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University of Mumbai

AMIGOS - Event Recommendation System

Submitted at the end of semester VIII in partial fulfilment of requirements

For the degree of

Bachelors in Technology

By

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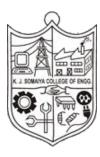
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Project Guide

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K. J. Somaiya College of Engineering, Mumbai-77
(Autonomous College Affiliated to University of Mumbai)
Batch 2017 -2021

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Certificate

This is to certify that the dissertation report entitled AMIGO - Event Recommendation System submitted by **Vikram Bhavsar**, **Vishvesh Pandey**, **Samiksha Gupta and Mohammed Anzal Shaikh** at the end of semester VIII of LY B. Tech is a bona fide record for partial fulfillment of requirements for the degree of Bachelors in Technology in Computer Engineering of University of Mumbai

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Place: Mumbai-77

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Certificate of Approval of Examiners

We certify that this dissertation report entitled **AMIGOS** - **Event Recommendation System is a bona fide record of project work done by Vikram Bhavsar, Vishvesh Pandey, Samiksha Gupta and Mohammed Anzal Shaikh** during semester VII. This project work is submitted at the end of semester VII in partial fulfilment of requirements for the degree of Bachelors in Technology in Computer Engineering of University of Mumbai.

Rohini Nair

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External/Internal Examiners

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Chapter 1: Introduction

1.1 Problem definition

To create a Personalized Event Recommendation System that recommends the events that are sorted according to the user based on his/her interests and also the events that are the most trending. This system would intend to save time for the user and also help the event organizers to boost their events and increase their reach.

1.2 Scope

- To suggest events to the users with as little interaction with the system as possible.
- To recommend events to the user which are highly personalized and based on the user's Interest
- To ensure only interesting events are recommended to the user.
- Also to recommend new trending events to the user that are out of the user's interest.
- Capturing implicit user behavior within the website and using that data to provide better recommendations

1.3 Hardware and software requirements

1.3.1 Software-

Python, Django, Google Chrome.

1.3.2 Hardware Requirements-(Minimum)

4GB-RAM.

I3-8th generation-Processor.

512 GB -Hard disk.

256GB -SSD.

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Chapter 2 : Literature Survey

In research, we have found 3 different applications which are used for recommending events.

2.1 BookMyShow website/app-

- It is majorly used for booking movie tickets and passes to various events. It uses
 the famous Amazon model that is based on a method called Collaborative Filtering.
- It takes items such as movies, books, and products that were rated highly by a set of users and recommends them to other users who also gave them high ratings. This method works well in domains where explicit ratings or implicit user actions can be gathered and analyzed.
- In our system, we focus on the implicit data such as what events are registered through the system, and use that to understand the interest of the user.

2.2 MagicPin app-

- MagicPin is more focused on inviting friends and getting discounts to visit different places based on the user's bills and locations visited by the user.
 Following is directly quoted from the app description of MagicPin on the play store-
- "Use the magic pin feed to find out what places & brands are trending among your friends & peers. Follow & interact with them by liking, commenting, & bookmarking their recommendations!".
- Hence, MagicPin is heavily based on recommending events and places based on what other people and friends are visiting. On the other hand, our system makes suggestions based on what's trending as well as the ones that the user is interested in.

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2.3 NearBuy App-

- The NearBuy app is similar to MagicPin in a way that It suggests various Shops and restaurants nearby based on the user's preference.
- There is a tab inside the app which provides suggestions based on what the user searches.
- But, these searches are from within the application whereas, in our case, the important information is taken from web history.

2.4 Papers referred-

- Users may often search for cars, so there must exist a spread of the keyword car
 based on time. From this, we can assume that since the user is searching for cars
 often then the user is interested in cars and we can recommend events related to
 cars to that user.
- This part of the module is heavily based on the paper Mining User Interests
 from Web History by Saurabh Kumar. The paper contains more details on
 the implementation of the above-mentioned.

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Chapter 3: Project Design

3.1 Proposed System model

3.1.1 Event Handler Side

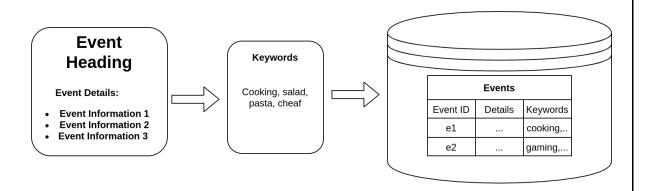


Figure (3.1.1): Keywords segregation

The above-proposed system depicts the Events Handler part where events details are specified as given above which displays heading, details, and event information. By using NLP's TF-IDF algorithm the important keywords are fetched or the data is cleaned and is stored in the database with the given above format. It will Extract information such as Nouns, and Named Entity using NLP.

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3.1.2 User's side

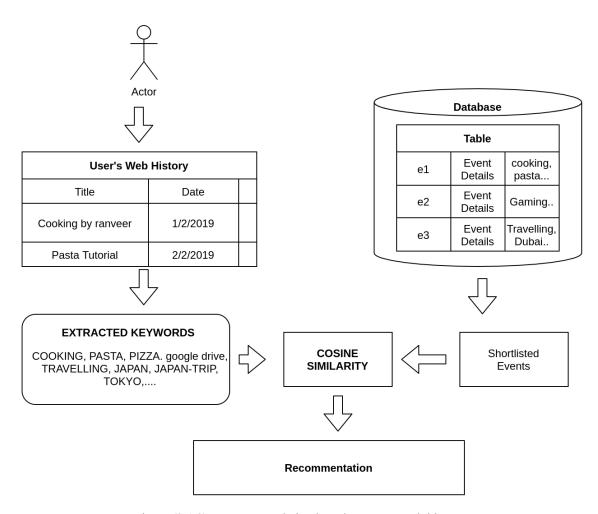


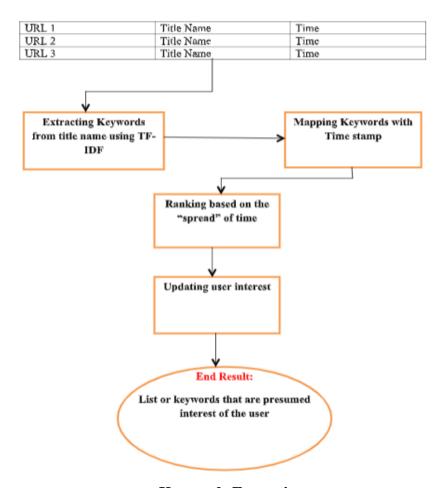
Figure (3.1.2): Recommendation based on user's web history

At the user's side, it will fetch the user's web history automatically using a web browser extension, on and based on the title and date from the history generated format it will apply the TF-IDF algorithm for extracting nouns, named entities, etc. which are our basic keywords required for our project. After fetching the important keywords from the user as well as from the event handler side our system will check using cosine similarity algorithm for identifying the same keywords and frequency of those words that occurred and similarly, it will recommend to the user based on his/her interest respectively as shown in the above figure.

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3.1.3 User's Web History Data Fetching

User's Web History



Keywords Extraction

Figure (3.1.3): User's web history data fetching and keyword extraction

The above-proposed model depicts the basic workflow of how the keywords will be extracted from the user's web history using an algorithm.

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3.2 Software Project Management Plan

3.2.1-Introduction

We aim to create a Personalized Event Recommendation System that recommends the events that are sorted according to the user based on his/her interests and also the events that are the most trending.

This system would intend to save time for the user and also helps the event organizers to boost their events and increase their reach and ensure that the user does not get recommended any events that he/she is not going to attend but, only the events according to his/her domain-interest along with a list of Trending Events. The system that we plan on developing is a hybrid recommendation system. This system will have two participants-Event Organizers and Users.

For Example: If a user on youtube is searching for Cake Baking videos it would be noted in the history. The system would then extract the title of that video and time and prepare a user profile based on similar searches. In the future when a new event is registered and that happens to be a workshop on cake baking then this event would be recommended to the user since that's one of his/her interests.

I. Project overview

Many times, we receive a large number of notifications about various events, exhibitions, and meetups happening all around us that are irrelevant to us because we simply are not interested in them. Various people have their importance of things that they are interested in and be notified of all these events and from them searching for something that might interest them will take a lot of time and sometimes does not provide any meaningful information.

In today's world, there is no such existing facility that notifies us about the various events that are tailored to our interest. Many popular Recommendation systems show

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their users various events and exhibitions based on what's trending and what their friends are likely interested in. Although various recommender systems such as google news provide users personalized news that they are interested in. There simply exists no system that focuses on the user's preference and gives personalized suggestions for various events.

II. Project Deliverables

- 1: Preliminary Project Plan
- 2: Requirement Specification
- 3: Timeline
- 4: Architecture Specification
- 5: Source Code
- 6: Test Plan
- 7: Final Product Submission

III. Schedule

September 15 - System Planning, System Flowchart, Approval Phase

September 30 - Technology Learning - Django, ER Diagram, Designing System Architecture

October 15 - Detailed Study on recommendation Systems, Database Design

October 30 - Technology Learning NLP

Gathering User's Data

November 15 - Implementation of Dummy user profiling and suggestion of trending events, Implementation of content Recommendation System.

January 30 - Detailed Study of collaborative Recommendation System and implementation

February 15 - UI Designing and UI implementation

February 30 - Final Implementation of knowledge-based being event recommendation System

March 15 - UAT testing, Alpha testing

March 30 - Change implementation

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April 15 - UAT testing

April 30 - Final Changes and Production deployment

3.2.2-Project organization

I. Process model

Our project is going to use the Agile process model because it is iterative and incremental. Our requirements and risks are not concrete and are changing to a small extent as we go on developing the project and hence agile process model is best for that matter.

II. Organizational structure

Team Members -

1821008 - Mohammed Anzal Shaikh

1821009 - Vishvesh Pandey

1821010 - Samiksha Gupta

1821003 - Vikram Bhaysar

III. Project responsibilities

- Project Plan Entire Team
- Analysis Entire Team
- Diagrams and figures Vishvesh Pandey
- Backend Intra-website search recommendation Anzal Shaikh and Vikram Bhavsar
- Backend intra-website browsing pattern recommendation Vishvesh Pandey and Samiksha Gupta
- Chrome web browser extension Anzal Shaikh and Samiksha Gupta
- Local events and user web history recommendation Vikram Bhavsar
- Front End UI Vishvesh Pandey and Samiksha Gupta
- Populating user history data Entire Team

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- Populating Event data Entire Team
- Testing Entire Team
- Documentation Entire Team

3.2.3-Managerial process

I. Management objectives and priorities

The management objective is to deliver the product in time and of high quality. The PM and QA work together to achieve this by respectively checking that progress is made as planned and monitoring the quality of the product at various stages.

II. Potential Risks

Potential Risks are mentioned below

i. Technology risk

Risk of unavailable libraries for other languages other than python such as javascript

ii. Structure/process risk

Since the project is covered with research as well as development, there's a risk of a lot of time to be taken by the project for implementation.

3.2.4-Technical process

I. Methods, tools, and techniques

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The project will be implemented using agile methodology and tools such as Github, Python, Javascript, HTML, CSS, Django, Postgresql, Visual Studio Code Jupyter Notebook.

3.3 Software Requirement Specification Document

3.3.1. Introduction

I. Intended Audience

This project is intended for the audience who want to explore the events of their interest all at one place.

II. Intended Use

Event organizers can use this project to post various events and increase their reach, whereas the users can get access to all the events of their interest which will save the user's time.

3.3.2. Overall Description

I. User Needs

This system will have two participants:

Event Organizers

We plan on making a website where both event organizers and users have to participate. If an organization's person or any party wants to organize an event then they have to register this event through a portal. The organizer would put in various information about the events such as what the event is about, what is the date, venue, and other information which they seem to fit.

Example of a typical event: If a Treasure hunt event is going to be organized by some organization then they'll put information about this event on our web portal.

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Users

We aim to suggest events to the users with as little interaction with the system as possible. Each user can log into the website and see the top suggested events for them. These suggested events will be the result of multiple algorithms. We plan on taking implicit data from the user and suggest them based on their likings. We plan on limiting the interaction between the user and the system as much as possible when it comes to learning about interest from the user and hence data from the user would be taken implicitly.

II. Assumptions and Dependencies

- This project assumes that it will be able to fulfill and suggest the events of the user's choice based on general user profiling.
- Also, it will be dependent on the user's browsing history such as keywords, titles,
 etc. for user profiling.
- This system also assumes that a user may be using a Web browser to browse the web and various things and events of his interest
- The Event data should be submitted by Event organizers to the website. The event data is not web scrapped from some third-party website or database.
- History Extractor extension is pre-installed when browsing the website.

3.3.3. System Features and Requirements

I. Functional Requirements

A fully functional laptop/PC with an internet connection and user's prior web history.

II. Non-functional Requirements

Performance - The process of getting a user's history and extracting important keywords and ranking them. Sending them to the server and matching the keywords with the keywords from the events should be fast and fluent. Should not take more time and make the user wait.

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Safety - Since the browser extension takes user history data user browser history data must be kept safe from various sniffing attacks to keep user privacy.

Security - Proper Security mechanism to ensure user data is not leaked.

Quality - Recommending only those events which can be of interest to users and saving time of the user by limiting irrelevant interaction between user and the proposed system.

3.4 Software Design Document (All applicable UML diagrams)

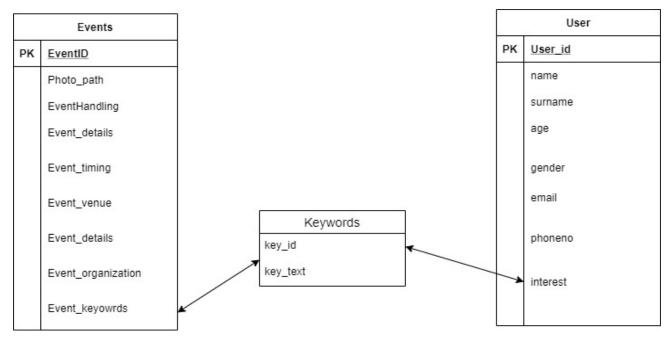


Figure (3.4.1): User's and Events data mapping based on keywords

Above is the schema of the proposed model. Here the Event and User table has many to many relationships with the keyword table.

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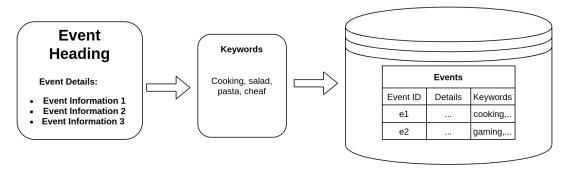


Figure (3.4.2): Keyword extraction from events and storing it into the database

Implementation at Event Organizers side

Event Details is a useful source for extracting keywords. When these event data are stored in the database, keywords related to those events would also be stored along.

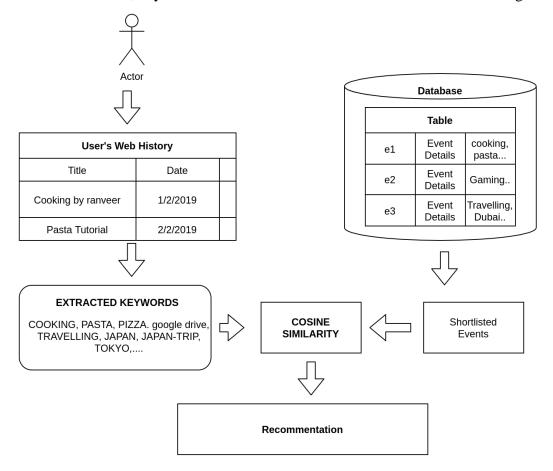


Figure (3.4.3): Recommendation based on Event Organizer's side

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Chapter 4: Implementation and Experimentation of Prototype Model

4.1 Structure of the project

4.1.1 File Structure

The main folder is Event_recommender_system

- history_extractor_extension this folder contains the browser extension.
- **recommender** This part of the code is responsible for providing recommendation based on the user's web browsing history
- **Collector** this folder is responsible for handling the implicit data that is gathered from the front-end side. Javascript makes ajax calls back to the server.
- static folder contains static files such as image, CSS, javascript.
- **Template** contains template code.
- **HelperPack** contains code related to stop words and calculating TD*IDF and cosine.
- **Events** main driver code. All requests are handled here.

4.1.2 How to run the project

Download the project using the below command:

Git clone https://github.com/vikramBhavsar/Event-Recommendation-System.git

Change directory using:

cd Event-Recommendation-System

Find out all the branches of the repo:

git branch -a

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Checkout to the remote master branch using the following command: git checkout remotes/origin/master

The project needs Django==3.0.7 (or above) and Python 3.5 and above.

It is recommended that the user runs the project in a virtualenv.

Check out the link below on setting up virtualenv using python

https://uoa-eresearch.github.io/eresearch-cookbook/recipe/2014/11/26/python-virtual-env/

Following are the list of dependencies that can be installed using **pip install** pip install pandas - (for handling CV data)
pip install sklearn - (for training and keyword extraction)
pip install geopandas - (for recommendation of local events)
pip install geopy - (for recommendation of local events)
pip install folium - (mostly not required)

Once all the dependencies are installed make sure the database is incorrect to state using the following commands:

python manage.py makemigrations python manage.py migrate

The GitHub repository contains pre-calculated similarity of events, keywords, etc. If in case a new CSV is provided or to be used then few scripts are needed to run before the execution of the project. These scripts calculate the similarities between events, extract keywords, etc and fill in the models with those data. Following is the order in which scripts need to be run. Also, make sure that the user is in virtualenv created using python.

```
python add_events_data_to_models_from_csv.py
python add_categories_to_events.py
python add_keywords_to_events.py
python add_similarity_to_events.py
```

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Once everything is set up, the user can run the project using the following command:

Python manage.py runserver <port-number>

Here port-number is optional. (Since browser extension uses port 8000 it is advised to use that port otherwise that link needs to be changed in background.js in extension.)

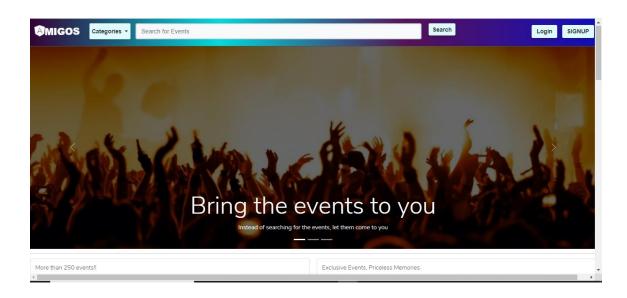
For personalized Recommendations, Recommendation install History Extractor Extension on our google chrome browser. Chromium browser does not work with handling location. To install the google chrome extension follow the below steps:

- Open chrome
- Click on the 3 dots on the top right-hand side and open settings
- Open Extensions from the left-hand menu.
- Activate the developer mode option on the top right side.
- Click on Load Unpacked.
- Go to our Event-Recommendation-System folder and select the folder which is named history_extractor_extension
- The browser extension will successfully install itself.

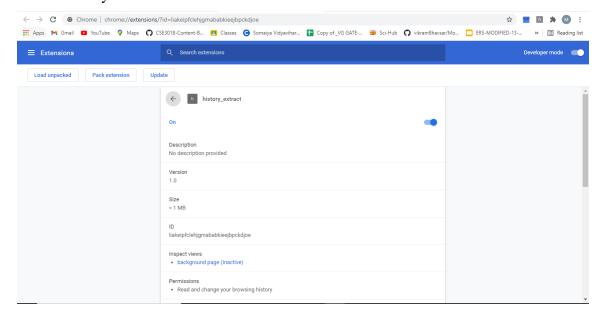
4.1.3 General Walkthrough of the project and basic functionality.

 When a new user comes to our website Home page is displayed along with all categories of Events

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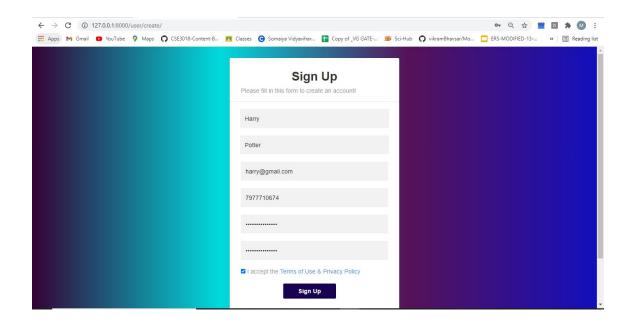


2) Then User needs to add an extension to the chrome browser for extracting personal history

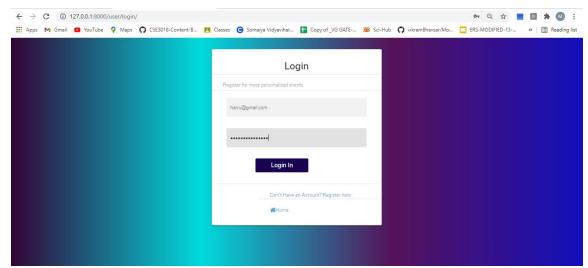


3) New User needs to SIGN UP first before moving ahead with events details and description

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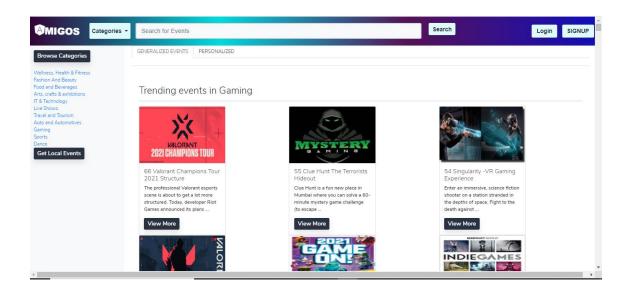
4) After signing up user needs to Login with proper credentials



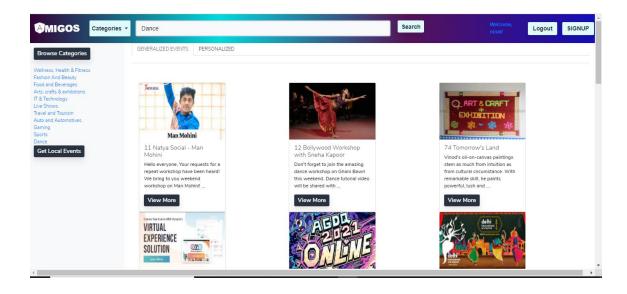
5) After logging in User gets all the trending events in each of the category mentioned on the homepage under the categories drop-down button

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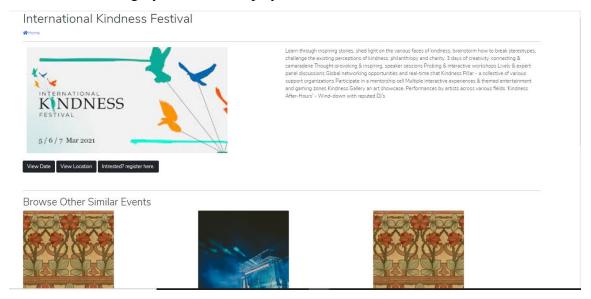


6) User can search for specific event category of his/her choice in the search bar

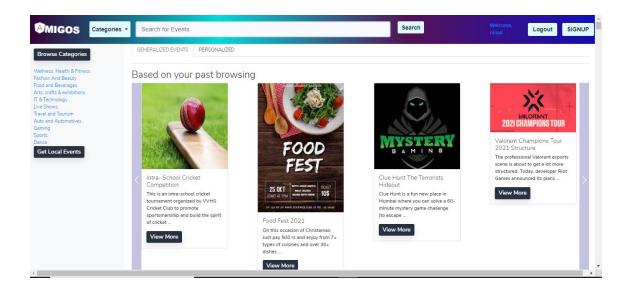


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7) By clicking on the View More option user can get a piece of detailed information about the event like date, venue and registration link, etc and also related to similar category events are displayed

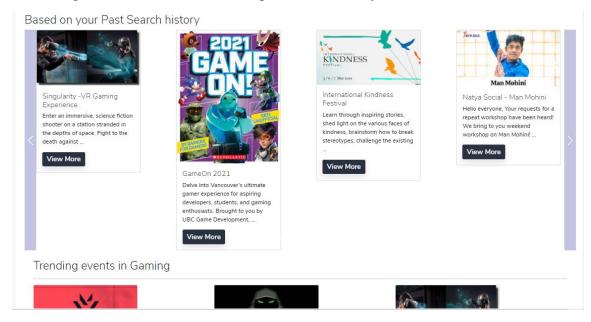


8) User also gets similar events based on past browsing history

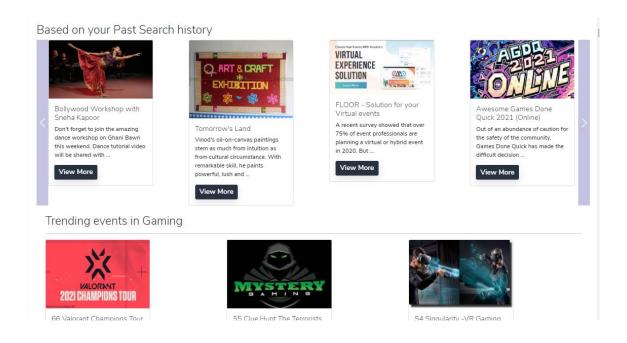


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9) User also gets similar events and only previous viewed event if he/she has not registered for it, into the tab of past search history

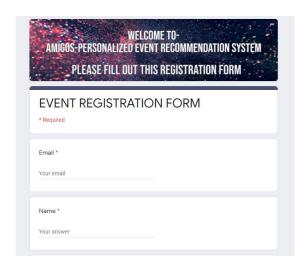


10) Once a user registers for a particular event then only similar events are displayed on to the past search browsing tab

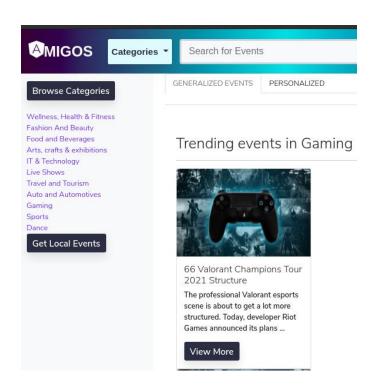


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11) Google form is displayed for registering for a particular event

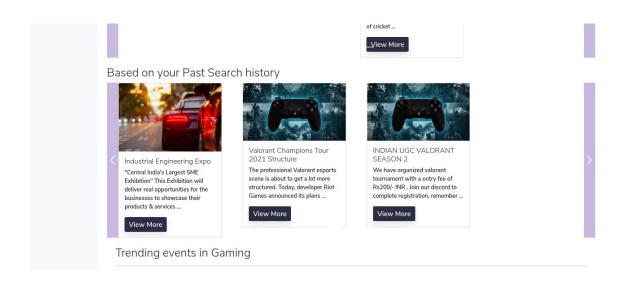


12) There's a button on the left hand side which says Get Local Events. When clicked the browser shares the location with the server.

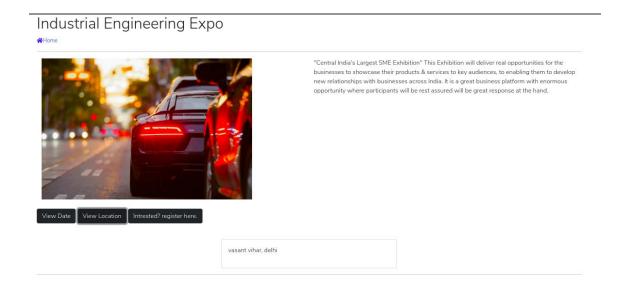


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13) Based on past search history the event **Industrial Engineering expo** is shown to the user.

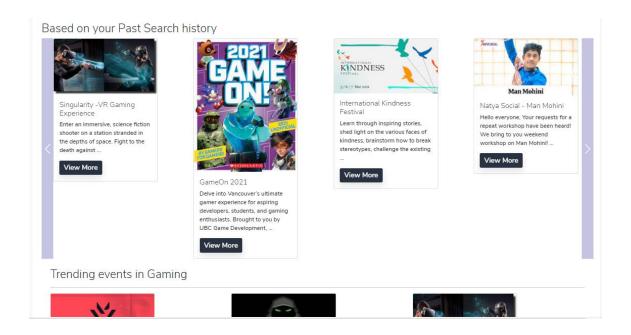


14) Here the venue for the event is **Vasant vihar, Delhi** but the user's location is Thane and hence it is not possible for the user to visit that event. Hence when clicked on **getting Local Events**

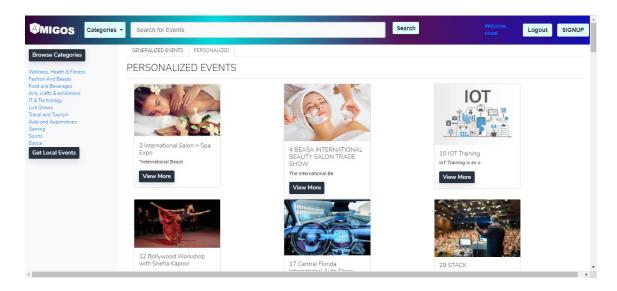


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15) That event is no longer shown.



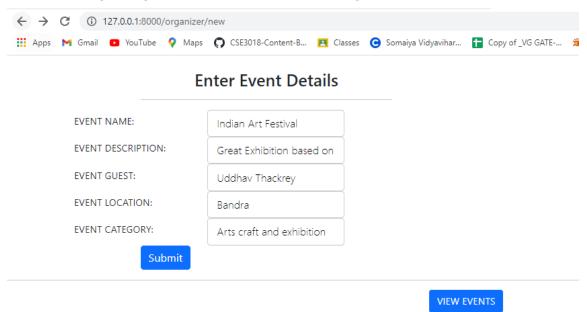
16) Personalized Tab shows all the recommendation based on user' web history



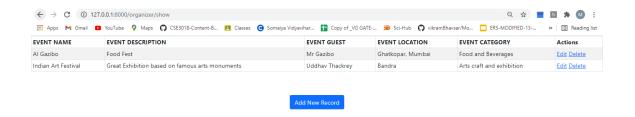
(Autonomous College Affiliated to University of Mumbai)

Organizer Side:

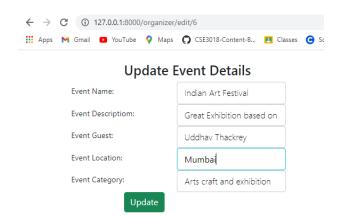
1) Creating/adding new events to our website from organizer's side



2) Organizer can view all events posted by him/her in View Events section



3) Organizer can edit as well as delete particular event from the list



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4) After deleting recently added event



4.2 Recommendations based on trending events

4.2.1 Architecture and logic

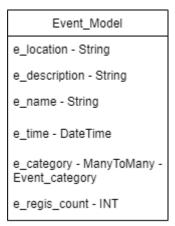
This is one of the simplest types of recommendation that is provided in this system. All the events are categorized. Available categories are:

- Wellness, Health & Fitness
- Fashion And Beauty
- Food and Beverages
- Arts, Crafts & exhibitions
- IT & Technology
- Live Shows
- Travel and Tourism
- Auto and Automotives
- Gaming
- Sports
- Dance

To provide this recommendation the system takes into account which active event has the largest number of registered users.

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4.2.2 Database Structure



Event_Category_Model
e_category - String

Figure (4.2.2): Database structure

Above is the schema of how the tables are organized.

4.2.3 Code

```
Models created in Django
Code in /events/models.py
```

Event Category Model

```
class Event_category_model(models.Model):
    e_category = models.CharField(max_length=60)

def __str__(self):
    return str(self.id) + " " + self.e_category
```

Event Model

```
# Create your models here.
class Events_model(models.Model):
    e_name = models.CharField(max_length=200)
    e_description = models.TextField()
```

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```
e_guest = models.CharField()
e_location = models.CharField()
e_time = models.DateTimeField()
e_category = models.ManyToManyField(Event_category_model)
e_regis_count = models.IntegerField(default=0)
```

Recommendation code:

Code in <u>events/views.py</u>

The above code first ensures that the user has not specifically requested any category and then it checks if the user has not provided any search query. If the user provides search query then we don't want to show trending events, we just want to show the search results. We loop each category from **Event_Category_Model** and get every event from that category by sorting it based on the registration count.

If in case the user is browsing any specific category then we show all trending events from that specific category. The following code is used:

Code in *events/views.py*

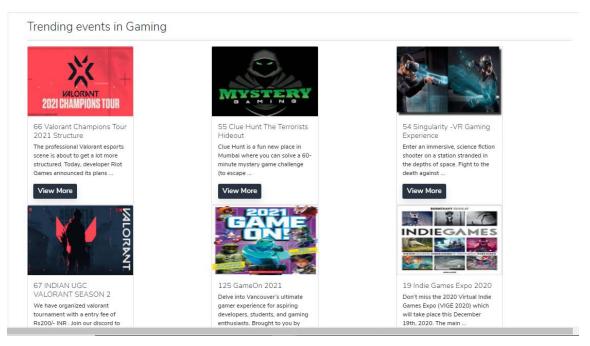
```
if 'category_pk' in self.kwargs:
   category_pk = self.kwargs["category_pk"]
```

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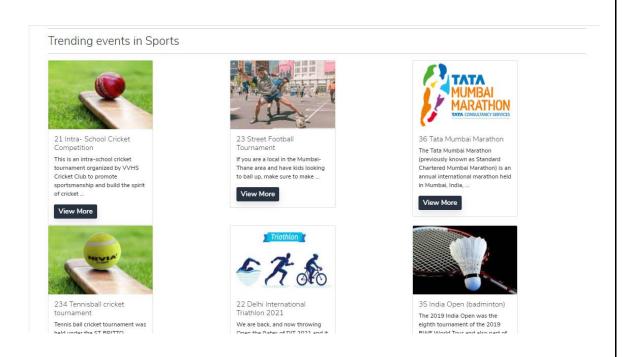
return

Events_model.objects.filter(e_category=Event_category_model.objects.get(pk=c ategory_pk)).order_by("-e_regis_count")

4.2.4 Output



Showing trending events from the gaming category in the **generalized tabs**.



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Showing trending events from the sports category in the generalized tabs.

4.3 Extracting keywords from event's description using TF*IDF Algorithm

4.3.1 Architecture and logic

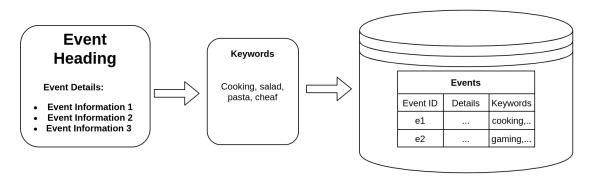


Figure (4.3.1): Keywords storing based on particular events

Here, Each event has useful information from which keywords can be extracted and its stored in a separated model known as **Event_keywords_model.**

4.3.2 Database Schema

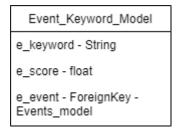


FIgure (4.3.2): Database Schema

The column **e_event** is a foreign key to the **Events_model** table.

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The column **e_keyword** is the keyword and **e_score** is the TF*IDF score of that keyword.

Example snapshot of the database looks like this:

Change event_	keywords_model	
E keyword:	mumbai	
E score:	0.06823233129080249	
E event:	234 Tennisball cricket tournament	v / +
Delete		

The image is the single entry inside the database. Here the keyword "Mumbai" and its score is given and also to which event this keyword belongs to.

4.3.3 Code

```
Model Code - Event_keywords_model

Code in <u>events/models.py</u>

class Event_keywords_model(models.Model):
    e_keyword = models.CharField(max_length=50)
    e_score = models.FloatField()
    e_event = models.ForeignKey(Events_model)
```

Following is the python script which generates keywords based on stored events inside the database and stores the updated keywords in **Event_keywords_model.**

Code in <u>add_keywords_to_events.py</u>

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Following is the class which was created:

```
class KeywordsExtraction:
  def get_record_split(self,sentence):
     rege = re.compile(r'[a-z0-9]{2,}')
     # returns list of words by removing any special characters from the string.
    return rege.findall(sentence)
  def getProcessedEventData(self,stop_words):
    # gets event details after pre-processing by removing stop words and combining
event
     # title and event Description together.
     return processed_events_list
  def tf_idf_get_keywords_list(self,event_data_list):
    # a list of keywords for a particular event that will added to the main data frame as a
sentence
    # and will be used during cosine similarity between events and interest extracted.
  def updateKeywordsAndScoreToDatabase(self,stop_words):
     # this method is responsible to update the keywords and their respected scores into
the database
       Below is the driver code.
       # Getting stop words for extraction
       from HelperPack import help
       stop_words = help.getStopWords()
       extractioner = KeywordsExtraction()
       extractioner.updateKeywordsAndScoreToDatabase(stop_words)
```

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Here HelperPack is a library from which help is imported. We need stopwords in order to calculate the keywords.

The first Method that is called is **updateKeywordsAndScoreToDatabase**.

updateKeywordsAndScoreToDatabase - it calls getProcessedEventData and
gives that data to tf_idf_get_keywords_list which returns a tuple. Which is a
keywords_list, scores_list.

Those two lists are then added as columns into the database.

getProcessedEventData - removes stop words and alpha numeric values from the description and title and then returns the list back to the caller method.

Tf_idf_get_keywords_list - uses the **TfidfVectorizer()** to generate a list of keywords and their corresponding scores of each keyword.

4.3.4 Output

EVENT_KEYWORDS_MODEL
☐ 16042 mumbai 0.06823233129080249
☐ 16041 teams 0.09142783312914754
☐ 16040 school 0.09142783312914754
☐ 16039 hands 0.09455026916896536
☐ 16038 participated 0.1085904780081596
☐ 16037 league 0.1085904780081596
☐ 16036 match 0.1085904780081596
☐ 16035 prizes 0.1085904780081596
☐ 16034 finally 0.1085904780081596
☐ 16033 chennai 0.1085904780081596

A table is generated after the script has been executed.

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4.4 Calculating Similarity of events

4.4.1 Architecture and logic

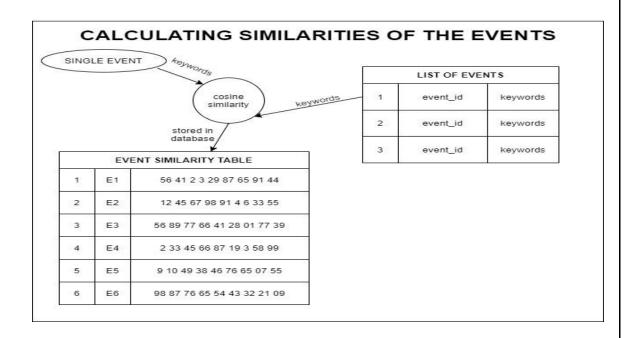


Figure (4.4.1): Calculating Similarity of Events

When a user views any event then the detail page of the website shows another 5 events which are similar to the event. This is because similarities between events are calculated and stored inside a table.

Calculated similarities are stored in the SimilarEvents model.

4.4.2 Database Schema

SimilarEvents

event-oneTOneEvents_model

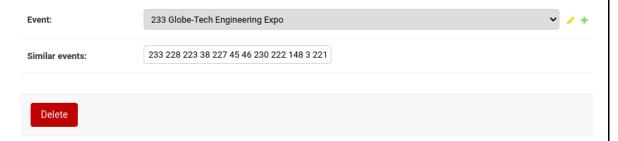
similar_events= String

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As shown in the above schema. The column **event** is oneToOne field to the **Events_model** and column **similar_events** is a string that contains space-separated ids of events in descending order of their similarities.

Example:

Change similar events



Here the 233 Globe-Tech Engineering Expo is the event and the list of similar events i.e 233 228 223 are other events that are most similar to that event. Also, client-side events happened more similar to 233 as compared to 223. It's in descending order of their similarity which makes sense as we only need the top 5 similar events.

4.4.3 Code

Model Code. Code in <u>recommender/models.py</u>

Create your models here.

class SimilarEvents(models.Model):

event =

models.OneToOneField(Events_model,on_delete=models.CASCADE,primary_k ey=True)

similar_events = models.CharField(max_length=200)

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Following is the code to get similar events based on the currently viewed event's primary key. The primary key of the event is taken from the URL.

Code in *events/views.py*

```
def get_context_data(self, **kwargs):
    cur_even_pk = self.kwargs['pk']
    similar_events_id = SimilarEvents.objects.get(pk=cur_even_pk)
    similar_events_id = similar_events_id.similar_events.split(' ')

# converting ids to numeric

# similar_events =

Events_model.objects.filter(pk__in=[similar_events_id_int[1:min(6,len(similar_events_id_int))]])
    similar_events =

Events_model.objects.filter(pk__in=similar_events_id_int[1:min(6,len(similar_events_id_int))])
    context['similar_events'] = similar_events
    return context
```

Once the context has been made and passed to the response then all the events are displayed on the client side using template code as follows:

Code in <u>templates/events/events_details.html</u>

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4.4.4 Output





Enter an immersive, science fiction shooter on a station stranded in the depths of space. Flight to the death against killer robots as you explore narrow corridors, teacherous life, and zero-gravity environments. Zero Latency is one of the world's best free-roam VR gaming experiences where you can roam around in an area CDO0 as, feet without writee! Feating state-of-the-stat, award-winning technology from one of the pioneses in VR entertainment, players can move around freely, as well as talk, interact and strategize with their team-mates in real-time. Choose from a variety of exciting multiplayer games ranging from cooperative like games surviving the zomble apocalypse or rescuing a space station from robots. "Selar into the world of Zero Latency where Your body is the controller and your mind believes it is real "EEXPLORE. EPLAY."

Browse Other Similar Events



19 Indie Games Expo 2020 Don't miss the 2020 Virtual Indie



OrcaCon 2021 is an inclusiv



The user is currently looking up an event on VR gaming experience and below the user can see other similar events which are Indie Games Expo, OrcaCon, etc.

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4.5 Recommendation of events based on past Intra-website search

4.5.1 Architecture and logic

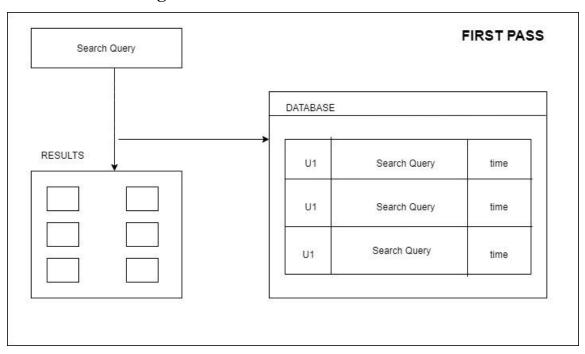


Figure (4.5.1.a): Architecture of Recommendation Based on Past Intra-Website Search

In Order to provide powerful recommendations to the user, this website takes into account what the user has searched within the website as well. If the user logs in and searches for any specific event then it is very evident that the user is interested in that search query. Hence these search queries are used to provide the results but these are stored in the database as well. As seen in the above diagram whenever the user searches for a keyword two things happens:

- Results of that search query is shown which is normal and expected
- The search query is stored inside a table in the database.

We have named this process as "First Pass" because to provide a recommendation based on an intra-website search query requires two passes.

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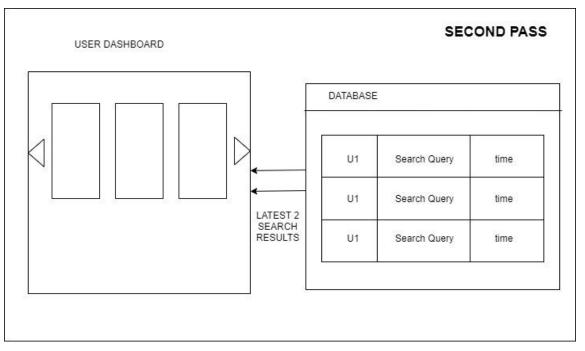
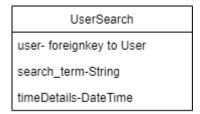


Figure (4.5.1.b): Based on past search query results from the database

Here in the second pass, The server checks whether the logged-in user has any intra-website search query stored inside the table. If there isn't then no results are shown specifically to this. If there are records then the server gets the latest two search queries and displays them back to the user. This is part two.

4.5.2 Database Schema



Above is the schema of the table. All the UserSearch is stored inside this table. The user is a foreign key to the main user model. It's not one to one Field since there will be multiple records of a single user. **search_term** is to store the actual term used by the user and time details is DateTimeField. Here, that is taken as default and the user does not have control over it. This field is required because we needed the latest two search results.

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4.5.3 Code

```
Database Model Code:
```

Code in <u>user_app/models.py</u>

```
class UserSearch(models.Model):
    user = models.ForeignKey(User)
    search_term = models.TextField()
    time_details = models.DateTimeField(default=now)
```

Code required for the first pass:

```
Code in events/views.py
Inside <u>EventsListView -> get_queryset() method.</u>
# checking if user has searched for any events
 if 'search_q' in self.request.GET:
    search_term = self.request.GET["search_q"]
    if self.request.user.is_authenticated:
      cu_user = User.get(pk=self.request.user.id)
      new_search =
UserSearch(user=cu_user,search_term=search_term,time_details=datetime.now()
)
            new_search.save()
          return Events_model.objects.filter(Q(e_name__contains=search_term)|
Q(e_description__contains=search_term))
       else:
          print("Normal search is taking place.")
          return super().get_queryset()
```

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Here the code checks if the user is authenticated, we get the user ID and search term from the request and then store it into the UserSearch model.

We then return the search result regardless if the user is logged in or not. If the user is not logged in then we just return the search results and if the user is logged in then we first save the search query and then return the results anyway.

Code for Pass two:

```
# getting results based on past search history
cu_user = User.objects.get(pk=self.request.user.id)
User_browse_recommends = UserSearch.objects.filter(user=cu_user).order_by("-
time_details")
       # to get only last 3 searches
  search\_count = 0
  for user_search in user_browse_recommends:
      search_count += 1
       # taking only last two search count to show.
           if search_count >= 4:
              break
search_events = Events_model.objects.filter(
Q(e_name__contains=user_search.search_term)
Q(e_description__contains=user_search.search_term))
## CODE TO HANDLE USER'S LOCATION REMOVED FOR SIMPLICITY.
More on this in "Recommending local events only"
```

Here if the user is logged in we get the user id from the request and get the user object. We then use that user object to filter **UserSearch** and order the records by

RETURNING THE SEARCH EVENTS

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time. We then use the search count to only use the latest two searches to show results. The list of events is then appended to the list.

```
browse_recs_carousal_active,browse_recs_carousal = get_courousel_list_for_objects(user_browse_recommendation_list)
```

Context["user_browse_recommendations_active"] =browse_recs_carousal_active context["user_browse_recommendations"] = browse_recs_carousal

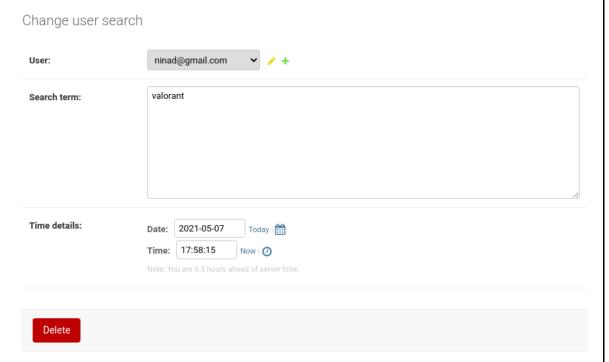
Since the Carousel we used in front end requires an "Active" list of events and then the rest of them we divide them in two parts and attach it to the context object. These two parts are named as **browse_recs_carousal_active** and **browse_recs_carousal.**

Code in <u>templates/events/events_list.html</u>

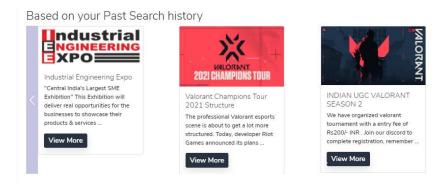
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Small snippet of code where the active carousel events are rendered on html

4.5.4 Output



This is a database snapshot of user ninad. The search term he used was **valorant** which is a popular game.



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Because of the above table, during pass two ninad sees this carousel where search results for **valorant** are shown.

4.6 Recommendation of events based on intra-website browsing patterns

This section is mainly focused on how to make sense of various activities that the user does and what it means. At the heart of a good recommendation system is to take implicit data from the user to provide recommendations. Whenever the user interacts with a particular section on the website, in our case any event, then we can safely assume that the user is interested in that particular section, or in our case an event. We collect data from the front end using javascript when the user interacts with various events on the website.

4.6.1 Architecture and logic

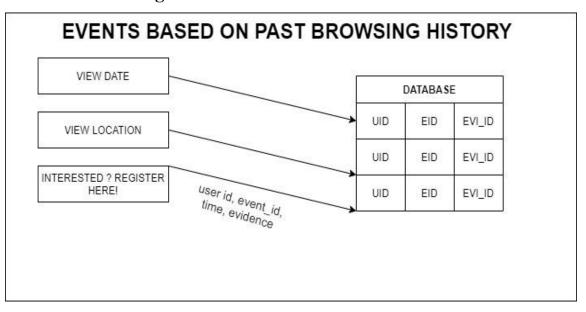


Figure (4.6.1.a): Architecture of Recommendation based on Past Browsing History

There are a total of 4 places from where data is collected from the user's side using Javascript. Which are:

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- When the user clicks on view more for an event and goes to the detail page of that event.
- When the user clicks on the **view date** of that event
- When the user clicks on the with **view location** of that event
- When the user finally registers for that event.

A table is maintained, known as **event_user_log** model, with 7 columns. 4 columns for 4 types of data collected from the user's side (mentioned above). And other details such as user id, event id which references the event which the user interacted with, and DateTime to order by latest events the user interacted with.

Each time the user clicks on any of the four "evidence" i.e (View details, View date, View location, and register?) the client sends the data back to the server where it is stored inside the table.

The primary key of this table is based on User ID and Event ID together since users will interact with multiple events and each event would be interacted by multiple users.

, = , , , , = 0, 0, ,	EVENTS
EGISTERED EVENTS	NON-REGSITETED EVENTS
REGIS=1	REGIS=0
	EGISTERED EVENTS REGIS=1

Figure (4.6.1.b): Types of Event

When collecting data on events there are two different types of events.

- Registered events
- Not registered events

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Registered events have the highest score, which will be explained below, but there's no point in recommending an event that has already been registered by the user. But we still take that data because now we know for sure what kind of events the user likes and hence we can show other similar events based on that registered event.

For recommending similar events refer to the section on *calculating the similarity* of events

SCORING EVENTS

EVENT1	EVENT2	EVENT3
VIEW DATE = 1	VIEW DATE = 1	VIEW DATE = 1
VIEW LOCATION = 1	VIEW LOCATION = 1	VIEW LOCATION = 0
VIEW REGISTRATION = 1	VIEW REGISTRATION = 0	VIEW REGISTRATION = 0
SCORE= 40 + 40 +100	SCORE = 40 + 40 + 0	SCORE = 40 + 0 + 0
= 180	= 80	= 40

Figure (4.6.1.c): Scoring Events

Scoring the events

All the evidence that is gathered from the front end is then used to score the events. Each piece of evidence has a weight assigned to it based on its importance. Here importance is how likely is that the user is interested. For example, if the user has registered to an event then it is very likely that the user is interested in that event but registered events cannot be recommended back to the user.

When the user reads the description and checks the date and location then it can be implied that the user is thinking of registering for that event. So evidence like **view**

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date and view location is also very important in terms of calculating the interest of the user. But at the same time, it is not as important as registering for the event. Similarly when the user clicks on an event to see what that event is about can be used to imply that the user is interested. But compared to other evidence types we have it is the least important of all.

Hence weight assigned to each of these pieces of evidence are:

- Is a user registered? Weight is 100
- If a user has a view date and location both? Weight is 80
- If a user has just viewed the date or location? Weight is 40
- If the user has just viewed the event? Weight is 20.

The final score of an event and user interaction is calculated by summing all of the above.

As seen in the diagram, Event 1 has a score of 180 because not only did the user inquire about the date and location together but he/she registered for the event as well. Hence 80 (40+40) +100 = 180.

For Event 2, the User has just viewed the date and location hence the score of that event for that particular user is 40 + 40 + 0 = 80.

For Event 3, the user has just viewed the date and hence the total score is 40 + 0 + 0 = 40.

Hence the recommendation of new events would be based on Event 1 and Event 2 rather than 3 since it is implied that the user is interested more in events 1 and 2 based on the evidence collected from the front end.

4.6.2 Database Schema

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EventUserlog

user-foreignkey to user model

event-foreignkey to EventsModel

timedetails-DateTime

viewDetails-Integer

viewDate-Integer

viewLocation- Integer

viewRegistration-Integer

The table is as above schema.

4.6.3 Code

Model Code

Code in *collector/models.py*

```
class Event_User_log(models.Model):
    user = models.ForeignKey(User)
    event = models.ForeignKey(Events_model)
    timedetails = models.DateTimeField(default=now)
    viewDetails = models.IntegerField(default=0)
    viewDate = models.IntegerField(default=0)
    viewLocation = models.IntegerField(default=0)
    viewRegistration = models.IntegerField(default=0)
```

Firstly, the data is collected from the client-side using javascript.

Code in <u>static/js/customJavaScript.js</u>

```
function sendEvidenceToCollector(userid,eventid,evidenceType){
  var xhttp = new XMLHttpRequest();
  xhttp.onreadystatechange = function(){
```

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```
xhttp.open('POST','/collector/put_evidence/',true);

xhttp.send(`user=${userid}&eventid=${eventid}&evidenceType=${evidenceType}');
}
```

This JS function is called from the HTML buttons whenever the user clicks on it.

```
Below code in <u>templates/events/events_list.html</u>
```

Here user's login id and events primary key are provided by Django using template language and the evidence ID here is **4**. Which is the View details Evidence.

We also check whether the user is logged in or not because we do not want to send Evidence since we wouldn't have the user id if the user has not logged in.

```
Code in templates/events/events_details.html
```

```
{% if user.is_authenticated %}
<button onClick=
"sendEvidenceToCollector({{user.id}},{{object.pk}},3)">View Date</button>
```

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```
<button
onClick="sendEvidenceToCollector({{user.id}},{{object.pk}},2)">View
Location</button>

<button
onClick="sendEvidenceToCollector({{user.id}},{{object.pk}},1)">Interested?
register here.</button>
{% else %}
// normal buttons
{% endif %}
```

Here the evidence id 3, 2, existsInare, and 1 is sent from the details page which refers to view date, view location, and Registration. Whenever the button is clicked the javascript makes an ajax request back to the server.

The ajax request is made to /collector/put_evidence/ link.

Below code in *collector/views.py*

The collector class is specifically made to collect the evidence.

Following is the post request in **CollectBrowsingData** view.

The below code is responsible for handling the post request made from the front end/client-side using javascript.

def post(self,request):

```
userid = int(request.POST["user"])
eventid = int(request.POST["eventid"])
evidenceType = int(request.POST["evidenceType"])

log_user = User.objects.get(pk=userid)
log_event = Events_model.objects.get(pk=eventid)
log_event.e_regis_count += 1
log_event.save()
```

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```
try:
# will be executed if data is already present in the database
event_log = Event_User_log.objects.get(user=log_user,event=log_event)
event_log.timedetails = datetime.now()
if evidenceType == 1:

    event_log.viewRegistration = event_log.viewRegistration + 1

/// for all type of evidence
event_log.save()
return response(JsonResponse({"status":'done'})
```

In the above code it is first checked whether the entry of a particular user and event already exists or not. If an entry exist then we update the existing entry by getting that object. If not then we create a new object and save that in the model.

At the end, we just respond with a JSON responseIn which doesn't do anything since in case an error occurs we don't really want to handle it.

Providing recommendation based on these evidences

```
Below code is in <a href="mailto:events/views.py">events/views.py</a>
Inside the <a href="mailto:get recs based on click events(user)">get recs based on click events(user)</a>: method

# calculating score:

for evidence in evidences:

    score = 0
    # print("[+] Investigation %s" % evidence)

if evidence.viewRegistration > 0:

    score += 100

# checking view date and location together. (very strong chance the user is interested)

if evidence.viewDate > 0 and evidence.viewLocation > 0:

    score += 80
```

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```
elif evidence.viewDate > 0 or evidence.viewLocation > 0:
       score += 40
if evidence.viewDetails > 0:
  score += 20
if score > 0:
   evidences_id[evidence.event.id] = score
  # sorting the dictionary based on values.
  sorted_evidence
                            sorted(evidences_id.items(),
                                                            key=lambda
                                                                             x:
x[1],reverse=True)
The scores are calculated and sorted in descending order since higher the score
more important.
        responsible to get the highest_scored_registered_events
   is
highest\_scored\_not\_registered\_event
We pass these two objects to the below function
def get_recs_from_regis_and_not_regis(high_regis_event,high_not_regis_event):
recs_list = []
if high_regis_event is not None and high_not_regis_event is not None:
    recs_list.append(high_not_regis_event)
similar_list = getTopFiveSimilarEvents(high_regis_event)
 for eve in similar list:
   recs_list.append(eve)
similar_list = getTopFiveSimilarEvents(high_not_regis_event)
```

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// FOR MORE CODE REFER THE GITHUB REPO return recs_list

The logic here is that instead of recommending registered events again back to the user, which serves no purpose, we show other similar events to the user based on that registered user.

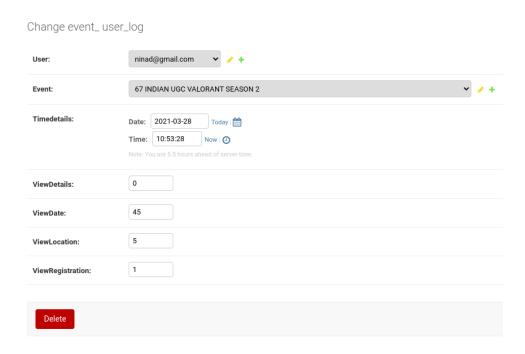
On the other hand, the below pieces of evidence the event which scores the highest is then again recommended to the user with the hopes that the user may register it now since they viewed it already.

4.6.4 Output

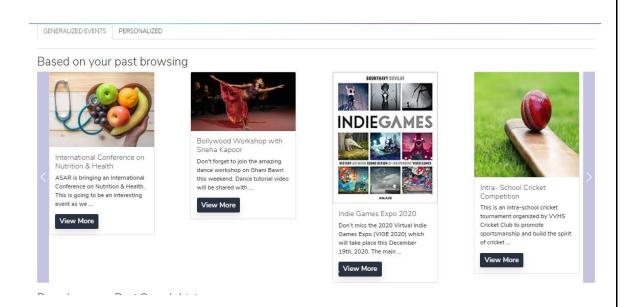
Action: Go 0 of 80 selected
□ EVENT_USER_LOG
harsh@gmail.com - 67 INDIAN UGC VALORANT SEASON 2
harsh@gmail.com - 28 Uttarakhand Darshan
harsh@gmail.com - 66 Valorant Champions Tour 2021 Structure
jethalal@gmail.com - 28 Uttarakhand Darshan
jethalal@gmail.com - 66 Valorant Champions Tour 2021 Structure
shamik@gmail.com - 23 Street Football Tournament
shamik@gmail.com - 28 Uttarakhand Darshan
shamik@gmail.com - 66 Valorant Champions Tour 2021 Structure
anzal@gmail.com - 147 Solo Dance Competition Online Powered By AYC
nidhi@gmail.com - 150 Groove Dance Competition

Above is the database snapshot of different users interacting with different events.

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The above is an example of user ninad interacting with the valorant season 2 event. And evidence are shown here.



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Based on your past browsing

Based on those data, scores are calculated and then events are recommended back to the user

4.7 Extracting User's Web browsing history

4.7.1 Architecture and logic

This is one of the most important parts of our project. We need a user's history in order to provide better recommendations based on their web browsing history. This is a complex logic and hence we need help with a browser extension because JavaScript has more power as a browser extension compared to normal web pages. We need access to the user's web browsing history which is simply not possible using normal JavaScript that we pass during serving requests.

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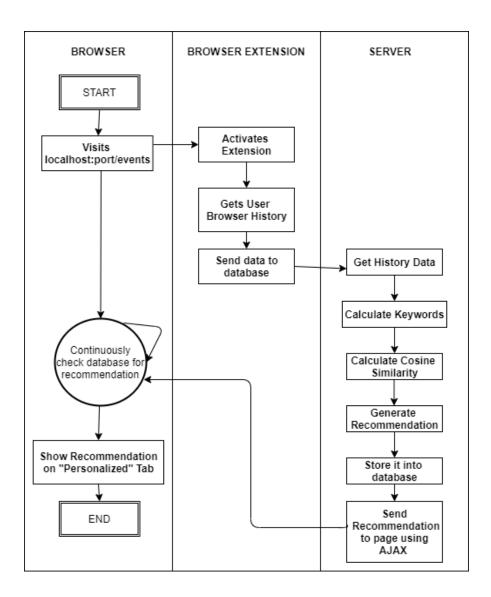


Figure (4.7.1): Architecture of Extracting User's Web History

Here whenever the client visits the link localhost: port/events the browser extension that is installed is automatically activated.

The task of browser extension is to gather the user's browsing history into a JSON data and then send it to the server. The browser extension also extracts User ID from the page visited which is a hidden field inside in the main page. If the hidden field is missing that means the user is not logged in and hence the extension won't send any data.

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While at the same time the front end webpage i.e /events continuously sends requests to the server checking if recommendations are available with respect to the user's browsing history.

When the browser extension has gathered all the History data it then sends it to the server. The Server calculates the recommendation and stores it temporarily inside the database.

When the data is ready the client side which is continuously checking the database for new records using ajax finds new records. Hence server replies with the new recommendation and the javascript on the front end render that on the client side access last access time two-three

4.7.2 Code

Here is the manifest file which is required to make web browser extensions Code in history_extractor_extension/manifest.json

```
{
  "name":"history_extract",
  "version":"1.0",
  "manifest_version":2,

  "background":{
      "scripts":["background.js"],
      "persistent": false
    },

  "content_scripts":[
      {
      "matches": ["http://127.0.0.1:8000/events/"],
      "js":["content.js"]
      }
    ],
```

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```
"permissions":[
    "history"
]
```

Here the inside the **content_scripts** the matches is the property which says that whenever the link http://127.0.0.1:8000/events/ is used inside the browser then execute the **content.js** file of the extension.

Below is the code responsible to check the hidden field for UserID. If the user id is not present then it means that the user has visited the link but he/she is not logged in.

Code in *history_extractor_extension/content.js*

```
var userHiddenField = document.getElementById("user_id_hidden_field");

if(userHiddenField.value != 'not_logged'){
    // alert(`User ID is ${userHiddenField.value}`);
    chrome.runtime.sendMessage({text: userHiddenField.value},
    function(response) {
        console.log(`${response}`) });}
```

When the user is indeed logged in then the **content.js** sends a message to the **background.js**

- Content.js Run in the context of the web pages.
- Background.js constantly runs in the background regardless of any website visited.

Background JS is then responsible for extracting the history.

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```
Code in <a href="https://extractor_extension/background.js">https://extractor_extension/background.js</a>
```

```
chrome.runtime.onMessage.addListener(function(msg, sender, sendResponse) {
var userId = msg["text"];
chrome.history.search({text:
                                                    9999999,startTime:
                                    maxResults:
                                                                           0},
function(data){
       // data is history
    data.forEach(function(page) {
if(lastGotIT == false){
 latestAccessedTime = page.lastVisitTime;}
singleHistory["title"] = page.title; // other information
historyJSON.push(singleHistory);
})
moreData = \{\}
moreData["latestAccessedTime"] = latestAccessedTime;
    historyJSON.push(moreData);
 sendHistoryToServerUsingPOST(historyJSON);});
 sendResponse("something"); });
```

Above is the method which gets executed when called by the content.js.

It accesses the user's browsing history and also access the **lastAccessTime** which is the date and time of the latest website visited by the user. This will be needed for calculating recommendations.

Once all the data is gathered the method internally calls **sendHistoryToServerUsingPOST(historyJSON)** where historyJSON is the web browser's history in json format.

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Within the same file, following method is used for sending data back to the server.

Code in hittory_extractor_extension/background.js

```
function sendHistoryToServerUsingPOST(historyJSON){
   var xhttp = new XMLHttpRequest();
   xhttp.onreadystatechange = function(){
      console.log(xhttp.responseText);
   }
   xhttp.open('POST','http://127.0.0.1:8000/recommender/index/',true);
   xhttp.send(JSON.stringify(historyJSON));
}
```

The next section discusses how the data is received through the POST and further calculations are done for event recommendations.

4.8 Calculating similarity between User's web browsing history and events and recommendations

Following section discusses how the recommendations are calculated based on the extracted history from the client browser.

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4.8.1 Architecture and logic

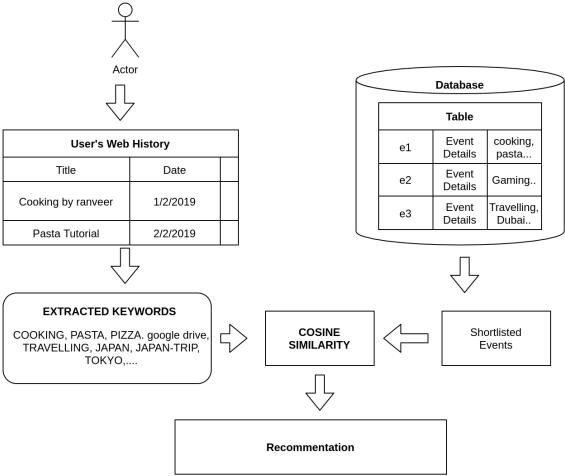
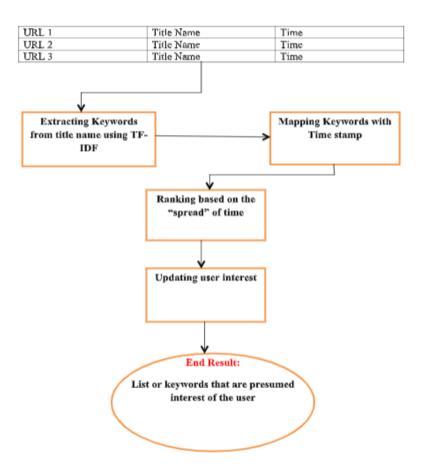


Figure (4.8.1): Architecture for Calculating similarity between user's web history and events

There's useful information on the title of the web page that the user visits. For example, The following the JSONconcerningfront-end the with pieces of evidence if the user happens to visit the news section of valorant game using the link: https://playvalorant.com/en-gb/news/ the title of that web page is "VALORANT: RIOT GAMES". Hence title is our only and a very good source of information for keywords. The keywords which are extracted from the user's history are then matched with the keywords extracted from various event's description and title. Cosine Similarity is used to calculate similarity between the two and based on that recommendations are provided.

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User's Web History



Here from the list of titles available. They are pre-processed such as removing stop words, removing numbers, the the two three, and converting every uppercase character to lowercase. Then each keyword is matched with the timestamp. The timestamp is the time when that particular keyword occurred. There may be a single keyword that would have multiple timestamps and that's the point. Keywords that are repeatedly found are of importance to us. Timestamp and time details are required to count the spread of that keyword. The understanding is that if repeated words have more spread it means that is something a user is interested in. If the spread is less but the count is more then it's just noise that the user was searching for or came across a topic for a day. For example: if a user searches about chairs for a single day but many times the count of the word chairs would be a lot but then again there isn't much "spread" of time between their access. They were

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accessed consistently and repeatedly in a short period which does not amount to something that the user may be interested in.

On the other hand, if the user searches about valorant game one day, then after two-three days and then again after a week, here not only the individual count rise but also the "spread" is more and can be implied that since user is interested in the topic he/she is regularly invested to research about that topic.

Any person interested in dance would look up dance videos and workshops regularly. Spread here would be more.

After the spread is calculated these keywords are ranked in descending order because the higher the spread score, the more important the keyword is to the user.

To keep the privacy of the user, the history is not stored inside the database. Whenever the server receives the history immediately the recommendations are calculated and then only event ids which are the calculated recommendations are stored inside the database keeping the privacy of the user.

They calculated keywords and the spread score and keywords from events and their score are taken together and cosine similarity is calculated to recommend events to the user.

4.8.2 Database Schema

HistoryRecommendedEvents
user oneToOne- User_model
rec_events-String
latest_update-DataField

Figure (4.8.2): Database Schema

Above is the schema used to store the calculated recommendations?

4.8.3 Code

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As seen in the earlier section, the browser extension makes a POST ajax call to the server giving the history data.

```
elif request.method == 'POST':
jsonHistory = json.loads(request.body)

// For simplicity code is reduced .refer the github repo
putRecInDatabase(sorted_recs,user_id)
return HttpResponse("Received Data from server")

The code makes use of the below of the helper pack.

class HistoryRecommendation():

def get_record_split(self,sentence):
def main_driver(self,records,latest_access_time,
```

Below code is in <u>recommender/views.py</u>

Here the main driver code is main_driver which produces **sorted_keywords**, **keywords,keyword_set,max_spread,min_spread** for us.

• Max_spread is the maximum spread

stop_words):

latest_access_time):

def getSpread(self,word_counts,

- Min_spread is the minimum spread
- Keyword_set is just a set of keywords to check whether any keyword is available. It is used for fast checking
- Keywords are the dictionary of keywords with their individual score.
- Sorted_keywords is the list of keywords which are sorted in descending order based on their score.

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Implementation of the main_driver function is as follows:

```
def main_driver(self,records,latest_access_time,stop_words): // Code is reduced for simplicity sake return sorted_keywords,keywords, keyword_set,max_spread,min_spread
```

To generate individual keywords properly below function is used which in turn

```
def get_record_split(self,sentence):
    rege = re.compile(r'[a-z0-9]{2,}')
    # returns list of words by removing any special characters from the string.
    return rege.findall(sentence)
```

It is responsible for splitting a sentence into words.

makes uses of regular expressions

The main driver also uses following function to calculate the spread of the keywords:

```
def getSpread(self,word_counts,latest_access_time):
    max_spread = 0
    // code is reduced for simplicity sake. Refer to the github repo
    return keywords, keyword_set, max_spread, min_spread
```

After getting these keywords from the user's history side we calculate the cosine similarity. Cosine similarity is calculated from the below function.

Code in <u>recommender/views.py</u>

```
def
getCosineBetHistoryAndEvents(sorted_keywords,keywords_dictionary,user_key
words_set,max_spread,min_spread):
```

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```
// for simplicity sake code is reduced. Refer to the github repo
return sorted_tuples
```

And then finally putting those recommendations into the database using the below function in the same file.

```
def putRecInDatabase(sorted_recs,user_id):
    cu_user = User.objects.get(id=user_id)
    // For simplicity sake code is reduces. Refer to the github repo
reco = HistoryRecommendedEvents(user=cu_user,
rec_events=recs)
reco.save()
```

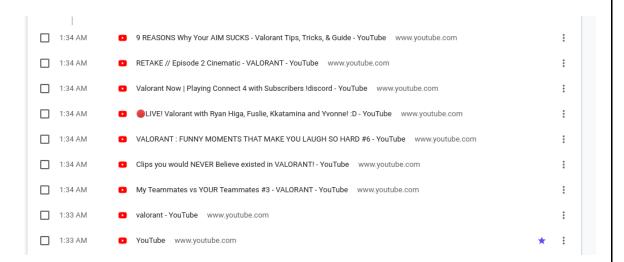
4.8.4 Output

Change history recommended events

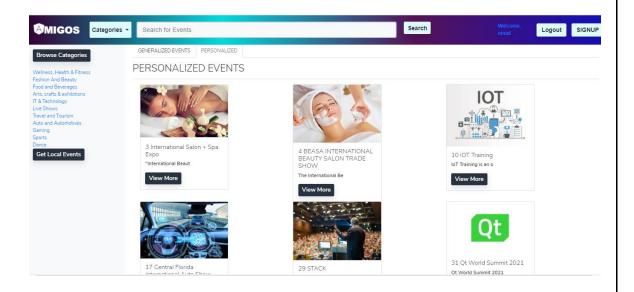
User:	ninad@gmail.com 🗸 🤌 🕂			
Rec events:	12 58 66 67 69 75 81 131 141 111 10			
Latest update:	2021-05-26 Today ### Note: You are 5.5 hours ahead of server time.			
Delete				

This is a snapshot of the database from Django Admin. Here the user is ninad and the calculated recommendation of the event is stored space separated as a CharField. Date is given to ensure the records are always updated.

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Above is the snapshot of the web history that is present in the browser. And below we can see that few valorant recommendations are provided along with others as well



The output of the personalized tab. Here the final recommendations are shown.

- 4.9 Recommending local events only.
- 4.9.1 Architecture and logic

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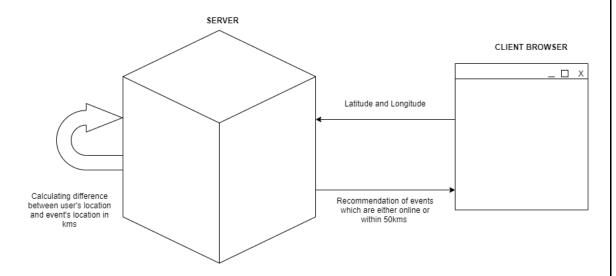


Figure (4.9.1): Architecture for recommending local events

Here the main page allows a button and option to show local events. When clicked on local events the browser automatically shares the user's latitude and longitude coordinates to the server. The server has event data and each event data has a location field. This could be Vasant Vihar, Delhi or Manpada, Mumbai etc. Based on this information the server is able to extract the coordinates of that location. So now the server has coordinates of the event as well as coordinates of the user and hence the server can calculate the distance between the event venue and user. Using this information server can calculate the distance between them in km. So with the aim of suggesting only the local events, the server does not recommend the event to the user where the distance is more than 50kms

4.9.2 Code

Following is the JavaScript code which is responsible for extracting the user's location.

Code in static/js/customJavascript.js

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```
function getLocation() {
  if (navigator.geolocation) {
navigator.geolocation.get Current Position (show Position, error);\\
  }else{
     alert("location not available");}}
function showPosition(position) {
  var hidden_location_field = document.getElementById("user_location");
  hidden_location_field.value
                                                   `${position.coords.latitude}--
${position.coords.longitude}`;
Code below is in <u>templates/events/events_list.html</u>
  <script type="text/javascript">
     getLocation();
  </script>
This code is at the bottom so it is called when all the above HTMLto suggest
elements are properly rendered.
There
                                 field inside the HTML. Code is
        is
                hidden
                          form
                                                                            in
             a
templates/events/events_list.html
<form>
<input type="hidden" id="user_location" name="user_location" value="_">
  <buttontype="submit">Get Local events</button>
</form>
```

When the page is rendered, the Html to that left-handthatDelhi javascript function extracts the latitude and longitude and replaces the value with those coordinates.

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```
<form>
    <input type="hidden" id="user_location" name="user_location"
value="19.2214739--72.9814762">
    <button class="btn" type="submit">Get Local events</button>
</form>
```

Like this and hence when the button is clicked the value is then shared with the server.

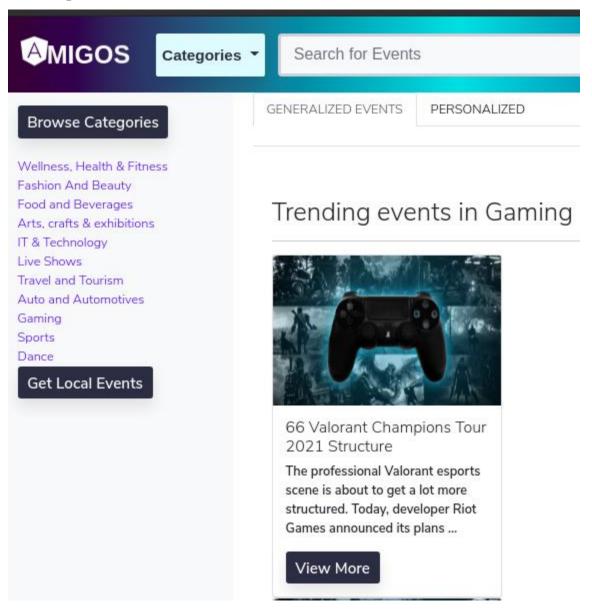
Code in *events/views.py*

```
if 'user_location' in self.request.GET:
    // for simplicity code is reduced. Refer the github repo.
user_clicks_recommends = user_clicks_recommends_temp
```

We don't want to restrict "online" events from the user hence that is excluded from the search. The code above is extra code that is added to remove events which are far away from the user.

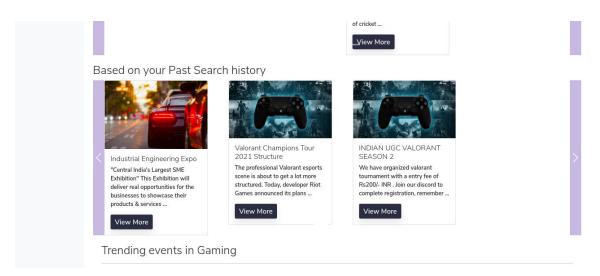
(Autonomous College Affiliated to University of Mumbai)

4.9.3 output

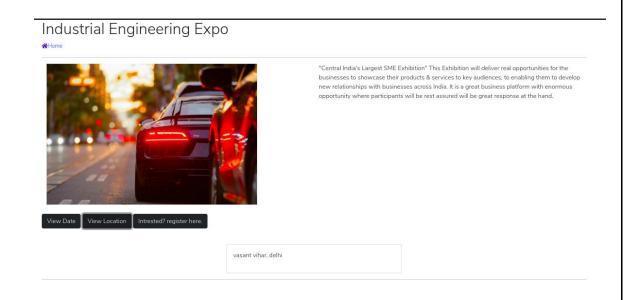


There's a button on the left hand side which says **Get Local Events**. When clicked the browser shares the location with the server.

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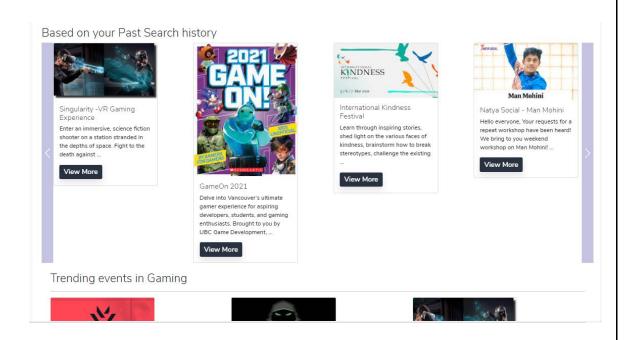


Based on past search history the event **Industrial Engineering expo** is shown to the user.



Here the venue for the event is **Vasant vihar**, **delhi** but the user's location is Thane and hence it is not possible for the user to visit that event. Hence when clicked on **getting Local Events**

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That event is no longer shown.

4.10 Testing

4.10.1 Black Box Testing

REGISTRATION OF USER AND EVENT ORGANIZER

→ Equivalence Class Partitioning to Verify UserID (only alphabets; 8-12 characters)

Valid	Invalid
a-z EC1	0-9 EC3
A-Z EC2	Special Chars EC4
-	Blank Field5 EC5

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TEST CASES-

TEST CASE ID	INPUT	EXPECTED OUTPUT	EQUIVALENCE CLASS-TESTED
#TC01	vishvesh	Registration will proceed	EC1
#TC02	SamikshA	Registration will proceed	EC2
#TC03	Vik09ram	Registration will not proceed	EC3
#TC04	@nz@l***	Registration will not proceed	EC4
#TC05	n id hi	Registration will not proceed	EC5

LOGIN

→ Equivalence Class Partitioning Verify Password (only alphabets ; 8-12 characters)

Valid	Invalid
a-z (EC1)	Blank Field (EC5)
A-Z (EC2)	
0 - 9 (EC3)	
Special Characters (EC4)	

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TEST CASES-

TEST CASE ID	INPUT	EXPECTED OUTPUT	EQUIVALENCE CLASS-TESTEDthat
#TC01	vishveshh	Registration will proceed	EC1
#TC02	SAmikSHA	Registration will proceed	EC2
#TC03	anzal123	Registration will proceed	EC3
#TC04	v!kram@5	Registration will proceed	EC4
#TC05	n! dh!5	Registration will not proceed	EC5

→ Boundary Value Analysis for Login

It accepts valid User Name and Password field and accepts minimum 8 characters and a maximum of 12 characters.

Valid range 8-12

Invalid range 7 or less than 7 and

Invalid range 13 or more than 13.

(BEST CASE)

Min = 8

Min+=9

Nom = 10

Max = 12

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Max - 11

N = 2 variables

(4n+1) = (4*2+1) = 9

Case	A	В	Expected Output
1	10	8	Valid
2	10	9	Valid
3	10	10	Valid
4	10	11	Valid
5	10	12	Valid
6	8	10	Valid
7	9	10	Valid
8	11	10	Valid
9	12	10	Valid

→ Decision Table for Login Screen (Username and Password)

- T Correct username/password
- F Wrong username/password
- E Error message is displayed
- H Home screen is displayed
 - 1) Rule 1 Username and password both were wrong. The user is shown an error message.

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- 2) Rule 2 The username of was correct, but the password was wrong. The user is shown an error message.
- 3) Rule 3 The username was wrong, but the password was correct. The user is shown an error message.
- 4) Rule 4 Username and password both were correct, and the user navigated to the homepage

Conditions	Rule 1	Rule 2	Rule 3	Rule 4
Username(T/F)	F	Т	F	Т
Password(T/F)	F	F	Т	Т
Actions				
Output(E/H)	Е	Е	Е	Н

EVENT LIST

There are a total 234 events stored in the database.



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→ ECP to check Event List:

TEST CASE ID	INPUT	EXPECTED OUTPUT	Remark
#TC01	0	404 Error	PASS
#TC02	1	Event 1 detail will be displayed	PASS
#TC03	115	Event 115 details will be displayed	PASS
#TC04	230	Event 230 details will be displayed	PASS
#TC05	231	404 Error	PASS

→ BVA to check Event List

Min = 1

Min+=2

Nom = 115

Max = 230

Max - 229

N = 1 variables

(4n+1) = (4*1+1) = 5 test cases

TEST CASE ID	INPUT	EXPECTED OUTPUT			Remark		
#TC01	1	Event	1	detail	will	be	PASS

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		displayed	
#TC02	2	Event 2 detail will be displayed	PASS
#TC03	115	Event 115 details will be displayed	PASS
#TC04	230	Event 230 details will be displayed	PASS
#TC05	229	Event 229 details will be displayed	PASS

BLACK BOX TESTING USING AUTOMATION TOOL

The tool used info, right Cypress.

```
New major version of npm available! 6.14.16 → 7.5.4
Changelog: https://github.com/npm/cli/releases/tag/v7.5.4
Run npm install -g npm to update!

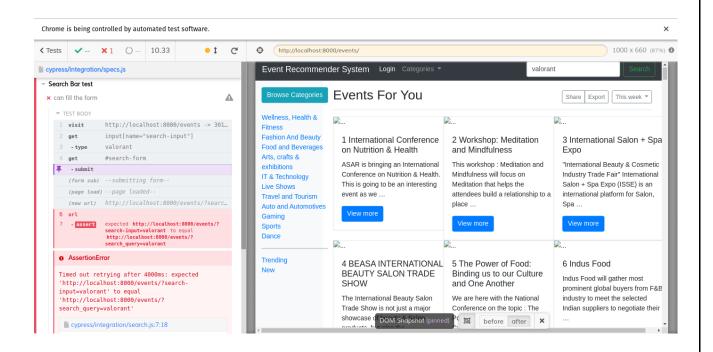
harrypop@harry-pop-os:~/Desktop/cypress_testing$ no
node node-gyp nodejs nodemon nohup nologin
harrypop@harry-pop-os:~/Desktop/cypress_testing$ ls
node_modules package.json package-lock.json
harrypop@harry-pop-os:~/Desktop/cypress_testing$ node_modules/
Display all 177 possibilities? (y or n)
harrypop@harry-pop-os:~/Desktop/cypress_testing$ node_modules/.bin/cypress open
It looks like this is your first time using Cypress: 6.5.0

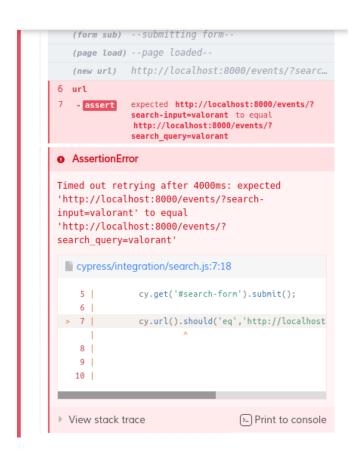
✓ Verified Cypress! /home/harrypop/.cache/Cypress/6.5.0/Cypress
```

Test Case 1:

The following test checks whether the **search query** provided in the website is reflected or not. After the search query is submitted the page is not reflected with the site. Hence the test case fails.

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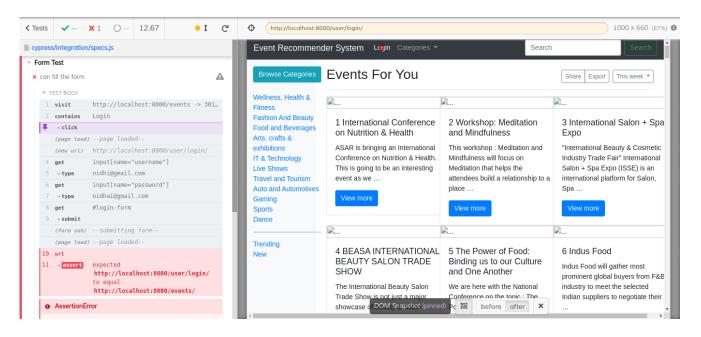
Following is the test code:

```
describe("Search Bar test",() =>{[
    it("can fill the form",() =>{
        cy.visit("http://localhost:8000/events");
        cy.get('input[name="search-input"]').type("valorant");
        cy.get('#search-form').submit();

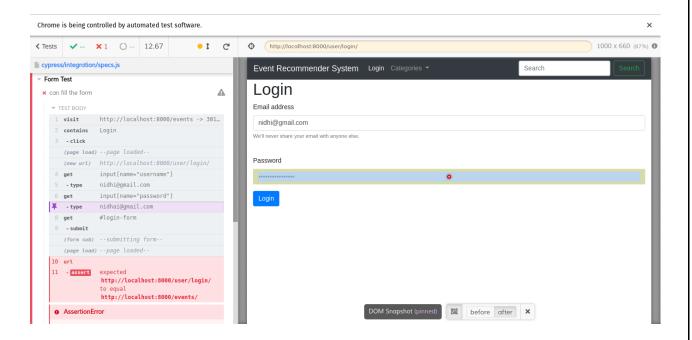
        cy.url().should('eq', 'http://localhost:8000/events/?search_query=valorant');
    })

})
```

Test Case 2:



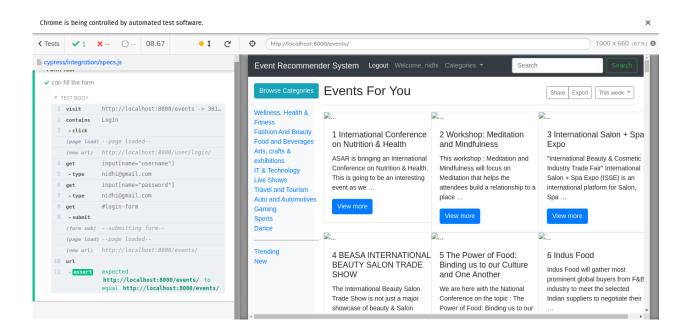
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The following test case 2 is to test whether if correct credentials are provided then it's reflected back or not. The username and password entered here are nid@gmail.com and nidha@gmail.com. Now for the sake of learning the tool, we have purposely entered the wrong data but if it was the right data and since it does not match the expected output on the website, the test fails.

Now below is an example of test which passes

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In this, when correct credentials are provided the website moves to the expected URL. Which the test recognizes and hence it shows the status that the test passed. Following is the code:

```
describe("Form Test",() =>{
    it("can fill the form",() =>{
        // cy.visit("http://localhost:8000/events");
        // cy.contains('Login').click();

    // cy.get('input[name="username"]').type("nidhi@gmail.com");
        // cy.get('input[name="password"]').type("nidhai@gmail.com");
        // cy.get('#login-form').submit();
        // cy.url().should('eq','http://localhost:8000/events/');

        cy.visit("http://localhost:8000/events");
        cy.contains('Login').click();

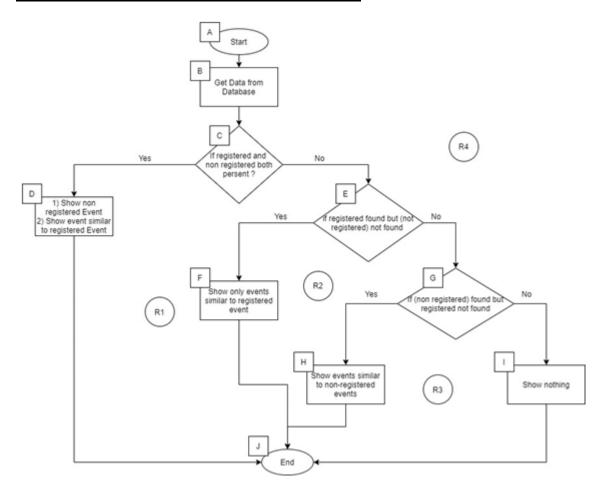
        cy.get('input[name="username"]').type("nidhi@gmail.com");
        cy.get('input[name="password"]').type("nidhi@gmail.com");
        cy.get('input[name="pass
```

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```
cy.get('#login-form').submit();
  cy.url().should('eq','http://localhost:8000/events/');
})
]})
```

4.10.2 White Box Testing

Recommendation based on past browsing patterns



→ Cyclomatic Complexity:

$$V(G) = E-N+2 = 12 - 10 + 2 = 4$$

$$V(G) = No.$$
 of Predicate Nodes $+1 = 3 + 1 = 4$

$$V(G) = No. of Regions = 4$$

→ Identification Of Different Paths-

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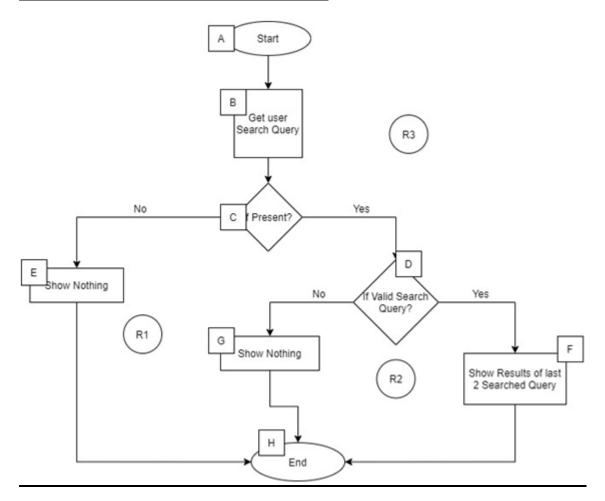
- A-B-C-D-J
- A-B-C-E-F-J
- A-B-C-E-G-H-J
- A-B-C-E-G-I-J

Test Cases

ID	Input Value	Expected Output	-	
1	List of events with values: Regis = 1 regis = 0	Show non registered event Show similar events based on registered event	Showed non registered event Showed similar events tp registered event	A-B-C-D-J
2	List of events with only regis = 1	Show similar events based on registered events	Showed similar events based on registered events	A-B-C-E-F-J
3	List of events with only regis = 0	Show similar events based on non registered events	Showed similar events based on non registered events	A-B-C-E-G-H-J
4	No events	Show nothing	Showed nothing	A-B-C-E-G-I-J

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Recommendation based on past search query



Cyclomatic Complexity:

$$V(G) = E-N+2 = 9 - 8 + 2 = 3$$

$$V(G) = No.$$
 of Predicate Nodes $+1 = 2 + 1 = 3$

$$V(G) = No. of Regions = 3$$

Identification Of Different Paths-

- A-B-C-E-H
- A-B-C-D-G-H
- A-B-C-D-F-H

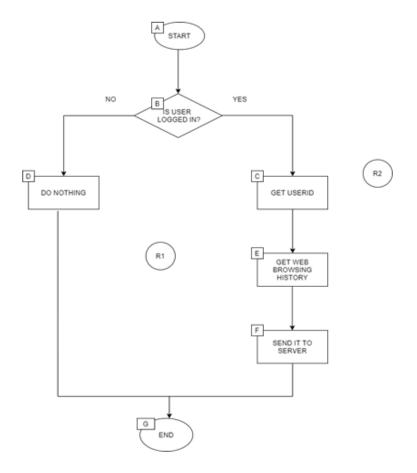
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Test Cases

ID	Input Value	Expected output	Actual output	Independent Path
1	No Data	Show nothing	Showed nothing	А-В-С-Е-Н
2	Invalid search query	Show nothing	Showed nothing	A-B-C-D-G-H
3	Valid Search query	Show results of the search query	Showed results of search query	A-B-C-D-F-H

Recommendation based on user's browsing history

Part 1



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Cyclomatic Complexity:

$$V(G) = E-N+2 = 7 - 7 + 2 = 2$$

$$V(G) = No.$$
 of Predicate Nodes $+1 = 1 + 1 = 2$

$$V(G) = No. of Regions = 2$$

Identification of Different Paths-

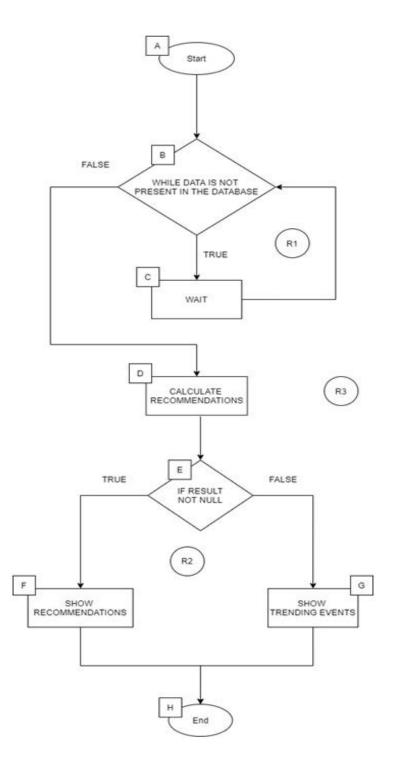
- A-B-D-G
- A-B-C-E-F-G

Test Cases

ID	Input Value	Expected output	Actual output	Independent Path
1	Valid User credentials	Passing history data to server	Passed history data to server	A-B-C-E-F-G
2	Not logged in	Do Nothing	Did nothing	A-B-D-G

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Part 2:



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Cyclomatic Complexity:

$$V(G) = E-N+2 = 9 - 8 + 2 = 3$$

$$V(G) = No.$$
 of Predicate Nodes $+1 = 2 + 1 = 3$

$$V(G) = No. of Regions = 3$$

Identification of Different Paths-

- A-B-D-E-F-H
- A-B-D-E-G-H
- A-B-C-B-D-E-F-H
- A-B-C-B-D-E-G-H

Test Cases -

ID	Input Values	Exptected output	Actual Output	Independent Path
1	Browsing before database is populated and no recommendation is calculated	Show nothing	Showed nothing	A-B-C-B-D- E-G-H
2	Browsing before database is populated and with calculated recommendation	Show nothing	Showed nothing	A-B-C-B-D- E-F-H
2	Browsing after data is populated but no recommendations calculated	Show nothing	Showed nothing	A-B-D-E-F-H

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3	Browsing after data is populated with calculated	Show recommendations	Showed recommendations	A-B-D-E-G- H
	recommendations			

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Chapter 5 : Conclusion

5.1 Conclusion

- Amigo-Event Recommendation System is a Personalized Event Recommendation System which makes use of concepts such as Cosine Similarity, TF-IDF algorithm which is used to Mine the user's interest from their web history to recommend them the events of their interests, the web history is extracted with the help of a web browser extension and then cleaned and processed accordingly to mine a user's interest.
- By default, any user who logins to the website is shown the list of Trending
 Events~ the events that are being registered the most by the current users of the
 System and also a list of events segregated categorically.
- A user is shown a list of Personalized Events based on his history and then if he/she
 is interested then a score is calculated which is further used for knowing up to
 which extent the user is interested in that event.
- From which the user is shown some unregistered events similar to those in which the user has explicitly shown interest.
- Apart from that, the user is also shown some events based on his search query on the website.
- An event organizer can register and post the events, which gets saved to our database from where they get into the list of events if the user is interested or if it belongs to any particular category.
- With the powerful location detection, up to the user's choice, can also view the events which are close to the user. (Within 50kms).

5.2 Scope for further work

5.2.1 Friends Concept

Our Project can be tuned and merged to further include the Friend's Concept.

The friend's concept includes the logic of suggesting the event attended by your friends. If you and your friend have the same taste of events and if your cosine similarity of events

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match along with the number of past events that you have attended together, then you are recommended the events attended by your friends and vice versa.

5.2.2 Intra-Somaiya Network

We can extend our project to make it functional for the Intra-Somaiya network. The idea is to make this project limited to the Intra-somaiya network which includes all the colleges under the Somaiya domain, so that any tech-events, lectures, exhibitions, competitions can be reachable to students and most importantly, it would only recommend the events according to the student's interest.

5.2.3 Large scale Implementation

If full-scale implementation istech events and our project is made more optimal and further performance-enhanced then, it can be used as a direct competitor to BookMyShow and MagicPin app, as it includes the main features of BookMyShow and MagicPin App along with suggesting only those events in which the user is interested in.

5.2.4 Adding events live to the website

Currently, all the events are taken from the CSV file. To add events live and dynamically by having event organizers register on our website and then update their events is out of the scope of this project. The Scope of this project is mainly to provide recommendations to the user using multiple techniques. Hence in future this can be elaborated and extended by having a live page for Event Organizers to register themselves and upload events on our website.

5.3 References

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