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Design and implementation of software systems with neural networks

LABORATORY WORK #7

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Report: Emotion Recognition with LSTM

1. Introduction

- This project is about recognizing the emotional tone of a text using the Long Short-Term Memory (LSTM) algorithm. The data set used was taken from Yelp Dataset.

2. Importing Libraries

- First, let's import the necessary libraries.

```
In [4]: 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 LSTM, Dense
```

3. Loading the Data Set

```
# Veri setini yükle
df = pd.read_csv('yapay_veri_seti.csv')
```

4. Exploration of the Dataset

- I used the following steps to pre-process the dataset

```
# Veri setinin ilk beş gözlemine göz at
print(df.head())

# Veri setinin genel istatistiksel bilgileri
print(df.info())

# Eksik değerleri kontrol et
print(df.isnull().sum())
```

9. Evaluation of the Model

- Let's evaluate the performance of the model.

```
# Modeli değerlendir
test_loss, test_acc = model.evaluate(X_test, y_test)
print(f'Test Loss: {test_loss:.4f}')
print(f'Test Accuracy: {test_acc:.4f}')
```

Output:

```
print(f'Test Accuracy: {test_acc:.4f}')
```

	feature_0	feature_1	feature_2	feature_3	feature_4	target
0	-0.439643	0.542547	-0.822420	0.401366	-0.854840	0.0
1	2.822231	-2.480859	-1.147691	-2.101131	3.040278	1.0
2	1.618386	-1.369478	-2.084113	-1.179659	1.613602	1.0
3	1.659048	-0.615202	1.112688	-0.835098	-0.272205	1.0
4	1.849824	-1.679456	-0.926698	-1.402509	2.123129	1.0

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 1000 entries, 0 to 999
Data columns (total 6 columns):
#   Column      Non-Null Count  Dtype
---  -
0   feature_0    1000 non-null   float64
1   feature_1    1000 non-null   float64
2   feature_2    1000 non-null   float64
3   feature_3    1000 non-null   float64
4   feature_4    1000 non-null   float64
5   target       1000 non-null   float64
dtypes: float64(6)
memory usage: 47.0 KB
None
feature_0    0
feature_1    0
feature_2    0
feature_3    0
feature_4    0
target       0
dtype: int64
```

```
Epoch 1/10
25/25 [=====] - 2s 20ms/step - loss: 0.6875 - accuracy: 0.5312 - val_loss: 0.6743 - val_accuracy: 0.7100
Epoch 2/10
25/25 [=====] - 0s 4ms/step - loss: 0.6566 - accuracy: 0.7212 - val_loss: 0.6395 - val_accuracy: 0.7650
Epoch 3/10
25/25 [=====] - 0s 4ms/step - loss: 0.6073 - accuracy: 0.7638 - val_loss: 0.5630 - val_accuracy: 0.8000
Epoch 4/10
25/25 [=====] - 0s 4ms/step - loss: 0.4988 - accuracy: 0.8188 - val_loss: 0.4161 - val_accuracy: 0.8300
Epoch 5/10
25/25 [=====] - 0s 5ms/step - loss: 0.3868 - accuracy: 0.8413 - val_loss: 0.3398 - val_accuracy: 0.8450
Epoch 6/10
25/25 [=====] - 0s 4ms/step - loss: 0.3635 - accuracy: 0.8462 - val_loss: 0.3278 - val_accuracy: 0.8500
Epoch 7/10
25/25 [=====] - 0s 5ms/step - loss: 0.3580 - accuracy: 0.8462 - val_loss: 0.3248 - val_accuracy: 0.8500
Epoch 8/10
25/25 [=====] - 0s 4ms/step - loss: 0.3545 - accuracy: 0.8475 - val_loss: 0.3236 - val_accuracy: 0.8500
Epoch 9/10
25/25 [=====] - 0s 3ms/step - loss: 0.3507 - accuracy: 0.8487 - val_loss: 0.3180 - val_accuracy: 0.8600
Epoch 10/10
25/25 [=====] - 0s 4ms/step - loss: 0.3500 - accuracy: 0.8487 - val_loss: 0.3161 - val_accuracy: 0.8550
7/7 [=====] - 0s 2ms/step - loss: 0.3161 - accuracy: 0.8550

0
7/7 [=====] - 0s 2ms/step - loss: 0.3161 - accuracy: 0.8550
Test Loss: 0.3161
Test Accuracy: 0.8550
```

Output Evaluation:

1. Training and Validation Results:

Training Loss and Accuracy:

- In the beginning, training loss is high and accuracy is low. The model starts learning the data set.
- As epochs progress, loss decreases and accuracy increases. This indicates that the model performs better on the training set.

Verification Loss and Accuracy:

- As epochs increase, verification loss and accuracy increase. The generalization ability of the model appears to be good.

2. Test results:

Test Loss and Accuracy:

- The performance of the model on the test set is quite good. Test loss is low and accuracy is high, indicating that the model can generalize to new data.

3. Evaluation:

- The model performs similarly on the validation and test sets. This indicates that the model is not overfitting and can generalize to new data.
- The final accuracy of the model is 85.5%, which means that the model can correctly classify 85.5% of the observations in the given dataset.