

دانشگاه صنعتی امیر کبیر ( پلی تکنیک تهران )

گزارش پروژه ی چهارم درس مبانی هوش مصنوعی استاد روشن فکر

آرمین ذوالفقاری داریانی ۹۷۳۱۰۸۲ و امیرحسین رجبپور ۹۷۳۱۰۸۵

## تاثیر حذف کلمات پرتکرار و کم تکرار در دقت بدست آمده:

• Epsilon: 0.2

Lambda: [L1:0.1, L2:0.3, L3:0.6]

Cut\_down = 0 Cut\_above = 0

```
# create bigram model object
lambda_arr = [0.1, 0.3, 0.6]  # [h0, h1, h2]
epsilon = 0.2
cut_down = 0
cut_above = 0
f_name_ == "_main_"
Main ×

C:\Users\ASUS\AppData\Local\Programs\Python\I
*** Bigram model ***
Accuracy in test set : 64.44652908067542
*** Unigram model ***
Accuracy in test set : 55.72232645403376
```

• Epsilon: 0.2

Lambda: [L1:0.1, L2:0.3, L3:0.6]

Cut\_down = 2 Cut\_above = 10

```
# create bigram model object
lambda_arr = [0.1, 0.3, 0.6]  # [h0, h1, h2]
epsilon = 0.2
cut_down = 2
cut_above = 10

f_name_ == "_main_"

Main ×

C:\Users\ASUS\AppData\Local\Programs\Python\F
*** Bigram model ***
Accuracy in test set : 68.29268292682927
*** Unigram model ***
Accuracy in test set : 58.724202626641656
```

تحلیل: می توان دید که با حذف کردن کلمات با تکرار بسیار کم (۲ به پایین (که معمولا اسامی خاص هستند و تاثیر چندانی در بار مثبت یا منفی داشتن جملات ندارند)) و تکرار بسیار زیاد

(۱۰ کلمه با بیشترین تکرار (این کلمات عموما stop words هستند و تاثیری در بار مثبت یا منفی داشتن جملات ندارند)) بازدهی بهتری داریم و مدل میتواند بهتر بار مثبت و یا منفی داشتن کامنتها را تشخیص دهد.

## تاثیر مقدار λ و اپسیلون دقت بدست آمده:

• Epsilon: 0.2

Lambda: [L1:0.33, L2:0.33, L3:0.33]

```
# create bigram model object
lambda_arr = [0.33, 0.33, 0.33] # [h0, h1, h2]
epsilon = 0.2
cut_down = 2
cut_above = 10
if __name__ == "__main_"
Main ×

C:\Users\ASUS\AppData\Local\Programs\Python\Pyt
*** Bigram model ***
Accuracy in test set : 64.9155722326454
*** Unigram model ***
Accuracy in test set : 57.223264540337716
```

• Epsilon: 0.2

Lambda: [L1:0.1, L2:0.3, L3:0.6]

```
# create bigram model object
lambda_arr = [0.1, 0.3, 0.6]  # [h0, h1, h2]
epsilon = 0.2
cut_down = 2
cut_above = 10

f__name__ == "__main__"

Main ×

C:\Users\ASUS\AppData\Local\Programs\Python\F
*** Bigram model ***
Accuracy in test set : 68.29268292682927
*** Unigram model ***
Accuracy in test set : 58.724202626641656
```

Lambda: [L1:0.2, L2:0.4, L3:0.4]

```
# create bigram model object
lambda_arr = [0.2, 0.4, 0.4]  # [h0, h1, h2]
epsilon = 0.2
cut_down = 2
cut_above = 10

f_name_ == "_main_"
Main ×

C:\Users\ASUS\AppData\Local\Programs\Python\
*** Bigram model ***
Accuracy in test set : 67.54221388367729
*** Unigram model ***
Accuracy in test set : 56.56660412757974
```

• Epsilon: 0.2

Lambda: [L1:0.6, L2:0.3, L3:0.1]

```
# create bigram model object
lambda_arr = [0.6, 0.3, 0.1]  # [h0, h1, h2]
epsilon = 0.2
cut_down = 2
cut_above = 10

f__name__ == "__main__"
Main ×

C:\Users\ASUS\AppData\Local\Programs\Python\F
*** Bigram model ***
Accuracy in test set : 63.789868667917446
*** Unigram model ***
Accuracy in test set : 57.692307692307686
```

Lambda: [L1:0.33, L2:0.33, L3:0.33]

```
# create bigram model object
lambda_arr = [0.33, 0.33, 0.33] # [h0, h1, h2]
epsilon = 0.5
cut_down = 2
cut_above = 10

f_name_ == "_main_"

Main ×

C:\Users\ASUS\AppData\Local\Programs\Python\Pyt
*** Bigram model ***
Accuracy in test set : 65.38461538461539
*** Unigram model ***
Accuracy in test set : 56.19136960600375
```

• Epsilon: 0.5

Lambda: [L1:0.1, L2:0.3, L3:0.6]

```
# create bigram model object
lambda_arr = [0.1, 0.3, 0.6]  # [h0, h1, h2]
epsilon = 0.5
cut_down = 2
cut_above = 10

f_name__ == "_main_"
Main ×

C:\Users\ASUS\AppData\Local\Programs\Python\
*** Bigram model ***
Accuracy in test set : 65.1031894934334
*** Unigram model ***
Accuracy in test set : 54.971857410881796
```

Lambda: [L1:0.2, L2:0.4, L3:0.4]

```
# create bigram model object
lambda_arr = [0.2, 0.4, 0.4]  # [h0, h1, h2]
epsilon = 0.5
cut_down = 2
cut_above = 10

f_name_ == "_main_"

Main ×

C:\Users\ASUS\AppData\Local\Programs\Python\I
*** Bigram model ***
Accuracy in test set : 66.22889305816135
*** Unigram model ***
Accuracy in test set : 57.692307692307686
```

• Epsilon: 0.5

Lambda: [L1:0.6, L2:0.3, L3:0.1]

```
# create bigram model object
lambda_arr = [0.6, 0.3, 0.1]  # [h0, h1, h2]
epsilon = 0.5
cut_down = 2
cut_above = 10

f_name_ == "_main_"

Main ×

C:\Users\ASUS\AppData\Local\Programs\Python\N
*** Bigram model ***
Accuracy in test set : 65.29080675422139
*** Unigram model ***
Accuracy in test set : 58.25515947467167
```

Lambda: [L1:0.33, L2:0.33, L3:0.33]

```
# create bigram model object
lambda_arr = [0.33, 0.33, 0.33] # [h0, h1, h2]
epsilon = 0.8
cut_down = 2
cut_above = 10

f_name_ == "_main_"

Main ×

C:\Users\ASUS\AppData\Local\Programs\Python\Pyt
*** Bigram model ***
Accuracy in test set : 62.007504690431524
*** Unigram model ***
Accuracy in test set : 55.34709193245778
```

• Epsilon: 0.8

Lambda: [L1:0.1 , L2:0.3 , L3:0.6]

```
# create bigram model object
lambda_arr = [0.1, 0.3, 0.6]  # [h0, h1, h2]
epsilon = 0.8
cut_down = 2
cut_above = 10

f_name_ == "_main_"

Main ×

C:\Users\ASUS\AppData\Local\Programs\Python\N
*** Bigram model ***
Accuracy in test set : 63.977485928705434
*** Unigram model ***
Accuracy in test set : 58.16135084427767
```

Lambda: [L1:0.2, L2:0.4, L3:0.4]

```
# create bigram model object
lambda_arr = [0.2, 0.4, 0.4]  # [h0, h1, h2]
epsilon = 0.8
cut_down = 2
cut_above = 10

f_name_ == "_main_"

Main ×

C:\Users\ASUS\AppData\Local\Programs\Python\U*** Bigram model ***
Accuracy in test set : 63.977485928705434
*** Unigram model ***
Accuracy in test set : 58.16135084427767
```

• Epsilon: 0.8

Lambda: [L1:0.6, L2:0.3, L3:0.1]

```
# create bigram model object
lambda_arr = [0.6, 0.3, 0.1]  # [h0, h1, h2]
epsilon = 0.8
cut_down = 2
cut_above = 10

f_name__ == "_main_"

Main ×

C:\Users\ASUS\AppData\Local\Programs\Python\0 *** Bigram model ***
Accuracy in test set : 64.25891181988743
*** Unigram model ***
Accuracy in test set : 56.94183864915572
```

این قسمت را به  $\pi$  بخش تقسیم کردیم (اپسیلون 0.0, 0.0, 0.0) و سپس در هر بخش  $\pi$  آرایه که لاندای متفاوت را تست کردیم که یکی با مقادیر صعودی، دیگری با مقادیر نرولی و دیگری با مقادیر مساوی و یکی با مقادیر [0.2, 0.4, 0.4] و دیدیم که در حالت صعودی دیگری با مقادیر مساوی و یکی با مقادیر [0.2, 0.4, 0.4] و دیدیم که در حالت صعودی (حالتی که بایگرام وزن بیشتری دارد) و اپسیلون مقدار کم تری ([0.2, 0.4]) دارد بهترین نتیجه را گرفتیم.

## بهترین دقت دستیافته و پارامترها در آن:

• Epsilon: 0.2

Lambda: [L1:0.1, L2:0.3, L3:0.6]

Cut\_down = 2 Cut\_above = 10

```
# create bigram model object
lambda_arr = [0.1, 0.3, 0.6]  # [h0, h1, h2]
epsilon = 0.2
cut_down = 2
cut_above = 10

f_name_ == "_main_"
Main ×

C:\Users\ASUS\AppData\Local\Programs\Python\F
*** Bigram model ***
Accuracy in test set : 68.29268292682927
*** Unigram model ***
Accuracy in test set : 58.724202626641656
```