07/10/20

CSEA022 - Notabral danguage Boussing

Digital Assignment

BBCE0715 SanjitCKS Page: 1

Developing on NLP Pipeline

Steps involved:

Input -> Centence -> POS -> demonstication

Sigmentation tagging

Name Entity_ Noun _ Dependency - Stop worlds
lecognition Phrases possing

Co-reference Data Structures

Resolution Depresenting Parsed

Text

(output)

Impact text Taken: (Text Document)

The Found is the reventh largest when agglomeration in Yamil Nodu. It is also to administrative Readquarters of France district. Fluode Ros a hilly totain with undulating topography and semi-ariol climate. Pivor Kaveri, flows through the city and an abundance of lineatone is

fourd in its beds. It is cocated centrally 18BCE0715 in the touth-Andrian Peningula. It is located Sayisticks at 80 km from Coimbatore and 50 km from Taruppus.

Being extenemely popular for the textile industry, a lost of cotton spinning, weaving and tenthing industries can be found in the Region this total early, it was part of the Kongu Nadu fagion in the Cangam age and was ruled by the correct of the was build with all independence was failed in 1947.

Step1: Sentence segmentation

The first step in the spipolite is to brokedown the sext into incluviolual sentences. Itach unit com be imaginal to be a separate idea. This divinding of text into meaningful units is achieved by using upunchadions (?,.,!) as delimites in English. With complex text processing techniques this com be achieved with suit accusacy and efficiency.

By Bhading down the input text we get the following rentences (first of one lixted).

- 1. Escole is the sevente langues when agglomeration in Tamil Nadu
- 2. It is also the administrative Roadquaters of the Freake district

- 3. Enode las a hilly tolsain with undulating Page 3 topography and semi-onix climate.
- A. River Kaveri flowe through the city and an abundance of limestone is found at its rocks.

Step a: Word Toleanisation

Each sentence foother has unite of meaning, that are words. By breaking down sentences into separate words (tokens). This is important for classifying and counting them for a posticular. Sentiment Tokenisation in singlish is easy because words are separated by spaces. Punctivations will also be tokenised because they contain senantic maning too.

from the 1st centence in the pseudine section, after toleonisation.

["Escole", is", "the", "seventh", "laggest", "when", "agglomeration", "in", "many "Tamil", "Nadu"]

Step 8: Posts- of Sporch Tagging for each Token.

To firstlus understood the meaning of the sidence, we read to know the state and world is playing total soft spend togging total with grammatical parts of spend togging total with grammatical parts of spend bable like - Noun, North, advert etc. This is done by with serical based, trulebased, probabilistic of over leading methods. The deep leading methods. The deep leading method is statistical and provides maximum efficiency.

After paroussing the parvious sentence & tagging, Page:4

France, is, the sevent about whom,

Propos Nown verb Intermines Adjective

agglomesodion, in, Tomil, Nadu

Pereposition Novem

Step A: Text demmatisation

noun

In national language worlds appear in different inflections.

It hadre to map their different inflections to the Lame roof world to make the computer consider them as the house world. This is the house forms to make the man removing plurality and tense to get the "lemmer". For example car' and cars' have a very minute difference but the computers clarifies them as different words.

By demmitigging the previous centence.

Frode is the seventhe longest votion agglomosation in Tamil Naolin (Frode be the seventh longest votion agglomosation in Tamil Naolin)

The 'is' is charged to lemma 'be'. Note thate be'(a state of being) is the commentered word of is', one', been', was et }

It is the administrative Readquarters of Excele district

We speak with a lot of Repetion and thought notwell largrages have a lot of maise in the form of fills, words like "and"; "he", "a" ptc. They occur in Dight frequency and can reduce the efficiency of marchine learning models used in the following stops: So they can be flaged as STOPWOPES and fillused out before statistical analysis.

A pre written list of step words are booked up a during sportsing to identify them. Deplending on the domain, the step words differ

By fillering out the previous sentencerce.

* Stop words | Sto

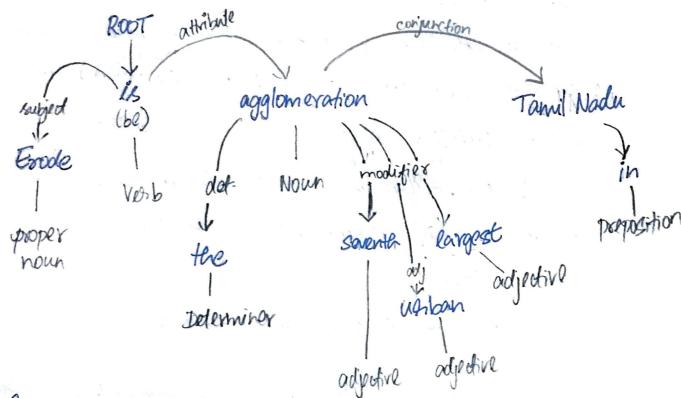
Step 6: Dependency Peorsing

On in this step to relationships of words in a sentence ook mapped livesthically that sulfs in a sold with roof to leaves. The root is the main vorb in the sentence.

-> Pasent world to each world is identified a mapped -> The type of relationship could also be mapped. -> First and group noun phraces.

finding grouping nown pheases that talk about the some thing,

you the first sentence in the text,



Grouping of Nounds

Tamil Nadu (Tomil + Nadeu)

when agglameration (when + agglameration).

Step 7: Named Entity Rocagnition

Some of the words in the processed text contain propornouns that people at the world entities. In this step of Named Entity Decognition these actual physical real world entities are obt detected and tagged.

This is achieved by a nombiration of statistical models and context processing. The NER system detects objects like manes of speople companies, splaysical locations, dates, names of events etc.

From the input text, the fixed line becomes,

Excelle is the seventh longest when application in Tomil
Node

Physical location

Other examples include: Alice went to the mister.

Cocholla was amozing.

physica

loation

Steps: Corefornce Resolution

At this point in the pipeline, most of the suncritic information from the sentence its obtained. But this brings another problem of pronoun usage in the and, and and following sentences 'It' is used to repost to Exade. This information requires passing accross one sentence.

With co-sequenced and nomed entity severation using deep learning more data can be extended. From the input text, the fixed I sentences,

[Exceld is the severall suggest values application in Tarrillandu. [at] is the administrative and quartus of Exade clients (at) has a any surrain with undulating topography and semi-daid climals

Sterre Exode

Implementation and Screenshots

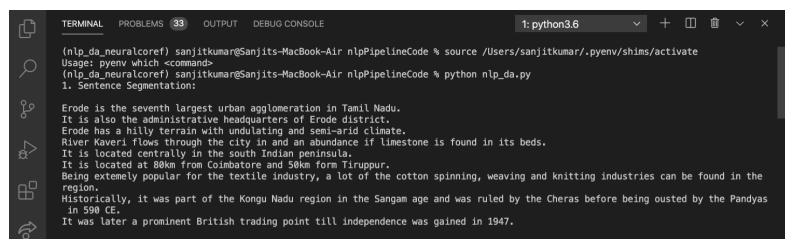
Source Code (Steps and Numbers in Comments)

```
import spacy
import nltk
from nltk.stem.porter import *
from spacy.lang.en import English
# NLP PIPELINE - STEP WISE - SANJIT C K S - 18BCE0715
# Load the large English NLP model
nlp = spacy.load('en core web lg')
# Input Text
text = u"""Erode is the seventh largest urban agglomeration in Tamil
Nadu. It is also the administrative headquarters of Erode district.
Erode has a hilly terrain with undulating and semi-arid climate. River
Kaveri flows through the city in and an abundance if limestone is found
in its beds. It is located centrally in the south Indian peninsula. It
is located at 80km from Coimbatore and 50km form Tiruppur. Being
extemely popular for the textile industry, a lot of the cotton spinning,
weaving and knitting industries can be found in the region.
Historically, it was part of the Kongu Nadu region in the Sangam age and
was ruled by the Cheras before being ousted by the Pandyas in 590 CE. It
was later a prominent British trading point till independence was gained
in 1947."""
# Parse the text with spaCy. This runs the entire pipeline.
doc = nlp(text)
# Segment Sentences
print("1. Sentence Segmentation:\n")
for sent in doc.sents:
    print(sent)
# Word Tokenization
print("\n\n")
print("2. Word Tokenization:\n")
word tokens = []
for word in doc:
    word tokens.append(word.text)
    print(word.text,end=",")
# POS Tagging
print("\n\n")
print("3. POS Tagging:\n")
for word in doc:
    print(word.text, word.pos_, end=", ")
# Lemmatisation
print("\n\n")
print("4. Lemmatisation:\n")
lemmatised = []
```

```
for word in doc:
    lemmatised.append(word.lemma )
    print(word.text + ' ===>', word.lemma_)
# Remove Stop Words
print("\n\n")
print("5. Remove Stop Words:\n")
from spacy.lang.en.stop words import STOP WORDS
filtered_sentence =[]
for word in word_tokens:
    lexeme = nlp.vocab[word]
    if lexeme.is_stop == False:
        filtered_sentence.append(word)
print("Before Removal: ",word_tokens)
print("After Removal: ",filtered_sentence)
# Dependency Parser
print("\n\n")
print("6. Dependency Parser:\n")
from spacy.pipeline import DependencyParser
from spacy import displacy
displacy.serve(doc, style='dep')
# parser = DependencyParser(nlp.vocab)
# processed = parser(doc)
# print(processed)
# NER
print("\n\n")
print("7. Name Entity Recognition:\n")
for entity in doc.ents:
    print(f"{entity.text} ({entity.label })")
# Co-reference Resolution
import neuralcoref
print("\n\n")
print("8. Co-reference resolution:\n")
nlp = en coref md.load()
doc = nlp(test sent)
print(doc._.has_coref)
print(doc._.coref_clusters)
```

Output

Part 1 - Sentence Segmentation



Part 2 - Word Tokenisation

2. Word Tokenization:

Erode, is, the, seventh, largest, urban, agglomeration, in, Tamil, Nadu,., It, is, also, the, administrative, headquarters, of, Erode, district,., Ero de, has, a, hilly, terrain, with, undulating, and, semi, -, arid, climate,., River, Kaveri, flows, through, the, city, in, and, an, abundance, if, limesto ne, is, found, in, its, beds,., It, is, located, centrally, in, the, south, Indian, peninsula,., It, is, located, at, 80, km, from, Coimbatore, and, 50, km, form, Tiruppur,., Being, extemely, popular, for, the, textile, industry,,,a, lot, of, the, cotton, spinning,,, weaving, and, knitting, industries, can, be, found, in, the, region,., Historically,,,it, was, part, of, the, Kongu, Nadu, region, in, the, Sangam, age, and, was, ruled, by, the, Cheras, before, being, ousted, by, the, Pandyas, in, 590, CE,., It, was, later, a, prominent, British, trading, point, till, independence, was, gained, in, 1947,.,

Part 3 - POS Tagging

3. POS Tagging:

Erode VERB, is AUX, the DET, seventh ADJ, largest ADJ, urban ADJ, agglomeration NOUN, in ADP, Tamil PROPN, Nadu PROPN, . PUNCT, It PRON, is AUX, also ADV, the DET, administrative ADJ, headquarters NOUN, of ADP, Erode PROPN, district NOUN, . PUNCT, Erode VERB, ha s AUX, a DET, hilly ADJ, terrain NOUN, with ADP, undulating ADJ, and CCONJ, semi ADJ, - ADJ, arid ADJ, climate NOUN, . PUNCT, Rivore PROPN, Kaveri PROPN, flows VERB, through ADP, the DET, city NOUN, in ADP, and CCONJ, an DET, abundance NOUN, if SCONJ, limestone NOUN, is AUX, found VERB, in ADP, its DET, beds NOUN, . PUNCT, It PRON, is AUX, located VERB, centrally ADV, in ADP, the DET, south ADJ, Indian ADJ, peninsula NOUN, . PUNCT, It PRON, is AUX, located VERB, at ADP, 80 NUM, km NOUN, from ADP, Coimbatore PROPN, and CCONJ, 50 NUM, km NOUN, form NOUN, Tiruppur PROPN, . PUNCT, Being AUX, extemely ADV, popular ADJ, for ADP, the DET, textile NOUN, in dustry NOUN, , PUNCT, a DET, lot NOUN, of ADP, the DET, cotton NOUN, spinning NOUN, , PUNCT, weaving NOUN, and CCONJ, knitting NOUN, industries NOUN, can VERB, be AUX, found VERB, in ADP, the DET, region NOUN, . PUNCT, Historically ADV, , PUNCT, it PRON, was AUX, part NOUN, of ADP, the DET, Kongu PROPN, Nadu PROPN, region NOUN, in ADP, the DET, Sangam PROPN, age NOUN, and CCONJ, was AUX, ru led VERB, by ADP, the DET, Cheras PROPN, before ADP, being AUX, ousted VERB, by ADP, the DET, Pandyas PROPN, in ADP, 590 NUM, CE PR OPN, . PUNCT, It PRON, was AUX, later ADV, a DET, prominent ADJ, British ADJ, trading NOUN, point NOUN, till SCONJ, independence NO UN, was AUX, gained VERB, in ADP, 1947 NUM, . PUNCT,

Part 4 - Lemmatisation

	city ===> city	in ===> in
4. Lemmatisation:	in ===> in	the ===> the
	and ===> and	region ===> region
Erode ===> erode	an ===> an	. ===> .
is ===> be	abundance ===> abundance	Historically ===> historically
the ===> the	if ===> if	, ===> , it ===> -PRON-
seventh ===> seventh	limestone ===> limestone	it ===> -PRON-
largest ===> large	is ===> be	was ===> be
urban ===> urban	found ===> find	part ===> part
agglomeration ===> agglomeration	in ===> in	of ===> of
in ===> in	its ===> -PRON-	the ===> the
Tamil ===> Tamil	beds ===> bed	Kongu ===> Kongu
Nadu ===> Nadu	. ===> .	Nadu ===> Nadu
. ===> .	It ===> -PRON-	region ===> region
It ===> -PRON-	is ===> be	in ===> in
is ===> be	located ===> locate	the ===> the
also ===> also	centrally ===> centrally	Sangam ===> Sangam
the ===> the	in ===> in	age ===> age
administrative ===> administrative	the ===> the	and ===> and
headquarters ===> headquarters	south ===> south	was ===> be
of ===> of	Indian ===> indian	ruled ===> rule
Erode ===> Erode	peninsula ===> peninsula	by ===> by
	. ===> .	the ===> the
district ===> district	It ===> -PRON-	Cheras ===> Cheras
. ===> .	is ===> be	before ===> before
Erode ===> erode	located ===> locate at ===> at	being ===> be
has ===> have	80 ===> 80	ousted ===> oust by ===> by
a ===> a	km ===> km	the ===> the
hilly ===> hilly	from ===> from	Pandyas ===> Pandyas
terrain ===> terrain	Coimbatore ===> Coimbatore	in ===> in
with ===> with	and ===> and	590 ===> 590
undulating ===> undulating	50 ===> 50	CE ===> CE
and ===> and	km ===> km	. ==> .
semi ===> semi	form ===> form	It ===> -PRON-
- ===> -	Tiruppur ===> Tiruppur	was ===> be
arid ===> arid	. ===> .	later ===> later
climate ===> climate	Being ===> be	a ===> a
. ===> .	extemely ===> extemely	prominent ===> prominent
River ===> River	popular ===> popular	British ===> british
Kaveri ===> Kaveri	for ===> for	trading ===> trading
flows ===> flow	the ===> the	point ===> point
through ===> through	textile ===> textile	till ===> till
the ===> the	industry ===> industry	independence ===> independence
	, ===> ,	was ===> be
	a ===> a	gained ===> gain
	lot ===> lot	in ===> in
	of ===> of	1947 ===> 1947

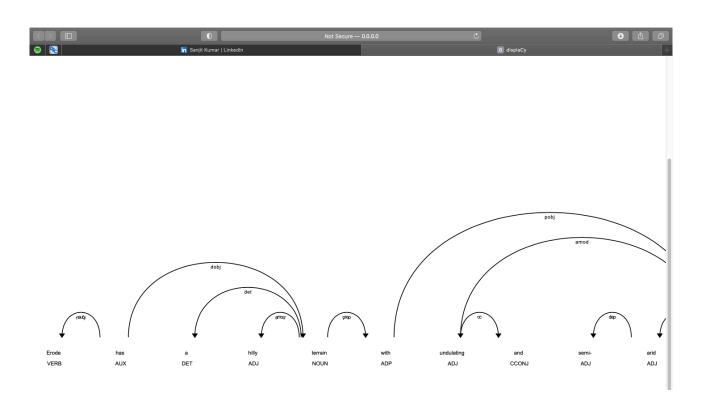
Part 5 - Remove Stop Words

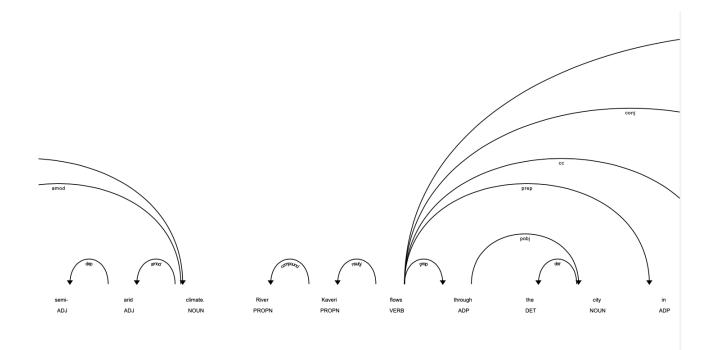
Before and After Removal

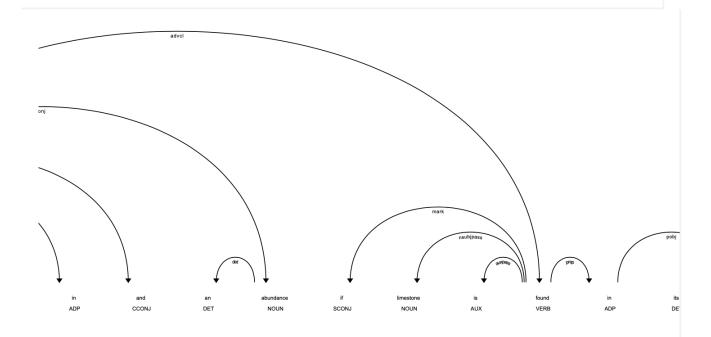
```
Before Removal: ['Erode', 'is', 'the', 'seventh', 'largest', 'urban', 'agglomeration', 'in', 'Tamil', 'Nadu', '.', 'It', 'is', 'also', 'the', 'administrative', 'head quarters', 'of', 'Erode', 'district', '.', 'Erode', 'has', 'a', 'hilly', 'terrain', 'with', 'undulating', 'and', 'semi', '--, 'arid', 'climate', '.', 'River', 'Kaveri ', 'flows', 'through', 'the', 'city', 'in', 'and', 'an', 'abundance', 'if', 'limestone', 'is', 'found', 'in', 'its', 'beds', '., 'It', 'is', 'located', 'centrally', 'in', 'the', 'south', 'Indian', 'peninsula', '.', 'It', 'is', 'located', 'at', '80', 'km', 'from', 'Coimbatore', 'and', '50', 'km', 'form', 'Tiruppur', '.', 'Being', 'extemely', 'popular', 'for', 'the', 'textile', 'industry', ', 'a', 'lot', 'off, 'the', 'cotton', 'spinning', ',', 'weaving', 'and', 'knitting', 'industries', 'can', 'be', 'found', 'in', 'the', 'region', '.', 'Historically', ',', 'it', 'was', 'part', 'of', 'the', 'Kongu', 'Nadu', 'region', 'in', 'the', 'Sangam', 'age', 'and', 'w as', 'ruled', 'by', 'the', 'Cheras', 'before', 'being', 'ousted', 'by', 'the', 'Pandyas', 'in', '590', 'CE', '.', 'It', 'was', 'later', 'a', 'prominent', 'British', 'trading', 'point', 'till', 'independence', 'was', 'gained', 'in', '1947', '.']

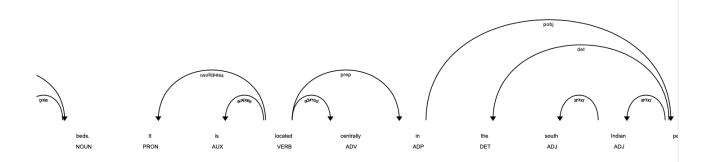
After Removal: ['Erode', 'seventh', 'largest', 'urban', 'agglomeration', 'Tamil', 'Nadu', '.', 'administrative', 'headquarters', 'Erode', 'district', '.', 'Erode', 'hilly', 'terrain', 'undulating', 'semi', '-', 'arid', 'climate', '.', 'River', 'Kaveri', 'flows', 'city', 'abundance', 'limestone', 'found', 'beds', '.', 'located', 'centrally', 'south', 'Indian', 'peninsula', '.', 'located', 'Row', 'Km', 'form', 'Tiruppur', '.', 'extemely', 'popular', 'textile', 'industry', ', 'lot', 'cotton', 'spinning', ', 'weaving', 'knitting', 'industries', 'found', 'region', '.', 'Historically', ', 'Kongu', 'Nadu', 'region', 'Sangam', 'age', 'ruled', 'Cheras', 'ousted', 'Pandyas', '590', 'CE', '.', 'later', 'prominent', 'British', 'trading', 'point', 'till', 'independence', 'gained', '1947', '.']
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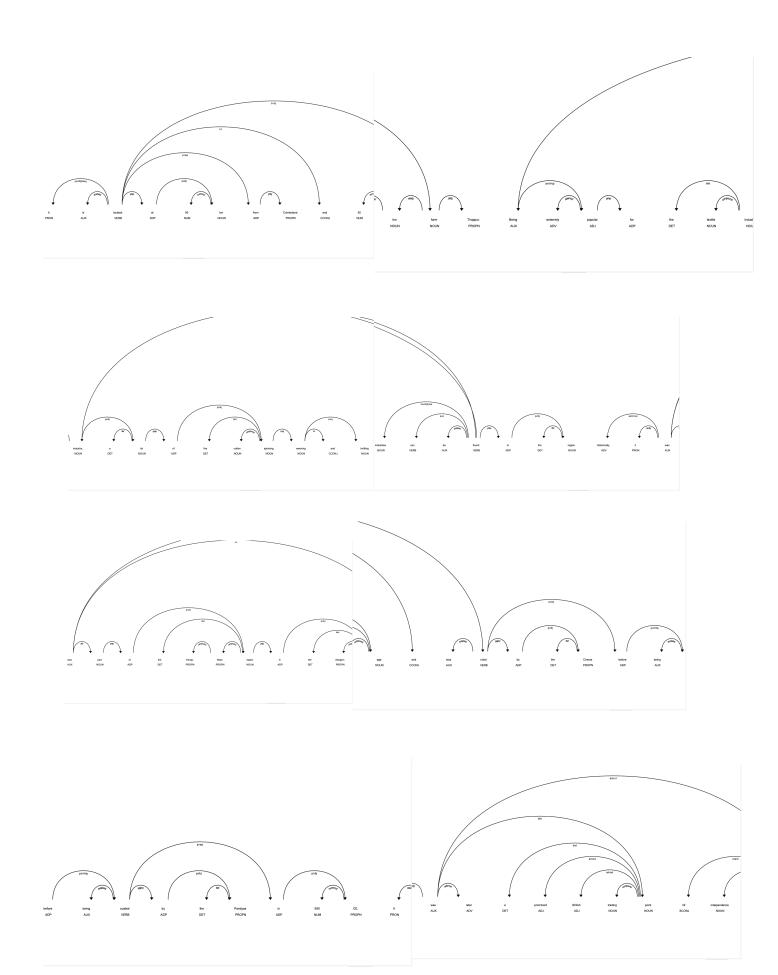
Part 6 - Dependancy Parsing and Graph











Part 7 - Name Entity Recognition

```
7. Name Entity Recognition:

seventh (ORDINAL)
Tamil Nadu (GPE)
Erode district (LOC)
River Kaveri (LOC)
Indian (NORP)
80km (QUANTITY)
Coimbatore (GPE)
50km (QUANTITY)
Tiruppur (GPE)
Kongu Nadu (GPE)
Sangam age (DATE)
Cheras (ORG)
Pandyas (LOC)
590 (CARDINAL)
British (NORP)
1947 (DATE)
```

Part 8 - Co-reference resolution

```
8. Co-reference Resolution:

[Erode: [Erode, It, Erode],
River Kaveri: [River Kaveri, its, It, It],
the Cheras: [the Cheras, they]]
```