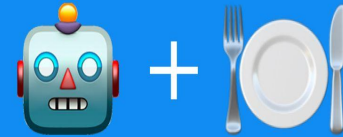


# Food Advisor Chatbot

Joe, Obinna, Michael

# Why we choose the topic

- Intimately connect with **business scenario**: meet customer's needs;
- **High Efficiency**: Using chatbot instead of service employee
- **NLP development**: Can understand customer more clearly



Chatbot for Restaurant

chatfuel

# Data Source and EDA

## Conclusion:

1. chatbot can deliver 3 parameters for recommendation value, can only for partly fitting customers' requirements
2. recommendation mainly focus on rank, and can get
3. can return address, mail to chatbot, and some restrictions

```
] df.to_csv('cleaned_tripadvisor.csv')
```

```
] df['primary_cus'].value_counts()
```

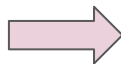
```
] Unknown      37972
   Japanese    7903
   French       6995
   Italian      5609
   Bar          3110
   ...
   Beijing cuisine    1
   Balti              1
   Fruit parlours     1
   Southwestern       1
   Burmese            1
Name: primary_cus, Length: 120, dtype: int64
```

**Data Source: Kaggle**, link

<https://www.kaggle.com/mikhailpustovalov/scraped-data-from-ta>

## EDA

1. Delete inactive, closed and name\_changed restaurants
2. Delete city with null value
3. rating, primary\_cus, city fields are useful for recommendation
4. About 50% cus\_rank is null, can't help to recommend
5. some restaurants have the same rank(mainly in 22500-25000)
6. ....



Got Data  
Characteristic

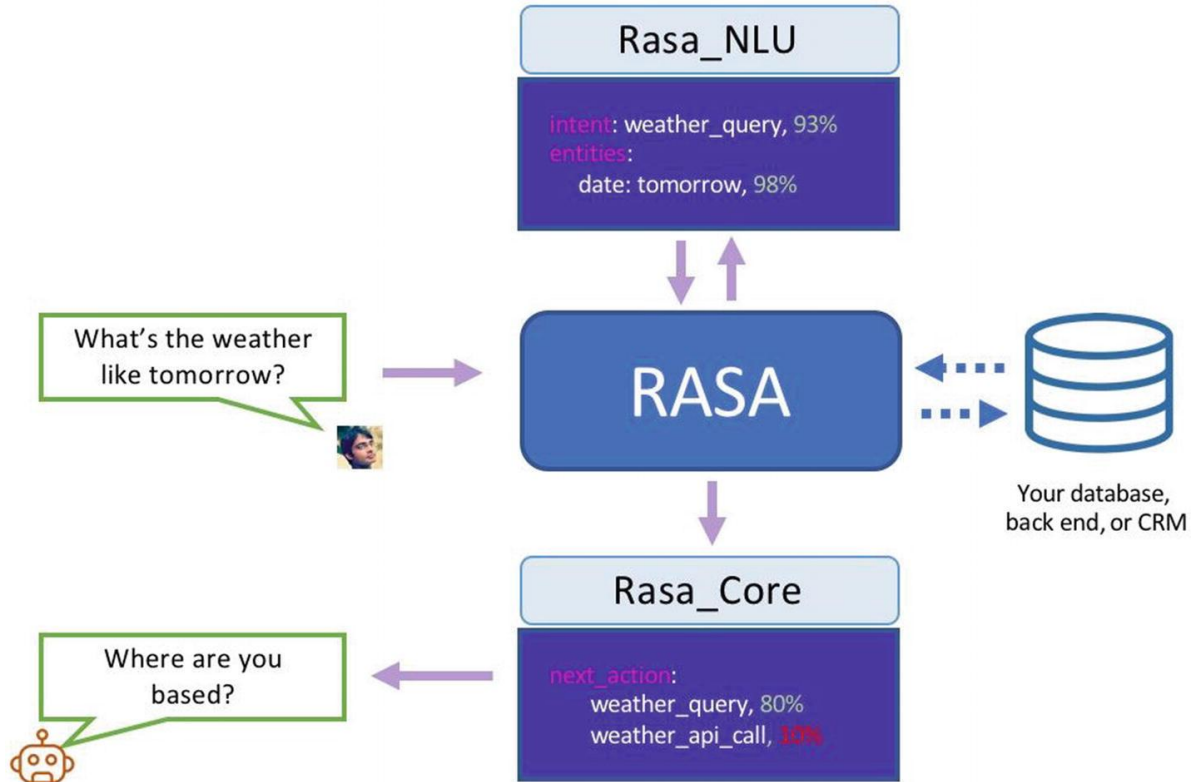


Cleaned Data

## Main Functions of the Chatbot

- **Search information** about restaurants and recipes.
- **Recommend** restaurants and recipes according customer's requirements and historical transactions.
- **Book** restaurants
- **Rate or commend** after consumption

# The main development tools: RASA



# RASA NLU

Intents: 33

Entities: 24

Stories: increase as training continue

Rules: 12

```
intents:  
- affirm  
- answer_date_time  
- answer_city  
- greet  
- answer_num_people  
- answer_name  
- search_restaurant_information  
- goodbye  
- deny  
- mood_great  
- mood_unhappy  
- bot_challenge  
- search_for_food  
- answer_calories  
- answer_carbohydrate  
- answer_cholesterol  
- answer_sugar  
- answer_protein  
- answer_recipe_category  
- reject_food_choices  
- accept_food_choices  
- ask_for_food_recommendation  
- answer_form  
- complain_about_something  
- complaint_target  
- complaint_name  
- complaint_aspect  
- recommend_restaurant  
- book_restaurant  
- cancel_booking  
- rate_restaurant
```

# RASA CORE and appended Database

Slots: 36

Customised Actions: 11

Database:

1. Transaction.csv
2. User\_record.csv
3. user\_complain\_record.csv

```
actions:  
- action_book_restaurant  
- action_cancel_booking  
- action_get_history_record  
- action_provide_other_food_choices  
- action_rating  
- action_recommend_food  
- action_recommend_restaurant  
- action_record_complaint  
- action_search_restaurant  
- action_select_food  
- action_suggest_food
```

# The interaction between Customised Actions and Database

## Customised Actions

1. Slots value get and assign
2. Direct Response
3. Condition Judgement
4. Recommendation according

### Domain Knowledge

- a. Distance Recommendation
- b. Random Recommendation
- c. Conditional Recommendation
- d. ....

## Actions and Database

1. Search Information
2. Store Information



# OUR FUTURE PLAN



1. More close to business scenario and logic
2. More Training to improve accuracy
3. Deploying with multiple communication tools