

**Experiment No. 5 **

Aim: Perform Sentiment Analysis in the network graph using RNN.

```

1 import numpy as np
2 import pandas as pd
3 import networkx as nx
4 import tensorflow as tf
5 import matplotlib.pyplot as plt
6 import seaborn as sns
7 from tensorflow.keras.preprocessing.text import Tokenizer
8 from tensorflow.keras.preprocessing.sequence import pad_sequences
9 from tensorflow.keras.models import Sequential
10 from tensorflow.keras.layers import Embedding, LSTM, Dense, SpatialDropout1D
11 from sklearn.model_selection import train_test_split
12 from sklearn.preprocessing import LabelEncoder
13 from sklearn.metrics import confusion_matrix

1 # Sample dataset (Replace with actual dataset)
2 data = {"text": ["I love this product!", "This is the worst experience ever.", "Amazing service and friendly staff.", "I hate waiting in
3           "sentiment": ["positive", "negative", "positive", "negative", "positive"]}
4 df = pd.DataFrame(data)

1 # Convert text data into a network graph
2 G = nx.Graph()
3 for text in df['text']:
4     words = text.lower().split()
5     for i in range(len(words) - 1):
6         G.add_edge(words[i], words[i+1])

1 # Text Preprocessing
2 max_words = 5000 # Vocabulary size
3 max_len = 20     # Max length of each sentence
4
5 tokenizer = Tokenizer(num_words=max_words, oov_token="<OOV>")
6 tokenizer.fit_on_texts(df['text'])
7 X = tokenizer.texts_to_sequences(df['text'])
8 X = pad_sequences(X, maxlen=max_len, padding='post')

1 # Encode sentiment labels
2 encoder = LabelEncoder()
3 y = encoder.fit_transform(df['sentiment'])
4 y = tf.keras.utils.to_categorical(y, num_classes=2)

1 # Train-test split
2 X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.2, random_state=42)

```

```

1 # Build RNN Model
2 model = Sequential([
3     Embedding(input_dim=max_words, output_dim=128, input_length=max_len),
4     SpatialDropout1D(0.2),
5     LSTM(100, dropout=0.2, recurrent_dropout=0.2, return_sequences=False),
6     Dense(2, activation='softmax')
7 ])
8
9 model.compile(loss='categorical_crossentropy', optimizer='adam', metrics=['accuracy'])

```

⚠ /usr/local/lib/python3.11/dist-packages/keras/src/layers/core/embedding.py:90: UserWarning: Argument `input_length` is deprecated. Just warnings.warn()

```

1 # Train Model
2 epochs = 5
3 batch_size = 4
4 model.fit(X_train, y_train, epochs=epochs, batch_size=batch_size, validation_data=(X_test, y_test), verbose=1)

```

⚡ Epoch 1/5
 1/1 ————— 6s 6s/step - accuracy: 0.2500 - loss: 0.7056 - val_accuracy: 0.0000e+00 - val_loss: 0.7353
 Epoch 2/5
 1/1 ————— 0s 124ms/step - accuracy: 0.7500 - loss: 0.6765 - val_accuracy: 0.0000e+00 - val_loss: 0.7932
 Epoch 3/5

```
1/1 ————— 0s 140ms/step - accuracy: 0.7500 - loss: 0.6559 - val_accuracy: 0.0000e+00 - val_loss: 0.8575
Epoch 4/5
1/1 ————— 0s 142ms/step - accuracy: 0.7500 - loss: 0.6217 - val_accuracy: 0.0000e+00 - val_loss: 0.9304
Epoch 5/5
1/1 ————— 0s 135ms/step - accuracy: 0.7500 - loss: 0.6088 - val_accuracy: 0.0000e+00 - val_loss: 1.0112
<keras.src.callbacks.history.History at 0x79d993f32cd0>
```

```
1 # Evaluate Model
2 loss, accuracy = model.evaluate(X_test, y_test, verbose=1)
3 print(f"Test Accuracy: {accuracy * 100:.2f}%")
```

```
1/1 ————— 0s 57ms/step - accuracy: 0.0000e+00 - loss: 1.0112
Test Accuracy: 0.00%
```

```
1 # Predict on test data
2 y_pred = model.predict(X_test)
3 y_pred_classes = np.argmax(y_pred, axis=1)
4 y_true = np.argmax(y_test, axis=1)
```

```
1/1 ————— 0s 38ms/step
```

```
1 # Confusion Matrix
2 cm = confusion_matrix(y_true, y_pred_classes)
3 plt.figure(figsize=(6, 4))
4 sns.heatmap(cm, annot=True, fmt='d', cmap='Blues', xticklabels=["Negative", "Positive"], yticklabels=["Negative", "Positive"])
5 plt.xlabel('Predicted label')
6 plt.ylabel('True label')
7 plt.title('Confusion Matrix')
8 plt.show()
```

