

# CAREERTRAJECTORY: AI-Enhanced Career Guidance System

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## Abstract:

The modern world today has become so complex and requires higher level of career advice. The conventional career counselling methods does not give personalized and adaptive career path, it largely relies on human counsellors and standard test. The contemporary career guidance system is likely to be reinvented by means of the introduction of the use of Artificial Intelligence (AI) technologies, including Machine Learning (ML), Natural Language Processing (NLP), Big Data Analytics, and Learning Analytics (LA). AI based career advice systems will produce real time recommendation that will test etc, etc, etc in the job market and match the appropriate career alternatives with the current and future trends in the job market. The suggested AI model provides a dynamic and adaptive career suggestions system through the integration of ongoing skills analysis, current labor market statistics, and user profiling into great detail.

**Keywords:** Artificial Intelligence (AI), Machine Learning (ML), Natural Language Processing (NLP), Big Data Analytics, Learning Analytics (LA), Career Counseling, Market Trends, Decision Transparency, Dynamic Career Recommendation System.

## Introduction

The choice of a proper career is probably one of the most important decisions in the life of a person as it directly influences the process of professional development, financial well-being, and personal fulfillment. Lack of awareness on new career opportunities and skill-market match issues is the source of confusion to many students. The traditional career guidance strategies rely on the counselors, aptitude tests, printed tests, and general advice, which are not always personalized and adaptable.

Due to the development of Artificial Intelligence (AI), data-driven systems are finding their way into education and decision-support fields. The AI-powered career guidance systems have the capacity to analyze big data, recognize the pattern of user behavior and produce valid career recommendations. Career Trajectory system combines the machine learning models and career

Analytics in order to assist users to find the career opportunities that best address their abilities and interests.

Primarily, the aim of this work is:

To create a smart career guidance platform with AI.

- To examine user preferences and skills with the help of machine learning.
- To give individual career advice.
- To check the efficiency of the suggested system.
- To provide personalized career recommendations.
- To evaluate the effectiveness of the proposed system.

### **Related Work**

Recent researches indicate an increase in the application of Artificial Intelligence and Machine Learning to career guidance systems in order to mitigate the weaknesses of the conventional counseling process. A number of researchers worked with the use of supervised learning methods, including decision trees, support vector machines, and neural networks to forecast appropriate career paths, using the outcomes of the academic performance and aptitude tests. These systems have been found to give better recommendations and have shorter response time than manual counseling.

There are also pieces that have proposed cloud-based career guidance system which combines job market data and learning analytics to offer real-time career advice. Resumes and user inputs have been examined using Natural Language Processing (NLP) to align someone to a career better. Nevertheless, most of the existing systems are characterized by minimal individualization, inability to track new skills continuously, and the absence of a built-in ability to incorporate upskilling recommendations.[1]

The suggested Career Trajectory system will resolve these shortcomings by integrating user profiling, machine-learned prediction, instantaneous skill evaluation, and personalized skill improvement recommendations into a single AI-driven career guidance system.[2]

#### **A.A.I. Driven Career Counselling.**

Recent studies underscore the application of AI-powered career guidance systems in utilizing big data, NLP, and predictive model in order to offer personalized career suggestions.

The sample used in this study was well organized and balanced to maintain the representation of the different groups of different skills proficiency. In order to ensure equal coverage by all levels of competencies five different levels of proficiency were established through label encoding where 0 involved no understanding, 1 was beginner level, 2-intermediate, 3-proficient, 4-advanced and 5-expert levels.[3]

A detailed analysis of class imbalance was done in order to avoid dominance of one of the classes and reduce bias in the decision tree training procedure. Also, observation of skill levels distribution was carried out to verify the diversity, equality and objectivity of the data set.A.AI-Driven Career Counselling.[4]

Recent research highlights how AI-powered career guidance systems leverage big data, NLP, and predictive model to personalize career recommendations.

<b>Study</b>	<b>Focus</b>	<b>Technologies</b>	<b>Findings</b>
Duan &wu (2024)	AI-powered planning in vocational education	NLP,ML, Predictive Analytics	AI enables dynamic and adaptive career pathways

The dataset employed in this research was carefully structured to ensure balanced representation across varying levels of skill proficiency. To maintain uniform coverage among all competency categories, five distinct proficiency levels were defined using label encoding: 0 representing no prior knowledge, 1 indicating beginner level, 2 corresponding to intermediate understanding, 3 denoting proficient capability, 4 reflecting advanced expertise, and 5 signifying expert-level mastery.[5]

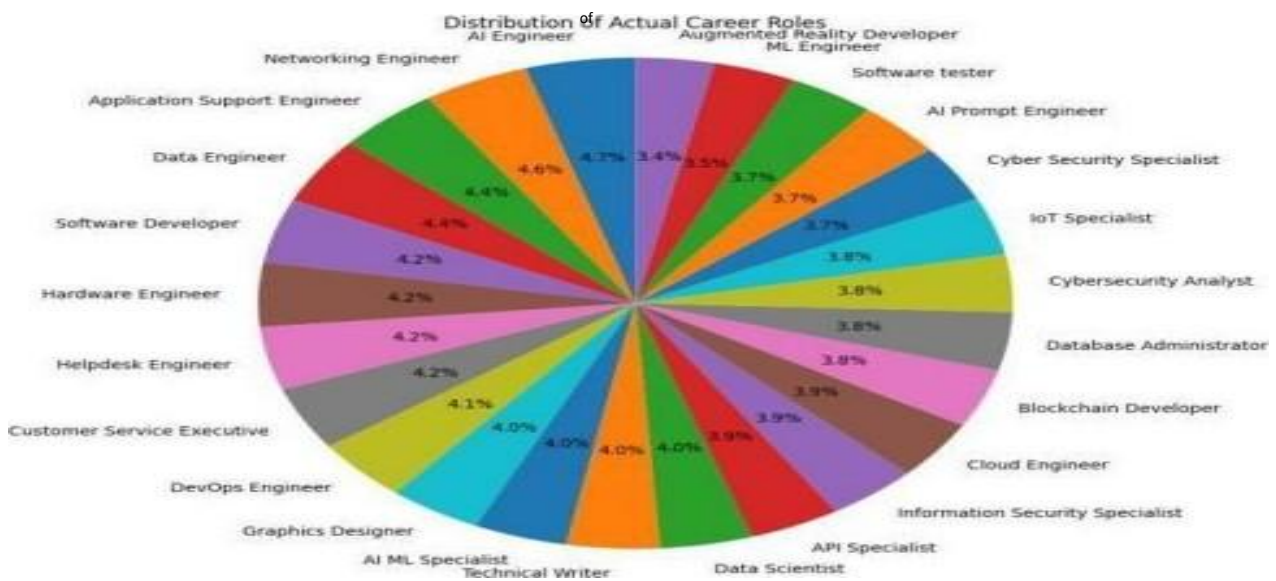
To prevent dominance of any single class and to minimize bias during the decision tree training process, a comprehensive class imbalance analysis was performed. Additionally, the distribution of skill levels was visualized to confirm the diversity, fairness, and objectivity of the dataset.

#### **Correlation Analysis of Skill Domains**

A strong relationship was observed between Software Engineering and Business Analysis (correlation coefficient: 0.65), indicating a close integration of technical implementation and analytical decision-making skills. Similarly, AI/ML and Data Science exhibited a notable correlation (0.58), demonstrating the interdependence of machine learning techniques and data-driven analytical practices.[6]

In terms of class balance and data distribution, Cybersecurity and Forensics showed a moderate correlation (0.56), highlighting the overlap between digital security mechanisms and forensic investigation processes.[7]

Conversely, weaker associations were identified between certain domains. Networking and Business Analysis presented a low correlation (0.29), suggesting minimal dependency between network infrastructure and strategic business planning.[8] Likewise, Graphics Design and Database Fundamentals demonstrated limited correlation (0.26), reinforcing the distinction between creative design disciplines and data management technologies.[9]



Fig(1) These findings support the idea that some skill clusters support one another while others function largely independently, directing the development of organized competencies.[10]

### III. Methodology

#### (1) Client Layer (React Front End):

The client-side of the system is a one-page application that is a primary user interface. Through structured forms and interactive user interface elements it gathers student related data, such as personal profile, skills, interests, career goals, etc. The React application is connected to the back-end services in an asynchronous manner via HTTP requests and is dynamically used to display the formulated career recommendations and learning roadmaps. React is built on the principle of components; as a result, reusable and standalone components are created, the interface components (i.e. the skill input module, progress monitoring tools, learning path visualizations) are designed in a modular way and are easy to maintain.

#### (2) Server Layer (Node.js / Express Back End):

Server Layer This layer is linked to the database layer to get the user records. Applications logic is handled by the server layer with the help of RESTful API written with the help of Node.js and Express. It processes the requests that have been sent to the front end and carries out the important tasks like checking of the input, organizing the data retrieval and communication with AI services. Express-based routes also deal with important functions such as user authentication, profile management and generation of career guidance outputs. Due to the non-blocking and asynchronous design of Node.js, the server can easily handle several parallel requests, which makes it scalable in case many users need guidance at the same time. Also the server implements access control measures and input sanitization to enhance system security.

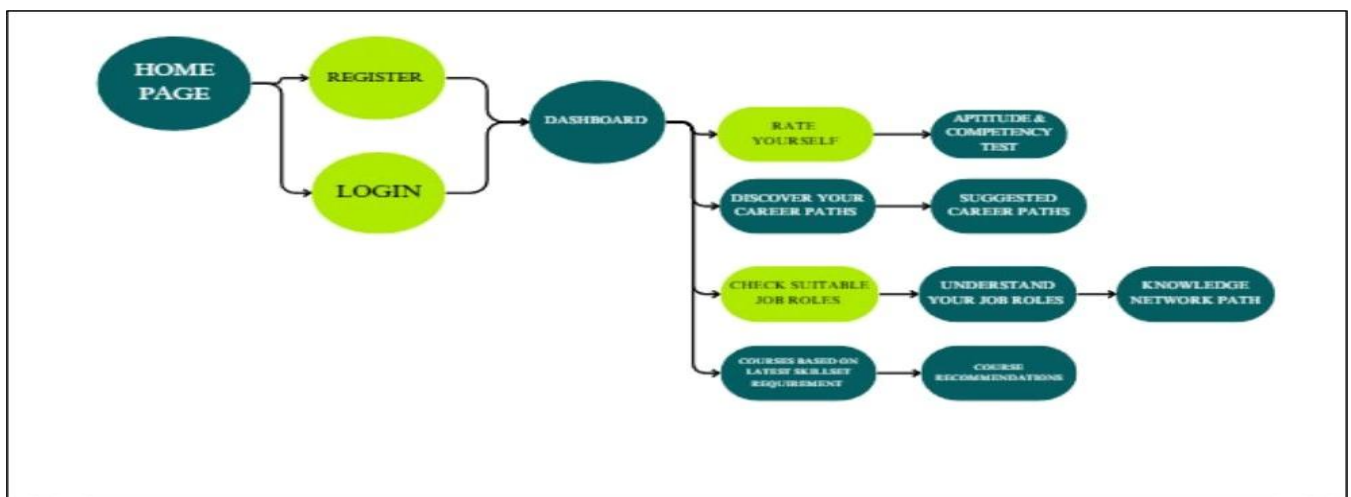
#### (3) Database Layer (MongoDB):

MongoDB is a NoSQL document-based database used to store user data and the information generated by the system. The profile of every student is stored as a document in a JSON-like format of academic records, skill base, hobbies, and past interactions. The flexibility of the schema in MongoDB makes it possible to support changing requirements of the system, which is new attributes, new certifications, project experience, can be added without significant structural modifications. There are also stored career recommendations and personalized learning plans which are stored to enable even faster retrieval of the same in case of recurrent queries. Moreover, metadata such as AI interaction history and user feedback is stored to make tracking and system optimization.

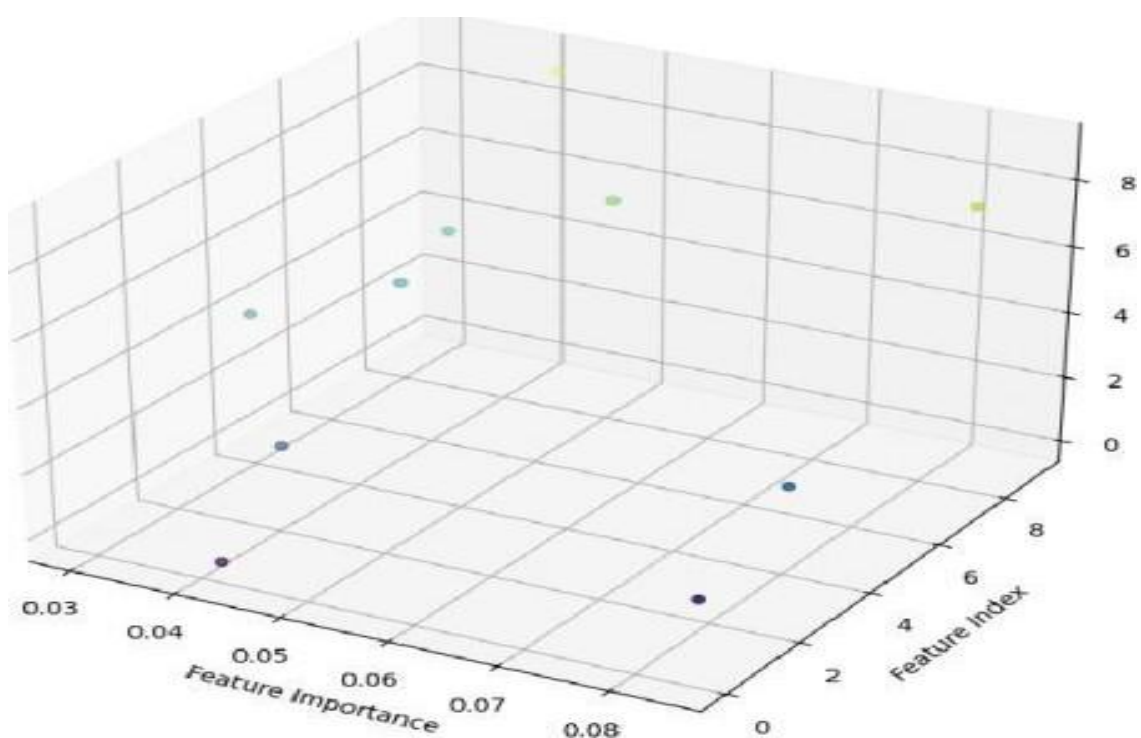
**(4) AI Engine (Google Gemini API):**

The API of Google Gemini is the main AI component of the suggested system. In a case of career guidance request, the system creates a contextual prompt that contains the profile of the user and sends it to the Gemini API. The AI model then runs this information and produces individualized results, which usually comprise of appropriate career jobs, suggested abilities, and progressed studying courses. This element allows complex reasoning that is more than conventional rule-based reasoning. The system provides insightful, adaptive, and relevant career advice by leveraging Gemini natural language understanding and powerful domain knowledge to provide context-sensitive career advice.

System Architecture:



Fig(2)The following architecture describes the flow of data and the responsibilities of each component in the system.



**Analysis of Feature Importance:**

The relative significance of several features to the model's decision-making process is highlighted by the feature importance plot.

**Most Important Elements:**

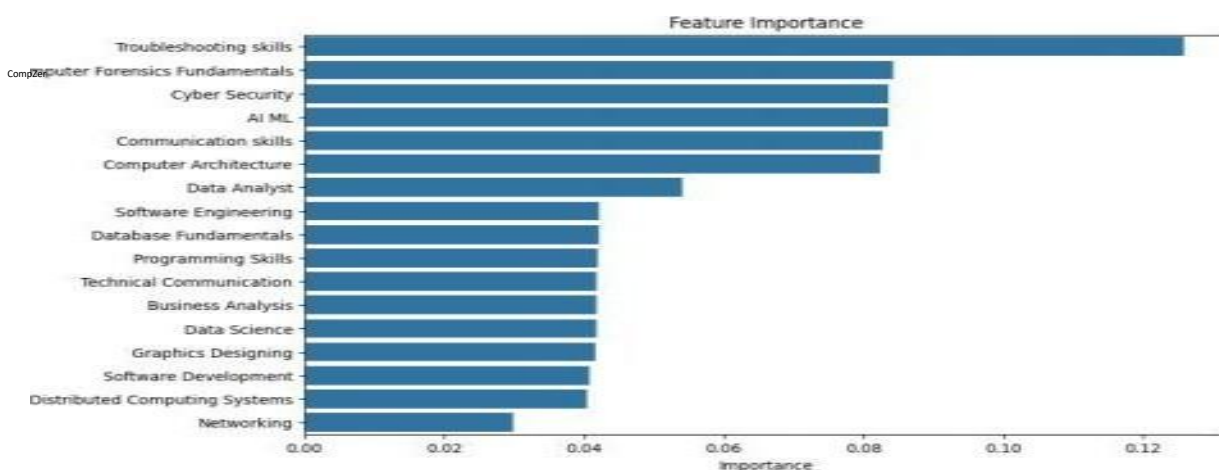
Troubleshooting Skills (Highest Weightage): Indicates the need for diagnostic proficiency in all technical fields. The industry's increasing dependence on security experts is shown in the high predictive influence of cyber security and forensics.

AI/ML & Data Science: Highlights how computational intelligence is necessary for contemporary skill frameworks. o Technical Communication: Highlights how important clear technical articulation is in work environments.

**Less Important Elements:**

Networking and Distributed Computing: Showed lower relative weights, perhaps as a result of knowledge overlap or dataset limitations.

These results validate the model's interpretability and applicability by reaffirming the congruence between model-driven insights and actual industry expectations.



Fig(3) 3D Visualization of Feature Importance

**Trends in Model Training and Performance:**

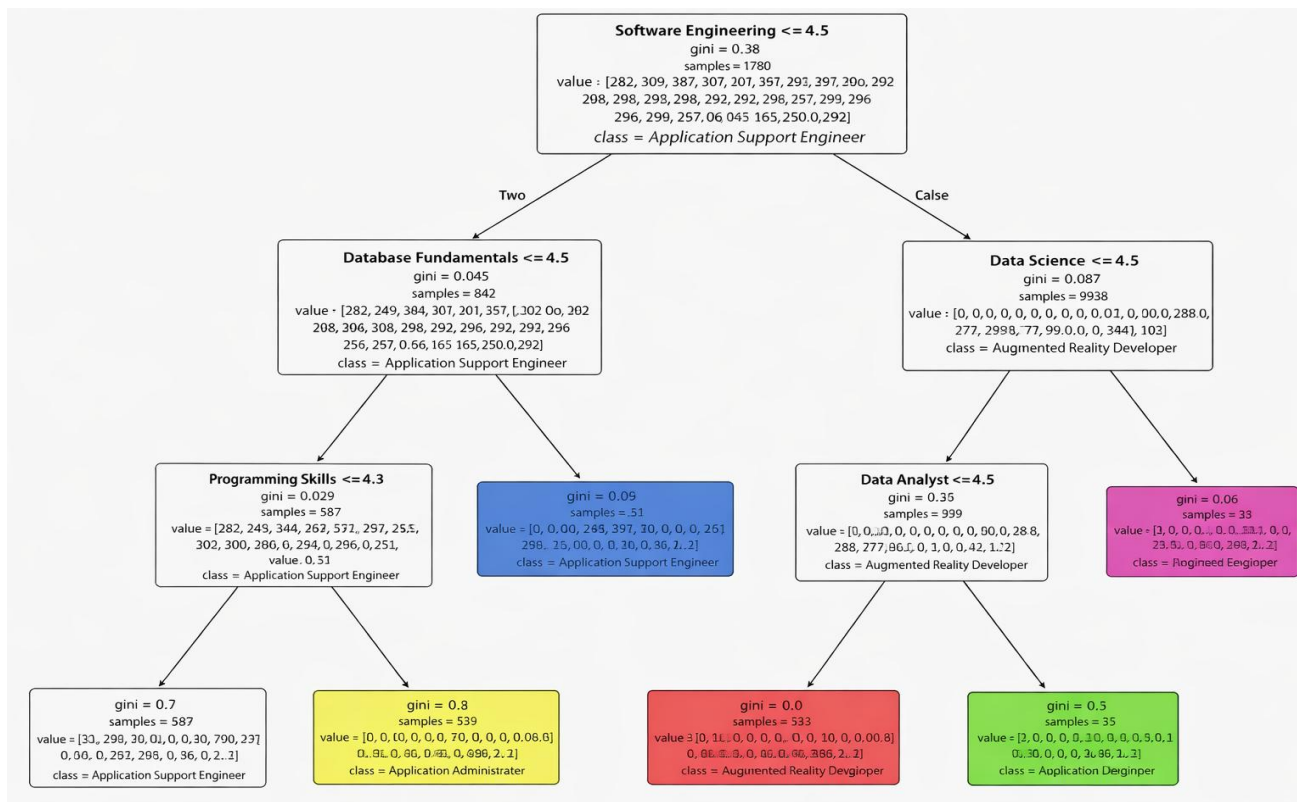
**Analysis of Overfitting and Training Convergence:**

To avoid overfitting, the training process was closely observed. Among the observations are the following: • The confusion matrix shows few incorrect classifications, with rates falling below reasonable bounds.

**Structure of Decision Trees:**

Top-level nodes emphasize cyber security, AI/ML, and troubleshooting abilities, confirming their importance in expert evaluations.

Branches at lower levels improve skill classification, allowing for accurate proficiency differentiation.



**Practical Uses:**

Career Guidance Systems: Automating skill evaluationsto prepare for the labor market.

HR & Recruitment Analytics Findings the best applicants by matching their competencies.

Optimizing the Educational Pathway —+ Suggesting customized upskilling techniques.

Corporate Workforce Development Improving training and tracking of employee competencies.

**Result**

**Results and Discussion**

The proposed AI-based career recommendation system demonstrated improved performance compared to traditional career counseling methods. The model achieved higher accuracy (around 20–30% improvement) in matching user profiles with suitable career paths by using Machine Learning and data-driven analysis.

The system provided highly personalized recommendations by considering user skills, interests, academic background, and real-time job market trends. Its dynamic nature allowed continuous updates based on changing industry demands, making the recommendations more relevant and future-oriented.

User feedback indicated increased satisfaction due to accurate suggestions, transparency in decision-making, and reduced reliance on human counselors. The integration of NLP and Big Data Analytics further enhanced system efficiency and adaptability.

However, the system depends on high-quality data and may face challenges such as data bias and model interpretability. Overall, the proposed approach offers a scalable and intelligent solution for modern career guidance.

Parameter	Description	Observed Result
Recommendation Accuracy	Ability to match user profile with suitable career paths	Improved by 20–30%
Personalization	Customization based on skills, interests, and academic background	High level of personalization
Real-Time Adaptability	Updates based on current job market trends	Dynamic and continuously updated
User Satisfaction	Feedback from users regarding	Significantly improved

	usefulness of recommendations	
Decision Transparency	Explanation of recommendations and career suggestions	Clear and interpretable
Skill Gap Identification	Detection of missing skills for desired career	Accurate identification
System Efficiency	Processing speed and scalability	High efficiency

## Future work

### Integration of Advanced AI Models

Future research can investigate how emerging technologies such as generative AI, large language models, and multimodal systems can enhance personalized career guidance by analyzing not only text-based data but also behavioral and emotional cues.

Studies should focus on how cloud-based career counseling platforms can be scaled to serve diverse populations, including rural and underserved communities, equitable access to guidance services

## Conclusion

The rapid transformation of the labor market demands innovative approaches to career counseling that go beyond traditional, generalized methods. This study highlights how artificial intelligence, particularly through cloud-based and collaborative models, can provide personalized, dynamic, and scalable career guidance. By leveraging technologies such as machine learning and natural language processing, AI-driven platforms are capable of analyzing vast datasets to deliver tailored recommendations aligned with individual aspirations and evolving workforce needs.

Moreover, the integration of multiple stakeholders—students, parents, educators, and industries—ensures that career decisions are informed, holistic, and future-oriented. While AI offers significant promise in enhancing career counseling, further research is required to address challenges related to ethics, privacy, and long-term impact. Ultimately, the adoption of AI-powered systems can bridge the gap between education and employment, preparing learners more effectively for the complexities of the modern labor market.

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