analysis-cross-entropy

November 1, 2020

1 Lambdas found by EM (with $\epsilon = 0.0001$)

- As expected, when we run EM on train data, we get $\lambda_3 = 1$.
- This is because this data has the same sequence of words, so the probabilities given by 3-gram is the highest (or at least it is newer lower than 2-gram)

```
10
                          11
  dataset
             from
                                     12
                                                13
0
                   0.0
                         0.0
                               0.000024
                                          0.999976
            train
       CZ
2
                   0.0
                               0.000023
                                         0.999977
            train
                         0.0
```

- On the other hand, when we use the *holdout data*, the result is different. Uniform probability is not very probable, which makes sense.
- We can see that the Czech language has much more weight on p_1 . Most probably, because the Czech language has a much larger vocabulary and in the $train\ data$, there are not enough combinations to even correctly model the p_2 . In the English language, there are fewer words, so probably the p_2 gives the best probability.
- Both languages do not have λ_3 the highest. I would say that if we had much more good training data, then maybe the λ_3 would be a bit higher.

```
from
                      10
  dataset
                                 11
1
       СZ
           holdout
                     0.0
                          0.445858
                                     0.450697
                                                0.103444
3
           holdout
                     0.0
                          0.165022
                                     0.674777
                                                0.160201
```

2 Cross-Entropy on the test data

- First, let's look at the default result (with unmodified λ_3).
- As stated above, Czech 2-grams and 3-grams are probably not very accurate for a lot of words, so 1-grams have the highest weight. Thus this leads to higher entropy than in the case of English.

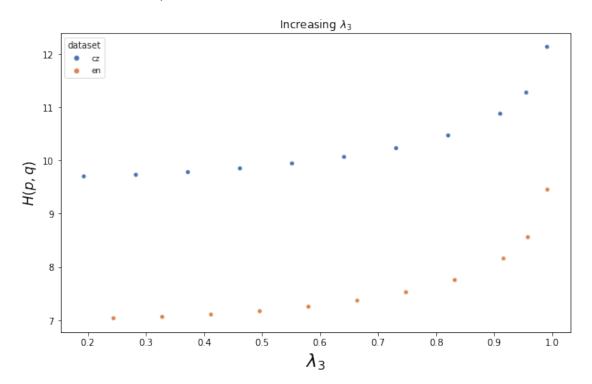
```
dataset modif modif_val L3_val cross_entropy
0 cz default 0.0 0.103444 9.696974
22 en default 0.0 0.160201 7.032398
```

2.1 Increasing and decreasing the λ_3 and and proportionally changing $\lambda_0, \lambda_1, \lambda_2$

 No we will increase and decrease the importance of the 3- gram and we will observe how the Cross-Entropy will change.

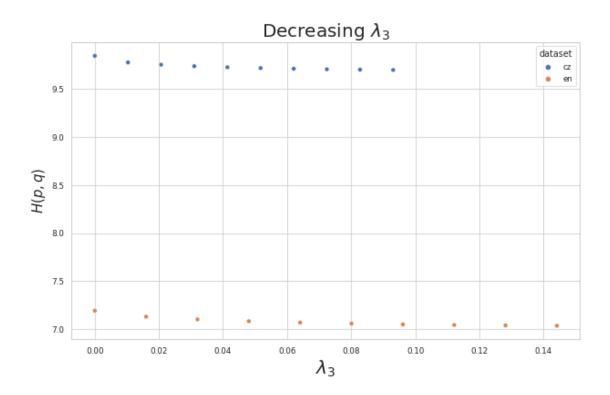
2.1.1 Increasing the λ_3

• In the figure below, we can see the results. We can see that for both languages, the increase of the λ_3 leads to higher entropy. This is probably caused because oftentimes the 3-gram does not know the answer producing 1/|V| and when most of the weight is on it, we are getting higher entropy. (But if we would perform the experiment on the *train data*, then the entropy would decrease to 0 ...)



2.1.2 Decreasing the λ_3

- On the figure below, we can see the resuts. We can se that for both languages, decreasement of the λ_3 also leads to higher entropy.
- This means, that the value assigned to the 3-gram is probably quite good. We can see, that both icreasement and decreasement of the λ_3 leads to incresement of cross entropy on the *test data*.
- This would suggest that the EM (using the *holdout data*) found quite optimal values of λ_i and it produces lowest entropy on *test data* also (at least w.r.t. icreasement or decreasement of λ_3)



2.1.3 Nicely plot of both decreasement and increasement together

• I Smoothed the results and plot it together to see the results better

