Implement a Transformer-based Language Model for text generation

```
# Install required libraries
!pip install tensorflow numpy
```

1. Import Libraries and Sample Data

```
import tensorflow as tf
import numpy as np

# Sample tiny corpus
sentences = [
    "hello world",
    "how are you",
    "hello how are you",
    "hello you",
    "are you there"
]
```

2. Tokenization & Padding

```
from tensorflow.keras.preprocessing.text import Tokenizer
from tensorflow.keras.preprocessing.sequence import pad_sequences

tokenizer = Tokenizer()
tokenizer.fit_on_texts(sentences)

sequences = tokenizer.texts_to_sequences(sentences)
max_len = max(len(seq) for seq in sequences)
padded_sequences = pad_sequences(sequences, maxlen=max_len, padding='post')

vocab_size = len(tokenizer.word_index) + 1
```

3. Prepare Input and Target

```
X = padded_sequences[:, :-1]  # All except last token
y = padded_sequences[:, 1:]   # All except first token

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

4. Build a Mini Transformer Model

```
from tensorflow.keras.layers import Input, Embedding, Dense, LayerNormalization, Dropout, MultiHeadAttention
from tensorflow.keras.models import Model

embed_dim = 64
num_heads = 2
ff_dim = 128

input_seq = Input(shape=(X.shape[1],))
embedding_layer = Embedding(input_dim=vocab_size, output_dim=embed_dim)(input_seq)

# Multi-Head Attention
attn_output = MultiHeadAttention(num_heads=num_heads, key_dim=embed_dim)(embedding_layer, embedding_layer)
attn_output = Dropout(0.1)(attn_output)
out1 = LayerNormalization(epsilon=1e-6)(embedding_layer + attn_output)

# Feed-Forward
ffn = Dense(ff_dim, activation='relu')(out1)
ffn = Dense(embed_dim)(ffn)
```

```
ffn_output = Dropout(0.1)(ffn)
out2 = LayerNormalization(epsilon=1e-6)(out1 + ffn_output)

# Output layer
final_output = Dense(vocab_size, activation='softmax')(out2)

model = Model(inputs=input_seq, outputs=final_output)
model.compile(optimizer='adam', loss='categorical_crossentropy', metrics=['accuracy'])

model.summary()
```

Model: "functional"

Layer (type)	Output Shape	Param #	Connected to
input_layer (InputLayer)	(None, 3)	0	-
embedding (Embedding)	(None, 3, 64)	448	input_layer[0][0]
multi_head_attenti… (MultiHeadAttentio…	(None, 3, 64)	33,216	embedding[0][0], embedding[0][0]
dropout_1 (Dropout)	(None, 3, 64)	0	multi_head_atten…
add (Add)	(None, 3, 64)	0	embedding[0][0], dropout_1[0][0]
layer_normalization (LayerNormalizatio…	(None, 3, 64)	128	add[0][0]
dense (Dense)	(None, 3, 128)	8,320	layer_normalizat…
dense_1 (Dense)	(None, 3, 64)	8,256	dense[0][0]
dropout_2 (Dropout)	(None, 3, 64)	0	dense_1[0][0]
add_1 (Add)	(None, 3, 64)	0	layer_normalizat… dropout_2[0][0]
layer_normalizatio… (LayerNormalizatio…	(None, 3, 64)	128	add_1[0][0]
dense_2 (Dense)	(None, 3, 7)	455	layer_normalizat…

Total params: 50,951 (199.03 KB)
Trainable params: 50,951 (199.03 KB)
Non-trainable params: 0 (0.00 B)

5. Train the Model

```
model.fit(X, y, batch_size=2, epochs=100, verbose=0)
print("Model Trained.")
```

```
Model Trained.
```

6. Generate Text (Prediction Function)

```
def generate_text(seed_text, next_words=5):
    for _ in range(next_words):
        token_list = tokenizer.texts_to_sequences([seed_text])[0]
        token_list = pad_sequences([token_list], maxlen=X.shape[1], padding='post')
        predicted_probs = model.predict(token_list, verbose=0)
        predicted_id = np.argmax(predicted_probs[0][-1])  # Get next word
        output_word = tokenizer.index_word.get(predicted_id, '')
        seed_text += " " + output_word
    return seed_text

print(generate_text("hello"))
```

```
hello
```

Start coding or generate with AI.