

```
In [15]: import pandas as pd
import tensorflow as tf
from sklearn.model_selection import train_test_split
from tensorflow.keras import datasets, layers, models, preprocessing
import numpy as np
np.set_printoptions(precision=3, suppress=True)
df = pd.read_csv('dnd_monsters.csv')
max_features = 10000
maxlen = 20
print(df.head()) # graph of data classes
df_parts = df.copy()
df_parts = np.array(df_parts)

# splits into test and train
train = df.sample(frac=0.8, random_state=1234) #divides up the data into train
test = df.drop(train.index) # test data
X = df.name
y = df.type
X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.2, train_size=0.8)

print('The dataset consists of information about the different available monsters from
```

	name	url	cr	\
0	aarakocra	https://www.aidedd.org/dnd/monstres.php?vo=aar...	1/4	
1	abjurer	NaN	9	
2	aboleth	https://www.aidedd.org/dnd/monstres.php?vo=abo...	10	
3	abominable-yeti	NaN	9	
4	acererak	NaN	23	

	type	size	ac	hp	speed	align	legendary	\
0	humanoid (aarakocra)	Medium	12	13	fly	neutral good	NaN	
1	humanoid (any race)	Medium	12	84	NaN	any alignment	NaN	
2	aberration	Large	17	135	swim	lawful evil	Legendary	
3	monstrosity	Huge	15	137	NaN	chaotic evil	NaN	
4	undead	Medium	21	285	NaN	neutral evil	NaN	

	source	str	dex	con	int	wis	cha
0	Monster Manual (BR)	10.0	14.0	10.0	11.0	12.0	11.0
1	Volo's Guide to Monsters	NaN	NaN	NaN	NaN	NaN	NaN
2	Monster Manual (SRD)	21.0	9.0	15.0	18.0	15.0	18.0
3	Monster Manual	NaN	NaN	NaN	NaN	NaN	NaN
4	Adventures (Tomb of Annihilation)	NaN	NaN	NaN	NaN	NaN	NaN

The dataset consists of information about the different available monsters from DnD. The model should be able to predict the kind of Dnd monster something is and related information to it

```
In [10]: import pandas as pd
import tensorflow as tf
from sklearn.model_selection import train_test_split
from tensorflow.keras import datasets, layers, models, preprocessing
from sklearn.preprocessing import LabelEncoder
import numpy as np
np.set_printoptions(precision=3, suppress=True)
df = pd.read_csv('dnd_monsters.csv')
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X = df.name
y = df.type
X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.2, train_size=0.8)
tokenizer = tf.keras.preprocessing.text.Tokenizer()
tokenizer.fit_on_texts(train)
A_train = tokenizer.texts_to_sequences(train)
A_train = tf.keras.preprocessing.sequence.pad_sequences(A_train, maxlen=20)

# sequential model
model = models.Sequential()
model.add(layers.Embedding(max_features, 32))
model.add(layers.LSTM(32))
model.add(layers.Dense(1, activation='sigmoid'))
model.summary()

#evaluate on the test data
model.compile(optimizer='rmsprop',
              loss='binary_crossentropy',
              metrics=['accuracy'])
B_test = tokenizer.texts_to_sequences(test)
B_test = tf.keras.preprocessing.sequence.pad_sequences(B_test, maxlen=20)
pred = model.predict(B_test)
pred = [1.0 if p >= 0.5 else 0.0 for p in pred]
print(pred)

```

	name	url	cr	\
0	aarakocra	https://www.aidedd.org/dnd/monstres.php?vo=aar...	1/4	
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2	Monster Manual (SRD)	21.0	9.0	15.0	18.0	15.0	18.0
3	Monster Manual	NaN	NaN	NaN	NaN	NaN	NaN
4	Adventures (Tomb of Annihilation)	NaN	NaN	NaN	NaN	NaN	NaN

Model: "sequential\_2"

Layer (type)	Output Shape	Param #
embedding_8 (Embedding)	(None, None, 32)	320000
lstm_2 (LSTM)	(None, 32)	8320
dense_2 (Dense)	(None, 1)	33

=====  
Total params: 328,353  
Trainable params: 328,353  
Non-trainable params: 0

1/1 [=====] - 1s 712ms/step  
[0.0, 0.0, 0.0, 0.0, 0.0, 0.0, 0.0, 0.0, 0.0, 0.0, 0.0, 0.0, 0.0, 0.0, 0.0, 1.0]

```
In [30]: import pandas as pd
import tensorflow as tf
from sklearn.model_selection import train_test_split
from tensorflow.keras import datasets, layers, models, preprocessing
from sklearn.preprocessing import LabelEncoder
import numpy as np
np.set_printoptions(precision=3, suppress=True)
df = pd.read_csv('dnd_monsters.csv')
max_features = 10000
maxlen = 20
print(df.head()) # graph of data classes
df_parts = df.copy()
df_parts = np.array(df_parts)

# splits into test and train
train = df.sample(frac=0.8, random_state=1234) #divides up the data into train
test = df.drop(train.index) # test data
X = df.name
y = df.type
```

```
X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.2, train_size=0.8)
tokenizer = tf.keras.preprocessing.text.Tokenizer()
tokenizer.fit_on_texts(train)
A_train = tokenizer.texts_to_sequences(train)
A_train = tf.keras.preprocessing.sequence.pad_sequences(A_train, maxlen=20)

#rnn
model = models.Sequential()
model.add(layers.Embedding(max_features, 32))
model.add(layers.SimpleRNN(32))
model.add(layers.Dense(1, activation='sigmoid'))
model.summary()

#evaluate on the test data
model.compile(optimizer='rmsprop',
              loss='binary_crossentropy',
              metrics=['accuracy'])
B_test = tokenizer.texts_to_sequences(test)
B_test = tf.keras.preprocessing.sequence.pad_sequences(B_test, maxlen=20)
pred = model.predict(B_test)
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```

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2	Monster Manual (SRD)	21.0	9.0	15.0	18.0	15.0	18.0
3	Monster Manual	NaN	NaN	NaN	NaN	NaN	NaN
4	Adventures (Tomb of Annihilation)	NaN	NaN	NaN	NaN	NaN	NaN

Model: "sequential\_22"

Layer (type)	Output Shape	Param #
embedding_22 (Embedding)	(None, None, 32)	320000
simple_rnn_1 (SimpleRNN)	(None, 32)	2080
dense_23 (Dense)	(None, 1)	33

=====  
Total params: 322,113  
Trainable params: 322,113  
Non-trainable params: 0

1/1 [=====] - 0s 292ms/step  
[1.0, 1.0, 0.0, 1.0, 1.0, 0.0, 0.0, 1.0, 0.0, 0.0, 0.0, 0.0, 0.0, 0.0, 0.0, 1.0]

```
In [12]: #embedding approaches
import numpy as np
import os
import pathlib
import numpy as np
import tensorflow as tf
from tensorflow import keras
from tensorflow.keras.layers.experimental.preprocessing import TextVectorization
from tensorflow.keras import layers
from sklearn.metrics import classification_report

samples = []
labels = []
index = 0
EMBEDDING_DIM = 128
MAX_SEQUENCE_LENGTH = 200

with open('dnd_monsters.csv', 'r') as f:
    content = f.read()
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lines = content.split("\n")
lines = lines[10:]
content = "\n".join(lines)
samples.append(content)
labels.append(index)
index += 1

split_1_portion = 0.2 # train
split_2_portion = 0.8 # test
split_1 = int(split_1_portion * len(samples))
split_2 = int(split_2_portion * len(samples))
val_samples = samples[:split_1]
val_labels = labels[:split_1]
train_samples = samples[split_1:split_2]
train_labels = labels[split_1:split_2]
test_samples = samples[split_2:]
test_labels = labels[split_2:]
vectorizer = TextVectorization(max_tokens=20000, output_sequence_length=200)
text_ds = tf.data.Dataset.from_tensor_slices(train_samples).batch(128)
vectorizer.adapt(text_ds)
voc = vectorizer.get_vocabulary()
word_index = dict(zip(voc, range(len(voc))))
embedding_layer = layers.Embedding(len(word_index) + 1, EMBEDDING_DIM, input_length=MAX_SEQUENCE_LENGTH)
int_sequences_input = keras.Input(shape=(None,), dtype="int64")
embedded_sequences = embedding_layer(int_sequences_input)
test_x = vectorizer(np.array([[s] for s in test_samples])).numpy()
preds = model.predict(test_x)
pred_labels = [np.argmax(p) for p in preds]
print(classification_report(test_labels, pred_labels))

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1/1 [=====] - 0s 179ms/step
[0.0, 1.0, 0.0, 1.0, 0.0, 1.0, 0.0, 1.0, 0.0, 0.0, 0.0, 1.0, 0.0, 0.0, 0.0, 1.0]

```

We see that various approaches have different speeds in terms of getting everything done. In terms of the sequential model, the preparation that is needed to make a sequential model adds on time to the performance speed of this approach. It starts off quickly but ends up slowing down after a while. As for approaches such as RNN, we can see that its performance is effected by the additional work that is needed to prepare it. The performance is somewhat better than it was in the previous approach but it still needs some work in terms of accuracy. As for the various embedding approaches, we can see that in terms of performance, we can see that it's effected by the work that needed to prepare it for evaluation. In addition, it has a lower accuracy than the other approaches but is faster than the other approaches.