

ABSTRACT

The Internet-based recruiting platforms become a primary recruitment channel in most companies. While such platforms decrease the recruitment time and advertisement cost, they suffer from an inappropriateness of traditional information retrieval techniques like the Boolean search methods. Consequently, a vast amount of candidates missed the opportunity of recruiting. The recommender system technology aims to help users in finding items that match their personnel interests; it has a successful usage in e-commerce applications to deal with problems related to information overload efficiently. In order to improve the e-recruiting functionality, many recommender system approaches have been proposed. This article will present a survey of e-recruiting process and existing recommendation approaches for building personalized recommender systems for candidates/job matching. In the last years, job recommender systems have become popular since they successfully reduce information overload by generating personalized job suggestions. Although in the literature exists a variety of techniques and strategies used as part of job recommender systems, most of them fail to recommending job vacancies that fit properly to the job seekers profiles. Thus, the contributions of this work are twofold, i) We made publicly available a new dataset formed by a set of job seekers profiles and a set of job vacancies collected from different job search engine sites. ii) We put forward the proposal of a framework for job recommendation based on professional skills of job seekers. iii) We provide a chatbot for user convenience, regarding their doubts to clarify. 4 We thus present a general panorama of job recommendation task aiming to facilitate research and real-world application design regarding this important issue.

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1.INTRODUCTION

1.1 PROJECT OVERVIEW

There has been a sudden boom in the technical industry and an increase in the number of good startups. Keeping track of various appropriate job openings in top industry names has become increasingly troublesome. This leads to deadlines and hence important opportunities being missed. Through this research paper, the aim is to automate this process to eliminate this problem. To achieve this, IBM cloud services like db2, Watson assistant, cluster, kubernetes have been used. A hybrid system of Content-Based Filtering and Collaborative Filtering is implemented to recommend these jobs. The intention is to aggregate and recommend appropriate jobs to job seekers, especially in the engineering domain. The entire process of accessing numerous company websites hoping to find a relevant job opening listed on their career portals is simplified. The proposed recommendation system is tested on an array of test cases with a fully functioning user interface in the form of a web application. It has shown satisfactory results, outperforming the existing systems. It thus testifies to the agenda of quality over quantity.

1.2 PURPOSE

- If To develop an end-to-end web application capable of displaying the current job openings based on the skillset of the users.
- **1** The users and their information are stored in the Database.
- An alert is sent when there is an opening based
- ∅ on the user skillset.

2. LITERATURE SURVEY

2.1 EXISTING PROBLEM

Building and managing recommender systems today requires specialized expertise in analytics, applied machine learning, software engineering, this makes it challenging regardless of your background or skillset.

Intelligent_ Chatbot Description A Chatbot is a software application that replaces a live human agent to conduct a conversation via text or text to speech. In this system, we demonstrate a chatbot that uses Artificial Intelligence to produce dynamic responses to online client enquiries. This web-based platform provides a vast intelligent base that can help humans to solve problems. The Chatbot recognizes the user's context, which prompts an intended response. Its objective is to reduce human dependency in every organization and reduce the need for different systems for different processes.

2.2 REFERENCES

- 1. Sha ha T Al-Ota and Mourad Ykhlef. "A survey of job recommender systems". In: International Journal of the Physical Sciences 7.29 (2012), pp. 5127—5142.
- 2. N Deniz, A Noyan, and O G Ertosun. "Linking Person-job Fit to Job Stress: The Mediating Effect of Perceived Person-organization Fit". In: Procedia Social and Behavioural Sciences 207 (2015), pp. 369—376.
- 3. M Diaby, E Viennet, and T Launay. "Toward the next generation of recruitment tools: An online social network-based job recommender system". In: Proc. of the 2013 IEEE/ACM Int. Conf. on Advances in Social Networks Analysis and Mining, ASONAM 2013 (2013), pp. 821–828. 8
- 4. M Diaby and E Viennet. "Taxonomy-based job recommender systems on Facebook and LinkedIn profiles". In: Proc. of Int. Conf. on Research Challenges in Information Science (2014), pp. 1–6. issn: 21511357. doi: 10.1109/RCIS.2014.6861048.
- 5. M Kusner et al. "From word embeddings to document distances". In: Proc. of the 32nd Int. Conf. on Machine Learning, ICML'15. 2015, pp. 957—966.
- 6. T Mikolov et al. "Distributed Representations of Words and Phrases and Their Compositionality". In: Proc. of the 26th Int. Conf. on Neural Information Processing Systems Volume 2. NIPS' 13. Lake Tahoe, Nevada, 2013, pp. 3111—3119.

2.3 PROBLEM STATEMENT DEFINITION

When we start to do a project, the first thing that we should always do is define the problem. It's not only about dividing the big project into small parts, but also representing how to think about the problem, which may have varying performance in our final solution.

To build a job recommendation system providing recommendations to millions of users with millions of items, the first thing is to define the problem. Dealing with the enormous amount of recruiting information on the Internet, a job seeker always spends hours to find useful ones. Many times, people who lack industry knowledge are unclear about what exactly they need to learn in order to get a suitable job for them. We address the problem of recommending suitable jobs to people who are seeking a new job. Job recommender technology aims to help job seekers in finding jobs that match their skills. The Internet caused a substantial impact on the recruitment process through the creation of e-recruiting platforms that become a primary recruitment channel in most companies. While companies established job positions on these portals, job-seeker uses them to publish their profiles. E-recruitment platforms accomplished clear advantages for 9 both recruiters and job-seekers by reducing the recruitment time and advertisement cost. Recommender system technology aims to help users in finding items that match their preferences; it has a successful usage in a widerange of applications to deal with problems related to information overload efficiently. In order to improve the e-recruiting functionality, many system approaches have been proposed. This paper will e-recruiting process and related issues for building personalized recommender systems of candidates/job matching.

3.IDEATION&PROPOSED SOLUTION

3.1 Empathy Map Canvas:

An empathy map is a simple, easy-to-digest visual that captures knowledge about a user's behaviours and attitudes.

It is a useful tool to helps teams better understand their users.

Creating an effective solution requires understanding the true problem and the person who is experiencing it. The exercise of creating the map helps participants consider things from the user's perspective along with his or her goals and challenges.

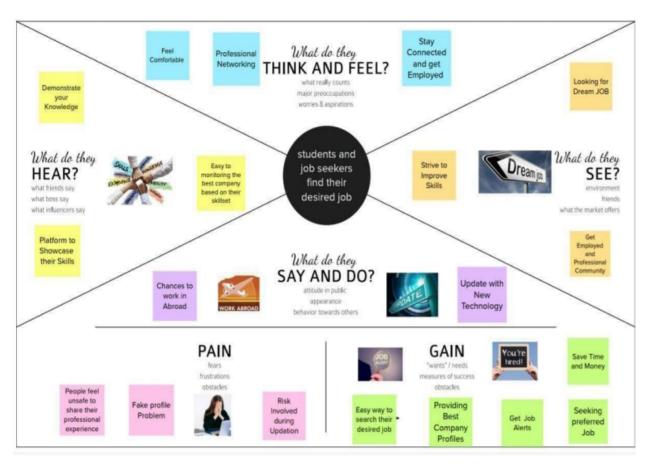


Fig 3.1 Empathy Map

3.2 IDEATION AND BRAINSTORMING

- ➤ E-recruitment platforms decrease the recruitment time and advertisement cost, they suffer from an inappropriateness of traditional information retrieval techniques like the Boolean search methods. Consequently, a vast amount of candidates missed the opportunity of recruiting. The recommender system technology aims to help users in finding items that match their personnel interests; it has a successful usage in e-commerce applications to deal with problems related to information overload efficiently
- > Job recommendation application with intelligence of chatbot. In this system, we demonstrate a chatbot that uses Artificial Intelligence to produce dynamic responses to online client enquiries. This web-based platform provides a vast intelligent base that can help humans to solve problems. The chatbot recognizes the user's context, which prompts an intended response. Because this is a dynamic response, the user's desired response will be generated. This also uses a machine-learning algorithm to learn the chatbot by experiencing various requests and responses. Chatbots come to use in numerous fields of our daily life. Because Al enhances the human touch in every communication, chatbots are becoming increasingly robust. It triggers accurate responses after understanding a user's query. Its objective is to reduce human dependency in every organization and reduce the need for different systems for different processes.
- > Job seekers struggling to get the desired job for skills they have. we are proposing an application which will help the students to give Suggestions on the jobs based the skills. In this application freshers or skilled person can sign up and find the jobs by using either the search option or they can directly interact with the chatbot and get their dream job. In this application freshers or skilled person can sign up and find the jobs by using either the search option or they can directly interact with the chatbot and get their dream job. An alert is sent when there is an opening based on the user skillset. Users will interact with the chatbot and can get the recommendations based on their skills. We can use a job search API to get the current job openings in the market which will fetch the data directly from websites.

Service of the control of the contro

Fig 3.2 Ideation and Brainstorming

3.3PROPOSED SOLUTION

Job recommender systems are desired to attain a high level of accuracy while making the predictions which are relevant to the customer, as it becomes a very tedious task to explore thousands of jobs, posted on the web, periodically. The web recommender System suffers from many challenges.

S.NO	PARAMETERS	DESCRIPTION
1	Problem Statement (Problem	To develop an end-to-end
	to be solved)	application capable of
		displaying the current job
		openings based on the user
		skillset
2	Idea / Solution description	In this paper, we proposed a
		framework for job
		recommendation task. This
		framework facilitates the
		understanding of job
		recommendation process as
		well as it allows the use of a
		variety of text processing and
		recommendation methods
		according to the preferences
		of the job recommender

		evetem decigner Marcover
		system designer. Moreover,
		we also contribute making
		publicly available a new
		dataset containing job
		seekers profiles and job
		vacancies. Future directions
		of our work will focus on
		performing a more exhaustive
		evaluation considering a
		greater amount of methods
		and data as well as a
		comprehensive evaluation of
		the impact of each
		professional skill of a job
		seeker on the received job
		recommendation.
3	Novelty/ Uniqueness	With the development of
		information technology and
		application of the Internet,
		People gradually entered the
		time of information overload
		from information scarcity.
		User satisfaction with
		recommender systems is
		•
		related not only to how
		related not only to how accurately the system
		accurately the system
		accurately the system recommends but also to how
		accurately the system recommends but also to how much it supports the user's
		accurately the system recommends but also to how much it supports the user's decision making. Novelty is
		accurately the system recommends but also to how much it supports the user's decision making. Novelty is one of the important metrics
		accurately the system recommends but also to how much it supports the user's decision making. Novelty is one of the important metrics of customer satisfaction.
		accurately the system recommends but also to how much it supports the user's decision making. Novelty is one of the important metrics

		Recommender Systems (RS)
		field that novelty is
		fundamental qualities of
		recommendation
		effectiveness and added-
		value. This paper combed
		research results about
		definition and algorithm of
		novel recommendation, and
		starting from the meaning of
		"novel", defined novelty of
		item in recommendation
		system. Experiment proved
		using the definition of novelty
		to recommend can effectively
		recognize the item that the
		user is familiar with and
		ensure certain accuracy
4	Social Impact/ Customer	we develop several
	Satisfaction	recommender systems and
		measure their ability to deliver
		accurate and diverse
		recommendations and their
		ability to generate customer
		satisfaction with diverse data
		sets. The results show that
		sets. The results show that accuracy and diversity
		accuracy and diversity
		accuracy and diversity positively affect customer
		accuracy and diversity positively affect customer satisfaction when applying a
		accuracy and diversity positively affect customer satisfaction when applying a deep learning-based recommender system. By
		accuracy and diversity positively affect customer satisfaction when applying a deep learning-based

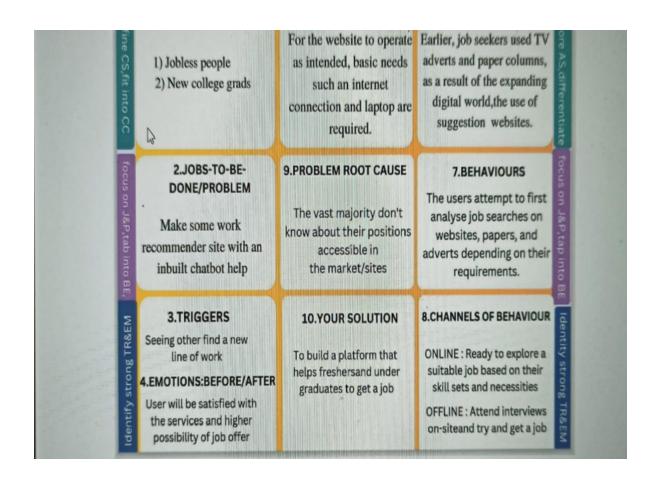
		satisfaction when applying
		traditional recommender
		systems. These results imply
		that developers or managers
		of recommender systems
		need to identify factors that
		further improve customer
		satisfaction with the
		recommender system and
		promote the sustainable
		development of e-commerce.
5	Business Model (Revenue	Recommendation systems
	Model)	allow brands to personalize
		the consumer experience and
		make suggestions for the
		information that make the
		most sense to them. A
		recommendation engine also
		lets businesses analyse the
		customer's current usage and
		past browsing history to
		deliver relevant service and
		product recommendations
6	Scalability of the Solution	Recommendation system is a
		which provides techniques
		with information, which
		he/she may be interested in
		or accessed in past.
		Traditional recommender
		techniques such as content
		and collaborative filtering
		used in various applications
1		

3.4 PROBLEM SOLUTION FIT

The Problem-Solution Fit simply means that you have found a problem with your customer and that the solution you have realized for it actually solves the customer's problem. It helps entrepreneurs, marketers and corporate innovators identify behavioural patterns and recognize what would work and why Purpose

- \square Solve complex problems in a way that fits the state of your customers.
- ☐ Succeed faster and increase your solution adoption by tapping into existing mediums and channels of behaviour.
- ☐ Sharpen your communication and marketing strategy with the right triggers and messaging.
- ☐ Increase touch-points with your company by finding the right problembehaviour fit and building trust by solving frequent annoyances, or urgent or costly problems.
- ☐ Understand the existing situation in order to improve it for your target group.

Table 3.4 Problem Solution Fit:



4. REQUIREMENT ANALYSIS

4.1 FUNCTIONAL REQUIREMENT

Following are the functional requirements of the proposed solution.

FR No.	Non-Functional	Description
	Requirement	
NFR-1	Usability	This application can be used by the job seekers to
		login and search for the job based on her Skills set.
NFR-2	Security	This application is secure with separate login for Job
		Seekers as well as Job Recruiters.
NFR-3	Reliability	This application is open-source and feel free to use,
		without need to pay anything. The enormous job
		openings will be provided to all the job seekers
		without any limitation.
NFR-4	Performance	The performance of this application is quicker
		response and takes lesser time to do any process.
NFR-5	Availability	This application provides job offers and
		recommends Skills for a Particular Job openings.
NFR-6	Scalability	The Response time of the application is quite faster
		compared to any other application.

5. PROJECT DESIGN

5.1 DATA FLOW DIAGRAMS

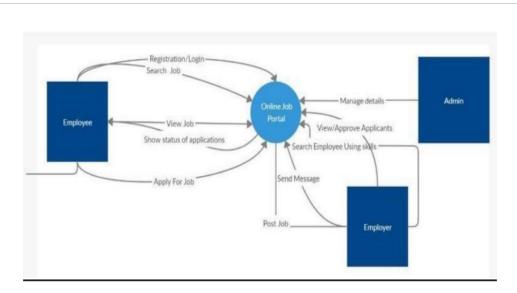


Fig 5.1 Data Flow Diagram

A Data Flow Diagram (DFD) is a traditional visual representation of the information flows within a system. A neat and clear DFD can depict the right amount of the system requirement graphically. It shows how data enters and leaves the system, what changes the information, and where data is stored.

5.2 SOLUTION AND TECHNICAL ARCHITECTURE

JOB RECOMMENDED APPLICATION: (SOLUTION ARCHITECTURE)

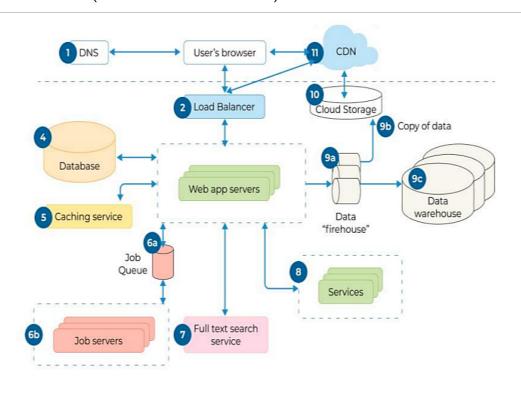


Fig 5.2: SOLUTION AND TECHNICAL ARCHITECTURE

5.3 USER STORIES

Use the below template to list all the user stories for the product.

Table 5.3 User Stories

User Type	Functional	User Story	User Story /	Acceptance	Priority	Release
	Requireme	Number	Task	criteria		
	nt (Epic)					
Customer	Registration	USN-1	As a user, I	I can	High	Sprint-1
(Mobile			can register	access my		
user)			for the job	account /		
			application	dashboard		
			by entering			
			my email,			
			password,			
			and			
			confirming			
			my			
			password.			
		USN-2	As a user, I	I can	High	Sprint-1
			will receive	receive		
			confirmati	confirmati		
			on email	on email &		
			once I have	click		
			registered	confirm		
			for the			
			application			
		USN-3	As a user, I	I can	Medium	Sprint-2
			can register	receive		
			for the job	confirmati		
			application	on Email		
			through	and apply		
			Gmail	for the job		
		USN-4	As a user, I	I can	Low	Sprint-1

			can register	register &		
			for the job	access the		
			application	dashboard		
			through	with		
			Facebook	Facebook		
				Login		
	Login	Login	As a user, I	I can apply	High	Sprint-1
			can log into	for a job		
			the			
			application			
			by entering			
			email &			
			password			
	Dashboard	USN-6	As a user, I	Update your	High	Sprint-1
			can Search	resume for		
			& Apply	latest jobs		
			jobs posted			
			by top			
			companies			
			&			
			consultants			
			as per your			
			skills			
Customer		USN-7	As a user, I	I can	High	Sprint-1
(Web user)			can limit	receive a		
			who can	information		
			see her	from		
			resume	company		
				can post		
				new job		
				openings		

Customer	USN-8	As a user, I	I can select	Medium	Sprint-2
Care		want to	a job based		
Executive		select a	on my		
		desired	skills		
		jobs			
	USN-9	As a user, I	l can	High	Sprint-1
		can Update	receive		
		my resume	confirmati		
		for latest	on form job		
		jobs	portal		
	USN-10	As a use, I	l can	High	Sprint-2
		want to	access and		
		read a	see the		
		privacy and	privacy		
		rules	statement		
			and read it		
			in the job		
			portal		
	USN-11	As a user, I	I can start	Medium	Sprint-1
		want to	searching		
		quickly and	in the job		
		easily apply	portable so		
		for a job	Its quickly		
			as possible		

6. PROJECT PLANNING & SCHEDULING

6.1 Sprint Planning & Estimation

6 .2 Sprint Delivery Schedule

Table 6.1: Sprint Planning &Estimation

Sprint	Functional	User Story	User Story	Story	Priority	Team
	Requireme	Number	/ Task	Points		Members
	nt (Epic)					
Sprint1	Registrati	USN-1	As a user, I	7	High	1
	on		can			
			register for			
			the			
			application			
			by entering			
			my email,			
			password,			
			and			
			confirming			
			my			
			password.			
Sprint1		USN-2	As a user, I	7	High	1
			will receive			
			confirmati			
			on email			
			once I have			
			registered			
			for the			
			application			
Sprint2		USN-4	As a user, I	5	Low	1
			can			
			register for			
			the			

			application			
			through			
			Facebook			
Sprint2		USN-5	As a user, I	5	Medium	1
			can			
			register for			
			the			
			application			
			through			
			Gmail			
Sprint2	Login	USN-6	As a user, I	10	High	1
			can log into			
			the			
			application			
			by entering			
			email &			
			password			
Sprint3	Profile and	USN-7	Update	7		1
	details		user skills			
			in their			
			account to			
			use it for			
			job search.			

Sprint3		USN-8	Make user	7	Low	1
			able to edit			
			their skill			
			set			
Sprint1	Communic	USN-3	A customer	6		1
	ation		care			
			executive			
			is a			
			profession			
			al			
			responsible			
			for			
			communic			
			ating the			
			how's and			
			why's			
			regarding			
			service			
			expectatio			
			ns within a			
			company			
Sprint3		USN15	Create a	6	Low	1
			chat			
			assistant			
			for the			
			users.			
Sprint4	Backend	USN10	Backend to	20	High	1
	processes		search job			
			based on			
			user skill			
			set.			
Sprint5	Deployme	USN13	Containeri	10	High	1
	nt		ze the			
			application.			
Sprint5		USN14	Deploy the	10	High	1
			application			
			for public			
			access.			

Project Tracker, Velocity & Burndown Chart:

Table 6.2: Project Tracker

Sprint	Total Story	Duration	Sprint Start Date	Sprint End Date	Story	Sprint Release Date
	Points			(Planned)	Points	(Actual)
					Completed	
					(as on	
					Planned	
					End Date)	
Sprint1	20	6 Days	24 Oct 2022	29 Oct 2022	20	02 Nov 2022
Sprint2	20	6 Days	31 Oct 2022	05 Nov 2022	20	06 Nov 2022
Sprint3	20	6 Days	07 Nov 2022	12 Nov 2022	20	
Sprint4	20	6 Days	14 Nov 2022	19 Nov 2022	20	

Velocity:

Imagine we have a 10-day sprint duration, and the velocity of the team is 20 (points per sprint). Let's calculate the team's average velocity (AV) per iteration unit (story points per day)

6.3 Reports From JIRA

Burndown Chart: A burn down chart is a graphical representation of work left to do versus time. It is often used in agile software development methodologies such as Scrum. However, burn down charts can be applied to any project containing measurable progress over time. Burndown Chart 30 25 20 15 10 5 Sprint 1 Sprint 2 Sprint 3 Sprint 4

Fig;6.3 Burndown chart

7. CODING & SOLUTIONING:

FEATURE 1: SIGN UP

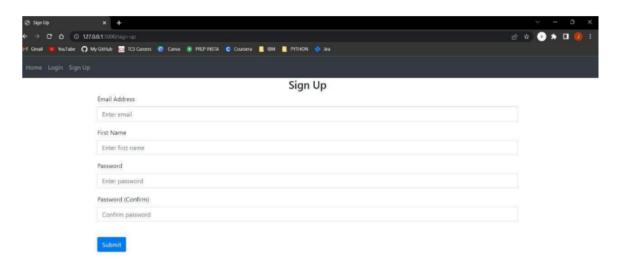
FEATURE 2: LOGIN

FEATURE 3: ADD SKILLS

FEATURE 4: APP WILL RECOMMEND THE JOBS ACCORDING TO THE SKILLS

FEATURE 5: USER CAN APPLY FOR JOBS.

SIGNUP PAGE



CODE

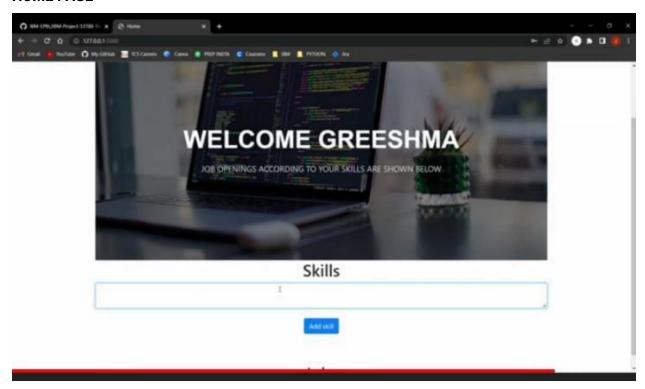
```
## wethod > Toroglates > O sign.up.html > ...

## vettods > No sign.up.html > ...

## dots | Sign.up.html | ...

## dot
```

HOME PAGE



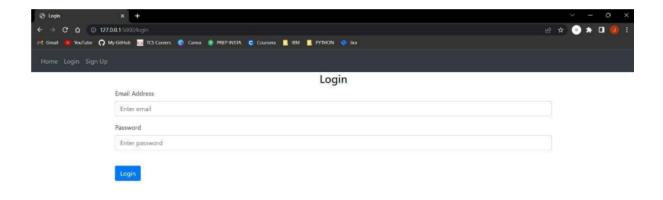
CODE

```
O home.html 3 × ...

website 3 templates > O home.html >...

1 (X extends "base.html" X) (X block title X)Home(X endblock X) (X block contered X) (X extends "base.html" X) (X block title X)Home(X endblock X) (X block contered X) (X extends "base.html" X) (X block title X)Home(X endblock X) (X block contered X) (X extends "base.html" X) (X block title X)Home(X endblock X) (X block contered X) (X extends "base.html" X) (X block title X)Home(X endblock X) (X block contered X) (X extends "base.html" X) (X block title X)Home(X endblock X) (X block contered X) (X extends "base.html" X) (X block title X)Home(X endblock X) (X block contered X) (X extends "base.html" X) (X block title X)Home(X endblock X) (X block contered X) (X extends "base.html" X) (X block title X)Home(X endblock X) (X block contered X) (X extends "base.html" X) (X block title X)Home(X endblock X) (X block contered X) (X extends "base.html" X) (X block title X)Home(X endblock X) (X block title X) (X block title X)Home(X endblock X) (X block title X)Home(X endb
```

LOGIN



CODE

```
| Description |
```

DATABASE SCHEMA:

```
website > \Phi models.py \ x
website > \Phi models.py > ...

from . import db
from flask_login import UserMixin
from sqlalchemy.sql import func

4

6
    class Note(db.Model):
    id = db.Column(db.Integer, primary_key=True)
    data = db.Column(db.String(19090))
    date = db.Column(db.DateTime(timezone=True), default=func.now())
    user_id = db.Column(db.Integer, db.ForeignKey('user.id'))

11

12
    class User(db.Model, UserMixin):
    id = db.Column(db.Integer, primary_key=True)
    email = db.Column(db.String(150), unique=True)
    password = db.Column(db.String(150))
    ifirst_name = db.Column(db.String(150))
    notes = db.relationship('Note')
```

8.TESTING

8.1 TEST CASES:

Table 8.1: Test Cases

1					Date	03-Nov-22	1					
2					Team ID	PNT2022TMID25046	1					
3					Project Name	Project - Skill/Job Recommender						
5	Test case ID	Feature Type	Component	Test Scenario	Pre-Requisite	4 marks Steps To Execute	Expected Result	Actual Result	Status	Commnets	TC for Automation(Y/N	Executed By
6	LoginPage_TC_00	Functional	Login page	Verify that after registration users are navigated to login page	Mail id, Username, Password, Phone number, Pin	Open the website and go to register page. Enter details and press register. Werify that users are navigated to registration page.	Users should be navigated to registration page	Working as expected	Pass	Excellent	'n	JEFRINJ
7	LoginPage_TC_00	UI	Home Page	Verify the UI elements in Login/Signup popup	Username & Password	Open the website Enter details and press login Verify that users are notified of login process	Users should be notified of login process	Working as expected	Pass	Good	N	ALBERT RAVIDOSS
	LoginPage_TC_OO 3	Functional	Home page	Verify user is able to log into application with Valid credentials		Open the website Enter details and press login Werify that users are logged into website properly	User should be logged into website properly	Working as expected	Pass	Good	N	ARUN K
9	HomePage_TC_00	Functional	Home Page	Verify that categories of skills and jobs are shown in homepage		Open the website Enter details and press login Verify that categories of are showing Jobs shown in	Categories of skills and jobs should be shown in homepage	Working as expected	Pass	Good	N	BALAMURUGAN K
10	HomePage_TC_00 2	Functional	Home page	Verify that jobs are displayed in homepage		Open the website Enter details and press login Werify that jobs are displayed in homepage	jobs should be displayed in homepage	Working as expected	Pass	Good	N	JEFRIN J
11	HomePage_TC_00 3	Functional	Home page	Verify that when clicked on jobs it is redirected to correct page		Open the website Enter details and press login Werify that when clicked on lobs it is redirected to correct	When clicked on job link it should be redirected to correct page	Working as expected	Pass	Excellent	N	JAYAKRISHNAN J

8.2 USER ACCEPTANCE TESTING:

Acceptance Testing UAT Execution & Report Submiss

- 1. Purpose of Document The purpose of this document is to briefly explain the test coverage and open issues of the Skills and Job Recommendation project at the time of the release to User Acceptance Testing (UAT).
- 2. Defect Analysis This report shows the number of resolved or closed bugs at each severity level, and how they were

Resolution	Severity 1	Severity 2	Severity 3	Severity 4	Subtotal
By Design	10	4	3	3	20
Duplicate	1	1	2	2	6
External	2	3	0	1	6
Fixed	11	2	4	20	17
Not	0	0	1	0	1
Reproduced					
Skipped	0	0	1	1	2
Won't Fix	0	5	2	1	8
Total	24	14	13	26	80

Table 8.2: User Acceptance Testing

8.3. Test Case Analysis

This report shows the number of test cases

Table 8.3: Test Case Analysis

Section	Total case	Not Tested	Fail	Pass
Print Engine	7	0	1	7
Client	51	1	0	51
Application				
Security	2	0	2	2
Outsource	3	0	1	3
shipping				
Exception	9	0	1	9
Reporting				
Final Report	4	0	1	4
Output				
Version Control	2	0	0	2

9.RESULTS

9.1 PERFORMANCE METRICS

Efficiency should be a priority for employees. This requires them to have a good sense of time management and resource utilization. They should be able to monitor missed deadlines and how well a certain task was executed. But what is efficiency? In simple terms, it is the output that you get after putting in a certain amount of input that contributes to the overall success of a business.

Here is how you can measure an employee's efficiency. For instance:

- Choose the number of tasks completed
- Measure the number of tasks completed during a period of one month.
- Measure the output against the average figure of the workplace. The average of the workplace is the benchmark to measure.
- Evaluate an employee's input which is the number of hours an employee puts in.
- Divide the output by the input to get the efficiency figure Remember, efficiency is a key indicator that reveals whether an employee is meeting expectations or not.

When measuring efficiency, remember to evaluate the following as well:

- The job description
- The nature of work
- Amount of work assigned
- Deadline for completing tasks
- · Quality of work done

10.ADVANTAGES AND DISADVANTAGES

ADVANTAGES:

It can speak volumes for a candidate-in-question when they are referred by an existing employee. Not only will the current employee, the referrer, likely want to add to—and not detract from—company culture, but they'll also vouch for required skillsets and competencies. Here are the top advantages of employee referrals:

1. Your company will save time and money.

Sourcing candidates requires a lot of effort, which means it can cost a company both time and money. It was found in one study that referred candidates are 55% faster to hire, compared with employees sourced through career sites. An advantage of employee referrals is that your current team member makes the connection and saves the recruiter that initial time of sourcing the candidate. Further, the candidate could be a better match compared to other candidates who apply externally. This will also help expedite the process and cut back on the need to find alternative options.

2. Your company will receive qualified, quality candidates.

Employees will want to work with someone who will improve their own output and day-to-day workload. So, in most cases, you can have more confidence in the candidate's ability to perform the necessary tasks. Further, according to research done by Zao, nearly three in ten employers have caught a fake reference on an application. So, a personal recommendation that is already within the company can instill confidence that the reference is in fact valid and reputable.

3. Retention rate is typically better.

After two years, retention of referred employees is 45% compared to 20% from job boards. Employee referrals tend to stay around longer, perhaps because 38 they are personally connected to their peers. That's not to mention that the referrer themselves may feel more respected and valued too after their company takes their recommendation. And when an employee feels respected and valued, they can become more dedicated in turn. You may also want to give an employee referrer a bonus to show your appreciation.

DISADVANTAGES:

To properly answer "What are the advantages and disadvantages of employee referral?" we must now also look at the disadvantages. The disadvantages of employee referrals do not outweigh the benefits, but there are still some to consider. Here are three employee referrals disadvantages to keep in mind when making a hiring decision:

1. You may get a recommendation based on bias.

While in most cases an employee's motives should be "pure," there may be circumstances where a person wants to just work with their friend or receive the referral bonus.

This can result in the candidate not being as qualified as either the referrer or referee said they were. The referrer may think that they can make up for the candidate's shortcomings or give them a crash course to level-set their skills. This can impact their own production in a negative way. And now your company may have two underperforming employees—and you may have to look to fill both of these positions in the not-so-far-off future.

2. Employee referrals can invite opportunity for negative company politics.

While an advantage of employee referrals is that they can positively impact peer morale, they can also cause unnecessary tension. The twosome can be negatively received by their peers especially if the external hire was chosen over an internal promotion. Further, the referrer may be afraid to offer critique to the person they referred. This kind of dynamic can negatively impact their work.

3. Your company could end up losing both the referrer and the referee.

When one goes, the other may follow. Whether one decides to leave because of company politics, personal reasons, or a better opportunity, there is a risk that their counterpart will follow suit. This chance may heighten if problems with team dynamics aren't addressed and resolved. So, it's important to stay involved with a new hire, beyond any initial onboarding and ensure they are connected to the company and not just the employee who referred them.

11. CONCLUSION

Conclusions and directions for further research in this paper, we have considered the job recommender system (JRS) literature from several perspectives. These include the influence of data science competitions, the effect of data availability on the choice of method and validation, and ethical considerations in job recommender systems. Furthermore, we branched the large class of hybrid recommender systems to obtain a better view on how these hybrid recommender systems differ. Both this multi-perspective view, and the new taxonomy of hybrid job recommender systems has not been discussed by previous reviews on job recommender systems. Application-oriented challenges in JRS were already highlighted in early JRS contributions, though, still most literature does not take these into account. Contributions that do take different views on the JRS problem, however, do show that such views can have considerable benefits. These benefits may include improved model performance (temporal perspective), improved distribution of candidates over a set of homogeneous vacancies (reciprocal perspective), or ensuring algorithm fairness (ethical perspective). Currently, most attention goes out to how to represent the substantial amount of textual data from both candidate profiles and vacancies to create job recommendations, for which recently especially deep representations have shown promising results. However, this focus may also create the illusion that this is the only perspective that is relevant. Especially in terms of fairness, such a single perspective can be considerably harmful. Although we are not aware of algorithm audits on job recommender systems, an audit on the candidate search engines of Indeed, Career builder, and Monster, did show significant results for both individual and group unfairness in terms of gender. The increased scientific attention towards algorithm fairness, however, does provide algorithms and metrics that can be applied to measure and ensure algorithm fairness. Hence, there is a research opportunity to study how these can be transferred to the job 41 recommender system domain. Many authors state in the introduction of their contribution that there is a vast amount of data available in the form of vacancies and job seeker profiles. However, there is a clear split in the literature with regards to contributions having also access to interaction data between these two, in particular in the form of clicks/skips on the recommendation list. Interaction data can resolve the language inconsistency between job seekers and recruiters, which is especially troublesome in content-based and some knowledgebased JRS. In case interaction data is missing, one common resort is to use one of the available datasets originating from JRS competitions, in particular the CareerBuilder 2012, Rec Sys 2016, and Rec Sys 2017 competitions, which therefore have had a considerable influence on the JRS literature. An interesting aspect with respect to the usage of these competition datasets, beyond the contributions to the competitions themselves, is that these datasets are mostly used for training, but rarely for validation. This is unfortunate, as the (to our knowledge) only contribution that compares JRS on different competition datasets shows that error metrics may differ substantially across different datasets. I.e., this raises questions with respect to the generalizability of JRS trained on one dataset. Another interesting guestion why (online)interaction data is sometimes not taken into account, or along the same line, why researchers often resort to the competition datasets, beyond the motives of contributing the

competition or for validation. Although there may be many valid reasons, we would like to hypothesize from anecdotal experience that it can be difficult to obtain such interaction datasets, as recruitment organizations are not always part of research communities, or given that these recruitment organizations have not always considered the implications of sharing data for research, either from a technical or legal point of view, making it difficult to use such datasets on a short term.

12.FUTURE SCOPE

Future directions of our work will focus on performing a more exhaustive evaluation considering a greater amount of methods and data as well as a comprehensive evaluation of the impact of each professional skill of a job seeker on the received job recommendation. We can use machine learning techniques to recommend data in a efficient way.

13.APPENDIX

SOURCE CODE

Front end

```
import { useToast } from "@chakra-ui/react"; import React,
{ useContext, useEffect, useState } from "react";
import { useNavigate } from "react-router-dom"; import
{ AppContext } from "../context/AppContext"; import {
loginUser } from "../proxies/backend_api"; import {
emailRegex } from "../utils/helper";
const Login = () => {
const toast = useToast();
   const { setUser } = useContext(AppContext);
   const navigate = useNavigate();
   const [inputs, setInputs] = useState({
email: "", password: "",
});
const [error, setErrors] = useState({
email: "", password: "",
});
const handleChange = ({ target: { name, value } }) => {
setErrors((prev) => { return { ...prev, [name]: ""
};
}); setInputs((prev) => ({ ...prev, [name]:
value }));
};
 const checkInputErrors = () => {
let status = true;
               (inputs.email.trim() ===
          if
                                                      Ш
!emailRegex.test(inputs.email.trim())) {
setErrors((prev) => {
        return { ...prev, email: "Enter a valid email" };
      });
status = false;
        }
        if (inputs.password.trim() === "") {
setErrors((prev) => {
       return { ...prev, password: "Enter a valid password"
};
       });
status = false;
    }
```

```
if (inputs.password.trim().length < 6) {
       setErrors((prev) => {
           return { ...prev, password: "Minimum 6 characters" }; }); status = false; } return status;
};
       });
status = false;
   } return
status;
};
const handleLogin = async () => {
if (checkInputErrors()) {
       const data = await loginUser(inputs);
if (data.error) {
                       toast({
title: data.error
                      , status:
"error"
               , duration: 3000,
                       variant: "left
isClosable: true,
accent",
               position: "top",
       });
return;
               }
setUser(data);
toast({
       title: `Welcome back ${data.name}`,
        status: "success",
duration: 3000, isClosable:
true, variant: "left-accent",
position: "top",
        });
       localStorage.setItem("user", JSON.stringify(data));
       navigate("/dashboard");
    }
};
return (
<>
       <div>
         <button className="big-base-300 rounded-box flex flex-row flex</pre>
flex -row justify-evenly items -center gap-10 px -10 py-5 wfit mx-auto">
```

```
<span>Sign in with Github</span>
   <img src={'github-dark.png} alt="github"</pre>
width="14%"/>
    </button>
    <div classname ="divider max -w-xs">or</div>
   <form
      onSumbit={(e)
                                        => e.preventfault()}
className="card bg-base-300 rounded-box flex flexcol
justify-center items-center gap-5 px-10 py-5 w-fit mxauto"
    >
      <div>
         <input
value={inputs.email}
type="text"
                    name="email"
     placeholder="email"
     className="input input -bordered input-primary
w-full"
      onChange={handleChange}
      />
     {error.email !== " "&& (
        fontmedium">
        {error.email}
       )}
</div>
```

```
<div>
<input
 value ={inputs .password }
type="password
                       name ="password"
placeholder="password"
           className="input input -boardered input -
primary w-full"
          onChange={handlechange}
          />
          {error.password !== " "&& (
              fontmedium">
           {error. password}
           )}
          </div>
          <div className="text-center">
             <but
type ="submit"
onClick={handleLogin}
       className="btn btn-sm btn-primary mb-4"
       >
         Login
         </button>
```

```
</div>
      </form>
    </div>
  </>
);
};
export default Navbar;
styling
@import
url("https://fonts.googleapis.com/css2?family=Ubuntu&displ
ay=swap");
@tailwind base;
@tailwind components;
@tailwind utilities;
           :root
{
font-family:
Inter, Avenir, Helvetica, Arial, sansserif; font-size: 16px;
line-height: 24px; font-weight:
400;
color-scheme: light;
/* color: rgba(255, 255, 255, 0.87); backgroundcolor: #242424; */
font-synthesis: none; text-
font-synthesis: none;
```

```
text-rendering: optimizeLegibility;
-webkit-font-smoothing: antialiased
; -moz-osx-font-smoothing: grayscale;
-webkit-text-size-adjust: 100%;
 *{
 margin: 0;
padding: 0;
font-family: "Ubuntu", sans-serif;
}
body::-webkit-scrollbar {
width: 5px; background-color:
none; border-radius: 20px;
}
body::-webkit-scrollbar-thumb { background-
color: #adada; border-radius: 20px;
} body { max-
height: 100vh;
}
Backend
from backend import create_app
app = create_app() if
__name__ == '__main__':
from waitress import serve
serve(app, port=5000)
```

```
from dotenv import dotenv_values
from flask import Flask from
flask_cors import CORS import
ibm_db
# Get the environment variables config =
dotenv_values("backend/.env")
# Connect to db try:
# conn = 'dd ' conn
= ibm_db.pconnect(
f"DATABASE= { config[ 'DB2_DATABASE ' ] };
HOSTNAME={conf ig[ 'DB2_HOSTNAME' ] };
PORT={ config['DB2_PORT']}; SECURITY=SSL;
SSLServerCertificate=backend/DigiCertGlobalRootCA .crt;UID=
{config ['DB2_USERNAME']};PWD={config['DB2_PASSWORD']}", ", ")
print("Connected to IBM_DB2 successfully!!")
print(conn) except:
print("Failed to connect to Database!")
def create_app():
# Tell flask to use the build directory of react to
serve static content
app = Flask(__name__, static_folder='../build',
static_url_path='/')
CORS(app)
```

```
# Set the secret key for flask
app.config['SECRET_KEY'] = config['APP_SECRET']
# Import and register auth_router
from .auth_router import auth
app.register_blueprint(auth, url_prefix='/api/auth')
from .files_router import files
app.register_blueprint(files, url_prefix='/api/files')
from .user_router import user
app.register_blueprint(user, url_prefix='/api/user')
# In production serve the index.html page at root
@app.route("/")
def home():
return app.send_static_file('index.html')
return app
auth = Blueprint("auth", __name__)
LOGIN_FEILDS = ('email', 'password')
SIGNUP_FEILDS = ('name', 'email', 'phone_number', 'password')
@auth.route("/login", methods=['POST']) def
login_user():
      # Check if all the required feild are present
for feild in LOGIN_FEILDS:
        if not (feild in request.json):
           return jsonify({"error": f"All feilds are required!"}), 409
      email =request.json['email']
```

```
password = request.json['password']
     sql = f"select * from users where email='{email}"
stmt = ibm_db.prepare(conn, sql)
 ibm_db.execute(stmt)
                             user =
ibm_db.fetch_assoc(stmt)
                              if not user:
    return jsonify({"error": "Invalid credentials!"}), 401
    if bcrypt.checkpw(password.encode('utf-8'),
user["PASSWORD"].encode('utf-8')):
     token = jwt.encode(
{"email": email},
config["APP_SECRET"],
algorithm="HS256"
)
                return jsonify({"name": user["NAME"],
"email":
                  "phone_number":
           email,
user["PHONE_NUMBER"],
"token":
token}), 200
else:
return jsonify({"error": "Invalid credentials!"}),
401
    @auth.route("/signup", methods=['POST']) def
register_user():
       # Check if all the required feild are present
for feild in SIGNUP_FEILDS:
```

```
if not (feild in request.json):
         return jsonify({"error": f"All feilds are required!"}),409
   email = request.json['email']
      phone_number = request.json['phone_number']
name = request.json['name'] password =
request.json['password']
# Sql stmt to check if email/number is already in use
sgl = f"select * from users where email='{email}'
or phone_number='{phone_number}' "
stmt = ibm_db.prepare(conn, sql)
ibm_db.execute(stmt)
                           user
= ibm_db.fetch_assoc(stmt) if user:
       return jsonify({"error": f"Email/Phone number is alread in use!"}), 409
          # If user does not exist, then create account
hashed_password = bcrypt.hashpw(
                password.encode('utf-8'), bcrypt.gensalt())
                         sql = f"insert into
users(name,email,phone_number,password)
values('{name}','{email}','{phone_number}',?)" stmt =
ibm_db.prepare(conn, sql) ibm_db.bind_param(stmt, 1,
hashed_password)
         ibm_db.execute(stmt)
          token = jwt.encode(
{"email": email},
```

```
config["APP_SECRET"],
algorithm="HS256"
)
    return jsonify({"name": name, "email": email,
"phone_number": phone_number, "token": token}), 200
```

GITHUB & PROJECT DEMO LINK:

GITHUB link - https://github.com/IBM-EPBL/IBM-Project-53788-1661495800

PROJECT DEMO VIDEO link-

