

Recurrent neural networks in Keras

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- ▶ Will focus exclusively on LSTMs here

LSTM layers

```
from keras.layers.recurrent import LSTM

LSTM(units,
      activation='tanh',
      recurrent_activation='hard_sigmoid',
      recurrent_initializer='orthogonal',
      recurrent_regularizer=None,
      dropout=0.0, recurrent_dropout=0.0,
      return_sequences=False)
```

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- ▶ Embed a vocabulary into a vector space, apply to sentences
- ▶ 2D input mapped to 3D output, connects to LSTMs.

Embedding layers

```
from keras.layers.embeddings import Embedding

Embedding(input_dim,          # Vocabulary size
           output_dim,        # Output vector length
           embeddings_initializer='uniform',
           mask_zero=False)    # Mask zero values
```

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- ▶ 25.000 movie reviews from IMDB, labelled good or bad
- ▶ Data available from `keras.datasets` module
- ▶ Data is preprocessed as sequences of integers
- ▶ Task: classify sentiment from review content
- ▶ Strategy: embed sentences, then learn structure with LSTM

Loading IMDB sentiment data

```
from keras.preprocessing import sequence
from keras.models import Sequential
from keras.layers import Dense, Embedding
from keras.layers import LSTM
from keras.datasets import imdb

max_features = 20000
maxlen = 80

(x_train, y_train), (x_test, y_test) = \
    imdb.load_data(num_words=max_features)
```

Padding sequences and defining LSTM model

```
x_train = sequence.pad_sequences(x_train, maxlen=maxlen)
x_test = sequence.pad_sequences(x_test, maxlen=maxlen)

model = Sequential()
model.add(Embedding(max_features, 128))
model.add(LSTM(128, dropout=0.2, recurrent_dropout=0.2))
model.add(Dense(1, activation='sigmoid'))
```

Run and evaluate model

```
model.compile(loss='binary_crossentropy',  
              optimizer='sgd',  
              metrics=['accuracy'])  
  
model.fit(x_train, y_train,  
          batch_size=32, epochs=15,  
          validation_data=(x_test, y_test))  
  
model.evaluate(x_test, y_test, batch_size=32)
```