# In20-S2-CS5616- Natural Language Processing

#### Assignment 1

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Module	CS5616

#### Steps that carried out

1. Calculate the frequencies of each word

```
all_the_content = dict()
for line in lines:
    result = ''.join([i for i in line if (not i.isdigit() and i not in set(string.punctuation) and i not in ['\"', '\"'])])
    words = result.split(' ')
    for word in words:
        if word not in all_the_content.keys():
            all_the_content[word] = 0
            all_the_content[word] += 1

del all_the_content['']
    del all_the_content['\n']
    print(len(all_the_content))

17153

('q@nd\u200ddsow': 72, 'molgap': 1217, 'e@esndveloo': 2, 'vorgeeto': 10, 'morgandamano': 16, '@rdpapen': 17, 'wot': 77
4, 'endral': 507, 'endral': 470, 'mor': 3029, 'endral': 194, 'd'manotaledamor': 91, 'qemanotal'\u200ddsow': 148, 'endral': 10, 'wordsow': 7, 'downoodsow': 14, 'endral': 10, 'endral': 14, 'endral': 1
```

2. Calculate the frequency of frequencies of each word

3. Using the frequencies, you calculate in 1, rank the words as shown in the example in lecture slides

```
import numpy as np
list_list = []
for key, val in freq_of_freq.items():
    list_list.append([int(key), val])

ary = np.array(list_list)
    ary = ary(np.argsort(ary[:,0])]
print(ary)

[[ 1 7566]
    [ 2 2758]
    [ 3 1326]
```

- 4. plot the rank vs frequency graph.
  - Calculate Rank and Frequency

```
rank_n_freq = []
for i, ele in enumerate(ary):
    rank_n_freq.append([i+1, ele[1]])
rank_n_freq = np.array(rank_n_freq)
```

#### • Plot the graph

```
from matplotlib import pyplot as plt
x = rank_n_freq[:,0]
y = rank_n_freq[:,1]
plt.plot(x, y)
plt.xlabel("Rank")
plt.ylabel("Frequency")
Text(0, 0.5, 'Frequency')
   7000
   6000
   5000
 4000
3000
   2000
   1000
     0
              50
                   100
                          150
                               200
                                      250
                                           300
                                                 350
```

### • Log scale graph

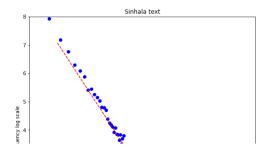
```
In [22]: # from matplotlib.pyplot import axes as ax

yy = [ math.log(ele) for ele in y]
    xx = [ math.log(ele) for ele in x]
    # plt.plot(xx, yy, color='blue')
    fig= plt.figure(figsize=(8, 8))

    axes= fig.add_axes([0.1, 0.1, 0.8, 0.8])

    axes.scatter(xx, yy, color='blue')
    plt.title("Sinhala text")
    plt.xlabel("Rank log scale")
    plt.ylabel("Frequency log scale")

    z = np.polyfit(xx, yy, 1)
    p = np.polyld(z)
    plt.plot(x,p(x),"r--")
    axes.set(xlim=(0, 8), ylim=(0, 8))
Out[22]: [(0, 8), (0, 8)]
```



# 5. What can you comment on the graphs?

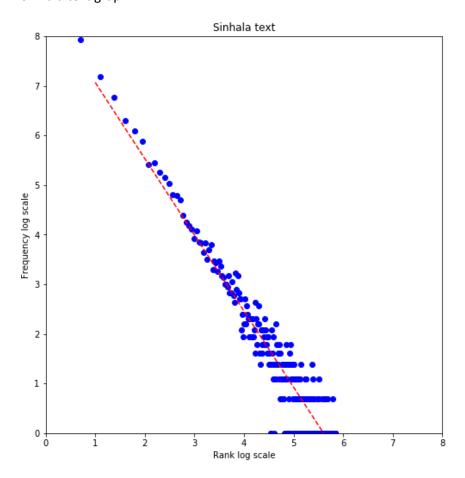
# Zipf's Law

This law states that given some word corpus a small number of words are used all the time, while the vast majority are used very rarely.

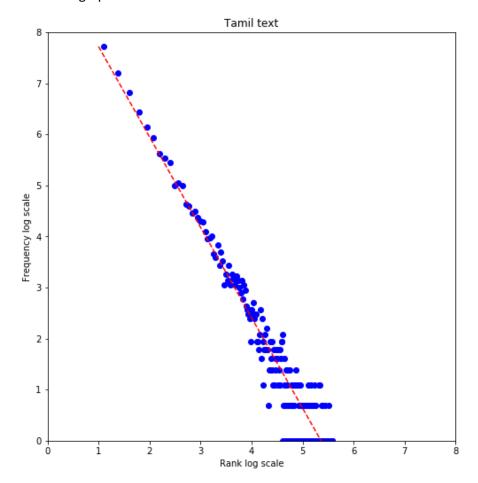
However, there is an interesting fact that in a given corpus of natural language utterance, the frequency of any word is inversely proportional to its rank in the frequency table. Following results exactly show this phenomenon.

#### **Results**

Sinhala text graph



# Tamil text graph



This law can be written as follows

The r<sup>th</sup> most frequent word has a frequency f(r) that scales according to

$$f(r) \propto \frac{1}{r^{\alpha}}$$

Where alpha ~ 1,

This seems very strange behavior, but this was observed for many other statistics that follow an exponential distribution.