Snackr: A swipe-based recipe recommendation app Al Romero and Mary Deignan Automations and Workflows December 17, 2024

1 Introduction

Home cooks around the world are constantly faced with the question of what to cook. This decision quickly becomes overwhelming as cooks factor in ingredients, cooking time, nutrition goals, and personal preference. This project intends to make finding new recipes fun, not exhausting.

Our final product is a user-friendly interface designed to bring more joy to meal planning. Rather than searching online endlessly, users can simply swipe through a list of recipes. As users interact by swiping, our application's recommendation engine learns from their preferences to suggest recipes they're most likely to enjoy. The user can take a closer look at each recipe with an information page before making their decisions. When a user "likes" a recipe, it is saved to another page within the application, so users can return to recipes they have liked for future reference. This page contains all of the necessary information about the recipe including directions, ingredients, and nutrition information.

2 Data Description

The data obtained from this project was scraped from Allrecipes.com. The scraping code was adapted from Dmitry Zub's GitHub [3]. Additional information on how we scraped and prepared the data can be found in the Methods section. Our raw data can be seen in Figure 1.

3 Methods

3.1 scraper.py and Feature Modification

Data was obtained using our custom python script, src/scraping/scraping.py. The script operates in three main parts.

The link collection stage begins by accessing the Allrecipes A-Z categories page. Using the get_category_links() function, it extracts URLs for all recipe categories. For each category, get_recipe_links_from_category() navigates to collect all recipe URLs on the category page. These URLs are stored in src/scraping/links/links.txt as tab-separated values, pairing each recipe URL with its category URL.

In the data extraction stage, the extract_recipe_data() function processes each recipe URL to collect structured data in seven categories: basic_info, prep_data, ingredients, nutritions, directions, image_filename, and category_url. The basic_info column

contains recipe details including title, category, rating, and rating count. The prep_data column includes preparation-related information such as cooking time and servings.

In the storage and organization stage, recipe data is structured and saved to a JSON file. This file is stored in scraping/recipes/scraped_recipes.json. Recipe images are downloaded as PNG files to scraping/images/. Image filenames in the JSON file correspond directly to the saved image files, allowing easy matching between recipes and their images. The data without any modifications can be seen in Figure 1.

Calcine Cheese Sloppy Cheese Sloppy Cheese Sloppy Calcine								
Cheese Sloppy Cype_Lime : 10 mins Cook_Lime : 10 mins Cook		basic_info	prep_data	ingredients	nutritions	directions	image_filename	category_url
Empanda Recipes, O	0	Cheese Sloppy		beef, 2 tablespoons	'fat': '39g',	over medium-high		https://www.allrecipes.com/recipes/14930/main
2 Butternut Squash Pasta', cook_lime'; '30 mins' cook_lime'; '45 mins. mystenetic facts'; '22. mystenetic facts'; '22. mystenetic facts'; '12. mystenetic fa	1	Empanada Recipes',						https://www.allrecipes.com/recipes/23131/appet
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Mexican Corn Bread, 'tat. Wexican Corn Bread Hash Wexican Corn Bread Hash Wexican Corn Bread Hash Wexican Corn Bread, 'tat. Wexican Corn Bread Hash Wexican Corn Bread Hash Wexican Corn Bread Hash Wexican Corn Bread Hash Wexican Corn Bread, 'tat. Wexican Corn Bread Hash Wexi	3	Brownies',		packed brown sugar,	'fat': '10g',	350 degrees F (175		https://www.allrecipes.com/recipes/2452/fruits
and Bacon Omelet Cups', cook_time': 10 mins', cook_time': 10 mins'	4	Mexican Corn		sweetened	'fat': '14g',	350 degrees F (175		https://www.allrecipes.com/recipes/342/bread/q
and Bacon Omelet Cups', cook_time': 10 mins', cook_time': 10 mins'								
19074 Freezer Jam; Coak Lime; '10 mins, coak Lime; '20 mins, coak Lime; '12 mins, coak Lime; '13 mins, coak Lime; '15 mins, additional Lime; '20 mins, coak Lime; '15 mins, additional Lime; '20 mins, coak Lime; '15 mins, additional Lime; '20 mins, coak Lime; '10 mins, coak Lime; '10 mins, additional Lime; '20 mins, coak Lime; '10 mins, coak Lime; '10 mins, additional Lime; '20 mins, additional Lime;	19073	and Bacon Omelet		cups frozen shredded	'fat': '22g',	425 degrees F (220		https://www.allrecipes.com/recipes/205/meat- an
19075 Chocolate Chip Oatmeel Cook_ime: 120 mins, cook_time: 120 mins, cook_time: 121 mins. packed light br 19076 Restaurant Slaw, Category: ("title: 'Sweet Cobbler, 'Category: ("title: 'Sweet Cobbler, 'Category: ("title: 'Sweet Cobbler, 'Category: ("title: 'Sweet Cobbler, 'Category: ("title: 'Sweet Sugar, 1 teaspoon ground ("dig "dig "d	19074	Freezer Jam',		strawberries, 1 1/2		ingredients., Dotdash Meredith	N/A	https://www.allrecipes.com/recipes/1961/side-d
19076 Restaurant Slaw; 'Category' 'Grep_time': '15 mins, 'additional_time': '2. 'tablespoors' 1 (the once) bag colesia wmix, 2 fatti': '12g', coles	19075	Chocolate Chip		softened, 1 cup	'fat': '8g', 'carbs':	ingredients. Preheat	N/A	https://www.allrecipes.com/recipes/851/dessert
19077 Cobbler, 'calegory': ("prep_time": '10 mins., 'sugar, 1 teaspoon 'fat': '1g, 'carbs': 350 degrees F (175 827642f75932475fb553990c838ddf83.jpg 'sugar, 1 teaspoon 'fat': '1g, 'carbs': 41g degree 41g degree	19076	Restaurant Slaw',		coleslaw mix, 2	'fat': '12g',	ingredients., Combine coleslaw	N/A	https://www.allrecipes.com/recipes/96/salad/
9078 rows × 7 columns	19077	Cobbler', 'category':		sugar, 1 teaspoon	'fat': '1g', 'carbs':	350 degrees F (175		https://www.allrecipes.com/recipes/361/dessert
	19078 ro	ws × 7 columns						

Figure 1: Raw Data

We separated these category groups so each feature has its own column (prep_time and cook_time have their own columns rather than being together in prep_data). We then removed any rows where there was no matching image so all recipes that appear have a corresponding image. We also removed all rows where any blank or null values appeared as well as their corresponding image. This was done to reduce the size of the data so this project can be stored on GitHub. After removing these rows, we reduced our dataset from 19,078 recipes to 6,360.

Our project goals required some feature modification in order to provide additional useful information about each recipe. We conducted three major feature modifications: verb_count, ingredient_count, and yield_servings_merge.

- 1. verb_count indicates the number of verbs in the directions column of a recipe. Since recipes sometimes hide a large number of steps in a single "step" designation, we used this to provide a more accurate picture of the number of steps necessary for a single recipe. This was determined using nltk part-of-speech tagging. This estimation is imperfect (due partially to the difference in sentence structure between recipes and normal sentences), but is close enough to be valuable.
- 2. ingredient_count indicates the number of items in the ingredients column of a recipe. This entry is formatted as a list of strings, so it simply counts the number of

- strings in the list. This attribute helps to quantify the ingredients list, so users can tell from a glance if the recipe requires a large amount of ingredients without having to parse the list itself.
- 3. yield_servings_merge indicates the yield of the recipe from the column yield, if it exists. If the yield is not available, it indicates the servings for the recipe from the column servings. We thought that the yield column contained more objective data, without making assumptions about how much a serving is. However, this data was sometimes missing, so we decided to supplement it with data from the servings column, which is still useful, but more subjective. We then saved the data as data/recipe_data.csv. The featured dataset can be seen in Figure 2.

	ingredients	directions	image_filename	category_url	title	category	rating	rating_count	cook_time	total_time	yield	servings	calories	fat :	carbs	protein	verb_count	ingredient_count	yield_servings_merge
C	[1 pound lean ground beef, 2 tablespoons dried	[Heat a large skillet over medium- high heat. C	5410153- e9807f995fb74fe58bd2fcd56b541010.jpg	https://www.allrecipes.com/recipes/14930/main	Grilled Cheese Sloppy Joes	Sloppy Joes				1 hr 10 mins	8 sandwiches			39g					8
ŝ	[1 cup butter, 1 cup packed brown sugar, 1 cup	[Preheat the oven to 350 degrees F (175 degree	1059053- 70a5c73ef725427e9770ad083c423e87.jpg	https://www.allrecipes.com/recipes/2452/fruits	Golden Yarn Brownies	Yarns			30 mins		24 servings								24
4		[Preheat the oven to 350 degrees F (175 degree	9281485-authentic-mexican-corn-bread- Lisa-Lode	https://www.allrecipes.com/recipes/342/bread/q	Authentic Mexican Corn Bread	Cornbread			45 mins		1 pan combread		448	14g	74g				12
e	[6 (5-inch) corn tortillas, 3 cups chopped coo	[Place a paper towel on a microwave- safe plate	5414541- 1e498b617cf84c479998b9903009a851.jpg	https://www.allrecipes.com/recipes/17874/main	Traditional Mexican Street Tacos	Tacos				10 mins	6 tacos			30g	44 g	64g			2
7	rolled oats, 1 ½ cups sifted pastry	[Preheat the oven to 350 degrees F (175 degree	7368576- 869e126fb14a489bbcb6df6fb3c827f6.jpg	https://www.allrecipes.com/recipes/841/holiday	Gramma's Date Squares	Christmas Cookies			25 mins	50 mins	12 squares				64g				12
-																			-
19061	[3 large eggs, 2 cups white sugar, 1 cup veget	[Preheat the oven to 350 degrees F (175 degree	459186-chocolate-zucchini-muffins-Pam- Ziegler	https://www.allrecipes.com/recipes/348/bread/q	Chocolate Zucchini Muffins	Zucchini Bread			20 mins	35 mins	2 dozen muffins				30g				24
19066	purpose	[Preheat the oven to 350 degrees F (175 degree	688015- 3c19325f6a6d4f93b0f8928cf0e47b2e.jpg	https://www.allrecipes.com/recipes/348/bread/q	Chocolate Chip Orange Zucchini Bread	Zucchini Bread			50 mins	1 hr 10 mins	2 9x5-inch loaves								20
19071	[2 pounds ground chuck, 2 tablespoons	CHUCK III a	2463212-tacos-de-matamoros-Allrecipes- Magazine	https://www.allrecipes.com/recipes/17874/main	Tacos De Matamoros	Tacos	4.3	111	1 hr 30 mins	2 hrs	6 servings		629	39g		38g			6

Figure 2: Modified Data

3.2 Recommendation system: classifier.py

The recommendation algorithm is based on a graph convolutional network, where the embeddings and the similarity between embeddings make up the graph. The goal of the recommendation algorithm is to make predictions while disrupting the user flow as little as possible. To do this, there are intervals in place to give the model time to train and make predictions.

The model trains on the first four selections by the user, then it begins predicting once the user has swiped ten times. Once predictions are made, a counter is reset in the interface, so that the model trains after the next four swipes and predicts after ten swipes. For efficiency, the model runs on separate QThreads from the interface for both training and prediction. QThreads allow the model to run while other actions within the interface occur. The model itself is composed of two classes: RecipeDataClassification and TwoLayerGCN.

RecipeDataClassification is the main class, which trains and makes predictions based on input data.

Training begins by creating BERT embeddings of each recipe's directions text, which captures both ingredients and cooking methods. For each recipe, these embeddings are averaged into a single vector for efficient comparison. A graph network is then constructed where nodes are recipes and edges represent recipe similarity, calculated using Euclidean distance between their embedding vectors. The TwoLayerGCN classifier, based on Kipf et al.'s 2017 work [1], processes this graph structure over 100 training epochs to learn patterns in user preferences. After training, the model is saved for later predictions.

For prediction, the system selects 20 random recipes and processes them through the same embedding pipeline. It converts directions to BERT embeddings and constructs a graph based on recipe similarities. The saved TwoLayerGCN then evaluates these recipes, outputting values between 0 and 1 representing the predicted likelihood of user preference for each recipe.

3.3 interface.py

The interface is the main route through which users can interact with the project. It was written using PyQt5, a package for designing apps with Python. The main goal we had when creating this interface was to build an easy-to-use and simple interface that users enjoyed swiping through. We modeled the interface design after popular swipe-based applications like Tinder, and created three major 'pages': the swiping interface, the 'liked recipes' page, and the 'further information' page with information about individual recipes. Users can swipe on recipes, view details about recipes they've 'liked' in the past, and view more information about the current recipe before they decide whether to swipe left or right. The swiping page has click-and-drag swiping as well as buttons for selecting user preferences, along with a button that provides more information about each recipe. The information button additionally provides many of the statistics calculated during feature modification. The Liked Recipes page provides more information about each recipe, including directions.

We wanted to implement a recommendation system within the interface that would adapt to the user's preferences without interfering with the user's experience. To do this, much of the recommendation system is run using QThreads, a PyQt5 class that encourages efficient resource management. This meant that we could run computationally expensive model training and prediction code without interfering with the user's experience. Essentially, the recommendation system is called based on the number of times the user has swiped. The model is first trained once there have been four swipes, then it makes its first predictions at ten swipes. Predictions are made on a random set of twenty recipes from the original dataframe data/recipe_data.csv. Once predictions have been successfully made, they are presented to the user in the order of most likely to like to least likely, and then the swipe counter is reset, so the process can continue.

4 Results

Our web scraping implementation successfully gathered a comprehensive dataset of 6,360 recipes from Allrecipes.com, which was used to create an interactive recipe swiping applica-

tion. The interface design focuses on presenting recipe information in a clear and user-friendly format.

Figure 3 shows the main interface page where users can view individual recipes. Each recipe is displayed with its image, title, rating, and total preparation time. The interface includes interaction buttons for accepting or rejecting recipes that can be used rather than the swiping feature. This page also shows an info button next to the recipe title. When this button is selected, a recipe details preview page comes up, as seen in Figure 4. This main page (Figure 3) also shows the "Liked Recipes" button which takes users to the liked recipes page, as shown in Figure 5. The liked recipes page allows users to easily view all recipes they have liked since they began using the application. Finally, users can select a given recipe title from the liked recipes list to view all relevant information needed to prepare the recipe, as seen in Figure 6. This recipe details page shows users all the information from the details preview page as well as the directions. Figure 7 shows a diagram laying out how the pages are connected.

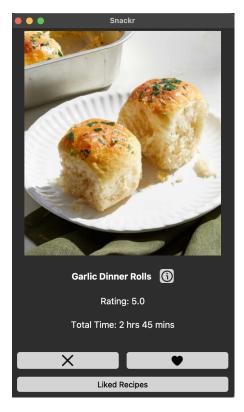


Figure 3: The main page of the interface. Users can swipe to like or dislike a recipe, click to display more information about the recipe, or open a list of their liked recipes.



Figure 4: The recipe details preview page. Users can view ingredients, preparation time, and other key recipe information before deciding to like or dislike.

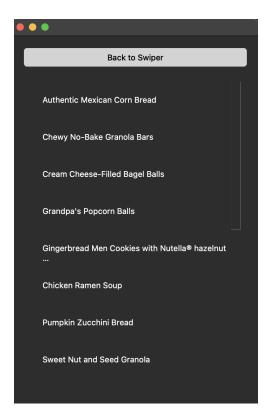


Figure 5: The liked recipes page showing a scrollable list of all recipes the user has liked. Users can click any recipe title to view its full details.

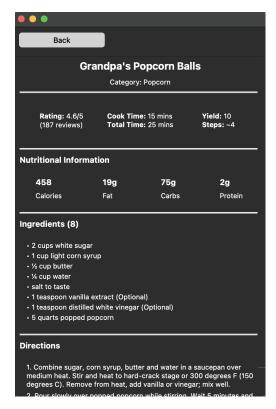


Figure 6: The complete recipe details page for liked recipes. Users can view all recipe information including full preparation directions and ingredient measurements.

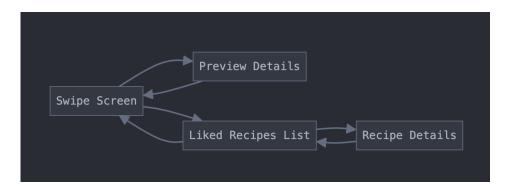


Figure 7: Diagram showing transitions between pages in the interface.

5 Future Work

Future work will focus on improving the recommendation system with more sophisticated algorithms, conducting more thorough stress testing of the interface, particularly around the

implementation of the recommender, and refining both the interface design and the code for improved performance. The recommender could be improved by using batch embedding to speed up training and prediction times, and a more sophisticated method of comparison between individual embeddings.

6 Contributions

Al has worked on feature engineering, built the recommender algorithm, implemented the recommender within the interface, and helped with the interface design and building. Mary has worked on scraping and documenting the scraping code, building the interface, and helping with the recommender algorithm.

References

- [1] Thomas N. Kipf and Max Welling. Semi-supervised classification with graph convolutional networks, 2017.
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