

# WorkHub: ML Based System for Client And Worker

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**Abstract** - In today's rapidly evolving business landscape, efficient collaboration and resource allocation are critical for success. This proposed System addresses the challenges faced by both clients and workers in finding suitable partnerships and optimizing work execution. The system offers an intuitive interface that empowers clients to post work requirements and objectives easily. Detailed work descriptions, timelines, and budgets can be specified, ensuring that workers have a clear understanding of the expectations from the outset. Simultaneously, workers can create personalized profiles highlighting their experience, and expertise, providing clients with comprehensive insights into their capabilities. Through the integrated search and matching algorithm, the System intelligently pairs clients with the most qualified workers for their projects. The algorithm takes into account factors such as location, availability, and previous reviews to make accurate and relevant recommendations. Overall, the Client-Worker System revolutionizes the way clients and workers connect and collaborate. By offering a seamless and efficient online platform, it enhances productivity, optimizes project outcomes, and fosters long-term professional relationships. In a competitive and dynamic business environment, this System is poised to become an indispensable tool for clients and workers alike.

**Keywords** – WorkHub, ML, Client, Worker.

## I. INTRODUCTION

In today's fast-paced gig economy, finding the right worker for a job can be a daunting task. That's where WorkHub steps in, offering a cutting-edge solution to this age-old problem. Using advanced Machine Learning (ML) technology, WorkHub simplifies the process of matching clients with skilled workers. Traditional methods of directory searches and outdated referral systems are rendered obsolete, replaced by WorkHub's streamlined approach. This platform simplifies the process, enabling clients to find the perfect match with just a few clicks. Our advanced ML algorithms consider an array of factors including skills, experience, location, and availability to ensure precise and optimal connections. Through data-driven decision-making, WorkHub optimizes the hiring process, reducing time-to-hire and enhancing overall satisfaction for both clients and workers. Clients can confidently entrust their projects to skilled professionals, while workers gain access to a broader range of opportunities that align with their expertise. This research paper delves into the architecture, algorithms, and implementation of WorkHub, presenting a comprehensive analysis of its impact on the modern gig economy. Through rigorous evaluation and empirical

evidence, we demonstrate how WorkHub represents not only a platform, but a paradigm shift in the way we envision and facilitate collaborative work in today's world.

## II. RELATED WORK

In the realm of online recruitment platforms, person-job fit stands as a pivotal technique, enhancing hiring efficacy through precise alignment of job roles with qualified candidates. Notably, prior research predominantly emphasizes recommendation scenarios, inadvertently overlooking the significance of the search channel in connecting positions with potential candidates. It is worth highlighting that user search histories encapsulate invaluable behavioral cues, serving as pivotal indicators of job seekers' intentions. [1].

This system has been designed and implemented a recommendation system for online job-hunting. In this paper, this system contrast user-based and item-based collaborative filtering algorithm to choose a better performed one. System also take background information including students' resumes and details of recruiting information into consideration, bring weights of co-apply users (the users who had applied the candidate jobs) and

weights of student used-liked jobs into the recommendation algorithm. [2].

In two-sided markets with incomplete information, enduring matches often prevail due to limited knowledge of parties' characteristics. To mitigate moral hazard, dynamic matching mechanisms have been proposed. This system aims to refine the matching process, fostering mutual learning while discouraging workers from obscuring their attributes through their actions.[3].

In addressing the challenge of skills-based matching, this study contemplates the potential impact on gender distribution within occupations. Referred to as gender mobility, it pertains to job seekers exploring roles aligned with their skill set rather than prior experience. This research augments existing discourse on occupational diversity and gender representation, contributing to a nuanced understanding of skills-driven employment dynamics. [4].

This research addresses a scenario where workers possess specialized skills, and tasks are time-bound, budget-constrained, and skill-specific, encompassing activities like repairs, decoration, and event performances .The surge in mobile device adoption and the proliferation of crowdsourcing platforms has spurred interest in spatial crowdsourcing within the database community. This paradigm involves dispatching location-based tasks to workers based on their real-timepositions. [5]

The proposed algorithm in this research uniquely addresses this multi-dimensional matching scenario, presenting a significant advancement in the field. This algorithm can perform matching for workers with multiple skills hired for multiple jobs with multiple requirements [6].

This emphasis aims to ensure that machine learning and algorithmic tools for worker hiring remain impartial, avoiding unintended bias based on factors such as nationality or gender. Researchers and practitioners have particularly focused on optimizing team formation, matching algorithms, and devising equitable incentive schemes for task completion. [7].

The proposed inference algorithm, which encompasses worker clustering, skill estimation, and weighted majority voting, operates independently of any prior knowledge regarding worker or task types. Significantly, this algorithm attains impressive recovery accuracy, demonstrating superior performance compared to existing methods. [8].

One of the potent personalization technologies powering the adaptive web is collaborative filtering. Collaborative filtering (CF) is the process of filtering or evaluating items through the opinions of other people. CFtechnology brings together the opinions of large interconnected communities on the web, supporting filtering of substantial quantities of data. [9].

The proposed solution is designed to maintain a high standard of customer experience while remaining cost-effective for platform adoption. It focuses on ensuring fair wages, reasonable working hours, and transparent work availability. [10].

### III. SYSTEM ARCHITECTURE

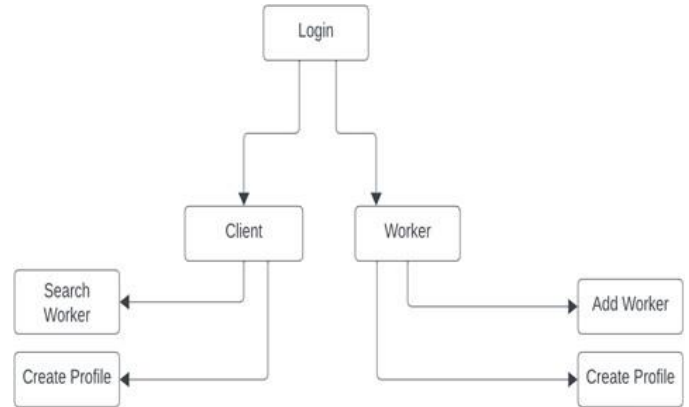


Fig. 3.1- Module 1: User Authentication and Dashboard

Module 1 encompasses the authentication and registration processes within the Android application. The core of the WorkHub experience lies in the shared dashboard accessible to both clients and workers. This dashboard provides a comprehensive view of available tasks and workers seeking opportunities.

Hire Worker (Client): Clients seeking workers for specific tasks can select this option to initiate the process of creating a task listing.

Add Worker (Worker): Workers looking to offer their services can utilize this option to register and create a comprehensive worker profile.

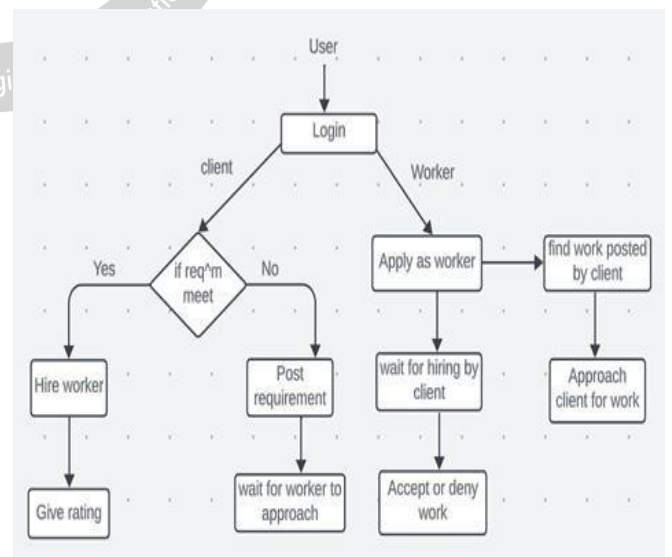


Fig 3.2 – Module2: Profile Management and Requirement Posting

Module 2 expands upon user functionalities, offering both clients and workers the ability to create comprehensive

profiles and conduct targeted searches.

### Task Posting and Requirement Submission

Clients have the capability to post new tasks directly from their profiles. This streamlined process allows clients to define task details, requirements, and compensation, ensuring that workers have clear instructions for task completion.

Clients can perform location-based searches to find workers in specific geographic areas, ensuring that tasks can be efficiently assigned to nearby workers.

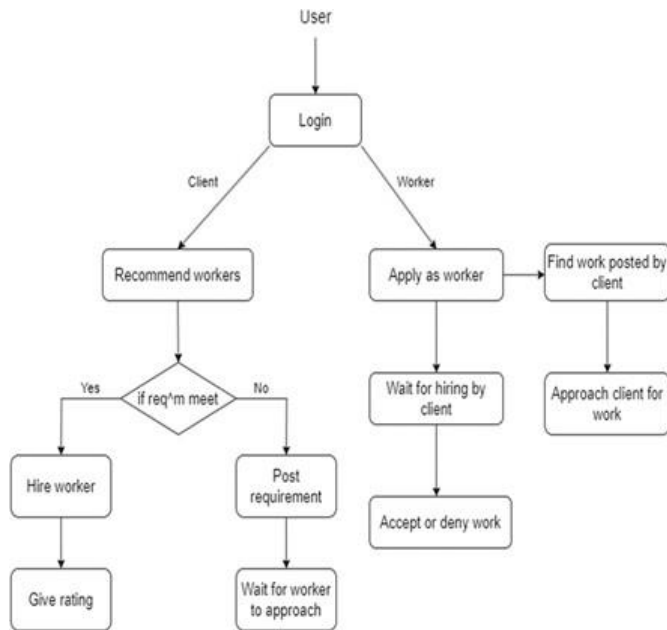


Fig 3.3- Module3 :Worker Recommendations and Hiring

Module 3 introduces enhanced communication features, facilitating seamless interactions between clients and workers.

Clients can now send SMS notifications to specific workers they wish to hire, streamlining the hiring process and ensuring timely communication.

Clients receive recommendations for workers based on their location and sorted by experience. This feature optimizes worker selection, providing clients with a curated list of potential candidates. Clients and Workers can update and delete their profiles, ensuring that their information remains accurate and relevant to potential clients.

## IV. IMPLEMENTATION

The system is developed using Kotlin and Java for mobile application development, with Firebase Database serving as the primary data management platform. The K-Nearest Neighbors (KNN) algorithm is integrated to calculate distances between clients and workers, facilitating precise recommendations based on geographical proximity and Experience.

Additionally, workers are categorized based on their type, category, ensuring that clients receive suggestions from the

most qualified candidates. This system is divided into 3 modules.

### Module 1: User Authentication and Dashboard

This module focuses on providing secure access to the WorkHub platform for both clients and workers. It includes:

**User Authentication:** Kotlin and Java are employed to manage user authentication, including functions for registration, login, and password recovery. Firebase authentication services provide a secure and reliable foundation for user management.

**Registration Process:** Guides new users through a registration process to create an account, providing necessary information and establishing their profile within the system.

**Dashboard:** Upon successful login, users were directed to their personalized dashboard. Notably, users had the ability to initiate two primary actions: "Add Worker" and "Hire Worker". These buttons facilitated the streamlined interaction between clients and workers.

### Module 2: Profile Management and Requirement Posting

**Profile Creation and Management:** Clients and workers have the ability to create, update, and delete their profiles. They can provide detailed information about their work, and contact details.

**Post Requirements:** Clients can post detailed work requirements, specifying work objectives, and budget constraints. This empowers them to clearly communicate their expectations from the outset.

**Provide Feedback:** Clients have the option to offer feedback to workers upon project completion. This helps build a feedback loop and ensures the quality of service delivery.

**View Posted Requirements:** Workers can view the work requirements posted by clients. This allows them to assess if the project aligns with their skills and availability before expressing interest.

**Receive Feedback:** After completing a project, workers receive feedback from clients. This serves as a valuable reputation-building tool and helps workers establish trust with future clients.

### Module 3: Worker Recommendations and Hiring

The system successfully generated worker recommendations for clients based on their specified location and sorted experience.

#### K-Nearest Neighbors Algorithm (KNN):

The K-Nearest Neighbors (KNN) algorithm is integrated to calculate distances between clients and workers. By computing distances based on location, and experience, the system provides accurate and relevant worker

recommendations.

**Worker Selection and Hiring:** Clients were able to review the recommended workers' profiles and select the most suitable candidates for their work. Clients can now send SMS notifications to specific workers they wish to hire, streamlining the hiring process and ensuring timely communication.

## V. RESULT & DISCUSSION

The WorkHub system presents a significant advancement in the domain of client-worker matchmaking through ML techniques, there are several avenues for future research and development that can enhance its capabilities and applicability. These potential areas of improvement are outlined below:

### Module1: User Authentication and Dashboard

In this section, we present the outcomes of the first module, which encompasses user authentication features, including login and registration, along with the user dashboard.



Fig 5.1 - Module1 :User Authentication and Dashboard

### Module 2: Profile Management and Requirement Posting

In this section, we present the outcomes of the second module, which encompasses profile management features for both clients and workers, as well as the functionality for clients to post project requirements and workers to view them.

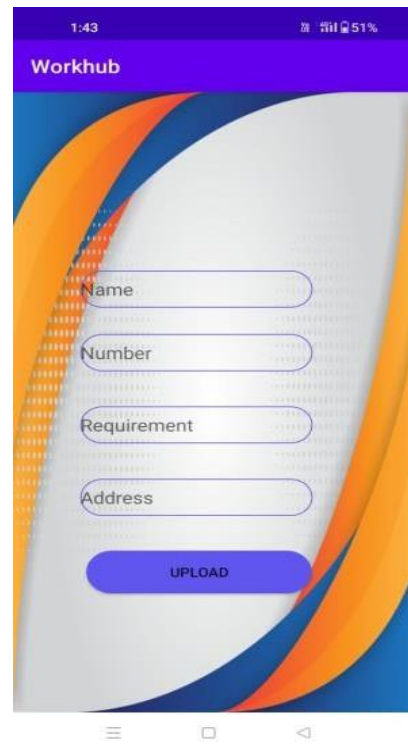


Fig 5.2 - Module 2: Profile Management and Requirement Posting

### Module3: Worker Recommendations and Hiring

In this section, we present the outcomes of the third module, which encompasses the functionality for clients to receive recommended workers based on location and sorted experience. Additionally, clients have the ability to hire workers for their work.

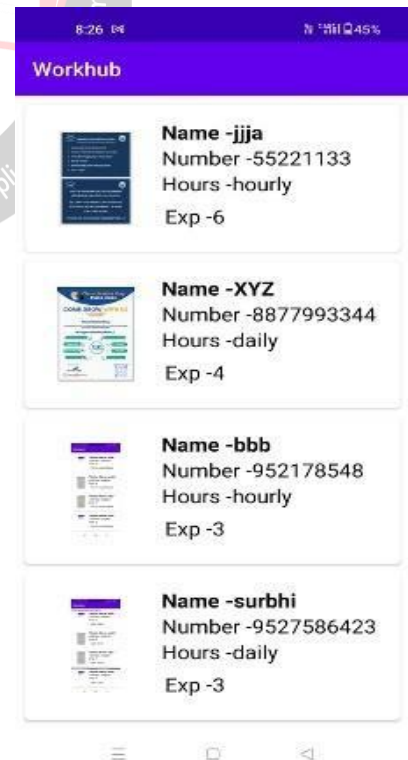


Fig 5.3 -. Module 3: Worker Recommendations and Hiring

## VI. CONCLUSION & FUTURE SCOPE

This system is a valuable resource for anyone who needs help finding or providing services. Whether you are a client looking for help with a specific task, or a worker looking for new opportunities, this system can help you connect with the right people. This system is committed to providing a safe and secure platform for our users, and System take great care to vet all of our workers. This System also offer a variety of features to help clients and workers find the perfect match, including: A detailed profile system that allows users to share their skills, experience, and availability and A search engine that allows users to find workers by skill, location, and availability. We are confident that our system can help you achieve your goals. Whether you are looking for help with a one-time work or a long-term commitment, this System can help you find the right person for the work.

**Integration of Payment Gateway:** Introduce a secure payment gateway to facilitate seamless transactions between clients and workers. This would enhance the platform's functionality and provide an end-to-end solution for task management.

**In-App Messaging System:** Develop an in-app messaging system to facilitate direct communication between clients and workers. This would streamline communication and reduce reliance on external messaging platforms.

## VII. REFERENCES

- [1] Hou, Y. et al. (2022). Leveraging Search History for Improving Person-Job Fit. In: Bhattacharya, A., et al. Database Systems for Advanced Applications. DASFAA 2022. Lecture Notes in Computer Science, vol 13245. Springer, Cham.
- [2] Zhang, Y., Yang, C., Niu, Z., A research of job recommendation system based on collaborative filtering. In: ISCID (2014)
- [3] Kartik Ahuja, Mihaela Van der Schaar: Dynamic Matching and Allocation of Tasks. in TEAC Volume 7, Issue 4(2019)
- [4] Ajaya Adhikari, Steven Vethman, Daan Vos, Marc Lenz, Ioana Cocu, Ioannis Tolios, Cor J. Veenman: Gender mobility in the labor market with skills-based matching models in AAAI Spring Symposium 2023.
- [5] TPeng Cheng; Xiang Lian; Lei Chen; Jinsong Han; Jizhong Zhaoask :Assignment on Multi- Skill Oriented Spatial Crowdsourcing in IEEE Transactions on Knowledge and Data Engineering ( Volume: 28, Issue: 8, 01 August 2016)
- [6] Philip Ekman, Sebastian Bellevis, Christos Dimitrakakism: Learning to Match at the VAMS Recsys workshop 2017.
- [7] Giorgio Barnabò, Adriano Fazzino, Stefano Leonardi : Algorithms for Fair Team Formation in Online Labour Marketplaces in WWW '19: The Web Conference(2019)
- [8] Doyeon Kim; Hye Won Chung.: Crowdsourced Labeling for Worker-Task Specialization Model in 2021 IEEE International Symposium on Information Theory (ISIT)
- [9] Schafer, J.B., Frankowski, D., Herlocker, J., Sen, S. (2007). Collaborative Filtering Recommender Systems. In: Brusilovsky, P., Kobsa, A., Nejdl, W. (eds) The Adaptive Web. Lecture Notes in Computer Science, vol 4321. Springer, Berlin, Heidelberg. [https://doi.org/10.1007/978-3-540-72079-9\\_9](https://doi.org/10.1007/978-3-540-72079-9_9)
- [10] Ashish Nair, Rahul Yadav, Anjali Gupta, Abhijnan Chakraborty, Sayan Ranu, Amitabha Bagchi : Gigs with Guarantees: Achieving Fair Wage for Food Delivery Workers Appeared in International Joint Conference on Artificial Intelligence (IJCAI) 2022