Untitled7.ipynb - Colab 9/29/24, 4:34 PM

Assignment No.: 3

Link to the Kaggle project: https://www.kaggle.com/barelydedicated/bank-customer-churn-modeling Perform following steps:

- 1. Read the dataset.
- 2. Distinguish the feature and target set and divide the data set into training and test sets.
- 3. Normalize the train and test data.
- 4. Initialize and build the model. Identify the points of improvement and implement the same.
- 5. Print the accuracy score and confusion matrix (5 points).

```
# 1.Read the dataset
import pandas as pd
data = pd.read_csv("/content/Churn_Modelling.csv")
data.head()
\rightarrow
      Show hidden output
data.isnull().sum()
\overline{\rightarrow}
      Show hidden output
data=data.replace({'Female': 0, 'Male': 1})
data.head()
```

→		RowNumber	CustomerId	Surname	CreditScore	Geography	Gender	Age	Tenure	Balance	Nu
	0	1	15634602	Hargrave	619	France	0	42	2	0.00	
	1	2	15647311	Hill	608	Spain	0	41	1	83807.86	
	2	3	15619304	Onio	502	France	0	42	8	159660.80	
	3	4	15701354	Boni	699	France	0	39	1	0.00	
	4	5	15737888	Mitchell	850	Spain	0	43	2	125510.82	
	4										•

Next steps:

Generate code with data



View recommended plots

New interactive sheet

2.Distinguish the feature and target set and divide the data set into training and from sklearn.model_selection import train_test_split

```
# Define feature set (X) and target variable (y)
X = data.drop(columns=['RowNumber', 'CustomerId', 'Surname', 'Geography', 'Exited'])
y = data['Exited']
```

```
# Split the dataset into training and test sets
X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.2, random_stat
```

3.Normalize the train and test data. from sklearn.preprocessing import StandardScaler

Normalize the features

```
scaler = StandardScaler()
X_train_scaled = scaler.fit_transform(X_train)
X_test_scaled = scaler.transform(X_test)
# 4.Initialize and build the model. Identify the points of improvement and implement
from tensorflow.keras.models import Sequential
from tensorflow.keras.layers import Dense
from tensorflow.keras.callbacks import EarlyStopping
# Build the model
model = Sequential()
model.add(Dense(32, activation='relu', input_shape=(X_train_scaled.shape[1],)))
model.add(Dense(16, activation='relu'))
model.add(Dense(1, activation='sigmoid')) # Binary classification
# Compile the model
model.compile(optimizer='adam', loss='binary crossentropy', metrics=['accuracy'])
# Early stopping to prevent overfitting
early stopping = EarlyStopping(monitor='val loss', patience=5, restore best weights=Ti
# Train the model
history = model.fit(X_train_scaled, y_train, epochs=100, validation_split=0.2, callba
\rightarrow
     Show hidden output
   5. Evaluate the Model
from sklearn.metrics import accuracy_score, confusion_matrix
# Make predictions
y pred = (model.predict(X test scaled) > 0.5).astype("int32")
# Calculate accuracy
accuracy = accuracy_score(y_test, y_pred)
print(f"Accuracy: {accuracy:.4f}")
# Confusion matrix
cm = confusion_matrix(y_test, y_pred)
print("Confusion Matrix:")
print(cm)
    63/63 ——
                      Os 2ms/step
    Accuracy: 0.8580
    Confusion Matrix:
    [[1538
           69]
     [ 215 178]]
```