

**Team ID: PNT2022TMID20463**

**Project Name:** SMART FASHION RECOMMENDER APPLICATION

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**Team Member 1:** Menati Vasudeva Reddy

**Team Member 2:** Loghadharshann D

**Team Member 3:** Rajavignesh M

## **Abstract**

Nowadays, fashion applications and e-commerce are growing more and more. And it also has some problems when finding the customer's wanted product in the web applications. Having a chatbot that understands the algorithm of a specific application can be of great aid. We are implementing such a chat bot in a web application, which is fed with the knowledge of the application's algorithm and helps the user completely from finding their needs to processing the payment and initiating delivery. It works as an advanced filter search that can bring the user what they want with the help of pictorial and named representation by getting simple user information and activities. The application also has two main UI interactions: one is the user panel and the other one is the admin panel. Users can interact with the chat bot to search for products, order them from the manufacturer or distributor through chatbot AI, and it can also make payment transactions, track the delivery, and so on. The admin interface enables the user to upload products' details ,user details, orders and find how many products have been bought, supervise the stock availability, and interact with the buyer regarding the product reviews.

## **Features of Chatbot :**

- Using chatbot we can manage user's choices and orders.
- The chatbot can give recommendations to the users based on their interests.
- It can promote the best deals and offers on that day.
- It will store the customer's details and orders in the database.
- The chatbot will send a notification to customers if the order is confirmed.
- Chatbots can also help in collecting customer feedback.

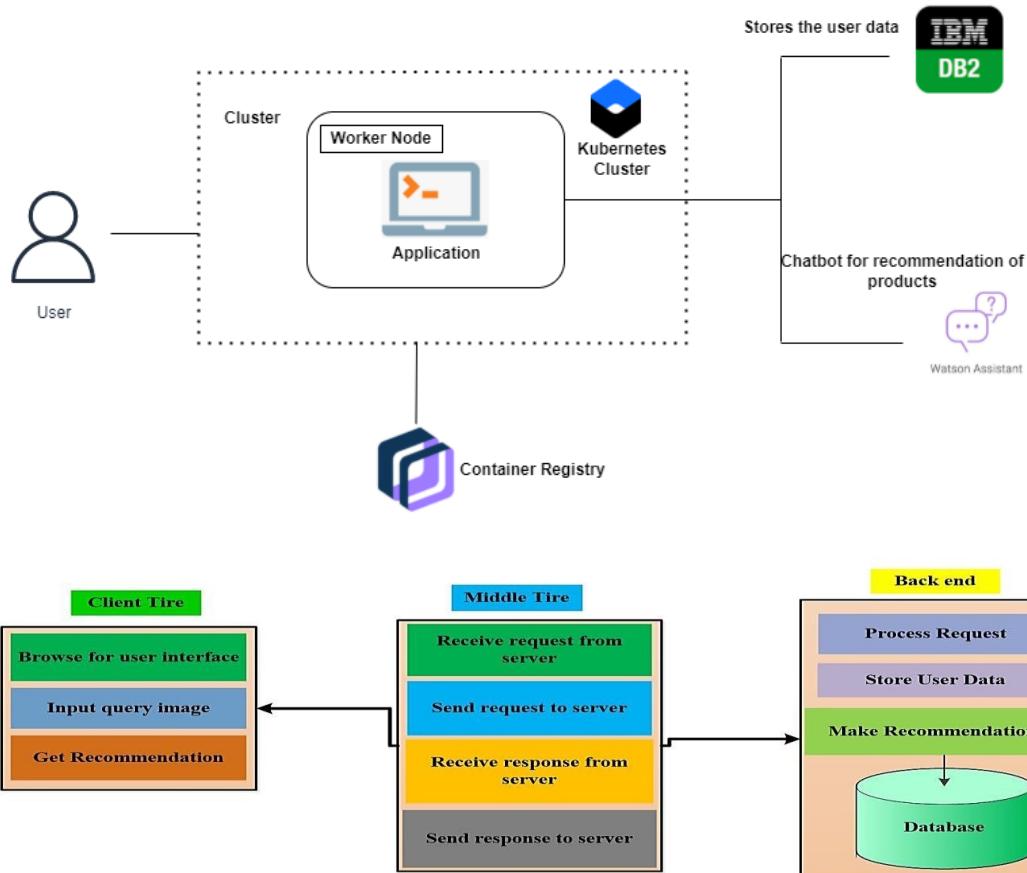
# **Introduction**

During the last years, online shopping has been growing. Both the customer and the business enterprise desire the client to easily discover applicable products or items both throughout search and when they are searching, and this is where recommender systems come into the picture. Recommendation systems make recommendations based on the information they are provided with and in the manner in which they are programmed. Going into details, most of the evaluation applied is independent coming up with a brand-new recommendation algorithm, system, or model. However, different researchers use already existing work as researchers use an already existing current piece of work to come up with a new diagram or to truly improve the current one. The present analysis model focuses on the use of a current algorithmic program and, consequently, the use of a new research concept comes up with a recommender system. Existing research and fashions have given us some inspirations of how to design fashion recommendation systems. The contribution of the research are follows:

- A scheme for improving a person's clothing style by removing the features he/she doesn't like from his/her clothing images.
- These attributes served to a similar model to retrieves similar images as recommendations.
- Combined with more common content-based recommendation systems, our model can help to extend robustness and performance, for example can suit a more pretentious style of a client.

## ● Proposed system architecture

The system architecture defines the hardware, software and network environment of the structure. The system will be web-based meaning that the users need to run the URL in order to run the system. The system will run both horizontally and vertically. The architecture used in the system is shown horizontally where the Model View Controller is explained as represented in Figure 2. The high-level part of the system is looked at using the vertical way.



The system comprises of the Client tire, which is the front end or View mode, middle tier which is the system controller and the backend tire which is the model. The client side is where the users/customers log in in the system, browse for the system interface, provide input query image to the system, and get recommendation according to the input query. The middle true is responsible for communication between the front end and the back end. It receives user requests and sends them to the back end and in turn accepts

responses from the backend and sends them to the user.

The back end which involves the data set and recommender algorithm deals with data storage, user input data storage, processing user requests, determining user input similarity, making recommendations and forwarding them to the middle tier which in turn sends them to the respective users. The internet works to provide access to the site with a strong security check, provided by both firewall and password protection policy. Any unauthorized access is detected and prevented by the firewall.

## **1.2 PURPOSE**

Clothing is a kind of symbol that represents people's internal perceptions through their outer appearance. It conveys information about their choices, faith, personality, profession, social status, and attitude towards life. Therefore, clothing is believed to be a nonverbal way of communicating and a major part of people's outer appearance. Recent technological advancements have enabled consumers to track current fashion trends around the globe, which influence their choices [2,3]. The fashion choices of consumers depend on many factors, such as demographics, geographic location, individual preferences, interpersonal influences, age, gender, season, and culture.

## **CHAPTER 2**

### **LITERATURE REVIEW**

[1] devised a parametric distance transformation that assigns a lower distance to garment pairings that fit well than to those that do not. And provided Image-based recommendations on styles and substitutes.

[2] conducted a preliminary investigation into personalised outfit recommendation. To

describe the user-item and item-item interactions, a functional tensor factorization method was presented. They proposed A functional tensor factorization approach. Veit et al.

[4] learned feature transformation for a compatibility measure between pairs of objects using a Siamese CNN architecture. All of these works focused solely on the compatibility of two things. Furthermore, they simply modelled broad matching criteria and ignored the issue of personalisation.

Thombre in [3] used image segmentation and Kalman filter to realize Human detection and tracking. Orrite-Urunuela proposed a statistical model for detection and tracking of human silhouette and the corresponding 3D skeletal structure in gait sequences [5]. How-Lung [6] provided an outdoor aquatic surveillance system for human motion tracking and detection. Ajmani et al.

[7] present a novel method for content based recommendation of media-rich commodities with the use of probabilistic multimedia ontology. Proposed an ontology based personalized garment recommendation system.

## CHAPTER 3

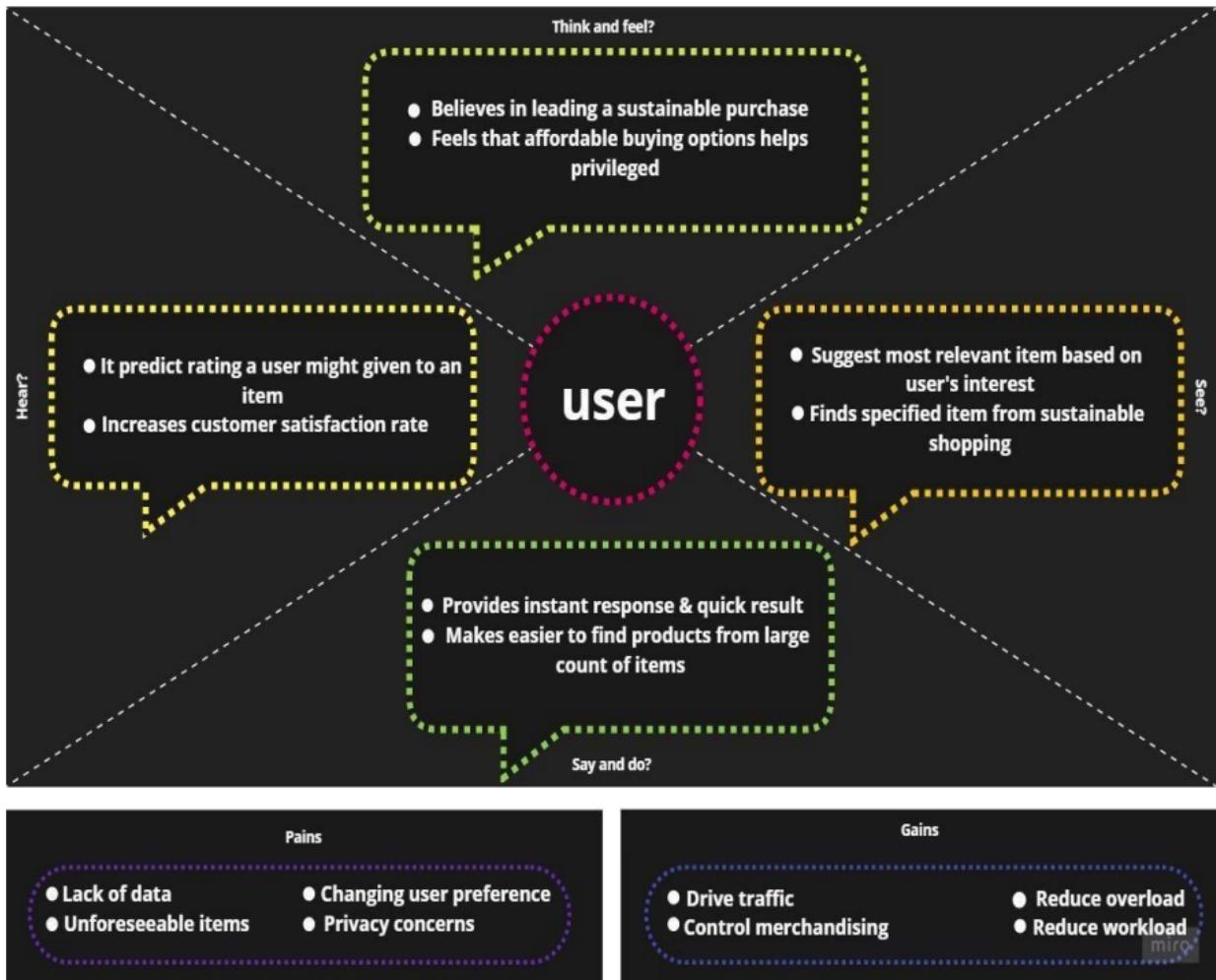
### IDEATION & PROPOSED SOLUTION

#### 3.1 Empathy Map Canvas

##### Goals and Objectives

- Recommendation system is to suggest relevant items to the user.
- To predict user's interest and recommend relevant items that quietly likely interesting

for them.



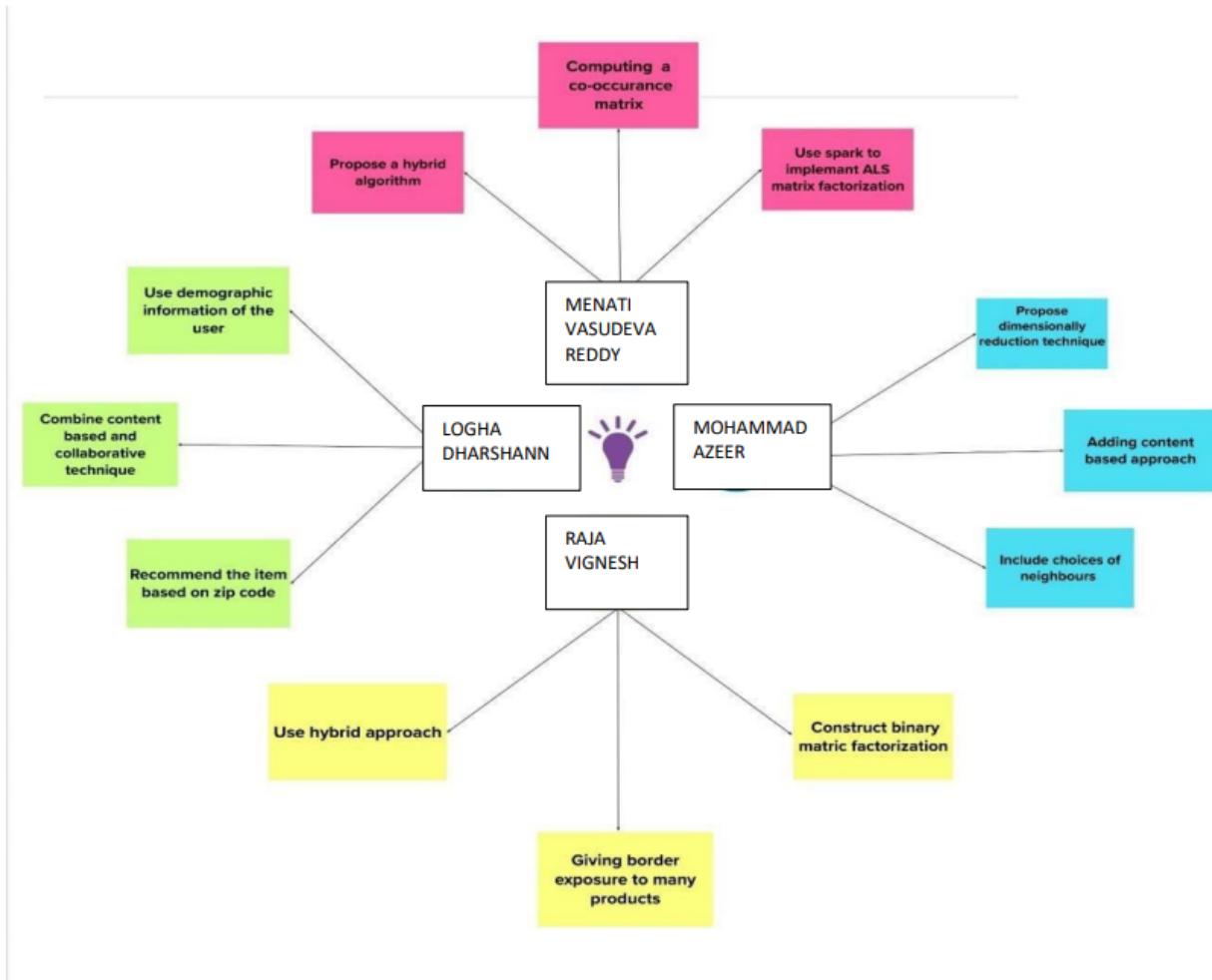
### 3.2 Ideation & Brainstorming

#### Step-1: Team Gathering, Collaboration and select the Problem statement

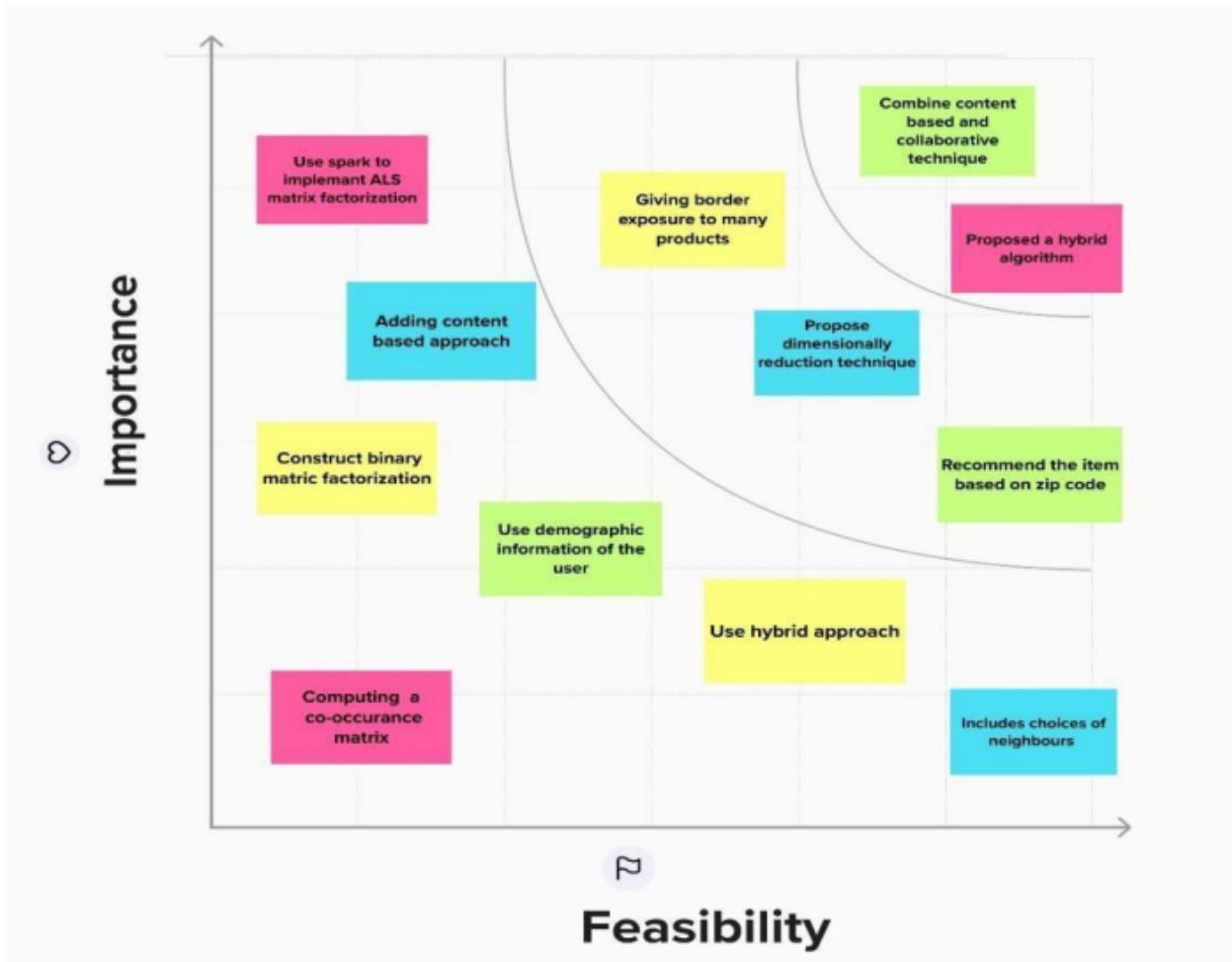
- o Team Members – MOHAMMED AZEER, MENATI VASUDEVA REDDY, LOGHADHARSHANN D, RAJAVIGNESH M
- o Goal – sorting data sparsity and lack of data analytics problem
- o Problem statement

- In recommendation system, sometimes recommend different items than user's preference.
- Privacy with minimum imposition of accuracy loss on the recommendation.

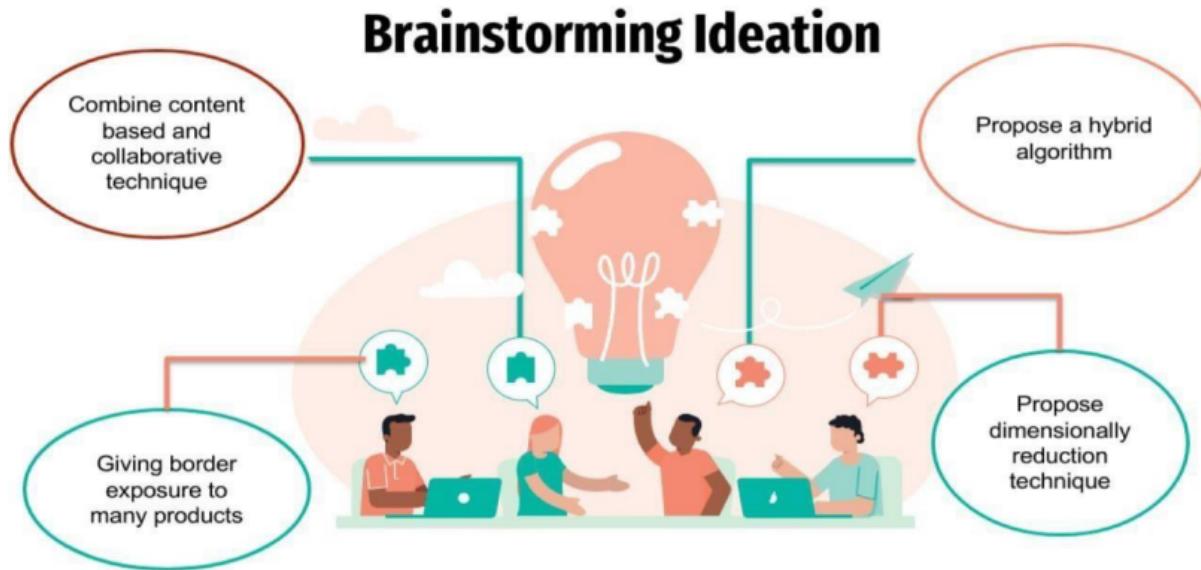
## Step-2: Brainstorm, Idea Listing and Grouping



### step-3: Idea Prioritization



## Brainstorming Ideation



### 3.3 Proposed Solution

#### 1) Problem Statement (Problem to be solved)

Improving users privacy with minimum imposition of accuracy loss on the recommendation.

- In the recommendation system the problem is trying to forecast the opinion of user will have on dissimilar substance and be able to recommend the finest items to each user.
- Another problems are data sparsity, gray sheep and scalability.
- Lack of data analytic capability.

#### 2. Idea / Solution description

- System will recommend the items based on the zip code.
- As the user click on link of any item a time session will be started to record how much time he has spent on particular page.

- When the time spent by the user crosses a certain threshold time, rating of the particular item will be increased by some measures.
- Distributed aggregation of user's profile.

### **3. Novelty / Uniqueness**

- Recommender system are very powerful to help a user find good products or items. the important thing the goals of recommender systems are to serve information

### **4. Social Impact / Customer Satisfaction**

- Recommender systems has consistently suggested that customer satisfaction will be highest when the recommendation algorithm is accurate and recommends a diversity of items.

### **5. Business Model (Revenue Model)**

- Researchers have studied and generate many algorithms to learn increasing rate for an online customer like Amazon site. Also, These algorithms study the difference between shopping online sites with others using recommendersystems for items to increase revenue by increasing the number of sales.

### **6. Scalability of the Solution**

- A recommendation technique that is efficient when the number of dataset is limited may be unable to generate satisfactory number of recommendations when the volume of dataset is increased

### 3.4 Problem Solution fit

<b>1 Customer Segment(s)</b>  CS <p>Retailers (merchant) &amp; person who sells and buy products online.</p>	<b>6 Customer Limitations</b> EG, Budget, Device  CL <p>Collaborative filtering is unable to discover the latent association between synonyms, so it will treat these products differently.</p>	<b>5 Available Solutions</b> Pros & Cons  AS <p><b>PROS</b></p> <p>Easy recommendations make less searches and some times end up good deals</p> <p>Speed up the process of decision and purchase based on the previous statistics</p> <hr/> <p><b>CONS</b></p> <p>If the system recommends products with bias, then customer will be landing into wrong deals</p> <p>Chances are that some websites may suggest products wrongly based on analysis of little information gathered</p>
<b>2 Problems \ Pains</b> It's Frequency  PR <ul style="list-style-type: none"> <li>Lack of data</li> <li>Unforeseeable items</li> <li>Privacy concerns</li> <li>Changing user preference</li> </ul>	<b>9 Problem Root \ Cause</b>  RC <p>Some offer up too many 'lowest common denominator' recommendations, some don't support The Long Tail enough and just recommend obvious items, outliers can be a problem</p>	<b>7 Behaviour</b> It's Intensity  BE <p>Algorithm also includes a way of providing implicit ratings considering the users' movements after receiving recommendations, aimed at measuring the users' interest for the recommended items. Conducted experiments measure the effectiveness and the efficiency of our recommender algorithm, as well as the impact of implicit ratings.</p>
<b>3 Trigger to act</b>  TR <ul style="list-style-type: none"> <li>Social proof.</li> <li>Usage.</li> <li>Ads</li> <li>Scarcity.</li> </ul> <hr/> <b>4 Emotions</b>  EM <p>Before Disappointed, Disgruntled, Frustrated</p> <p>After Gratitude, Fulfilled</p>	<b>10 Your Solution</b>  SL <ul style="list-style-type: none"> <li>Combine content based and collaborative technique</li> <li>Propose a hybrid algorithm</li> <li>Giving border exposure to many product</li> <li>Propose dimensionality reduction technique</li> </ul>	<b>8 Channels of Behaviour</b>  CH <p><b>Online</b> software that analyzes available data to make suggestions for something that a website user might be interested in</p> <hr/> <p><b>Offline</b> data is used to estimate how a user might have reacted to a different set of recommendations placed in front of them at a certain point in time, by using the knowledge of what they really did react to later.</p>

## Solution architecture



## CHAPTER 4

### REQUIREMENT ANALYSIS

#### 4.1 Functional requirement

Following are the functional requirements of the proposed solution.

FR No.	Functional Requirement (Epic)	Sub Requirement (Story / Sub-Task)
FR-1	User Registration	Registration through mail Registration through Gmail
FR-2	User Confirmation	Confirmation via Email Confirmation via OTP
FR-3	User Login	Login using username & Password
FR-4	Personal Details	Personal details through Form Personal details through UI Tab
FR-5	Delivery Confirmation	Confirmation via Email Confirmation via Phone

#### 4.2 Non-Functional requirements

Following are the non-functional requirements of the proposed solution.

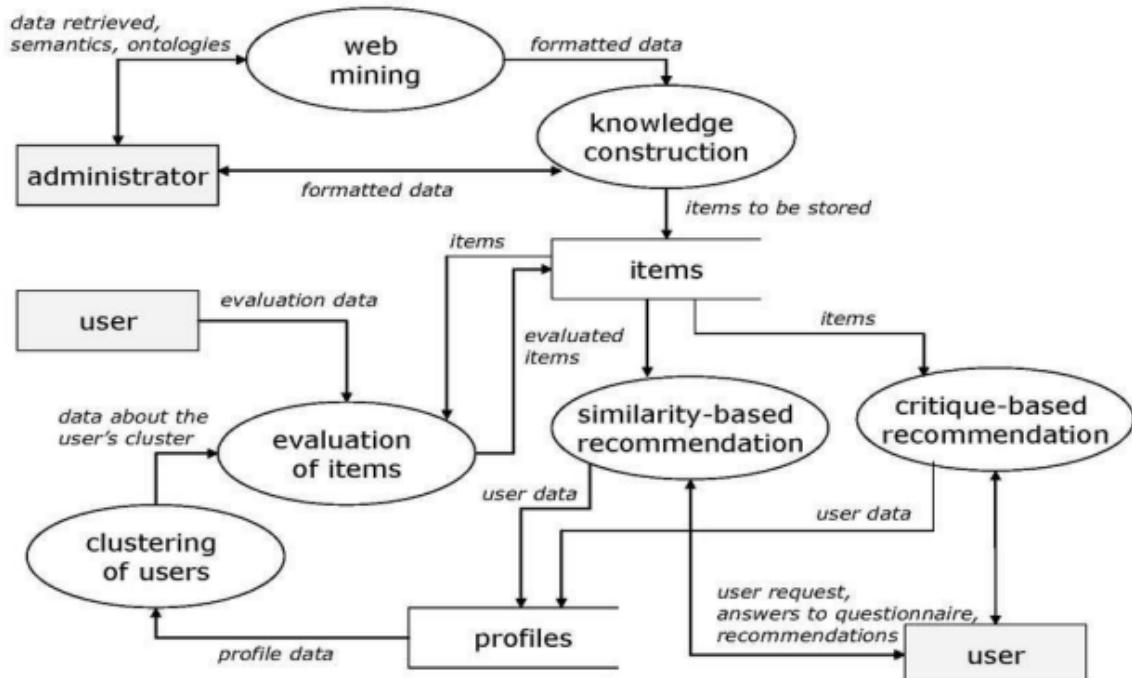
FR No.	Non-Functional Requirement	Description
NFR-1	<b>Usability</b>	Ease of use of the application for the user
NFR-2	<b>Security</b>	User privacy is the highest priority of the application. Security measures are undertaken for the user
NFR-3	<b>Reliability</b>	It can handle more than 2000 users at a time. It can process and initialize most functions.
NFR-4	<b>Performance</b>	The application can handle complex tasks and supports multi-tasking.
NFR-5	<b>Availability</b>	It is a free web and application available on all platforms.
NFR-6	<b>Scalability</b>	With higher workloads the user will experience a 10 to 17% drop in performance.

## CHAPTER 5

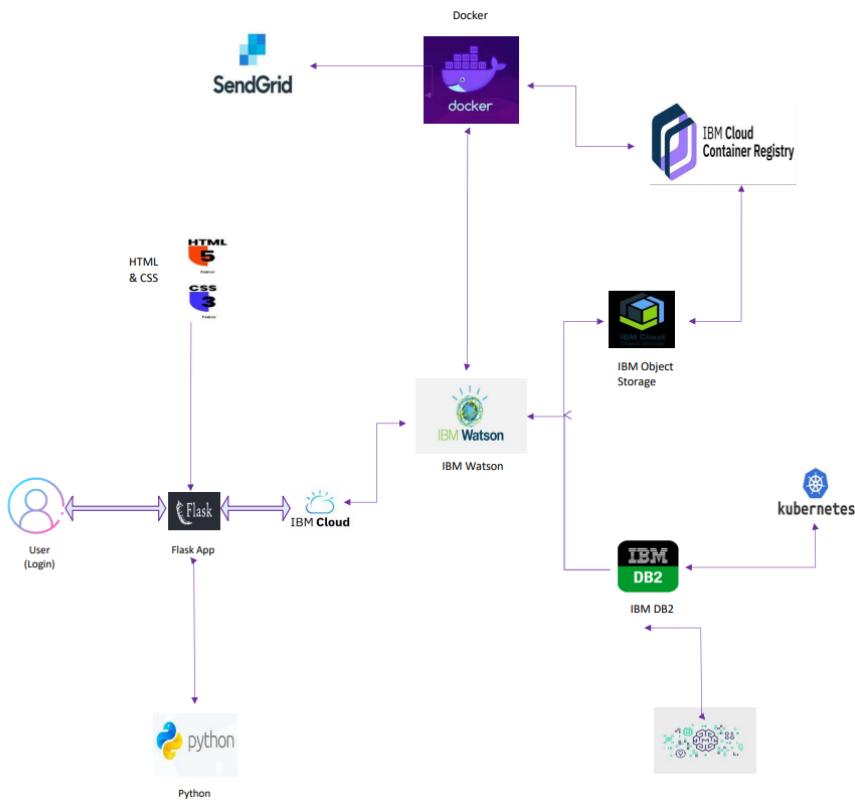
### PROJECT DESIGN

#### 5.1 Data Flow Diagrams

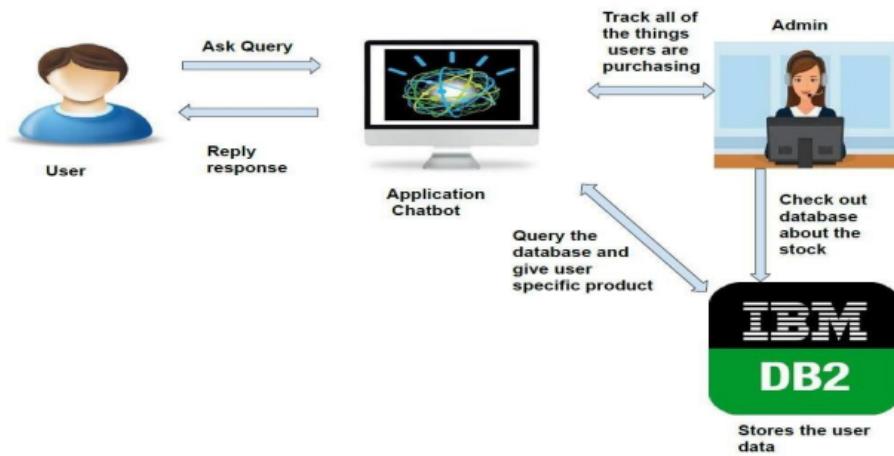
DFD of Fashion Recommender (Industry Standard)



## 5.2 Solution & Technical Architecture



## 5.3 User Stories



## CHAPTER 6

### PROJECT PLANNING & SCHEDULING

#### 6.1 Sprint Planning & Estimation

Sprint	Functional Requirement (Epic)	User Story Number	User Story / Task	Story points	Priority	Team Members
Sprint-1	Setting up App environment	USN-1	As a user, I can register in ICTA Academy and create IBM cloud account.	2	High	MENATI VASUDEVA REDDY MOHAMMAD AZEER
Sprint-1		USN-2	As a user, I will create a flask project	1	Low	LOGHA DHARSHANN D RAJA VIGNESH M
Sprint-1		USN-3	As a user, I will install IBM CloudCLI	2	Medium	LOGHA DHARSHANN D MENATI VASUDEVA REDDY
Sprint-2	Setting up App environment	USN-4	As a user, I can install Docker CLI	1	Low	MOHAMMAD AZEER RAJA VIGNESH M
Sprint-2		USN-5	As a user, I will Create an accountin sendgrid	2	Medium	LOGHA DHARSHANN D MOHAMMAD AZEER

Sprint-3	Implementing web application	USN-6	As a user, I Create UI to interactwith the application	1	High	MENATI VASUDEVA REDDY MOHAMMAD AZEER
Sprint-3		USN-7	As a user, I Create IBM DB2 and connect with Python	3	High	LOGHA DHARSHANN D
Sprint-3	Integrating sendgridservice	USN-8	As a user, I will integrating sendgridwith python code	2	High	MOHAMMAD AZEER
Sprint-3	Developing a chatbot	USN-9	As a user, I have to build a chatbotand Integrate to application	1	Medium	MENATI VASUDEVA REDDY
Sprint-4	Development of App inIBM Cloud	USN-10	As a user, I will Containerize theApp	1	Low	RAJA VIGNESH M
Sprint-4		USN-11	As a user, I will upload image toIBM Container registry	2	Medium	LOGHA DHARSHANN D
Sprint-4		USN-12	As a user, I will deploy App in Kebernetes cluster	3	High	MOHAMMAD AZEER
Sprint-4	User panel		As a user ● Register, Login, Email, Verification ● Manual Search ● Order placement, Order Details	3	High	MENATI VASUDEVA REDDY LOGHA DHARSHANN D RAJA VIGNESH M

## 6.2 Sprint Delivery Schedule

Sprint	Total Story Points	Duration	Sprint Start Date	Sprint End Date (Planned)	Story Points Completed (as on Planned End Date)	Sprint Release Date (Actual)
Sprint-1	18	6 Days	24 Oct 2022	29 Oct 2022	24	29 Oct 2022
Sprint-2	18	6 Days	31 Oct 2022	05 Nov 2022	24	05 Nov 2022
Sprint-3	18	6 Days	07 Nov 2022	12 Nov 2022	24	12 Nov 2022
Sprint-4	18	6 Days	14 Nov 2022	19 Nov 2022	24	19 Nov 2022

## 7. CODING & SOLUTIONING (Explain the features added in the project along with code)

### 7.1 Features

```
from cs50 import SQL
from flask_session import Session
from flask import Flask, render_template, redirect, request, session, jsonify
from datetime import datetime

## Instantiate Flask object named app
app = Flask(__name__)

## Configure sessions
app.config["SESSION_PERMANENT"] = False
app.config["SESSION_TYPE"] = "filesystem"
Session(app)

# Creates a connection to the database
db = SQL( "sqlite:///data.db" )

@app.route("/")
def index():
    shirts = db.execute("SELECT * FROM shirts ORDER BY onSalePrice")
```

```

shirtsLen = len(shirts)
# Initialize variables
shoppingCart = []
shopLen = len(shoppingCart)
totItems, total, display = 0, 0, 0
if 'user' in session:
    shoppingCart = db.execute("SELECT samplename, image, SUM(qty), SUM(subTotal), price, id FROM
cart GROUP BY samplename")
    shopLen = len(shoppingCart)
    for i in range(shopLen):
        total += shoppingCart[i]["SUM(subTotal)"]
        totItems += shoppingCart[i]["SUM(qty)"]
    shirts = db.execute("SELECT * FROM shirts ORDER BY onSalePrice ASC")
    shirtsLen = len(shirts)
    return render_template ("index.html", shoppingCart=shoppingCart, shirts=shirts, shopLen=shopLen,
shirtsLen=shirtsLen, total=total, totItems=totItems, display=display, session=session )
    return render_template ( "index.html", shirts=shirts, shoppingCart=shoppingCart, shirtsLen=shirtsLen,
shopLen=shopLen, total=total, totItems=totItems, display=display)

```

```

@app.route("/buy/")
def buy():
    # Initialize shopping cart variables
    shoppingCart = []
    shopLen = len(shoppingCart)
    totItems, total, display = 0, 0, 0
    qty = int(request.args.get('quantity'))
    if session:
        # Store id of the selected shirt
        id = int(request.args.get('id'))
        # Select info of selected shirt from database
        goods = db.execute("SELECT * FROM shirts WHERE id = :id", id=id)
        # Extract values from selected shirt record
        # Check if shirt is on sale to determine price
        if(goods[0]["onSale"] == 1):
            price = goods[0]["onSalePrice"]

```

```

else:
    price = goods[0]["price"]
    samplename = goods[0]["samplename"]
    image = goods[0]["image"]
    subTotal = qty * price
    # Insert selected shirt into shopping cart
    db.execute("INSERT INTO cart (id, qty, samplename, image, price, subTotal) VALUES (:id, :qty,
:samplename, :image, :price, :subTotal)", id=id, qty=qty, samplename=samplename, image=image,
price=price, subTotal=subTotal)
    shoppingCart = db.execute("SELECT samplename, image, SUM(qty), SUM(subTotal), price, id FROM
cart GROUP BY samplename")
    shopLen = len(shoppingCart)
    # Rebuild shopping cart
    for i in range(shopLen):
        total += shoppingCart[i]["SUM(subTotal)"]
        totItems += shoppingCart[i]["SUM(qty)"]
    # Select all shirts for home page view
    shirts = db.execute("SELECT * FROM shirts ORDER BY samplename ASC")
    shirtsLen = len(shirts)
    # Go back to home page
    return render_template ("index.html", shoppingCart=shoppingCart, shirts=shirts, shopLen=shopLen,
shirtsLen=shirtsLen, total=total, totItems=totItems, display=display, session=session )

```

```

@app.route("/update/")
def update():
    # Initialize shopping cart variables
    shoppingCart = []
    shopLen = len(shoppingCart)
    totItems, total, display = 0, 0, 0
    qty = int(request.args.get('quantity'))
    if session:
        # Store id of the selected shirt
        id = int(request.args.get('id'))
        db.execute("DELETE FROM cart WHERE id = :id", id=id)
        # Select info of selected shirt from database
        goods = db.execute("SELECT * FROM shirts WHERE id = :id", id=id)

```

```

# Extract values from selected shirt record
# Check if shirt is on sale to determine price
if(goods[0]["onSale"] == 1):
    price = goods[0]["onSalePrice"]
else:
    price = goods[0]["price"]
samplename = goods[0]["samplename"]
image = goods[0]["image"]
subTotal = qty * price
# Insert selected shirt into shopping cart
db.execute("INSERT INTO cart (id, qty, samplename, image, price, subTotal) VALUES (:id, :qty,
:samplename, :image, :price, :subTotal)", id=id, qty=qty, samplename=samplename, image=image,
price=price, subTotal=subTotal)
shoppingCart = db.execute("SELECT samplename, image, SUM(qty), SUM(subTotal), price, id FROM
cart GROUP BY samplename")
shopLen = len(shoppingCart)
# Rebuild shopping cart
for i in range(shopLen):
    total += shoppingCart[i]["SUM(subTotal)"]
    totItems += shoppingCart[i]["SUM(qty)"]
# Go back to cart page
return render_template ("cart.html", shoppingCart=shoppingCart, shopLen=shopLen, total=total,
totItems=totItems, display=display, session=session )

@app.route("/filter/")
def filter():
    if request.args.get('typeClothes'):
        query = request.args.get('typeClothes')
        shirts = db.execute("SELECT * FROM shirts WHERE typeClothes = :query ORDER BY samplename
ASC", query=query )
        if request.args.get('sale'):
            query = request.args.get('sale')
            shirts = db.execute("SELECT * FROM shirts WHERE onSale = :query ORDER BY samplename ASC",
query=query)
        if request.args.get('id'):
            query = int(request.args.get('id'))

```

```

shirts = db.execute("SELECT * FROM shirts WHERE id = :query ORDER BY samplename ASC",
query=query)

if request.args.get('kind'):
    query = request.args.get('kind')
    shirts = db.execute("SELECT * FROM shirts WHERE kind = :query ORDER BY samplename ASC",
query=query)

if request.args.get('price'):
    query = request.args.get('price')
    shirts = db.execute("SELECT * FROM shirts ORDER BY onSalePrice ASC")

shirtsLen = len(shirts)

# Initialize shopping cart variables
shoppingCart = []
shopLen = len(shoppingCart)
totItems, total, display = 0, 0, 0

if 'user' in session:
    # Rebuild shopping cart
    shoppingCart = db.execute("SELECT samplename, image, SUM(qty), SUM(subTotal), price, id FROM
cart GROUP BY samplename")
    shopLen = len(shoppingCart)
    for i in range(shopLen):
        total += shoppingCart[i]["SUM(subTotal)"]
        totItems += shoppingCart[i]["SUM(qty)"]

    # Render filtered view
    return render_template ("index.html", shoppingCart=shoppingCart, shirts=shirts, shopLen=shopLen,
shirtsLen=shirtsLen, total=total, totItems=totItems, display=display, session=session )

    # Render filtered view
    return render_template ( "index.html", shirts=shirts, shoppingCart=shoppingCart, shirtsLen=shirtsLen,
shopLen=shopLen, total=total, totItems=totItems, display=display)

```

```

@app.route("/checkout/")
def checkout():

    order = db.execute("SELECT * from cart")

    # Update purchase history of current customer
    for item in order:
        db.execute("INSERT INTO purchases (uid, id, samplename, image, quantity) VALUES(:uid, :id,

```

```
:samplename, :image, :quantity)", uid=session["uid"], id=item["id"], samplename=item["samplename"],  
image=item["image"], quantity=item["qty"] )
```

```
# Clear shopping cart
```

```
db.execute("DELETE from cart")
```

```
shoppingCart = []
```

```
shopLen = len(shoppingCart)
```

```
totItems, total, display = 0, 0, 0
```

```
# Redirect to home page
```

```
return redirect('/')
```

```
@app.route("/remove/", methods=["GET"])
```

```
def remove():
```

```
# Get the id of shirt selected to be removed
```

```
out = int(request.args.get("id"))
```

```
# Remove shirt from shopping cart
```

```
db.execute("DELETE from cart WHERE id=:id", id=out)
```

```
# Initialize shopping cart variables
```

```
totItems, total, display = 0, 0, 0
```

```
# Rebuild shopping cart
```

```
shoppingCart = db.execute("SELECT samplename, image, SUM(qty), SUM(subTotal), price, id FROM cart
```

```
GROUP BY samplename")
```

```
shopLen = len(shoppingCart)
```

```
for i in range(shopLen):
```

```
    total += shoppingCart[i]["SUM(subTotal)"]
```

```
    totItems += shoppingCart[i]["SUM(qty)"]
```

```
# Turn on "remove success" flag
```

```
display = 1
```

```
# Render shopping cart
```

```
return render_template ("cart.html", shoppingCart=shoppingCart, shopLen=shopLen, total=total,  
totItems=totItems, display=display, session=session )
```

```
@app.route("/login/", methods=["GET"])
```

```
def login():
```

```
    return render_template("login.html")
```

```

@app.route("/new/", methods=["GET"])
def new():
    # Render log in page
    return render_template("new.html")

@app.route("/logged/", methods=["POST"] )
def logged():
    # Get log in info from log in form
    user = request.form["username"].lower()
    pwd = request.form["password"]
    #pwd = str(hashlib.sha1(request.form["password"].encode('utf-8')).hexdigest())
    # Make sure form input is not blank and re-render log in page if blank
    if user == "" or pwd == "":
        return render_template ( "login.html" )

    # Find out if info in form matches a record in user database
    query = "SELECT * FROM users WHERE username = :user AND password = :pwd"
    rows = db.execute ( query, user=user, pwd=pwd )

    # If username and password match a record in database, set session variables
    if len(rows) == 1:
        session['user'] = user
        session['time'] = datetime.now( )
        session['uid'] = rows[0]['id']

    # Redirect to Home Page
    if 'user' in session:
        return redirect ( "/" )

    # If username is not in the database return the log in page
    return render_template ( "login.html", msg="Wrong username or password." )

@app.route("/history/")
def history():
    # Initialize shopping cart variables
    shoppingCart = []
    shopLen = len(shoppingCart)
    totItems, total, display = 0, 0, 0
    # Retrieve all shirts ever bought by current user

```

```

myShirts = db.execute("SELECT * FROM purchases WHERE uid=:uid", uid=session["uid"])
myShirtsLen = len(myShirts)
# Render table with shopping history of current user
return render_template("history.html", shoppingCart=shoppingCart, shopLen=shopLen, total=total,
totItems=totItems, display=display, session=session, myShirts=myShirts, myShirtsLen=myShirtsLen)

@app.route("/logout/")
def logout():
    # clear shopping cart
    db.execute("DELETE from cart")
    # Forget any user_id
    session.clear()
    # Redirect user to login form
    return redirect("/")

@app.route("/register/", methods=["POST"])
def registration():
    # Get info from form
    username = request.form["username"]
    password = request.form["password"]
    confirm = request.form["confirm"]
    fname = request.form["fname"]
    lname = request.form["lname"]
    email = request.form["email"]
    # See if username already in the database
    rows = db.execute( "SELECT * FROM users WHERE username = :username ", username = username )
    # If username already exists, alert user
    if len( rows ) > 0:
        return render_template ( "new.html", msg="Username already exists!" )
    # If new user, upload his/her info into the users database
    new = db.execute ( "INSERT INTO users (username, password, fname, lname, email) VALUES (:username,
:password, :fname, :lname, :email)",
                     username=username, password=password, fname=fname, lname=lname, email=email )
    # Render login template
    return render_template ( "login.html" )

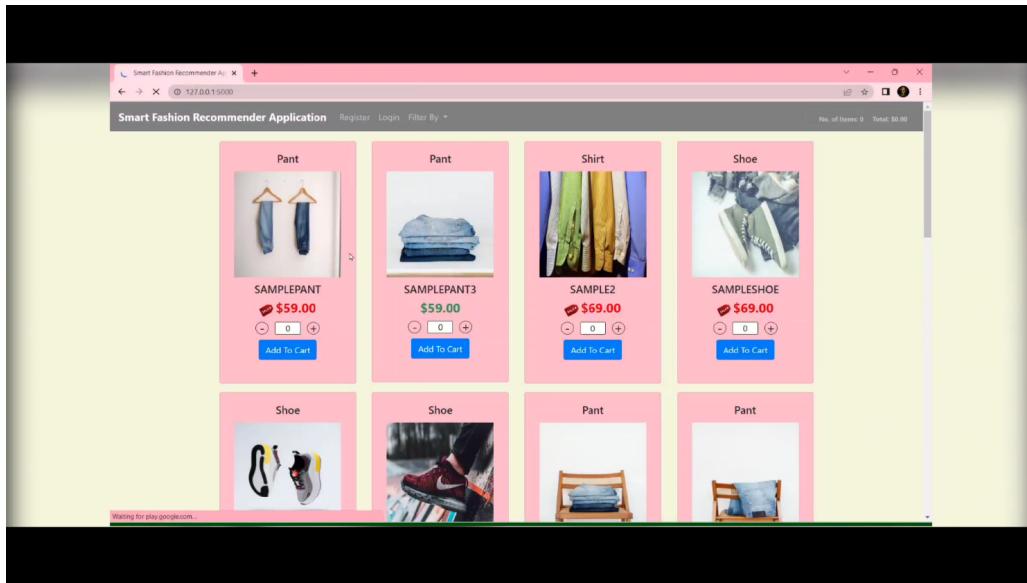
```

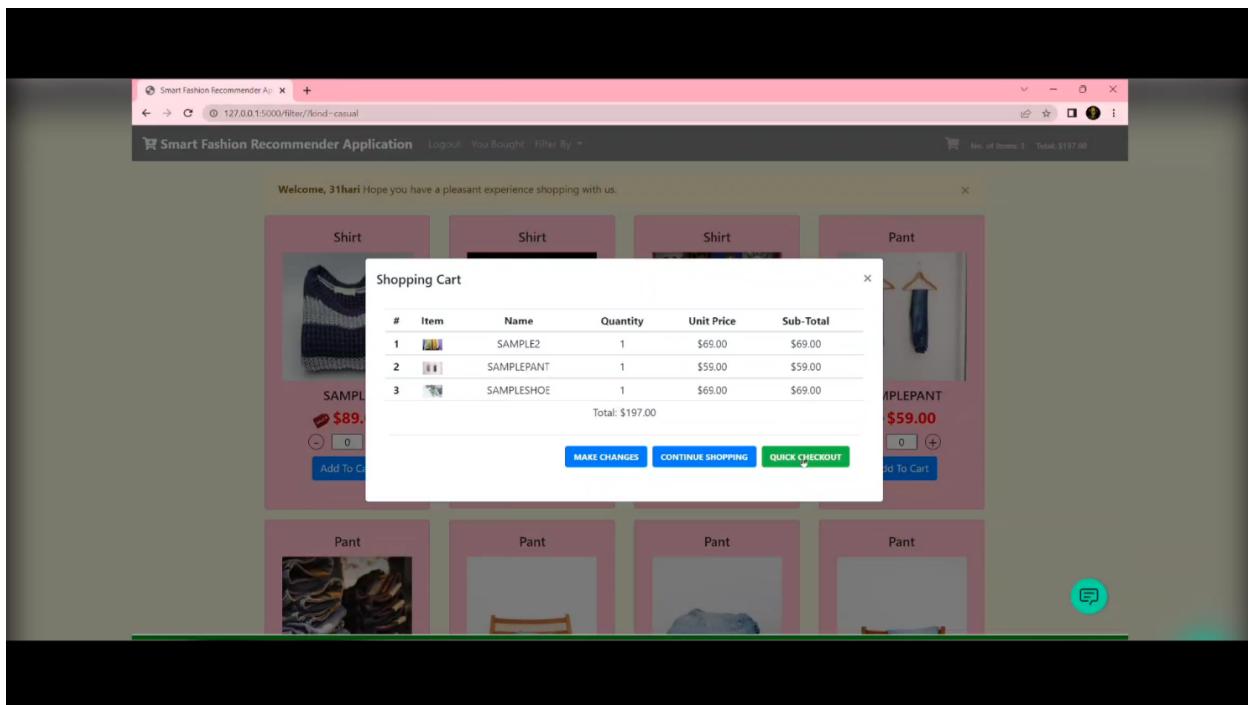
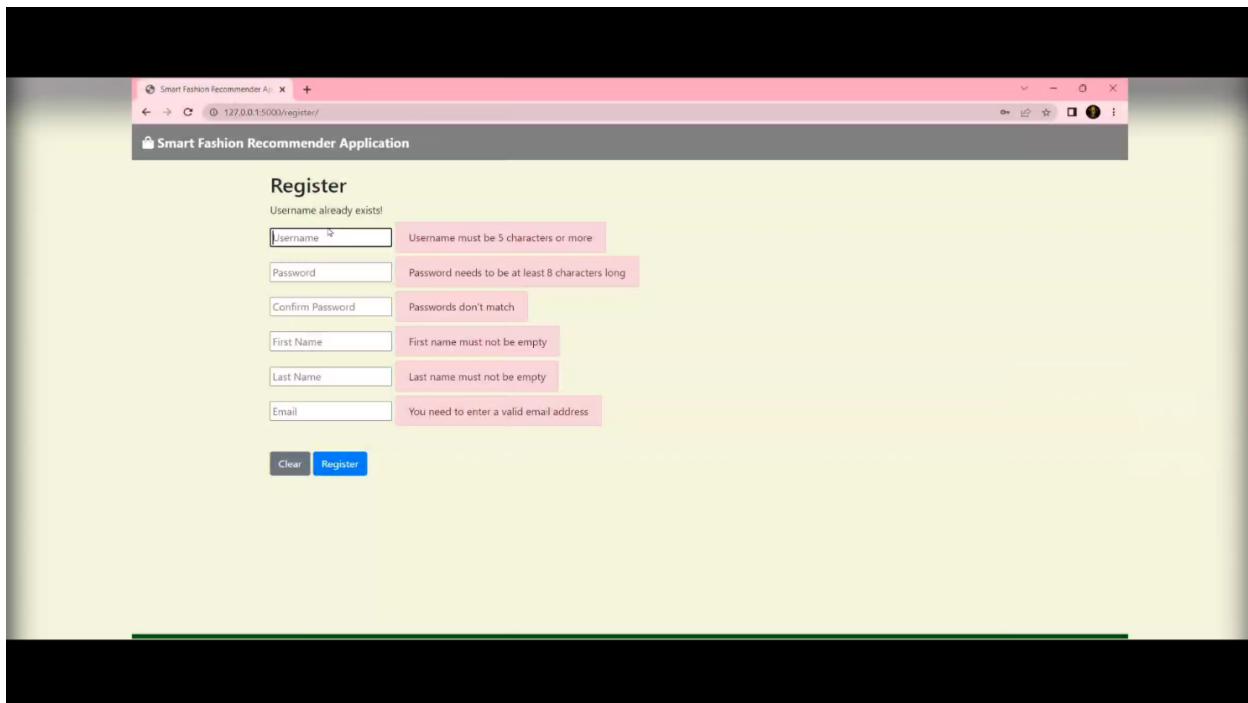
```

@app.route("/cart/")
def cart():
    if 'user' in session:
        # Clear shopping cart variables
        totItems, total, display = 0, 0, 0
        # Grab info currently in database
        shoppingCart = db.execute("SELECT samplename, image, SUM(qty), SUM(subTotal), price, id FROM
        cart GROUP BY samplename")
        # Get variable values
        shopLen = len(shoppingCart)
        for i in range(shopLen):
            total += shoppingCart[i]["SUM(subTotal)"]
            totItems += shoppingCart[i]["SUM(qty)"]
        # Render shopping cart
        return render_template("cart.html", shoppingCart=shoppingCart, shopLen=shopLen, total=total,
        totItems=totItems, display=display, session=session)

```

## OUTPUT SCREENSHOTS





Smart Fashion Recommender Application

Your Shopping History

Items you've bought in the past.

#	Item	Name	Quantity	Date	Action
1		SAMPLEPANT	1	2022-11-13	<button>Buy Again</button>
2		SAMPLE2	1	2022-11-13	<button>Buy Again</button>
3		SAMPLESHOE	1	2022-11-13	<button>Buy Again</button>

© Smart Fashion Recommender Application

127.0.0.1:5000/history/

Welcome, 31hari! Hope you have a pleasant experience shopping with us.

<p>Shirt</p> <p>SAMPLE \$89.00</p> <p>(-) 0 (+)</p> <p>Add To Cart</p>	<p>Shirt</p> <p>SAMPLE3 \$89.00</p> <p>(-) 0 (+)</p> <p>Add To Cart</p>	<p>Shirt</p> <p>SAMPLE4 \$79.00</p> <p>(-) 0 (+)</p> <p>Add To Cart</p>	<p>Watson Assistant</p> <p>Hai Handsome. Here are few Recommendations for you.</p> <p>Please select your age Range:</p> <p>1 - 10 11 - 20 21 - 30 31 - Above</p> <p>&gt; 10</p> <p>Here is your Recommend products:</p> <p>Modern Western Traditional Formal Casual</p> <p>Do you want to continue?</p> <p>yes no</p> <p>Type something... Built with IBM Watson®</p>
<p>Pant</p> <p>Pant</p>	<p>Pant</p> <p>Pant</p>	<p>Pant</p> <p>Pant</p>	

127.0.0.1:5000/filter/?kind=casual

## DEPLOY IN KUBERNETES CLUSTER:

```
Hostname: echoserver-deployment-859b75d8c4-w75jn

Pod Information:
  node name:      10.94.21.13
  pod name:       echoserver-deployment-859b75d8c4-w75jn
  pod namespace:  default
  pod IP:        172.30.45.7

Server values:
  server_version=nginx: 1.13.3 - lua: 10008

Request Information:
  client_address=172.30.45.5
  method=GET
  real path/
  query=
  request_version=1.1
  request_scheme=http
  request_uri=http://echoserver.arpod-ipvs-test-aug14.us-south.containers.appdomain.cloud:8080/

Request Headers:
  accept=/*
  host=echoserver.arpod-ipvs-test-aug14.us-south.containers.appdomain.cloud
  user-agent=curl/7.54.0
  x-forwarded-for=195.21.1.1
  x-forwarded-host=echoserver.arpod-ipvs-test-aug14.us-south.containers.appdomain.cloud
  x-forwarded-port=443
  x-forwarded-proto=https
  x-global-k8f0ic-transaction-id=fc9b6d1f0ac1b7b63bf96cbf02396378
  x-real-ip=195.21.1.1

Request Body:
  -no body in request-
```

```
$ curl http://169.61.18.4:1884

Hostname: echoserver-deployment-859b75d8c4-r6s62

Pod Information:
  node name:      10.73.115.27
  pod name:       echoserver-deployment-859b75d8c4-r6s62
  pod namespace:  default
  pod IP:        172.30.154.209

Server values:
  server_version=nginx: 1.13.3 - lua: 10008

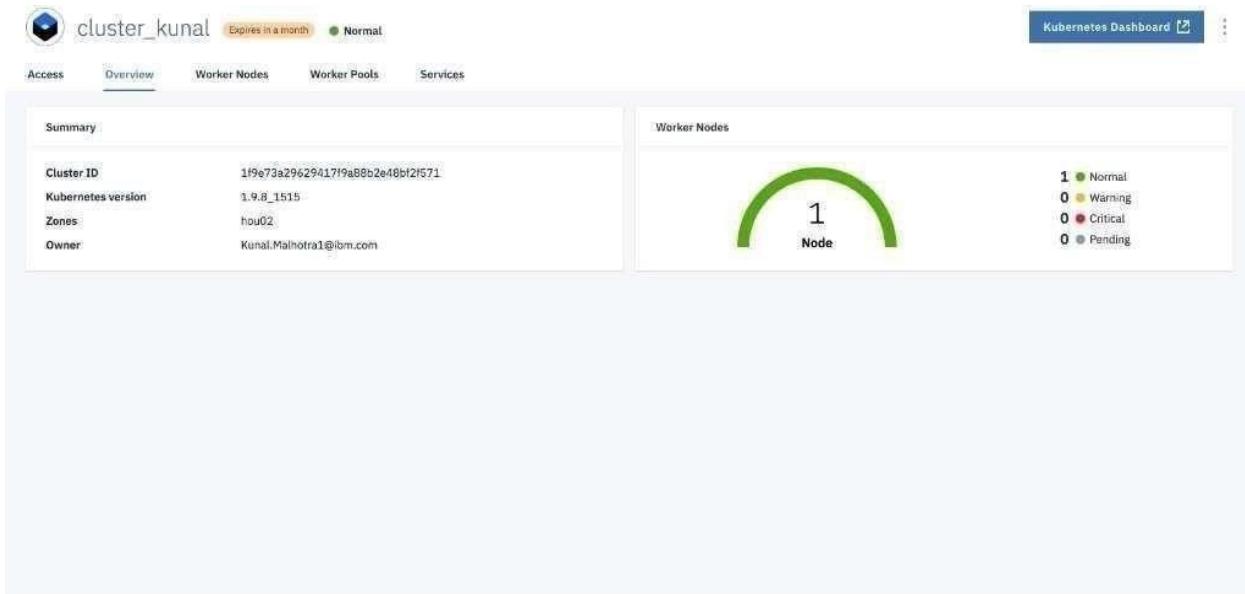
Request Information:
  client_address=195.21.1.1
  method=GET
  real path/
  query=
  request_version=1.1
  request_scheme=http
  request_uri=http://169.61.18.4:8080/

Request Headers:
  accept=/*
  host=169.61.18.4:1884
  user-agent=curl/7.54.0

Request Body:
  -no body in request-
```

# RESULTS

## 8.1 Performance Metrics



The screenshot shows the Kubernetes Dashboard's Worker Nodes page for the same cluster.

**Worker Nodes:**

Name	Status	Worker Pool	Zone	Private IP	Public IP	Kubernetes Version
w1	Normal	default	hou02	10.47.79.201	184.172.233.151	1.9.8_1517

Items per page: 10 | 1 of 1 pages

## **CHAPTER 9**

### **CONCLUSION**

The present paper presents the development of a system that recognizes fashion similar images. We accomplish this by implementing an already existing CNN model with transfer learning for cloth image recognition using different libraries. For this purpose, we created a plan for collecting data and for developing the steps needed for preprocessing and cleaning up the data. We took into account features like patterns, machine, fabric, style etc. After extensive preprocessing and cleaning of data in a dataset, we constructed the model of stacked CNN to predict the features specific to these attributes and to train the models with the dataset to generate accurate predictions regarding almost all forms of images. A stacked CNN was used and implemented, with the help of this algorithm through which the system can recommend similar images. This is the last test to assess if deep learning for style recovery is at a high development and can be utilized in making fashion choices.

## **CHAPTER 10**

### **FUTURE SCOPE**

Our project has a wide scope in future which indeed results in large revolution in E-commerce platform. In our project we have to improvise with machine learning which results in easy way of interaction and using IBM cloud, suggestions for purchasing and chat bots for customer convenience. This would help in targeting audience who are interested in which specific types of products and also help in tempting others those who don't believe in online shopping this would increase the rate of online E-commerce users and thus directly proportional to profit of E-Commerce as well as customer satisfaction.

## REFERENCES

1. Abadi, M., Agarwal, A., Barham, P., Brevdo, E., Chen, Z., Citro, C., Corrado, G. S., Davis, A., Dean, J. & Devin, M. (2016). *Tensorflow: Large-scale machine learning on heterogeneous distributed systems*, pp. 265–283. arXiv preprint arXiv:1603.04467.
2. Alkhawlani, M., Elmogy, M. & EL Bakry, H. (2015). *Text-based, content-based, and semantic-based image retrievals: a survey*. Int. J. Comput. Inf. Technol, 4, pp. 58-66.
3. Beel, J., Langer, S., Genzmehr, M., Gipp, B., Breitinger, C. & Nürnbergger (2013). A. *Research paper recommender system evaluation: a quantitative literature survey*. Proceedings of the International Workshop on Reproducibility and Replication in Recommender SystemsEvaluation, 2013, pp. 15-22.
4. Bobadilla, J., Hernando, A., Ortega, F. & Bernal, J. (2011). *A framework for collaborative filtering recommender systems*. ExpertSystems with Applications, 38, pp. 14609-14623.
5. Burke, R. (2002). *Interactive critiquing forcatalog navigation in e-commerce*. Artificial Intelligence Review, 18, pp. 245-267.
6. Claypool, M., Gokhale, A., Miranda, T., Murnikov, P., Netes, D. & Sartin,M. (1999). *Combing content-based and collaborative filters in an online newspaper*. Proc. of Workshop on Recommender Systems-Implementation and Evaluation, 1999, pp. 6-22.
7. Guo, G., Zhang, J. & Thalmann, D. (2014). *Merging trust in collaborative filtering to alleviatedata sparsity and cold start*. Knowledge-Based Systems, 57, pp. 57-68.
8. Iglesias, F. & Kastner, W. (2013). *Analysis of similarity measures in times series clustering for the discovery of building energy patterns*. Energies, 6(2), pp. 579-597.
9. Isinkaye, F. O., Folajimi, Y. & Ojokoh,B. A. (2015). *Recommendation systems: Principles,methodsand evaluation*. Egyptian informatics journal, 16, pp. 261-273.

10. Jannach, D. & Friedrich, G. (2013). *Tutorial: recommender systems*. International Joint Conference on Artificial Intelligence Beijing, 2013, pp. 1-144.
11. Krizhevsky, A., Sutskever, I. & Hinton, G. E. (2017). *ImageNet classification with deep convolutional neural networks*. Communications of the ACM, 60, pp. 84-90.
12. Lee, J. S. (2012). *Survey of Recommender Systems*. Citeseer.
13. Massa, P. & Bhattacharjee, B. (2004). *Using trust in recommender systems: an experimental analysis*. International conference on trust management, 2004. Springer, Berlin, Heidelberg, pp. 221-235.
14. Melville, P. & Sindhwani, V. (2010). *Recommender systems*. Encyclopedia of machine learning, vol. 1, pp. 829-838.
15. Mirescu, S. V. & Maiorescu, T. (2010). The premises and the evolution of electronic commerce. *Journal of knowledge management, economics and information technology*, vol. 1(1), pp. 44-56.
16. Park, D. H., Choi, I. Y., Kim, H. K. & Kim, J. K. (2011). *A review and classification of recommender systems research*. International Proceedings of Economics Development & Research, 5 (1), p. 290.
17. Pazzani, M. J. (1999). *A framework for collaborative, content-based and demographic filtering*. *Artificial intelligence review*, vol. 13, pp. 393-408.
18. Pine, I. (1993). BJ (1993). *Mass customization: The new frontier in business competition*, Vol. 17, issue 2, pp. 271-283.
19. Reshma, C. & Patil, A. (2012). *Content based image retrieval using color and shape features*. International Journal of Advanced Research in Electrical, Electronics and Instrumentation Engineering, 1, pp. 386-392.
20. Schafer, J. B., Konstan, J. & Riedl, J. (1999). *Recommender systems in e-commerce*. Proceedings of the 1st ACM conference on Electronic commerce, 1999, pp. 158-166.