nlp hw2

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1 manual classification

guess: R1 is possitive

possible motivation: these may result in a positive sentiment identification:

- positive words: R1 has
- double denial: "If Busy doesnt get a nomination...it would be a disaster.
- comparison: "[another thing] which wasnt all that great"
- punctuation mark: exclamation mark and question mark may indicate excitement, strong dissatisfaction, or doubts. This can relate to nagetive sentiment. so absence of these marks indecate positive sentiment in this way.

2 Tokenization

2.1 making your own tokenizer

v1:

```
text = text.replace('.','')
text = text.replace(',','')
tokenized_text = str.split(text,' ')
sample_string1 = "If you have the chance, watch it. Although, a warning, you'll cry your eyes or
sample_string2 = "Whatever is worth doing is worth doing well." #Write two more sample sentences
sample_string3 = "The hard part isn't making the decision. It's living with it."
```

description:

we designed it mainly because a sentence is natually consisted by words which are seperated by comma and period. So if you replace all the marks with space, than split the string, you may get a series of words.

```
['If', 'you', 'have', 'the', 'chance', 'watch', 'it', 'Although', 'a', 'warning', "you'll", 'cry', 'your', 'eyes', 'out']
```

```
['Whatever', 'is', 'worth', 'doing', 'is', 'worth', 'doing', 'well']
['The', 'hard', 'part', 'isn't', 'making', 'the', 'decision', 'It's', 'living', 'with', 'it']
```

v2:

```
regular=re.compile("[a-zA-z']+|[0-9]+\.[0-9]+?|[^a-z0-9A-Z\s]+")
tokenized_text=regular.findall(text)
sample_string1 = "If you have the chance, watch it. Although, a warning, you'll cry your eyes or
sample_string2 = "Night gathers, and now my watch begins. It shall not end until my death. I sha
sample_string3 = "Imagination is more important than knowledge"
```

description:

this is an updated version, we use regular expression to match words instead of split sentence. This can recognize words, numbers, then other combinations of characters other than spaces. And the three matching methods have priority to prevent token like "you?".

```
['If', 'you', 'have', 'the', 'chance', ',', 'watch', 'it', '.', 'Although', ',', 'a', 'warning', ',', "you'll", 'cry', 'your', 'eyes', 'out', '.']

['Night', 'gathers', ',', 'and', 'now', 'my', 'watch', 'begins', '.', 'It', 'shall', 'not', 'end', 'until', 'my', 'death', '.', 'I', 'shall', 'take', 'no', 'wife', ',', 'hold', 'no', 'lands', ',', 'father', 'no', 'children', '.', 'I', 'shall', 'wear', 'no', 'crowns', 'and', 'win', 'no', 'glory', '.', 'I', 'shall', 'live', 'and', 'die', 'at', 'my', 'post', '.', 'I', 'am', 'the', 'sword', 'in', 'the', 'darkness', '.', 'I', 'am', 'the', 'watcher', 'on', 'the', 'walls', '.', 'I', 'am', 'the', 'fire', 'that', 'burns', 'against', 'the', 'cold', ',', 'the', 'light', 'that', 'brings', 'the', 'dawn', ',', 'the, 'horn', 'that', 'wakes', 'the', 'sleepers', ',', 'the', 'shield', 'that', 'guards', 'the', 'realms', 'of', 'men', '.', 'I', 'pledge', 'my', 'life', 'and', 'honor', 'to', 'the', 'Night', "", 's', 'Watch', ',', 'for', 'this', 'night', 'and', 'all', 'the', 'nights', 'to', 'come', '.']

['lmagination', 'is', 'more', 'important', 'than', 'knowledge']

description:after do
```

2.2 Using an off-the-shelf tokenizer

- differences and which better:
 - clearly, the mature API considered more situations like proper nouns and it is not a perfect way to replace the comma and period because firstly it directly changes some words(like C.I.A to CIA), secondly loses imformation like number of sentences. So we developed it and have a version 2 using regular expression to include numbers, titles with '.',and other conbinations. Also, our v1 tokenizer didn't consider other marks like "?/n:". So in this case, the function word_tokenize() from nltk is definately better.
- even when we update our own tokenizer, we still find it difficult to manage it perfectly. Still some
 words are special like C.I.A. which can't be recognized. Maybe a vocabulary including special
 words is necessary. Also we found the nltk is not always perfect.

see OMG #Twitter is sooooo coooool <3 :-) <-- lol...why do i write like this idk right? :) nltk tokenizer seperate :-) into ':', '-', ')' but it should be a character emotions. But our tokenize did well.

• Tokenizer should base on the corpus to achieve optimal performance. So there is no perfect tokenizer at all.

2.3 vocabulary

• q1:

unigrams: 148127 bigrams: 148126 trigrams 148125

• q2:

the 8437

and 4144

a 4091

of 3673

to 3324

is 2731

in 2386

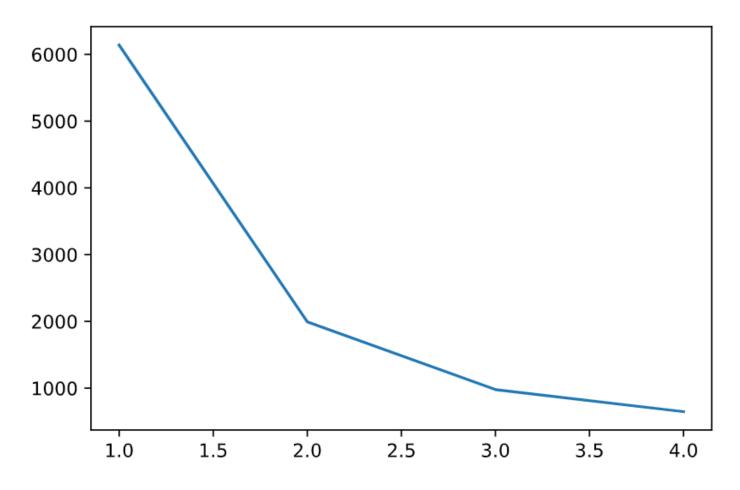
it 2178

i 2028

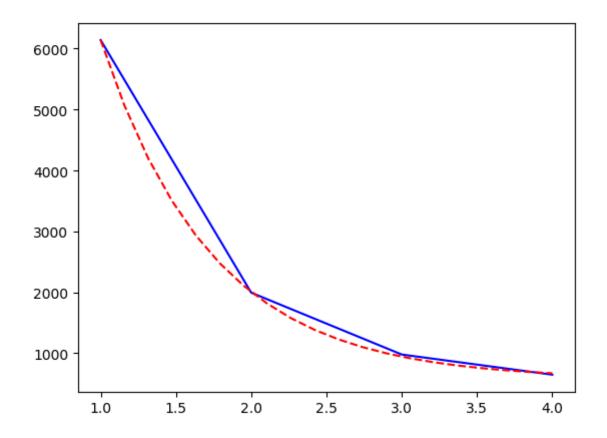
this 1738

This words are not specific verbs, noun, or adjective. They are article, preposition, linking verb or pronoun.

q3
 times for occuring 1,2,3,4 are [6140, 1994, 979, 648]
 it looks like



we thought this is a **Exponential distribution** assume func:y=a * np.exp(-b * x) + c using curve_fit(func, xdata, ydata)



[a,b,c]=[2.16187846e+04 1.35754115e+00 5.76131550e+02]

3 Text classification with a unigram language model

k = 1

Incorrect N-test3

Incorrect P-test32

Incorrect P-test33

Incorrect P-test34

Incorrect P-test35

Incorrect P-test36

Incorrect P-test37

Incorrect P-test39

Incorrect P-test40

Incorrect P-test43

Incorrect P-test44

Incorrect P-test45

Incorrect P-test47

Incorrect P-test48

Incorrect P-test49

35 out of 50 were correct! accuracy 0.7

k = 5

Incorrect N-test3

Incorrect P-test32

Incorrect P-test33

Incorrect P-test34

Incorrect P-test35

Incorrect P-test37

Incorrect P-test39

Incorrect P-test40

Incorrect P-test43

Incorrect P-test44

Incorrect P-test45

Incorrect P-test47

Incorrect P-test48

Incorrect P-test49

• 36 out of 50 were correct! accuracy 0.72

4 Text classification with a bigram language model

4.1 Report how you compute p(w|w')

is the number of times the word appears

is the number of times two word appears together

4.2 What is the accuracy of this second model on the test?

k=1

Incorrect N-test0

Incorrect N-test13

Incorrect N-test2

Incorrect P-test25

Incorrect P-test28

Incorrect P-test30

Incorrect P-test31

Incorrect P-test32

Incorrect P-test35

Incorrect P-test37

Incorrect P-test40

Incorrect P-test42

Incorrect P-test43

Incorrect P-test44

Incorrect P-test45

Incorrect P-test47

Incorrect P-test48

Incorrect P-test49

 32 out of 50 were correct! accuracy 0.64

5 Improving the classifier

- First, we can increase the parameter k of laplace smoothing.
- Second, we can set some stop words, very frequent words like the and a.
- Third, use binary NB,

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