

```
import numpy as np
import tensorflow as tf
from tensorflow.keras.models import Sequential
from tensorflow.keras.layers import LSTM, Dense, Embedding
from tensorflow.keras.preprocessing.text import Tokenizer
from tensorflow.keras.preprocessing.sequence import pad_sequences
import random
```

```
text = """
artificial intelligence is transforming modern society.
it is used in healthcare finance education and transportation.
machine learning allows systems to improve automatically with experience.
data plays a critical role in training intelligent systems.
large datasets help models learn complex patterns.
deep learning uses multi layer neural networks.
neural networks are inspired by biological neurons.
each neuron processes input and produces an output.
training a neural network requires optimization techniques.
gradient descent minimizes the loss function.
```

```
natural language processing helps computers understand human language.
text generation is a key task in nlp.
language models predict the next word or character.
recurrent neural networks handle sequential data.
lstm and gru models address long term dependency problems.
however rnn based models are slow for long sequences.
```

```
transformer models changed the field of nlp.
they rely on self attention mechanisms.
attention allows the model to focus on relevant context.
transformers process data in parallel.
this makes training faster and more efficient.
modern language models are based on transformers.
```

```
education is being improved using artificial intelligence.
intelligent tutoring systems personalize learning.
automated grading saves time for teachers.
online education platforms use recommendation systems.
technology enhances the quality of learning experiences.
```

```
ethical considerations are important in artificial intelligence.
fairness transparency and accountability must be ensured.
ai systems should be designed responsibly.
data privacy and security are major concerns.
researchers continue to improve ai safety.
```

```
text generation models can create stories poems and articles.
they are used in chatbots virtual assistants and content creation.
generated text should be meaningful and coherent.
evaluation of text generation is challenging.
human judgement is often required.
```

```
continuous learning is essential in the field of ai.
research and innovation drive technological progress.
students should build strong foundations in mathematics.
programming skills are important for ai engineers.
practical experimentation enhances understanding.
"""
```

```
tokenizer = Tokenizer()
tokenizer.fit_on_texts([text])

total_words = len(tokenizer.word_index) + 1

input_sequences = []

for line in text.split("\n"):
    token_list = tokenizer.texts_to_sequences([line])[0]
    for i in range(1, len(token_list)):
        n_gram = token_list[:i+1]
        input_sequences.append(n_gram)

max_seq_len = max(len(seq) for seq in input_sequences)
```

```
input_sequences = pad_sequences(input_sequences, maxlen=max_seq_len, padding='pre')

X = input_sequences[:, :-1]
y = input_sequences[:, -1]

y = tf.keras.utils.to_categorical(y, num_classes=total_words)
```

```
from tensorflow.keras.layers import LayerNormalization, MultiHeadAttention, Dropout
```

```
model = Sequential([
    Embedding(total_words, 100, input_length=max_seq_len-1),
    LSTM(150),
    Dense(total_words, activation='softmax')
])

model.compile(loss='categorical_crossentropy', optimizer='adam', metrics=['accuracy'])
model.summary()
```

/usr/local/lib/python3.12/dist-packages/keras/src/layers/core/embedding.py:97: UserWarning: Argument `input\_length` is deprecated  
warnings.warn(

Model: "sequential\_1"

Layer (type)	Output Shape	Param #
embedding_1 ( <a href="#">Embedding</a> )	?	0 (unbuilt)
lstm ( <a href="#">LSTM</a> )	?	0 (unbuilt)
dense_3 ( <a href="#">Dense</a> )	?	0 (unbuilt)

Total params: 0 (0.00 B)

Trainable params: 0 (0.00 B)

```
model.fit(X, y, epochs=200, verbose=1)
```

```

8/8 ----- 0s 8ms/step - accuracy: 0.9811 - loss: 0.0515
Epoch 195/200
8/8 ----- 0s 7ms/step - accuracy: 0.9948 - loss: 0.0295
Epoch 196/200
8/8 ----- 0s 7ms/step - accuracy: 0.9938 - loss: 0.0300
Epoch 197/200
8/8 ----- 0s 8ms/step - accuracy: 0.9841 - loss: 0.0444
Epoch 198/200
8/8 ----- 0s 8ms/step - accuracy: 0.9841 - loss: 0.0447
Epoch 199/200
8/8 ----- 0s 8ms/step - accuracy: 0.9875 - loss: 0.0294
Epoch 200/200
8/8 ----- 0s 8ms/step - accuracy: 0.9808 - loss: 0.0348
<keras.src.callbacks.history.History at 0x7cfb30775070>

```

```

def generate_text(seed_text, next_words=20):
    for _ in range(next_words):
        token_list = tokenizer.texts_to_sequences([seed_text])[0]
        token_list = pad_sequences([token_list], maxlen=max_seq_len-1, padding='pre')
        predicted = np.argmax(model.predict(token_list), axis=-1)[0]

        output_word = ""
        for word, index in tokenizer.word_index.items():
            if index == predicted:
                output_word = word
                break

        seed_text += " " + output_word

    return seed_text

print(generate_text("artificial intelligence"))

```

```

1/1 ----- 0s 418ms/step
1/1 ----- 0s 81ms/step
1/1 ----- 0s 70ms/step
1/1 ----- 0s 44ms/step
1/1 ----- 0s 31ms/step
1/1 ----- 0s 28ms/step
1/1 ----- 0s 29ms/step
1/1 ----- 0s 27ms/step
1/1 ----- 0s 28ms/step
1/1 ----- 0s 27ms/step
1/1 ----- 0s 28ms/step
1/1 ----- 0s 29ms/step
1/1 ----- 0s 30ms/step
1/1 ----- 0s 28ms/step
1/1 ----- 0s 28ms/step
1/1 ----- 0s 28ms/step
1/1 ----- 0s 28ms/step
1/1 ----- 0s 29ms/step
1/1 ----- 0s 29ms/step
1/1 ----- 0s 27ms/step
artificial intelligence is transforming modern society society society engineers engineers problems problems problems problems p

```

```

def positional_encoding(length, depth):
    depth = depth/2
    positions = np.arange(length)[: , np.newaxis]
    depths = np.arange(depth)[np.newaxis, :] / depth
    angle_rates = 1 / (10000**(depths))
    angle_rads = positions * angle_rates
    pos_encoding = np.concatenate(
        [np.sin(angle_rads), np.cos(angle_rads)],
        axis=-1)
    return tf.cast(pos_encoding, dtype=tf.float32)

```

```

class TransformerBlock(tf.keras.layers.Layer):
    def __init__(self, embed_dim, num_heads, ff_dim):
        super().__init__()
        self.att = MultiHeadAttention(num_heads=num_heads, key_dim=embed_dim)
        self.ffn = Sequential([
            Dense(ff_dim, activation="relu"),
            Dense(embed_dim),
        ])
        self.layernorm1 = LayerNormalization()
        self.layernorm2 = LayerNormalization()
        self.dropout1 = Dropout(0.1)
        self.dropout2 = Dropout(0.1)

```

```
def call(self, inputs, training=False):
    attn_output = self.att(inputs, inputs)
    attn_output = self.dropout1(attn_output, training=training)
    out1 = self.layernorm1(inputs + attn_output)
    ffn_output = self.ffn(out1)
    ffn_output = self.dropout2(ffn_output, training=training)
    return self.layernorm2(out1 + ffn_output)
```

```
embed_dim = 64
num_heads = 2
ff_dim = 128

inputs = tf.keras.Input(shape=(max_seq_len-1,))
embedding_layer = Embedding(total_words, embed_dim)
x = embedding_layer(inputs)

x = x + positional_encoding(max_seq_len-1, embed_dim)

transformer_block = TransformerBlock(embed_dim, num_heads, ff_dim)
x = transformer_block(x)

x = tf.keras.layers.GlobalAveragePooling1D()(x)
outputs = Dense(total_words, activation="softmax")(x)

transformer_model = tf.keras.Model(inputs=inputs, outputs=outputs)
transformer_model.compile(loss='categorical_crossentropy', optimizer='adam', metrics=['accuracy'])
transformer_model.summary()
```

Model: "functional\_1"

Layer (type)	Output Shape	Param #
input_layer (InputLayer)	(None, 9)	0
embedding (Embedding)	(None, 9, 64)	12,480
add (Add)	(None, 9, 64)	0
transformer_block (TransformerBlock)	(None, 9, 64)	50,048
global_average_pooling1d (GlobalAveragePooling1D)	(None, 64)	0
dense_2 (Dense)	(None, 195)	12,675

Total params: 75,203 (293.76 KB)  
Trainable params: 75,203 (293.76 KB)  
Non-trainable params: 0 (0.00 KB)

```
transformer_model.fit(X, y, epochs=200, verbose=1)
```

```

Epoch 162/200
8/8 ————— 0s 5ms/step - accuracy: 0.9929 - loss: 0.0464
Epoch 163/200
8/8 ————— 0s 5ms/step - accuracy: 0.9632 - loss: 0.0584
Epoch 164/200
8/8 ————— 0s 6ms/step - accuracy: 0.9938 - loss: 0.0509
Epoch 165/200
8/8 ————— 0s 6ms/step - accuracy: 0.9894 - loss: 0.0408
Epoch 166/200
8/8 ————— 0s 5ms/step - accuracy: 0.9876 - loss: 0.0484
Epoch 167/200
8/8 ————— 0s 5ms/step - accuracy: 0.9918 - loss: 0.0414
Epoch 168/200
8/8 ————— 0s 6ms/step - accuracy: 0.9798 - loss: 0.0588
Epoch 169/200
8/8 ————— 0s 5ms/step - accuracy: 0.9648 - loss: 0.0596
Epoch 170/200
8/8 ————— 0s 5ms/step - accuracy: 0.9844 - loss: 0.0475
Epoch 171/200
8/8 ————— 0s 5ms/step - accuracy: 0.9897 - loss: 0.0421
Epoch 172/200
8/8 ————— 0s 5ms/step - accuracy: 0.9651 - loss: 0.0685
Epoch 173/200
8/8 ————— 0s 5ms/step - accuracy: 0.9827 - loss: 0.0440
Epoch 174/200
8/8 ————— 0s 5ms/step - accuracy: 0.9839 - loss: 0.0482
Epoch 175/200
8/8 ————— 0s 5ms/step - accuracy: 0.9733 - loss: 0.0488
Epoch 176/200
8/8 ————— 0s 5ms/step - accuracy: 0.9740 - loss: 0.0514

```

```

def generate_text_transformer(seed_text, next_words=20):
    for _ in range(next_words):
        token_list = tokenizer.texts_to_sequences([seed_text])[0]
        token_list = pad_sequences([token_list], maxlen=max_seq_len-1, padding='pre')
        predicted = np.argmax(transformer_model.predict(token_list), axis=-1)[0]

        for word, index in tokenizer.word_index.items():
            if index == predicted:
                seed_text += " " + word
                break

    return seed_text

print(generate_text_transformer("machine learning"))

```

```

1/1 ————— 2s 2s/step
1/1 ————— 0s 30ms/step
1/1 ————— 0s 30ms/step
1/1 ————— 0s 28ms/step
1/1 ————— 0s 29ms/step
1/1 ————— 0s 32ms/step
1/1 ————— 0s 29ms/step
1/1 ————— 0s 28ms/step
1/1 ————— 0s 28ms/step
1/1 ————— 0s 29ms/step
1/1 ————— 0s 29ms/step
1/1 ————— 0s 28ms/step
1/1 ————— 0s 28ms/step
1/1 ————— 0s 28ms/step
1/1 ————— 0s 28ms/step
1/1 ————— 0s 29ms/step
1/1 ————— 0s 29ms/step
1/1 ————— 0s 28ms/step
1/1 ————— 0s 27ms/step
1/1 ————— 0s 29ms/step
1/1 ————— 0s 28ms/step
1/1 ————— 0s 33ms/step
machine learning allows systems to improve automatically with experience experience experience experience systems sho

```

