# **EX 2: Language Models and Entropy**

Jing Huang (933039277)

# 1. Training Language Models

2.

#### unigram:

```
Derivations found for all 100 inputs

Viterbi (best path) product of probs=e^-27039.4213064869, probability=2^-39009.6

per-input-symbol-perplexity(N=9480)=2^4.11494 per-line-perplexity(N=100)=2^390.096
```

entropy: 4.11494

## bigram:

```
Derivations found for all 100 inputs

Viterbi (best path) product of probs=e^-22539.9452505938, probability=2^-32518.3

per-input-symbol-perplexity(N=9480)=2^3.4302 per-line-perplexity(N=100)=2^325.1

83
```

entropy: 3.4302

#### trigram:

```
Derivations found for all 100 inputs

Viterbi (best path) product of probs=e^-19222.7607662978, probability=2^-27732.6

per-input-symbol-perplexity(N=9480)=2^2.92538 per-line-perplexity(N=100)=2^277.

326
```

entropy: 2.92538

# 3. unigram:

```
Number of states in result: 3
Number of arcs in result: 29
Number of paths in result (valid for acyclic only; a cycle means infinitely many): 1
Number of cycle-causing arcs in result: 27
```

## bigram:

```
Number of states in result: 30
Number of arcs in result: 785
Number of paths in result (valid for acyclic only; a cycle means infinitely many
): 3812798743586.9
Number of cycle-causing arcs in result: 378
```

#### trigram:

```
Number of states in result: 759

Number of arcs in result: 21197

Number of paths in result (valid for acyclic only; a cycle means infinitely many
): e^253.822874920708

Number of cycle-causing arcs in result: 9504
```

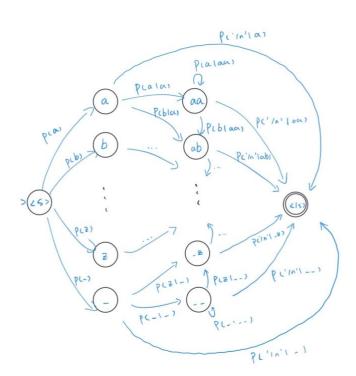
4.

I used the Witten-Bell smoothing.

$$P(w_i|w_{i-1} = \lambda_{w_{i-1}} P_{ML}(w_i|w_{i-1} + (1 - \lambda_{w_{i-1}}) P(w_i)$$
$$\lambda_{w_{i-1}} = 1 - \frac{T(w_{i-1})}{T(w_{i-1}) + N(w_{i-1})}$$

I used 'train.txt' for training, 'dev.txt' for fiddling  $\lambda$ , and 'test.txt' for testing.

5.



# 2. Using Language Models

1. unigram: doesn't make sense.

```
<s> c _ e s b o _ a o l e a l d s n a e o _ a l _ i a w i _ s n s i _ s a n _ a a d _ f t _ e e
v a h w </s> e^-147.783516374817
<s> - _ _ e i _ _ r e t o n _ n g i h e t b a y r n n t t n o e v n g a i _ i e e u _ s e o d o d
e s _ i a a m h d _ i r e w _ i h r n p d i e n t e _ u r n n f m w _ g s s n h _ p _ c w t _ h </s>
e^-276.984808986575

<s> h m _ r d s n n o r i r a i n w r d e _ k e o l e p a o </s> e^-91.3337275834848
<s> _ n i e s s n d m a _ a t _ o _ g k u s w e t l _ t p e t s v _ r t _ o _ k _ i _ l o n a m
t a e e a _ a w o e w s a s </s> e^-178.29466683483

<s> > o t t l n t s s n _ u a </s> 3.27846625368167e-17

<s> h o _ o s _ n _ a y _ u _ a _ n e d e _ e i c l _ s h _ g _ t _ i b f o s s _ _ _ t _ r
l a g _ i o o _ i m _ a s _ o t o m _ o _ _ o o e n e a </s> e^-202.814946785389

<s> > v h a _ w o n o m w _ s a _ _ t n e _ w s o _ t v g _ o w a o i </s> e^-95.871901999315|
<s> > e s e f _ r n _ e n r y l c t t _ o t e d s t e _ p e i o </s> e^-82.927094800324

<s> > w t _ u y e e s _ _ r </s> 6.95326573608822e-16

<s> i o e t c _ _ n _ n _ r n g n _ h </s> 4.116185084895e-22

<s> i o e t c _ _ n _ n _ r n g n _ h </s> 4.116185084895e-22

<s> i o a t c _ o a e _ b g _ k _ s u e l _ s t m h t n m r e o i m r n _ t _ i b p _ p p o _

s r l c n c t _ e r e i a s f t r n u a g r i _ i i _ i h t r _ r _ c _ h s </s>

e^-399.526631317511

<s> m _ _ b _ g _ u b _ y _ e f i u i t _ _ s e r r u i n b r e f a u _ l _ a s a l i l s c g r h _

u a i w m m r n s a o o d r t n e o h d o c e h y n n e s t h p e a w i g r p v o o i _ h g l t _ n

d e l t r t _ s e b h e r </s> e^-339.228875992406

<s> n m _ o o o g o i f r o n _ i e r o _ t _ n t t t t r s _ i e t b s i _ i n _ m a c w

i _ e n n c _ g e y i a _ d a t w o _ o _ _ h t _ n _ e e _ u h l e t d l i i n n o u o e

c - 112.0704022737306

<s> c w p h e o n t a i _ e _ u u s n i o s o y i e n _ b </s> e^-84.0880627735608

<s> c w p h e o n t a i _ e _ u u s n i o s o y i e n _ b </s> e^-84.0880627735608

<s> c - 140.499007051967</s>
```

bigram: has correctly spelled a few words (me, be, the)

trigram: makes more sense, appears more correct words (them, hick, man)

2.

- 1. Use 'sed -e 's/[aeiou]//g' test.txt > test.txt.novowels' to remove vowels.
- 2. Use 'carmel wfsa remove-vowels.fst > wfsa.fst' to combine fsa and fst
- 3. Use 'cat test.txt.novowels | sed -e 's/ /\_/g;s/\(.\)/\1 /g' | awk '{printf("<s> %s </s>\n",
- \$0)}' | carmel -sribIEWk 1 wfsa.fst > test.txt.vowel\_restored. {uni,bi,tri}' to restore.
- 4. Use 'cat test.txt | sed -e 's/ /\_/g;s/\(.\)/\1 /g' | awk '{printf("<s> %s </s>\n", \$0)}' > test\_formatted.txt' to format the test.txt for evaluation.
  - 5. Use 'python eval.py test\_formatted.txt test.txt.vowel\_restored.uni' to get the accuracy.

## Unigram:

```
jing:~/Dropbox/0SU/2019/Fall/cs539 - NPL/exs/ex2/ex2-data$ python
eval.py test_formatted.txt test.txt.vowel_restored.uni
0.0081351689612
```

## Bigram:

```
jing:~/Dropbox/OSU/2019/Fall/cs539 - NPL/exs/ex2/ex2-data$ python
eval.py test_formatted.txt test.txt.vowel_restored.bi
0.141426783479
```

#### Trigram:

```
jing:~/Dropbox/OSU/2019/Fall/cs539 - NPL/exs/ex2/ex2-data$ python
eval.py test_formatted.txt test.txt.vowel_restored.tri
0.435544430538
```

3.

- 1. Use 'sed -e 's/[]//g' test.txt > test.txt.nospaces' to remove spaces.
- 2. Use 'carmel wfsa remove-spaces.fst > wfsa.fst' to combine fsa and fst
- 3. Use 'cat test.txt.nospaces | sed -e 's/ /\_/g;s/\(.\)/\1 /g' | awk '{printf("<s> %s </s>\n", \$0)}' | carmel -sribIEWk 1 wfsa.fst > test.txt.space restored. {uni,bi,tri}' to restore.
- 4. Use 'cat test.txt | sed -e 's/ /\_/g;s/\(.\)/\1 /g' | awk '{printf("<s> %s </s>\n", \$0)}' > test formatted.txt' to format the test.txt for evaluation.
  - 5. Use 'python eval.py test formatted.txt test.txt.space restored.uni' to get the accuracy.

#### Unigram:

```
jing:~/Dropbox/OSU/2019/Fall/cs539 - NPL/exs/ex2/ex2-data$ python
eval.py test_formatted.txt test.txt.space_restored.uni
0.0
```

#### Bigram:

```
jing:~/Dropbox/OSU/2019/Fall/cs539 - NPL/exs/ex2/ex2-data$ python
eval.py test_formatted.txt test.txt.space_restored.bi
0.0866071428571
```

#### Trigram:

```
jing:~/Dropbox/OSU/2019/Fall/cs539 - NPL/exs/ex2/ex2-data$ python
eval.py test_formatted.txt test.txt.space_restored.tri
0.247375328084
```

#### Sentences:

#### Unigram:

```
Input line 1: <s> this is because the human mind does not readevery letter by itself but the word as a whole </s>
(150 states / 297 arcs)
<s> this is because the human mind does not readevery letter by itself but the word as a whole </s>
Derivations found for all 1 inputs
```

#### Bigram:

```
Input line 1: <s> this is because the human mind does not readevery letter by itself but the word as a whole </s>
(150 states / 297 arcs)
<s> this is _ because _ the _ human _ mind _ does _ not readevery _ le _ t _ ter _ by _ itself _ but _ the _ word _ as _ awhole </s>
Derivations found for all 1 inputs
```

#### Trigram:

```
Input line 1: <s> this is because the humanmind does notreadevery letter by itself but the word as a whole </s>
(297 states / 591 arcs)
<s> this _ is _ because _ the _ human _ mind _ does _ not _ readevery _ letter _ by _ it _ self _ but _ the _ word _ as _ a _ whole </s>
Derivations found for all 1 inputs
```

4. From the observations in these experiments, the accuracy results of restoring vowels are better than the results of restoring space. In addition, the number of states and arcs of generated vowels.fst are greater than the generated space.fst's. In other words, the restoring of vowels has more conditions to judge than the restoring of space. As a result, restoring vowels is easier.