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Branch/Sec : IT-A

Lab Experiment 6

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In [2]: # Import the necessary libraries
import pandas as pd
import numpy as np
from sklearn.model_selection import train_test_split
from sklearn.preprocessing import StandardScaler
from keras.models import Sequential
from keras.layers import Dense
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In [3]: # Step 1: Load the data
bank_data = pd.read_csv("BankMarketing.csv", delimiter=";")
print(bank_data.head())
```

	age	job	marital	education	default	balance	housing	loan	\
0	58	management	married	tertiary	no	2143	yes	no	
1	44	technician	single	secondary	no	29	yes	no	
2	33	entrepreneur	married	secondary	no	7	yes	yes	
3	47	blue-collar	married	unknown	no	1506	yes	no	

	contact	day	month	duration	campaign	pdays	previous	y
0	unknown	5	may	151	1	-1	0	unknown
1	unknown	5	may	286	1	-1	0	unknown
2	unknown	5	may	92	1	-1	0	unknown
3	unknown	5	may	198	1	-1	0	unknown

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In [4]: # Step 2: Preprocess the data
# Convert categorical variables to numerical variables using one-hot encoding
bank_data = pd.get_dummies(bank_data, columns=["job", "marital", "education", "default"])
# Convert target variable from yes/no to 1/0
bank_data["y"] = bank_data["y"].map({"yes": 1, "no": 0})
```

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In [5]: # Step 3: Split the data into training and testing sets
X = bank_data.drop("y", axis=1)
y = bank_data["y"]
X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.2, random_state=42)
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In [6]: # Step 4: Build the feedforward neural network model
model = Sequential()
model.add(Dense(64, input_dim=X_train.shape[1], activation="relu"))
model.add(Dense(32, activation="relu"))
model.add(Dense(1, activation="sigmoid"))
model.compile(loss="binary_crossentropy", optimizer="adam", metrics=["accuracy"])

In [7]: # Step 5: Train the model on the training data
scaler = StandardScaler()
X_train = scaler.fit_transform(X_train)
X_test = scaler.transform(X_test)
model.fit(X_train, y_train, epochs=50, batch_size=32, validation_data=(X_test, y_test))
Epoch 45/50
1131/1131 [=====] - 2s 2ms/step - loss: 0.1298 - accuracy: 0.9446 - val_loss: 0.2913 - val_accuracy: 0.8946
Epoch 46/50
1131/1131 [=====] - 2s 2ms/step - loss: 0.1287 - accuracy: 0.9454 - val_loss: 0.3008 - val_accuracy: 0.8948
Epoch 47/50
1131/1131 [=====] - 2s 2ms/step - loss: 0.1278 - accuracy: 0.9450 - val_loss: 0.2979 - val_accuracy: 0.8943
Epoch 48/50
1131/1131 [=====] - 2s 2ms/step - loss: 0.1263 - accuracy: 0.9463 - val_loss: 0.3062 - val_accuracy: 0.8966
Epoch 49/50
1131/1131 [=====] - 2s 2ms/step - loss: 0.1264 - accuracy: 0.9474 - val_loss: 0.2984 - val_accuracy: 0.8928
Epoch 50/50
1131/1131 [=====] - 2s 2ms/step - loss: 0.1254 - accuracy: 0.9465 - val_loss: 0.3029 - val_accuracy: 0.8955

Out[7]: <keras.callbacks.History at 0x202d25a7310>
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In [8]: # Step 6: Evaluate the model on the testing data
loss, accuracy = model.evaluate(X_test, y_test)
print("Test Loss:", loss)
print("Test Accuracy:", accuracy)

283/283 [=====] - 0s 1ms/step - loss: 0.3029 - accuracy: 0.8955
Test Loss: 0.30293911695480347
Test Accuracy: 0.8954992890357971
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In [ ]:
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