Dataset:

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Lab: 5: Text corpus creation and binary classification using $\mathbf{DNN}\P$

In [6]: M import nltk import pandas as pd In [7]: M from nltk.corpus import stopwords from sklearn.model_selection import train_test_split from nltk.stem import WordNetLemmatizer In [8]: import warnings warnings.filterwarnings("ignore", category=YourWarningCategory, action="once") In [9]: M nltk.download('stopwords') stop words = set(stopwords.words('english')) nltk.download('wordnet') nltk.download('omw-1.4') In [10]: H df=pd.read csv("Motive - Sheet1.csv",encoding='cp1252') H In [11]: df.shape In [12]: H df.head()

```
H
In [13]:
df.groupby('Label').count()
Pre-processing:
In [8]:
                                                                                        H
X=df.Statement
y=df.Label
In [9]:
                                                                                        M
lemmatizer=WordNetLemmatizer()
In [10]:
                                                                                        H
def clean_review(review):
    tokens = review.lower().split()
    filtered_tokens = [lemmatizer.lemmatize(w) for w in tokens if w not in stop_words]
    return " ".join(filtered_tokens)
In [11]:
                                                                                        M
temp=X.tolist()
fax=[]
for i in temp:
    fax.append(clean_review(i))
n_X=pd.Series(fax)
```

```
H
In [12]:
from sklearn.feature_extraction.text import TfidfVectorizer
import pandas as pd
tfidf = TfidfVectorizer()
vectors = tfidf.fit_transform(n_X)
features_names = tfidf.get_feature_names_out()
text_vect = pd.DataFrame(vectors.todense(), columns=features_names)
text_vect
Out[12]:
     abilities
              achieve
                         afraid anything
                                          believe
                                                   bother capable challenge
                                                                              change
                                                                                         come
 0 0.000000 0.550965 0.000000
                               0.550965
                                        0.626798 0.000000
                                                           0.00000
                                                                     0.00000 0.000000 0.000000
  1 0.000000 0.000000 0.000000 0.000000
                                        0.000000
                                                 0.000000
                                                          0.00000
                                                                     0.00000
                                                                             0.000000 0.000000
 2 0.000000 0.000000 0.000000 0.000000
                                        0.000000
                                                 0.000000
                                                           0.00000
                                                                     0.00000
                                                                             0.000000 0.000000
   0.000000 0.000000 0.000000
                              0.000000
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                                                                             0.000000 0.519707
                                        0.000000
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                                                                     0.00000
    0.000000 0.000000 0.000000 0.000000
                                                          0.00000
                                                                     0.00000 0.000000 0.000000
                                        0.000000
                                                 0.000000
    0.000000 0.000000 0.421482 0.000000
                                                           0.00000
                                                                     0.00000 0.000000 0.000000
                                        0.000000
                                                 0.000000
    0.000000 0.000000 0.000000 0.000000
                                                                     0.57735 0.000000 0.000000
                                        0.000000
                                                 0.000000
                                                           0.57735
                                                                     0.00000
    0.000000
            0.000000
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    0.000000 0.000000 0.000000 0.000000
                                        0.000000
                                                 0.000000
                                                           0.00000
                                                                     0.00000 0.000000 0.000000
                                                                                                H
In [13]:
import tensorflow as tf
temp = tf.Variable(text_vect)
                                                                                                H
In [14]:
X_train,X_test,y_train,y_test=train_test_split(text_vect,y,train_size=0.75,test_size=0.2
                                                                                                H
In [15]:
print(X_train.shape)
(15, 62)
                                                                                                H
In [16]:
print(y_train.shape)
(15,)
```

```
In [17]:
                                                                                  M
print(X_test.shape)
(5, 62)
In [18]:
                                                                                  M
print(y_test.shape)
(5,)
Model Build
In [19]:
                                                                                  M
import tensorflow as tf
from tensorflow.keras import Sequential
from keras.layers import Dense,Activation
Model 1 : Hidden Layer : 8
                                                                                  H
In [20]:
model1 = Sequential()
model1.add(Dense(8, activation='relu',input_dim=X_train.shape[1]))
model1.add(Dense(2, activation='sigmoid'))
model1.summary()
Model: "sequential"
 Layer (type)
                            Output Shape
                                                     Param #
______
 dense (Dense)
                            (None, 8)
                                                     504
 dense 1 (Dense)
                            (None, 2)
                                                     18
Total params: 522 (2.04 KB)
Trainable params: 522 (2.04 KB)
Non-trainable params: 0 (0.00 Byte)
In [27]:
                                                                                  M
from sklearn.preprocessing import LabelEncoder
label_encoder = LabelEncoder()
y_train = label_encoder.fit_transform(y_train)
y_test = label_encoder.fit_transform(y_test)
```

```
In [28]:
model1.compile(loss='sparse_categorical_crossentropy',optimizer='adam',metrics=['accurac
history=model1.fit(X_train,y_train,epochs=100,verbose=2,validation_split=0.2,batch_size=
Epoch 54/100
1/1 - 0s - loss: 0.2691 - accuracy: 1.0000 - val_loss: 0.9616 - val_ac
curacy: 0.3333 - 55ms/epoch - 55ms/step
Epoch 55/100
1/1 - 0s - loss: 0.2671 - accuracy: 1.0000 - val_loss: 0.9638 - val_ac
curacy: 0.3333 - 58ms/epoch - 58ms/step
Epoch 56/100
1/1 - 0s - loss: 0.2651 - accuracy: 1.0000 - val loss: 0.9660 - val ac
curacy: 0.3333 - 56ms/epoch - 56ms/step
Epoch 57/100
1/1 - 0s - loss: 0.2630 - accuracy: 1.0000 - val_loss: 0.9682 - val ac
curacy: 0.3333 - 60ms/epoch - 60ms/step
Epoch 58/100
1/1 - 0s - loss: 0.2611 - accuracy: 1.0000 - val loss: 0.9705 - val ac
curacy: 0.3333 - 55ms/epoch - 55ms/step
Epoch 59/100
1/1 - 0s - loss: 0.2591 - accuracy: 1.0000 - val_loss: 0.9728 - val_ac
curacy: 0.3333 - 78ms/epoch - 78ms/step
Epoch 60/100
1/1 - 0s - loss: 0.2571 - accuracy: 1.0000 - val loss: 0.9752 - val ac
In [29]:
model1.evaluate(X_test,y_test)
```

Out[29]:

[1.2599141597747803, 0.20000000298023224]

Model 2: Hidden Layer: 16

In [30]:

```
model2 = Sequential()
model2.add(Dense(16, activation='relu',input_dim=X_train.shape[1]))
model2.add(Dense(8, activation='relu'))
model2.add(Dense(2, activation='sigmoid'))
model2.summary()
```

Model: "sequential_1"

Layer (type)	Output Shape	Param #
dense_2 (Dense)	(None, 16)	1008
dense_3 (Dense)	(None, 8)	136
dense_4 (Dense)	(None, 2)	18
=======================================		=========

Total params: 1162 (4.54 KB) Trainable params: 1162 (4.54 KB) Non-trainable params: 0 (0.00 Byte)

In [31]: M

model2.compile(loss='sparse_categorical_crossentropy',optimizer='adam',metrics=['accurac history=model2.fit(X train,y train,epochs=100,verbose=2,validation split=0.2,batch size=

```
Epoch 1/100
1/1 - 1s - loss: 0.7032 - accuracy: 0.5000 - val_loss: 0.6688 - val_ac
curacy: 0.6667 - 1s/epoch - 1s/step
Epoch 2/100
1/1 - 0s - loss: 0.6996 - accuracy: 0.5000 - val_loss: 0.6696 - val_ac
curacy: 0.6667 - 76ms/epoch - 76ms/step
Epoch 3/100
1/1 - 0s - loss: 0.6961 - accuracy: 0.5000 - val_loss: 0.6702 - val_ac
curacy: 0.6667 - 54ms/epoch - 54ms/step
Epoch 4/100
1/1 - 0s - loss: 0.6926 - accuracy: 0.5000 - val loss: 0.6709 - val ac
curacy: 0.6667 - 56ms/epoch - 56ms/step
Epoch 5/100
1/1 - 0s - loss: 0.6892 - accuracy: 0.5833 - val_loss: 0.6715 - val_ac
curacy: 0.6667 - 60ms/epoch - 60ms/step
Epoch 6/100
1/1 - 0s - loss: 0.6857 - accuracy: 0.5833 - val loss: 0.6721 - val ac
curacy: 0.6667 - 54ms/epoch - 54ms/step
Epoch 7/100
```

In [32]:

```
model2.evaluate(X_test,y_test)
```

Out[32]:

[0.9271775484085083, 0.20000000298023224]

Model 3: Hidden Layer: 32

In [33]: ▶

```
model3 = Sequential()
model3.add(Dense(32, activation='relu',input_dim=X_train.shape[1]))
model3.add(Dense(16, activation='relu'))
model3.add(Dense(8, activation='relu'))
model3.add(Dense(2, activation='sigmoid'))
model3.summary()
```

Model: "sequential_2"

Layer (type)	Output Shape	Param #
dense_5 (Dense)	(None, 32)	2016
dense_6 (Dense)	(None, 16)	528
dense_7 (Dense)	(None, 8)	136
dense_8 (Dense)	(None, 2)	18

Total params: 2698 (10.54 KB)
Trainable params: 2698 (10.54 KB)
Non-trainable params: 0 (0.00 Byte)

localhost:8888/notebooks/225229101 PDL Lab-5.ipynb#

```
In [34]:
model3.compile(loss='sparse_categorical_crossentropy',optimizer='adam',metrics=['accurac
history=model3.fit(X_train,y_train,epochs=100,verbose=2,validation_split=0.2,batch_size=
Epoch 1/100
1/1 - 2s - loss: 0.6831 - accuracy: 0.5833 - val_loss: 0.7003 - val_ac
curacy: 0.3333 - 2s/epoch - 2s/step
Epoch 2/100
1/1 - 0s - loss: 0.6792 - accuracy: 0.6667 - val_loss: 0.6990 - val_ac
curacy: 0.3333 - 64ms/epoch - 64ms/step
Epoch 3/100
1/1 - 0s - loss: 0.6756 - accuracy: 0.6667 - val_loss: 0.6979 - val_ac
curacy: 0.3333 - 58ms/epoch - 58ms/step
Epoch 4/100
1/1 - 0s - loss: 0.6718 - accuracy: 0.6667 - val_loss: 0.6969 - val_ac
curacy: 0.3333 - 64ms/epoch - 64ms/step
Epoch 5/100
1/1 - 0s - loss: 0.6677 - accuracy: 0.6667 - val loss: 0.6959 - val ac
curacy: 0.3333 - 65ms/epoch - 65ms/step
Epoch 6/100
1/1 - 0s - loss: 0.6637 - accuracy: 0.6667 - val_loss: 0.6954 - val_ac
curacy: 0.3333 - 67ms/epoch - 67ms/step
Epoch 7/100
In [35]:
model3.evaluate(X_test,y_test)
1/1 [================ ] - 0s 45ms/step - loss: 1.6447 - accu
```

Out[35]:

racy: 0.2000

[1.6446993350982666, 0.20000000298023224]

Model 4: Hidden Layer: 64

In [36]: ▶

```
model4 = Sequential()
model4.add(Dense(64, activation='relu',input_dim=X_train.shape[1]))
model4.add(Dense(32, activation='relu'))
model4.add(Dense(16, activation='relu'))
model4.add(Dense(8, activation='relu'))
model4.add(Dense(2, activation='sigmoid'))
model4.summary()
```

Model: "sequential_3"

Layer (type)	Output Shape	Param #
dense_9 (Dense)	(None, 64)	4032
dense_10 (Dense)	(None, 32)	2080
dense_11 (Dense)	(None, 16)	528
dense_12 (Dense)	(None, 8)	136
dense_13 (Dense)	(None, 2)	18

Total params: 6794 (26.54 KB)
Trainable params: 6794 (26.54 KB)
Non-trainable params: 0 (0.00 Byte)

In [37]:

H

model4.compile(loss='sparse_categorical_crossentropy',optimizer='adam',metrics=['accurac
history=model4.fit(X_train,y_train,epochs=100,verbose=2,validation_split=0.2,batch_size=

```
Epoch 1/100
1/1 - 2s - loss: 0.6941 - accuracy: 0.3333 - val_loss: 0.7014 - val_ac
curacy: 0.3333 - 2s/epoch - 2s/step
Epoch 2/100
1/1 - 0s - loss: 0.6841 - accuracy: 0.4167 - val loss: 0.7032 - val ac
curacy: 0.3333 - 78ms/epoch - 78ms/step
Epoch 3/100
1/1 - 0s - loss: 0.6763 - accuracy: 0.6667 - val_loss: 0.7050 - val_ac
curacy: 0.3333 - 60ms/epoch - 60ms/step
Epoch 4/100
1/1 - 0s - loss: 0.6693 - accuracy: 0.8333 - val loss: 0.7073 - val ac
curacy: 0.3333 - 61ms/epoch - 61ms/step
1/1 - 0s - loss: 0.6638 - accuracy: 0.9167 - val_loss: 0.7095 - val_ac
curacy: 0.3333 - 54ms/epoch - 54ms/step
Epoch 6/100
1/1 - 0s - loss: 0.6587 - accuracy: 0.9167 - val loss: 0.7112 - val ac
curacy: 0.3333 - 55ms/epoch - 55ms/step
Epoch 7/100
```

In [38]:

```
model4.evaluate(X_test,y_test)
```

Out[38]:

[1.7601642608642578, 0.0]

Model 5: Hidden Layer: 128

In [39]: ▶

```
model5 = Sequential()
model5.add(Dense(128, activation='relu',input_dim=X_train.shape[1]))
model5.add(Dense(64, activation='relu'))
model5.add(Dense(32, activation='relu'))
model5.add(Dense(16, activation='relu'))
model5.add(Dense(8, activation='relu'))
model5.add(Dense(2, activation='sigmoid'))
model5.summary()
```

Model: "sequential_4"

Layer (type)	Output Shape	Param #
dense_14 (Dense)	(None, 128)	8064
dense_15 (Dense)	(None, 64)	8256
dense_16 (Dense)	(None, 32)	2080
dense_17 (Dense)	(None, 16)	528
dense_18 (Dense)	(None, 8)	136
dense_19 (Dense)	(None, 2)	18

Total params: 19082 (74.54 KB)
Trainable params: 19082 (74.54 KB)
Non-trainable params: 0 (0.00 Byte)

```
In [40]:
model5.compile(loss='sparse_categorical_crossentropy',optimizer='adam',metrics=['accurac
history=model5.fit(X_train,y_train,epochs=100,verbose=2,validation_split=0.2,batch_size=
Epoch 1/100
1/1 - 2s - loss: 0.6907 - accuracy: 0.6667 - val_loss: 0.6951 - val_ac
curacy: 0.6667 - 2s/epoch - 2s/step
Epoch 2/100
1/1 - 0s - loss: 0.6832 - accuracy: 1.0000 - val_loss: 0.6965 - val_ac
curacy: 0.3333 - 84ms/epoch - 84ms/step
Epoch 3/100
1/1 - 0s - loss: 0.6769 - accuracy: 1.0000 - val_loss: 0.6981 - val_ac
curacy: 0.3333 - 72ms/epoch - 72ms/step
Epoch 4/100
1/1 - 0s - loss: 0.6708 - accuracy: 1.0000 - val_loss: 0.7008 - val_ac
curacy: 0.3333 - 86ms/epoch - 86ms/step
Epoch 5/100
1/1 - 0s - loss: 0.6648 - accuracy: 1.0000 - val loss: 0.7040 - val ac
curacy: 0.3333 - 82ms/epoch - 82ms/step
Epoch 6/100
1/1 - 0s - loss: 0.6587 - accuracy: 1.0000 - val_loss: 0.7074 - val_ac
curacy: 0.3333 - 79ms/epoch - 79ms/step
Epoch 7/100
In [41]:
model5.evaluate(X_test,y_test)
1/1 [=============== ] - 0s 46ms/step - loss: 2.7402 - accu
```

```
racy: 0.0000e+00
```

Out[41]:

[2.7402262687683105, 0.0]

Model 6: Hidden Layer: 256

In [42]:

```
model6 = Sequential()
model6.add(Dense(256, activation='relu',input_dim=X_train.shape[1]))
model6.add(Dense(128, activation='relu'))
model6.add(Dense(64, activation='relu'))
model6.add(Dense(32, activation='relu'))
model6.add(Dense(16, activation='relu'))
model6.add(Dense(6, activation='relu'))
model6.add(Dense(2, activation='sigmoid'))
model6.summary()
```

Model: "sequential_5"

Layer (type)	Output Shape	Param #
dense_20 (Dense)	(None, 256)	16128
dense_21 (Dense)	(None, 128)	32896
dense_22 (Dense)	(None, 64)	8256
dense_23 (Dense)	(None, 32)	2080
dense_24 (Dense)	(None, 16)	528
dense_25 (Dense)	(None, 6)	102
dense_26 (Dense)	(None, 2)	14

Total params: 60004 (234.39 KB)
Trainable params: 60004 (234.39 KB)
Non-trainable params: 0 (0.00 Byte)

```
In [43]:
model6.compile(loss='sparse_categorical_crossentropy',optimizer='adam',metrics=['accurac
history=model6.fit(X_train,y_train,epochs=100,verbose=2,validation_split=0.2,batch_size=
Epoch 5/100
1/1 - 0s - loss: 0.6802 - accuracy: 1.0000 - val_loss: 0.6893 - val_ac
curacy: 0.3333 - 58ms/epoch - 58ms/step
Epoch 6/100
1/1 - 0s - loss: 0.6770 - accuracy: 1.0000 - val_loss: 0.6902 - val_ac
curacy: 0.3333 - 56ms/epoch - 56ms/step
Epoch 7/100
1/1 - 0s - loss: 0.6730 - accuracy: 1.0000 - val_loss: 0.6907 - val_ac
curacy: 0.6667 - 61ms/epoch - 61ms/step
Epoch 8/100
1/1 - 0s - loss: 0.6698 - accuracy: 1.0000 - val_loss: 0.6923 - val_ac
curacy: 0.6667 - 53ms/epoch - 53ms/step
Epoch 9/100
1/1 - 0s - loss: 0.6646 - accuracy: 1.0000 - val_loss: 0.6947 - val_ac
curacy: 0.6667 - 59ms/epoch - 59ms/step
Epoch 10/100
1/1 - 0s - loss: 0.6603 - accuracy: 1.0000 - val_loss: 0.6956 - val_ac
curacy: 0.6667 - 54ms/epoch - 54ms/step
Epoch 11/100
1/1 - 0s - loss: 0.6555 - accuracy: 1.0000 - val_loss: 0.6961 - val_ac
In [44]:
model6.evaluate(X_test,y_test)
1/1 [================ ] - 0s 45ms/step - loss: 5.1345 - accu
```

```
racy: 0.0000e+00
```

Out[44]:

[5.134530067443848, 0.0]

Model 7: Hidden Layer: 512

In [45]:

```
model7 = Sequential()
model7.add(Dense(512, activation='relu',input_dim=X_train.shape[1]))
model7.add(Dense(256, activation='relu'))
model7.add(Dense(128, activation='relu'))
model7.add(Dense(64, activation='relu'))
model7.add(Dense(32, activation='relu'))
model7.add(Dense(16, activation='relu'))
model7.add(Dense(8, activation='relu'))
model7.add(Dense(2, activation='relu'))
model7.add(Dense(2, activation='sigmoid'))
model7.summary()
```

Model: "sequential_6"

Layer (type)	Output Shape	Param #
dense_27 (Dense)	(None, 512)	32256
dense_28 (Dense)	(None, 256)	131328
dense_29 (Dense)	(None, 128)	32896
dense_30 (Dense)	(None, 64)	8256
dense_31 (Dense)	(None, 32)	2080
dense_32 (Dense)	(None, 16)	528
dense_33 (Dense)	(None, 8)	136
dense_34 (Dense)	(None, 2)	18

Total params: 207498 (810.54 KB)
Trainable params: 207498 (810.54 KB)
Non-trainable params: 0 (0.00 Byte)

```
In [46]:
model7.compile(loss='sparse_categorical_crossentropy',optimizer='adam',metrics=['accurac
history=model7.fit(X_train,y_train,epochs=100,verbose=2,validation_split=0.2,batch_size=
Epoch 1/100
1/1 - 3s - loss: 0.6870 - accuracy: 0.6667 - val_loss: 0.7017 - val_ac
curacy: 0.3333 - 3s/epoch - 3s/step
Epoch 2/100
1/1 - 0s - loss: 0.6729 - accuracy: 0.6667 - val_loss: 0.7075 - val_ac
curacy: 0.3333 - 70ms/epoch - 70ms/step
Epoch 3/100
1/1 - 0s - loss: 0.6533 - accuracy: 0.6667 - val_loss: 0.7161 - val_ac
curacy: 0.3333 - 74ms/epoch - 74ms/step
Epoch 4/100
1/1 - 0s - loss: 0.6324 - accuracy: 0.6667 - val_loss: 0.7281 - val_ac
curacy: 0.3333 - 76ms/epoch - 76ms/step
Epoch 5/100
1/1 - 0s - loss: 0.6070 - accuracy: 0.6667 - val loss: 0.7417 - val ac
curacy: 0.3333 - 80ms/epoch - 80ms/step
Epoch 6/100
1/1 - 0s - loss: 0.5782 - accuracy: 0.6667 - val_loss: 0.7613 - val_ac
curacy: 0.3333 - 66ms/epoch - 66ms/step
Epoch 7/100
In [47]:
model7.evaluate(X_test,y_test)
1/1 [=============== ] - 0s 54ms/step - loss: 4.6134 - accu
racy: 0.0000e+00
```

Out[47]:

[4.6134138107299805, 0.0]

Model 8: Hidden Layer: 1028

In [48]:

M

```
model8 = Sequential()
model8.add(Dense(1028, activation='relu',input_dim=X_train.shape[1]))
model8.add(Dense(512, activation='relu'))
model8.add(Dense(356, activation='relu'))
model8.add(Dense(128, activation='relu'))
model8.add(Dense(64, activation='relu'))
model8.add(Dense(32, activation='relu'))
model8.add(Dense(16, activation='relu'))
model8.add(Dense(8, activation='relu'))
model8.add(Dense(2, activation='relu'))
model8.add(Dense(2, activation='relu'))
model8.summary()
```

Model: "sequential_7"

Layer (type)	Output Shape	Param #
dense_35 (Dense)	(None, 1028)	64764
dense_36 (Dense)	(None, 512)	526848
dense_37 (Dense)	(None, 356)	182628
dense_38 (Dense)	(None, 128)	45696
dense_39 (Dense)	(None, 64)	8256
dense_40 (Dense)	(None, 32)	2080
dense_41 (Dense)	(None, 16)	528
dense_42 (Dense)	(None, 8)	136
dense_43 (Dense)	(None, 2)	18

Total params: 830954 (3.17 MB)
Trainable params: 830954 (3.17 MB)
Non-trainable params: 0 (0.00 Byte)

localhost:8888/notebooks/225229101 PDL Lab-5.ipynb#

```
In [49]:
model8.compile(loss='sparse_categorical_crossentropy',optimizer='adam',metrics=['accurac
history=model8.fit(X_train,y_train,epochs=100,verbose=2,validation_split=0.2,batch_size=
Epoch 1/100
1/1 - 3s - loss: 0.6959 - accuracy: 0.3333 - val_loss: 0.6933 - val_ac
curacy: 0.6667 - 3s/epoch - 3s/step
Epoch 2/100
1/1 - 0s - loss: 0.6935 - accuracy: 0.3333 - val_loss: 0.6926 - val_ac
curacy: 0.6667 - 87ms/epoch - 87ms/step
Epoch 3/100
1/1 - 0s - loss: 0.6894 - accuracy: 0.5000 - val_loss: 0.6934 - val_ac
curacy: 0.3333 - 86ms/epoch - 86ms/step
Epoch 4/100
1/1 - 0s - loss: 0.6864 - accuracy: 0.7500 - val_loss: 0.6946 - val_ac
curacy: 0.3333 - 88ms/epoch - 88ms/step
Epoch 5/100
1/1 - 0s - loss: 0.6798 - accuracy: 0.8333 - val loss: 0.6952 - val ac
curacy: 0.3333 - 101ms/epoch - 101ms/step
Epoch 6/100
1/1 - 0s - loss: 0.6720 - accuracy: 1.0000 - val_loss: 0.6953 - val_ac
curacy: 0.3333 - 111ms/epoch - 111ms/step
Epoch 7/100
In [50]:
model8.evaluate(X_test,y_test)
```

Out[50]:

[16.292612075805664, 0.20000000298023224]

Model 1 : Outer Layer : 2

```
H
In [51]:
model = Sequential()
model.add(Dense(32, activation='relu',input_dim=X_train.shape[1]))
model.add(Dense(2, activation='sigmoid'))
model.summary()
Model: "sequential_8"
Layer (type)
                          Output Shape
                                                 Param #
______
dense 44 (Dense)
                          (None, 32)
                                                 2016
dense_45 (Dense)
                          (None, 2)
                                                 66
______
Total params: 2082 (8.13 KB)
Trainable params: 2082 (8.13 KB)
Non-trainable params: 0 (0.00 Byte)
                                                                            M
In [52]:
model.compile(loss='sparse_categorical_crossentropy',optimizer='adam',metrics=['accuracy
history=model.fit(X_train,y_train,epochs=100,verbose=2,validation_split=0.2,batch_size=1
Epoch 1/100
1/1 - 1s - loss: 0.6854 - accuracy: 0.5833 - val loss: 0.7520 - val ac
curacy: 0.3333 - 1s/epoch - 1s/step
Epoch 2/100
1/1 - 0s - loss: 0.6794 - accuracy: 0.5833 - val_loss: 0.7539 - val_ac
curacy: 0.3333 - 68ms/epoch - 68ms/step
Epoch 3/100
1/1 - 0s - loss: 0.6735 - accuracy: 0.5833 - val_loss: 0.7558 - val_ac
curacy: 0.3333 - 63ms/epoch - 63ms/step
Epoch 4/100
1/1 - 0s - loss: 0.6678 - accuracy: 0.5833 - val_loss: 0.7578 - val_ac
curacy: 0.3333 - 62ms/epoch - 62ms/step
Epoch 5/100
1/1 - 0s - loss: 0.6621 - accuracy: 0.6667 - val loss: 0.7598 - val ac
curacy: 0.3333 - 63ms/epoch - 63ms/step
Epoch 6/100
1/1 - 0s - loss: 0.6564 - accuracy: 0.6667 - val_loss: 0.7619 - val_ac
curacy: 0.3333 - 63ms/epoch - 63ms/step
Epoch 7/100
In [53]:
model.evaluate(X test,y test)
racy: 0.2000
```

```
localhost:8888/notebooks/225229101 PDL Lab-5.ipynb#
```

[0.987899124622345, 0.20000000298023224]

Out[53]:

Model 2 : Outer Layer : 3

```
M
In [54]:
model0 = Sequential()
model0.add(Dense(32, activation='relu',input_dim=X_train.shape[1]))
model0.add(Dense(3, activation='sigmoid'))
model0.summary()
```

Model: "sequential 9"

Layer (type)	Output Shape	Param #
dense_46 (Dense)	(None, 32)	2016
dense_47 (Dense)	(None, 3)	99
Total params: 2115 (8.2	======================================	

Trainable params: 2115 (8.26 KB) Non-trainable params: 0 (0.00 Byte)

In [55]: M

```
model0.compile(loss='sparse categorical crossentropy',optimizer='adam',metrics=['accuracy
```

```
history=model0.fit(X_train,y_train,epochs=100,verbose=2,validation_split=0.2,batch_size=
1/1 - 0s - loss: 1.0254 - accuracy: 0.5000 - val_loss: 1.0224 - val_ac
curacy: 0.6667 - 109ms/epoch - 109ms/step
Epoch 7/100
1/1 - 0s - loss: 1.0177 - accuracy: 0.7500 - val_loss: 1.0215 - val ac
curacy: 0.6667 - 69ms/epoch - 69ms/step
Epoch 8/100
1/1 - 0s - loss: 1.0100 - accuracy: 0.7500 - val_loss: 1.0206 - val_ac
curacy: 0.6667 - 59ms/epoch - 59ms/step
Epoch 9/100
1/1 - 0s - loss: 1.0023 - accuracy: 0.7500 - val_loss: 1.0197 - val_ac
curacy: 0.6667 - 56ms/epoch - 56ms/step
Epoch 10/100
1/1 - 0s - loss: 0.9947 - accuracy: 0.7500 - val loss: 1.0187 - val ac
curacy: 0.6667 - 56ms/epoch - 56ms/step
Epoch 11/100
1/1 - 0s - loss: 0.9872 - accuracy: 0.7500 - val_loss: 1.0179 - val_ac
curacy: 0.6667 - 53ms/epoch - 53ms/step
Epoch 12/100
1/1 - 0s - loss: 0.9796 - accuracy: 0.7500 - val loss: 1.0171 - val ac
```

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----- 0 (((7

In [56]: ▶

```
model0.evaluate(X_test,y_test)
```

```
1/1 [============ ] - 0s 49ms/step - loss: 1.2712 - accu
```

racy: 0.4000

Out[56]:

[1.27116060256958, 0.4000000059604645]

Model 3: Outer Layer: 4

In [57]: ▶

```
model01 = Sequential()
model01.add(Dense(32, activation='relu',input_dim=X_train.shape[1]))
model01.add(Dense(4, activation='sigmoid'))
model01.summary()
```

Model: "sequential_10"

Layer (type)	Output Shape	Param #
dense_48 (Dense)	(None, 32)	2016
dense_49 (Dense)	(None, 4)	132

Total params: 2148 (8.39 KB)
Trainable params: 2148 (8.39 KB)
Non-trainable params: 0 (0.00 Byte)

```
In [58]:
model01.compile(loss='sparse_categorical_crossentropy',optimizer='adam',metrics=['accura
history=model01.fit(X_train,y_train,epochs=100,verbose=2,validation_split=0.2,batch_size
Epoch 1/100
1/1 - 1s - loss: 1.4518 - accuracy: 0.0833 - val_loss: 1.4335 - val_ac
curacy: 0.0000e+00 - 1s/epoch - 1s/step
Epoch 2/100
1/1 - 0s - loss: 1.4416 - accuracy: 0.0833 - val_loss: 1.4308 - val_ac
curacy: 0.0000e+00 - 73ms/epoch - 73ms/step
Epoch 3/100
1/1 - 0s - loss: 1.4315 - accuracy: 0.1667 - val_loss: 1.4282 - val_ac
curacy: 0.0000e+00 - 61ms/epoch - 61ms/step
Epoch 4/100
1/1 - 0s - loss: 1.4215 - accuracy: 0.2500 - val_loss: 1.4255 - val_ac
curacy: 0.0000e+00 - 59ms/epoch - 59ms/step
Epoch 5/100
1/1 - 0s - loss: 1.4115 - accuracy: 0.2500 - val loss: 1.4228 - val ac
curacy: 0.0000e+00 - 59ms/epoch - 59ms/step
Epoch 6/100
1/1 - 0s - loss: 1.4016 - accuracy: 0.2500 - val_loss: 1.4201 - val_ac
curacy: 0.0000e+00 - 59ms/epoch - 59ms/step
Epoch 7/100
In [59]:
model01.evaluate(X_test,y_test)
1/1 [=============== ] - 0s 46ms/step - loss: 1.4003 - accu
```

```
racy: 0.2000
```

Out[59]:

[1.4003479480743408, 0.20000000298023224]

Model 4: Outer Layer: 5

```
7/25/23, 9:13 PM
                                      225229101 PDL Lab-5 - Jupyter Notebook
                                                                                H
 In [60]:
 model02 = Sequential()
 model02.add(Dense(32, activation='relu',input_dim=X_train.shape[1]))
 model02.add(Dense(5, activation='sigmoid'))
 model02.summary()
 Model: "sequential_11"
  Layer (type)
                            Output Shape
                                                    Param #
 ______
  dense 50 (Dense)
                            (None, 32)
                                                    2016
  dense_51 (Dense)
                            (None, 5)
                                                    165
 ______
 Total params: 2181 (8.52 KB)
 Trainable params: 2181 (8.52 KB)
 Non-trainable params: 0 (0.00 Byte)
 In [61]:
 model02.compile(loss='sparse_categorical_crossentropy',optimizer='adam',metrics=['accura
 history=model02.fit(X_train,y_train,epochs=100,verbose=2,validation_split=0.2,batch_size
 1/1 - של - באפט - accuracy: אועד - val_toss: 1.4883 - val_ac
 curacy: 0.6667 - 65ms/epoch - 65ms/step
 Epoch 4/100
 1/1 - 0s - loss: 1.5866 - accuracy: 0.2500 - val_loss: 1.4838 - val_ac
 curacy: 0.6667 - 55ms/epoch - 55ms/step
 Epoch 5/100
 1/1 - 0s - loss: 1.5743 - accuracy: 0.3333 - val_loss: 1.4794 - val_ac
 curacy: 0.6667 - 62ms/epoch - 62ms/step
```

```
Epoch 6/100
1/1 - 0s - loss: 1.5621 - accuracy: 0.5000 - val_loss: 1.4750 - val_ac
curacy: 0.6667 - 62ms/epoch - 62ms/step
Epoch 7/100
1/1 - 0s - loss: 1.5500 - accuracy: 0.5833 - val loss: 1.4706 - val ac
curacy: 0.6667 - 53ms/epoch - 53ms/step
Epoch 8/100
1/1 - 0s - loss: 1.5380 - accuracy: 0.5833 - val_loss: 1.4663 - val_ac
curacy: 0.6667 - 67ms/epoch - 67ms/step
Epoch 9/100
1/1 - 0s - loss: 1.5261 - accuracy: 0.5833 - val_loss: 1.4620 - val_ac
curacy: 1.0000 - 54ms/epoch - 54ms/step
```

```
In [62]:
                                                                                               H
```

```
model02.evaluate(X test,y test)
```

```
racy: 0.4000
```

Out[62]:

[1.4643977880477905, 0.4000000059604645]

In []:	H