# Food Review & Analysis

Guide
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Team Members

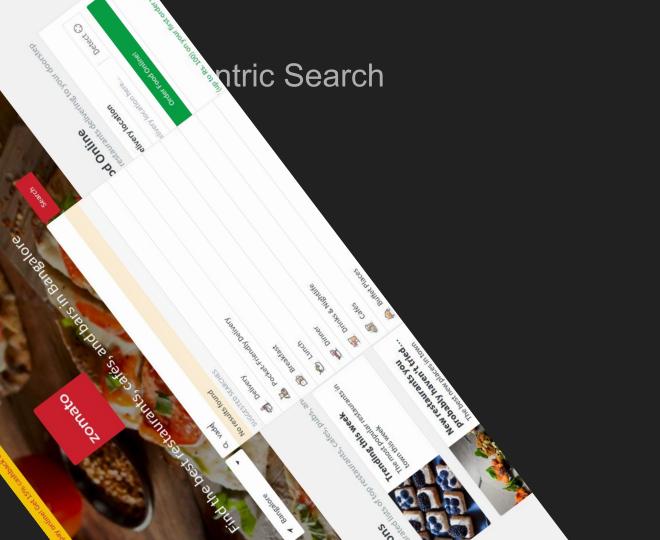
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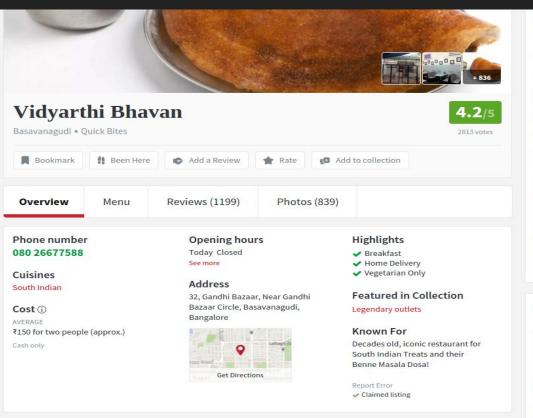
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### The IDEA...

Two things missing in today's Food Searching Platforms:



### No mention of specialities of restaurant





Empire Restaurant JAYANAGAR Birvani, Kerala



CALL NOW!





JAYANAGAR North Indian. Chinese



AB's - Absolute Barbecues JAYANAGAR





Cable Car JAVANAGAR North Indian, Italian



Reetle Juice Bar - Springs Hotel & BASAVANAGUDI Finger Food



Addhuri Udupi Bhojana





Timepass Dinner BASAVANAGUDI North Indian



Aagaman Veg JAYANAGAR Buffets starting

#### Cafes

Coffee, snacks & beverages



Unique Brew Sports Cafe JAYANAGAR Fun, Food and



Leanin' Tree Art Cafe BASAVANAGUDI Cafe, Continental



Costa Coffee JAYANAGAR Cafe

### Problem Statement

I. Tagging of food menu items by user reviews for restaurants, to identify best/worst dishes of a place as mentioned by users and also to identify best places serving a particular dish as per user experience.

### DataSet

- I. Restaurants 17
- II. 40 reviews per restuarants
- III. A total of 715 menu items
- \*All humanely annotated

# Restaurants we thought of ...

Vidyarthi Bhavan <4.2> Tim Tai <4.2>

Taaza Thindi <4.9> Mamagoto <4.3>

Indian Coffee house <3.6> Kapoor's cafe <4>

Shri Sagar <4.7> Punjab grill <4.2>

Brahmin's Coffee bar <4.9> Kopper kadai <4.1>

Chetty's Coffee <3.3> Kesariya <3.9>

Sri laxmi venkatesh coffee bar <4.4> Dal tadkaa <2.5>

Milano ice cream <4.9> Belgvum <4.7>

filano ice cream <4.9> Belgyum <4.7>

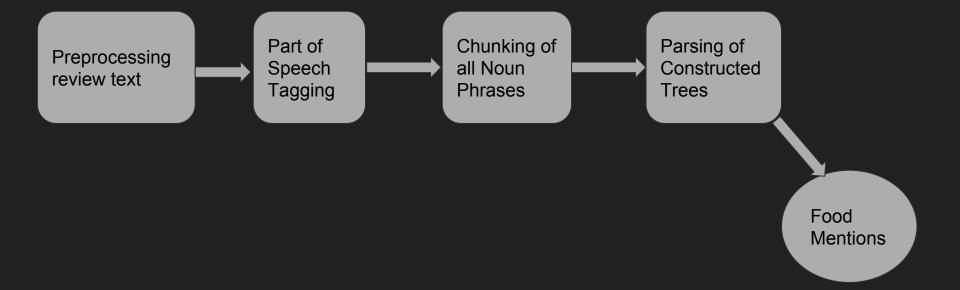
Mr. Idli <2.6>.

Demo ...

#### Model / Pipeline Menu Items For Each Restaurant & Price User Custom Reviews NER Food Mentions Tagging Analysis in Reviews Box

### Custom Named Entity Recognizer

> Not learnt (due to unavailability of dataset), but rule based



### Tagging Box

- ExactMatch ('Noodles','Noodles')
- II. SubstringMatch ('Chef's Special Bhaji Pao', 'Bhaji Pao')
- III. PartialMatch ('Fetter Cheese Pasta', 'Cheese Paste')
- IV. FuzzyMatch ('Masale Dose', 'Masala Dosa')
  - V. PercentMatch ('Red Pasta', 'Past')

### DataStructure

**ALIAS** 

RestName: Id

**DISH SEARCH** 

MenuItemName:[(RestId, RestRating, Price, MenuItemRating, MenuItemPopularity)]

**REVIEWS** 

Restld:

[(Reviews,Rating])]

**RESTAURANT SEARCH** 

RestId:(Rest\_Rating,{MenuItem:(Price,Rating,Popularity)})

**MENU** 

RestId: (Rest\_rating,[MenuItem, Price])

**MENTION** 

RestId:{Mention:[Popularity, Rating]}

### Hurdles {Data : The Missing Piece}

- Zomato API erroneous
- II. Image to Text Cluttered
- III. Zomato US Menu Empty Calls
- IV. Yelp Dataset (has latitude & longitude of restaurants, not menus)
- V. Yelp API does not exposes Menus, obtained from Loci
- VI. Third Party Indian Menu APIs hoaxes
- VII. Last Resort Manual Creation

### Visualization Inferences

- 1. Graphical Representation helps decipher hidden correlations among entities.
- We modelled some visualizations to infer certain obscure correlations.

#### **Bubble Plot**

- 1. Represents 3-dimensions
- Average Price, Average Review Length & Avg Rating
- 3. Modelled for individual restaurant

#### Inferences:

- Smaller bubbles have lower ratings, bigger have larger.
- Heavy Price doesn't mean review is good.
- When people write more, means they loved the place :)

### Scatter Plot [Price vs Review Length]

#### Inferences:

• When customer pay more, they don't talk much about it.

# Bar Charts [Restaurants]

Visual comparison of different restaurants.

## Bar Graph - Review Length vs Average Rating

X-axis - Average review length

Y-axis - Rating associated with the review.

Inferences -

- Unremarkably higher number of more expressive reviews for highly rated dishes.
- 2. Also interesting is the many reviews with high number of words for dishes rated as low as 1.5 (possibly expressing the discontent).

### Donut - Rating vs Number of Reviews

Serves to highlight the percentage of the number of reviews for a particular rating.

Inferences -

1. A general trend observed is the high number of reviews for dishes that are rated highly by the users. (And consequently the less number of reviews for dishes with low ratings)

### Scatter plot - Price vs Rating

X-axis - Price of a menu item

Y-axis - Rating of a menu item

Inferences -

- 1. An upward trend among the ratings as the price of the menu items increases.
- However reasonably high ratings are observed for a number of lower prices menu items as well.

### Scatter plot - Price vs Popularity

X-axis - Price of a menu item

Y-axis - Popularity(total number of reviews) of a menu item

#### Inferences

- 1. The number of reviews for most of the dishes fall within a specific range of 10-15 reviews. Not very strongly dependent on the price of the dish.
- 2. Very high number of reviews seen for a few dishes is likely due to the generally high demand/popularity of the dish.

### Scope

Machine Learning Components we skipped because of data unavailability:

I. NER

II. SVM to identify mentions of food mentions

III. Tagging Box Implementation via SVM with algorithm outputs as features.

PS - Both II & III need human annotation of sizable data as features for predictions.

# Thank you