SURIYA S

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

Out[2]:

	age	sex	ср	trestbps	chol	fbs	restecg	thalach	exang	oldpeak	slope	са	thal	target
0	63	1	3	145	233	1	0	150	0	2.3	0	0	1	1
1	37	1	2	130	250	0	1	187	0	3.5	0	0	2	1
2	41	0	1	130	204	0	0	172	0	1.4	2	0	2	1
3	56	1	1	120	236	0	1	178	0	0.8	2	0	2	1
4	57	0	0	120	354	0	1	163	1	0.6	2	0	2	1

part 2

```
In [3]: X = data
y = data.pop('target')
```

In [4]: X

Out[4]:

	age	sex	ср	trestbps	chol	fbs	restecg	thalach	exang	oldpeak	slope	са	thal
0	63	1	3	145	233	1	0	150	0	2.3	0	0	1
1	37	1	2	130	250	0	1	187	0	3.5	0	0	2
2	41	0	1	130	204	0	0	172	0	1.4	2	0	2
3	56	1	1	120	236	0	1	178	0	0.8	2	0	2
4	57	0	0	120	354	0	1	163	1	0.6	2	0	2
				•••									
298	57	0	0	140	241	0	1	123	1	0.2	1	0	3
299	45	1	3	110	264	0	1	132	0	1.2	1	0	3
300	68	1	0	144	193	1	1	141	0	3.4	1	2	3
301	57	1	0	130	131	0	1	115	1	1.2	1	1	3
302	57	0	1	130	236	0	0	174	0	0.0	1	1	2

303 rows × 13 columns

```
In [ ]:
```

```
In [5]: from sklearn.model_selection import train_test_split
```

```
In [6]: X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.33, random]
In [7]: X_train.shape
Out[7]: (203, 13)
In [8]: X_test.shape
Out[8]: (100, 13)
```

part 3

```
In [9]: from tensorflow.keras.models import Sequential
from tensorflow.keras.layers import Dense
```

```
In [10]: model = Sequential()
    model.add(Dense(8, input_dim=13, activation='relu'))
    model.add(Dense(1, activation='sigmoid'))
```

```
In [11]: from tensorflow import keras
```

```
In [12]: optimizer = keras.optimizers.RMSprop(learning_rate=0.001)
```

```
In [13]: model.compile(loss='mse', optimizer=optimizer, metrics=['accuracy'])
       model.fit(X_train, y_train, epochs=10, batch_size=30, verbose=1)
       Epoch 1/10
       0.5271
       Epoch 2/10
       0.5271
       Epoch 3/10
       7/7 [================ ] - 0s 2ms/step - loss: 0.4729 - accuracy:
       0.5271
       Epoch 4/10
       7/7 [============ ] - 0s 2ms/step - loss: 0.4729 - accuracy:
       0.5271
       Epoch 5/10
       7/7 [============== ] - 0s 2ms/step - loss: 0.4729 - accuracy:
       0.5271
       Epoch 6/10
       7/7 [================ ] - 0s 1ms/step - loss: 0.4729 - accuracy:
       0.5271
       Epoch 7/10
       7/7 [=============== ] - 0s 1ms/step - loss: 0.4729 - accuracy:
       0.5271
       Epoch 8/10
       7/7 [============== ] - 0s 1ms/step - loss: 0.4729 - accuracy:
       0.5271
       Epoch 9/10
       7/7 [=============== ] - 0s 1ms/step - loss: 0.4729 - accuracy:
       0.5271
       Epoch 10/10
       7/7 [=============== ] - 0s 2ms/step - loss: 0.4729 - accuracy:
       0.5271
Out[13]: <keras.callbacks.History at 0x21433922380>
In [14]: model.evaluate(X test, y test)
       4/4 [=========== ] - 0s 2ms/step - loss: 0.4200 - accuracy:
       0.5800
Out[14]: [0.41999998688697815, 0.5799999833106995]
```

```
In [15]: model.summary()
```

Model: "sequential"

Layer (type)	Output Shape	Param #
dense (Dense)	(None, 8)	112
dense_1 (Dense)	(None, 1)	9

Total params: 121 Trainable params: 121 Non-trainable params: 0

part 6

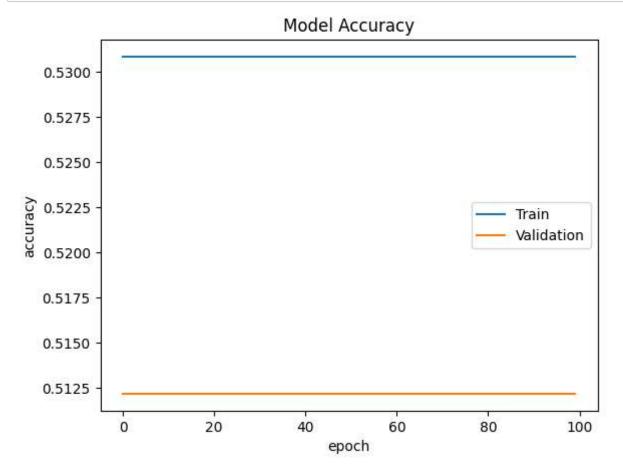
```
In [16]: model.compile(loss='mse', optimizer=optimizer, metrics=['accuracy'])
        model.fit(X_train, y_train, epochs=200, batch_size=10, verbose=1)
        y: 0.5271
        Epoch 20/200
        21/21 [=============== ] - 0s 1ms/step - loss: 0.4729 - accurac
        y: 0.5271
        Epoch 21/200
        21/21 [============ ] - 0s 1ms/step - loss: 0.4729 - accurac
        y: 0.5271
        Epoch 22/200
        21/21 [============ ] - 0s 1ms/step - loss: 0.4729 - accurac
        y: 0.5271
        Epoch 23/200
        21/21 [============ ] - 0s 1ms/step - loss: 0.4729 - accurac
        y: 0.5271
        Epoch 24/200
        21/21 [============== ] - 0s 1ms/step - loss: 0.4729 - accurac
        y: 0.5271
        Epoch 25/200
        21/21 [============ ] - 0s 1ms/step - loss: 0.4729 - accurac
        v: 0.5271
        Enach 26/200
In [17]: model.evaluate(X_test, y_test)
        4/4 [============= ] - 0s 2ms/step - loss: 0.4200 - accuracy:
        0.5800
Out[17]: [0.41999998688697815, 0.5799999833106995]
```

```
In [18]: history = model.fit(X_train, y_train, validation_split=0.2, epochs=100, batch_siz
        y: 0.5309 - val loss: 0.4878 - val accuracy: 0.5122
        Epoch 10/100
        17/17 [============= ] - 0s 3ms/step - loss: 0.4691 - accurac
        y: 0.5309 - val_loss: 0.4878 - val_accuracy: 0.5122
        Epoch 11/100
        17/17 [================ ] - 0s 3ms/step - loss: 0.4691 - accurac
        y: 0.5309 - val_loss: 0.4878 - val_accuracy: 0.5122
        Epoch 12/100
        17/17 [============ ] - 0s 3ms/step - loss: 0.4691 - accurac
        y: 0.5309 - val_loss: 0.4878 - val_accuracy: 0.5122
        Epoch 13/100
        y: 0.5309 - val_loss: 0.4878 - val_accuracy: 0.5122
        Epoch 14/100
        17/17 [============ ] - 0s 3ms/step - loss: 0.4691 - accurac
        y: 0.5309 - val_loss: 0.4878 - val_accuracy: 0.5122
        Epoch 15/100
        17/17 [============ ] - 0s 3ms/step - loss: 0.4691 - accurac
        y: 0.5309 - val loss: 0.4878 - val accuracy: 0.5122
        Epoch 16/100
```

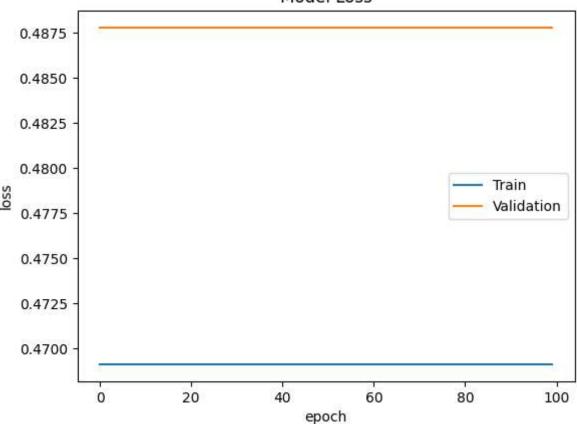
part 8

```
In [20]: history.history.keys()
Out[20]: dict_keys(['loss', 'accuracy', 'val_loss', 'val_accuracy'])
In [21]: import matplotlib.pyplot as plt
```

```
In [22]: plt.plot(history.history['accuracy'])
    plt.plot(history.history['val_accuracy'])
    plt.title('Model Accuracy')
    plt.ylabel('accuracy')
    plt.legend(['Train', 'Validation'])
    plt.show()
    plt.plot(history.history['loss'])
    plt.plot(history.history['val_loss'])
    plt.title('Model Loss')
    plt.ylabel('loss')
    plt.xlabel('epoch')
    plt.legend(['Train', 'Validation'])
    plt.show()
```



Model Loss



```
In [23]: model1 = Sequential()

model1.add(Dense(16, input_dim=13, activation='relu'))
model1.add(Dense(8, activation='relu'))
model1.add(Dense(1, activation='sigmoid'))
```

```
In [24]:
       model1.compile(loss='mse', optimizer=optimizer, metrics=['accuracy'])
       model1.fit(X train, y train, epochs=10, batch size=30, verbose=1)
       Epoch 1/10
       0.4729
       Epoch 2/10
       0.4729
       Epoch 3/10
       7/7 [================= ] - 0s 2ms/step - loss: 0.5271 - accuracy:
       0.4729
       Epoch 4/10
       7/7 [============ ] - 0s 2ms/step - loss: 0.5271 - accuracy:
       0.4729
       Epoch 5/10
       7/7 [============== ] - 0s 2ms/step - loss: 0.5271 - accuracy:
       0.4729
       Epoch 6/10
       7/7 [================ ] - 0s 1ms/step - loss: 0.5239 - accuracy:
       0.4680
       Epoch 7/10
       7/7 [================ ] - 0s 1ms/step - loss: 0.4925 - accuracy:
       0.4039
       Epoch 8/10
       7/7 [=============== ] - 0s 1ms/step - loss: 0.3548 - accuracy:
       0.5074
       Epoch 9/10
       7/7 [=============== ] - 0s 1ms/step - loss: 0.2705 - accuracy:
       0.5813
       Epoch 10/10
       7/7 [=============== ] - 0s 2ms/step - loss: 0.2505 - accuracy:
       0.5862
Out[24]: <keras.callbacks.History at 0x214389a1630>
In [25]: |model1.evaluate(X test, y test)
       4/4 [=========== ] - 0s 2ms/step - loss: 0.2092 - accuracy:
       0.6400
Out[25]: [0.20918434858322144, 0.6399999856948853]
```

```
In [26]: history1 = model.fit(X_train, y_train, validation_split=0.2, epochs=100, batch_s:
        y: 0.5309 - val_loss: 0.4878 - val_accuracy: 0.5122
        Epoch 9/100
        17/17 [============= ] - 0s 3ms/step - loss: 0.4691 - accurac
        y: 0.5309 - val_loss: 0.4878 - val_accuracy: 0.5122
        Epoch 10/100
        17/17 [============= ] - 0s 3ms/step - loss: 0.4691 - accurac
        y: 0.5309 - val_loss: 0.4878 - val_accuracy: 0.5122
        Epoch 11/100
        17/17 [============ ] - 0s 3ms/step - loss: 0.4691 - accurac
        y: 0.5309 - val_loss: 0.4878 - val_accuracy: 0.5122
        Epoch 12/100
        17/17 [=========== ] - 0s 3ms/step - loss: 0.4691 - accurac
        y: 0.5309 - val_loss: 0.4878 - val_accuracy: 0.5122
        Epoch 13/100
        17/17 [============ ] - 0s 3ms/step - loss: 0.4691 - accurac
        y: 0.5309 - val_loss: 0.4878 - val_accuracy: 0.5122
        Epoch 14/100
        17/17 [============== ] - 0s 3ms/step - loss: 0.4691 - accurac
        y: 0.5309 - val_loss: 0.4878 - val_accuracy: 0.5122
        Epoch 15/100
```

In [27]: model1.summary()

Model: "sequential_1"

Layer (type)	Output Shape	Param #
dense_2 (Dense)	(None, 16)	224
dense_3 (Dense)	(None, 8)	136
dense_4 (Dense)	(None, 1)	9

Total params: 369
Trainable params: 369
Non-trainable params: 0

In [30]: ls = history1.history

Out[29]:

	loss	accuracy	val_loss	val_accuracy
0	0.469136	0.530864	0.487805	0.512195
1	0.469136	0.530864	0.487805	0.512195
2	0.469136	0.530864	0.487805	0.512195
3	0.469136	0.530864	0.487805	0.512195
4	0.469136	0.530864	0.487805	0.512195
95	0.469136	0.530864	0.487805	0.512195
96	0.469136	0.530864	0.487805	0.512195
97	0.469136	0.530864	0.487805	0.512195
98	0.469136	0.530864	0.487805	0.512195
99	0.469136	0.530864	0.487805	0.512195

100 rows × 4 columns

```
In [31]: model2 = Sequential()
    model2.add(Dense(32, input_dim=13, activation='relu'))
    model2.add(Dense(16, activation='relu'))
    model2.add(Dense(8, activation='relu'))
    model2.add(Dense(1, activation='sigmoid'))
```

```
In [32]: model2.compile(loss='mse', optimizer=optimizer, metrics=['accuracy'])
       model2.fit(X train, y train, epochs=10, batch size=30, verbose=1)
       Epoch 1/10
       0.4729
       Epoch 2/10
       0.5567
       Epoch 3/10
       7/7 [================ ] - 0s 1ms/step - loss: 0.3305 - accuracy:
       0.5862
       Epoch 4/10
       7/7 [============ ] - 0s 1ms/step - loss: 0.2764 - accuracy:
       0.6059
       Epoch 5/10
       7/7 [============== ] - 0s 1ms/step - loss: 0.2707 - accuracy:
       0.6256
       Epoch 6/10
       7/7 [================ ] - 0s 2ms/step - loss: 0.2521 - accuracy:
       0.6207
       Epoch 7/10
       7/7 [=============== ] - 0s 2ms/step - loss: 0.2707 - accuracy:
       0.6158
       Epoch 8/10
       7/7 [=============== ] - 0s 2ms/step - loss: 0.2417 - accuracy:
       0.6798
       Epoch 9/10
       7/7 [=============== ] - 0s 2ms/step - loss: 0.2416 - accuracy:
       0.6502
       Epoch 10/10
       7/7 [=============== ] - 0s 2ms/step - loss: 0.2411 - accuracy:
       0.6207
Out[32]: <keras.callbacks.History at 0x21439b2fa60>
In [34]: model2.evaluate(X test, y test)
       4/4 [=========== ] - 0s 2ms/step - loss: 0.1724 - accuracy:
       0.7300
Out[34]: [0.17240825295448303, 0.7300000190734863]
```

In [35]: model2.summary()

Model: "sequential_2"

Layer (type)	Output Shape	Param #
dense_5 (Dense)	(None, 32)	448
dense_6 (Dense)	(None, 16)	528
dense_7 (Dense)	(None, 8)	136
dense_8 (Dense)	(None, 1)	9
		========

Total params: 1,121 Trainable params: 1,121 Non-trainable params: 0

```
In [36]: model3 = Sequential()
    model3.add(Dense(64, input_dim=13, activation='relu'))
    model3.add(Dense(32, activation='relu'))
    model3.add(Dense(16, activation='relu'))
    model3.add(Dense(8, activation='relu'))
    model3.add(Dense(1, activation='sigmoid'))
```

```
In [37]: model3.compile(loss='mse', optimizer=optimizer, metrics=['accuracy'])
       model3.fit(X train, y train, epochs=10, batch size=30, verbose=1)
       Epoch 1/10
       0.4926
       Epoch 2/10
       0.5025
       Epoch 3/10
       0.5123
       Epoch 4/10
       7/7 [============ ] - 0s 2ms/step - loss: 0.3367 - accuracy:
       0.5567
       Epoch 5/10
       7/7 [============== ] - 0s 2ms/step - loss: 0.3232 - accuracy:
       0.5764
       Epoch 6/10
       7/7 [=============== ] - 0s 2ms/step - loss: 0.3048 - accuracy:
       0.6059
       Epoch 7/10
       7/7 [=============== ] - 0s 2ms/step - loss: 0.3274 - accuracy:
       0.5813
       Epoch 8/10
       7/7 [============== ] - 0s 2ms/step - loss: 0.3085 - accuracy:
       0.5714
       Epoch 9/10
       7/7 [============== ] - 0s 2ms/step - loss: 0.3222 - accuracy:
       0.6010
       Epoch 10/10
       7/7 [=============== ] - 0s 2ms/step - loss: 0.2586 - accuracy:
       0.6355
Out[37]: <keras.callbacks.History at 0x2143ae7ceb0>
In [38]: model3.evaluate(X test, y test)
       4/4 [=========== ] - 0s 2ms/step - loss: 0.1892 - accuracy:
       0.7200
Out[38]: [0.18918751180171967, 0.7200000286102295]
```

In [39]: model3.summary()

Model: "sequential_3"

Layer (type)	Output Shape	Param #
dense_9 (Dense)	(None, 64)	896
dense_10 (Dense)	(None, 32)	2080
dense_11 (Dense)	(None, 16)	528
dense_12 (Dense)	(None, 8)	136
dense_13 (Dense)	(None, 1)	9

Total params: 3,649 Trainable params: 3,649 Non-trainable params: 0

In []: