

Amardeep Singh

E23CSEU2189

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# ===== IMPORTS =====
import tensorflow as tf
import numpy as np

# ===== LOAD DATASET FROM FILE =====
with open("sample text.txt", "r", encoding="utf-8") as f:
    text = f.read().lower()

# ===== PREPROCESSING =====
chars = sorted(list(set(text)))
vocab_size = len(chars)

char_to_idx = {c: i for i, c in enumerate(chars)}
idx_to_char = {i: c for i, c in enumerate(chars)}

encoded = np.array([char_to_idx[c] for c in text])

# ===== CREATE SEQUENCES =====
seq_len = 50
X, y = [], []

for i in range(len(encoded) - seq_len):
    X.append(encoded[i:i + seq_len])
    y.append(encoded[i + seq_len])

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

# ===== RNN MODEL =====
model_rnn = tf.keras.Sequential([
    tf.keras.layers.Embedding(vocab_size, 64, input_length=seq_len),
    tf.keras.layers.SimpleRNN(128),
    tf.keras.layers.Dense(vocab_size, activation="softmax")
])

model_rnn.compile(
    loss="sparse_categorical_crossentropy",
    optimizer="adam"
)

# ===== TRAIN =====
model_rnn.fit(X, y, epochs=30, batch_size=64)

# ===== TEMPERATURE SAMPLING =====
def sample_with_temperature(preds, temperature=0.8):
    preds = np.asarray(preds).astype("float64")
    preds = np.log(preds + 1e-8) / temperature
    exp_preds = np.exp(preds)
    preds = exp_preds / np.sum(exp_preds)
    return np.random.choice(len(preds), p=preds)

# ===== TEXT GENERATION =====
def generate_text(seed, length=600, temperature=0.8):
    result = seed

    for _ in range(length):
        encoded_seed = [char_to_idx[c] for c in seed]
        encoded_seed = np.array(encoded_seed).reshape(1, -1)

        preds = model_rnn.predict(encoded_seed, verbose=0)[0]
        next_idx = sample_with_temperature(preds, temperature)
        next_char = idx_to_char[next_idx]

        result += next_char
        seed = seed[1:] + next_char

    return result

# ===== OUTPUT =====
# Use a seed that exists in the corpus
seed_text = text[:50]

print("\nGenerated Text (Simple RNN):\n")
print(generate_text(seed_text))

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Epoch 8/30          0s 6ms/step - loss: 2.2049
34/34 [=====] 0s 6ms/step - loss: 2.0853
Epoch 9/30          0s 6ms/step - loss: 2.0853
34/34 [=====] 0s 6ms/step - loss: 2.0108
Epoch 10/30         0s 6ms/step - loss: 2.0108
34/34 [=====] 0s 6ms/step - loss: 1.8539
Epoch 11/30         0s 7ms/step - loss: 1.8539
34/34 [=====] 0s 6ms/step - loss: 1.8255
Epoch 12/30         0s 6ms/step - loss: 1.8255
34/34 [=====] 0s 6ms/step - loss: 1.7434
Epoch 13/30         0s 6ms/step - loss: 1.7434
34/34 [=====] 0s 6ms/step - loss: 1.6936
Epoch 14/30         0s 6ms/step - loss: 1.6936
34/34 [=====] 0s 6ms/step - loss: 1.5760
Epoch 15/30         0s 6ms/step - loss: 1.5760
34/34 [=====] 0s 6ms/step - loss: 1.5157
Epoch 16/30         0s 6ms/step - loss: 1.5157
34/34 [=====] 0s 6ms/step - loss: 1.4555
Epoch 17/30         0s 6ms/step - loss: 1.4555
34/34 [=====] 0s 6ms/step - loss: 1.4182
Epoch 18/30         0s 6ms/step - loss: 1.4182
34/34 [=====] 0s 6ms/step - loss: 1.3489
Epoch 19/30         0s 6ms/step - loss: 1.3489
34/34 [=====] 0s 6ms/step - loss: 1.2761
Epoch 20/30         0s 6ms/step - loss: 1.2761
34/34 [=====] 0s 7ms/step - loss: 1.1845
Epoch 21/30         0s 7ms/step - loss: 1.1845
34/34 [=====] 0s 6ms/step - loss: 1.1582
Epoch 22/30         0s 6ms/step - loss: 1.1582
34/34 [=====] 0s 6ms/step - loss: 1.0562
Epoch 23/30         0s 6ms/step - loss: 1.0562
34/34 [=====] 0s 7ms/step - loss: 1.0060
Epoch 24/30         0s 7ms/step - loss: 1.0060
34/34 [=====] 0s 7ms/step - loss: 0.9745
Epoch 25/30         0s 7ms/step - loss: 0.9745
34/34 [=====] 0s 6ms/step - loss: 0.8970
Epoch 26/30         0s 6ms/step - loss: 0.8970
34/34 [=====] 0s 7ms/step - loss: 0.8491
Epoch 27/30         0s 7ms/step - loss: 0.8491
34/34 [=====] 0s 7ms/step - loss: 0.8341
Epoch 28/30         0s 7ms/step - loss: 0.8341
34/34 [=====] 0s 6ms/step - loss: 0.7562
Epoch 29/30         0s 6ms/step - loss: 0.7562
34/34 [=====] 0s 6ms/step - loss: 0.7382
34/34 [=====] 0s 6ms/step - loss: 0.7382
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Generated Text (Simple RNN):

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