# Machine Learning approach to POS Tagging in Bengali Language

### PROJECT REPORT

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### **BONAFIDE CERTIFICATE**

Certified that this project report "Machine Learning Approach to POS Tagging in Bengali" is the bona fide work of Rajdeep Das, Arghyadeep Banerjee, Soham Chakraborty, Tanmay Guchhait, Debabrata Maity, Alik Sarkar, Sanju Manna, students of MCA 5<sup>th</sup> Semester of Heritage Institute of Technology, Kolkata, who carried out the project work under my supervision.

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# **Declaration by Student**

This is to declare that this report has been written by us. No part of the report is plagiarized from other sources. All information included from other sources has been duly acknowledged. We aver that if any part of the report is found to be plagiarized, we'll take full responsibility for it.

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# Table of Contents:

Content	Page Number
1. Abstract & Introduction	6
2. POS Tagging on English Text	6-7
3. Bengali Corpus POS Tagging:	7-8
Reasons behind Bengali POS Tagging Challenges of Bengali POS Tagging	
4. BNLP Toolkit	9
5. Data Collection & Processing	10-13
6. Features and Approaches:	13-16
· Rule-based approach · Conditional Random Field based approach	
4. Sentence-piece Tokenizer	16
5. Confusion Matrices and Performance Analysis	17-18
7. Code-snippets and explanation	18-21
8. Accuracy	22-23
9. References	24-25

# **ABSTRACT:**

Part-of-speech (POS) tagging is a popular Natural Language Processing technique which refers to categorizing words in a text (corpus) in correspondence with a particular part of speech, depending on the definition of the word and its context. It consists of assigning to each word of a text the proper morphosyntactic tag in its context of appearance. It is very useful for a number of NLP applications: as a pre-processing step to syntactic parsing, in information extraction and retrieval (e.g. document classification in internet searchers), text to speech systems, corpus linguistics, etc.

POS tagging techniques, including eight stochastic based methods and eight transformation-based methods.

A comparative analysis of the tagging methods is performed with accuracy measures. CRF shows the highest accuracy of 80% among all the tagging techniques.

# **INTRODUCTION:**

Natural language processing is a subfield of linguistics, computer science, and artificial intelligence concerned with the interactions between computers and human language.

Now Natural language processing (NLP) tools have sparked a great deal of interest due to rapid improvements in information and communications technologies.

As a result, many different NLP tools are being produced. However, there are many challenges for developing efficient and effective NLP tools that accurately process natural languages. One such tool is part of speech (POS) tagging.

# **POS Tagging on English Text:**

A set of all POS tags used in a corpus is called a tagset. Tagsets for different languages are typically different. They can be completely different for unrelated languages and very similar for similar languages, but this is not always the rule. Tagsets can also go to a different level of detail. Basic tagsets may only include tags for the most common parts of speech (N for noun, V for verb, A for adjective etc.). In traditional POS Tagging, English is the most suitable to mostly used easy

to tagging corpus for POS Tagging. Because all the POS tags like nouns, pronouns, verbs, adverbs, adjectives are more compatible for tagging English dataset than any other language. For example,

In the above example, we have used NLTK (Natural Language Toolkit) to tag a English raw text and the outcome is as shown in the above picture.

# **Bengali Corpus POS Tagging:**

Bengali is the second most widely spoken of the 22 scheduled languages of India. With approximately 300 million native speakers and another 37 million as second language speakers, Bengali is the fifth most-spoken native language and the sixth most spoken language by total number of speakers in the world. Therefore, in addition to expanding the field of Bengali language research for this large number of people, it is necessary to enhance modern artificial intelligence-based technology.

Bengali is a highly inflectional language with more than 160 different inflected forms for verbs and 36 different forms for nouns, and 24 different forms for pronouns. Developing an efficient POS tagger is a challenging task for resource-scarce languages like Bengali. Like in English language, the Part of speeches is

noun, verb, adjective, adverb, pronoun, preposition, conjunction, interjection etc which is similar in the Bengali language.

In Bengali we can also tokenize & classify words into their corresponding Parts of Speech tags. Such classifiers help computers in understanding Bengali language, used in different regional real life applications.

# **Challenges in Bengali POS Tagging:**

- The main challenge of POS tagging comes due to the ambiguity in language.
   In Bengali, a large percentage of the words in a corpus are ambiguous.
- Bengali is a morphologically rich language. It is a highly agglutinative language, because of which the vocabulary size grows of a high rate with increase in the size of the corpus.
- Bengali is a relatively free word order language in comparison with European Languages.

As an example, we can consider the English Sentence:

I eat rice (PRP VB NN)

We have the following possible Bengali equivalents of the sentence:

- Ami vat khai (I rice eat) (PPR NC VM)
- Ami khai vat (I eat rice) (PPR VM NC)
- Vat ami khai (rice I eat) (NC PPR VM)
- Vat khai ami (rice eat I ) (NC VM PPR)
- Khai ami vat (Eat I rice) (VM PPR NC)
- Khai vat ami (Eat rice I) (VM NC PPR)

Hence, Part of Speech tagging using Linguistic rules is a difficult problem for such kinds of languages.

```
In [4]: from bnlp import POS
bn_pos = POS()
model_path = "bn_pos.pkl"
text = "আমি ভাত খাই!"
res = bn_pos.tag(model_path, text)
print(res)
[('আমি', 'PPR'), ('ভাত', 'NC'), ('খাই', 'VM'), ('!', 'PU')]
```

# **BNLP Toolkit:**

BNLP is an open source language processing toolkit for Bengali language consisting with tokenization, word embedding, POS tagging, NER tagging facilities. BNLP provides pre-trained model with high accuracy to do model based tokenization, embedding, POS tagging, NER tagging task for Bengali language. BNLP pre-trained model achieves significant results in Bengali text tokenization, word embedding, POS tagging and NER tagging task.

BNLP is providing different tokenization method to tokenize Bengali text efficiently

- BNLP Provides different types of embedding method using their pre-trained models to embed Bengali word it also provides an option to train and embedding model from scratch.
- It provides hands on start option for pos tagging or Name Entity Recognition (NER) tagging of Bengali sentences and also provides an option for training Conditional Random Field based (CRF) approach to pos tagger or NER tagger model from scratch.
- BNLP also provides some utility methods like to remove stop-words from Bengali text, to get Bengali letters list or punctuation list.
- BNLP libraries have a permissive MIT license. BNLP is easy to install via pip or by cloning repository with any python projects.

# Installation

```
pypi package installer(python 3.6, 3.7, 3.8 tested okay)
pip install bnlp_toolkit
or Upgrade
pip install -U bnlp_toolkit
```

# DATA COLLECTION AND PROCESSING:

Each word has its own lexical term written underneath, however, having to constantly write out these full terms when we perform text analysis can very quickly become cumbersome — especially as the size of the corpus grows. Thence, we use a short representation referred to as "tags" to represent the categories. Part-of-speech tags describe the characteristic structure of lexical terms within a sentence or text; therefore, we can use them for making assumptions about semantics. When we perform POS tagging, it's often the case that our tagger will encounter words that were not within the vocabulary that was used. Consequently, augmenting your dataset to include unknown word tokens will aid the tagger in selecting appropriate tags for those words. For POS tagging we used NLTR 8 datasets which contains total of 2997 sentences.

We split that datasets into 2247 train and 750 tests set and train our POS tagging model. BNLP toolkit has been used to tag Bengali words & punctuations using training dataset of NLTR with 80% accuracy & has Bengali CRF NER Tagging which was trained with this data with 90% accuracy. We divided data into 75% train and 25% test. Our evaluation result for POS tagging model is 80.75 F1.

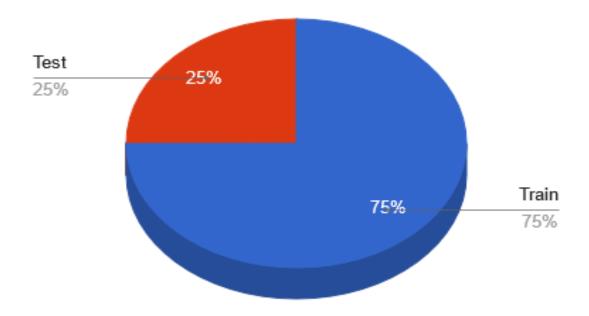
	Sentences	Train	Test	
POS	2997	2247	750	

### Statistics of POS Dataset from NLTR

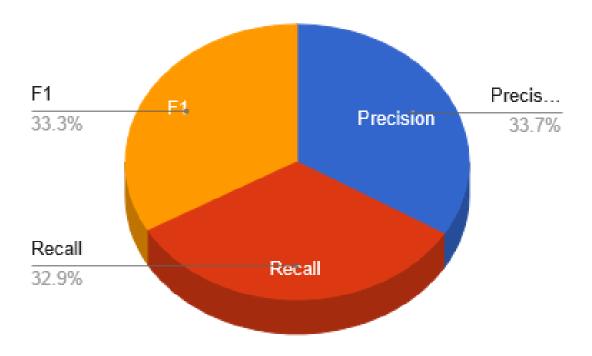
	Precision	Recall	F1	
POS	81.74	79.78	80.75	

**Evaluation result of POS** 

**BNLP POS Tagging using NLTR Dataset** 

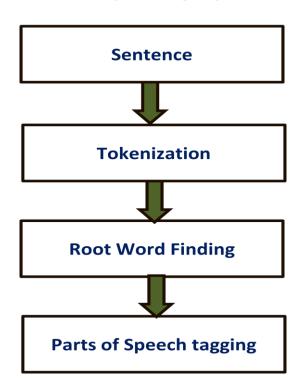


Evaluated result of BNLP POS Tagging using NLTR Dataset



# **Preprocessing on Raw dataset**

# **Raw Text**



# **Sample Testing on Preprocessing**

Raw text = ""আমি ভাত খাই।"

Taking above sentence to apply POS Tagging.

Sentence= "আমি ভাত খাই।"

### Tokenization:

```
In [3]: from bnlp import BasicTokenizer
basic_t = BasicTokenizer()
raw_text = "আমি ভাত খাই।"
tokens = basic_t.tokenize(raw_text)
print(tokens)

['আমি', 'ভাত', 'খাই', '!']
```

# Parts of Speech Tagging

```
In [4]: from bnlp import POS
bn_pos = POS()
model_path = "bn_pos.pkl"
text = "আমি ভাত খাই!"
res = bn_pos.tag(model_path, text)
print(res)

[('আমি', 'PPR'), ('ভাত', 'NC'), ('খাই', 'VM'), ('I', 'PU')]
```

So here, we have taken a raw text as a sentence to apply first POS Tagging where first we have applied tokenization using BasicTokenizer(). After tokenization, we have applied POS() for POS Tagging.

```
[('আমি', 'PPR'), ('ভাত', 'NC'), ('খাই', 'VM'), ('l', 'PU')]
```

# **Features and Approaches:**

# Rule-based approach:

One of the oldest techniques of tagging is rule-based POS tagging. Rule-based taggers use a dictionary (i.e. it can store a number of words) or lexicon for getting possible tags for tagging each word. If the word has more than one possible tag, then rule-based taggers use hand-written rules to identify the correct tag. For example, suppose if the preceding word of a word is an article then the word must be a noun. All such kind of information in rule-based POS tagging is coded in the form of rules. These rules may be either –

- Context-pattern rules
- Or, as Regular expression compiled into finite-state automata, intersected with lexically ambiguous sentence representation.

```
In [4]: import nltk
          from nltk.tokenize import word_tokenize
          text = word_tokenize(" Rabindranath Tagore, Bengali Rabindranath Thākur, (born May 7, 1861, Calcutta [now Kolkata], India-diec
         nltk.pos_tag(text)
Out[4]: [('Rabindranath', 'NNP'),
           ('Tagore', 'NNP'),
           (',', ','),
           ('Bengali', 'NNP'),
           ('Rabindranāth', 'NNP'),
           ('Thākur', 'NNP'),
           (',', ','),
('(', '('),
           ('born', 'VBN'),
('born', 'VBN'),
('May', 'NNP'),
('7', 'CD'),
(',', ','),
('1861', 'CD'),
             ',', ','),
            ('Calcutta', 'NNP'),
           ('[', 'NNP'),
            ('now', 'RB'),
            ('Kolkata', 'NNP'),
            (']', 'NNP'),
```

Now we have test this same approach on Bengali Text POS Tagging using BNLP Toolkit.

BNLP provides three different tokenization options to tokenize Bengali text. Under rule based approach BNLP provides "Basic Tokenizer", a "Punctuation Splitting Tokenizer" and "NLTK Tokenizer".

1. Using Basic Tokenizer we get:

```
In [2]: from bnlp import BasicTokenizer
basic_t = BasicTokenizer()
raw_text = "আমি বাংলায় গান গাই।"
tokens = basic_t.tokenize(raw_text)
print(tokens)
['আমি', 'বাংলায়', 'গান', 'গাই', '|']
```

Here, under rule based approach, we have tokenized each word & punctuation using 'BasicTokenizer' from BNLP on the following raw text.

### 2. Using NLTK Tokenizer we get:

```
In [3]: from bnlp import NLTKTokenizer

bnltk = NLTKTokenizer()

text = "আমি ভাত খাই। সে বাজারে যায়। তিনি কি সত্যিই ভালো মানুষ?"

word_tokens = bnltk.word_tokenize(text)
sentence_tokens = bnltk.sentence_tokenize(text)
print(word_tokens)
print(sentence_tokens)

['আমি', 'ভাত', 'খাই', '।', 'সে', 'বাজারে', 'যায়', '।', 'তিনি', 'কি', 'সত্যিই', 'ভালো', 'মানুষ', '?']
['আমি ভাত খাই।', 'সে বাজারে যায়।', 'তিনি কি সত্যিই ভালো মানুষ?']
```

Under Rule based approach, we have tokenized the above raw text using 'NLTKTokenizer' in two phases. One is Word Tokenization and another is Sentence Tokenizer. In Word Toknization, we have tokenized each word from each sentence and in sentence tokenization; we have tokenized 3 sentences into 3 tokens.

# **Conditional Random Field based approach:**

A CRF is a sequence modeling algorithm which is used to identify entities or patterns in text, such as POS tags. This model not only assumes that features are dependent on each other, but also considers future observations while learning a pattern. In terms of performance, it is considered to be the best method for entity recognition.

- CRF is a discriminative probabilistic classifier
- In CRF based models, the input is:
  - A set of feature derived from the input sequence.
  - Weights associated with the features.
  - Previous label
- Our task is to predict the correct label

The feature functions express certain characteristics of the sequence. Example: the tag sequence noun-> verb -> adjective

```
In [5]: from bnlp import POS
bn_pos = POS()
model_path = "bn_pos.pkl"
text = "আমি বাংলায় গান গাই।"
res = bn_pos.tag(model_path, text)
print(res)
[('আমি', 'PPR'), ('বাংলায়', 'NP'), ('গান', 'NC'), ('গাই', 'NC'), ('I', 'PU')]
```

# SentencePiece Tokenizer:

SentencePiece is an unsupervised text tokenizer and detokenizer mainly for Neural Network-based text generation systems where the vocabulary size is predetermined prior to the neural model training. SentencePiece implements subword units (e.g., byte-pair-encoding (BPE).

Here we have also SentencePiece Tokenizer on Bengali Raw Text as follows:

```
In [6]: from bnlp import SentencepieceTokenizer

bsp = SentencepieceTokenizer()
model_path = "bn_spm.model"
input_text = "আমি ভাত খাই। সে বাজারে যায়।"
tokens = bsp.tokenize(model_path, input_text)
print(tokens)

['_আমি', '_ভাত', '_খাই', '|', '_সে', '_বাজারে', '_যায়', '|']
```

# **Confusion Matrices:**

In the field of machine learning and specifically the problem of statistical classification, a confusion matrix, also known as an error matrix, is a specific table layout that allows visualization of the performance of an algorithm, typically a supervised learning one.

By definition a confusion matrix C is such that C(i, j) is equal to the number of observations known to be in group i, and predicted to be in group j.

Thus in binary classification, the count of

- True negatives is C(0,0) TN
- False negatives is C(1,0) FN
- True positives is C(1,1) TP
- False positives is C(0,1) FP

In this project, while using pre-trained model of Bengali POS Tagger, we have retrieved our values from Precision and Recall from following calculations:

Precision= 81.74: TP/(TP+ FP)

Recall=79.78: TP/(TP+FN)

```
In [5]: from bnlp import POS
bn_pos = POS()
model_path = "bn_pos.pkl"
text = "আমি বাংলায় গান গাই!"
res = bn_pos.tag(model_path, text)
print(res)

[('আমি', 'PPR'), ('বাংলায়', 'NP'), ('গান', 'NC'), ('গাই', 'NC'), ('I', 'PU')]
```

In the above image, the term "গাই" is expected to be tagged as VM (Verb) but it is tagged as NC (Noun). That means, it can be considered as FALSE POSITIVE.

```
In [2]: from bnlp import POS
bn_pos = POS()
model_path = "bn_pos.pkl"
text = "আমি ভাত খাই!"
res = bn_pos.tag(model_path, text)
print(res)
[('আমি', 'PPR'), ('ভাত', 'NC'), ('খাই', 'VM'), ('!', 'PU')]
```

Unlike the previous one, here the term "খাই" is correctly tagged as VM (Verb). So it is TRUE POSITIVE.

# **Code-Snippet**

We have collected corpus from NLTR on which we have applied all our above mentioned POS Tagging approaches using BNLP Toolkit.

# 1. Code-snippet 1

```
In [4]: from bnlp import BasicTokenizer
basic_t = BasicTokenizer()
raw_text = "বাংলা সাহিত্যের নবজাগরণের পথিকৃৎ-কর্মী বঙ্কিমচন্দ্র চট্টোপাধ্যায়। বঙ্কিমচন্দ্রের জন্ম ১৮৩৮ সালের ২৬ শে জুন, অধুনা চব্বিশ পরগণা জেলার হ
tokens = basic_t.tokenize(raw_text)
print(tokens)

['বাংলা', 'সাহিত্যের', 'নবজাগরণের', 'পথিকৃৎ', '-', 'কর্মী', 'বঙ্কিমচন্দ্র', 'চট্টোপাধ্যায়', '।', 'বঙ্কিমচন্দ্রের', 'জন্ম', '১৮৩৮', 'সালের', '
২৬', 'শে', 'জুন', ',', 'অধুনা', 'চব্বিশ', 'পরগণা', 'জেলার', 'অন্তর্গত', 'নৈহাটির', 'কঠিলপাড়া', 'গ্রমে', '।', 'বাবা', 'বাদবচন্দ্র', '
চট্টোপাধ্যায়', 'ছিলেন', 'মেদিনীপুরের', 'কলন্ট্রর', '।', 'নিয়মমাফিক', 'পড়াশোনা', 'শুরুন', 'কর্মছল', 'মেদিনীপুর', 'জেলার', 'এক
', 'ইংরেজি', 'ক্কুলো', '।', 'পবে', 'কঠিলপাড়ায়', 'ফিরে', 'হগলি', 'কলেজে', '।', '১৮৫৬', 'সালে', 'বঙ্কিমচন্দ্র', 'আইন', 'পড়বার', '
জন্ম', 'মেসিডেন্সি', 'কলেজে', 'ভর্তি', 'হন', 'এবং', '১৮৫৭তে', 'সেখান', 'থেকে', 'প্রথম', 'বিভাগে', 'এর্দ্রান্তর্গ, 'পর্জাক্রা', 'বঙ্কিমচন্দ্র', '
দ্বিতীয়', 'বিভাগে', 'প্রথম', 'স্থান', 'অধিকার', 'করেন', '।', 'আইন', 'পড়া', 'শেষ', 'হওয়ার', 'আগেই', 'যশোরের', 'ডেপুটি', 'ম্যাজি
সেইট', '৪', 'ডেপুটি', 'কলেক্ট্রের', 'চাকরি', 'পান', '।']
```

# **Explanation:**

In the above code snippet, we have applied BasicTokenizer().

First we have imported BasicTokenizer function from BNLP.

Taking a variable basic\_t we have assigned the BasicTokenizer function inside it.

Taking a raw text on Bankim Chandra Chattopadhyay's lifestory in raw\_text variable,

Stored the result in tokens variable using basic\_t.tokenize() function with raw\_text variable.

### 2. Code-snippet 2

```
In [4]: from bnlp import NLTKTokenizer

bnltk = NLTKTokenizer()

text = "বাংলা সাহিত্যের নবজাগরণের পথিকৃৎ-কর্মী বস্কিমচন্দ্র চট্টোপাধ্যায়। বস্কিমচন্দ্রের জন্ম ১৮৩৮ সালের ২৬ শে জুন, অধুনা চব্দিশ পরগণা জেলার অন্তর্গ

word tokens = bnltk.word tokenize(text)
sentence_tokens = bnltk.sentence_tokenize(text)
print(word tokens)
print(sentence_tokens)

(

['বাংলা', 'সাহিত্যের', 'নবজাগরণের', 'পাথকৃৎ-কর্মী', 'বাঙ্কামচন্দ্র', 'চট্টোপাধ্যায়', '।', 'বাঙ্কামচন্দ্রের', 'জন্ম', '১৮৩৮', 'সালের', '২৬', '
শে', 'জুন', ',', 'অধুনা', 'চব্বিশ', 'পরগণা', 'ডেলার', 'অন্তর্গতি', 'নৈহার্ত্রির', 'কাইলাপাড়া', 'গ্রীমে', '।', 'বাবান', 'বামনিপুর', 'ডেলার', 'বজুন', '১, বজন', 'হংরে
জি', 'ক্কুলে', '।', 'পরে', 'কাইলাপাড়ায়', 'কিরে', 'ছগলি', 'কলেজে', '।', '১৮৫৬', সালে', 'বেমিনিপুর', 'জনার', 'পড়বার', 'পড়বার', 'জনার', '
প্রেসিডেনি', 'কলেজে', 'ভর্তি', 'হন', 'এবং, '১৮৫৭তে', 'সেখান', '(থকে', 'প্রথম', 'বিভাগে', 'এইমচন্দ্র', 'অন্তর্গন', 'পনার', 'পান', ''
'হান', 'অধিকার', 'করেন', '।', 'আইন', 'পড়া', 'কোকাতা', 'বিশ্ববিদ্যালারেন', 'বেমে', '১৮৫৮', 'মালা', 'বিশ্বমিন্দ্রালার', 'বাবার', 'ক্রিমচন্দ্র', 'জিনিস্র', 'করেন', '।'

['বাংলা সার্বিত্যের নবজাগরণের পথিকৃৎ-কর্মী বিশ্বমিন্দ্রের 'ক্রের', 'এপম', 'বিভাগে', 'ব্রের্জান্তর্গন 'বিভাগে', 'প্রথম', 'ভরের', 'ডপটি', 'কলেই
রের', 'ভাকবি', 'পান', '।''

['বাংলা সার্বিত্যের নবজাগরণের পথিকৃৎ-কর্মী বিশ্বমিন্দ্র ক্রেরার', 'আগেই', 'বানার', 'বেজ্বান্তর্গন বিশ্বমিন্দ্রালার অন্তর্গ
ত নৈহাটির কাঁঠালপাড়া প্রমো', 'বাবা যাদবচন্দ্র চট্টোপাথ্যায়।', 'বিশ্বমচন্দ্র জলা ব্রুম ক্রন্দ্র ব্যাবার কর্মছিল ব্রের্জান ক্রিমচন্দ্র ছিল ক্রের্জন বিশ্বমিন্দ্র জলা
র এক ইংরেজি ক্কুলো', 'পান করেন।', 'আইন ক্রিমচন্দ্র ছিলি কলেজে।', '১৮৫৬ সালে বিশ্বমিন্দ্রালারর প্রথম বি,র প্রীক্রার করিম চন্দ্র হিনিড বিশ্বমিন্দ্র জলা
র এক ইংরেজি ক্কুলো', 'পান করিন।' পান করিন।', '১৮৫৬ সালে বিশ্বমিন্দ্রালার প্রথম বি,র প্রীক্রার বিশ্বমিন।
ব্রুম চন্দ্র বিল্বার (রুমিনিডারি) করেন।' '১৮৫৬ সালে বিশ্বমিন্দ্র জনের ক্রম বিশ্বমিনালারের প্রথম বি,র প্রিমচন্দ্র ছিলি
বিত্ত সেখান বেকে প্রথম বিন্তন করিন।', 'আইন পড়া শেক হেগার আগের তেপুটি মাাজিডেটিটি কলেজিরের চাকরি পান।')
র বিত্ত সেখান
```

### **Explanation:**

In the above code snippet, we have applied NLTKTokenizer().

First we have imported NLTKTokenizer function from BNLP.

Taking a variable bnltk, we have assigned the NLTKTokenizer function inside it.

We have taken two variables, such as word\_tokens and sentence\_tokens and assigned two functions bnltk.word\_tokenize and bnltk.sentence\_tokenize respectively inside it with corresponding Bengali corpus.

Taking a raw text on Bankim Chandra Chattopadhyay's lifestory in the 'text' variable.

We have tokenized it in two phases. One is Word Tokenization and another is Sentence Tokenization.

# 3. Code-snippet 3

```
In [4]: from bnlp import POS
bn_pos = POS()
model_path = "bn_pos.pkl"
text = "airen | mixiona নবজাগরণের পৃথিকৃৎ-কর্মী বৃদ্ধিমচন্দ্র চট্টোপাধ্যায়। বৃদ্ধিমচন্দ্রের জন্ম ১৮৩৮ সালের ২৬ শে জুন, অধুনা চবিনশ পরগণা জেলার অন্তর্গ
res = bn_pos.tag(model_path, text)
print(res)

【('বাংলা', 'NP'), ('সাহিত্যের', 'NC'), ('নবজাগরণের', 'NC'), ('পথিকৃৎ', 'NC'), ('১৮৩৮', 'RDF'), ('সালের', 'NC'), ('বর্জিমচন্দ্রের', 'NP'), ('তেট্টোপাধ্যায়', 'NP'), ('', 'NP'), ('বর্জিমচন্দ্রের', 'NP'), ('জর্মা, 'NC'), ('উল্ডেল্), '(সরগণা', 'NC'), ('জলার', 'NC'), ('অন্তর্গত', 'JJ
'), ('নবর্গার্জ', 'NC'), ('জুন', 'NP'), ('NC'), ('অনুনা', 'NC'), ('চবিনশ', 'NC'), ('বরার', 'NC'), ('অনুনা', 'NC'), ('অনুনা', 'NC'), ('বরার', 'NC'), ('ব
```

### **Explanation:**

In the above code snippet, we have applied POS().

First we have imported POS function from BNLP.

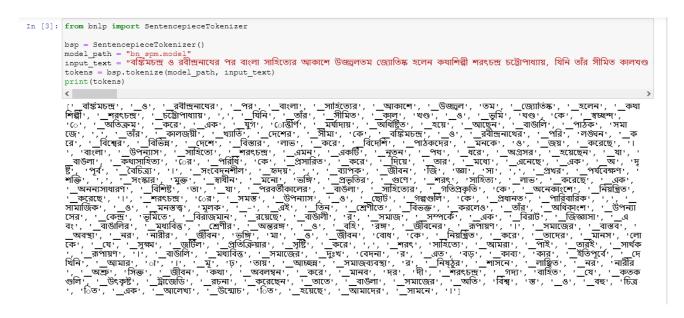
Taking a variable bn\_pos, we have assigned the POS() function inside it.

We have taken a pre-trained model named 'bn\_pos.pkl' in model\_path variable to execute it on the corpus.

Taking a raw text on Bankim Chandra Chattopadhyay's lifestory in the 'text' variable, we have tagged each and every word into different Parts of Speech classifications.

Used bn\_pos.tage function with model path and text and stored the output in res variable.

# 4. Code-snippet 4



# **Explanation:**

In the above code snippet, we have applied SentencepieceTokenizer().

First we have imported SentencepieceTokenizer function from BNLP.

Taking a variable bsp, we have assigned the SentencepieceTokenizer function inside it.

We have taken a pre-trained model names 'bn\_spm.model' in model\_path variable to execute it on the corpus.

Taking a raw text on Bankim Chandra Chattopadhyay's lifestory in input\_text variable, we have tokenized and detokenized into Sentence-pieces.

Used bsp.tokenize() function with model\_path and input\_text variables and stored the outcome in tokens variable.

# **Accuracy:**

Tagger: Unigram Tagger

0.647873865265 0.645268663814 0.634671890304 0.661375661376 0.64314516129 0.6666666666667 0.654210028382 0.625187406297

0.657087189479 0.658976930792

Accuracy Score: 0.649446346367

Tagger: Unigram-Bigram Tagger

0.637048192771 0.652387986214 0.672032193159 0.663613231552 0.662399241347 0.653611393693 0.654751131222 0.66061522945 0.647086914995 0.654237288136

Accuracy Score: 0.655778280254

Tagger: Unigram-Tigram Tagger

0.66021708353 0.65369261477 0.656352555087 0.646687697161 0.643867924528 0.645937358148 0.643706640238 0.678863745787 0.652109911678 0.648975791434

Accuracy Score: 0.653041132236

Tagger: Unigram Bigram Trigram Tagger

0.658385093168 0.648886827458 0.670961347869 0.66004842615 0.661089866157 0.660611854685 0.64859437751 0.639512195122 0.658722592946 0.639906103286

Accuracy Score: 0.654671868435

Tagger: Affix based tagger

0.334864726901 0.355238095238 0.332535885167 0.346898263027 0.335504885993 0.3415 0.354997538159 0.346314325452 0.341048653755 0.337493759361

Accuracy Score: 0.342639613305

Tagger: Affix Unigram Bigram Tigram Tagger

0.773422562141 0.778386454183 0.766697163769 0.756005652379 0.783231083845 0.766909975669 0.778246601031 0.757142857143

0.773729626079 0.776486988848

Accuracy Score: 0.771025896509

Tagger: Brill Based Tagger with AUBT as the trainer Tagger 0.759141882009

 $0.774105930285\ 0.768367346939\ 0.770405727924\ 0.764204545455$ 

 $0.78997020854\ 0.782851344495\ 0.76412649041\ 0.770356816102$ 

0.758321273517

Accuracy Score: 0.770185156567

# References:

# **Rural based Approach:**

https://www.tutorialspoint.com/natural language processing/natural language processing part of speech tagging.htmhttps://www.tutorialspoint.com/natural language processing/natural language processing part of speech tagging.htm

Conditional Random Field: <a href="https://towardsdatascience.com/pos-tagging-using-crfs-ea430c5fb78b">https://towardsdatascience.com/pos-tagging-using-crfs-ea430c5fb78b</a>

**Available Corpus: -**

Bengali Stop-words Collected from: <a href="https://github.com/stopwords-iso/stopwords-bn">https://github.com/stopwords-iso/stopwords-bn</a>

Bengali letters and vowel mark collected from

https://github.com/MinhasKamal/BengaliDictionary/blob/master/BengaliCharacterCombinations.txt

All BNLP Corpus list:

https://github.com/sagorbrur/bnlp/tree/master/bnlp/corpus

Pre-trained models-

Bengali SentencePiece: https://github.com/sagorbrur/bnlp/tree/master/model

**Bengali POS Tagging:** 

https://github.com/sagorbrur/bnlp/blob/master/model/bn\_pos.pkl

Bengali POS (Parts-Of-Speech) Tagging using Indian corpus-

https://medium.com/analytics-vidhya/bengali-pos-part-of-speech-tagging-using-indian-corpus-e85f47d3ad65

Required Training sets: <a href="https://github.com/sagorbrur/bnlp/tree/master/model">https://github.com/sagorbrur/bnlp/tree/master/model</a>

# **Towards POS Tagging Methods for Bengali Language: A Comparative Analysis.**

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