

GRU for Text Generation

E23CSEU2354 ADITI JAIN

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# =====  
# GRU BASED TEXT GENERATION  
# =====  
  
import numpy as np  
import tensorflow as tf  
from tensorflow.keras.models import Sequential  
from tensorflow.keras.layers import GRU, Dense, Embedding  
  
# -----  
# DATASET (SAME AS BEFORE)  
# -----  
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.
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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.
"""

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text = text.lower().replace("\n", " ")
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# -----  
# CHARACTER TOKENIZATION  
# -----
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chars = sorted(list(set(text)))  
char_to_index = {c: i for i, c in enumerate(chars)}  
index_to_char = {i: c for c, i in char_to_index.items()}
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vocab_size = len(chars)  
seq_length = 40
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# -----  
# CREATE INPUT-OUTPUT SEQUENCES  
# -----  
X, y = [], []
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for i in range(len(text) - seq_length):  
    X.append([char_to_index[c] for c in text[i:i+seq_length]])  
    y.append(char_to_index[text[i+seq_length]])
```

```
X = np.array(X)  
y = np.array(y)
```

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# -----  
# BUILD GRU MODEL  
# -----  
model = Sequential([  
    Embedding(vocab_size, 64),  
    GRU(128),
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        Dense(vocab_size, activation='softmax')
    ])

model.compile(
    loss='sparse_categorical_crossentropy',
    optimizer='adam'
)

# -----
# TRAIN MODEL
# -----
model.fit(X, y, epochs=15, batch_size=64)

# -----
# TEXT GENERATION FUNCTION
# -----
def generate_text(seed_text, length=300):
    output = seed_text
    for _ in range(length):
        seq = np.array([[char_to_index[c] for c in output[-
seq_length:]]])
        prediction = model.predict(seq, verbose=0)
        next_char = index_to_char[np.argmax(prediction)]
        output += next_char
    return output

# -----
# GENERATE TEXT
# -----
seed = "artificial intelligence "
generated_text = generate_text(seed)

print("Generated Text using GRU:\n")
print(generated_text)

Epoch 1/15
34/34 _____ 1s 6ms/step - loss: 3.2153
Epoch 2/15
34/34 _____ 0s 6ms/step - loss: 2.8862
Epoch 3/15
34/34 _____ 0s 6ms/step - loss: 2.7275
Epoch 4/15
34/34 _____ 0s 5ms/step - loss: 2.5949
Epoch 5/15
34/34 _____ 0s 5ms/step - loss: 2.4868
Epoch 6/15
34/34 _____ 0s 5ms/step - loss: 2.3954
Epoch 7/15
34/34 _____ 0s 5ms/step - loss: 2.3598
Epoch 8/15

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34/34 _____ 0s 6ms/step - loss: 2.3043
Epoch 9/15
34/34 _____ 0s 6ms/step - loss: 2.2097
Epoch 10/15
34/34 _____ 0s 5ms/step - loss: 2.1788
Epoch 11/15
34/34 _____ 0s 5ms/step - loss: 2.1352
Epoch 12/15
34/34 _____ 0s 5ms/step - loss: 2.0808
Epoch 13/15
34/34 _____ 0s 5ms/step - loss: 1.9832
Epoch 14/15
34/34 _____ 0s 5ms/step - loss: 1.9646
Epoch 15/15
34/34 _____ 0s 5ms/step - loss: 1.8692
Generated Text using GRU:
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Limitations of GRU

- GRU models are still computationally expensive compared to simple RNNs.
- Performance depends heavily on the size and diversity of the training dataset.
- GRUs process sequences sequentially, limiting parallel computation.
- Generated text may still suffer from repetition when greedy decoding is used.
- Transformer-based models outperform GRUs on large-scale text generation tasks.