### EX 9 TEXT GENERATION USING LSTM NETWORKS

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#### PROBLEM STATEMENT

Build a text generation model using Long Short-Term Memory (LSTM) networks. Train the model on a text corpus to generate coherent sequences of text and evaluate the output for fluency and coherence.

Suggested Dataset: Shakespeare Corpus

## Objectives:

- Understand sequential modeling for natural language generation.
- 2. Train a character-level LSTM model to learn language patterns.
- 3. Generate text using a seed prompt and evaluate the results.
- 4. Analyze the fluency and creativity of LSTM-generated outputs.

## Scope:

Text generation is a foundational task in natural language processing. This experiment demonstrates how LSTMs can learn syntactic and semantic patterns over time and generate believable sequences of text. The use of character-level modeling helps capture detailed language structures.

Tools and Libraries Used:

- 1. Python 3.x
- 2. TensorFlow / Keras
- 3. NumPy
- Shakespeare Text Corpus (Tiny Shakespeare)

## Implementation Steps:

## Step 1: Load and Preprocess the Dataset

import tensorflow as tf import numpy as np

text = tf.keras.utils.get\_file('shakespeare.txt',
 'https://raw.githubusercontent.com/karpathy/charrnn/master/data/tinyshakespeare/input.txt')
text = open(text, 'r').read().lower()
chars = sorted(set(text))
c2i = {c: i for i, c in enumerate(chars)}
i2c = {i: c for i, c in enumerate(chars)}

## Step 2: Create Input and Output Sequences

```
seq len = 40
X = \prod
y = []
for i in range(len(text) - seq_len):
  input seq = text[i:i + seq len]
  target char = text[i + seq len]
  X.append([c2i[c] for c in input_seq])
  v.append(c2i[target char])
X = np.array(X)
y = np.array(y)
Step 3: Build the LSTM Model
model = tf.keras.Sequential([
  tf.keras.layers.Embedding(len(chars), 64, input length=seq len),
  tf.keras.layers.LSTM(128),
  tf.keras.layers.Dense(len(chars), activation='softmax')
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model.compile(loss='sparse categorical crossentropy', optimizer='adam')
model.fit(X, y, batch_size=128, epochs=1)
Step 4: Define the Text Generation Function
def generate(seed, length=300):
  seq = [c2i[c] for c in seed.lower()]
  for in range(length):
    inp = np.array(seq[-seq_len:]).reshape(1, -1)
    pred = model.predict(inp, verbose=0)[0]
    next_idx = np.random.choice(len(pred), p=pred)
    seq.append(next_idx)
```

# Step 5: Generate and Display Text

```
print("\nGenerated Text:\n")
print(generate("shall i compare thee to a summer's day?\n"))
```

return seed + ".join(i2c[i] for i in seq[len(seed):])

# **Output:**

Generated Text:

shall i compare thee to a summer's day?

kind worldmbly: be the was of before spyech will of beopker, i ayf.

lucendeo: that what would thim anst marbned unto you.

vicinent: the fyore!

duke intimnes:
we'se sither, and immalio,
foil i for is of autenel, go but, i deas
them our lieg. ruclio?
our to younce a face and poling,
and this the h