

# universe

July 14, 2024

```
[ ]: import pandas as pd
import numpy as np
from keras.models import Sequential
from keras.layers import LSTM, Dense
```

```
[ ]: data = pd.read_csv('/home/darkcover/Documentos/Out/dados/Parte2/matrix.csv')
data.head()
```

```
[ ]:      Unnamed: 0      Listas
0      0  [1, 1, 1, 0, 1, 1, 0, 1, 1, 1, 1, 0, 1, 1, 1, ...
1      1  [1, 1, 0, 1, 1, 1, 0, 0, 0, 1, 1, 1, 1, 1, 1, ...
2      2  [1, 1, 0, 1, 1, 0, 1, 0, 1, 0, 0, 0, 1, 0, 1, ...
3      3  [1, 1, 0, 0, 0, 1, 1, 1, 0, 0, 1, 1, 1, 1, 1, ...
4      4  [1, 1, 0, 1, 1, 1, 1, 0, 0, 1, 1, 1, 1, 0, 1, ...
```

```
[ ]: data = data.drop(columns=['Unnamed: 0'])
data.head()
```

```
[ ]:      Listas
0  [1, 1, 1, 0, 1, 1, 0, 1, 1, 1, 1, 0, 1, 1, 1, ...
1  [1, 1, 0, 1, 1, 1, 0, 0, 0, 1, 1, 1, 1, 1, 1, ...
2  [1, 1, 0, 1, 1, 0, 1, 0, 1, 0, 0, 0, 1, 0, 1, ...
3  [1, 1, 0, 0, 0, 1, 1, 1, 0, 0, 1, 1, 1, 1, 1, ...
4  [1, 1, 0, 1, 1, 1, 1, 0, 0, 1, 1, 1, 1, 0, 1, ...
```

```
[ ]: data = data.drop(data.index[-1])
```

```
[ ]: data.describe
```

```
[ ]: <bound method NDFrame.describe of
Listas
0  [1, 1, 1, 0, 1, 1, 0, 1, 1, 1, 1, 0, 1, 1, 1, ...
1  [1, 1, 0, 1, 1, 1, 0, 0, 0, 1, 1, 1, 1, 1, 1, ...
2  [1, 1, 0, 1, 1, 0, 1, 0, 1, 0, 0, 0, 1, 0, 1, ...
3  [1, 1, 0, 0, 0, 1, 1, 1, 0, 0, 1, 1, 1, 1, 1, ...
4  [1, 1, 0, 1, 1, 1, 1, 0, 0, 1, 1, 1, 1, 0, 1, ...
.. ..
```

```

410 [1, 1, 1, 1, 0, 1, 0, 1, 1, 0, 0, 1, 1, 1, 1, ...
411 [1, 0, 0, 0, 0, 0, 1, 1, 0, 0, 1, 0, 1, 0, 1, ...
412 [1, 0, 1, 0, 0, 0, 1, 1, 1, 0, 0, 1, 1, 1, 0, ...
413 [1, 0, 1, 0, 1, 1, 0, 1, 0, 1, 0, 1, 1, 0, 0, ...
414 [1, 1, 1, 0, 1, 1, 1, 1, 0, 1, 1, 1, 1, 1, 1, ...

```

```
[415 rows x 1 columns]>
```

```

[ ]: # Supondo que seus dados estão em um DataFrame chamado 'df'
     # Convertendo a coluna de listas para um array numpy
     data = data['Listas'].apply(lambda x: np.array(eval(x))).values
     data = np.array([list(item) for item in data])

```

```

[ ]: import numpy as np
     import pandas as pd
     from keras.models import Sequential
     from keras.layers import LSTM, Dense

     # Número de entradas iniciais a serem usadas
     k = 160

     # Dividindo dados em entradas (X) e saídas (y)
     X = data[:, :k]
     y = data[:, k:]

     # Reformatar X para [samples, timesteps, features]
     X = X.reshape((X.shape[0], X.shape[1], 1))

     # Construindo o modelo LSTM
     model = Sequential()
     model.add(LSTM(640, activation='relu', input_shape=(k, 1)))
     model.add(Dense(y.shape[1], activation='sigmoid')) # Sigmoid para saída binária
     model.compile(optimizer='adam', loss='binary_crossentropy',
                   metrics=['accuracy'])

     # Treinando o modelo
     model.fit(X, y, epochs=50, verbose=1, batch_size=640)

     # Fazendo previsões
     predictions = (model.predict(X) > 0.5).astype(int) # Convertendo
                   ↳probabilidades em 0s e 1s

     # Avaliando a precisão
     accuracy = np.mean(predictions == y)
     print(f"Accuracy: {accuracy:.2f}")

```

```
/home/darkcover/Documentos/Out/venv/lib/python3.10/site-  
packages/keras/src/layers/rnn/rnn.py:204: UserWarning: Do not pass an  
`input_shape`/`input_dim` argument to a layer. When using Sequential models,  
prefer using an `Input(shape)` object as the first layer in the model instead.  
    super().__init__(**kwargs)
```

Epoch 1/50

```
1/1          24s 24s/step -  
accuracy: 0.0000e+00 - loss: 0.6932
```

Epoch 2/50

```
1/1          21s 21s/step -  
accuracy: 0.0000e+00 - loss: 0.6917
```

Epoch 3/50

```
1/1          22s 22s/step -  
accuracy: 0.0000e+00 - loss: 0.6898
```

Epoch 4/50

```
1/1          35s 35s/step -  
accuracy: 0.0000e+00 - loss: 0.6870
```

Epoch 5/50

```
1/1          20s 20s/step -  
accuracy: 0.0000e+00 - loss: 0.6813
```

Epoch 6/50

```
1/1          21s 21s/step -  
accuracy: 0.0000e+00 - loss: 0.6429
```

Epoch 7/50

```
1/1          16s 16s/step -  
accuracy: 0.0000e+00 - loss: 0.6752
```

Epoch 8/50

```
1/1          16s 16s/step -  
accuracy: 0.0000e+00 - loss: 0.6789
```

Epoch 9/50

```
1/1          21s 21s/step -  
accuracy: 0.0000e+00 - loss: 0.6802
```

Epoch 10/50

```
1/1          19s 19s/step -  
accuracy: 0.0000e+00 - loss: 0.6806
```

Epoch 11/50

```
1/1          20s 20s/step -  
accuracy: 0.0000e+00 - loss: 0.6806
```

Epoch 12/50

```
1/1          15s 15s/step -  
accuracy: 0.0000e+00 - loss: 0.6802
```

Epoch 13/50

```
1/1          14s 14s/step -  
accuracy: 0.0000e+00 - loss: 0.6797
```

Epoch 14/50

```
1/1          14s 14s/step -  
accuracy: 0.0000e+00 - loss: 0.6789
```

Epoch 15/50  
1/1 21s 21s/step -  
accuracy: 0.0000e+00 - loss: 0.6778  
Epoch 16/50  
1/1 15s 15s/step -  
accuracy: 0.0000e+00 - loss: 0.6763  
Epoch 17/50  
1/1 15s 15s/step -  
accuracy: 0.0000e+00 - loss: 0.6741  
Epoch 18/50  
1/1 20s 20s/step -  
accuracy: 0.0000e+00 - loss: 0.6710  
Epoch 19/50  
1/1 15s 15s/step -  
accuracy: 0.0000e+00 - loss: 0.6662  
Epoch 20/50  
1/1 15s 15s/step -  
accuracy: 0.0000e+00 - loss: 0.6572  
Epoch 21/50  
1/1 19s 19s/step -  
accuracy: 0.0000e+00 - loss: 0.6305  
Epoch 22/50  
1/1 14s 14s/step -  
accuracy: 0.0000e+00 - loss: 0.6461  
Epoch 23/50  
1/1 21s 21s/step -  
accuracy: 0.0000e+00 - loss: 0.6474  
Epoch 24/50  
1/1 15s 15s/step -  
accuracy: 0.0000e+00 - loss: 0.6447  
Epoch 25/50  
1/1 16s 16s/step -  
accuracy: 0.0000e+00 - loss: 0.6387  
Epoch 26/50  
1/1 20s 20s/step -  
accuracy: 0.0000e+00 - loss: 0.6289  
Epoch 27/50  
1/1 21s 21s/step -  
accuracy: 0.0000e+00 - loss: 1.0566  
Epoch 28/50  
1/1 20s 20s/step -  
accuracy: 0.0000e+00 - loss: 0.6394  
Epoch 29/50  
1/1 20s 20s/step -  
accuracy: 0.0000e+00 - loss: 0.6527  
Epoch 30/50  
1/1 19s 19s/step -  
accuracy: 0.0000e+00 - loss: 0.6600

Epoch 31/50  
1/1 21s 21s/step -  
accuracy: 0.0000e+00 - loss: 0.6645  
Epoch 32/50  
1/1 15s 15s/step -  
accuracy: 0.0000e+00 - loss: 0.6675  
Epoch 33/50  
1/1 17s 17s/step -  
accuracy: 0.0000e+00 - loss: 0.6696  
Epoch 34/50  
1/1 19s 19s/step -  
accuracy: 0.0000e+00 - loss: 0.6711  
Epoch 35/50  
1/1 16s 16s/step -  
accuracy: 0.0000e+00 - loss: 0.6722  
Epoch 36/50  
1/1 20s 20s/step -  
accuracy: 0.0000e+00 - loss: 0.6731  
Epoch 37/50  
1/1 21s 21s/step -  
accuracy: 0.0000e+00 - loss: 0.6737  
Epoch 38/50  
1/1 19s 19s/step -  
accuracy: 0.0000e+00 - loss: 0.6741  
Epoch 39/50  
1/1 22s 22s/step -  
accuracy: 0.0000e+00 - loss: 0.6745  
Epoch 40/50  
1/1 20s 20s/step -  
accuracy: 0.0000e+00 - loss: 0.6747  
Epoch 41/50  
1/1 21s 21s/step -  
accuracy: 0.0000e+00 - loss: 0.6748  
Epoch 42/50  
1/1 21s 21s/step -  
accuracy: 0.0000e+00 - loss: 0.6749  
Epoch 43/50  
1/1 20s 20s/step -  
accuracy: 0.0000e+00 - loss: 0.6749  
Epoch 44/50  
1/1 15s 15s/step -  
accuracy: 0.0000e+00 - loss: 0.6748  
Epoch 45/50  
1/1 16s 16s/step -  
accuracy: 0.0000e+00 - loss: 0.6746  
Epoch 46/50  
1/1 18s 18s/step -  
accuracy: 0.0000e+00 - loss: 0.6744

```
Epoch 47/50
1/1          12s 12s/step -
accuracy: 0.0000e+00 - loss: 0.6741
Epoch 48/50
1/1          24s 24s/step -
accuracy: 0.0000e+00 - loss: 0.6736
Epoch 49/50
1/1          16s 16s/step -
accuracy: 0.0000e+00 - loss: 0.6731
Epoch 50/50
1/1          21s 21s/step -
accuracy: 0.0000e+00 - loss: 0.6724
13/13        10s 734ms/step
Accuracy: 0.68
```

```
[ ]: model.summary
```

```
[ ]: <bound method Model.summary of <Sequential name=sequential, built=True>>
```