

## **MOM - 28 Feb 2020**

Problem statement -

**\*\*which recipe belongs to which region? (Based on Ingredients)**

**\*\*literature review based on context vector concept at least (8 papers in text analytics)**

**\*\* If possible find the datasets**

Share chosen papers by Monday 2/3/2020 EOD

1 paper should concentrate on theoretical aspect and 1 on technical

Summarize these papers.

Nishant to share the excel sheet

Next Meet - Important Stuffs - 5th March 12:00

Abstract, Introduction, Experimentation, Evaluation steps, We can add results after 9th March

## **MoM 5<sup>th</sup> March**

Ideas Discussion - Need to improvise so guys try to break it from all possible points

We have dish A

It has ingredients [1,2,3]

Recipe A has 1000 reviews

So we do text analytics on those 1000 reviews and generate binary label for that dish A

Positive

Negative

So now we have

Data row like

1. Dish A

$[1,2,3] = 1$

Assume 1 positive

So we have n dishes represented by context vectors of ingredients and their labels

So now when we get a new dish B

And it's ingredients

$[1,3,4]$  we can predict if these

Dish or combination of ingredients would be liked or disliked by the customers/ppl

## **MoM 7<sup>th</sup> March 2020**

Research questions

A list of research questions we would like to address during the project:

What are the most commonly used ingredients?

What are the most distinctive ingredients?

How can we represent cuisines as a network of ingredients ?

How can we cluster cuisine in terms of their recipe components?

How do cuisines influence each other?

## **MoM 15 March 2020**

### Scope of the project -

Can we predict the contents on the quizzine based on recipe/integradients context vectors?

Lujain - Continue to build taxonomy

Aakash - Word2Vec check for the cosine distances (Wikipedia Taxonomy)

Nishant - Data cleaning

Oommen – Understand Word2Vec and Glove Methodology

Rohan - To create Context vectors

## **MoM 25 March –**

**Covid** –Continue working on the same.

Nishant Plus Rohan – Check the feasibility of the code.

## **MoM 2<sup>nd</sup> April 2020**

Lujain --→ Nishant/Rohan --→ Oommen/Aakash

Lujain to generate data...Send data to Nishant to clean and rohan to build vectors, Oommen/Aakash to Generate Charts and visualizations

Next Meet – 6 April 5pm

## **MoM 6<sup>th</sup> April 2020**

Drop an email - Modification in the problem statement(discard taxonomy)

Compare results based on Word2Vec and GloVe - Nishant and Rohan (Implementation and Evaluation)

Look into the Visualization Task (Recipe Vectors) - Oommen

Find existing paper on Bag of Words. Explain in the paper as well

Modification in Literature Survey (Lujain and Aakash)

- (Try to answer 15 Ques in the list)
- Modification in Abstract, Introduction, keywords and Formatting
- Previous work on frequency based models, word2vec, etc, etc

Pseudo Code - Word2Vec, GloVe and Our Algorithm

### **MoM 11 April –**

Lujain, Aakash - Continue Building the paper

Nishant, Rohan - Test Context Vectors

Oommen – Continue building cluster visualization

Next Meet 13 April 13, 2020

### **MoM 2<sup>nd</sup> April 2020**

Proofreading and Final Actions

Submit Articles with solving below errors -

1. Change in title!

From: Classifying Recipes into Cuisines Using Context Vectors

To: Classifying Recipes into Cuisines Using Context Vectors of Ingredients

2. Abstract first line the word "like" -> "such as"

3. Line no 5 Abstract: of cuisines using the context vectors -> of ingredients

4. In abstract which baseline method: traditional frequentist approach using bag of words

5. Intro second para last line: a small corpus per entry

6. Intro third para " We will be following that with our results, analyse them along with any errors. "

7. e large corpus, like Wikipedia section 2.1.1

8. 2.2 related work no space

oss the world have attempted to predict a dish's cuisine based on a recipe's ingredients as

9. related work : xgboost mai dala but what was put words or count vectors

10. Sectoi 3.2

. The prediction accuracy for this method was found to be 78%. (classification)

### 11. 3.3 Cleaning Up The Data

we first remove

12. by adding should we mention

13. 3.4

Not split randomly