

# Identifying Key Entities in Recipe Data - Report

## Business Objective

The source dataset comprises culinary recipes with a focus on ingredient extraction and analysis. Each recipe features a structured ingredient list with labelled components, identifying ingredients, quantities and units. This diverse collection supports tasks such as understanding recipes and discovering culinary knowledge, enabling the development of models for information extraction in the culinary domain.

## Input Data

- The input data comprises a Json file with two columns input & pos with 285 rows
  - Column input comprises all the ingredients used for the Recipe & the Column pos comprises the corresponding POS tag.

```
# print the information of the dataframe  
inputDF.info()
```

```
<class 'pandas.core.frame.DataFrame'>  
RangeIndex: 285 entries, 0 to 284  
Data columns (total 2 columns):  
 #   Column Non-Null Count Dtype  
 ---  -----  -----  
 0   input    285 non-null   object  
 1   pos      285 non-null   object  
dtypes: object(2)  
memory usage: 4.6+ KB
```

- There were 81 records for which the pos\_tokens length was not equal to input\_text length. So, they were dropped.

## Data Analysis

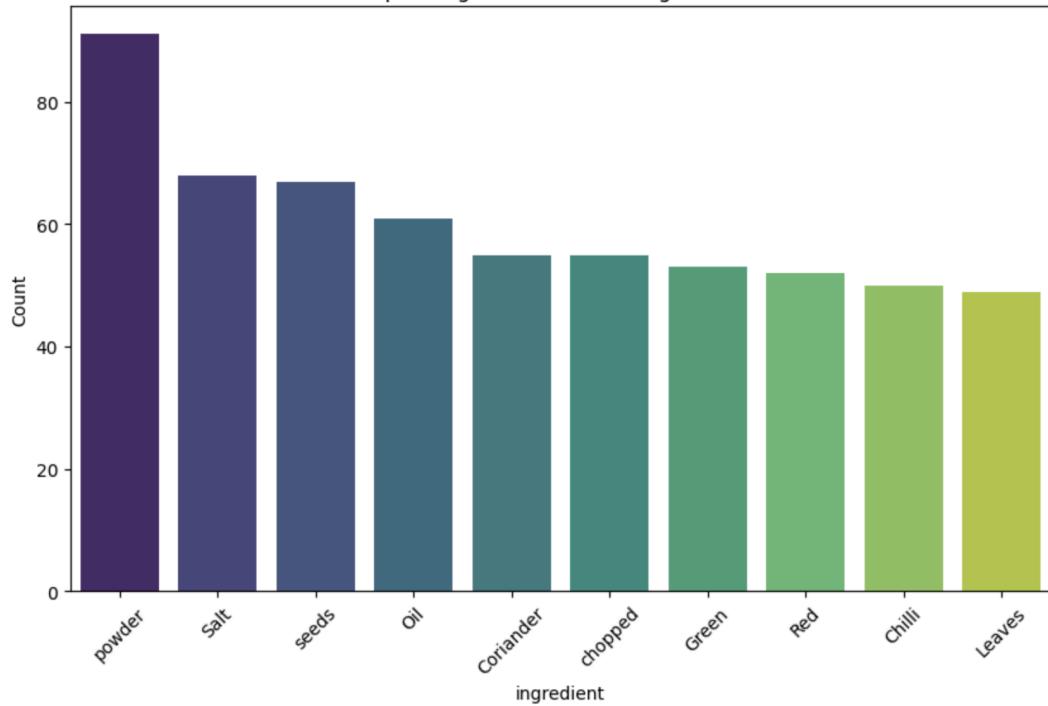
---

```
First 10 records in Training dataset:  
Input Token --> POS Token  
1 --> quantity  
cup --> unit  
Gram --> ingredient  
flour --> ingredient  
besan --> ingredient  
1/2 --> quantity  
teaspoon --> unit  
Ajwain --> ingredient  
Carom --> ingredient  
seeds --> ingredient
```

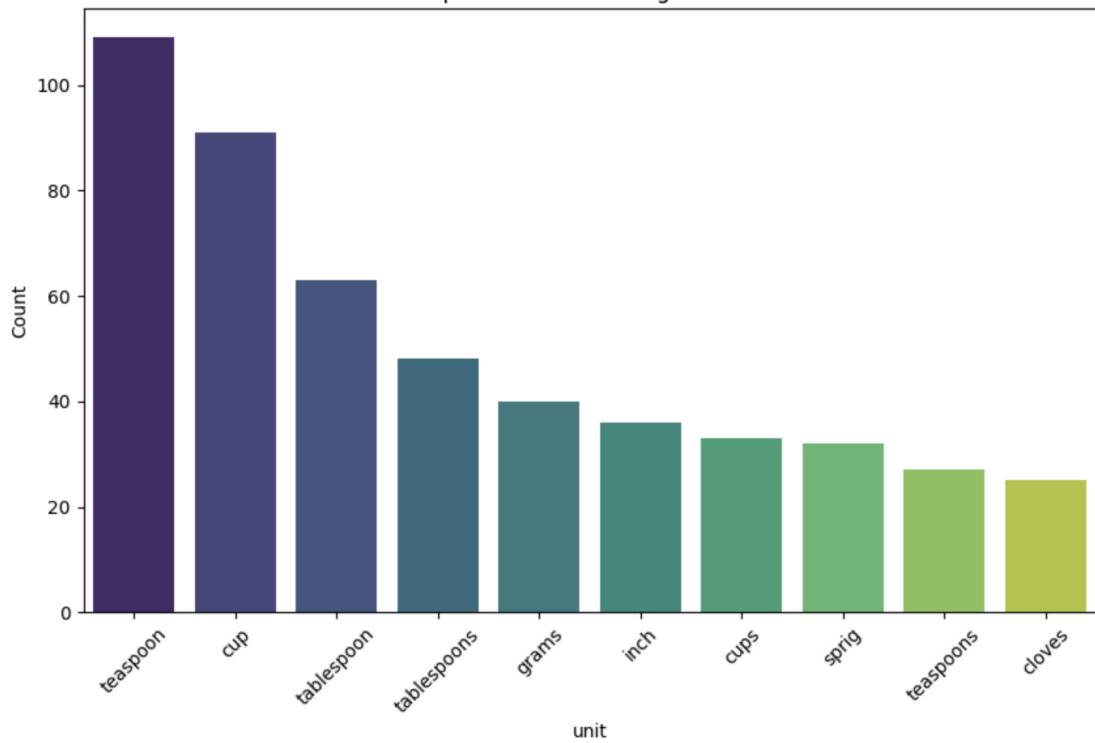
---

```
First 10 records in Validation dataset:  
Input Token --> POS Token  
1 --> quantity  
cup --> unit  
rice --> ingredient  
12 --> quantity  
small --> unit  
onions --> ingredient  
2 --> quantity  
cloves --> ingredient  
garlic --> ingredient  
inch --> unit
```

Top 10 ingredients in Training dataset



Top 10 units in Training dataset



## Observations

- Powder is the top most ingredient in the entire dataset
- Along with powder, Salt & Seeds are also mostly commonly used
- Teaspoon, cup & tablespoon are the most units

## Feature Engineering

- We identified common Unit and Quantity keywords by analyzing frequent terms.
- We defined regular expressions to match different forms of Units
- We also converted words to features to analyse the impact

## Core Features

- Token properties
- Character patterns

## Quantity & Unit Detection

- Regular Expressions for Quantities
- Numeric pattern recognition for Units

## Contextual Features

- Boundary Markers
- Previous & Next words/tokens

## Model Training

CRF model is used to train the recipe data. Below are the parameters initialized:

```
# initialise CRF model with the specified hyperparameters and use weight_dict
crfModel = sklearn_crfsuite.CRF(
    algorithm='lbgfgs',
    c1=0.5,
    c2=1.0,
    max_iterations=100,
    all_possible_transitions=True
)
# train the CRF model with the weighted training data
crfModel.fit(XTrainWeightedFeatures, yTrainLabels)
```

```
CRF
CRF(algorithm='lbgfgs', all_possible_transitions=True, c1=0.5, c2=1.0,
max_iterations=100)
```

## Model Evaluation

### Training Data Results

	precision	recall	f1-score	support
ingredient	1.00	1.00	1.00	3411
quantity	1.00	1.00	1.00	658
unit	0.99	1.00	1.00	563
accuracy			1.00	4632
macro avg	1.00	1.00	1.00	4632
weighted avg	1.00	1.00	1.00	4632

**Confusion Matrix:**

```
Labels ['unit', 'ingredient', 'quantity']
[[ 562    0    1]
 [  0 3411    0]
 [  3    0 655]]
```

Validation Data Results

```
# specify flat classification report
yValTrueFlat = [label for sent in yValLabels for label in sent]
yValPredFlat = [label for sent in yValPred for label in sent]
print(classification_report(yValTrueFlat, yValPredFlat))
```

	precision	recall	f1-score	support
ingredient	1.00	1.00	1.00	1611
quantity	1.00	0.99	1.00	294
unit	0.99	1.00	1.00	244
accuracy			1.00	2149
macro avg	1.00	1.00	1.00	2149
weighted avg	1.00	1.00	1.00	2149

**Confusion Matrix:**

```
Labels ['unit', 'ingredient', 'quantity']
[[ 244    0    0]
 [  0 1611    0]
 [  2    0 292]]
```

Validation Accuracy is very good - **0.9906**

This indicates that the Model has been trained very well.

## Error Analysis

```
Label: quantity, ErrorCount: 2, Class Weight: 7.0395
```

```
Error DataFrame:
```

	token	true_label	predicted_label	prev_token	next_token	class_weight
0	little	quantity		unit	leaves	Salt
1	taste	quantity		unit	per	1/2

## Conclusion

- Model shows strong performance in recognising recipe entities.
- Model Accuracy is very high on both Training & Test data sets
- Recommend this model to be tested on other real world practical data.