

Today's Menu

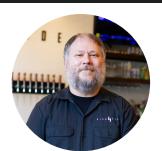
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Meet the Team



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Grant Fitch Project Lead Iron Chef



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Sommelier / Cicerone



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Conseiller Culinaire



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Problem Statement

Despite the fact that dining out offers a rich experience—bringing people together and enhancing social bonding—rising inflation has shifted the behavior of many Americans. With inflation up and restaurant prices increasing by 4.1%, 68% of Americans are now choosing to eat at home. [8] This means that people are missing out on new culinary experiences and the well-being benefits of communal dining.[4] Furthermore, with the overwhelming number of restaurant choices and generic reviews, finding the perfect dining option has become a risky financial decision, leaving many diners hesitant to explore new venues.

Problem Characteristics

High Financial risk:

With restaurant prices outpacing inflation, dining out has become a more expensive and risky decision for the average consumer. In recent years, United States food prices rose by 25%. [11]

Overwhelming Choice:

Customers experience indecision when selecting dishes, making it difficult to confidently choose meals they will enjoy based on taste.

Generic Reviews:

Online reviews may not accurately reflect the customer's personal taste, leading to dissatisfaction in the dining experience. About 30% of online reviews are fabricated. [9] How do you know which reviews to believe?

Group Indecision:

Studies show that group decisions regarding where to eat are heavily influenced by social environment.^[12] Can lead to individuals eating at places they do not enjoy just to fit in with the group and avoid conflict.

CS 410 TasteBuddies - Team Iron

Solution: Dine With Confidence

TasteBuddies is a smartphone app that will provide tailored restaurant and dish recommendations based on taste profiles. TasteBuddies will use data clustering to connect users with others who share similar preferences, offering relevant suggestions rather than generic reviews. TasteBuddies dynamically enhances user confidence with real-time feedback from crowdsourced data on dish quality and level of business, adding a layer of insight to support an optimal dining experience.

By using TasteBuddies, diners are more likely to end up with a meal they truly enjoy, while reducing the stress of sifting through irrelevant reviews, enhancing their overall dining experience. Restaurants will also benefit from fewer complaints, less food waste, and happier customers who are more likely to return, give positive reviews, and tip well.

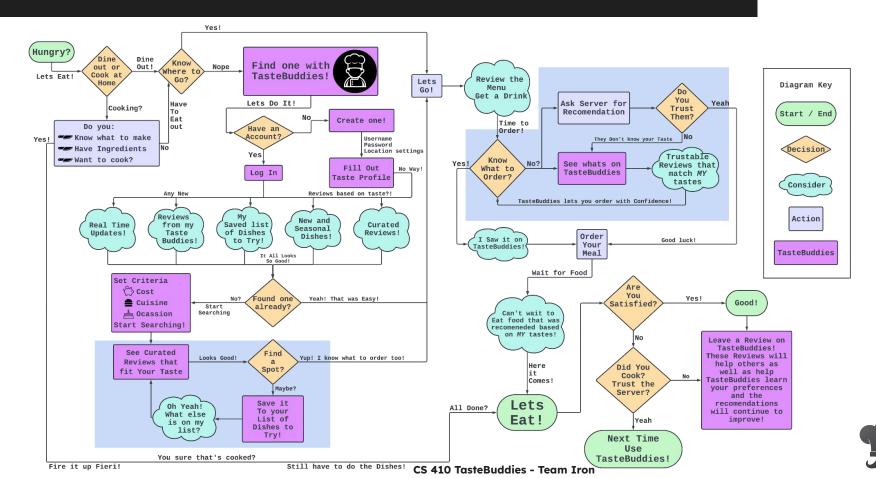


Solution Characteristics

- Personalization: Our revolutionary app will provide personalized dish and restaurant recommendations tailored to individual tastes rather than offering a one-size fits all approach
- **Tailored Recommendations:** Instead of relying on broad, generic reviews, the app connects users with others who have aligned taste profiles, offering relevant reviews and a customized dining experience
- **Customer Satisfaction:** By offering recommendations based on individual preferences, the app helps customers get a better value for their money, and food they truly enjoy, enhancing the dining experience.
- **Reduced Waste:** With more accurate recommendations, fewer dishes are sent back due to dissatisfaction, reducing food waste and lost revenue for restaurants.
- Increased Tips: Happier customers lead to increased tips for working staff and more positive reviews, benefiting restaurant owners and workers.
- Crowdsourced Real-Time Updates: Our platform empowers users to share live updates on restaurant conditions, from wait times and menu availability to special events, ensuring a dynamic and responsive platform that adapts to users' real-time dining needs.



Solution Process Flow



Product Prototype Ingredients

Category	Features	RWP	Prototype	Additional Notes
	Account Creation	•	•	
Account	Login / Authentication	•	Eliminated	
Management	Access Permissions and Preferences	•	Partially Implemented	Access Permissions required for database
	Taste Profile	•	•	
	Social Engagement	•	Partially Implemented	Find friends only for group matching
	Daily Dish Feed	•	•	
	Group Restaurant Matching	•	Partially Implemented	Implementation is time dependent
	Dish Recommendations	•	•	
	Taste Profile Builder	•	•	
Mobile App Features	Reviews	•	Partially implemented Mock data for compatibility matching	
redidies	Community Updates	•	Eliminated	
	Dish Validation	•	Eliminated	
	Taste Matching	•	•	
	Notification Features	•	Eliminated	
	Engagement Features	•	Eliminated	
	Data Analytics	•	Eliminated	
DataBase	Data Privacy and Security	•	•	
Management	Trend Reports	•	Eliminated	
	Data Backups	•	•	

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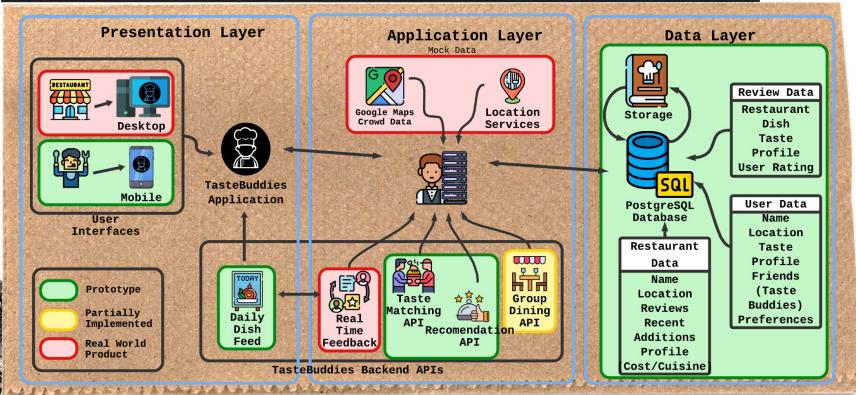
Co	ategory	Features	RWP	Prototype	Additional Notes
		TasteBuddies	•	•	· · · · · · · · · · · · · · · · · · ·
	Control	Super TasteBuddies	•	Partially implemented	Hard coded
Social Engagement	Add/Find Buddies	•	•		
		Follow TasteBuddy	•	Eliminated	
		Follow Restaurant	•	Eliminated	
		Add Kudos	•	Eliminated	
Live Interactive Updates	Daily Dish feed	•	•		
	Add reviews	•	Partially implemented	Mock data provided	
	Post restaurant update	•	Eliminated		
	Post dish update	•	Eliminated		
xpanaea Iser		Notifications	•	Eliminated	
lobile		Taste Profile	•	•	
.pp		Read Reviews	•	Partially Implemented	
eatures		Taste Matching	•	•	
	Recommend- ation Engine	Dish Recommendation	•	•	
	dhon Engine	Group Restaurant Matching	•	Partially Implemented	Implementation time dependent
		Rewards	•	Eliminated	
		Adaptive Taste Profile personalization	•	Eliminated	Need active data over time
		Restaurant filtering	•	Eliminated	
	Search	Dish filtering	•	Eliminated	
		Rewards	•	Eliminated	
	Engagement features	Badges	•	Eliminated	
	legiules	Challenges	•	Eliminated	

Kitchen Aids (Development Tools)

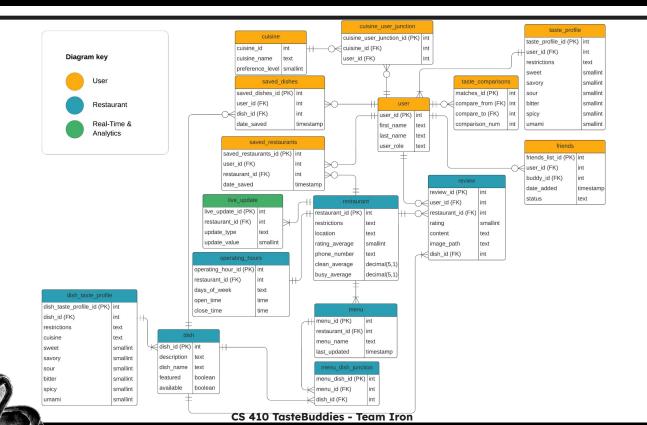
		Planned	Actual
Frontend	Framework:	React	Flask with Jinja2
Fromena	Languages:	HTML, CSS, Javascript	HTML, CSS, Javascript
Backend	Framework:	Flask	Flask with SQLAlchemy
Buckellu	Languages:	Python	Python
Testing Framewor	ks	Pytest, Jest & Maestro	PyTest
Story & Issue Trac	ker	Trello & Github Issues	Github Issues
Version Control		Git through Github	Git through Github
Documentation To	ool	Pydoc, JSDoc	PyDoc
Database		PostgreSQL	SQLite



Major Functional Components Diagram



Database Schema







Whats on the Grill

	Task	Status
	Register User & Delete User in Database	Completed
Account Creation	Login & Logout with Sessions	Completed
	Connected to Taste Profile	Completed
Taste	Setup Questions & UI	Completed
Profile:	Store in DB attached to User	Completed
Add Oliosa	UI for Finding, adding and viewing buddies	Completed
Add/View Buddies:	Basic Search and View Logic	Completed
	Add Buddy - Connect to User Buddy List	Completed



Sprint Breakdown



- Version Control, Git
- Trello Board & Github Issues
- Initial Repository Structure

Sprint 2: User Accounts (2/8 - 2/19)

- Account Creation
- Taste profile setup
- Add/View Buddies



Sprint 1: Foundation (1/24 - 2/7)

- Framework for Frontend UI
- Framework for Backend
- Setup database schema
- Connect Everything



Feb 26 Demo 1 Sprint 3: Reviews & Social (2/20 - 3/5)

- Dish Review Function
- Mock Data for Dishes, Restaurants, Reviews and User Taste Profiles
- Simple Taste Matching





Sprint Breakdown

Sprint 4: Recommendations (3/6 - 3/19)

Implement the Daily dish

feed

Dish recommendation

Sprint 6: Polish & Finalize (4/10 - 4/30)

Testing & Bug fixes

Polish UI

Review and Improve algorithms

March 26 Demo 2



Sprint 5: Daily Dish Feed (3/20 - 4/9)

Suggest Tastebuddies

Resolve UI decisions

Refine Taste Matching

(Group Dining)

May 6 Final Prototype



User Stories: TasteBuddy

As a TasteBuddy diner, I shall...

Fully **Implemented** create a personalized taste profile so I can get tailored restaurant recommendations

find people I know on the app.

see dish recommendations and how compatible I am with a dish based on my taste profile.

receive a restaurant recommendation when I create a group with other TasteBuddies.



modify and further build my taste profile

view a live feed called the Daily Dish feed that provides recommendations and live updates from users and restaurants I follow.

connect with other diners who share my taste preferences so I can get relevant recommendations

Partially Implemented find restaurants that other people with similar tastes enjoy so that I can dine with confidence



see ratings and reviews from people with similar taste preferences so I can make informed dining decisions receive notifications about specials from restaurants that match my taste profile

rate and review restaurants to help other diners with similar tastes

Not implemented have my taste profile altered based on my dish feedback and reviews.

search for specific dishes near me to find restaurants that serve food I like track my dining experiences and preferences to improve future recommendations

earn rewards through the reward system so I can get discounts at restaurants I enjoy

filter recommendations based on cuisine type, location, and price range participate in the crowdsourcing system to help keep information accurate and up-to-date

User Stories: Restaurant

As a Restaurant owner/manager, I shall...

Not Implemented reach diners who are most likely to enjoy my restaurant's offerings

maintain an updated profile of my restaurant's menu and specials

receive feedback from diners who match our restaurant's taste profile understand what dishes are most popular among different taste profiles

participate in the reward system to encourage customer loyalty

access analytics about customer preferences and dining patterns

respond to customer reviews and ratings

showcase our restaurant's specialties to targeted customers

receive notifications when our information needs updating

connect with other diners who share my taste preferences so I can get relevant recommendations

verify information reported by users about our restaurant

view data about what potential customers in our area are searching for



User Stories: Administrator

As an administrator, I shall

Fully **Implemented**



Not

implemented



access analytics about customer preferences and dining patterns implement and maintain data privacy measures ensure data accuracy and system security backup system data

analyze app usage patterns to improve user experience manage the reward system

maintain and optimize the taste matching algorithm

monitor and validate the crowdsourcing system

handle user support requests and feedback

maintain API integrations for location services and mapping monitor system performance and implement optimizations

manage database operations and maintenance

view data about what potential customers in our area are searching for

manage user profiles and authentication systems

generate reports on system usage and user engagement



User Stories to Tasks





	Account	· Management	
Feature	Task	User Story	User Type
Account Creation	Build a taste profile tied to account creation	create a personalized taste profile so I can get tailored restaurant recommendations	TasteBuddy
Access Permissions	Set up access permissions	Implement and maintain data privacy measures	Administrator
Taste Profile	Build and implement algorithm for developing taste profile	modify and further build my taste profile	TasteBuddy



User Stories to Tasks





Mobile App Features

Feature	Task	User Story	User Type
Dish Recommendations	Implement item-based collaborative filtering	see dish recommendations and how compatible I am with a dish based on my taste profile.	TasteBuddy
Taste Matching	Build algorithm that matches users to similar tastes to match dishes using k-means clustering	connect with other diners who share my taste preferences so I can get relevant recommendations	TasteBuddy
Daily Dish Feed	Implement the Daily Dish Feed to show restaurant and dish recommendations	view a live feed called the Daily Dish feed that provides recommendations and live updates from users and restaurants I follow.	TasteBuddy
Group Restaurant Matching	Build and implement group matching algorithm	receive a restaurant recommendation when I create a group with other TasteBuddies.	TasteBuddy
Social Engagement	Implement basic search	find people I know on the app.	TasteBuddy

User Stories to Tasks





	Database Maria	gement	
Feature	Task	User Story	User Type
Data analytics	Use database to analyse data	access analytics about customer preferences and dining patterns	Administrator
Data Privacy and Security	Comply with data privacy regulations and implement access permissions based on user roles	implement and maintain data privacy measures	Administrator
	Create different levels of access permissions and type checks	ensure data accuracy and system security	Administrator
Data Backups	Implement method and policy for backing up data in timely intervals	backup system data	Administrator

Database Management



Grant's orders



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	Maria III					
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	Buil	d inte	o revi	ew		
-	onc	e bas	ic			
	revi	ew st	ructu	re		
	ic c	reate	d.			
	Allo	w neg	rative			
	revi	ews t	o alte	r		
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9233						

Curact Charle





Colette's orders











Ashley's orders



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ार) ज्या		www.roya	alpaper.cor	n	GC3	632-1

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	Receipt					
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Ben's orders



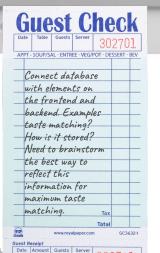








Nate's orders





Date	Table	Guests	Server			-				
Dutt	lubic	ouests	oci ici	30	270	1				
APPT	- SOUP/S	AL - ENTR	EE - VEG/	POT - DE	SSERT -	BEV				
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	how we store the taste matches. Not									
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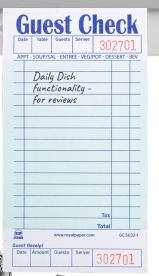
Cuant Charle

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APPT	- SOUP/S	AL - ENTR	EE - VEG/	POT - DESS	ERT - BEV
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Date	Amount	Guests	Server	302	7 - 7

Cuant Chank



Oronde's orders



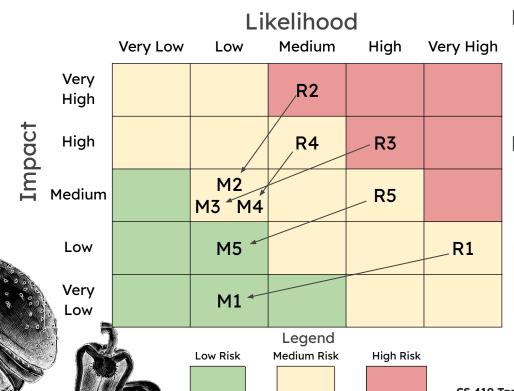








User Risk Matrix



Risks Prototype Partial/Mock Real World Product

R1: Fake Profiles and Reviews

R2: User satisfaction with their recommendations

R3: User Participation and Retention

R4: Users evolving taste preferences

R5: Users not wanting to share data

Mitigations

M1: Checking for duplicate accounts and emails, ensuring emails are unique by checking for dot agnostic and + tags for potential false emails.

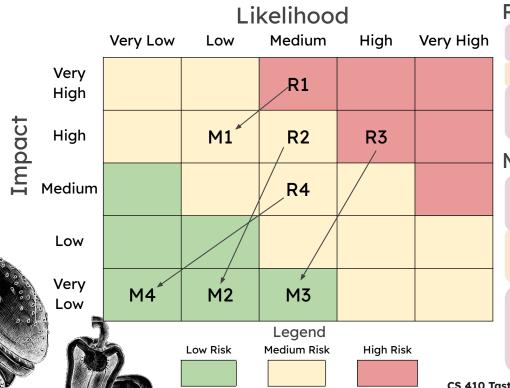
M2: Allow user feedback on recommendations

M3: Implement milestone-based Badges and rewards system for leaving reviews, and send reminder notifications to encourage participation

M4: Allow users to update preferences and periodically prompt users for updates, "Do you still like..."

M5: Allow users to opt in or out of data collection for specific features

Customer Risk Matrix



Risks Prototype Partial/Mock Real World Product

R1: Restaurants will not upgrade to premium accounts

R2: Matching with closed restaurants

R3: Matching with outdated dishes

R4: Users may feel overwhelmed by excessive notifications

Mitigations

M1: Offer trial subscription to premium accounts and data-driven insights for the success of other restaurants with premium accounts

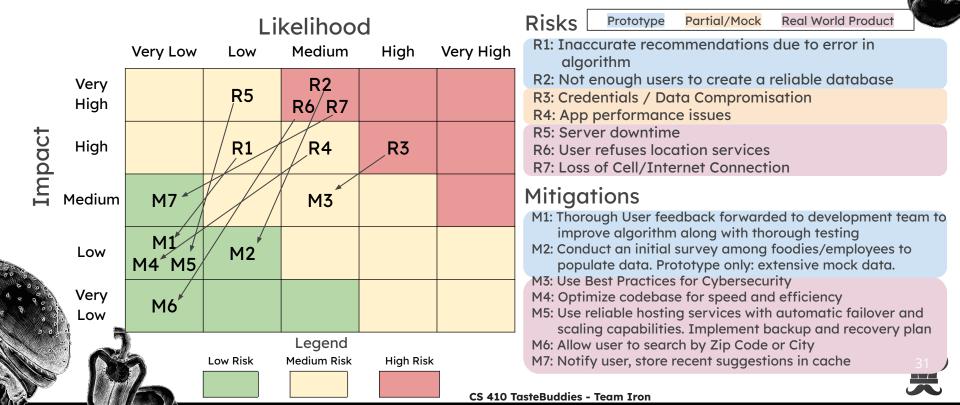
M2: Enable users to report a restaurant closure with a review process to prevent misuse, as well as reviews triggered by events (ending subscription)

M3: Enable user feedback as well as send periodic reminders to restaurants to verify dish availability

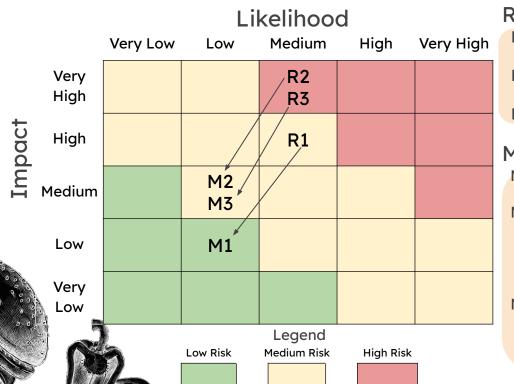
M4: Allow users to customize notification settings
Limit notifications restaurants can send
based on subscription tier

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Technical Risk Matrix



Legal & Security Risk Matrix



Risks Prototype Partial/Mock Real World Product

- R1: Civil lawsuits against the app including potential user disputes or trademark violations
- R2: Data privacy regulations and potential mishandling of user data
- R3: Allergens not listed in dish description

Mitigations

- M1: Detailed terms and conditions for both users and restaurant that must be agreed to before use.
- M2: Obtain explicit user consent in the initial terms and conditions before collecting any personal information to ensure compliance with data privacy laws, including GDPR, CDPA, and the Privacy Act of 1974
- M3: Require restaurant provided dishes to be tagged with any allergens. User submitted dishes will be tagged as 'Unverified Allergens' until the restaurant provides appropriate tags

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TasteBuddies

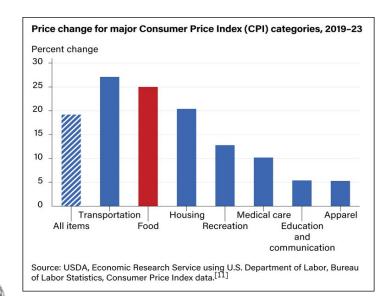


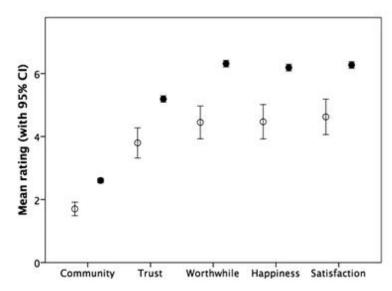
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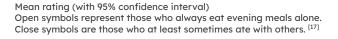


Appendix













Appendix (Matching Algorithms)

Collaborative Filtering

Collaborative filtering is one of the most common algorithms used for recommendation systems and can be applied to matching TasteBuddies.

User-Based Collaborative Filtering:

Identifies users who have similar taste preferences (e.g., similar restaurant ratings or dish preferences).

Matches users based on their shared preferences, creating a group of TasteBuddies who enjoy similar dining experiences.

Item-Based Collaborative Filtering:

Analyzes similarities between restaurants or dishes based on user ratings. Groups users who rate similar items positively, assuming that they share similar tastes.

Implementation:

Use cosine similarity, Pearson correlation, or Jaccard index to measure the similarity between users.

Matrix Factorization

Matrix factorization is a machine learning technique commonly used in recommendation systems.

How It Works:

Decomposes a user-item interaction matrix (e.g., ratings of dishes or restaurants) into latent factors.

Matches users with similar latent factors, representing hidden patterns in preferences.

Algorithms:

Singular Value Decomposition (SVD) Alternating Least Squares (ALS)

Benefit:

Captures complex relationships between users and preferences beyond simple correlations.

Content-Based Filtering

This algorithm focuses on matching users based on the attributes of their taste profiles and dining preferences.

How It Works:

Uses the attributes of a user's taste profile (e.g., preference for spicy, salty, sweet dishes, or dietary restrictions).

Matches users with similar attributes and preferences.

Implementation:

Represent user preferences as vectors and use cosine similarity or Euclidean distance to find the closest matches.

Clustering algorithms group users into clusters based on their taste profiles and preferences.

K-Means Clustering:

Clustering Algorithms

Groups users into clusters based on their taste preferences. Users in the same cluster are matched as TasteBuddies.

Hierarchical Clustering:

Creates a hierarchy of user groups based on their preferences, allowing for finer granularity in matches.

DBSCAN:

Groups users with dense taste similarity while ignoring outliers.

Implementation:

Use user profile data as input features for clustering. Cluster users and recommend TasteBuddies within the same group.

Graph-Based Algorithms

Graph-based approaches model user relationships and interactions as a network.

How It Works:

Represent users and their interactions (e.g., shared preferences or mutual likes) as a graph. Apply graph algorithms to identify similar users or clusters.

Algorithms:

PageRank:

Identifies influential users (Super TasteBuddies) based on their connections within the graph. **Community Detection:**

Identifies tightly connected groups of users with shared preferences.

Implementation:

Use libraries like NetworkX (Python) to build and analyze user graphs.

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