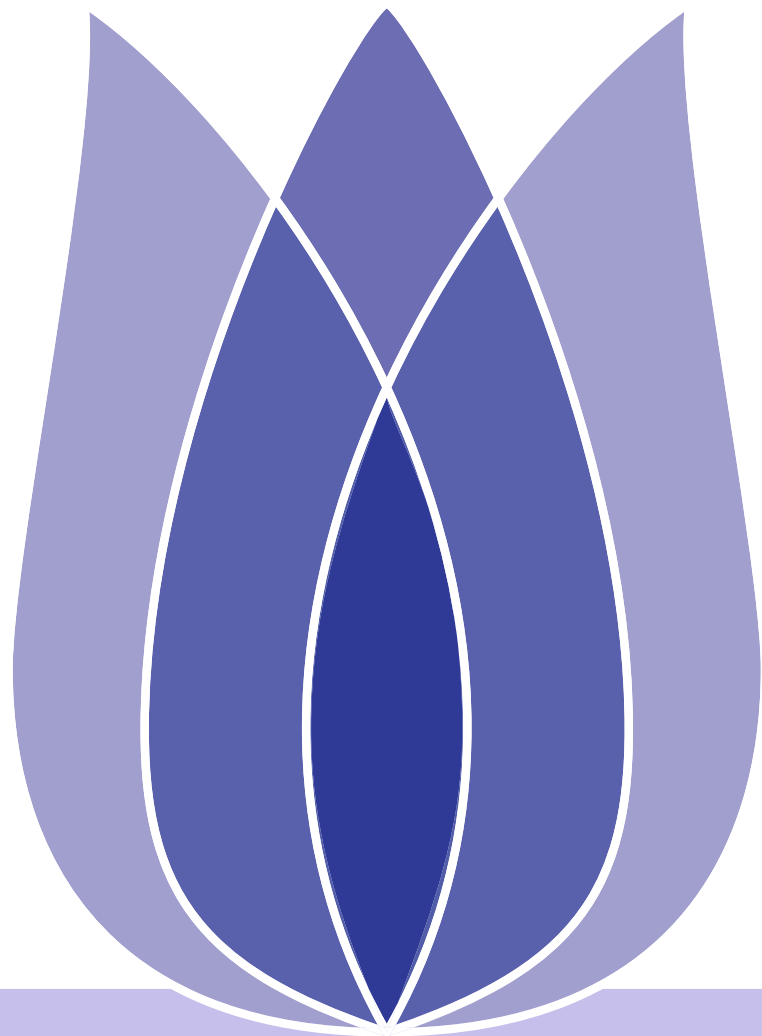




# FLIP01 FINAL PRESENTATION

Zhaoyang Wang  
Xi'an Shiyou University

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## Problem Statement

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# Problem Statement



# Problem Definition

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Some of our strongest geographic and cultural associations are tied to a region’s local foods. This playground competitions asks you to predict the category of a dish’s cuisine given a list of its ingredients.



# Data Set

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## ■ train data

Table 1: The head of the train data

	cuisine	id	ingredients
0	greek	10259	[romaine lettuce, black olives, grape tomatoes...
1	southern_us	25693	[plain flour, ground pepper, salt, tomatoes, g...
2	filipino	20130	[eggs, pepper, salt, mayonaise, cooking oil, g...
3	indian	22213	[water, vegetable oil, wheat, salt]
4	indian	13162	[black pepper, shallots, cornflour, cayenne pe...

## ■ Display the data set

Table 2: The head of the test data

	id	ingredients
0	18009	[baking powder, eggs, all-purpose flour, raisi...
1	28583	[sugar, egg yolks, corn starch, cream of tarta...
2	41580	[sausage links, fennel bulb, fronds, olive oil...
3	29752	[meat cuts, file powder, smoked sausage, okra,...
4	35687	[ground black pepper, salt, sausage casings, l...



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# Text Preprocessing



# Preprocessing

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- stopwords
- regularization
- convert to lowercase letters

Since my text data is relatively clean, I only used the stopwords method



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By using wordcloud to discribe the frequency of text data.

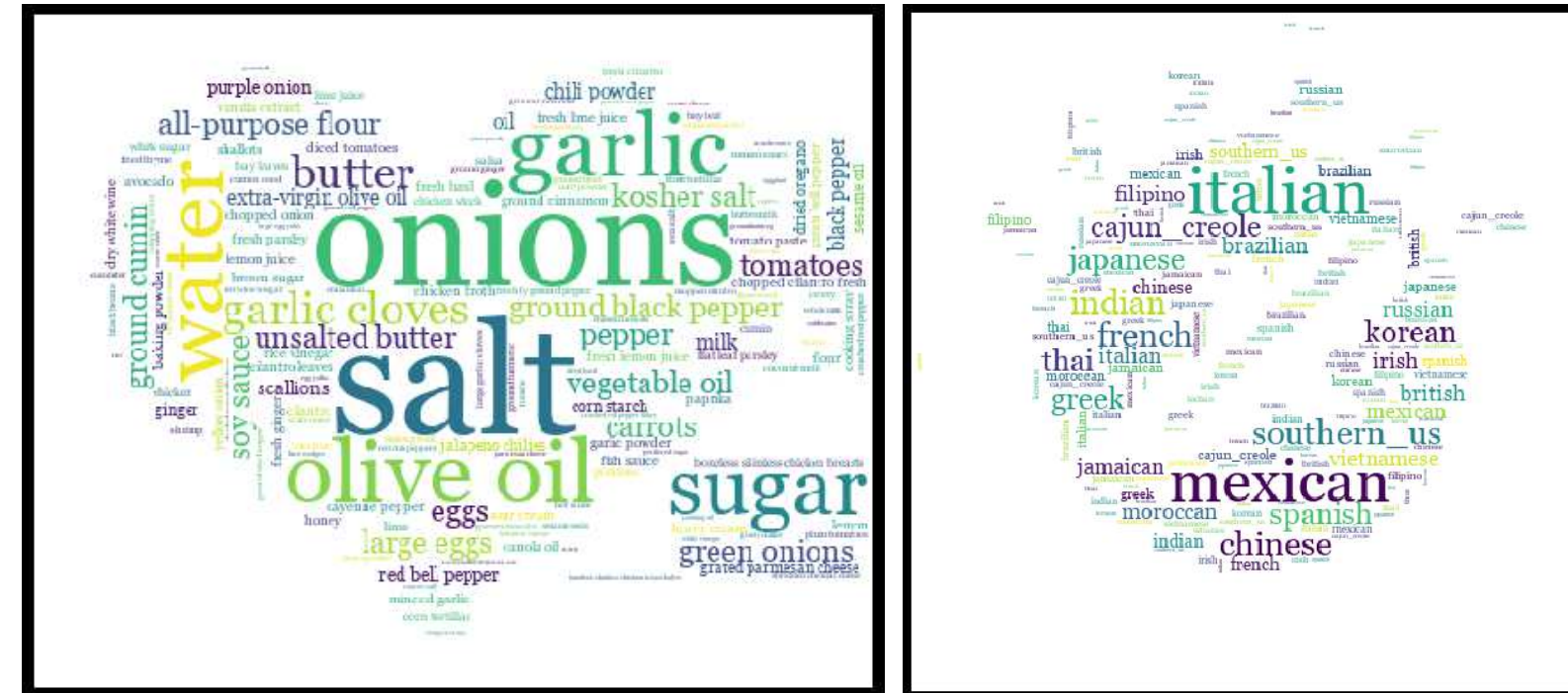


Figure 1: Displaying the words in text



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# Text feature extraction



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■  
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Replace text labels in the data set with numbers. Transform text in feature values into word vectors.

- unique() and apply()
- word2vec



# Replace text labels

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By using the unique() and apply() can replace the texe label into figures.

Table 3: Replace the text label

	cuisine	label
0	irish	16
1	italian	6
2	irish	16
3	chinese	8
4	mexican	7



# word2vec

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Use word2vec to convert text to word vectors. And convert word vectors to sentence vector.and then for each sentence vector we have one label for it.

- vector size 300
- mean



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# Modeling



# Modeling

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- problem analysis

This is a text classification problem.we can use many ways to solve text classification problems.

- Logistic Regression
- KNN
- Random forest
- SVM
- CNN





# Step

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- Divide training data and test data
- Do a model training
- Model evaluation
- Model prediction





# The score of models

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Table 4: The score of models		
	model	score
1	Logistic Regression	0.729
2	KNN	0.740
3	Random forest	0.739
4	SVM	0.736
5	CNN	0.753



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# Conclusion



# Conclusion

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- 1** Using the Word2vec to help us process the textdata.If the text data is Chinese, we can use jieba for word segmentation.
- 2** There are many ways to deal with text classification in machine learning .we can select suitable ways on combination with the problem.
- 3** In this problem, i use the mean of each words vector to caculate the sentence vector. Maybe this is the question why accuracy is lower than my espect





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