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import tensorflow as tf
import numpy as np
# Positional Encoding
def positional encoding(max position, d model):
angles = np.arange(max_position)[:, np.newaxis] / np.power(10000, (2 *
(np.arange(d_model)[np.newaxis, :] // 2)) / d_model)
pos encoding = np.zeros((max position, d model))
pos\_encoding[:, 0::2] = np.sin(angles[:, 0::2])
pos\_encoding[:, 1::2] = np.cos(angles[:, 1::2])
return tf.constant(pos_encoding[np.newaxis, ...], dtype=tf.float32)
# Multi-Head Attention
class MultiHeadAttention(tf.keras.layers.Layer):
def ___init___(self, d_model, num_heads):
super(MultiHeadAttention, self). init ()
self.num heads = num heads
self.d \mod el = d \mod el
assert d model \% num heads ==0
self.depth = d \mod l / num heads
self.wq = tf.keras.layers.Dense(d model)
self.wk = tf.keras.layers.Dense(d model)
self.wv = tf.keras.layers.Dense(d model)
self.dense = tf.keras.layers.Dense(d model)
def split heads(self, x, batch size):
x = tf.reshape(x, (batch\_size, -1, self.num\_heads, self.depth))
return tf.transpose(x, perm=[0, 2, 1, 3])
def call(self, q, k, v, mask=None):
batch size = tf.shape(q)[0]
q = self.split heads(self.wq(q), batch size)
k = self.split\_heads(self.wk(k), batch\_size)
v = self.split heads(self.wv(v), batch size)
matmul qk = tf.matmul(q, k, transpose b=True)
dk = tf.cast(tf.shape(k)[-1], tf.float32)
scaled attention logits = matmul qk / tf.math.sqrt(dk)
if mask is not None:
scaled attention logits += (mask * -1e9)
attention_weights = tf.nn.softmax(scaled_attention_logits, axis=-1)
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scaled attention = tf.matmul(attention weights, v)
scaled attention = tf.transpose(scaled attention, perm=[0, 2, 1, 3])
concat attention = tf.reshape(scaled attention, (batch size, -1, self.d model))
return self.dense(concat attention)
# Point-wise Feed-Forward Network
def point wise feed_forward_network(d_model, dff):
return tf.keras.Sequential([
tf.keras.layers.Dense(dff, activation='relu'),
tf.keras.layers.Dense(d model)
# Encoder Layer
class EncoderLayer(tf.keras.layers.Layer):
def init (self, d model, num heads, dff, dropout rate=0.1):
super(EncoderLayer, self).___init___()
self.mha = MultiHeadAttention(d model, num heads)
self.ffn = point_wise_feed_forward_network(d_model, dff)
self.layernorm1 = tf.keras.layers.LayerNormalization(epsilon=1e-6)
self.layernorm2 = tf.keras.layers.LayerNormalization(epsilon=1e-6)
self.dropout1 = tf.keras.layers.Dropout(dropout rate)
self.dropout2 = tf.keras.layers.Dropout(dropout rate)
def call(self, x, training, mask):
attn output = self.mha(x, x, x, mask)
attn_output = self.dropout1(attn_output, training=training)
out1 = self.layernorm1(x + attn\_output)
ffn \quad output = self.ffn(out1)
ffn output = self.dropout2(ffn output, training=training)
return self.layernorm2(out1 + ffn_output)
# Encoder
class Encoder(tf.keras.layers.Layer):
def init (self, num layers, d model, num heads, dff, input vocab size,
maximum position encoding, dropout rate=0.1):
super(Encoder, self).___init___()
self.d \mod el = d \mod el
self.num\_layers = num\_layers
self.embedding = tf.keras.layers.Embedding(input vocab size, d model)
self.pos encoding = positional encoding(maximum position encoding,
d model)
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self.enc layers = [
EncoderLayer(d model, num heads, dff, dropout rate) for in range(num layers)
self.dropout = tf.keras.layers.Dropout(dropout rate)
def call(self, x, training, mask):
seq len = tf.shape(x)[1]
x = self.embedding(x)
x *= tf.math.sqrt(tf.cast(self.d_model, tf.float32))
x += self.pos\_encoding[:, :seq\_len, :]
x = self.dropout(x, training=training)
for i in range(self.num layers):
x = self.enc layers[i](x, training, mask)
return x
# Decoder Layer
class DecoderLayer(tf.keras.layers.Layer):
def ___init___(self, d_model, num_heads, dff, dropout_rate=0.1):
super(DecoderLayer, self). init ()
self.mha1 = MultiHeadAttention(d model, num heads)
self.mha2 = MultiHeadAttention(d model, num heads)
self.ffn = point_wise_feed_forward_network(d_model, dff)
self.layernorm1 = tf.keras.layers.LayerNormalization(epsilon=1e-6)
self.layernorm2 = tf.keras.layers.LayerNormalization(epsilon=1e-6)
self.layernorm3 = tf.keras.layers.LayerNormalization(epsilon=1e-6)
self.dropout1 = tf.keras.layers.Dropout(dropout rate)
self.dropout2 = tf.keras.layers.Dropout(dropout rate)
self.dropout3 = tf.keras.layers.Dropout(dropout rate)
def call(self, x, enc output, training, look ahead mask, padding mask):
attn1 = self.mha1(x, x, x, look\_ahead\_mask)
attn1 = self.dropout1(attn1, training=training)
out1 = self.layernorm1(x + attn1)
attn2 = self.mha2(out1, enc_output, enc_output, padding_mask)
attn2 = self.dropout2(attn2, training=training)
out2 = self.layernorm2(out1 + attn2)
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ffn\_output = self.ffn(out2)
ffn output = self.dropout3(ffn output, training=training)
return self.layernorm3(out2 + ffn output)
# Decoder
class Decoder(tf.keras.layers.Layer):
def init (self, num layers, d model, num heads, dff, target vocab size,
maximum position encoding, dropout rate=0.1):
super(Decoder, self).___init___()
self.d \mod el = d \mod el
self.num\_layers = num\_layers
self.embedding = tf.keras.layers.Embedding(target vocab size, d model)
self.pos encoding
                   = positional encoding(maximum position encoding,
d model)
self.dec layers = [
DecoderLayer(d_model, num_heads, dff, dropout_rate) for _ in range(num_layers)
self.dropout = tf.keras.layers.Dropout(dropout rate)
def call(self, x, enc_output, training, look_ahead_mask, padding_mask):
seq len = tf.shape(x)[1]
attention weights = \{\}
x = self.embedding(x)
x *= tf.math.sqrt(tf.cast(self.d_model, tf.float32))
x += self.pos\_encoding[:, :seq\_len, :]
x = self.dropout(x, training=training)
for i in range(self.num layers):
x = self.dec layers[i](x, enc output, training, look ahead mask, padding mask)
return x
# Transformer
class Transformer(tf.keras.Model):
def init (self, num layers, d model, num heads, dff, input vocab size,
target\_vocab\_size, \ pe\_input, \ pe\_target, \ dropout\_rate=0.1):
super(Transformer, self). init ()
self.encoder = Encoder(num layers, d model, num heads, dff, in-
put_vocab_size, pe_input, dropout_rate)
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self.decoder = Decoder(num layers, d model, num heads, dff, tar-
get_vocab_size, pe_target, dropout_rate)
self.final layer = tf.keras.layers.Dense(target vocab size)
def call(self, inp, tar, training, enc_padding_mask, look_ahead_mask,
dec_padding_mask):
enc output = self.encoder(inp, training, enc padding mask)
dec output = self.decoder(tar, enc output, training, look ahead mask,
dec_padding_mask)
return self.final_layer(dec_output)
# Example Usage
sample transformer = Transformer(
num_layers=4,
d model=128,
num_heads=8,
dff = 512,
input_vocab_size=8500,
target vocab size=8000,
pe input=10000,
pe target=6000
sample input = tf.random.uniform((64, 37), dtype=tf.int64, minval=0,
\max = 200
sample_target = tf.random.uniform((64, 35), dtype=tf.int64, minval=0,
\max = 200
output = sample_transformer(sample_input, sample_target, training=False,
enc_padding_mask=None, look_ahead_mask=None, dec_padding_mask=None)
print(output.shape)
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