

Recipe for Success

Optimizing meals under dietary constraints

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Introduction

Finding a healthy recipe online that aligns with your personal preferences and dietary restrictions can be a daunting task. Using data from allrecipes.com, we constructed machine learning models that map recipes (as captured by their constituent ingredients) to success as measured by online ratings. Using vector representations of our ingredients, we develop a methodology for detecting logical ingredient substitutions.

Dataset and Features



Lasagna: 259

Brownies: 383

Cookies: 4703

"Banana Nut Brownies"



Rating: 4.39
Reviews: 1322
Servings: 20

Ingredients: [2 cups white sugar, 1 cup butter, 1 1/2 cups all-purpose flour, ... 1 ripe banana, mashed]

Servings: 20

["2 cups white sugar",
"1 cup butter",
"1 1/2 cups all-purpose flour",
...
"1 ripe banana, mashed"]

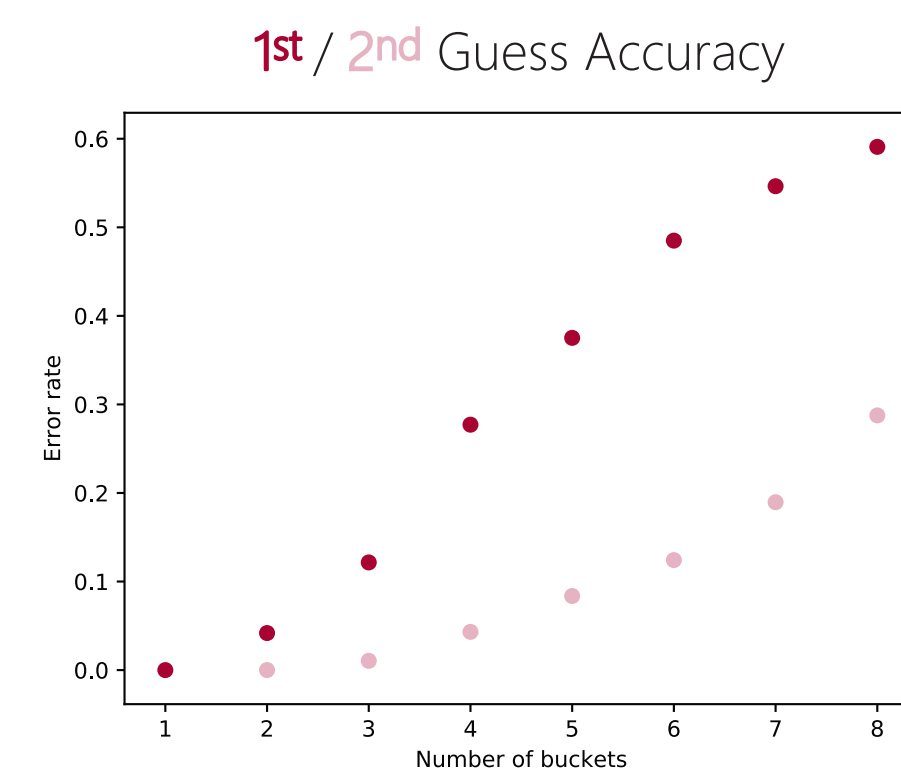
[[0.67, "sugar"],
[0.33, "butter"],
[0.5, "flour"],
...
[0.83, "banana"]]

Recipe Rating Prediction

Naive Bayes

Model: We apply the Naive Bayes multinomial model with Laplace smoothing. A flexible bucketing schema discretizes the continuous ranges of ratings from one to five stars.

Data and Features: We filter the recipes to include only ingredients that appear in at least 10 meals. Recipes are randomly partitioned into training (80%) and test (20%) sets.



Best Cookie Ingredients

- "cream cheese" (3.39)
- "peanut butter" (3.47)
- "semi-sweet chocolate" (3.51)
- "marshmallow" (3.54)
- "peanut butter cup" (3.76)

Worst Cookie Ingredients

- "coconut sugar" (1.92)
- "peppermint extract" (1.94)
- "anise oil" (1.95)
- "orange extract" (1.95)
- "almond milk" (1.95)

Sample Generated 5-star Recipe

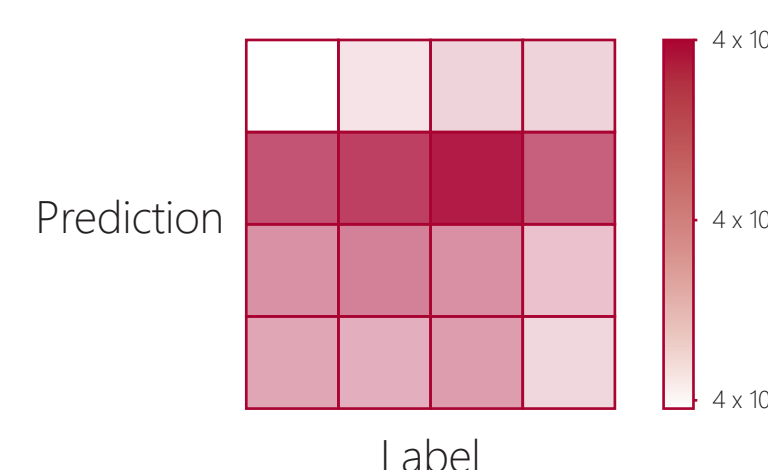
- | | | |
|------------------------------|------------------------------|-------------------------------|
| 2 tbsp vegetable oil | 1/2 cup butter | 2 tbsp pumpkin |
| 2 tbsp condensed milk | 1 floz vanilla extract | 2 tbsp confectioners' sugar |
| 2 tbsp marshmallow | 3/4 cups of white sugar | 2 tbsp water |
| 2 tbsp brown sugar | 2 tbsp almond | 3/4 cups semi-sweet chocolate |
| 2 tbsp walnut | 4 tbsp of chocolate cake mix | 3 tbsp of peanut butter cup |
| 2 tbsp shortening | 2 tbsp cocoa powder | 3 eggs |
| 1 1/3 cups all-purpose flour | 2 tbsp raisin | |

Rating Neural Network

Model: We implemented two neural networks—one for classification, and one for regression. Both use one hidden layer and sigmoid activation.

Data and Features: Again, we filter the recipes to include to only track ingredients that appear in at least 10 recipes. Recipes are randomly partitioned into training (70%) and test (30%) sets.

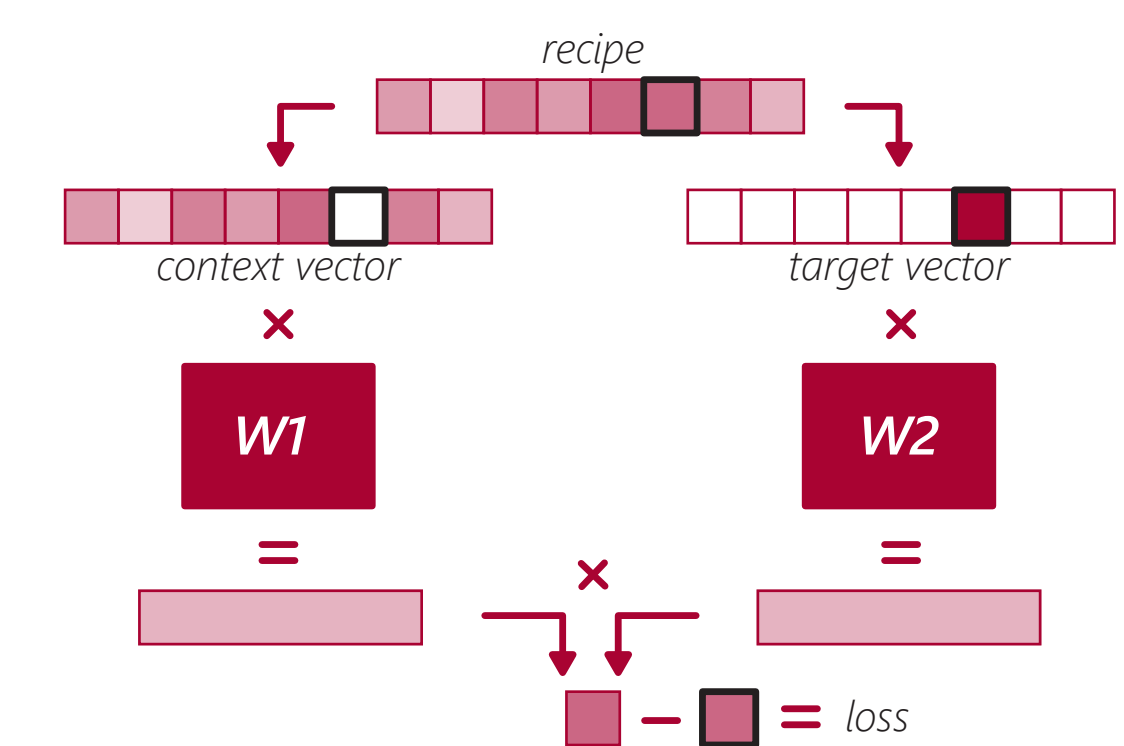
Classification Accuracy: 53%



Ingredient Substitution

Word2vec Neural Network

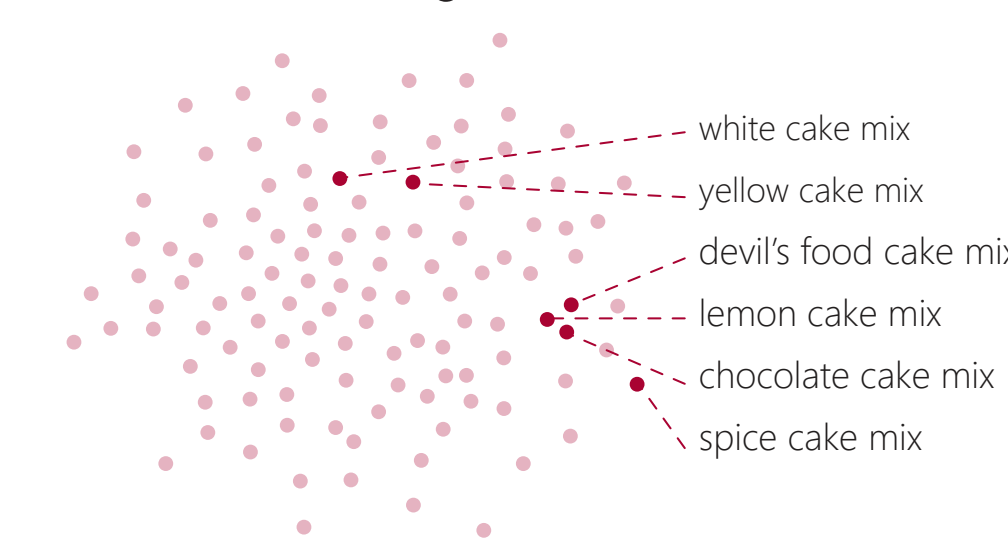
Model: We adapted Mikolov *et al.*'s word2vec model for generating vector representations of features, allowing us to synthesize ingredients as vectors in a high-dimensional space.



Substitutes for "chocolate cake mix"

- | | |
|---------------------------|--------------------------|
| Best Ingredients | Worst Ingredients |
| • "devil's food cake mix" | • "semi-sweet chocolate" |
| • "lemon cake mix" | • "white sugar" |
| • "coconut" | • "brown sugar" |
| • "chocolate pudding mix" | • "peanut butter cup" |
| • "marshmallow" | • "all-purpose flour" |

Embeddings of "cake Mix"



Embeddings of "flour"



Future Directions

Future work would involve integrating our models for rating prediction and ingredient substitution into one tool to generate highly-rated recipes given a set of dietary constraints. Doing this would likely require more data for many types of meals. There are a number of more nuanced approaches that can be further explored, such as reverse-engineering our rating network. The tools we've explored should make this process fairly straightforward.