Restaurant Reccomendation System

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PROTOTYPE SELECTION

ABSTRACT

A restaurant recommender system is a tool that leverages machine learning algorithms to suggest restaurants based on user preferences. The system is designed to provide personalized recommendations that match the user's tastes and preferences while also taking into account other factors such as availability, popularity, and proximity. The system consists of several key components, including data collection, preprocessing, the recommendation engine, and the user interface. Data collection involves gathering data about both the users and the restaurants from a variety of sources such as online reviews, social media posts, and user surveys. Preprocessing is the step that cleans, filters, and organizes the data to remove any noise or irrelevant information. The recommendation engine is the core component of the system responsible for analyzing the user's data and generating personalized recommendations. The engine may use a variety of algorithms such as collaborative filtering, content-based filtering, or hybrid approaches. The user interface is the part of the system that interacts with the user, presenting them with recommendations and allowing them to provide feedback and refine their preferences. This provides users with a more personalized and relevant dining experience, which can lead to increased customer satisfaction, positive reviews, and word-of-mouth referrals. Overall, restaurant recommender systems have numerous benefits, such as personalization, time-saving, increased revenue, and improved customer satisfaction. However, there are also limitations to consider, such as the quality and quantity of data available, over-reliance on data, and lack of transparency. Despite these limitations, restaurant recommender systems are a valuable addition to the dining industry.

Key Components of Restaurant Recommendation Systems

- 1. Data Collection: Restaurant recommendation systems rely on a vast amount of data, including restaurant details, user reviews, ratings, and user preferences. This data is collected from various sources such as online review platforms, reservation systems, and user feedback.
- 2. User Profiling: User profiling involves capturing and analyzing user preferences, dining history, demographics, and other relevant information.

- By understanding user preferences and behaviour patterns, recommendation systems can deliver personalized recommendations that match individual tastes and requirements.
- 3. Recommendation Algorithms: Advanced recommendation algorithms play a crucial role in restaurant recommendation systems. Collaborative filtering, content-based filtering, and hybrid approaches are commonly used algorithms. Collaborative filtering analyzes user behaviour and preferences to recommend restaurants based on similar users' choices. Content-based filtering focuses on restaurant attributes such as cuisine, location, and ambience to suggest relevant options. Hybrid approaches combine multiple algorithms to improve recommendation accuracy.
- 4. Feature Engineering: Feature engineering involves extracting meaningful information from raw data to create relevant features for the recommendation system. This may include transforming and normalizing data, generating features from text data (e.g., using TF-IDF), and incorporating contextual information like time and location.
- 5. Evaluation Metrics: Evaluation metrics are used to measure the performance and effectiveness of restaurant recommendation systems. Common metrics include precision, recall, F1 score, and Mean Average Precision (MAP). These metrics help assess how accurately the system predicts user preferences and how well it provides personalized recommendations.
- 6. Real-time Updates: Restaurant recommendation systems should adapt to changing user preferences and new restaurant offerings. Regular updates are required to incorporate new data, refine algorithms, and improve the accuracy of recommendations. Systems can leverage techniques such as online learning and incremental updates to keep up with evolving user preferences.
- 7. User Feedback and Ratings: Collecting user feedback and ratings on recommended restaurants is valuable for system refinement. Feedback helps the system understand the quality and relevance of recommendations and allows for continuous improvement of the recommendation algorithm.
- 8. Ethical Considerations: Restaurant recommendation systems should prioritize user privacy, transparency, and fairness. User data should be handled securely, and clear explanations of the recommendation process

should be provided to users. Avoiding biases in recommendations is crucial to ensure fair and inclusive recommendations for all users.

Some of the main restaurant recommendation system prototypes are:

- Collaborative Filtering: Collaborative filtering is a popular technique used in recommendation systems. It analyzes user behaviour and preferences to make recommendations. In the context of restaurant recommendations, it looks at patterns such as user ratings, reviews, and dining history to suggest similar restaurants that other users with similar tastes have enjoyed.
- 2. Content-Based Filtering: Content-based filtering focuses on the characteristics and attributes of restaurants. It considers factors such as cuisine type, location, price range, ambience, and specific features to recommend restaurants that align with a user's preferences. By analyzing user-profiles and historical data, the system can match the user's stated preferences with the restaurant attributes.
- Hybrid Approaches: Many recommendation systems combine collaborative filtering and content-based filtering techniques to benefit from their respective strengths. Hybrid approaches leverage both user preferences and restaurant attributes to provide more accurate and diverse recommendations.
- 4. Machine Learning and Natural Language Processing (NLP): Machine learning and NLP techniques are often used in recommendation systems to process and analyze large amounts of data, such as user reviews, ratings, and textual information about restaurants. These techniques can extract meaningful insights, identify patterns, and improve the accuracy of recommendations.
- 5. Location-Based Recommendations: Location-based recommendations take into account the user's geographical location to suggest nearby restaurants. By utilizing location data, such as GPS coordinates or user-provided addresses, the system can recommend restaurants within a certain radius or in specific neighbourhoods.
- Personalization and User Profiles: User profiles play a crucial role in restaurant recommendation systems. The system collects information about users' preferences, dietary restrictions, previous dining experiences, and other relevant data to create personalized recommendations. This

- personalization helps tailor recommendations to each user's specific tastes and needs.
- 7. Integration with External Data Sources: Restaurant recommendation systems often integrate with external data sources, such as online review platforms, social media, or restaurant APIs, to gather up-to-date information about restaurants. This integration ensures that the system has access to accurate and timely data for making recommendations.

For our project, we decided to move on with content-based filtering.

Problem Statement

The restaurant industry is highly competitive, with numerous options available to customers in any given location. As a result, it can be challenging for customers to find restaurants that match their preferences and satisfy their needs. Additionally, the time and effort required to search for and evaluate restaurant options can be a significant barrier to making a decision and can lead to frustration and indecisiveness.

Market/Customer/Business Need Assessment Market Assessment

The restaurant industry is a large and growing market, with millions of restaurants worldwide and billions of dollars in revenue generated each year. Within this market, there is a growing demand for personalized experiences and convenience, which can be met through the development of restaurant recommender systems. These systems can provide customers with personalized recommendations based on their preferences, saving them time and effort in the decision-making process.

Customer Assessment

Understanding the needs, preferences, and behaviours of customers can help ensure that the system provides personalized and effective recommendations. One key aspect of customer assessment is gathering data on customer preferences and behaviour. This can be done through surveys, online reviews, social media, and other sources. The data can then be used to develop algorithms that analyze the customer's data and generate personalized recommendations. Another important consideration is the user interface of the system. The system should be user-friendly and intuitive, allowing customers to

easily navigate and provide feedback on their preferences. The system should also provide transparency in the recommendation process, allowing customers to understand why certain recommendations were made. Privacy is also an important consideration in customer assessment. Customers should have control over their data and be able to choose what information is shared with the system. Clear communication about how the data is being used and protected is crucial to building trust with customers.

Business Assessment

It involves evaluating the potential impact of the system on the restaurant industry and identifying opportunities for revenue generation and growth. One key benefit of a restaurant recommender system is increased customer satisfaction and loyalty. By providing personalized recommendations, the system can help customers find restaurants that meet their preferences and needs, leading to a better dining experience. This can result in positive reviews, increased customer retention, and word-of-mouth referrals. Another potential benefit is increased revenue. A restaurant recommender system can promote specific menu items or promotions to customers, leading to increased sales. The system can also be used to target specific customer segments or geographic locations. providing opportunities for targeted marketing and customer acquisition. Additionally, the restaurant recommender system can provide valuable data insights to restaurants. By analyzing customer data, restaurants can gain insights into customer preferences and behaviours, allowing them to make data-driven decisions about menu items, pricing, and marketing strategies. However, there are also potential challenges and limitations to consider. Developing an effective recommender system requires significant investment in data collection, algorithm development, and system implementation. There may also be concerns around data privacy and security, which need to be addressed to build trust with customers.

Restaurant Recommendation Final Prototype

Content-based filtering relies on analyzing the characteristics and attributes of restaurants, such as cuisine type, location, price range, ambience, and specific features. It doesn't require extensive user data or complex algorithms to make recommendations. Instead, it focuses on matching the user's stated preferences with the restaurant attributes.

Implementing a content-based filtering prototype involves the following steps:

- 1. Data collection: Gather data about restaurants, including their attributes such as cuisine type, location, price range, and features. This data can be obtained from publicly available sources, restaurant APIs, or web scraping.
- 2. User profiling: Allow users to create profiles and specify their preferences for cuisine type, location, price range, and other relevant attributes. This information will serve as the basis for making personalized recommendations.
- 3. Similarity calculation: Compute the similarity between the user's profile and the attributes of restaurants. Various similarity metrics, such as cosine similarity or Jaccard similarity, can be used to determine how closely the user's preferences align with each restaurant's attributes.
- 4. Ranking and recommendation generation: Rank the restaurants based on their similarity to the user's preferences and present the top-ranked options as recommendations. This can be done by assigning weights to different attributes and calculating an overall score for each restaurant.
- 5. User feedback and iteration: Incorporate user feedback to continuously improve the recommendations. Users can provide ratings or indicate whether they liked or disliked the recommended restaurants. This feedback can be used to refine the prototype and enhance the accuracy of future recommendations.

Data Collection

The secondary data was collected from Kaggle and data cleaning was done using python.

	Nan	ne	Links	Cost		Colle	ections		Cuisines		Timings
0	Beyond Flavou	nrs https://www.zo	mato.com/hyderabad/beyond- flavou	800	Foo Restaurants in	d Hygiene Hyderab			inental, Kebab, pean, South I		3:30pm, 6:30pm 0pm (Mon-Sun)
1	Paradi	se https://www.zon	nato.com/hyderabad/paradise- gach	800	Нус	lerabad's	Hottest	Biryani, North I	ndian, Chinese		11 AM to 11 PM
2	Flecha	zo https://www.zon	nato.com/hyderabad/flechazo- gach	1,300	Great Buff		rabad's Hottest		erranean, North ndian, Desserts	11:30 AM t	o 4:30 PM, 6:30 PM to 11 PM
3	Shah Ghouse Ho & Restaura		.zomato.com/hyderabad/shah- ghouse-h	800	Late N	light Rest	aurants	Biryani, North Ir Se	ndian, Chinese, eafood, Bever	1	2 Noon to 2 AM
4	Over The Moon Bre Compa		mato.com/hyderabad/over-the- moon	1,200	Best Ba Hygiene Ra	ars & Pub ated Rest			ntinental, North Chinese, Med		1pm (Mon, Tue, u, Sun), 12no
	Restaurant	Reviewer			Review	Rating		Metadata	Time	Pictures	
0	Beyond Flavours	Rusha Chakraborty	The ambience was good, foo	d was q	uite good . h	5	1 Revie	ew , 2 Followers	5/25/2019 15:54	0	
1	Beyond Flavours	Anusha Tirumalaneedi	Ambience is too good for a	pleasar	nt evening. S	5	3 Review	ws , 2 Followers	5/25/2019 14:20	0	
2	Beyond Flavours	Ashok Shekhawat	A must try great food gre	eat ambi	ence. Thnx f	5	2 Review	ws , 3 Followers	5/24/2019 22:54	0	
3	Beyond Flavours	Swapnil Sarkar	Soumen das and Arun was a	great gu	y. Only beca	5	1 Rev	iew , 1 Follower	5/24/2019 22:11	0	
4	Beyond Flavours	Dileep	Food is good.we ordered Ko	di drums	sticks and ba	5	3 Review	ws , 2 Followers	5/24/2019 21:37	0	

The two datasets were joined to get better data so that we can build a recommender system.

Analysis of restaurant recommendation system

https://github.com/Himanshu091002/Restaurant-Recommender-System-with-Machine-Learning/blob/main/Restaurant%20Recommender%20System.ipynb

The content-based recommender system uses various libraries such as

- re For pattern matching and text manipulation
 stopwords To filter out stopwords from your text data during preprocessing or text analysis tasks.
 - linear_kernel for computing pairwise similarities or distances between samples.
 - CountVectorizer It is used for text feature extraction based on the bag-of-words approach.
 - TfidfVectorizer It is used for text feature extraction based on the Term Frequency-Inverse Document Frequency (TF-IDF) approach. It converts a collection of text documents into a matrix of TF-IDF features, which reflects the importance of each word in the text data.

```
def recommend(name, cosine_similarities = cosine_similarities):
   # Create a list to put top 10 restaurants
  recommend_restaurant = []
   # Find the index of the hotel entered
  idx = indices[indices == name].index[0]
  # Find the restaurants with a similar cosine-sim value and order them from bigges number
  score_series = pd.Series(cosine_similarities[idx]).sort_values(ascending=False)
  # Extract top 30 restaurant indexes with a similar cosine-sim value
  top30_indexes = list(score_series.iloc[0:31].index)
  # Names of the top 30 restaurants
  for each in top30_indexes:
       recommend_restaurant.append(list(df.index)[each])
   # Creating the new data set to show similar restaurants
  df_new = pd.DataFrame(columns=['Cuisines', 'Mean Rating', 'Cost', 'Timings'])
   # Create the top 30 similar restaurants with some of their columns
  for each in recommend_restaurant:
       df_new = df_new.append(pd.DataFrame(df[['Cuisines','Mean Rating', 'Cost', 'Timings']][df.index == each].sample()))
  # Drop the same named restaurants and sort only the top 10 by the highest rating
df_new = df_new.drop_duplicates(subset=['Cuisines','Mean Rating', 'Cost'], keep=False)
  df_new = df_new.sort_values(by='Mean Rating', ascending=False).head(10)
   print('TOP %s RESTAURANTS LIKE %s WITH SIMILAR REVIEWS: ' % (str(len(df_new)), name))
  return df new
```

Words are weighed and quantified using the TF-IDF algorithm. In order to apply mathematics in our recommender system, we represent each word (or pair of words, etc.) with a number. Simply said, the rarer and more significant the term, and vice versa, the greater the TF*IDF score (weight).

Regardless of the size of the documents, cosine similarity is a statistic used to determine how similar they are.

To Calculate Cosine Similarity

similarity(A,B) =
$$\frac{A \cdot B}{\|A\| \times \|B\|} = \frac{\sum_{i=1}^{n} A_i \times B_i}{\sqrt{\sum_{i=1}^{n} A_i^2} \times \sqrt{\sum_{i=1}^{n} B_i^2}}$$

Once the recommender model is built, we save the model to a .pkl file extension or to a pickle file so that we can use it to build a model in our recommender application.

Streamlit library was used to build the content-based recommender model. Necessary libraries were imported first.

```
app.py > ...
1 v import streamlit as st
2 import pandas as pd
3 from sklearn.feature_extraction.text import TfidfVectorizer
4 from sklearn.metrics.pairwise import linear_kernel
5 import pickle
6 import base64
```

To make the web app more appealing, CSS elements were also included.

```
st.markdown(
f"""

style>
.stApp {{
    background-image: url(data:image/{"jpg"};base64,{encoded_string.decode()});
    background-size: cover;
    background-position: center;
    background-position: center;
    font-weight: bold !important;
    color: white;
    font-family: 'Arial', sans-serif;

// syle>
""",
unsafe_allow_html=True

// add_bg_from_local(r"D:\Feynn Labs\golden-cutlery-with-textile-plate-dark-background-top-view.jpg")
```

Restaurant Recommendation App - Web view

	urant name:	mmendation App
10 Downing Street		
Top 10 restau	ırants similar to	10 Downing Street:
		Mean Rating Cost Timings
	north indian continental italian	3.3300 1500 12noon to 12midnight (Mon-Sun)
Olive Garden	north indian chinese continental biryani	3.1900 700 12 Noon to 3:30 PM, 7 PM to 11 PM
	north indian biryani chinese	3.1600 500 11 AM to 11 PM
Being Hungry	north indian chinese	3.0300 450 12 Noon to 3 PM, 7 PM to 11:30 PM
Absolute Sizzlers	continental american chinese	2.9700 750 11:30 AM to 1 AM
Gal Punjab Di	north indian continental	2.9500 800 11:55 AM to 4 PM, 7 PM to 11:15 PM
13 Dhaba	north indian	2.7400 450 12:30 PM to 10 PM (Tue-Sun), Mon Closed
Tiki Shack	continental	2.7400 1000 12Noon to 11PM (Mon- Thu),12Noon to 11:30PM (Fri-Sun)
	andhra biryani hyderabadi north indian	2.7300 500 12 Noon to 4 PM, 7 PM to 11 PM
Shah Ghouse Spl Shawarma	lebanese	2.6600 300 12 Noon to 12 Midnight

GitHub 𝚱(Click here)

Restaurant Recommendation App

Enter a restaurant name:

10 Downing Street

Top 10 restaurants similar to 10 Downing Street:

	Cuisines	Mean Rating	Cost	Timings
The Tilt Bar Republic	north indian continental italian	3.3300	1500	12noon to 12midnight (Mon-Sun)
Olive Garden	north indian chinese continental biryani	3.1900	700	12 Noon to 3:30 PM, 7 PM to 11 PM
Biryanis And More	north indian biryani chinese	3.1600	500	11 AM to 11 PM
Being Hungry	north indian chinese	3.0300	450	12 Noon to 3 PM, 7 PM to 11:30 PM
Absolute Sizzlers	continental american chinese	2.9700	750	11:30 AM to 1 AM
Gal Punjab Di	north indian continental	2.9500	800	11:55 AM to 4 PM, 7 PM to 11:15 PM
13 Dhaba	north indian	2.7400	450	12:30 PM to 10 PM (Tue-Sun), Mon Closed
Tiki Shack	continental	2.7400	1000	12Noon to 11PM (Mon- Thu),12Noon to 11:30PM (Fri-Sun)
Kritunga Restaurant	andhra biryani hyderabadi north indian	2.7300	500	12 Noon to 4 PM, 7 PM to 11 PM
Shah Ghouse Spl Shawarma	lebanese	2.6600	300	12 Noon to 12 Midnight

BUSINESS MODELLING

A commission-based business model for a restaurant recommender system can be an effective way to generate revenue. Here's how it could work:

- Partner with Restaurants: Establish partnerships with restaurants and food establishments that are interested in attracting new customers. Negotiate commission rates for each successful referral or transaction made through the recommendation system.
- 2. **User Referrals:** Encourage users to discover and try out recommended restaurants through the platform. When a user makes a reservation or places an order at a partner restaurant based on the system's recommendation, the system earns a commission from the restaurant.
- 3. Tracking and Attribution: Implement a tracking system to monitor and attribute referrals accurately. This can be achieved through unique referral codes, tracking links, or customized landing pages that identify the source of the referral. This ensures that the system receives proper credit for generating the customer.
- 4. Transparent Commission Structure: Clearly define the commission structure for partner restaurants. It can be a percentage of the total bill amount or a fixed amount per transaction. Ensure that the terms and conditions regarding commission payments are transparent and communicated to the partners upfront.
- Payment Processing: Set up a streamlined payment processing system to handle commission payments. Regularly reconcile referral data with the partner restaurants and initiate timely commission pay-outs based on the agreed-upon terms.
- 6. **Continuous Relationship Building:** Maintain strong relationships with partner restaurants. Provide them with regular reports on referral performance, customer feedback, and other relevant data to demonstrate the

value generated by the recommendation system. This can help build trust and foster long-term partnerships.

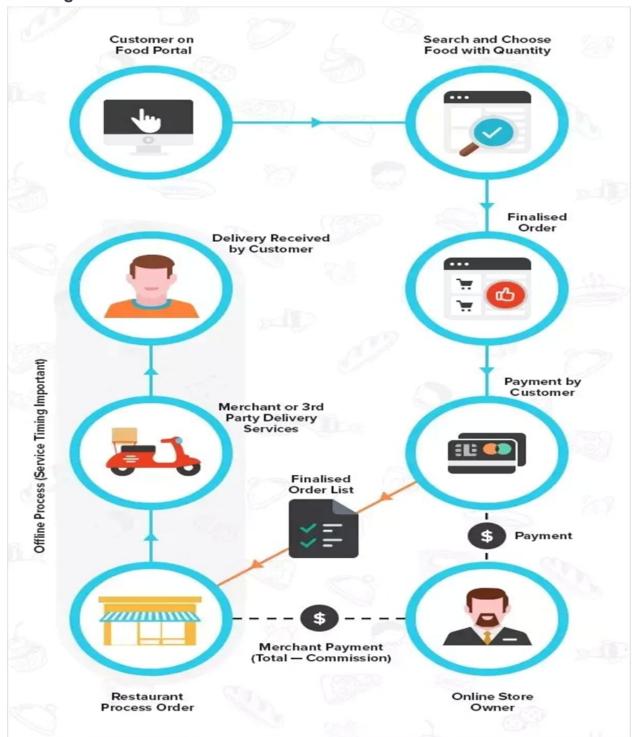
- 7. Expansion and Scalability: As the recommendation system gains traction and more restaurants partner with the platform, consider expanding to new geographical areas or offering additional services like food delivery. Scaling the business can lead to increased revenue opportunities and a wider customer base.
- 8. Quality Control and User Experience: It's crucial to maintain the integrity of the recommendation system by ensuring the restaurants recommended are of high quality and meet user expectations. Regularly monitor user feedback, ratings, and reviews to ensure customer satisfaction and avoid negative experiences that could harm the reputation of the platform remember to adapt and iterate your business model based on the market dynamics, customer preferences, and the specific needs of your restaurant partners. Regularly assess the effectiveness of the commission-based model and make adjustments as necessary to optimize revenue generation and sustain the recommender system's growth.

It is essential to be flexible and responsive to market dynamics, customer preferences, and the unique requirements of restaurant partners. Continuously evaluate the performance of your commission-based business model and make necessary modifications to enhance revenue generation and ensure the ongoing growth of your restaurant recommender system.

PRODUCT DESCRIPTION

The restaurant recommendation website is designed to connect users with the perfect dining experiences while generating revenue through successful referrals. With a user-friendly interface, we help individuals discover and explore a wide range of restaurants based on their preferences, location, and culinary interests. Through strategic partnerships with restaurants, we offer a curated selection of establishments that have agreed to collaborate with our platform. As users browse through our recommendations, make reservations, or place orders at partner restaurants, we earn a commission for each successful transaction.

Working model Flowchart



MARKET ANALYSIS

The market for restaurant recommender systems has experienced significant growth in recent years. As consumers increasingly rely on online platforms and

mobile apps to make dining decisions, the demand for personalized restaurant recommendations has surged.

Key factors driving the market include the widespread use of smartphones and internet platforms, the need for personalized recommendations, and the competitive landscape. Emerging technologies like AI and ML are enhancing the accuracy and relevance of recommendations. It is essential to consider regional differences, regulatory compliance, and staying ahead of technological advancements to succeed in this market. Overall, the restaurant recommender system market offers significant growth potential by catering to consumers' preferences and helping restaurants attract new customers.

OPERATING PLAN

- 1. Develop an intuitive platform for users to access restaurant recommendations, menus, reviews, and reservations.
- 2. Collect and analyse user data to generate personalized recommendations.
- 3. Establish partnerships with a variety of restaurants, negotiating commission rates or revenue-sharing agreements.
- 4. Streamline the onboarding process for partner restaurants and provide training and support.
- 5. Implementing marketing strategies to attract and engage users, utilizing digital channels and user reviews.
- 6. Ensure recommended restaurants meet high-quality standards through regular monitoring and user feedback.
- 7. Provide responsive customer support channels for user queries and technical issues.
- 8. Generate revenue through commission-based models, referral tracking, and exploring additional revenue streams.
- 9. Continuously update and optimize the system based on user feedback and emerging technologies.
- 10. Conduct regular market analysis to identify market opportunities.

MARKETING PLAN

- 1. Identify the target audience and their preference such as food enthusiasts, travellers, or locals seeking recommendations.
- 2. Clearly communicate the unique benefits of the recommender system.
- 3. Establish an intuitive website with a compelling design and intuitive user experience.

- 4. Create engaging content about dining experiences, food trends, and partner restaurant features.
- 5. Leverage social media for community engagement and influencer collaborations.
- 6. Seek partnerships with local businesses and travel agencies for cross-promotion.

FINANCIAL MODELLING

A commission-based financial model for a restaurant recommendation system can be an effective way to generate revenue. Here's how it could work:

- 1. **Revenue Projections:** To provide revenue projections for the restaurant recommendation system, we'll need to make some assumptions. Let's assume the following:
 - 1. Average commission rate per successful referral: \$10 (this could be a percentage of the total bill amount, but for simplicity, let's assume a fixed amount).
 - 2. **An average number of monthly successful referrals:** 500 (based on market research and projections).
 - 3. **Payment processing fees:** 2% of the revenue from successful referrals.

we can calculate the revenue projections:

- Monthly revenue from successful referrals: Average commission rate per successful referral * Average number of monthly successful referrals = \$10 * 500 = \$5,000
- Cost of Sales: Payment processing fees = 2% of the revenue from successful referrals. The monthly cost of sales = 2% * \$5,000 = \$100 Based on these projections, the monthly revenue from successful referrals would be \$5,000, and the monthly cost of sales would be \$100.

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2. Cost of Sales: The cost of sales in the restaurant recommendation system model typically includes payment processing fees associated with the revenue generated from successful referrals. Here's a more detailed breakdown of the cost of sales:

a. Payment Processing Fees:

- i. When a user makes a reservation or places an order at a partner based restaurant on the system's recommendation, the system earns a This commission. commission is percentage of the total bill amount or a fixed amount per transaction, as agreed upon with the partner restaurants.
- ii. Payment processing providers, such as credit card processors or online payment gateways, usually charge a fee for processing these transactions.
- iii. The cost of sales for the recommendation system includes these payment processing fees, which are typically a percentage of the revenue earned from successful referrals.
- iv. The specific percentage may vary based on the payment processor and the negotiated terms.

For example:

- Average commission rate per successful referral: 10%
- - Monthly revenue from successful referrals: \$10,000
- - Payment processing fees: 2%

In this case, the calculation for the cost of sales would be:

Monthly cost of sales = (Monthly revenue from successful referrals) * (Payment processing fees/100)

= \$10,000 * (2/100)

= \$200

This means that the recommendation system incurs a cost of \$200 per month for payment processing fees.

3. Operating Expenses:

a. Development and Maintenance Costs:

- i. Website development: Including design, front-end and back-end development, hosting, and domain registration.
- ii. Platform maintenance and updates: Regular maintenance, bug fixes, and updates to ensure the platform operates smoothly.

b. Marketing and Advertising Expenses:

- i. Digital marketing campaigns: Including paid advertising on platforms like Google Ads or social media platforms to increase user acquisition and engagement.
- ii. Content creation: Developing engaging content such as blog posts, articles, videos, and social media posts related to dining experiences, food trends, and partner restaurant features.

iii. Social media management: Managing and maintaining active social media accounts to engage with the community, respond to user inquiries, and share updates about new restaurants and offers.

c. Partnership Management Costs:

- i. Personnel: Allocating resources to manage partnerships, negotiate commission rates, onboard new restaurants, and maintain ongoing relationships.
- ii. Relationship building: Building and nurturing relationships with partner restaurants through regular communication, meetings, and providing them with performance reports and feedback.

d. Customer Support Costs:

- i. Personnel: Hiring customer support representatives to handle user queries, technical issues, and provide assistance with reservations or orders.
- ii. Infrastructure: Providing necessary tools and systems to manage customer support channels such as email, live chat, or phone support.

e. Other Operating Expenses:

i. Legal and Compliance: Seeking legal advice, ensuring compliance with local regulations, and maintaining necessary licenses and permits.

- ii. Accounting: Engaging accounting services for bookkeeping, financial statements, and tax filings.
- iii. Office Expenses: Rent, utilities, office supplies, and other miscellaneous expenses related to running the business.

4. Profit/Loss Calculation:

a. Gross Profit Calculation:

- i. Gross profit is the revenue generated from successful referrals minus the cost of sales. The cost of sales in this case refers to the payment processing fees associated with the commission-based model.
- ii. Gross profit = Monthly revenue from successful referrals Monthly cost of sales

b. Net Profit Calculation:

- i. Net profit is the gross profit minus the total monthly operating expenses. It represents the overall profitability of the business after accounting for all costs and expenses.
- ii. Net profit = Gross profit -Total monthly operating expenses
- iii. The total monthly operating expenses include development and maintenance costs, marketing and

advertising expenses, partnership management costs, customer support costs, and other operating expenses.

c. Cash Flow Considerations:

- i. Cash flow is an important aspect of financial modelling as it represents the actual inflow and outflow of cash in the business. It helps determine the sustainability and liquidity of the operation.
- ii. Initial investment refers to the upfront costs required to set up the business, including equipment, website development, and initial marketing expenses.
- iii. Monthly cash inflow is the revenue generated from successful referrals on a monthly basis.
- iv. Monthly cash outflow is the total monthly operating expenses required to run the business.
- v. Net cash flow is the difference between monthly cash inflow and monthly cash outflow, representing the net amount of cash generated or consumed each month.
- vi. Cumulative cash flow is the running total of net cash flow over time, taking into account the initial investment.

d. Break-even Analysis:

i. The break-even point is the point at which the revenue generated

from successful referrals equals the total monthly operating expenses. It helps determine the minimum number of successful referrals needed to cover the costs and achieve profitability.

ii. Breakeven referrals = Monthly cost of sales / Average commission rate per successful referral

e. Financial Ratios and Metrics:

- i. Return on Investment (ROI): Net profit / Initial investment * 100
- ii. ROI indicates the profitability of the investment relative to its cost.
- iii. Payback Period: Initial investment / Monthly net cash flow
- iv. The payback period represents the time it takes to recover the initial investment based on the monthly net cash flow.

5. Cash Flow Considerations:

a. Initial Investment:

- i. Determine the initial investment required to set up the platform, develop the website, and cover initial marketing expenses.
- ii. Consider costs such as software development, hosting, domain

registration, branding, and marketing campaigns.

iii. Calculate the total amount needed for the initial investment, denoted as \$F.

b. Monthly Cash Inflow:

- i. The primary source of cash inflow is the revenue generated from successful referrals made through the platform.
- ii. Calculate the average commission rate per successful referral (e.g., \$X) and estimate the average number of monthly successful referrals (e.g., Y).
- iii. Multiply the commission rate by the number of successful referrals to determine the monthly revenue from successful referrals.

c. Monthly Cash Outflow:

- i. Consider the various operating expenses required to run the restaurant recommendation system on a monthly basis.
- ii. Sum up the costs associated with development and maintenance (\$A), marketing and advertising (\$B), partnership management (\$C), customer support (\$D), and other operating expenses (\$E).

iii. Calculate the total monthly operating expenses, denoted as the sum of \$A, \$B, \$C, \$D, and \$E.

d. Net Cash Flow:

- i. Subtract the total monthly operating expenses from the monthly revenue from successful referrals to calculate the net cash flow.
- ii. Net cash flow = Monthly revenue from successful referrals Total monthly operating expenses.

e. Cumulative Cash Flow:

- i. Track the cumulative cash flow over time to understand the financial progress of the business.
- ii. Start with the initial investment (\$F) and add the net cash flow from each subsequent month.
- iii. Cumulative cash flow = Net cash flow + Initial investment.
- 6. **Break-even Analysis:** Let's calculate the break-even point for the restaurant recommendation system based on the information provided:

Assumptions:

- Average commission rate per successful referral: \$X
- Monthly cost of sales: (X * Y) * (Z/100)
- Monthly operating expenses: Total monthly operating expenses

 Break-even referrals = Monthly cost of sales / Average commission rate per successful referral

For example, if the monthly cost of sales is \$5,000 and the average commission rate per successful referral is 10% (0.1), the break-even point can be calculated as follows:

Break-even referrals = \$5,000 / 0.1 = 50,000 referrals

This means that in order to cover the monthly operating expenses, the restaurant recommendation system needs to generate at least 50,000 successful referrals per month.

• **Break-even revenue** = Break-even referrals * Average transaction value

For example, if the average transaction value is \$50, the break-even revenue can be calculated as:

Break-even revenue = 50,000 referrals * \$50 = \$2,500,000

Therefore, the restaurant recommendation system needs to generate at least \$2,500,000 in monthly revenue to cover the monthly operating expenses and break even.

7. Financial Ratios and Metrics:

1. Gross Profit Margin:

- Formula: (Monthly revenue from successful referrals Monthly cost of sales) / Monthly revenue from successful referrals * 100
- 2. This ratio measures the profitability of each successful referral after accounting for the cost of generating that revenue.

2. Net Profit Margin:

- Formula: Net profit / Monthly revenue from successful referrals
 * 100
- 2. This ratio indicates the overall profitability of the business after considering all operating expenses.

3. Return on Investment (ROI):

- 1. Formula: Net profit / Initial investment * 100
- 2. ROI measures the profitability of the business relative to the initial investment made.

4. Cash Flow Break-even Point:

1. This metric determines the number of successful referrals needed to cover the monthly operating expenses. It indicates the minimum level of business activity required to reach profitability.

5. Payback Period:

- 1. Formula: Initial investment / Monthly net cash flow
- 2. The payback period represents the time it takes for the initial investment to be recovered from the monthly net cash flows generated by the business.

6. Customer Acquisition Cost (CAC):

- Formula: Total marketing and advertising expenses / Number of acquired customers
- 2. CAC measures the cost incurred to acquire each new customer for the recommendation system. It helps assess the efficiency of your marketing efforts.

7. Customer Lifetime Value (CLV):

- 1. Formula: Average revenue per successful referral * Average customer retention period
- 2. CLV estimates the total revenue expected to be generated from a single customer during their engagement with the recommendation system.

8. Return on Marketing Investment (ROMI):

- 1. Formula: (Net profit from marketing efforts Marketing expenses) / Marketing expenses * 100
- 2. ROMI measures the effectiveness of marketing campaigns by evaluating the return generated relative to the marketing expenses incurred.

9. Break-even Point in Time:

1. This metric determines the time it takes for the net cash flows to reach a breakeven point, where the total inflows equal the total outflows.