

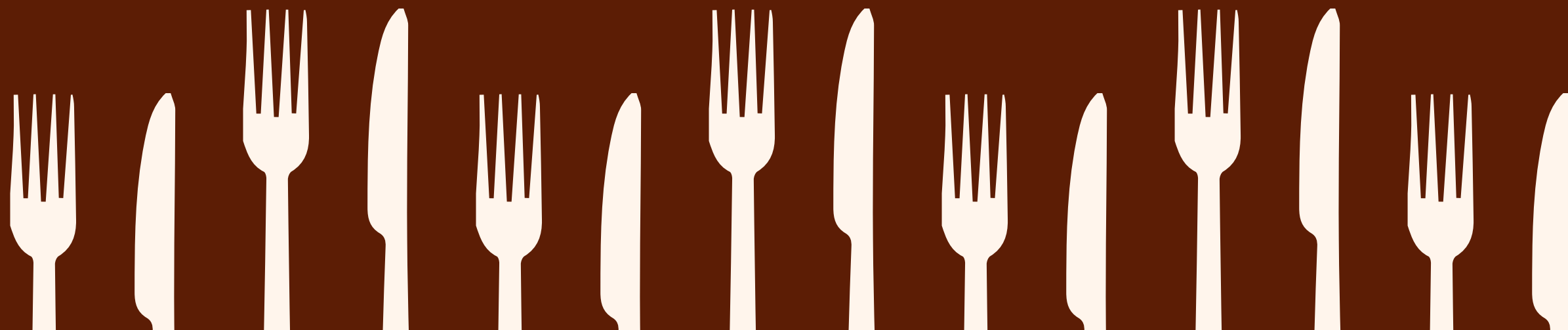
BANA 277 CUSTOMER & SOCIAL ANALYTICS PROJECT

# ANALYZING RECIPE SUCCESS

A Deep Dive into User Ratings

TEAM TRACE3B:

Vy Nguyen, Kento Morita, Qui Nguyen, Aria Zhou  
Sahil Chennadi, Yuheshwar Kamakkapalayam Subramani



# Introduction & Data Overview

**Purpose** Analyze recipe data on [Kaggle](#) to gain insights that can help recipe creators and platforms optimize recipes for higher ratings.

○ **What recipe attributes influence ratings?**

RAW\_recipes.csv

- 231,637 rows, 12 columns
- Unique ids: user & recipe

○ **How do user sentiment impact ratings?**

RAW\_interactions.csv

- 1,158,039 rows, 5 columns
- Unique id: user & recipe

| Sentiment | Description              |
|-----------|--------------------------|
| rating    | User rating from 0 to 5. |
| review    | User review in text.     |

| Attribute     | Description   |
|---------------|---|
| name          | The name of the recipe.   |
| minutes       | Minutes required to prepare the recipe.   |
| tags          | Food.com tags associated with the recipe (e.g. 'course')                                  |
| nutrition     | calories (#), total fat (PDV), sugar (PDV) , sodium (PDV) , protein (PDV) , saturated fat |
| n_steps       | Number of steps required to prepare the recipe.   |
| steps         | The text description of each step in the recipe.  |
| description   | User-provided description for the recipe.   |
| ingredients   | A list of ingredients required for the recipe.  |
| n_ingredients | The total number of ingredients in the recipe.  |

# Recipe Attributes Analysis **PROCESS**

## 1. WRANGLING

- Outliers (z-score)
- Missing values

## 2. DATA TRANSFORMATION

- Submitted Date: Recalculate years since recipe publication (from 2019).
- Tag: Count the number of tags per recipe.
- Nutrition: Sum up the total nutrition values per recipe.
- Ingredients: Count the number of ingredients per recipe.

## 4. ML MODELING

- $Y \Rightarrow$  rating (0-5)
- Linear Regression, GLM, Neural Network, Random Forest

## 3. DATA PREPROCESSING

- Standardization
- Oversampling

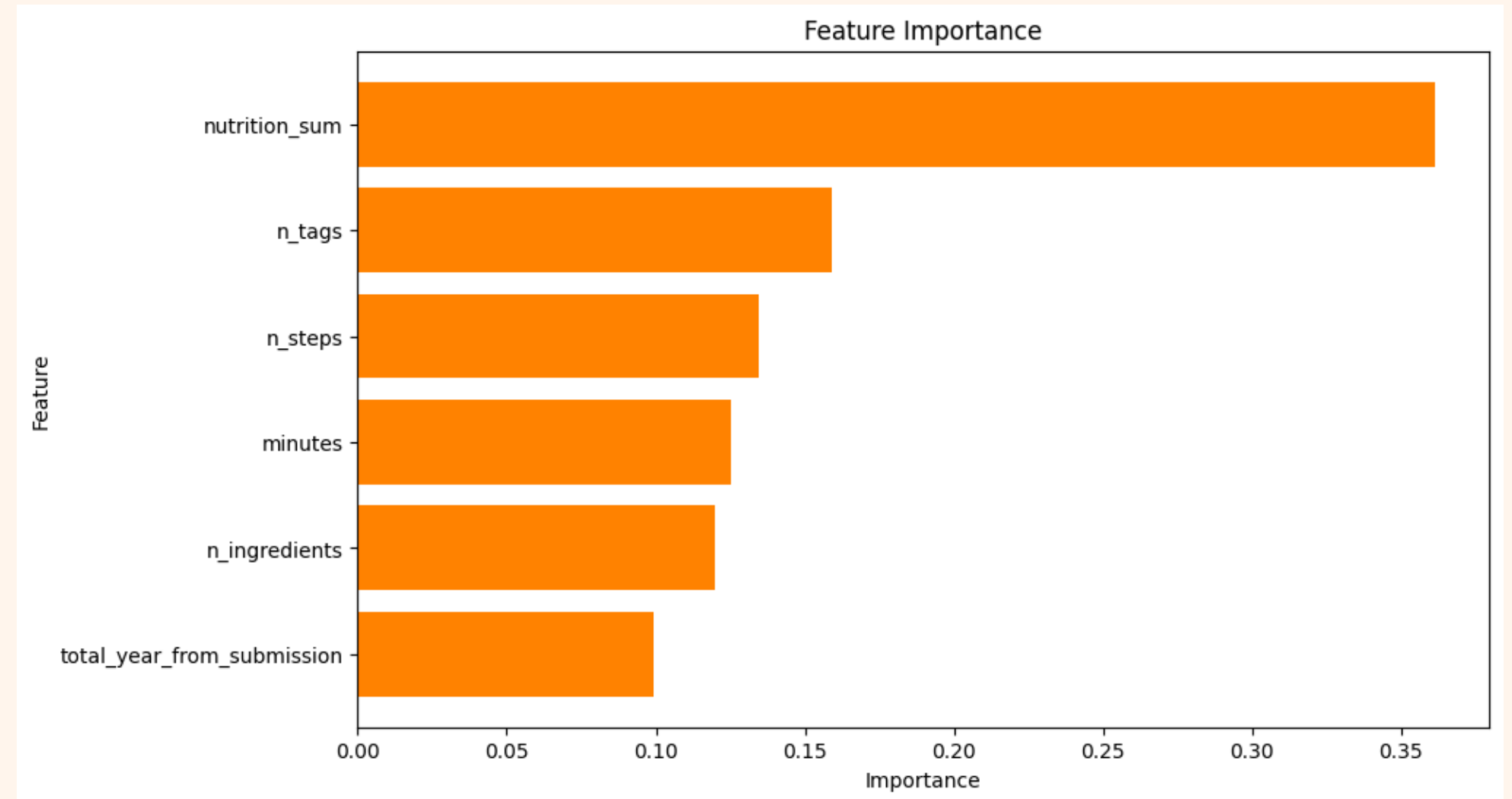
# Recipe Attributes Analysis

## KEY FINDINGS

| Models                   | MSE     |
|--------------------------|---------|
| Linear Regression        | 127.941 |
| Generalized Linear Model | 0.980   |
| Neural Network           | 0.977   |
| Random Forest            | 0.145   |

- Best model: Random Forest
- Stronger nonlinear relationship

Data was probably easy  
BUT we are talented!



- Most important factors for rating: nutrition and number of tags and steps
- User loves easy, healthy food=> create nutrition focus tags

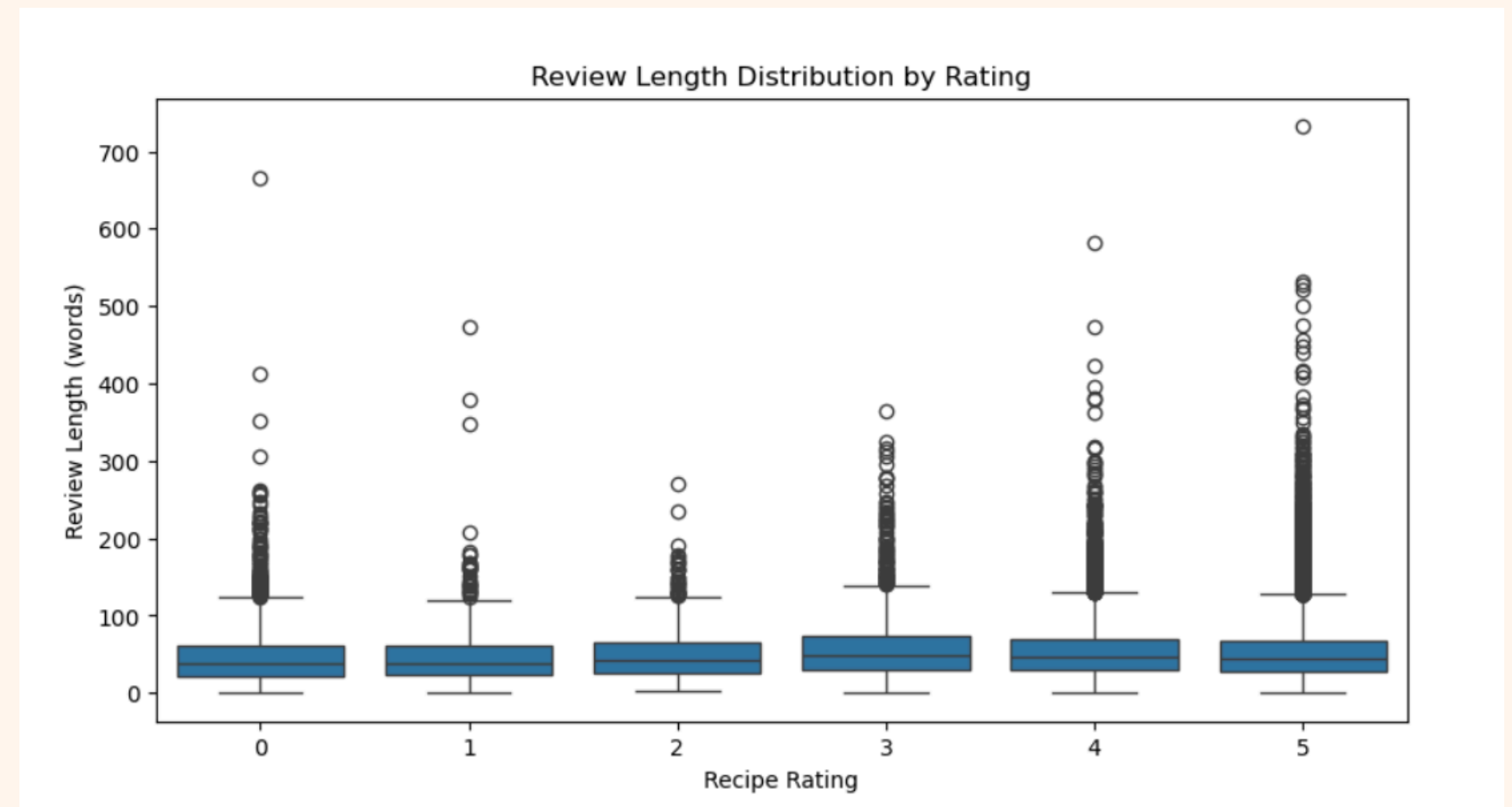
# Review Sentiment Analysis MODELS

## Random Forest Classification Report Across Ratings 1-5

|                   |      |
|-------------------|------|
| Total Accuracy    | 0.93 |
| Macro Average     | 0.96 |
| Weighted Average  | 0.93 |
| Precision Average | 0.96 |

- The model is good at correctly predicting ratings with minimal misclassifications
- Best-performing attributes consistently correlate with positive sentiment and high-rated reviews.

## Review Length Distribution by Rating

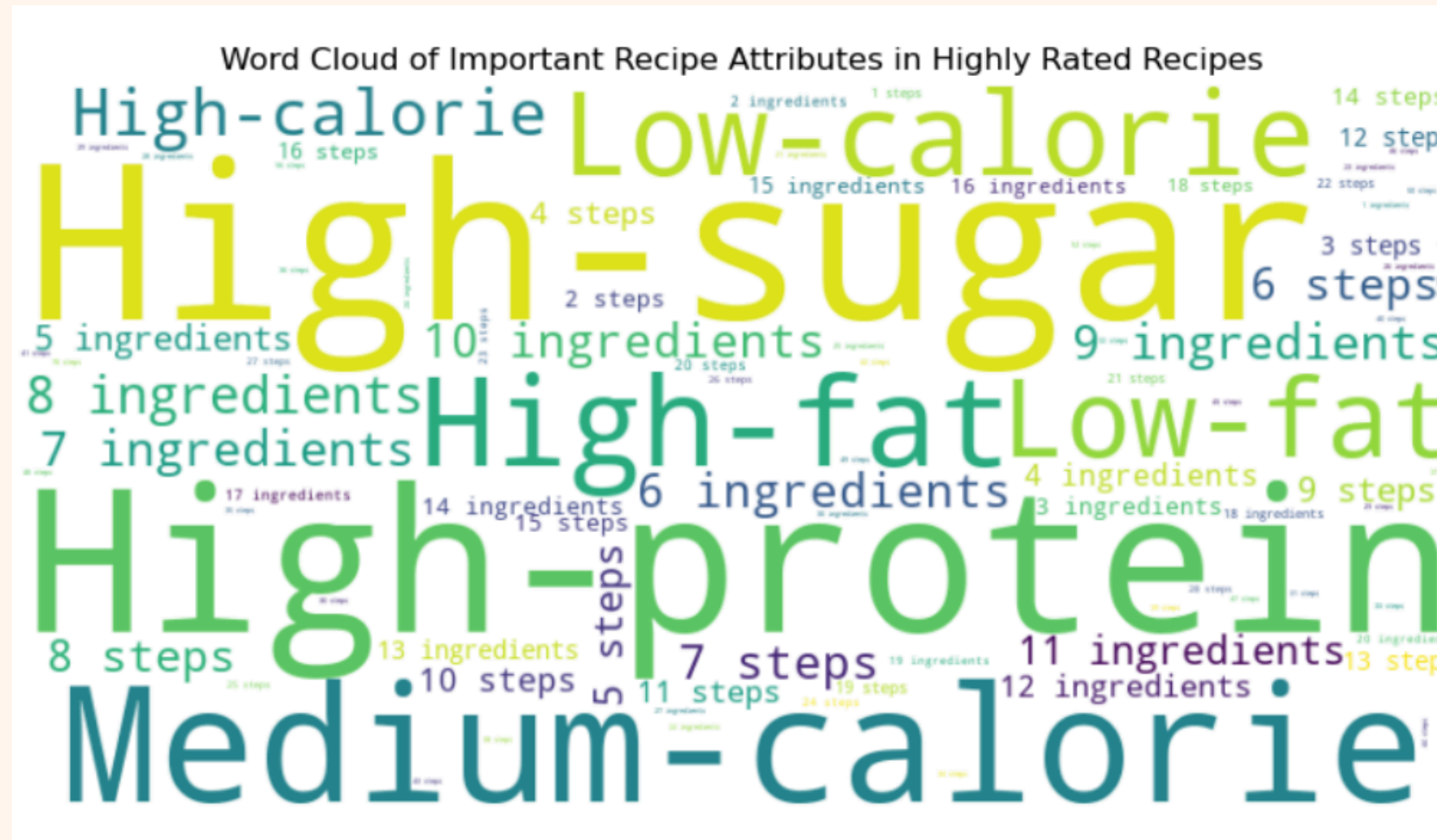


- This analysis highlights the importance of encouraging user feedback to better understand their preferences
- Ratings in the 4-5 range have more outliers => Longer reviews may be associated with higher ratings



# Review Sentiment Analysis

## KEY FINDINGS

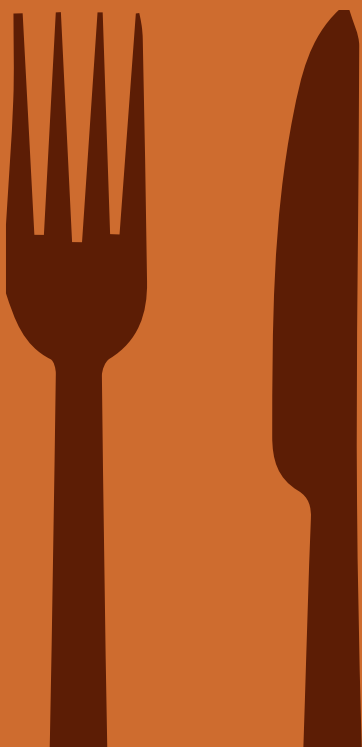


- High-protein recipes are commonly found among highly rated dishes
- Moderate ingredient and step counts are common among highly rated recipes



- "recipe," "made," "used," "added," "thank," "easy," "good," "great," "delicious," "flavor"
- Taste and ease of preparation are the biggest factors influencing positive feedback.

# Recommendations



## Recipe Creators and Platforms should:

### Focus on Taste and Ease

- Positive language: "delicious," "easy," and "flavorful" in description
- 7-9 ingredients and 5-8 steps of preparation

### Leverage Nutritional Appeal

- Highlight moderate caloric and fat content
- Avoid overemphasizing sodium or sugar levels

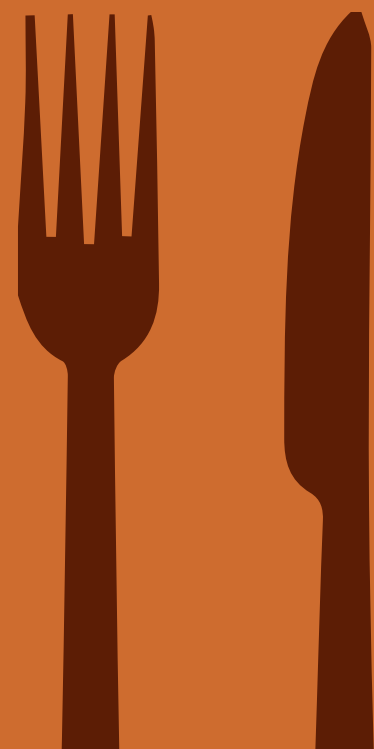
### Encourage Detailed Reviews

- Prompt users to leave detailed feedback, as longer reviews often indicate stronger opinions
- Identify and address common pain points in low-rated recipes

### Optimize Recipe Presentation

- Use tags effectively to categorize recipes (e.g. "low-carb", "30-minute meal")
- Ensure recipe steps are clear and easy to follow

# Conclusion & Next Steps



**Taste, ease of preparation, and user experience drive high ratings more than recipe complexity or nutritional content**



**Scaling our insights, further analysis can be conducted:**

## **Incorporate User Demographics**

- How age, location, or dietary preferences influence ratings?

## **Integrate Advanced NLP Techniques**

- Extract deeper insights from user reviews

## **Conduct A/B Testing**

- Test different recipe formats (e.g, varying ingredient counts, step complexity) to validate findings.