitely was also # to the two little boy's that played the # of norman and paul they were just brilliant children are often left out of the # list i think because the stars that play them all grown up are such a big profile for the whole film but these children are amazing and should be praised for what they have done don't you think the whole story was so lovely because it was true and was someone's life after all that was shared with us all
```

```
In [15]: index
```

```
Out[15]: {'fawn': 34701,
'tsukino': 52006,
'nunnery': 52007,
'sonja': 16816,
'vani': 63951,
'woods': 1408,
'spiders': 16115,
'hanging': 2345,
'woody': 2289,
'trawling': 52008,
'hold's': 52009,
'comically': 11307,
'localized': 40830,
'disobeying': 30568,
'royale': 52010,
'harpo's': 40831,
'canet': 52011,
'aileen': 19313,
'acurately': 52012,
'diplomatic': 52013}
```

```
In [16]: reverse_index
```

```
Out[16]: {34701: 'fawn',
52006: 'tsukino',
52007: 'nunnery',
16816: 'sonja',
63951: 'vani',
1408: 'woods',
16115: 'spiders',
2345: 'hanging',
2289: 'woody',
52008: 'trawling',
52009: "hold's",
11307: 'comically',
40830: 'localized',
30568: 'disobeying',
52010: "'royale",
40831: "harpo's",
52011: 'canet',
19313: 'aileen',
52012: 'acurately',
52013: "it's a bit of a"
```

```
In [17]: decoded
```

```
Out[17]: "# this film was just brilliant casting location scenery story direction eve
ryone's really suited the part they played and you could just imagine being
there robert # is an amazing actor and now the same being director # father
came from the same scottish island as myself so i loved the fact there was a
real connection with this film the witty remarks throughout the film were gr
eat it was just brilliant so much that i bought the film as soon as it was r
eleased for # and would recommend it to everyone to watch and the fly fishin
g was amazing really cried at the end it was so sad and you know what they s
ay if you cry at a film it must have been good and this definitely was also
# to the two little boy's that played the # of norman and paul they were jus
t brilliant children are often left out of the # list i think because the st
ars that play them all grown up are such a big profile for the whole film bu
t these children are amazing and should be praised for what they have done d
on't you think the whole story was so lovely because it was true and was som
eone's life after all that was shared with us all"
```

```
In [18]: data = vectorize(data)
label = np.array(label).astype("float32")
```

```
In [19]: # Creating train and test data set
from sklearn.model_selection import train_test_split
X_train, X_test, y_train, y_test = train_test_split(data,label, test_size=0.2)
```

```
In [20]: X_train.shape
```

```
Out[20]: (40000, 10000)
```

```
In [21]: X_test.shape
```

```
Out[21]: (10000, 10000)
```

```
In [22]: from keras.utils import to_categorical
from keras import models
from keras import layers
```

```
In [23]: model = models.Sequential()
# Input - Layer
model.add(layers.Dense(50, activation = "relu", input_shape=(10000, )))
# Hidden - Layers
model.add(layers.Dropout(0.3, noise_shape=None, seed=None))
model.add(layers.Dense(50, activation = "relu")) #ReLU stands for Rectified
model.add(layers.Dropout(0.2, noise_shape=None, seed=None))
model.add(layers.Dense(50, activation = "relu"))
# Output- Layer
model.add(layers.Dense(1, activation = "sigmoid")) #adds another Dense Layer
model.summary()
```

Model: "sequential"

Layer (type)	Output Shape	Param #
=====		
dense (Dense)	(None, 50)	500050
dropout (Dropout)	(None, 50)	0
dense_1 (Dense)	(None, 50)	2550
dropout_1 (Dropout)	(None, 50)	0
dense_2 (Dense)	(None, 50)	2550
dense_3 (Dense)	(None, 1)	51

=====

Total params: 505,201  
Trainable params: 505,201  
Non-trainable params: 0

=====

```
In [24]: import tensorflow as tf
callback = tf.keras.callbacks.EarlyStopping(monitor='loss', patience=3)
```

```
In [25]: model.compile(  
    optimizer = "adam",  
    loss = "binary_crossentropy",  
    metrics = ["accuracy"]  
)
```

```
In [26]: results = model.fit(  
    X_train, y_train,  
    epochs= 2,  
    batch_size = 500,  
    validation_data = (X_test, y_test),  
    callbacks=[callback]  
)
```

```
Epoch 1/2  
80/80 [=====] - 5s 46ms/step - loss: 0.4119 - accur  
acy: 0.8105 - val_loss: 0.2616 - val_accuracy: 0.8968  
Epoch 2/2  
80/80 [=====] - 3s 38ms/step - loss: 0.2191 - accur  
acy: 0.9158 - val_loss: 0.2548 - val_accuracy: 0.8989
```

```
In [27]: print(np.mean(results.history["val_accuracy"]))
```

```
0.897849977016449
```

```
In [28]: model.predict(X_test)
```

```
313/313 [=====] - 1s 3ms/step
```

```
Out[28]: array([[0.1319962 ],  
    [0.9938973 ],  
    [0.8705446 ],  
    ...,  
    [0.95068085],  
    [0.9893535 ],  
    [0.99133193]], dtype=float32)
```