```
import pandas as pd
import numpy as np
import random
import matplotlib.pyplot as plt
```

1) Dataset used is car.data

```
df=pd.read_csv('car.data')
In [70]:
```

df.head(100)

Out[70]:

In [69]:

	vhigh	vhigh.1	2	2.1	small	low	unacc
0	vhigh	vhigh	2	2	small	med	unacc
1	vhigh	vhigh	2	2	small	high	unacc
2	vhigh	vhigh	2	2	med	low	unacc
3	vhigh	vhigh	2	2	med	med	unacc
4	vhigh	vhigh	2	2	med	high	unacc
95	vhigh	vhigh	5more	4	big	low	unacc
96	vhigh	vhigh	5more	4	big	med	unacc
97	vhigh	vhigh	5more	4	big	high	unacc
98	vhigh	vhigh	5more	more	small	low	unacc
99	vhigh	vhigh	5more	more	small	med	unacc

100 rows × 7 columns

2)

```
In [71]:
X=df.drop('unacc', axis=1)
```

In [72]:
X

Out[72]:

	vhigh	vhigh.1	2	2.1	small	low
0	vhigh	vhigh	2	2	small	med
1	vhigh	vhigh	2	2	small	high
2	vhigh	vhigh	2	2	med	low
3	vhigh	vhigh	2	2	med	med
4	vhigh	vhigh	2	2	med	high

```
1722 vhigh vhidhy 5more mere small med
1723
      low
                            med high
                5more more
1724
      low
                5more more
                             big low
1725
            low 5more more
                             big med
      low
1726
                             big high
      low
            low 5more more
1727 rows × 6 columns
In [73]:
from sklearn.preprocessing import LabelEncoder
In [74]:
X=np.array(X)
In [75]:
le=LabelEncoder()
In [76]:
X[:,0]=le.fit transform(X[:,0])
X[:,1] = le.fit transform(X[:,1])
X[:,2]=le.fit\_transform(X[:,2])
X[:,3]=le.fit transform(X[:,3])
X[:,4]=le.fit_transform(X[:,4])
X[:,5]=le.fit transform(X[:,5])
In [77]:
Χ
Out[77]:
array([[3, 3, 0, 0, 2, 2],
       [3, 3, 0, 0, 2, 0],
       [3, 3, 0, 0, 1, 1],
       [1, 1, 3, 2, 0, 1],
       [1, 1, 3, 2, 0, 2],
       [1, 1, 3, 2, 0, 0]], dtype=object)
In [78]:
from sklearn.preprocessing import LabelEncoder
In [79]:
le=LabelEncoder()
In [80]:
Y=df.iloc[:,-1].values
In [81]:
Υ
Out[81]:
array(['unacc', 'unacc', 'unacc', 'good', 'vgood'],
      dtype=object)
In [82]:
from sklearn.preprocessing import LabelEncoder
```

```
In [83]:
le=LabelEncoder()
In [84]:
Y=le.fit transform(Y)
In [85]:
Out[85]:
array([2, 2, 2, ..., 2, 1, 3])
In [86]:
from sklearn.model selection import train test split
In [87]:
X_train, X_test, y_train, y_test = train_test_split(X, Y, test_size=0.33, random_state=4
In [88]:
print("Train shape", X train.shape)
print("Test shape", X test.shape)
print(y train.shape)
print(y test.shape)
Train shape (1157, 6)
Test shape (570, 6)
(1157,)
(570,)
In [89]:
X train=np.asarray(X train).astype(np.int)
y train=np.asarray(y train).astype(np.int)
C:\Users\Surya\AppData\Local\Temp/ipykernel 9300/2533269013.py:1: DeprecationWarning: `np
.int` is a deprecated alias for the builtin `int`. To silence this warning, use `int` by
itself. Doing this will not modify any behavior and is safe. When replacing `np.int`, you
may wish to use e.g. `np.int64` or `np.int32` to specify the precision. If you wish to re
view your current use, check the release note link for additional information.
Deprecated in NumPy 1.20; for more details and quidance: https://numpy.org/devdocs/releas
e/1.20.0-notes.html#deprecations
  X_train=np.asarray(X_train).astype(np.int)
C:\Users\Surya\AppData\Local\Temp/ipykernel_9300/2533269013.py:3: DeprecationWarning: `np .int` is a deprecated alias for the builtin `int`. To silence this warning, use `int` by
itself. Doing this will not modify any behavior and is safe. When replacing `np.int`, you
may wish to use e.g. `np.int64` or `np.int32` to specify the precision. If you wish to re
view your current use, check the release note link for additional information.
Deprecated in NumPy 1.20; for more details and guidance: https://numpy.org/devdocs/releas
e/1.20.0-notes.html#deprecations
  y_train=np.asarray(y_train).astype(np.int)
In [90]:
X test=np.asarray(X test).astype(np.int)
y test=np.asarray(y test).astype(np.int)
C:\Users\Surya\AppData\Local\Temp/ipykernel_9300/1019634495.py:1: DeprecationWarning: `np .int` is a deprecated alias for the builtin `int`. To silence this warning, use `int` by
itself. Doing this will not modify any behavior and is safe. When replacing `np.int`, you
may wish to use e.g. `np.int64` or `np.int32` to specify the precision. If you wish to re
```

view your current use, check the release note link for additional information.

Deprecated in NumPy 1.20; for more details and guidance: https://numpy.org/devdocs/releas

```
e/1.20.0-notes.html#deprecations
 X test=np.asarray(X test).astype(np.int)
C:\Users\Surya\AppData\Local\Temp/ipykernel 9300/1019634495.py:3: DeprecationWarning: `np
.int` is a deprecated alias for the builtin `int`. To silence this warning, use `int` by
itself. Doing this will not modify any behavior and is safe. When replacing `np.int`, you
may wish to use e.g. `np.int64` or `np.int32` to specify the precision. If you wish to re
view your current use, check the release note link for additional information.
Deprecated in NumPy 1.20; for more details and guidance: https://numpy.org/devdocs/releas
e/1.20.0-notes.html#deprecations
 y test=np.asarray(y test).astype(np.int)
In [91]:
import tensorflow as tf
from tensorflow.keras.models import Sequential
from tensorflow.keras.layers import Flatten
from tensorflow.keras.layers import Dense
from tensorflow.keras.layers import Activation
In [92]:
model=Sequential()
In [93]:
model.add(Dense(10, input shape = (6,), activation = 'sigmoid'))
model.add(Dense(10, activation = 'sigmoid'))
model.add(Dense(10, activation = 'sigmoid'))
model.add(Dense(10, activation = 'sigmoid'))
model.add(Dense(10, activation = 'sigmoid'))
model.add(Dense(4))
In [94]:
model.summary()
Model: "sequential 2"
Layer (type)
                            Output Shape
                                                      Param #
                            (None, 10)
dense 12 (Dense)
                                                      70
dense 13 (Dense)
                            (None, 10)
                                                      110
dense 14 (Dense)
                            (None, 10)
                                                      110
                            (None, 10)
dense 15 (Dense)
                                                      110
dense 16 (Dense)
                            (None, 10)
                                                      110
dense 17 (Dense)
                            (None, 4)
______
Total params: 554
Trainable params: 554
Non-trainable params: 0
```

In [95]:

```
from tensorflow.keras import optimizers
sgd = optimizers.SGD(learning_rate = 0.01)
model.compile(optimizer = sgd, loss = 'binary crossentropy', metrics = ['accuracy'])
```

In [96]:

```
train=model.fit(X train, y train, batch size = 50, epochs = 100, verbose = 1)
Epoch 1/100
```

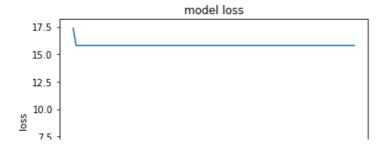
Epoch 2/100

```
Epoch 3/100
Epoch 4/100
Epoch 5/100
Epoch 6/100
Epoch 7/100
Epoch 8/100
Epoch 9/100
Epoch 10/100
Epoch 11/100
Epoch 12/100
Epoch 13/100
Epoch 14/100
Epoch 15/100
Epoch 16/100
Epoch 17/100
Epoch 18/100
Epoch 19/100
Epoch 20/100
Epoch 21/100
Epoch 22/100
Epoch 23/100
Epoch 24/100
Epoch 25/100
Epoch 26/100
Epoch 27/100
Epoch 28/100
Epoch 29/100
Epoch 30/100
Epoch 31/100
Epoch 32/100
Epoch 33/100
Epoch 34/100
Epoch 35/100
Epoch 36/100
Epoch 37/100
Epoch 38/100
```

```
Epoch 39/100
Epoch 40/100
Epoch 41/100
Epoch 42/100
Epoch 43/100
Epoch 44/100
Epoch 45/100
Epoch 46/100
Epoch 47/100
Epoch 48/100
Epoch 49/100
Epoch 50/100
Epoch 51/100
Epoch 52/100
Epoch 53/100
Epoch 54/100
Epoch 55/100
Epoch 56/100
Epoch 57/100
Epoch 58/100
Epoch 59/100
Epoch 60/100
Epoch 61/100
Epoch 62/100
Epoch 63/100
Epoch 64/100
Epoch 65/100
Epoch 66/100
Epoch 67/100
Epoch 68/100
Epoch 69/100
Epoch 70/100
Epoch 71/100
Epoch 72/100
Epoch 73/100
Epoch 74/100
```

```
Epoch 75/100
Epoch 76/100
Epoch 77/100
Epoch 78/100
Epoch 79/100
Epoch 80/100
Epoch 81/100
Epoch 82/100
Epoch 83/100
Epoch 84/100
Epoch 85/100
Epoch 86/100
Epoch 87/100
Epoch 88/100
Epoch 89/100
Epoch 90/100
Epoch 91/100
Epoch 92/100
Epoch 93/100
Epoch 94/100
Epoch 95/100
Epoch 96/100
Epoch 97/100
Epoch 98/100
Epoch 99/100
Epoch 100/100
In [297]:
import matplotlib.pyplot as plt
plt.plot(train.history['loss'])
plt.plot(train.history['accuracy'])
plt.title('model loss')
```

```
plt.ylabel('loss')
plt.xlabel('epoch')
plt.show()
```



```
5.0
  2.5
  0.0
            20
                                    100
                         60
                               80
                    epoch
In [98]:
results = model.evaluate(X test, y test)
In [99]:
print(model.metrics names)
                         # list of metric names the model is employing
print(results)
['loss', 'accuracy']
[15.977612495422363, 0.699999988079071]
In [100]:
print("The Accuracy is ",results[1]*100,"%")
The Accuracy is 69.9999988079071 %
3)
In [134]:
model1=Sequential()
In [122]:
from tensorflow.keras import initializers
In [109]:
# Initializing Initializers
```

```
# Initializing Initializers
initializer1 = tf.keras.initializers.RandomNormal(mean=0., stddev=1.)
initializer2 = tf.keras.initializers.RandomUniform(minval=0., maxval=1.)
initializer3 = tf.keras.initializers.TruncatedNormal(mean=0., stddev=1.)
initializer4 = tf.keras.initializers.Zeros()
initializer5 = tf.keras.initializers.Ones()
initializer6 = tf.keras.initializers.GlorotNormal()
initializer7 = tf.keras.initializers.GlorotUniform()
initializer8 = tf.keras.initializers.HeNormal()
initializer9 = tf.keras.initializers.HeUniform()
initializer10 = tf.keras.initializers.Identity()
initializer11 = tf.keras.initializers.Orthogonal()
```

Random Normal Entropy

```
In [135]:
```

.

```
model1.add(Dense(10, input_shape = (6,), activation = 'sigmoid', kernel_initializer=initi
alizer1))
model1.add(Dense(10, activation = 'sigmoid', kernel_initializer=initializer1))
model1.add(Dense(4))
```

```
In [136]:
```

```
model1.summary()
```

Model: "sequential 8"

Layer (ty	vpe)	Output	Shape	Param #
dense_48	(Dense)	(None,	10)	70
dense_49	(Dense)	(None,	10)	110
dense_50	(Dense)	(None,	10)	110
dense_51	(Dense)	(None,	10)	110
dense_52	(Dense)	(None,	10)	110
dense_53	(Dense)	(None,	4)	44

Total params: 554
Trainable params: 554
Non-trainable params: 0

In [137]:

```
sgd1 = optimizers.SGD(learning_rate = 0.01)
model1.compile(optimizer = sgd1, loss = 'binary_crossentropy', metrics = ['accuracy'])
```

In [138]:

```
train1=model1.fit(X train, y train, batch size = 50, epochs = 100, verbose = 1)
Epoch 1/100
Epoch 2/100
Epoch 3/100
Epoch 4/100
Epoch 5/100
Epoch 6/100
Epoch 7/100
Epoch 8/100
Epoch 9/100
Epoch 10/100
Epoch 11/100
Epoch 12/100
Epoch 13/100
Epoch 14/100
Epoch 15/100
Epoch 16/100
Epoch 17/100
Epoch 18/100
Epoch 19/100
Epoch 20/100
```

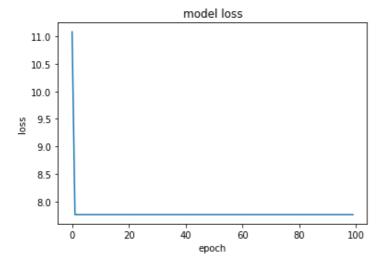
```
Epoch 21/100
Epoch 22/100
Epoch 23/100
Epoch 24/100
Epoch 25/100
Epoch 26/100
Epoch 27/100
Epoch 28/100
Epoch 29/100
Epoch 30/100
Epoch 31/100
Epoch 32/100
Epoch 33/100
Epoch 34/100
Epoch 35/100
Epoch 36/100
Epoch 37/100
Epoch 38/100
Epoch 39/100
Epoch 40/100
Epoch 41/100
Epoch 42/100
Epoch 43/100
Epoch 44/100
Epoch 45/100
Epoch 46/100
Epoch 47/100
Epoch 48/100
Epoch 49/100
Epoch 50/100
Epoch 51/100
Epoch 52/100
Epoch 53/100
Epoch 54/100
Epoch 55/100
Epoch 56/100
```

```
Epoch 57/100
Epoch 58/100
Epoch 59/100
Epoch 60/100
Epoch 61/100
Epoch 62/100
Epoch 63/100
Epoch 64/100
Epoch 65/100
Epoch 66/100
Epoch 67/100
Epoch 68/100
Epoch 69/100
Epoch 70/100
Epoch 71/100
Epoch 72/100
Epoch 73/100
Epoch 74/100
Epoch 75/100
Epoch 76/100
Epoch 77/100
Epoch 78/100
Epoch 79/100
Epoch 80/100
Epoch 81/100
Epoch 82/100
Epoch 83/100
Epoch 84/100
Epoch 85/100
Epoch 86/100
Epoch 87/100
Epoch 88/100
Epoch 89/100
Epoch 90/100
Epoch 91/100
Epoch 92/100
```

```
Epoch 93/100
Epoch 94/100
     =======] - 0s 8ms/step - loss: 7.7604 - accuracy: 0.0337
24/24 [=====
Epoch 95/100
Epoch 96/100
Epoch 97/100
Epoch 98/100
Epoch 99/100
Epoch 100/100
```

In [144]:

```
plt.plot(train1.history['loss'])
plt.title('model loss')
plt.ylabel('loss')
plt.xlabel('epoch')
plt.show()
```



In [291]:

```
results1 = model1.evaluate(X test, y test)
print(model1.metrics names)
print(results1)
print("The Accuracy is ",results[1]*100,"%")
['loss', 'accuracy']
[7.7624125480651855, 0.045614033937454224]
The Accuracy is 4.561403393745422 %
```

RandomUniform

In [116]:

```
model2=Sequential()
```

In [117]:

```
model2.add(Dense(10, input shape = (6,), activation = 'sigmoid', kernel initializer=initi
alizer2))
model2.add(Dense(10, activation = 'sigmoid', kernel initializer=initializer2))
model2.add(Dense(10, activation = 'sigmoid', kernel initializer=initializer2))
model2.add(Dense(10, activation = 'sigmoid', kernel initializer=initializer2))
model2.add(Dense(10, activation = 'sigmoid', kernel_initializer=initializer2))
model2.add(Dense(4))
```

In [118]:

```
model2.summary()
```

Model: "sequential 5"

Layer (ty	vpe)	Output	Shape	Param #
dense_30	(Dense)	(None,	10)	70
dense_31	(Dense)	(None,	10)	110
dense_32	(Dense)	(None,	10)	110
dense_33	(Dense)	(None,	10)	110
dense_34	(Dense)	(None,	10)	110
dense_35	(Dense)	(None,	4)	44

Total params: 554
Trainable params: 554
Non-trainable params: 0

In [119]:

```
sgd2 = optimizers.SGD(learning_rate = 0.01)
model2.compile(optimizer = sgd2, loss = 'binary_crossentropy', metrics = ['accuracy'])
```

In [120]:

```
train2=model2.fit(X train, y train, batch size = 50, epochs = 100, verbose = 1)
Epoch 1/100
Epoch 2/100
Epoch 3/100
Epoch 4/100
Epoch 5/100
Epoch 6/100
Epoch 7/100
Epoch 8/100
Epoch 9/100
Epoch 10/100
Epoch 11/100
Epoch 12/100
Epoch 13/100
Epoch 14/100
Epoch 15/100
Epoch 16/100
Epoch 17/100
Epoch 18/100
Epoch 19/100
01/01
       0 - 1 0 - - / - + - -
          1 - - - . 0 0701
```

```
Epoch 20/100
Epoch 21/100
Epoch 22/100
Epoch 23/100
Epoch 24/100
Epoch 25/100
Epoch 26/100
Epoch 27/100
Epoch 28/100
Epoch 29/100
Epoch 30/100
Epoch 31/100
Epoch 32/100
Epoch 33/100
Epoch 34/100
Epoch 35/100
Epoch 36/100
Epoch 37/100
Epoch 38/100
Epoch 39/100
Epoch 40/100
Epoch 41/100
Epoch 42/100
Epoch 43/100
Epoch 44/100
Epoch 45/100
Epoch 46/100
Epoch 47/100
Epoch 48/100
Epoch 49/100
Epoch 50/100
Epoch 51/100
Epoch 52/100
Epoch 53/100
Epoch 54/100
Epoch 55/100
```

_____1

0- 0---/---

1 - - - . 0 0701

01/01

```
Epoch 56/100
Epoch 57/100
Epoch 58/100
Epoch 59/100
Epoch 60/100
Epoch 61/100
Epoch 62/100
Epoch 63/100
Epoch 64/100
Epoch 65/100
Epoch 66/100
Epoch 67/100
Epoch 68/100
Epoch 69/100
Epoch 70/100
Epoch 71/100
Epoch 72/100
Epoch 73/100
Epoch 74/100
Epoch 75/100
Epoch 76/100
Epoch 77/100
Epoch 78/100
Epoch 79/100
Epoch 80/100
Epoch 81/100
Epoch 82/100
Epoch 83/100
Epoch 84/100
Epoch 85/100
Epoch 86/100
Epoch 87/100
Epoch 88/100
Epoch 89/100
Epoch 90/100
Epoch 91/100
```

_____1

0- 0--/---

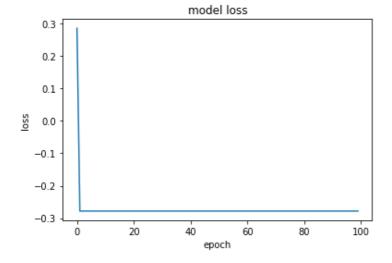
1 - - - - 0 0701

01/01 [-

```
Epoch 92/100
Epoch 93/100
Epoch 94/100
Epoch 95/100
Epoch 96/100
Epoch 97/100
Epoch 98/100
Epoch 99/100
Epoch 100/100
```

In [127]:

```
import matplotlib.pyplot as plt
plt.plot(train2.history['loss'])
plt.title('model loss')
plt.ylabel('loss')
plt.xlabel('epoch')
plt.show()
```



In [173]:

TruncatedNormal

```
In [139]:
```

```
model3=Sequential()
```

In [140]:

```
model3.add(Dense(10, input_shape = (6,), activation = 'sigmoid', kernel_initializer=initi
alizer3))
model3.add(Dense(10, activation = 'sigmoid', kernel_initializer=initializer3))
```

```
model3.add(Dense(10, activation = 'sigmoid', kernel_initializer=initializer3))
model3.add(Dense(10, activation = 'sigmoid', kernel_initializer=initializer3))
model3.add(Dense(10, activation = 'sigmoid', kernel initializer=initializer3))
model3.add(Dense(4))
```

In [141]:

```
model3.summary()
```

Model: "sequential 9"

Layer (type)	Output Shape	Param #
dense_54 (Dense)	(None, 10)	70
dense_55 (Dense)	(None, 10)	110
dense_56 (Dense)	(None, 10)	110
dense_57 (Dense)	(None, 10)	110
dense_58 (Dense)	(None, 10)	110
dense_59 (Dense)	(None, 4)	44

Total params: 554 Trainable params: 554 Non-trainable params: 0

In [142]:

```
sgd3 = optimizers.SGD(learning rate = 0.01)
model3.compile(optimizer = sgd3, loss = 'binary crossentropy', metrics = ['accuracy'])
```

In [143]:

```
train3=model3.fit(X train, y train, batch size = 50, epochs = 100, verbose = 1)
Epoch 1/100
Epoch 2/100
Epoch 3/100
Epoch 4/100
Epoch 5/100
Epoch 6/100
Epoch 7/100
Epoch 8/100
Epoch 9/100
Epoch 10/100
Epoch 11/100
Epoch 12/100
Epoch 13/100
Epoch 14/100
Epoch 15/100
Epoch 16/100
Epoch 17/100
```

```
Epoch 18/100
Epoch 19/100
Epoch 20/100
Epoch 21/100
Epoch 22/100
Epoch 23/100
Epoch 24/100
Epoch 25/100
Epoch 26/100
Epoch 27/100
Epoch 28/100
Epoch 29/100
Epoch 30/100
Epoch 31/100
Epoch 32/100
Epoch 33/100
Epoch 34/100
Epoch 35/100
Epoch 36/100
Epoch 37/100
Epoch 38/100
Epoch 39/100
Epoch 40/100
Epoch 41/100
Epoch 42/100
Epoch 43/100
Epoch 44/100
Epoch 45/100
Epoch 46/100
Epoch 47/100
Epoch 48/100
Epoch 49/100
Epoch 50/100
Epoch 51/100
Epoch 52/100
```

Epoch 53/100

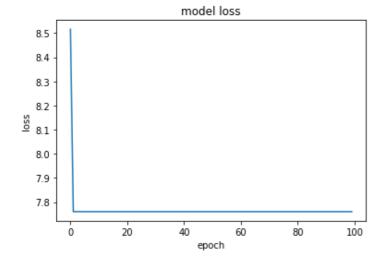
```
Epoch 54/100
Epoch 55/100
Epoch 56/100
Epoch 57/100
Epoch 58/100
Epoch 59/100
Epoch 60/100
Epoch 61/100
Epoch 62/100
Epoch 63/100
Epoch 64/100
Epoch 65/100
Epoch 66/100
Epoch 67/100
Epoch 68/100
Epoch 69/100
Epoch 70/100
Epoch 71/100
Epoch 72/100
Epoch 73/100
Epoch 74/100
Epoch 75/100
Epoch 76/100
Epoch 77/100
Epoch 78/100
Epoch 79/100
Epoch 80/100
Epoch 81/100
Epoch 82/100
Epoch 83/100
Epoch 84/100
Epoch 85/100
Epoch 86/100
Epoch 87/100
Epoch 88/100
```

Epoch 89/100

```
Epoch 90/100
Epoch 91/100
Epoch 92/100
Epoch 93/100
Epoch 94/100
Epoch 95/100
Epoch 96/100
Epoch 97/100
Epoch 98/100
Epoch 99/100
Epoch 100/100
```

In [146]:

```
import matplotlib.pyplot as plt
plt.plot(train3.history['loss'])
plt.title('model loss')
plt.ylabel('loss')
plt.xlabel('epoch')
plt.show()
```



The Accuracy is 22.280701994895935 %

In [170]:

Zeros

In [148]:

```
model4=Sequential()
```

In [149]:

```
model4.add(Dense(10, input_shape = (6,), activation = 'sigmoid', kernel_initializer=initi
alizer4))
model4.add(Dense(10, activation = 'sigmoid', kernel_initializer=initializer4))
model4.add(Dense(4))
```

In [150]:

```
model4.summary()
```

Model: "sequential 10"

Layer (type)	Output Shape	Param #
dense_60 (Dense)	(None, 10)	70
dense_61 (Dense)	(None, 10)	110
dense_62 (Dense)	(None, 10)	110
dense_63 (Dense)	(None, 10)	110
dense_64 (Dense)	(None, 10)	110
dense_65 (Dense)	(None, 4)	44
 Total params: 554		

Trainable params: 554
Non-trainable params: 0

In [151]:

Epoch 15/100

```
sgd4 = optimizers.SGD(learning_rate = 0.01)
model4.compile(optimizer = sgd4, loss = 'binary_crossentropy', metrics = ['accuracy'])
```

```
In [152]:
train4=model4.fit(X train, y train, batch size = 50, epochs = 100, verbose = 1)
Epoch 1/100
Epoch 2/100
Epoch 3/100
Epoch 4/100
Epoch 5/100
Epoch 6/100
Epoch 7/100
Epoch 8/100
Epoch 9/100
Epoch 10/100
Epoch 11/100
Epoch 12/100
Epoch 13/100
Epoch 14/100
```

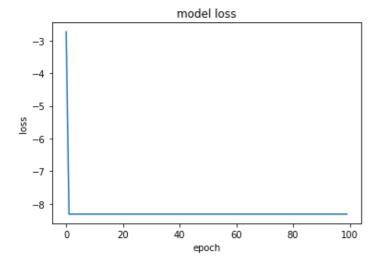
```
Epoch 16/100
Epoch 17/100
Epoch 18/100
Epoch 19/100
Epoch 20/100
Epoch 21/100
Epoch 22/100
Epoch 23/100
Epoch 24/100
Epoch 25/100
Epoch 26/100
Epoch 27/100
Epoch 28/100
Epoch 29/100
Epoch 30/100
Epoch 31/100
Epoch 32/100
Epoch 33/100
Epoch 34/100
Epoch 35/100
Epoch 36/100
Epoch 37/100
Epoch 38/100
Epoch 39/100
Epoch 40/100
Epoch 41/100
Epoch 42/100
Epoch 43/100
Epoch 44/100
Epoch 45/100
Epoch 46/100
Epoch 47/100
Epoch 48/100
Epoch 49/100
Epoch 50/100
Epoch 51/100
```

```
Epoch 52/100
Epoch 53/100
Epoch 54/100
Epoch 55/100
Epoch 56/100
Epoch 57/100
Epoch 58/100
Epoch 59/100
Epoch 60/100
Epoch 61/100
Epoch 62/100
Epoch 63/100
Epoch 64/100
Epoch 65/100
Epoch 66/100
Epoch 67/100
Epoch 68/100
Epoch 69/100
Epoch 70/100
Epoch 71/100
Epoch 72/100
Epoch 73/100
Epoch 74/100
Epoch 75/100
Epoch 76/100
Epoch 77/100
Epoch 78/100
Epoch 79/100
Epoch 80/100
Epoch 81/100
Epoch 82/100
Epoch 83/100
Epoch 84/100
Epoch 85/100
Epoch 86/100
Epoch 87/100
```

```
Epoch 88/100
Epoch 89/100
Epoch 90/100
Epoch 91/100
Epoch 92/100
Epoch 93/100
Epoch 94/100
Epoch 95/100
Epoch 96/100
Epoch 97/100
Epoch 98/100
Epoch 99/100
Epoch 100/100
```

In [154]:

```
import matplotlib.pyplot as plt
plt.plot(train4.history['loss'])
plt.title('model loss')
plt.ylabel('loss')
plt.xlabel('epoch')
plt.show()
```



The Accuracy is 22.280701994895935 %

In [169]:

Ones

In [156]:

```
model5=Sequential()

In [157]:

model5 add (Dance (10 input shape = (6)) activation = Laigneid kernel initializar-initializar-initializar-initializar-initializar-initializar-initializar-initializar-initializar-initializar-initializar-initializar-initializar-initializar-initializar-initializar-initializar-initializar-initializar-initializar-initializar-initializar-initializar-initializar-initializar-initializar-initializar-initializar-initializar-initializar-initializar-initializar-initializar-initializar-initializar-initializar-initializar-initializar-initializar-initializar-initializar-initializar-initializar-initializar-initializar-initializar-initializar-initializar-initializar-initializar-initializar-initializar-initializar-initializar-initializar-initializar-initializar-initializar-initializar-initializar-initializar-initializar-initializar-initializar-initializar-initializar-initializar-initializar-initializar-initializar-initializar-initializar-initializar-initializar-initializar-initializar-initializar-initializar-initializar-initializar-initializar-initializar-initializar-initializar-initializar-initializar-initializar-initializar-initializar-initializar-initializar-initializar-initializar-initializar-initializar-initializar-initializar-initializar-initializar-initializar-initializar-initializar-initializar-initializar-initializar-initializar-initializar-initializar-initializar-initializar-initializar-initializar-initializar-initializar-initializar-initializar-initializar-initializar-initializar-initializar-initializar-initializar-initializar-initializar-initializar-initializar-initializar-initializar-initializar-initializar-initializar-initializar-initializar-initializar-initializar-initializar-initializar-initializar-initializar-initializar-initializar-initializar-initializar-initializar-initializar-initializar-initializar-initializar-initializar-initializar-initializar-initializar-initializar-initializar-initializar-initializar-initializar-initializar-initializar-initializar-initializar-initializar-
```

```
model5.add(Dense(10, input_shape = (6,), activation = 'sigmoid', kernel_initializer=initi
alizer5))
model5.add(Dense(10, activation = 'sigmoid', kernel_initializer=initializer5))
model5.add(Dense(4))
```

In [158]:

```
model5.summary()
```

Model: "sequential 11"

Trainable params: 554
Non-trainable params: 0

Layer (type)	Output Shape	Param #
dense_66 (Dense)	(None, 10)	70
dense_67 (Dense)	(None, 10)	110
dense_68 (Dense)	(None, 10)	110
dense_69 (Dense)	(None, 10)	110
dense_70 (Dense)	(None, 10)	110
dense_71 (Dense)	(None, 4)	44
Total params: 554		

In [159]:

```
sgd5 = optimizers.SGD(learning_rate = 0.01)
model5.compile(optimizer = sgd5, loss = 'binary_crossentropy', metrics = ['accuracy'])
```

In [160]:

```
train5=model5.fit(X train, y train, batch size = 50, epochs = 100, verbose = 1)
Epoch 1/100
Epoch 2/100
Epoch 3/100
Epoch 4/100
Epoch 5/100
Epoch 6/100
Epoch 7/100
Epoch 8/100
Epoch 9/100
Epoch 10/100
Epoch 11/100
Epoch 12/100
Epoch 13/100
```

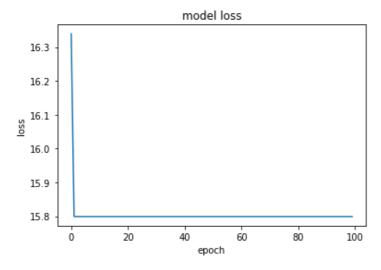
```
00 0m0/000p 1000. 10.7505 accaracy. 0.7001
ے ہے رہے ا
Epoch 14/100
Epoch 15/100
Epoch 16/100
Epoch 17/100
Epoch 18/100
Epoch 19/100
Epoch 20/100
Epoch 21/100
Epoch 22/100
Epoch 23/100
Epoch 24/100
Epoch 25/100
Epoch 26/100
Epoch 27/100
Epoch 28/100
Epoch 29/100
Epoch 30/100
Epoch 31/100
Epoch 32/100
Epoch 33/100
Epoch 34/100
Epoch 35/100
Epoch 36/100
Epoch 37/100
Epoch 38/100
Epoch 39/100
Epoch 40/100
Epoch 41/100
Epoch 42/100
Epoch 43/100
Epoch 44/100
Epoch 45/100
Epoch 46/100
Epoch 47/100
Epoch 48/100
Epoch 49/100
```

```
00 0m0/000p 1000. 10.7505 accaracy. 0.7001
ے ہے رہے ا
Epoch 50/100
Epoch 51/100
Epoch 52/100
Epoch 53/100
Epoch 54/100
Epoch 55/100
Epoch 56/100
Epoch 57/100
Epoch 58/100
Epoch 59/100
Epoch 60/100
Epoch 61/100
Epoch 62/100
Epoch 63/100
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Epoch 66/100
Epoch 67/100
Epoch 68/100
Epoch 69/100
Epoch 70/100
Epoch 71/100
Epoch 72/100
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Epoch 74/100
Epoch 75/100
Epoch 76/100
Epoch 77/100
Epoch 78/100
Epoch 79/100
Epoch 80/100
Epoch 81/100
Epoch 82/100
Epoch 83/100
Epoch 84/100
Epoch 85/100
```

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      00 01110/00CP
        TODD. TO.1909
Epoch 86/100
Epoch 87/100
Epoch 88/100
Epoch 89/100
Epoch 90/100
Epoch 91/100
Epoch 92/100
Epoch 93/100
Epoch 94/100
Epoch 95/100
Epoch 96/100
Epoch 97/100
Epoch 98/100
Epoch 99/100
Epoch 100/100
```

In [161]:

```
import matplotlib.pyplot as plt
plt.plot(train5.history['loss'])
plt.title('model loss')
plt.ylabel('loss')
plt.xlabel('epoch')
plt.show()
```



In [166]:

```
results5 = model5.evaluate(X_test, y_test)
print(model5.metrics_names)
print(results)
print("The Accuracy is ",results[1]*100,"%")
```

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```
In [179]:
```

```
model6=Sequential()
```

In [180]:

```
model6.add(Dense(10, input_shape = (6,), activation = 'sigmoid', kernel_initializer=initi
alizer6))
model6.add(Dense(10, activation = 'sigmoid', kernel_initializer=initializer6))
model6.add(Dense(4))
```

In [181]:

```
model6.summary()
```

Model: "sequential 12"

Layer (type)	Output Shape	Param #
dense_72 (Dense)	(None, 10)	70
dense_73 (Dense)	(None, 10)	110
dense_74 (Dense)	(None, 10)	110
dense_75 (Dense)	(None, 10)	110
dense_76 (Dense)	(None, 10)	110
dense_77 (Dense)	(None, 4)	44

Total params: 554
Trainable params: 554
Non-trainable params: 0

In [182]:

```
sgd6 = optimizers.SGD(learning_rate = 0.01)
model6.compile(optimizer = sgd6, loss = 'binary_crossentropy', metrics = ['accuracy'])
```

In [183]:

Epoch 11/100

```
train6=model6.fit(X train, y train, batch size = 50, epochs = 100, verbose = 1)
Epoch 1/100
Epoch 2/100
Epoch 3/100
Epoch 4/100
Epoch 5/100
Epoch 6/100
Epoch 7/100
Epoch 8/100
Epoch 9/100
Epoch 10/100
```

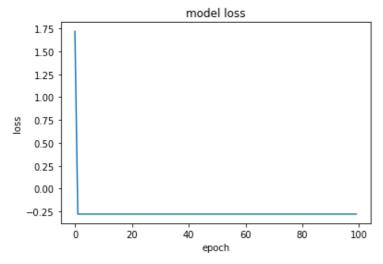
```
Epoch 12/100
Epoch 13/100
Epoch 14/100
Epoch 15/100
Epoch 16/100
Epoch 17/100
Epoch 18/100
Epoch 19/100
Epoch 20/100
Epoch 21/100
Epoch 22/100
Epoch 23/100
Epoch 24/100
Epoch 25/100
Epoch 26/100
Epoch 27/100
Epoch 28/100
Epoch 29/100
Epoch 30/100
Epoch 31/100
Epoch 32/100
Epoch 33/100
Epoch 34/100
Epoch 35/100
Epoch 36/100
Epoch 37/100
Epoch 38/100
Epoch 39/100
Epoch 40/100
Epoch 41/100
Epoch 42/100
Epoch 43/100
Epoch 44/100
Epoch 45/100
Epoch 46/100
Epoch 47/100
```

```
Epoch 48/100
Epoch 49/100
Epoch 50/100
Epoch 51/100
Epoch 52/100
Epoch 53/100
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Epoch 55/100
Epoch 56/100
Epoch 57/100
Epoch 58/100
Epoch 59/100
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Epoch 74/100
Epoch 75/100
Epoch 76/100
Epoch 77/100
Epoch 78/100
Epoch 79/100
Epoch 80/100
Epoch 81/100
Epoch 82/100
Epoch 83/100
```

```
Epoch 84/100
Epoch 85/100
Epoch 86/100
Epoch 87/100
Epoch 88/100
Epoch 89/100
Epoch 90/100
Epoch 91/100
Epoch 92/100
Epoch 93/100
Epoch 94/100
Epoch 95/100
Epoch 96/100
Epoch 97/100
Epoch 98/100
Epoch 99/100
Epoch 100/100
```

In [185]:

```
import matplotlib.pyplot as plt
plt.plot(train6.history['loss'])
plt.title('model loss')
plt.ylabel('loss')
plt.xlabel('epoch')
plt.show()
```



In [186]:

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```
In [187]:
```

```
model7=Sequential()
```

In [188]:

```
model7.add(Dense(10, input_shape = (6,), activation = 'sigmoid', kernel_initializer=initi
alizer7))
model7.add(Dense(10, activation = 'sigmoid', kernel_initializer=initializer7))
model7.add(Dense(4))
```

In [189]:

```
model7.summary()
```

Model: "sequential_13"

Layer (type)	Output Shape	Param #
dense_78 (Dense)	(None, 10)	70
dense_79 (Dense)	(None, 10)	110
dense_80 (Dense)	(None, 10)	110
dense_81 (Dense)	(None, 10)	110
dense_82 (Dense)	(None, 10)	110
dense_83 (Dense)	(None, 4)	44

Total params: 554
Trainable params: 554
Non-trainable params: 0

In [190]:

```
sgd7 = optimizers.SGD(learning_rate = 0.01)
model7.compile(optimizer = sgd7, loss = 'binary_crossentropy', metrics = ['accuracy'])
```

In [191]:

```
train7=model7.fit(X train, y train, batch size = 50, epochs = 100, verbose = 1)
Epoch 1/100
Epoch 2/100
Epoch 3/100
Epoch 4/100
Epoch 5/100
Epoch 6/100
Epoch 7/100
Epoch 8/100
Epoch 9/100
```

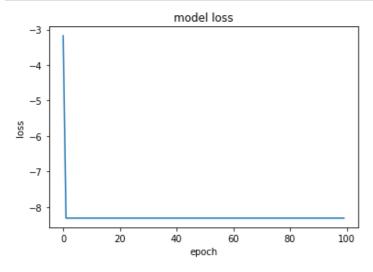
```
Epoch 10/100
Epoch 11/100
Epoch 12/100
Epoch 13/100
Epoch 14/100
Epoch 15/100
Epoch 16/100
Epoch 17/100
Epoch 18/100
Epoch 19/100
Epoch 20/100
Epoch 21/100
Epoch 22/100
Epoch 23/100
Epoch 24/100
Epoch 25/100
Epoch 26/100
Epoch 27/100
Epoch 28/100
Epoch 29/100
Epoch 30/100
Epoch 31/100
Epoch 32/100
Epoch 33/100
Epoch 34/100
Epoch 35/100
Epoch 36/100
Epoch 37/100
Epoch 38/100
Epoch 39/100
Epoch 40/100
Epoch 41/100
Epoch 42/100
Epoch 43/100
Epoch 44/100
Epoch 45/100
```

```
Epoch 46/100
Epoch 47/100
Epoch 48/100
Epoch 49/100
Epoch 50/100
Epoch 51/100
Epoch 52/100
Epoch 53/100
Epoch 54/100
Epoch 55/100
Epoch 56/100
Epoch 57/100
Epoch 58/100
Epoch 59/100
Epoch 60/100
Epoch 61/100
Epoch 62/100
Epoch 63/100
Epoch 64/100
Epoch 65/100
Epoch 66/100
Epoch 67/100
Epoch 68/100
Epoch 69/100
Epoch 70/100
Epoch 71/100
Epoch 72/100
Epoch 73/100
Epoch 74/100
Epoch 75/100
Epoch 76/100
Epoch 77/100
Epoch 78/100
Epoch 79/100
Epoch 80/100
Epoch 81/100
```

```
Epoch 82/100
Epoch 83/100
Epoch 84/100
Epoch 85/100
Epoch 86/100
Epoch 87/100
Epoch 88/100
Epoch 89/100
Epoch 90/100
Epoch 91/100
Epoch 92/100
Epoch 93/100
Epoch 94/100
Epoch 95/100
Epoch 96/100
Epoch 97/100
Epoch 98/100
Epoch 99/100
Epoch 100/100
```

In [192]:

```
import matplotlib.pyplot as plt
plt.plot(train7.history['loss'])
plt.title('model loss')
plt.ylabel('loss')
plt.xlabel('epoch')
plt.show()
```



In [193]:

```
results = model7.evaluate(X_test, y_test)
print(model7.metrics_names)
print(results)
print("The Accuracy is ",results[1]*100,"%")
```

HeNormal

```
In [194]:
```

```
model8=Sequential()
```

In [195]:

```
model8.add(Dense(10, input_shape = (6,), activation = 'sigmoid', kernel_initializer=initi
alizer8))
model8.add(Dense(10, activation = 'sigmoid', kernel_initializer=initializer8))
model8.add(Dense(4))
```

In [196]:

```
model8.summary()
```

Model: "sequential 14"

Layer (ty	vpe)	Output	Shape	Param #
dense_84	(Dense)	(None,	10)	70
dense_85	(Dense)	(None,	10)	110
dense_86	(Dense)	(None,	10)	110
dense_87	(Dense)	(None,	10)	110
dense_88	(Dense)	(None,	10)	110
dense_89	(Dense)	(None,	4)	44

Total params: 554
Trainable params: 554
Non-trainable params: 0

In [197]:

```
sgd8 = optimizers.SGD(learning_rate = 0.01)
model8.compile(optimizer = sgd8, loss = 'binary_crossentropy', metrics = ['accuracy'])
```

In [198]:

```
train8=model8.fit(X train, y train, batch size = 50, epochs = 100, verbose = 1)
Epoch 1/100
Epoch 2/100
Epoch 3/100
Epoch 4/100
Epoch 5/100
Epoch 6/100
Epoch 7/100
```

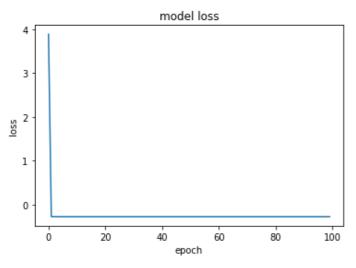
```
Fbocu @\Inn
Epoch 9/100
Epoch 10/100
Epoch 11/100
Epoch 12/100
Epoch 13/100
Epoch 14/100
Epoch 15/100
Epoch 16/100
Epoch 17/100
Epoch 18/100
Epoch 19/100
Epoch 20/100
Epoch 21/100
Epoch 22/100
Epoch 23/100
Epoch 24/100
Epoch 25/100
Epoch 26/100
Epoch 27/100
Epoch 28/100
Epoch 29/100
Epoch 30/100
Epoch 31/100
Epoch 32/100
Epoch 33/100
Epoch 34/100
Epoch 35/100
Epoch 36/100
Epoch 37/100
Epoch 38/100
Epoch 39/100
Epoch 40/100
Epoch 41/100
Epoch 42/100
Epoch 43/100
```

```
Fbocii 44/Tnn
Epoch 45/100
Epoch 46/100
Epoch 47/100
Epoch 48/100
Epoch 49/100
Epoch 50/100
Epoch 51/100
Epoch 52/100
Epoch 53/100
Epoch 54/100
Epoch 55/100
Epoch 56/100
Epoch 57/100
Epoch 58/100
Epoch 59/100
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Epoch 67/100
Epoch 68/100
Epoch 69/100
Epoch 70/100
Epoch 71/100
Epoch 72/100
Epoch 73/100
Epoch 74/100
Epoch 75/100
Epoch 76/100
Epoch 77/100
Epoch 78/100
Epoch 79/100
```

```
Fbocu an/Inn
Epoch 81/100
Epoch 82/100
Epoch 83/100
Epoch 84/100
Epoch 85/100
Epoch 86/100
Epoch 87/100
Epoch 88/100
Epoch 89/100
Epoch 90/100
Epoch 91/100
Epoch 92/100
Epoch 93/100
Epoch 94/100
Epoch 95/100
Epoch 96/100
Epoch 97/100
Epoch 98/100
Epoch 99/100
Epoch 100/100
```

In [199]:

```
import matplotlib.pyplot as plt
plt.plot(train8.history['loss'])
plt.title('model loss')
plt.ylabel('loss')
plt.xlabel('epoch')
plt.show()
```



In [200]:

results = model8.evaluate(X test. v test)

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```
In [202]:
```

```
model9=Sequential()
```

In [203]:

```
model9.add(Dense(10, input_shape = (6,), activation = 'sigmoid', kernel_initializer=initi
alizer9))
model9.add(Dense(10, activation = 'sigmoid', kernel_initializer=initializer9))
model9.add(Dense(4))
```

In [204]:

```
model9.summary()
```

Model: "sequential 16"

Layer (type)	Output Shape	Param #
dense_90 (Dense)	(None, 10)	70
dense_91 (Dense)	(None, 10)	110
dense_92 (Dense)	(None, 10)	110
dense_93 (Dense)	(None, 10)	110
dense_94 (Dense)	(None, 10)	110
dense_95 (Dense)	(None, 4)	44

Total params: 554
Trainable params: 554
Non-trainable params: 0

In [205]:

```
sgd9 = optimizers.SGD(learning_rate = 0.01)
model9.compile(optimizer = sgd9, loss = 'binary_crossentropy', metrics = ['accuracy'])
```

In [206]:

```
Epoch 7/100
Epoch 8/100
Epoch 9/100
Epoch 10/100
Epoch 11/100
Epoch 12/100
Epoch 13/100
Epoch 14/100
Epoch 15/100
Epoch 16/100
Epoch 17/100
Epoch 18/100
Epoch 19/100
Epoch 20/100
Epoch 21/100
Epoch 22/100
Epoch 23/100
Epoch 24/100
Epoch 25/100
Epoch 26/100
Epoch 27/100
Epoch 28/100
Epoch 29/100
Epoch 30/100
Epoch 31/100
Epoch 32/100
Epoch 33/100
Epoch 34/100
Epoch 35/100
Epoch 36/100
Epoch 37/100
Epoch 38/100
Epoch 39/100
Epoch 40/100
Epoch 41/100
```

Epoch 42/100

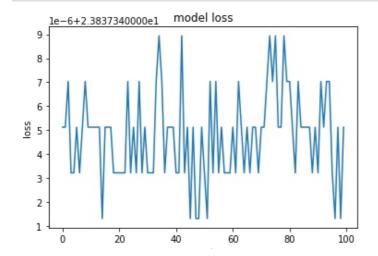
```
Epoch 43/100
Epoch 44/100
Epoch 45/100
Epoch 46/100
Epoch 47/100
Epoch 48/100
Epoch 49/100
Epoch 50/100
Epoch 51/100
Epoch 52/100
Epoch 53/100
Epoch 54/100
Epoch 55/100
Epoch 56/100
Epoch 57/100
Epoch 58/100
Epoch 59/100
Epoch 60/100
Epoch 61/100
Epoch 62/100
Epoch 63/100
Epoch 64/100
Epoch 65/100
Epoch 66/100
Epoch 67/100
Epoch 68/100
Epoch 69/100
Epoch 70/100
Epoch 71/100
Epoch 72/100
Epoch 73/100
Epoch 74/100
Epoch 75/100
Epoch 76/100
Epoch 77/100
```

Epoch 78/100

```
Epoch 79/100
Epoch 80/100
Epoch 81/100
Epoch 82/100
Epoch 83/100
Epoch 84/100
Epoch 85/100
Epoch 86/100
Epoch 87/100
Epoch 88/100
Epoch 89/100
Epoch 90/100
Epoch 91/100
Epoch 92/100
Epoch 93/100
Os - loss: 23.4459 - accuracy: 0.2
Epoch 94/100
Epoch 95/100
Epoch 96/100
Epoch 97/100
Epoch 98/100
Epoch 99/100
Epoch 100/100
```

In [207]:

```
import matplotlib.pyplot as plt
plt.plot(train9.history['loss'])
plt.title('model loss')
plt.ylabel('loss')
plt.xlabel('epoch')
plt.show()
```



epoch

Epoch 3/100

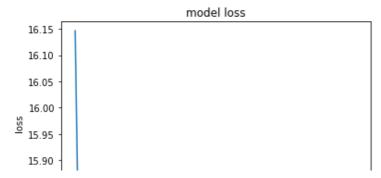
```
In [208]:
results = model9.evaluate(X test, y test)
print(model9.metrics names)
print(results)
print("The Accuracy is ",results[1]*100,"%")
['loss', 'accuracy']
[24.192813873291016, 0.22280701994895935]
The Accuracy is 22.280701994895935 %
Identity
In [209]:
model10=Sequential()
In [210]:
model10.add(Dense(10, input shape = (6,), activation = 'sigmoid', kernel initializer=init
ializer10))
model10.add(Dense(10, activation = 'sigmoid', kernel initializer=initializer10))
model10.add(Dense(10, activation = 'sigmoid', kernel_initializer=initializer10))
model10.add(Dense(10, activation = 'sigmoid', kernel initializer=initializer10))
model10.add(Dense(10, activation = 'sigmoid', kernel initializer=initializer10))
model10.add(Dense(4))
In [211]:
model10.summary()
Model: "sequential 17"
Layer (type)
                        Output Shape
                                              Param #
dense 96 (Dense)
                        (None, 10)
                                              70
                        (None, 10)
dense 97 (Dense)
                                              110
dense 98 (Dense)
                        (None, 10)
                                              110
dense 99 (Dense)
                        (None, 10)
                                              110
dense 100 (Dense)
                        (None, 10)
                                              110
dense 101 (Dense)
                        (None, 4)
                                              44
______
Total params: 554
Trainable params: 554
Non-trainable params: 0
In [212]:
from tensorflow.keras import optimizers
sgd10 = optimizers.SGD(learning rate = 0.01)
model10.compile(optimizer = sgd10, loss = 'binary_crossentropy', metrics = ['accuracy'])
In [213]:
train10=model10.fit(X train, y train, batch size = 50, epochs = 100, verbose = 1)
Epoch 1/100
Epoch 2/100
```

```
Epoch 4/100
Epoch 5/100
Epoch 7/100
Epoch 8/100
Epoch 9/100
Epoch 10/100
Epoch 11/100
Epoch 12/100
Epoch 13/100
Epoch 14/100
Epoch 15/100
Epoch 16/100
Epoch 17/100
Epoch 18/100
Epoch 19/100
Epoch 20/100
Epoch 21/100
Epoch 22/100
Epoch 23/100
Epoch 24/100
Epoch 25/100
Epoch 26/100
Epoch 27/100
Epoch 28/100
Epoch 29/100
Epoch 30/100
Epoch 31/100
Epoch 32/100
Epoch 33/100
Epoch 34/100
Epoch 35/100
Epoch 36/100
Epoch 37/100
Epoch 38/100
Epoch 39/100
```

```
Epoch 40/100
Epoch 41/100
Epoch 42/100
Epoch 43/100
Epoch 44/100
Epoch 45/100
Epoch 46/100
Epoch 47/100
Epoch 48/100
Epoch 49/100
Epoch 50/100
Epoch 51/100
Epoch 52/100
Epoch 53/100
Epoch 54/100
Epoch 55/100
Epoch 56/100
Epoch 57/100
Epoch 58/100
Epoch 59/100
Epoch 60/100
Epoch 61/100
Epoch 62/100
Epoch 63/100
Epoch 64/100
Epoch 65/100
Epoch 66/100
Epoch 67/100
Epoch 68/100
Epoch 69/100
Epoch 70/100
Epoch 71/100
Epoch 72/100
Epoch 73/100
Epoch 74/100
Epoch 75/100
```

```
Epoch 76/100
Epoch 77/100
Epoch 78/100
Epoch 79/100
Epoch 80/100
Epoch 81/100
Epoch 82/100
Epoch 83/100
Epoch 84/100
Epoch 85/100
Epoch 86/100
Epoch 87/100
Epoch 88/100
Epoch 89/100
Epoch 90/100
Epoch 91/100
Epoch 92/100
Epoch 93/100
Epoch 94/100
Epoch 95/100
Epoch 96/100
Epoch 97/100
Epoch 98/100
Epoch 99/100
Epoch 100/100
In [214]:
```

```
import matplotlib.pyplot as plt
plt.plot(train10.history['loss'])
plt.title('model loss')
plt.ylabel('loss')
plt.xlabel('epoch')
plt.show()
```



```
15.85 - 15.80 - 20 40 60 80 100 epoch
```

```
In [215]:
```

Orthogonal

In [216]:

```
model11=Sequential()
```

In [217]:

```
model11.add(Dense(10, input_shape = (6,), activation = 'sigmoid', kernel_initializer=init
ializer11))
model11.add(Dense(10, activation = 'sigmoid', kernel_initializer=initializer11))
model11.add(Dense(4))
```

In [218]:

```
model11.summary()
```

Model: "sequential 18"

Layer (type)	Output	Shape	Param #				
dense_102 (Dense)	(None,	10)	70				
dense_103 (Dense)	(None,	10)	110				
dense_104 (Dense)	(None,	10)	110				
dense_105 (Dense)	(None,	10)	110				
dense_106 (Dense)	(None,	10)	110				
dense_107 (Dense)	(None,	4)	44				
Total params: 554							

Trainable params: 554
Non-trainable params: 0

In [219]:

```
sgd11 = optimizers.SGD(learning_rate = 0.01)
model11.compile(optimizer = sgd11, loss = 'binary_crossentropy', metrics = ['accuracy'])
```

In [220]:

```
train11=model11.fit(X_train, y_train, batch_size = 50, epochs = 100, verbose = 1)
```

Epoch 1/100

```
Epoch 2/100
Epoch 3/100
Epoch 4/100
Epoch 5/100
Epoch 6/100
Epoch 7/100
Epoch 8/100
Epoch 9/100
Epoch 10/100
Epoch 11/100
Epoch 12/100
Epoch 13/100
Epoch 14/100
Epoch 15/100
Epoch 16/100
Epoch 17/100
Epoch 18/100
Epoch 19/100
Epoch 20/100
Epoch 21/100
Epoch 22/100
Epoch 23/100
Epoch 24/100
Epoch 25/100
Epoch 26/100
Epoch 27/100
Epoch 28/100
Epoch 29/100
Epoch 30/100
Epoch 31/100
Epoch 32/100
Epoch 33/100
Epoch 34/100
Epoch 35/100
Epoch 36/100
Epoch 37/100
```

```
Epoch 38/100
Epoch 39/100
Epoch 40/100
Epoch 41/100
Epoch 42/100
Epoch 43/100
Epoch 44/100
Epoch 45/100
Epoch 46/100
Epoch 47/100
Epoch 48/100
Epoch 49/100
Epoch 50/100
Epoch 51/100
Epoch 52/100
Epoch 53/100
Epoch 54/100
Epoch 55/100
Epoch 56/100
Epoch 57/100
Epoch 58/100
Epoch 59/100
Epoch 60/100
Epoch 61/100
Epoch 62/100
Epoch 63/100
Epoch 64/100
Epoch 65/100
Epoch 66/100
Epoch 67/100
Epoch 68/100
Epoch 69/100
Epoch 70/100
Epoch 71/100
Epoch 72/100
Epoch 73/100
```

```
Epoch 74/100
Epoch 75/100
Epoch 76/100
Epoch 77/100
Epoch 78/100
Epoch 79/100
Epoch 80/100
Epoch 81/100
Epoch 82/100
Epoch 83/100
Epoch 84/100
Epoch 85/100
Epoch 86/100
Epoch 87/100
Epoch 88/100
Epoch 89/100
Epoch 90/100
Epoch 91/100
Epoch 92/100
Epoch 93/100
Epoch 94/100
Epoch 95/100
Epoch 96/100
Epoch 97/100
Epoch 98/100
Epoch 99/100
Epoch 100/100
In [221]:
import matplotlib.pyplot as plt
plt.plot(train11.history['loss'])
plt.title('model loss')
plt.ylabel('loss')
```

```
plt.xlabel('epoch')
plt.show()
```

```
model loss
17.4
17.2
17.0
16.8
```

```
16.6

16.4

16.2

16.0

15.8

0 20 40 60 80 100

epoch
```

In [286]:

Random Uniform, Ones, HeNormal, Identity and Orthogonal Gives an accuracy of 70% accuracy

4)

In [243]:

```
modela1=Sequential()
```

In [245]:

```
modela1.add(Dense(10, input_shape = (6,), activation = 'relu', kernel_initializer=initial
izer11))
modela1.add(Dense(10, activation = 'relu', kernel_initializer=initializer11))
modela1.add(Dense(4))
```

In [246]:

```
modelal.summary()
```

Model: "sequential 23"

Layer (typ	oe)	Output	Shape	Param #			
dense_132	(Dense)	(None,	10)	70			
dense_133	(Dense)	(None,	10)	110			
dense_134	(Dense)	(None,	10)	110			
dense_135	(Dense)	(None,	10)	110			
dense_136	(Dense)	(None,	10)	110			
dense_137	(Dense)	(None,	4)	44			
Total params: 554							

Trainable params: 554
Non-trainable params: 0

In [247]:

```
sgda1 = optimizers.SGD(learning_rate = 0.01)
```

In [248]:

```
trainal=modela1.fit(X train, y train, batch size = 50, epochs = 100, verbose = 1)
Epoch 1/100
Epoch 2/100
Epoch 3/100
Epoch 4/100
Epoch 5/100
Epoch 6/100
Epoch 7/100
Epoch 8/100
Epoch 9/100
Epoch 10/100
Epoch 11/100
Epoch 12/100
Epoch 13/100
Epoch 14/100
Epoch 15/100
Epoch 16/100
Epoch 17/100
Epoch 18/100
Epoch 19/100
Epoch 20/100
Epoch 21/100
Epoch 22/100
Epoch 23/100
Epoch 24/100
Epoch 25/100
Epoch 26/100
Epoch 27/100
Epoch 28/100
Epoch 29/100
Epoch 30/100
Epoch 31/100
Epoch 32/100
Epoch 33/100
```

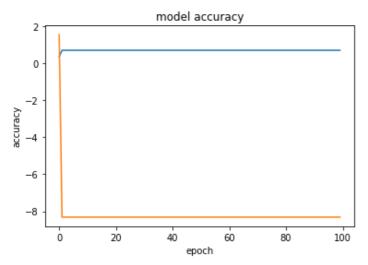
```
Epoch 34/100
Epoch 35/100
Epoch 36/100
Epoch 37/100
Epoch 38/100
Epoch 39/100
Epoch 40/100
Epoch 41/100
Os - loss: -7.7771 - accuracy: 0.6
Epoch 42/100
Epoch 43/100
Epoch 44/100
Epoch 45/100
Epoch 46/100
Epoch 47/100
Epoch 48/100
Epoch 49/100
Epoch 50/100
Epoch 51/100
Epoch 52/100
Epoch 53/100
Epoch 54/100
Epoch 55/100
Epoch 56/100
Epoch 57/100
Epoch 58/100
Epoch 59/100
Epoch 60/100
Epoch 61/100
Epoch 62/100
Epoch 63/100
Epoch 64/100
Epoch 65/100
Epoch 66/100
Epoch 67/100
Epoch 68/100
Epoch 69/100
```

```
Epoch 70/100
Epoch 71/100
Epoch 72/100
Epoch 73/100
Epoch 74/100
Epoch 75/100
Epoch 76/100
Epoch 77/100
Epoch 78/100
Epoch 79/100
Epoch 80/100
Epoch 81/100
Epoch 82/100
Epoch 83/100
Epoch 84/100
Epoch 85/100
Epoch 86/100
Epoch 87/100
Epoch 88/100
Epoch 89/100
Epoch 90/100
Epoch 91/100
Epoch 92/100
Epoch 93/100
Epoch 94/100
Epoch 95/100
Epoch 96/100
Epoch 97/100
Epoch 98/100
Epoch 99/100
Epoch 100/100
```

In [295]:

```
import matplotlib.pyplot as plt
plt.plot(trainal.history['accuracy'])
plt.plot(trainal.history['loss'])
plt.title('model accuracy')
plt.ylabel('accuracy')
```

```
plt.xlabel('epoch')
plt.show()
```



In [277]:

[-8.667988777160645, 0.699999988079071] The Accuracy is 69.9999988079071 %

5)

In [265]:

```
modelb1=Sequential()
```

In [266]:

```
modelb1.add(Dense(10, input_shape = (6,), activation = 'relu', kernel_initializer=initial
izer11))
modelb1.add(Dense(10, activation = 'relu', kernel_initializer=initializer11))
modelb1.add(Dense(4))
```

In [267]:

```
modelb1.summary()
```

Model: "sequential 26"

Layer (typ	pe)	Output	Shape	Param #
dense_150	(Dense)	(None,	10)	70
dense_151	(Dense)	(None,	10)	110
dense_152	(Dense)	(None,	10)	110
dense_153	(Dense)	(None,	10)	110
dense_154	(Dense)	(None,	10)	110
dense_155	(Dense)	(None,	4)	44
	554			

Total params: 554
Trainable params: 554

Non-trainable params: 0

In [269]:

```
adam = optimizers.Adam(learning_rate = 0.01)
modelb1.compile(optimizer = adam, loss = 'binary_crossentropy', metrics = ['accuracy'])
```

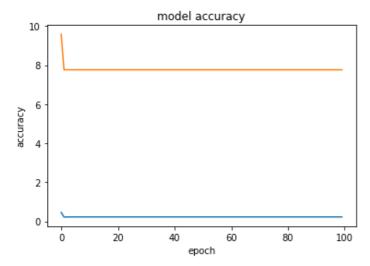
In [270]:

```
trainb1=modelb1.fit(X_train, y_train, batch_size = 50, epochs = 100, verbose = 1)
Epoch 1/100
Epoch 2/100
Epoch 3/100
Epoch 4/100
Epoch 5/100
Epoch 6/100
Epoch 7/100
Epoch 8/100
Epoch 9/100
Epoch 10/100
Epoch 11/100
Epoch 12/100
Epoch 13/100
Epoch 14/100
Epoch 15/100
Epoch 16/100
Epoch 17/100
Epoch 18/100
Epoch 19/100
Epoch 20/100
Epoch 21/100
Epoch 22/100
Epoch 23/100
Epoch 24/100
Epoch 25/100
Epoch 26/100
Epoch 27/100
Epoch 28/100
Epoch 29/100
Epoch 30/100
```

```
Epoch 31/100
Epoch 32/100
Epoch 33/100
Epoch 34/100
Epoch 35/100
Epoch 36/100
Epoch 37/100
Epoch 38/100
Epoch 39/100
Epoch 40/100
Epoch 41/100
Epoch 42/100
Epoch 43/100
Epoch 44/100
Epoch 45/100
Epoch 46/100
Epoch 47/100
Epoch 48/100
Epoch 49/100
Epoch 50/100
Epoch 51/100
Epoch 52/100
Epoch 53/100
Epoch 54/100
Epoch 55/100
Epoch 56/100
Epoch 57/100
Epoch 58/100
Epoch 59/100
Epoch 60/100
Epoch 61/100
Epoch 62/100
Epoch 63/100
Epoch 64/100
Epoch 65/100
Epoch 66/100
```

	[_	Vδ	amp/preh	-	TO22:	/ • / ७०५	_	асситасу:	U. ZJIU
24/24	67/100 [=======]	_	0s	8ms/step	_	loss:	7.7604	-	accuracy:	0.2316
-	68/100 [========]	_	0s	8ms/step	_	loss:	7.7604	_	accuracy:	0.2316
-	69/100 [======]	_	0s	8ms/step	_	loss:	7.7604	_	accuracy:	0.2316
Epoch	70/100 [=======]			_					_	
Epoch	71/100			-					_	
Epoch	[=========] 72/100									
	[=====================================	-	0s	8ms/step	-	loss:	7.7604	-	accuracy:	0.2316
	[=======] 74/100	-	0s	8ms/step	-	loss:	7.7604	-	accuracy:	0.2316
24/24	[=====================================	-	0s	8ms/step	-	loss:	7.7604	-	accuracy:	0.2316
24/24	[=======] 76/100	-	0s	8ms/step	-	loss:	7.7604	-	accuracy:	0.2316
24/24	[=======]	-	0s	8ms/step	-	loss:	7.7604	-	accuracy:	0.2316
24/24	77/100 [=======]	_	0s	8ms/step	_	loss:	7.7604	_	accuracy:	0.2316
	78/100 [=======]	_	0s	8ms/step	_	loss:	7.7604	_	accuracy:	0.2316
-	79/100	_	0s	8ms/step	_	loss:	7.7604	_	accuracy:	0.2316
Epoch	80/100 [=======]			_					_	
Epoch	81/100 [======]									
Epoch	82/100									
Epoch	[=====================================									
	[=====================================	-	0s	8ms/step	-	loss:	7.7604	-	accuracy:	0.2316
	[=======] 85/100	-	0s	8ms/step	-	loss:	7.7604	-	accuracy:	0.2316
24/24	[=====================================	-	0s	8ms/step	-	loss:	7.7604	-	accuracy:	0.2316
24/24	[=======]	-	0s	8ms/step	-	loss:	7.7604	-	accuracy:	0.2316
24/24	87/100 [========]	-	0s	9ms/step	_	loss:	7.7604	-	accuracy:	0.2316
24/24	88/100 [=======]	_	0s	8ms/step	_	loss:	7.7604	_	accuracy:	0.2316
-	89/100 [========]	_	0s	8ms/step	_	loss:	7.7604	_	accuracy:	0.2316
	90/100	_	0s	8ms/step	_	loss:	7.7604	_	accuracy:	0.2316
Epoch	91/100									
Epoch	92/100 [=======]			_					_	
Epoch	93/100			_					_	
Epoch	[========] 94/100			_					_	
Epoch	[=====================================									
	[======] 96/100	-	0s	8ms/step	-	loss:	7.7604	-	accuracy:	0.2316
24/24	[=====================================	-	0s	9ms/step	-	loss:	7.7604	-	accuracy:	0.2316
24/24	[========] 98/100	-	0s	8ms/step	-	loss:	7.7604	-	accuracy:	0.2316
24/24	[=======]	-	0s	8ms/step	-	loss:	7.7604	-	accuracy:	0.2316
24/24	99/100	-	0s	8ms/step	-	loss:	7.7604	-	accuracy:	0.2316
	100/100	_	0s	8ms/step	_	loss:	7.7604	_	accuracy:	0.2316

```
import matplotlib.pyplot as plt
plt.plot(trainbl.history['accuracy'])
plt.plot(trainbl.history['loss'])
plt.title('model accuracy')
plt.ylabel('accuracy')
plt.xlabel('epoch')
plt.show()
```



In [282]:

Write down your inferences.

[7.7624125480651855, 0.224561408162117] The Accuracy is 22.4561408162117 %

From the above experiments we infer that the relu function gives an accuracy of 69% with Orthogonal Initializer.