COG-DUO

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Overview

- An app which can intelligently recommend places to explore, based on current (or any other) location.
- Real-time knowledge of customer's context (current location, preferences, etc.)
- Customer segmentation by analyzing usage patterns

tripadvisor

Understanding the problems

- Improve an industry's (or company's in specific) outreach and quality of service (QoS).
- Limited opportunities to satisfy customers on a person-to-person level.
- Hospitality based businesses like hotels and eateries depend on customer satisfaction.
- Understanding customer becoming a more important field with advancement in ML techniques.

Goals

01 Personalized suggestions, offers, and alerts

02 Recommendation engines

03 Predicting customer ROI and lifetime value

04 Predictive inventory planning

Business Model:

Reviews + place_id

NLU

Sentiment scores and grouping.

Ranking algorithm and rating

Sort Out the negative comments and get the burning issues

Tools being used

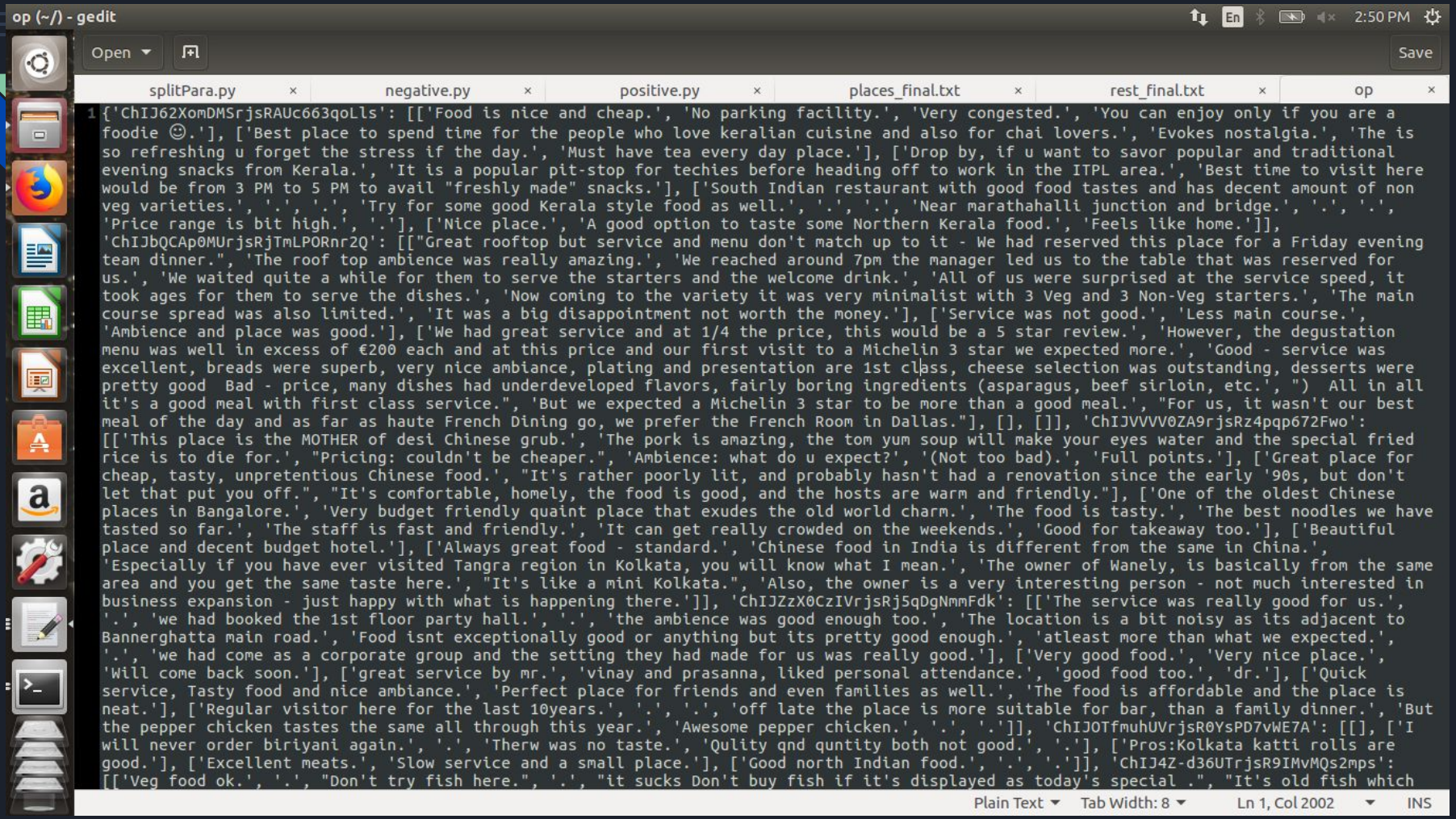
- APIs from IBM Analytics:
  A. Natural Language Classifier(NLC) (to train a model based on the classes and keywords) (future application)
  B. Natural language understanding (NLU) for analyzing the sentiment of the review.

- APIs from other vendors:
  A. Google Places API
  - used to get all the places visited by the customer
  - duration, place and mode of stay to learn about the customer's specific behaviour.

Getting the data

- The data we require is user reviews. We used Google Places API to get the user reviews .

- Reviews are taken for a specific place. This place is part of a list of places which were captured by issuing a request which gives a set of places (of interest and or restaurants for now).
  a
- Then we scrapped out the required fields which were the place id and the user reviews of that place. There are many other fields like userid and photos of the places etc., which can be put in order to enhance the UI.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negative.py
positive.py
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rest_final.txt
op
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'ChIJbQCAp0MUrjsRjTmLPORnr2Q': [["Great rooftop but service and menu don't match up to it - We had reserved this place for a Friday evening team dinner.", 'The roof top ambience was really amazing.', 'We reached around 7pm the manager led us to the table that was reserved for us.', 'We waited quite a while for them to serve the starters and the welcome drink.', 'All of us were surprised at the service speed, it took ages for them to serve the dishes.', 'Now coming to the variety it was very minimalist with 3 Veg and 3 Non-Veg starters.', 'The main course spread was also limited.', 'It was a big disappointment not worth the money.'], ['Service was not good.', 'Less main course.', 'Ambience and place was good.'], ['We had great service and at 1/4 the price, this would be a 5 star review.', 'However, the degustation menu was well in excess of €200 each and at this price and our first visit to a Michelin 3 star we expected more.', 'Good - service was excellent, breads were superb, very nice ambiance, plating and presentation are 1st class, cheese selection was outstanding, desserts were pretty good Bad - price, many dishes had underdeveloped flavors, fairly boring ingredients (asparagus, beef sirloin, etc.', ") All in all it's a good meal with first class service.", 'But we expected a Michelin 3 star to be more than a good meal.', "For us, it wasn't our best meal of the day and as far as haute French Dining go, we prefer the French Room in Dallas."], [], []],
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'ChIJZzX0CzIVrjsRj5qDgNmmFdk': [['The service was really good for us.', '.', 'we had booked the 1st floor party hall.', '.', 'the ambience was good enough too.', 'The location is a bit noisy as its adjacent to Bannerghatta main road.', 'Food isnt exceptionally good or anything but its pretty good enough.', 'atleast more than what we expected.', '.', 'we had come as a corporate group and the setting they had made for us was really good.'], ['Very good food.', 'Very nice place.', 'Will come back soon.'], ['great service by mr.', 'vinay and prasanna, liked personal attendance.', 'good food too.', 'dr.'], ['Quick service, Tasty food and nice ambiance.', 'Perfect place for friends and even families as well.', 'The food is affordable and the place is neat.'], ['Regular visitor here for the last 10years.', '.', '.', 'off late the place is more suitable for bar, than a family dinner.', 'But the pepper chicken tastes the same all through this year.', 'Awesome pepper chicken.', '.', '.']],
'ChIJOTfmuhUVrjsR0YsPD7vWE7A': [[], ['I will never order biriyani again.', '.', 'Therw was no taste.', 'Qulity qnd quntity both not good.', '.'], ['Pros:Kolkata katti rolls are good.'], ['Excellent meats.', 'Slow service and a small place.'], ['Good north Indian food.', '.', '.']],
'ChIJ4Z-d36UTrjsR9IMvMQs2mps': [['Veg food ok.', '.', "Don't try fish here.", '.', "it sucks Don't buy fish if it's displayed as today's special .", "It's old fish which
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Approach

- Pass the reviews to NLU and get the sentiment and its score.
- Then for each user review , create a positive , negative score by categorizing the keywords and their scores into positive , negative and neutral .
- Then we rated the particular place based on the place_id , by using our rating technique where we combine the aggregates of positive and negative scores of a review and get the final score .
- Sort the final scores and give them their ranks.
- Then finally we can recommend customers based on our approach which is enriched with NLP brain of Watson.

Terminal

Open Save

ibm.js zomato.py ibm.py

```
var NaturalLanguageUnderstandingV1 = require('watson-developer-cloud/natural-language-understanding/v1.js');
var natural_language_understanding = new NaturalLanguageUnderstandingV1({'username': 'a79238a0-5e8d-4b5b-8101-825adeff273e','password':
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var parameters = {
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      'sentiment': true,
      'limit': 2
    },
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      'emotion': true,
      'sentiment': true,
      'limit': 2
    }
  }
}

natural_language_understanding.analyze(parameters, function(err, response) {
  if (err)
    console.log('error:', err);
  else
    console.log(JSON.stringify(response, null, 2));
});
```

sameer@sameer-Inspiron-3543: ~

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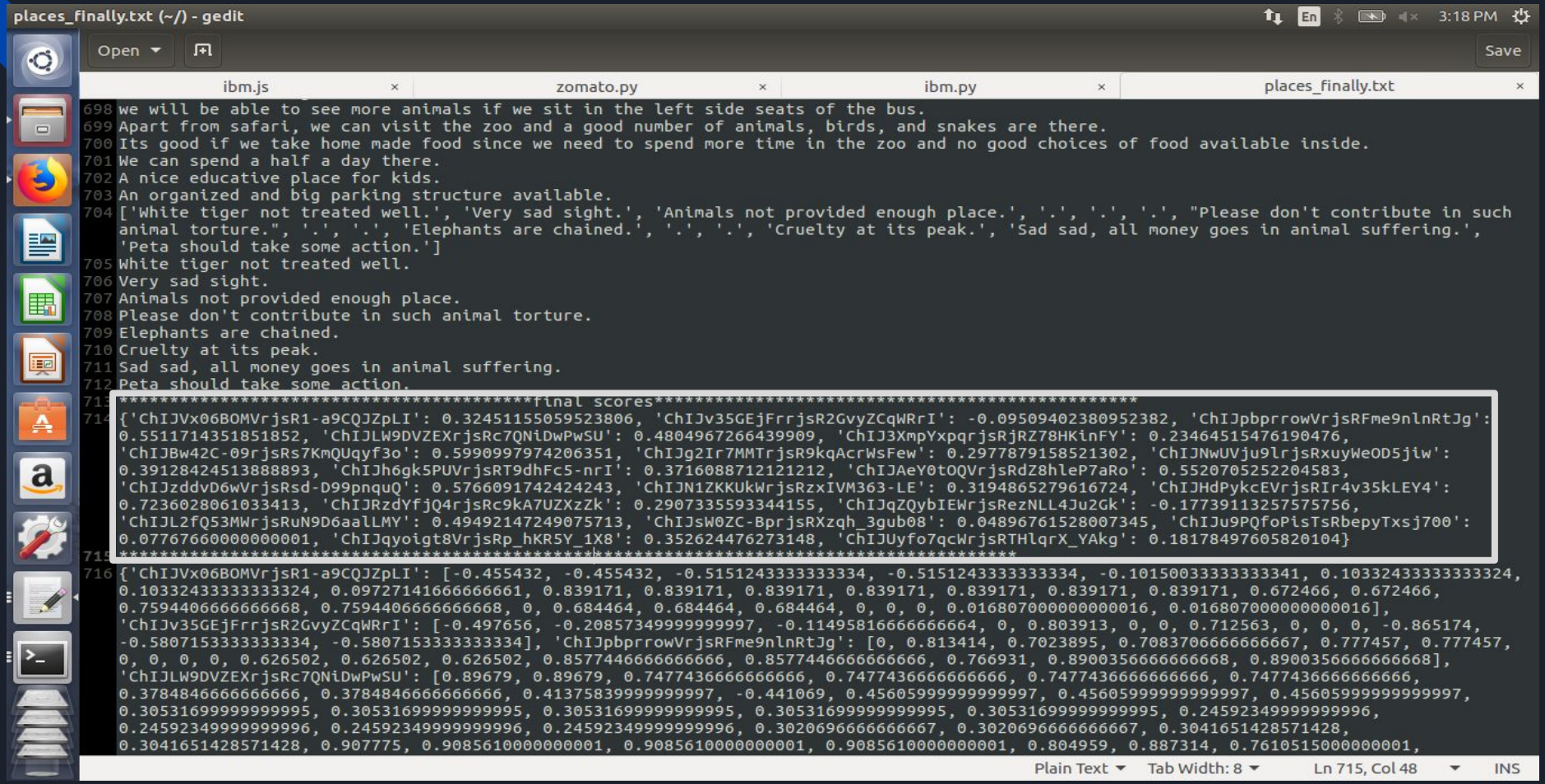
Final Scores of Places



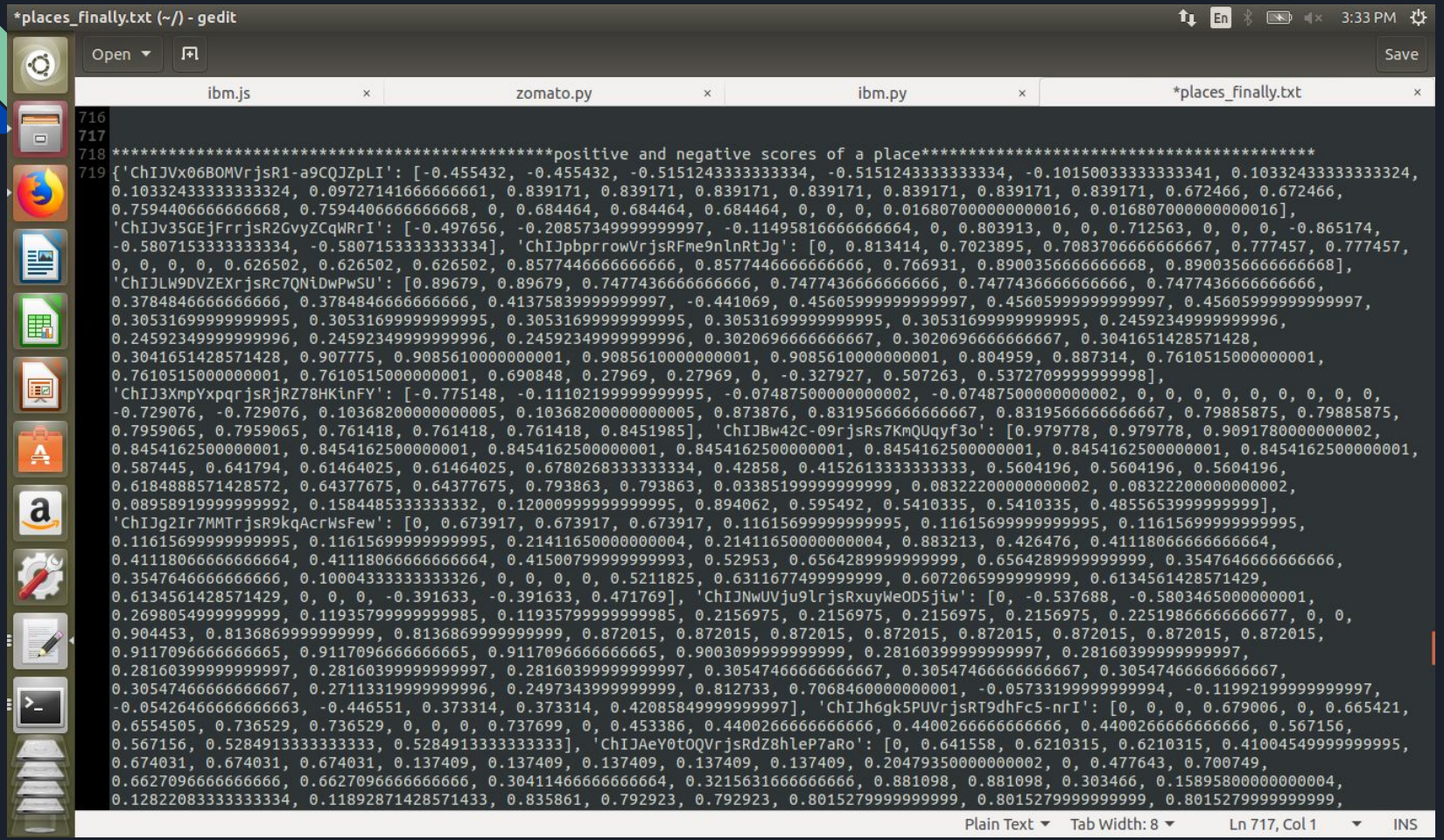
Approaching Burning Issues:

- After ranking , our aim is to let the authority know about the negative issues which people are facing at their place.

- So , as our approach creates a list of negative keywords along with their scores , we will take all the negative keywords from all the users and see which of them are more prevalent among all of them.

- Since we have the respective places_ids mapped to the negative keywords , we can submit the top issues to the organization and let them improvise on these issues.

Look + vs -


*places_finally.txt (~/) - gedit
3:33 PM
Open
Save
ibm.js
zomato.py
ibm.py
*places_finally.txt
716
717
718 **********positive and negative scores of a place**********
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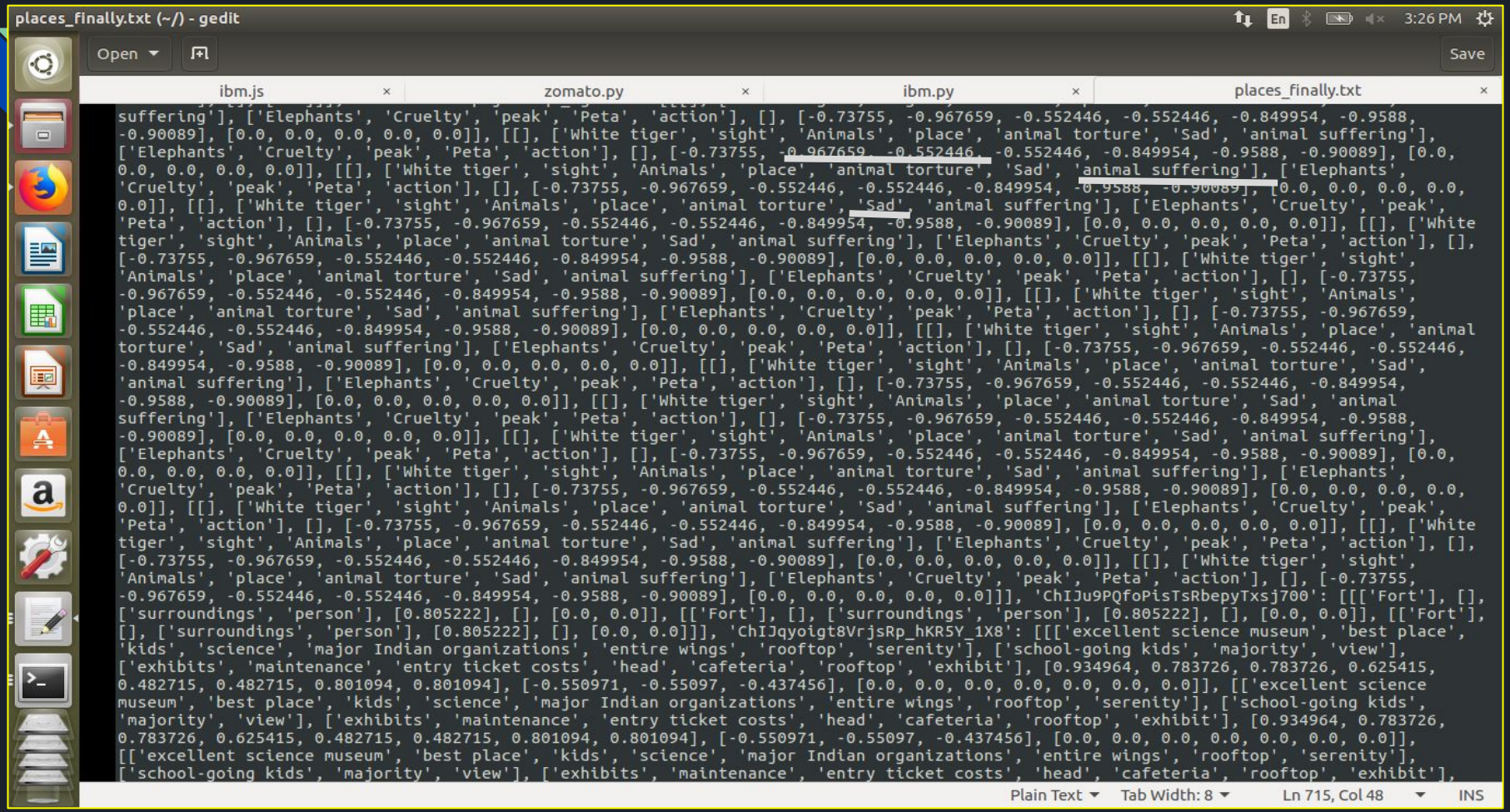
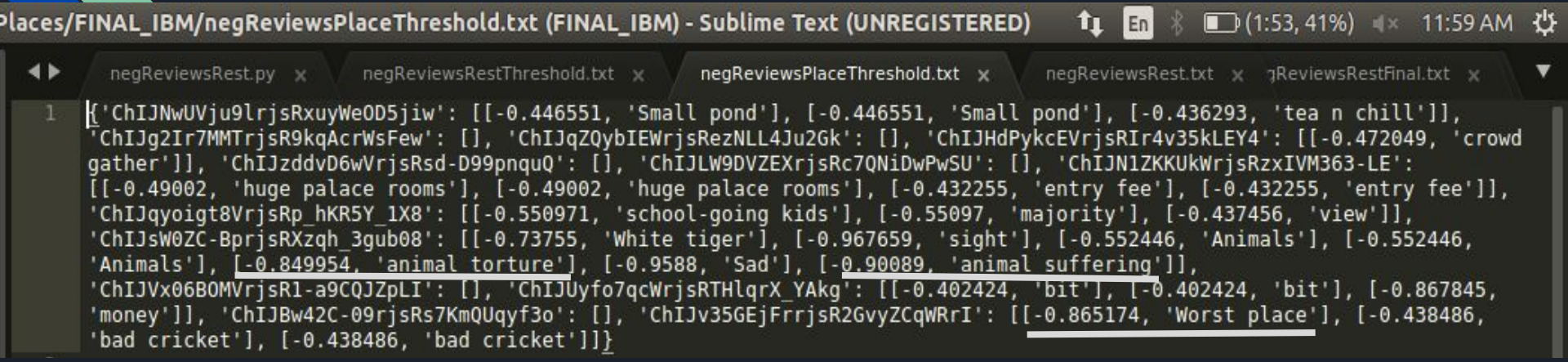
Negative Issues



Which Issues to choose???



Future Enhancements

- Extend it for transportation , so that it can become something like a “***DEEP - Travel Agent***” .
- Put NLC and take the user reviews and categorize accordingly to a scoring formula, much like a ranking algorithm.
- Placing it on to chat agents like ***Cortona*** , ***Siri*** , ***Google now*** for ease of use. (chat bot with watson would be helpful for easy integration with NLU).

Thank you.