Pratical no: 01

1. Write a program to implement sentence segmentation and word tokenization.

```
import nltk
from nltk.tokenize import sent tokenize, word tokenize
# Download the necessary resources
nltk.download('punkt')
def segment sentences(text):
  Segment the input text into sentences.
  :param text: A string containing the text to be segmented.
  :return: A list of sentences.
  sentences = sent tokenize(text)
  return sentences
def tokenize words(sentences):
  Tokenize the input sentences into words.
  :param sentences: A list of sentences.
  :return: A list of lists, where each inner list contains the words of the corresponding
sentence.
  word tokens = [word tokenize(sentence) for sentence in sentences]
  return word tokens
if name == "_main__":
  text = "Hello world! This is a test sentence. Sentence segmentation and word
tokenization are important preprocessing steps."
  # Segment the text into sentences
  sentences = segment sentences(text)
  print("Sentences:")
  for i, sentence in enumerate(sentences):
    print(f"{i+1}: {sentence}")
  # Tokenize each sentence into words
  word tokens = tokenize words(sentences)
  print("\nWord Tokens:")
```

```
for i, words in enumerate(word_tokens):
    print(f"Sentence {i+1} words: {words}")
```

Pratical no: 02

2. Write a program to implement stemming and lemmatization.

```
import nltk
from nltk.stem import PorterStemmer, WordNetLemmatizer
from nltk.tokenize import word tokenize
# Download the necessary resources
nltk.download('punkt')
nltk.download('wordnet')
nltk.download('omw-1.4')
def perform stemming(words):
  Perform stemming on the input words.
  :param words: A list of words to be stemmed.
  :return: A list of stemmed words.
  stemmer = PorterStemmer()
  stemmed_words = [stemmer.stem(word) for word in words]
  return stemmed words
def perform_lemmatization(words):
  Perform lemmatization on the input words.
  :param words: A list of words to be lemmatized.
  :return: A list of lemmatized words.
  lemmatizer = WordNetLemmatizer()
  lemmatized words = [lemmatizer.lemmatize(word) for word in words]
  return lemmatized words
if name == " main ":
  text = "The striped bats are hanging on their feet for best"
  # Tokenize the text into words
  words = word tokenize(text)
  print("Original Words:")
  print(words)
  # Perform stemming
  stemmed words = perform stemming(words)
  print("\nStemmed Words:")
```

```
print(stemmed_words)

# Perform lemmatization
lemmatized_words = perform_lemmatization(words)
print("\nLemmatized Words:")
```

print(lemmatized_words)

Practical no:03

3. Write a program to implement Pos tagging using HMM and Neural Model.

Input:

a) PoS Tagging using HMM with NLTK

```
import nltk
from nltk.tag import hmm
from nltk.corpus import treebank
# Download necessary resources
nltk.download('treebank')
nltk.download('universal tagset')
# Prepare the training data
train data = treebank.tagged sents(tagset='universal')
# Train an HMM tagger
trainer = hmm.HiddenMarkovModelTrainer()
hmm tagger = trainer.train(train data)
# Test the HMM tagger
test_sentence = "The quick brown fox jumps over the lazy dog".split()
hmm tags = hmm tagger.tag(test sentence)
print("HMM Tagging:")
print(hmm tags)
```

B) PoS Tagging using a Neural Model with PyTorch

```
import torch
import torch.nn as nn
import torch.optim as optim
from torch.utils.data import DataLoader, Dataset
import nltk
from nltk.corpus import treebank
from nltk import word tokenize
# Download necessary resources
nltk.download('treebank')
nltk.download('universal tagset')
# Prepare the data
tagged sentences = treebank.tagged sents(tagset='universal')
vocab = set(word for sentence in treebank.sents() for word in sentence)
tags = set(tag for sentence in treebank.tagged sents(tagset='universal') for word, tag
in sentence)
word to ix = \{word: i \text{ for } i, \text{ word in enumerate}(vocab)\}
tag to ix = \{tag: i \text{ for } i, tag \text{ in enumerate}(tags)\}
class POSTaggingDataset(Dataset):
  def init (self, tagged sentences):
     self.sentences = [[word for word, tag in sentence] for sentence in
tagged sentences]
     self.tags = [[tag for word, tag in sentence] for sentence in tagged sentences]
  def len (self):
     return len(self.sentences)
  def getitem (self, idx):
     sentence = self.sentences[idx]
     tags = self.tags[idx]
     return torch.tensor([word_to_ix[word] for word in sentence], dtype=torch.long),
torch.tensor([tag to ix[tag] for tag in tags], dtype=torch.long)
dataset = POSTaggingDataset(tagged sentences)
dataloader = DataLoader(dataset, batch size=1, shuffle=True)
class LSTMTagger(nn.Module):
  def init (self, vocab size, tagset size, embedding dim=64, hidden dim=128):
     super(LSTMTagger, self). init ()
     self.embedding = nn.Embedding(vocab size, embedding dim)
     self.lstm = nn.LSTM(embedding dim, hidden dim, batch first=True)
     self.hidden2tag = nn.Linear(hidden dim, tagset size)
```

```
def forward(self, sentence):
    embeds = self.embedding(sentence)
    lstm out, = self.lstm(embeds)
    tag space = self.hidden2tag(lstm out)
    tag scores = nn.functional.log softmax(tag space, dim=2)
    return tag scores
model = LSTMTagger(len(vocab), len(tag to ix))
loss function = nn.NLLLoss()
optimizer = optim.SGD(model.parameters(), lr=0.1)
# Training the model
for epoch in range(10):
  for sentence, tags in dataloader:
    model.zero grad()
    tag scores = model(sentence)
    loss = loss function(tag scores.view(-1, len(tag_to_ix)), tags.view(-1))
    loss.backward()
    optimizer.step()
# Testing the model
def predict(model, sentence):
  with torch.no grad():
    inputs = torch.tensor([word to ix[word] for word in word tokenize(sentence)],
dtype=torch.long).unsqueeze(0)
    tag scores = model(inputs)
    , predicted tags = torch.max(tag scores, dim=2)
    predicted tags = predicted tags.squeeze().tolist()
    return [(word, list(tag to ix.keys())[tag]) for word, tag in
zip(word tokenize(sentence), predicted tags)]
test sentence = "The quick brown fox jumps over the lazy dog"
neural tags = predict(model, test sentence)
print("Neural Model Tagging:")
print(neural tags)
```

Pratical:04

4. Write a program to Implement syntactic parsing of a given text.

```
import nltk
from nltk import CFG
from nltk.parse.generate import generate
# Define a simple grammar
grammar = CFG.fromstring("""
  S \rightarrow NP VP
  VP -> V NP | V NP PP
  PP -> P NP
  V -> "saw" | "ate" | "walked"
  NP -> "John" | "Mary" | "Bob" | Det N | Det N PP
  Det -> "a" | "an" | "the" | "my"
  N -> "man" | "dog" | "cat" | "telescope" | "park"
  P -> "in" | "on" | "by" | "with"
# Create a parser
parser = nltk.ChartParser(grammar)
# Define a test sentence
sentence = "John saw the man in the park".split()
# Parse the sentence
parses = list(parser.parse(sentence))
# Display the parse trees
for tree in parses:
  print(tree)
  tree.draw()
# If you want to generate all possible sentences according to the grammar
print("Generated sentences:")
for sentence in generate(grammar, n=10):
  print(' '.join(sentence))
```

Pratical no:05

5. Write a program to Implement dependency parsing of a given text.

Input:

```
import spacy
# Load the pre-trained spaCy model
nlp = spacy.load("en_core_web_sm")

# Define a test sentence
sentence = "John saw the man in the park."

# Parse the sentence
doc = nlp(sentence)

# Display the syntactic structure
for token in doc:
    print(f"{token.text:10} {token.dep_:10} {token.head.text:10} {token.head.pos_:10} {[child for child in token.children]}")

# Visualize the parse tree
spacy.displacy.serve(doc, style="dep")
```

```
*IDLE Shell 3.11.1*
File Edit Shell Debug Options Window Help
   Python 3.11.1 (tags/v3.11.1:a7a450f, Dec 6 2022, 19:58:39) [MSC v.1934 64 bit (AMD64)] on win32 Type "help", "copyright", "credits" or "license()" for more information.
          ROOT
                                    VERB
                                                [John, man, .]
              det
                         man
                                    NOUN
                                                []
[the, in]
              dobi
                                    VERB
                        saw
                       man
                                    NOUN
              prep
det
                                                [park]
                         park
                                                [the]
              pobj
   Using the 'dep' visualizer
    Serving on http://0.0.0.0:5000 ...
```

Practical:06

6. Write a program to Implement Named Entity Recognition (NER).

Input:

```
import spacy
# Load the pre-trained spaCy model
nlp = spacy.load("en_core_web_sm")

# Define a test sentence
text = "Apple is looking at buying U.K. startup for $1 billion. Barack Obama was
born on August 4, 1961."

# Process the text
doc = nlp(text)

# Display the named entities
print("Named Entities, their labels, and explanations:")
for ent in doc.ents:
    print(f"{ent.text:20} {ent.label_:10} {spacy.explain(ent.label_)}")

# Visualize the named entities
spacy.displacy.serve(doc, style="ent")
```

```
*IDLE Shell 3.11.1*
File Edit Shell Debug Options Window Help
   Python 3.11.1 (tags/v3.11.1:a7a450f, Dec 6 2022, 19:58:39) [MSC v.1934 64 bit (AMD64)] on win32
   Type "help", "copyright", "credits" or "license()" for more information.
   ======= RESTART: C:/Users/pavan/Downloads/nlp/tweleve.py =========
   Named Entities, their labels, and explanations:
            ORG Companies, agencies, institutions, etc.
                      GPE
                                Countries, cities, states
   U.K.
                   MONEY Monetary values, including unit
PERSON People, including fictional
   $1 billion
   Barack Obama
   August 4, 1961 DATE Absolute or relative dates or periods
   Using the 'ent' visualizer
   Serving on http://0.0.0.0:5000 ...
```