# E1246 - Natural Language Understanding Assignment1 : Language Models

Rupali Dakhane (SR Number 14601)

### **Abstract**

In this assignment I have experimented with various language models on **D1: Brown D2: Gutenberg**.

The best language model is obtained for the given datasets. The tasks performed are:

- Splitting datasets into train, dev and test and building best language model using the evaluation criterion of perplexity.
  - S1: Train: D1-Train, Test: D1-Test
  - S2: Train: D2-Train, Test: D2-Test
  - S3: Train: D1-Train + D2-Train,
     Test: D1-Test
  - S4: Train: D1-Train + D2-Train,Test: D2-Test
- Generating a sentence of 10 tokens using the built language model

#### 1 Text Preprocessing

The following tasks are performed on text obtained from Gutenberg and Brown Corpus

Word Tokenisation

041

044

• Removal of Punctuations

## 2 Handling Unknown Words

Unknown words are handled using absolute discounting method. In which I subtracted 0.75 from each word count and then this count was given to unseen words of test data.

## 3 Implementation

I have tried following experiments with corpus-

where

C(x) = number of times x appears in training  $w_i = i\hbar$  word in the given context

Figure 1: Kartz Backoff

$$eta_{w_{i-n+1}\cdots w_{i-1}} = 1 - \sum_{\{w_i: C(w_{i-n+1}\cdots w_i) > k\}} d_{w_{i-n+1}\cdots w_i} rac{C(w_{i-n+1}\cdots w_{i-1}w_i)}{C(w_{i-n+1}\cdots w_{i-1})}$$

Figure 2: Value of Beta for Kartz Backoff

$$lpha_{w_{i-n+1}\cdots w_{i-1}} = rac{eta_{w_{i-n+1}\cdots w_{i-1}}}{\sum_{\{w_i: C(w_{i-n+1}\cdots w_i) \leq k\}} P_{bo}(w_i \mid w_{i-n+2}\cdots w_{i-1})}$$

Figure 3: Value of Alpha for Kartz Backoff

- Stupid Backoff with absolute discounting on trigram model
  - $-\alpha = 0.4$  is used in figure 1
- Kartz Backoff with absolute discounting on trigram model
  - To compute  $\alpha$ , the value of  $\beta$  is calculated as given in figure 2, And figure 3 gives equation to compute  $\alpha$

#### 4 Perplexity

I have used **perplexity** as metric which is given by following equation.

$$Perplexity(C) = \sqrt[N]{\frac{1}{P(s_1, s_2, ..., s_m)}}$$

```
P(s_1, ..., s_m) = \prod_{i=1}^m p(s_i)
Perplexity(C) = \sqrt[N]{\frac{1}{\prod_{i=1}^m p(s_i)}}
= 2^{\log_2 \left[\prod_{i=1}^m p(s_i)\right]^{-N}}
= 2^{-\frac{1}{N} \log_2 \left[\prod_{i=1}^m p(s_i)\right]}
= 2^{-\frac{1}{N} \sum_{i=1}^m \log_2 p(s_i)}
```

## 5 Result

S1: train = D1 train and test = D1 test	
Stupid Backoff	334.50
Karts Backoff	85.16
S2 : train = D2 train and test = D2 test	
Stupid Backoff	350.34
Karts Backoff	123.65
S3 : train = D1 train+D2 train and test = D1 test	
Stupid Backoff	200.95
Karts Backoff	90.30
S4: train = D1 train+D2 train and test = D2 test	
Stupid Backoff	385.86
Karts Backoff	85.74

#### Sentence Genration

I have used **Shannon method** for generating sentences

- · Generate random sentences:
- · Choose a random bigram "<s> w" according to its probability
- · Now choose a random bigram "w x" according to its probability
- And so on until we randomly choose "</s>"
- · Then string the words together

```
<s> I
    I want
    want to
    to eat
    eat Chinese
    Chinese food
    food </s>
```

## 7 Git Link

https://github.com/Rupali0408/NLU\_Assignment\_1