```
In [1]: import numpy as np
    import matplotlib.pyplot as plt
    import tensorflow as tf
    import os
        from tensorflow.keras import Sequential
        from tensorflow.keras.layers import Attention, Dense, Embedding, GRU, L
        STM, SimpleRNN
In [2]: # Ensure GPU is being recognized
    print("Num GPUs available: ", len(tf.config.experimental.list_physical_devices('GPU')))

Num GPUs available: 1
```

# Prepare dataset

#### **Encode Features**

```
In [5]: # Preview first 1000 characters
        print(sherlock data[:500])
                  CHAPTER I
                  Mr. Sherlock Holmes
             In the year 1878 I took my degree of Doctor of Medicine of the
             University of London, and proceeded to Netley to go through the
             course prescribed for surgeons in the army. Having completed my
             studies there, I was duly attached to the Fifth Northumberland
             Fusiliers as Assistant Surgeon. The regiment was stationed in In
        dia
             at the time, and before I could join it, the second Afghan war h
        ad
             broken out. On landing a
In [5]: # Encode characters for training
        index mapping = {char:i for i, char in enumerate(unique chars)}
        sherlock index = np.array([index_mapping[char] for char in sherlock_dat
        a]) # map features to index
        char index = np.array(unique chars)
        print('Character Representation: ', sherlock data[30:42])
        print('Integer Representation: ', sherlock index[30:42])
        Character Representation: Mr. Sherlock
        Integer Representation: [38 72 11 1 44 62 59 72 66 69 57 65]
```

## **Create Training Batches**

```
In [8]: # Break dataset into batches
BATCH_SIZE = 64
BUFFER_SIZE = 10000

dataset = dataset.shuffle(BUFFER_SIZE).batch(BATCH_SIZE, drop_remainder
=True)
dataset
Out[8]: <BatchDataset shapes: ((64, 100), (64, 100)), types: (tf.int32, tf.int32)>
```

### **Train Models**

return model

ly=True)

```
In [12]: # Create the model
        model = build GRU model(vocab size=len(unique chars),
                          embedding dim=embedding dim,
                          rnn units=rnn units,
                          batch size=BATCH SIZE)
        model.summary()
        Model: "sequential"
        Layer (type)
                                  Output Shape
                                                          Param #
        ______
        embedding (Embedding)
                                  (64, None, 256)
                                                          24832
                                  (64, None, 1024)
        gru (GRU)
                                                          3938304
                                  (64, None, 97)
        dense (Dense)
                                                         99425
        ______
        Total params: 4,062,561
        Trainable params: 4,062,561
        Non-trainable params: 0
In [10]: # Use sparse categorical crossentropy as loss
        def loss(labels, logits):
           return tf.keras.losses.sparse_categorical_crossentropy(labels, logi
        ts, from logits=True)
In [14]: # Compile the model
        model.compile(optimizer='adam', loss=loss)
In [15]: | # Save checkpoints to access weights later
        checkpoint dir = './training checkpoints'
        checkpoint prefix = os.path.join(checkpoint dir, 'ckpt {epoch}')
        checkpoint callback = tf.keras.callbacks.ModelCheckpoint(filepath=check
        point prefix,
                                                           save weights on
```

```
In [16]: # Train model
       EPOCHS = 15
       history = model.fit(dataset, epochs=EPOCHS, callbacks=[checkpoint callb
       ack])
       Train for 522 steps
       Epoch 1/15
       522/522 [============ ] - 45s 86ms/step - loss: 2.01
       17
       Epoch 2/15
       522/522 [============ ] - 41s 78ms/step - loss: 1.35
       Epoch 3/15
       522/522 [============ ] - 41s 78ms/step - loss: 1.22
       Epoch 4/15
       522/522 [============ ] - 41s 79ms/step - loss: 1.16
       Epoch 5/15
       522/522 [============ ] - 41s 79ms/step - loss: 1.12
       63
       Epoch 6/15
       522/522 [============= ] - 41s 78ms/step - loss: 1.09
       70
       Epoch 7/15
       522/522 [============ ] - 41s 78ms/step - loss: 1.07
       31
       Epoch 8/15
       522/522 [============= ] - 40s 76ms/step - loss: 1.05
       19
       Epoch 9/15
       522/522 [============ ] - 39s 74ms/step - loss: 1.03
       Epoch 10/15
       Epoch 11/15
       522/522 [============ ] - 40s 76ms/step - loss: 1.00
       15
       Epoch 12/15
       522/522 [============= ] - 40s 77ms/step - loss: 0.98
       72
       Epoch 13/15
       522/522 [============= ] - 42s 80ms/step - loss: 0.97
       52
       Epoch 14/15
       522/522 [============ ] - 42s 81ms/step - loss: 0.96
       Epoch 15/15
       522/522 [============= ] - 42s 80ms/step - loss: 0.95
```

```
In [17]: model = build GRU model (vocab size, embedding dim, rnn units, batch siz
        e=1)
        model.load weights(tf.train.latest checkpoint(checkpoint dir))
        model.build(tf.TensorShape([1, None]))
In [18]: | model.summary()
        Model: "sequential 1"
        Layer (type)
                                  Output Shape
                                                         Param #
        ______
                                 (1, None, 256)
        embedding 1 (Embedding)
                                                         24832
        gru 1 (GRU)
                                  (1, None, 1024)
                                                         3938304
        dense 1 (Dense)
                                 (1, None, 97)
                                                         99425
        ______
        Total params: 4,062,561
        Trainable params: 4,062,561
        Non-trainable params: 0
In [17]: def generate text (model, start string, num generate=1000, temperature=
            input eval = [index mapping[c] for c in start string]
            input eval = tf.expand dims(input eval, 0)
           generated text = []
           model.reset states()
            for i in range(num generate):
               predictions = model(input eval)
               predictions = tf.squeeze(predictions, 0)
               predictions = predictions / temperature
               predicted id = tf.random.categorical(predictions, num samples=
        1) [-1, 0].numpy()
               input eval = tf.expand dims([predicted id], 0)
```

generated text.append(char index[predicted id])

return(start string + ''.join(generated text))

```
In [20]: print(generate text(model, start string=u'There '))
         There was Mr. Reuland I
              should not have something to decrepatid you will fail to tell yo
         u than I.
              "'That with a shrink hat one
              generally is looking at the apologiency quite a bit.
              They appeared.
              Let it threatened by their light her for the tragedy.
              "When a man on the Underground, outwin lire adjustivess
              into the garret was able body and should hive reached this
              creature upon this mental detach."
              "That was the use of the express be put down your lather has
              been mistaken, but as I heard often and I was filled into it. I
         gather that he
              walked so hu round, though I can be rather belay to be an idea.
         The case besides, with a wooden saw the enormous
              reason why did you expect Dr. Professor Coram, will bring a
              return back to have been into the matter so such an unfortunate
         Manor of the tould
              of us out of the latesteeparis again, Dr round the room caught
         the edme baskets. Fou before ever you want to
              introduce mutton, that
In [66]: # Build RNN with LSTM
         def build LSTM model(vocab size, embedding dim, rnn units, batch size):
             model = tf.keras.Sequential([
                 tf.keras.layers.Embedding(vocab size, embedding dim, batch inpu
         t shape=[batch size, None]),
                 tf.keras.layers.LSTM(rnn units,
                                     return sequences=True,
```

```
stateful=True,
                         recurrent initializer='glorot uniform'),
    tf.keras.layers.Dense(vocab size)
1)
return model
```

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```
In [29]: model2 = build LSTM model(vocab size=len(unique chars),
                      embedding dim=embedding dim,
                      rnn_units=rnn_units,
                      batch size=BATCH SIZE)
       model2.summary()
       Model: "sequential 4"
                                           ____
Param #
                            Output Shape
       Layer (type)
       ______
       embedding 4 (Embedding)
                            (64, None, 256)
                                                24832
       lstm_2 (LSTM)
                            (64, None, 1024)
                                                 5246976
       dense 4 (Dense)
                            (64, None, 97)
                                           99425
       _____
       Total params: 5,371,233
       Trainable params: 5,371,233
       Non-trainable params: 0
```

In [30]: model2.compile(optimizer='adam', loss=loss)

```
In [31]: model2.fit(dataset, epochs=EPOCHS)
       Train for 522 steps
       Epoch 1/15
       522/522 [============ ] - 51s 98ms/step - loss: 1.91
       41
       Epoch 2/15
       522/522 [=========== ] - 50s 96ms/step - loss: 1.34
       45
       Epoch 3/15
       522/522 [============ ] - 50s 95ms/step - loss: 1.22
       Epoch 4/15
       522/522 [============ ] - 49s 94ms/step - loss: 1.16
       Epoch 5/15
       522/522 [============ ] - 49s 94ms/step - loss: 1.12
       Epoch 6/15
       522/522 [============= ] - 49s 94ms/step - loss: 1.09
       71 1s -
       Epoch 7/15
       522/522 [============ ] - 49s 95ms/step - loss: 1.07
       23
       Epoch 8/15
       522/522 [============ ] - 49s 93ms/step - loss: 1.05
       06
       Epoch 9/15
       522/522 [============= ] - 50s 95ms/step - loss: 1.03
       13
       Epoch 10/15
       522/522 [============ ] - 51s 97ms/step - loss: 1.01
       Epoch 11/15
       522/522 [============ ] - 53s 101ms/step - loss: 0.9
       Epoch 12/15
       522/522 [============ ] - 50s 97ms/step - loss: 0.97
       76
       Epoch 13/15
       522/522 [============= ] - 50s 97ms/step - loss: 0.96
       12
       Epoch 14/15
       522/522 [============ ] - 51s 98ms/step - loss: 0.94
       49
       Epoch 15/15
       522/522 [============ ] - 49s 94ms/step - loss: 0.92
Out[31]: <tensorflow.python.keras.callbacks.History at 0x202c122f688>
In [32]: model2.save weights('./training checkpoints/LSTM checkpoint')
```

#### In [13]: model2.summary()

Model: "sequential"

Layer (type)	Output Shape	Param #
embedding (Embedding)	(1, None, 256)	24832
lstm (LSTM)	(1, None, 1024)	5246976
dense (Dense)	(1, None, 97)	99425

Total params: 5,371,233
Trainable params: 5,371,233
Non-trainable params: 0

```
In [16]: print(generate text(model2, start string='There'))
         There got away into the house," said Holmes. "I wion my friend jealou
         s, as you tell me
              that he and I have seen anything more like a long one to applaus
         e. It was
              some story of the middle of it.
              "What is it, then?"
              "Well, you dejected that, there would be all that is better than
         ks."
              There was a few months back to the neck in front. Take my ears a
              ended by crumpling of the sharing brow. The afternoon was in
              Mortimer Spanish desires. I understood that we shall show you to
         the Franco-Midland
              Hill."
              "There is."
              "It was I have no means of geores buttons upon a new base, and h
         0
              was regomsed for a money and took a
              stream of fierce-eyed, wicked chindering from as to their
              business. It has been committed in the outside-shever. There's n
         o one their own
              countries from a clothes does. It ran
              argument which brought was married. A lady was
              largely in an ordinary but, as the cloud of light in the doar, w
         hich lead
              in
In [11]: # Build RNN with basic RNN implementation
         def build RNN model (vocab size, embedding dim, rnn units, batch size):
             model = tf.keras.Sequential([
                 tf.keras.layers.Embedding(vocab size, embedding dim, batch inpu
         t shape=[batch size, None]),
                 tf.keras.layers.SimpleRNN(rnn units,
                                     return sequences=True,
                                      stateful=True,
                                     recurrent initializer='glorot uniform'),
                 tf.keras.layers.Dense(vocab size)
             1)
             return model
```

Model: "sequential"

Layer (type)	Output Shape	Param #
embedding (Embedding)	(64, None, 256)	24832
simple_rnn (SimpleRNN)	(64, None, 1024)	1311744
dense (Dense)	(64, None, 97)	99425

Total params: 1,436,001 Trainable params: 1,436,001 Non-trainable params: 0

```
In [13]: | model3.compile(optimizer='adam', loss=loss)
       model3.fit(dataset, epochs=15)
       Train for 522 steps
       Epoch 1/15
       522/522 [============ ] - 39s 76ms/step - loss: 2.36
       Epoch 2/15
       522/522 [============= ] - 38s 73ms/step - loss: 1.58
       Epoch 3/15
       522/522 [============== ] - 38s 73ms/step - loss: 1.40
       Epoch 4/15
       522/522 [============= ] - 38s 73ms/step - loss: 1.31
       86
       Epoch 5/15
       522/522 [============ ] - 38s 72ms/step - loss: 1.26
       87
       Epoch 6/15
       522/522 [============== ] - 38s 72ms/step - loss: 1.23
       Epoch 7/15
       522/522 [=========== ] - 38s 72ms/step - loss: 1.21
       Epoch 8/15
       522/522 [============= ] - 38s 72ms/step - loss: 1.19
       Epoch 9/15
       522/522 [============ ] - 38s 72ms/step - loss: 1.17
       Epoch 10/15
       522/522 [============== ] - 38s 72ms/step - loss: 1.16
       59
       Epoch 11/15
       522/522 [============ ] - 38s 72ms/step - loss: 1.15
       56
       Epoch 12/15
       522/522 [============ ] - 38s 72ms/step - loss: 1.14
       60
       Epoch 13/15
       522/522 [============== ] - 38s 72ms/step - loss: 1.13
       Epoch 14/15
       522/522 [============= ] - 38s 73ms/step - loss: 1.13
       Epoch 15/15
       522/522 [============ ] - 38s 72ms/step - loss: 1.12
       44
Out[13]: <tensorflow.python.keras.callbacks.History at 0x14746279388>
In [14]: model3.save_weights('./training_checkpoints/RNN_checkpoint')
```

```
In [15]: model3 = build RNN model(vocab size=len(unique chars),
                             embedding dim=embedding dim,
                             rnn units=rnn units,
                             batch size=1)
         model3.load weights ('./training checkpoints/RNN checkpoint')
         model3.build(tf.TensorShape([1, None]))
In [18]: print(generate text(model3, start string='There'))
         There LNd, I want of his pocket, I pessired bygeal.
              "A very solution."
              "In the colligants, and maid doubt the faces as it was there and
         something was evidently
              had a most serious rest up, and
              the weight would have had py
              round with a very right had ruch for his mind.
              Holmes stopped at our dark man's hand. A confession of the relac
         itions. I must assure no means, who had first might trie, there was a
         lossly barrates for with the door and shutters below any step had reg
         ret him should, had occasional
              twission to
              he live up to me that
              they are brows most look in with convally last
              things are in his eyes e might imagines we will seek-heavy stagg
         erings and outside, house, as business may
              believe to say that is the paper answered. ' and was
              wrong. They saw that we are dealingshe
              object to acced is too well within the room which seemed to me t
         hat I am a companion of Agent Prower McMurdo, but I shall suddenly st
         aring up as his foolsca
In [52]: # Build RNN with both GRU and LSTM
         def build LSTM GRU model (vocab size, embedding dim, rnn units, batch si
         ze):
             model = tf.keras.Sequential([
                 tf.keras.layers.Embedding(vocab size, embedding dim, batch inpu
         t shape=[batch size, None]),
                 tf.keras.layers.LSTM(rnn units,
                                      return sequences=True,
                                       stateful=True,
                                       recurrent initializer='glorot uniform'),
                 tf.keras.layers.GRU(rnn units,
                                      return sequences=True,
                                      stateful=True,
                                      recurrent initializer='glorot uniform'),
                 tf.keras.layers.Dense(vocab size)
             1)
             return model
```

Model: "sequential\_10"

Layer (type)	Output Shape	Param #
embedding_11 (Embedding)	(64, None, 256)	24832
lstm_6 (LSTM)	(64, None, 1024)	5246976
gru_2 (GRU)	(64, None, 1024)	6297600
dense_10 (Dense)	(64, None, 97)	99425

Total params: 11,668,833 Trainable params: 11,668,833

Non-trainable params: 0

```
In [61]: model4.compile(optimizer='adam', loss=loss)
       model4.fit(dataset, epochs=EPOCHS)
       Train for 522 steps
       Epoch 1/15
       522/522 [============ ] - 158s 302ms/step - loss: 2.
       Epoch 2/15
       522/522 [============= ] - 157s 301ms/step - loss: 1.
       Epoch 3/15
       522/522 [============= ] - 153s 294ms/step - loss: 1.
       Epoch 4/15
       522/522 [============== ] - 153s 292ms/step - loss: 1.
       1404
       Epoch 5/15
       522/522 [============ ] - 155s 296ms/step - loss: 1.
       1017
       Epoch 6/15
       522/522 [============== ] - 158s 303ms/step - loss: 1.
       0710
       Epoch 7/15
       522/522 [============ ] - 164s 314ms/step - loss: 1.
       Epoch 8/15
       522/522 [============= ] - 163s 312ms/step - loss: 1.
       Epoch 9/15
       522/522 [=========== ] - 153s 294ms/step - loss: 0.
       9877
       Epoch 10/15
       522/522 [============== ] - 162s 309ms/step - loss: 0.
       9599
       Epoch 11/15
       522/522 [=========== ] - 161s 309ms/step - loss: 0.
       9334
       Epoch 12/15
       522/522 [=========== ] - 155s 297ms/step - loss: 0.
       9064
       Epoch 13/15
       522/522 [============== ] - 155s 297ms/step - loss: 0.
       Epoch 14/15
       522/522 [============= ] - 156s 299ms/step - loss: 0.
       8544
       Epoch 15/15
       522/522 [============== ] - 155s 296ms/step - loss: 0.
       8304
Out[61]: <tensorflow.python.keras.callbacks.History at 0x202c78cecc8>
In [62]: | model4.save weights('./training checkpoints/LSTM GRU checkpoint')
```

```
In [53]: model4 = build LSTM GRU model(vocab size=len(unique chars),
                             embedding dim=embedding dim,
                             rnn units=rnn units,
                             batch size=1)
         model4.load weights('./training checkpoints/LSTM GRU checkpoint')
         model4.build(tf.TensorShape([1, None]))
In [54]: print(generate text(model4, start string='There'))
         There was no one upon him?"
              Holmes nodded applied "There, you name!" said Holmes, "I shall s
         ee him even sooner
              there shared in the street. It loved means
              that it was to far me, and his eyes fell go to keep her it upon
         the
              ground followed by the mentte who had met Mr. Barclay lumps no
              telegraph with it."
              "That you," said Holmes, smiling, answer in the exerces of the f
         irm. Her
              sleep would stand through my head, pael face, and against them,
         es
              upon it, in spite of its
              instrument.
              "Your own villainy," he white figure. "Can't you, if you wonder
         to his ruin, and
              running through the squasters?"
              "Well, it is likely a little throughory at ten o'clock't.
              Their keys and looked sternly at her excited, heard, high injury
         ty was a quick step now in it was
              covered with the Navy, invite that they were told of my father
              Barrymore of Napoleon, and his persuady has gone this idea. Noth
         ing more showing than
              singular knill I
In [46]: def build bidirectional model (vocab size, embedding dim, rnn units, bat
         ch size):
             model = tf.keras.Sequential([
                 tf.keras.layers.Embedding(vocab size, embedding dim, batch inpu
         t shape=[batch size, None]),
                 tf.keras.layers.Bidirectional(tf.keras.layers.LSTM(rnn units //
         2, return sequences=True)),
                 tf.keras.layers.Dense(vocab size)
             return model
```

Model: "sequential\_12"

Layer (type)	Output Shape	Param #
embedding_11 (Embedding)	(64, None, 256)	24832
bidirectional_16 (Bidirectio	(64, None, 1024)	3149824
dense_12 (Dense)	(64, None, 97)	99425

Total params: 3,274,081 Trainable params: 3,274,081 Non-trainable params: 0

\_\_\_\_\_

```
In [48]: model5.compile(optimizer='adam', loss=loss)
model5.fit(dataset, epochs=15)
```

```
Train for 522 steps
Epoch 1/15
WARNING: tensorflow: Unresolved object in checkpoint: (root).optimizer
WARNING: tensorflow: Unresolved object in checkpoint: (root).optimizer.
learning rate
WARNING: tensorflow: Unresolved object in checkpoint: (root).optimizer'
s state 'm' for (root).layer with weights-0.embeddings
WARNING: tensorflow: Unresolved object in checkpoint: (root).optimizer'
s state 'm' for (root).layer_with_weights-2.kernel
WARNING: tensorflow: Unresolved object in checkpoint: (root).optimizer'
s state 'm' for (root).layer with weights-2.bias
WARNING: tensorflow: Unresolved object in checkpoint: (root).optimizer'
s state 'm' for (root).layer with weights-1.forward layer.cell.kernel
WARNING: tensorflow: Unresolved object in checkpoint: (root).optimizer'
s state 'm' for (root).layer with weights-1.forward layer.cell.recurr
ent kernel
WARNING: tensorflow: Unresolved object in checkpoint: (root).optimizer'
s state 'm' for (root).layer with weights-1.forward layer.cell.bias
WARNING: tensorflow: Unresolved object in checkpoint: (root).optimizer'
s state 'm' for (root).layer with weights-1.backward layer.cell.kerne
WARNING: tensorflow: Unresolved object in checkpoint: (root).optimizer'
s state 'm' for (root).layer with weights-1.backward layer.cell.recur
rent kernel
WARNING: tensorflow: Unresolved object in checkpoint: (root).optimizer'
s state 'm' for (root).layer with weights-1.backward layer.cell.bias
WARNING: tensorflow: Unresolved object in checkpoint: (root).optimizer'
s state 'v' for (root).layer with weights-0.embeddings
WARNING: tensorflow: Unresolved object in checkpoint: (root).optimizer'
s state 'v' for (root).layer with weights-2.kernel
WARNING: tensorflow: Unresolved object in checkpoint: (root).optimizer'
s state 'v' for (root).layer with weights-2.bias
WARNING: tensorflow: Unresolved object in checkpoint: (root).optimizer'
s state 'v' for (root).layer with weights-1.forward layer.cell.kernel
WARNING: tensorflow: Unresolved object in checkpoint: (root).optimizer'
s state 'v' for (root).layer with weights-1.forward layer.cell.recurr
ent kernel
WARNING: tensorflow: Unresolved object in checkpoint: (root).optimizer'
s state 'v' for (root).layer with weights-1.forward layer.cell.bias
WARNING: tensorflow: Unresolved object in checkpoint: (root).optimizer'
s state 'v' for (root).layer_with_weights-1.backward_layer.cell.kerne
WARNING: tensorflow: Unresolved object in checkpoint: (root).optimizer'
s state 'v' for (root).layer_with_weights-1.backward layer.cell.recur
rent kernel
WARNING: tensorflow: Unresolved object in checkpoint: (root).optimizer'
s state 'v' for (root).layer with weights-1.backward layer.cell.bias
WARNING: tensorflow: A checkpoint was restored (e.g. tf.train. Checkpoin
```

```
t.restore or tf.keras.Model.load weights) but not all checkpointed va
        lues were used. See above for specific issues. Use expect partial() o
        n the load status object, e.g. tf.train.Checkpoint.restore(...).expec
        t partial(), to silence these warnings, or use assert consumed() to m
        ake the check explicit. See https://www.tensorflow.org/guide/checkpoi
        nt#loading mechanics for details.
        522/522 [============ ] - 40s 77ms/step - loss: 0.48
        97
        Epoch 2/15
        522/522 [============ ] - 37s 71ms/step - loss: 0.02
        Epoch 3/15
        522/522 [============= ] - 37s 71ms/step - loss: 0.02
        Epoch 4/15
        522/522 [=========== ] - 37s 71ms/step - loss: 0.01
        Epoch 5/15
        522/522 [============ ] - 37s 71ms/step - loss: 0.01
        Epoch 6/15
        522/522 [============= ] - 38s 72ms/step - loss: 0.01
        Epoch 7/15
        522/522 [============= ] - 38s 73ms/step - loss: 0.01
        70
        Epoch 8/15
        522/522 [============ ] - 38s 72ms/step - loss: 0.01
        Epoch 9/15
        522/522 [============ ] - 38s 72ms/step - loss: 0.01
        Epoch 10/15
        522/522 [============= ] - 38s 72ms/step - loss: 0.01
        Epoch 11/15
        522/522 [============= ] - 37s 72ms/step - loss: 0.01
        35
        Epoch 12/15
        522/522 [============ ] - 37s 71ms/step - loss: 0.01
        24
        Epoch 13/15
        522/522 [============= ] - 37s 71ms/step - loss: 0.01
        11
        Epoch 14/15
        522/522 [=========== ] - 38s 73ms/step - loss: 0.00
        Epoch 15/15
        522/522 [=========== ] - 37s 71ms/step - loss: 0.00
Out[48]: <tensorflow.python.keras.callbacks.History at 0x11dd7b3d248>
In [49]: model5.save weights('./training checkpoints/bidirectional checkpoint')
```

```
In [50]: model5 = build bidirectional model(vocab size=len(unique chars),
                             embedding dim=embedding dim,
                              rnn units=rnn units,
                             batch size=1)
         model5.load weights('./training checkpoints/bidirectional checkpoint')
         model5.build(tf.TensorShape([1, None]))
In [51]: print(generate text(model5, start string='There'))
         There hexalopexanythadves?"
          tot vene livesumaringroxinemy'HAmond
          rFPreneneme, menfedonenebes
          H fatonodenetha
          mivenemermey
          nexar--I Brivesananes cèinèSclvif Miny -d
          stesad benereöGeereve mensAnenCin tingy petoréAsus°à'üI fa winotesed
         Pon'din bfined mesinYonen.
          anexadrenoxpare, I oborexarineme! Wmes?"I ugen Ohhexaloro-Clinsunea
         mexarinelachestealenx12edI Yoryon, us.
          \verb|finncimalifid--cr`'| Trestheved rvenend zzzzzzzzzzzzzenes hancorend.
          agy, junear ZVelinemes'm, Cagrex Agy t y udwabobes; sveved, Chiswheme
         nesmady cudd.
          ddy taloponellemed finerenobal½?½Dy I sishen'sos, Trabunefey hes'Ja
         minere.
          *7lin ItenoneGonconanequncrineperinenennd mesmuneshe'Slaîtowad
          umery Heroushachelo, necloryotoggreswime ey he! gringes manenedas,
         fusony
          bin'Siny.
          Cine ocralinunshes' estedodrinedenve
          gre-glalrengeney gineny masthwheaney y hesgrmerinw-finasathenoxc."Gi
         nd.
          tar` menores semonanexaloruala cholexzenenenelalos!alinopunths; puss
         tonedy Cinetenexalonenellle CDuned Mreblory, I unesPenedy umer
```

```
In [57]: model6 = build GRU model(vocab size=len(unique chars),
                         embedding dim=embedding dim,
                         rnn units=rnn units,
                         batch size=BATCH SIZE)
       model6.compile(optimizer='RMSprop', loss=loss)
       model6.fit(dataset, epochs=15)
       Train for 522 steps
       Epoch 1/15
       522/522 [============ ] - 39s 75ms/step - loss: 1.83
       4.3
       Epoch 2/15
       522/522 [============ ] - 40s 77ms/step - loss: 1.29
       01
       Epoch 3/15
       522/522 [============ ] - 41s 78ms/step - loss: 1.18
       Epoch 4/15
       522/522 [============= ] - 39s 74ms/step - loss: 1.13
       Epoch 5/15
       522/522 [============= ] - 38s 73ms/step - loss: 1.09
       Epoch 6/15
       522/522 [============ ] - 38s 74ms/step - loss: 1.07
       0.4
       Epoch 7/15
       522/522 [============ ] - 39s 74ms/step - loss: 1.04
       61
       Epoch 8/15
       522/522 [============= ] - 39s 74ms/step - loss: 1.02
       48
       Epoch 9/15
       522/522 [=========== ] - 38s 73ms/step - loss: 1.00
       53
       Epoch 10/15
       522/522 [============= ] - 39s 74ms/step - loss: 0.98
       Epoch 11/15
       522/522 [=========== ] - 39s 74ms/step - loss: 0.97
       Epoch 12/15
       522/522 [============= ] - 39s 74ms/step - loss: 0.95
       76
       Epoch 13/15
       522/522 [=========== ] - 41s 79ms/step - loss: 0.94
       50
       Epoch 14/15
       522/522 [============= ] - 40s 76ms/step - loss: 0.93
       32
       Epoch 15/15
       522/522 [============ ] - 41s 78ms/step - loss: 0.92
Out[57]: <tensorflow.python.keras.callbacks.History at 0x11ddf2046c8>
```

```
In [58]: model6.save weights ('./training checkpoints/GRU RSM checkpoint')
In [60]: | model6 = build GRU model(vocab size=len(unique chars),
                             embedding dim=embedding dim,
                             rnn units=rnn units,
                             batch size=1)
         model6.load weights('./training checkpoints/GRU RSM checkpoint')
         model6.build(tf.TensorShape([1, None]))
In [64]: print(generate text(model6, start string='There'))
         There a cloud heavily
              canable. Even now--n the key as he pays."
              "Oh! I realized that my wife was about to realize of my advice,
         Watson?"
              I loved all the views, staring otherwise beyond their solurn con
         versation. They brought no
              common man had blown to the stair.
              "That will do," said I, "for we can then take them
              at least you are, and that there are time it is
              evident that his visit has been entirely drugged to this group.
              And then anno name I have the meaning of the facts of mankind ch
         ief from the West End. Lawurnil
              maid shook his head with his flight."
              "And how did he hear a letter you think before. Of course this b
         eautiful
              perhaps she has spoken to the spot mixty hobbling but not blowni
         ng on to the
              Barrymores.
              He approached it, and a husband, coming down to conceal it, but
         gathered through the
              whole of that last man this morning.
              "I may add that I had been the moat will find a new mornings at
```

Baker Street

Train for 522 steps

```
Epoch 1/15
       93 3s - loss:
       Epoch 2/15
       Epoch 3/15
       522/522 [============== ] - 49s 94ms/step - loss: 1.21
       Epoch 4/15
       522/522 [============= ] - 49s 94ms/step - loss: 1.14
       Epoch 5/15
       522/522 [============ ] - 49s 94ms/step - loss: 1.10
       88 Os - loss: 1.1
       Epoch 6/15
       522/522 [============ ] - 49s 95ms/step - loss: 1.07
       82
       Epoch 7/15
       522/522 [============ ] - 50s 95ms/step - loss: 1.05
       Epoch 8/15
       522/522 [============ ] - 49s 95ms/step - loss: 1.02
       Epoch 9/15
       522/522 [============ ] - 49s 94ms/step - loss: 1.00
       Epoch 10/15
       522/522 [============= ] - 50s 95ms/step - loss: 0.98
       68 1s - lo - ETA: Os - loss:
       Epoch 11/15
       522/522 [============ ] - 49s 94ms/step - loss: 0.96
       78
       Epoch 12/15
       522/522 [============== ] - 49s 95ms/step - loss: 0.94
       91
       Epoch 13/15
       522/522 [============== ] - 49s 94ms/step - loss: 0.93
       16
       Epoch 14/15
       522/522 [============ ] - 49s 94ms/step - loss: 0.91
       Epoch 15/15
       522/522 [============= ] - 49s 95ms/step - loss: 0.89
Out[67]: <tensorflow.python.keras.callbacks.History at 0x11de35343c8>
In [68]: model7.save weights('./training checkpoints/LSTM RMS checkpoint')
```

```
In [69]: model7 = build LSTM model(vocab size=len(unique chars),
                             embedding dim=embedding dim,
                             rnn units=rnn units,
                             batch size=1)
         model7.load weights('./training checkpoints/LSTM RMS checkpoint')
         model7.build(tf.TensorShape([1, None]))
In [70]: print(generate text(model7, start string='There'))
         There came down to
              Rolding's six; but I was a sinking out a slip of paper and the
                    Rylodes will be back?"
              "By thumb," he said. "But the lady is sure that you would. You
              slipped at a glovely interesting path, and y-- Young October sin
         ce Dr. Mortimer told
              me hell upon so much as material, red before the treasure
              is gieaties. As he said, however, otherwise, on which was they
              glancing backwaze nor any elecwing girl. It is a stout fanging i
         n front of
              yellow-shot, forbidden it is close to me with the auraction of a
              large idea that he did."
              "You may make it worth this security for?" asked Holmes.
              "Father--link above. She was told in one of his
              met. Even within each business must blame it itsed window of the
         heavy brows,
              and begins to clear it up. With our friend
              could lie down without her, but indoors I can."
              "What else it is ubout war, for the ladisch door over the walls
              t it. And I
              think. We must in that
```

Train for 522 steps

```
Epoch 1/15
       522/522 [============ ] - 39s 74ms/step - loss: 2.59
       35
       Epoch 2/15
       Epoch 3/15
       522/522 [============== ] - 37s 72ms/step - loss: 1.56
       Epoch 4/15
       522/522 [============= ] - 37s 71ms/step - loss: 1.48
       Epoch 5/15
       522/522 [============ ] - 38s 72ms/step - loss: 1.43
       75
       Epoch 6/15
       522/522 [============ ] - 37s 72ms/step - loss: 1.40
       37
       Epoch 7/15
       522/522 [============ ] - 37s 72ms/step - loss: 1.37
       88
       Epoch 8/15
       522/522 [============ ] - 37s 72ms/step - loss: 1.35
       Epoch 9/15
       522/522 [=========== ] - 38s 72ms/step - loss: 1.34
       Epoch 10/15
       522/522 [============ ] - 37s 71ms/step - loss: 1.33
       06
       Epoch 11/15
       90
       Epoch 12/15
       522/522 [============== ] - 38s 72ms/step - loss: 1.31
       01
       Epoch 13/15
       522/522 [============= ] - 38s 72ms/step - loss: 1.30
       13
       Epoch 14/15
       522/522 [============ ] - 38s 72ms/step - loss: 1.29
       Epoch 15/15
       522/522 [============ ] - 37s 71ms/step - loss: 1.28
Out[74]: <tensorflow.python.keras.callbacks.History at 0x11dc748ee88>
In [75]: model8.save weights('./training checkpoints/RNN RMS checkpoint')
```

```
In [76]: model8 = build RNN model(vocab size=len(unique chars),
                             embedding dim=embedding dim,
                             rnn units=rnn units,
                             batch size=1)
         model8.load weights('./training checkpoints/RNN RMS checkpoint')
         model8.build(tf.TensorShape([1, None]))
In [78]: print(generate text(model8, start string='There'))
         There all
              to but to be disguiseous aspackmissist
              know. The
              window lodge play she causence that one passive."
              "You will would be dark I can Alvarourous oftei. But they took a
         ll you, Sir Claimanion. That's
              look between from his
              atmospage of
              the end of them against his hury told your confided job-gover ab
              long prescinger which can she was enter off and dumbled
              of Afficeales to have you, Souses, and also question had never h
         eard I've back.
              "I know that he had take tracked upon the
              last quiting. On the
              vague deal him the orders and soul-wream of this!" Sand-house in
         spector to started.
              "That is you, Formall
              down gales sinctions should have no derately up the
              discover the will launch
              in the man
              admital charactering that
              det in an accues marious satually come. I think that he sporting
         his friendly known house. His own
              dars and my alsoftened at, well, and
              you will trusting as not with the ins
In [ ]:
```

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