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RECIPE RECOMMENDATION ENGINE

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ABSTRACT: The widening variety of gastronomic preferences and the accessibility of enormous ingredient datasets have led to a considerable increase in the use of recipe suggestion systems in recent years. This study proposes a novel method for recommending tailored recipe recommendations based on user-supplied input ingredients. The system analyzes ingredient patterns and builds links between them using cutting-edge machine learning algorithms the suggested technique uses a multi-step methodology to produce customized recipe recommendations. First, a thorough ingredient database that contains details on various recipes and the ingredients that go into them is developed. After that, a powerful natural language processing (NLP) module pulls useful information from the user's input ingredients so that the system can recognize the flavors and culinary aesthetic that are intended. The collaborative filtering technique at the center of the recommendation system makes use of the ingredient data gathered to find recipes that are comparable to the user's input. The system may provide recipes that complement the user's taste preferences while also being related by ingredient overlap by using collaborative filtering.

KEYWORDS: recipe suggestion, Ingredients, Collective filtration, Nutritional imitation, Profiling of flavors;

I. INTRODUCTION

A wide variety of cuisines and flavors from all around the world are now included in the culinary discovery and art that has grown greatly over time. Technology's development and the ubiquitous availability of ingredients have given people access to an unmatched range of culinary creations. For home cooks and food enthusiasts, choosing a recipe that matches their unique preferences and accessible components might be difficult amidst this abundance.

A solution to this problem has developed in the form of recipe recommendation systems, which use machine learning and data analytics to provide personalized recipes. suggestions that are catered to each person's preferences and cooking style.

II. LITERATURE REVIEW

The primary objective of the project is to create a recommendation system that assists users in making the most of their available ingredients, minimizing food waste, and promoting creative cooking. By leveraging a vast dataset of recipes and implementing machine learning algorithms such as K-nearest neighbors (KNN) with means and cosine similarity index, we can generate accurate and relevant recipe recommendations.

III. LITERATURE REVIEW

2.1 Present System

In the current situation, the Food Donation System Rarely The culinary world has expanded tremendously over the years, embracing a wide variety of foods and flavors from all around the world. People now have access to an unrivaled range of culinary creations thanks to the development of technology and the ubiquitous availability of ingredients. To choose a dish that matches their own preferences and available components can, however, frequently be difficult for home cooks and food enthusiasts in the midst of this abundance.

In order to solve this problem, recipe recommendation systems have evolved. These systems use machine learning and data analytics to generate personalized recipe suggestions that are catered to each person's preferences and cooking preferences.



The suggested recipe recommendation system impressed testers with its performance by providing users with appealing recipe choices based on their input items. The model's versatility and flexibility make it possible for it to consider a variety of cuisines, different ingredient lists, and changing culinary trends.

Overall, this research provides a powerful and user-friendly recipe suggestion system that makes use of collaborative filtering and machine learning to cater to people looking for culinary inspiration while reducing food waste by making inventive use of readily available resources.

2.2 Proposed System

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IV.PARTS

Login with e-mail

User ID and Password which will be created by user.

1. User Input:

This represents the interface through which users input their available ingredients into the system. It can be web form or any other user input method.

2. Data Processing:

This process receives the user input and performs data processing tasks. It may involve cleaning and validating the input, extracting relevant information and preparing it for further analysis.

3. Recipe Recommendation:

This process utilizes the processed data to generate recipe recommendations based on the ingredient matching algorithms, such as KNN with means and cosine similarity index. It retrieves relevant recipes from the recipe dataset and applies the algorithm to determine suitable recommendations.

4. Display Results:

This represents the interface where the recommended recipes are displayed to the user. It can be a web page or any other form of visual representation.

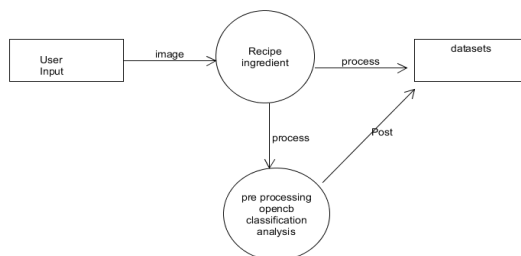


Fig 2. Activity Diagram for user

In these diagram new user will register using their e-mail and give the password & once again confirm the password. Next the website will provide the login page user can login and prediction will be shown for the user selected companies

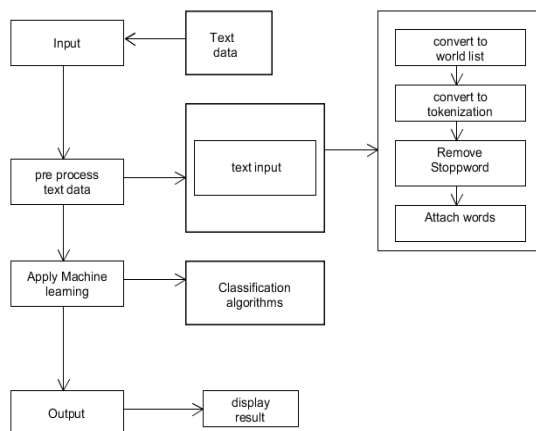


Fig 3. A Design of System Architecture

In this system user will be login with the registered login credentials, it is user friendly, easy to use, user interface is good for using the process. This project is socially useful for many people who would like to explore new food as the saying goes, “good food is good mood.” As the user login, the outlook of the system would show a page for recommendations, prediction with the input area for ingredients.

Once the user gets inside the system they will provide the ingredients on hand to the recommendation system. And the system will respond with predicted recipes. The system does not just predict for the given list of ingredients but also provides the related recipes for the ingredients given by the user in textual form.

As it provides the information of the recipes it also provides the YouTube link for the preparation of the particular dish or a recipe recommended by the system which is very conventional and user friendly. If the user does not want to spend time on the video looking for then there is a way out with the text information for cooking the mentioned recipe.

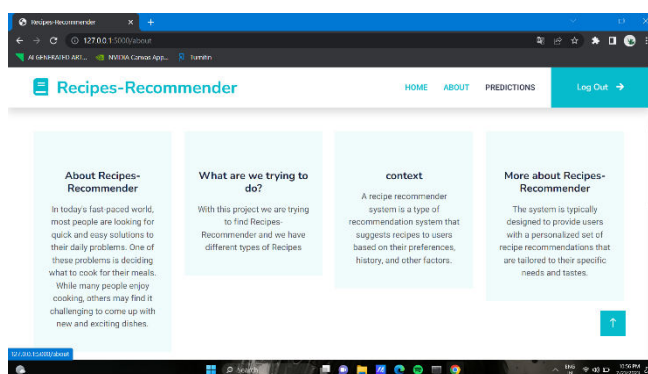


Fig 4. Home Page

In this home page it holds route login page and about the system built. Where a new user can get registered and get access to the system.

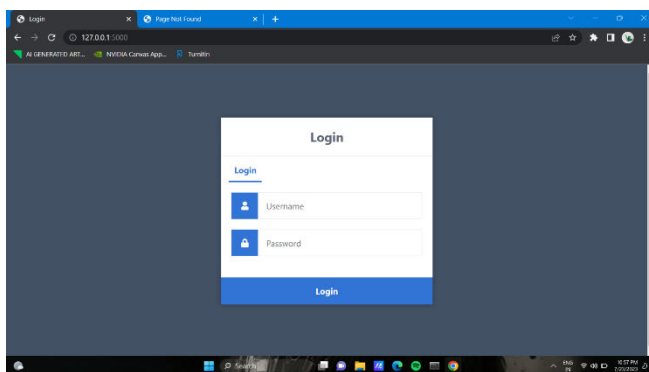


Fig 5. User login page

These page only for registered user. User will fill the above log-in details user name and password for logging in prediction account.

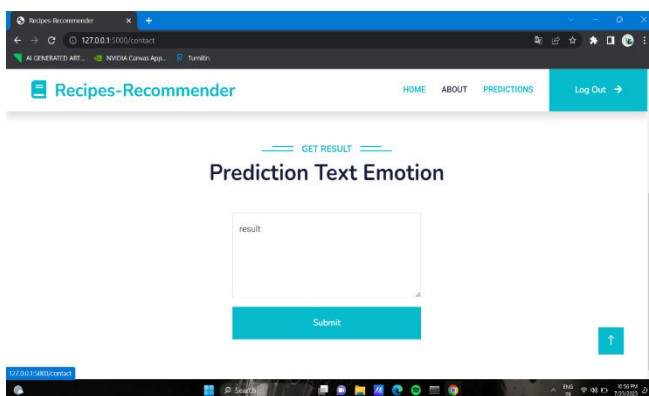


Fig 6. Ingredients Input

In this page of the system the user will provide the ingredients list with the comma separators. Where once the user clicks on the submit button the ingredients to which the recipe should be recommended will enter the system and provides the list of information.

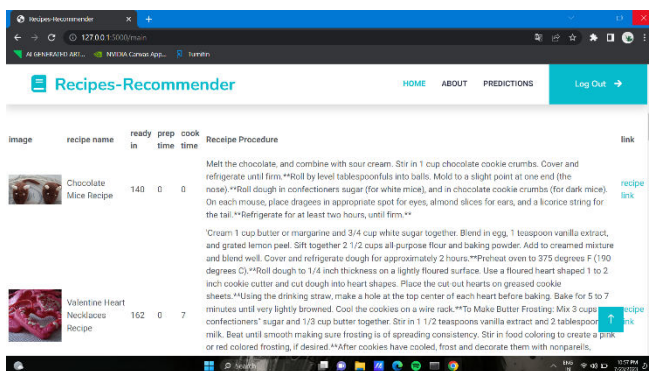


Fig 7 Top Recipes Recommended

The once user will login the prediction account they provide the recommended list of recipes. User can select the any recipe from the predicted list for further information.

Here this page not just provides the recipe but also helps with time required to cook the recommended recipe and the way to cook or prepare the recipe in textual information.

As it provides the information it will also add up the You Tube link for preparation of that particular recipe.



V. SYSTEM LIMITATIONS

There are various phases involved in putting a Recipe Recommendation System Based on Ingredients into action. An overview of the system implementation procedure is provided below:

Gather a broad collection of recipes, along with their ingredient lists, cooking instructions, and other pertinent metadata.

To eliminate duplication, handle missing values, and standardize ingredient names and measures, clean and preprocess the recipe data.

Construction of the Ingredient Database:

To store the preprocessed recipe data, create an organized debasement.

Food Trend Inclusion and Diversity:

Utilize strategies to guarantee that recipe ideas span a variety of cuisines and consider recent culinary trends.

For a variety of options, group recipes based on flavors and cooking methods using clustering algorithms.

Creative Ingredient Association: A module that proposes recipes using the user's input items in inventive ways encourages users to experiment with their available ingredients and decreases food waste.

Infrastructure that is both scalable and cloud-based: The system architecture should be built to support both an increasing user base and a growing ingredient database. A scalable and cloud-based architecture gives the system the ability to deploy resources effectively and react to changing demands.

Training and Updating Machine Learning Models: To stay up with user behavior and shifting gastronomic trends, the collaborative filtering and NLP algorithms may need regular training and updating. The recommendation engine's accuracy and correctness are maintained by an automated pipeline for model training and updates.

VI. FUTURE SCOPE AND IMPROVEMENTS

- ▶ Improved NLP and collaborative filtering techniques utilized in the recommendation system can produce more precise and context-sensitive recipe suggestions. Modern machine learning methods, including transformer-based topologies, can be incorporated to further improve the system's comprehension of user preferences and ingredient connections.
- ▶ Deep Learning for Recipe Generation: By using deep learning models to create novel recipes depending on the items at hand, users may be inspired to try out inventive and one-of-a-kind dishes. To develop dishes that suit different palates, these algorithms might learn from huge recipe libraries and user feedback.
- ▶ Nutritional analysis integration: Future systems might include nutritional analysis to inform users of the nutrient makeup of suggested dishes. Users with special dietary objectives or health concerns would benefit from this functionality.
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- ▶ Integration of User Context: The system can provide more accurate and customized recipe recommendations by considering user context, such as the time of day, the weather, and the user's location. For instance, it might offer cool drinks on a hot day or warming stews on a chilly day.
- ▶ Social Sharing and Community Engagement: Including social sharing capabilities and developing a user community can encourage communication and participation. Users might create a vibrant and encouraging cooking community by sharing their preferred recipes, cooking advice, and adaptations of suggested dishes.



VII. CONCLUSION

The system can recommend meals that are in line with the user's ingredient preferences and culinary preferences by curating an extensive ingredient library and using collaborative filtering algorithms. The recommendations are further improved by incorporating user feedback and customization, guaranteeing that the system's accuracy and relevancy improve over time.

The suggested recipes are secure and suitable for each person's needs because the system considers dietary restrictions and allergies. Furthermore, its ability to support various cuisines and culinary fads expands the range of cooking options, appealing to a variety of users with disparate preferences and interests.

Additionally, the system promotes inventive ingredient pairings, encouraging users to experiment with the items they have on hand and minimizing food waste by making the best use of resources.

The topic of Recipe Recommendation Based on Ingredients has a huge future potential for advancements and improvements as technology and data continue to grow. The system can offer even more precise, context-aware, and customized suggestions by combining deep learning models, augmented reality, nutritional analysis, and user context.

Additionally, encouraging social sharing and community involvement can build a vibrant and helpful cooking community where users can share their culinary knowledge and experiences.

Future improvements to the system will focus on sustainability and eco-awareness, encouraging the use of locally sourced and seasonally appropriate ingredients to encourage environmentally responsible cooking techniques.

Overall, Recipe Recommendation Based on Ingredients gives consumers the freedom to experiment with new flavors, explore a wide range of culinary possibilities, and go on thrilling culinary adventures without leaving the comfort of their own homes. This system has the potential to transform the way people cook by adopting cutting-edge technologies and user-centric features, cutting down on food waste and raising the enjoyment of cooking to new levels.

REFERENCES

- [1] Wang, Haoyu, et al. (2018) "A Stock Recommendation System Using with Distributed Graph Computation and Trust Model-Collaborative Filtering Algorithm." 2018 2nd IEEE Advanced Information Management, Communicates, Electronic and Automation Control Conference (IMCEC). IEEE
- [2] Thakkar, Priyank, et al. "Combining -Based and Item-Based Collaborative Filtering Using Machine Learning." Information and Communication Technology for Intelligent Systems. Springer, Singapore, 2019. 173-180.
- [3] Pereira, Nymphia, and Satish Kumar L. Varma. "Financial Planning Recommendation System Using Content-Based Collaborative and Demographic Filtering." Smart Innovations in Communication and Computational Sciences. Springer, Singapore, 2019. 141-151
- [4] Pereira, Nymphia, and Satish Kumar L. Varma. "Financial Planning Recommendation System Using Content-Based Collaborative and Demographic Filtering." Smart Innovations in Communication and Computational Sciences. Springer, Singapore, 2019. 141-151
- [5] Qian, Yongfeng, et al. "EARS: Emotion-aware recommender system based on hybrid information fusion." Information Fusion 46 (2019): 141-146.
- [6] Kolla, Bhanu Prakash, and Arun Raja Raman. "Data Engineered Content Extraction Studies for Indian Web Pages." Computational Intelligence in Data Mining. Springer, Singapore, 2019. 505-512
- [7] Patel, Ankit Dilip, and Yogesh Kumar Sharma. "Web Page Classification on News Feeds Using Hybrid Technique for Extraction." Information and Communication Technology for Intelligent Systems. Springer, Singapore, 2019. 399-405 www.ijcrt.org © 2021 IJCRT | Volume 9, Issue 6 June 2021 | ISSN: 2320-2882 IJCRT2106580 International Journal of Creative Research Thoughts (IJCRT) www.ijcrt.org f6
- [8] Goswami, Saptarsi, et al. "A review on application of data mining techniques to combat natural disasters." Ain Shams Engineering Journal 9.3 (2018): 365378.
- [9] Zhao, Rui, and Kezhi Mao. "Fuzzy bag-of-words model for document representation." IEEE Transactions on Fuzzy Systems 26.2 (2018): 794-804.
- [10] Sang, Jitao, Ming Yan, and Changsheng Xu. "Understanding Dynamic Cross OSN Associations for Cold-start Recommendation." IEEE Transactions on Multimedia (2018).



- [11] Khan, Sadik, Yashpal Singh, and Kalpana Sharma. "Role of Web Usage Mining Technique for Website Structure Redesign." International Journal of Scientific Research in Computer Science, Engineering and Information Technology 3.1 (2018)
- [12] Logesh, R., and V. Subramaniya Swamy. "Exploring Hybrid Recommender Systems for Personalized Travel Applications." Cognitive Informatics and Soft Computing. Springer, Singapore, 2019. 535-544. [13] Sun, Y., Fang, M. and Wang, X., 2018. A novel stock recommendation system using Guba sentiment analysis. Personal and Ubiquitous Computing, 22(3), pp.575-587.



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