CIS550

PENNDASH

Restaurant Review Website



GROUP 33

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PROJECT DESCRIPTION

The goal of this group project is to develop PennDash, a user-friendly restaurant review platform focusing on the state of Pennsylvania (PA, USA). The website will enable users to explore restaurants based on various filters, such as type of cuisine, location, ratings, and enable them to bookmark and favorite their desired restaurants.

The platform will incorporate the following key functionalities:

- Detailed restaurant reviews and ratings along with their business information (e.g., Wi-Fi availability, outdoor seating, vehicle parking, etc.)
- Location mapping via Google Maps integration.
- Users can save their favorite restaurants and bookmark them.
- Secure login/logout functionality for an optimized user experience.
- Restaurant analytics from Yelp's public dataset will be shown to the user.







Diners in Pennsylvania lack access to platforms providing hyper-local, tailored restaurant recommendations based on preferences and amenities.

Existing platforms focus on national or global markets but fail to optimize for state-specific data, making it harder for users to find relevant dining options.

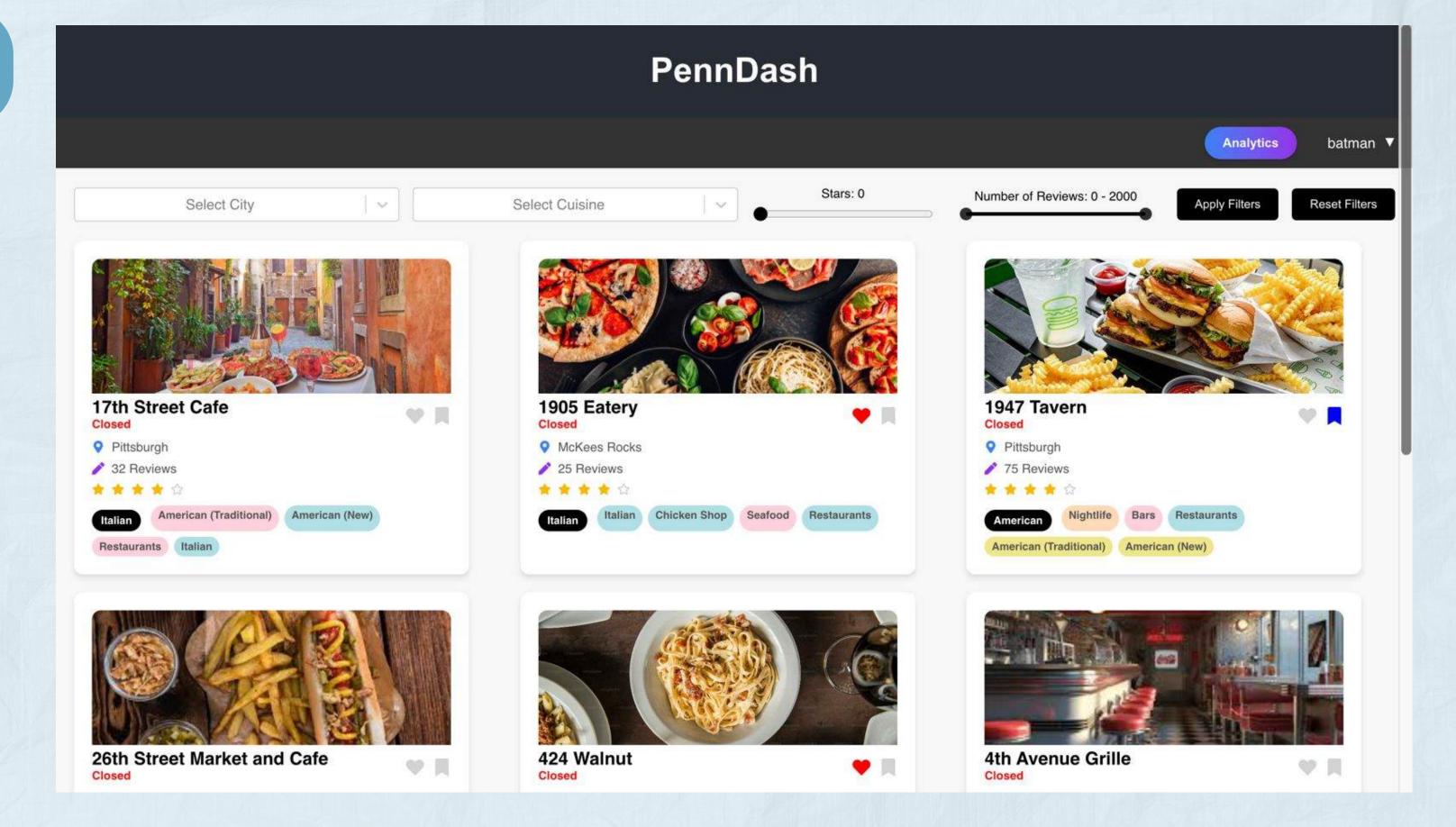
PennDash Solution:

A Pennsylvania-specific restaurant recommendation platform that,

- Combines user-friendly filters and trending insights.
- Can bookmark/favorite restaurants for future reference.
- Incorporates navigation and amenity-based reviews.

PREVIEW OF OUR WEBSITE







Dataset Files: https://www.yelp.com/dataset

yelp_academic_dataset_business.json

- Description: Contains business-level information (e.g., business ID, name, address, category, latitude, longitude, attributes like Wi-Fi or outdoor seating availability).
- Size: ~150,000 businesses across multiple regions, including Pennsylvania.
- Attributes Summary :
 - Mean rating: 3.8
 - Average review count: 120

yelp_academic_dataset_review.json

- Description: Includes detailed reviews such as review text, rating, and user ID.
- Size: 7 million reviews (~5 GB).
- Attributes Summary :
 - Average review length: 200 words
 - Star rating distribution: Normally distributed, with the mean around 3.8

yelp_academic_dataset_user.json

- ◆ Descript. : Details user profiles, including no. of reviews written and their friends.
- Size: 2 million user profiles.
- Attributes Summary :
 - Average useful votes per user: 4.2
 - Mean friend count: 7

yelp_academic_dataset_checkin.json

- Descript.: Contains the business ID and timestamps for when users checked in.
- ♦ Size: ~1.2 million check-ins.

yelp_academic_dataset_tip.json

- Descript.: Contains shorter, informal tips by users, offering concise restaurant feedback.
- ♦ Size : ~2 million tips.







Basic Descriptive Analysis of the Data

Restaurant Types:

 Includes diverse
 categories such as
 Bakeries, Japanese,

 Burger joints, Chinese
 restaurants, Thai,
 Vietnamese, Sushi Bars

 and Cafes.

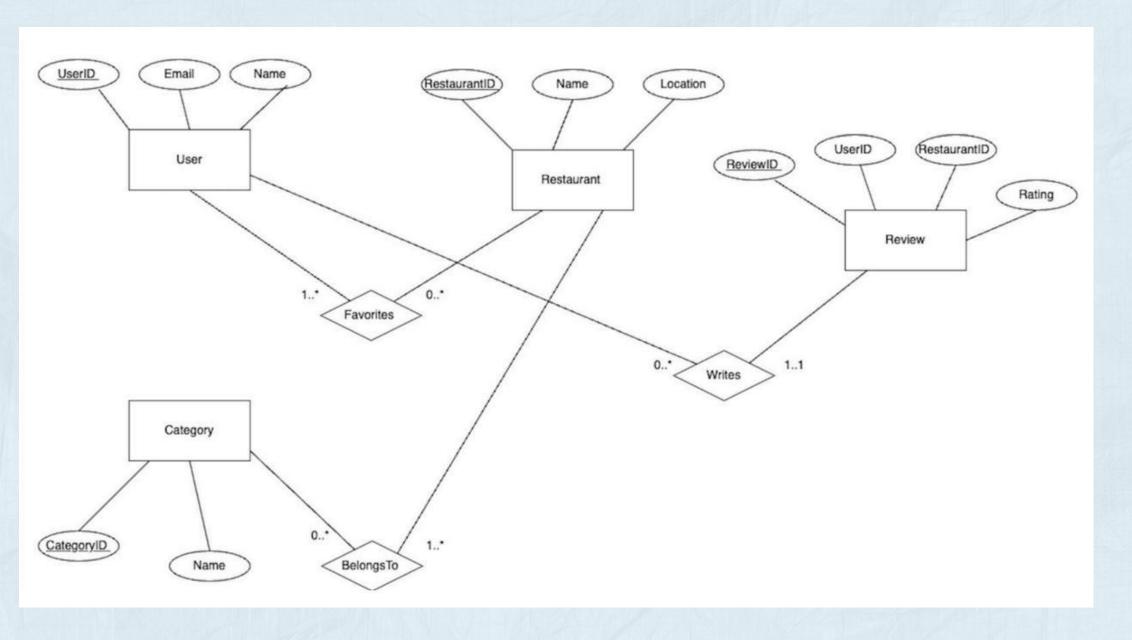
Rating Trends:
 Distribution shows a
 concentration around 3
 to 5 stars.

• Regional Breakdown:

The dataset includes
multiple US and
Canadian states, but we
will filter restaurants
within Pennsylvania.

SCHEMA DESIGN

The schema will include entities such as User, Restaurant, Review, Category, Check-in, and Tip, with relationships among users, restaurants through reviews, favorites, check-in.





NORMALIZATION

Normalization:

Table Name	Action	Reason	
businesses	Changed	Split location data into business_locations to remove redundancy.	
business_locations	New	Isolated location-specific data for reusability and integrity.	
business_attributes	Changed	Renamed columns for consistency; structure is unchanged.	
business_hours	Changed	Normalized days into rows for easier filtering and querying.	
checkin	Changed	Added checkin_id for unique identification and changed hour to TIME.	
users, user_favorites, user_stats, tips,	Unchanged	Already normalized and structured efficiently.	
user_reviews	Changed	Normalized text for better storage and querying.	



3NF PROOF

1. Business Table

Functional Dependencies:

- business_id → name, address, city, state, postal_code, latitude, longitude, stars, review_count, is_open
- business_id is the primary key.
- All non-prime attributes (e.g., name, stars, address) depend only on business_id.
- There are no transitive dependencies.

Conclusion: The Business table is in 3NF.

2. Users Table

Functional Dependencies:

- user_id → name, review_count, yelping_since, useful, funny, cool, fans, average_stars, compliments
- user_id is the primary key.
- All non-prime attributes depend solely on user_id.
- There are no transitive dependencies.

Conclusion: The Users table is in 3NF.

3. Review Table

Functional Dependencies:

- review_id → user_id, business_id, stars, useful, funny, cool, text, date
- review_id is the primary key.
- All non-prime attributes depend solely on review_id.
- user_id and business_id are foreign keys, maintaining relationships without introducing transitive dependencies.

Conclusion: The Review table is in 3NF.

4. Business_Category Table (Many-to-Many Relationship)

Functional Dependencies:

- Composite Key: (business_id, category_id)
- No non-prime attributes exist.
- Each row contains only foreign keys linking businesses and categories.

Conclusion: The Business_Category table is in 3NF.

5. Business_Attributes Table

Functional Dependencies:

- business_id → AcceptsInsurance, WiFi, RestaurantsTakeOut, GoodForKids, etc.
- business_id is the primary key.
- Each non-prime attribute depends solely on business_id.
- No transitive dependencies exist.

Conclusion: The Business_Attributes table is in 3NF.

SNF PROOF

6. Business_Hours Table

Functional Dependencies:

- Composite Key: (business_id, day_of_week) → opening_time, closing_time
- All non-prime attributes depend on the full composite key (business_id, day_of_week).
- There are no transitive dependencies.

Conclusion: The Business_Hours table is in 3NF.

7. Check-in Table

Functional Dependencies:

- checkin_id → business_id, date
- checkin_id is the primary key.
- Non-prime attributes depend solely on checkin_id.

Conclusion: The Check-in table is in 3NF.

9. Category Table

Functional Dependencies:

- category_id → category_name
- category_id is the primary key.
- category_name is atomic and depends only on category_id.

Conclusion: The Category table is in 3NF.

8. Tip Table

Functional Dependencies:

- tip_id → user_id, business_id, text, date, compliment_count
- tip_id is the primary key.
- All non-prime attributes depend solely on tip_id.
- user_id and business_id are foreign keys without introducing transitive dependencies.

Conclusion: The Tip table is in 3NF.

All tables in the PennDash database schema adhere to Third Normal Form (3NF).

- Each non-prime attribute depends only on the primary key.
- There are no partial dependencies or transitive dependencies.

EXAMPLE OF A COMPLEX QUERY

Top 5 restaurants reviewed most by users whose friends also reviewed them

```
const query = `
  WITH UserFriendsReviews AS (
    SELECT r.business_id, uf.user_id, COUNT(DISTINCT r.review_id) AS
review_count_by_friends
    FROM normalized_user_reviews r
    JOIN user_stats uf ON r.user_id = uf.user_id
    GROUP BY r.business_id, uf.user_id
   PopularWithFriends AS (
    SELECT ufr.business_id, SUM(ufr.review_count_by_friends) AS
total_reviews_by_friends
    FROM UserFriendsReviews ufr
    GROUP BY ufr.business_id
    ORDER BY total_reviews_by_friends DESC
    LIMIT 5
  SELECT b.name AS restaurant_name, bl.city as city,
pwf.total_reviews_by_friends
  FROM PopularWithFriends pwf
  JOIN normalized_businesses b ON pwf.business_id = b.business_id
  JOIN business_locations bl ON b.location_id = bl.location_id;
```



PERFORMANCE



	ORIGINAL	OPTIMIZED
COMPLEX QUERY 1	5.7sec	0.3sec
COMPLEX QUERY 2	9.3sec	0.5sec
COMPLEX QUERY 3	5.8sec	0.2sec
COMPLEX QUERY 4	5.5sec	0.1sec
COMPLEX QUERY 5	18.3sec	1.1sec

TECHNICAL CHALLENGES

Handling millions of records **Large Dataset** efficiently required database Management optimization. Complex queries (e.g., CTEs, joins) **Query Performance** needed optimization for low latency. Preprocessing Yelp data to resolve **Data Cleaning** missing values and inconsistencies. Balancing core features with time **Feature Prioritization** constraints (e.g., Google Maps).





Feel free to ask any questions!