

# Alumni Connect: An Automated Platform for Alumni Engagement and Career Readiness

Renuka Gavli<sup>1</sup>, Yash Rakhunde<sup>2</sup>, Pavan Tamhane<sup>3</sup>, Sourabh Nayak<sup>4</sup>

<sup>1</sup>*Professor, Department. of Computer Engineering, Ajeenkya D Y Patil School of Engineering, Pune India.*

<sup>2,3,4</sup>*Department. of Computer Engineering, Ajeenkya D Y Patil School of Engineering, Pune, India.*

**Abstract**—In the contemporary educational ecosystem, alumni networks represent a significantly underutilized asset for student career development and institutional growth. Traditional alumni management systems, reliant on manual processes and static databases, suffer from data inaccuracy, poor engagement, and a lack of personalized interaction. This paper proposes "Alumni Connect," an intelligent and automated platform designed to bridge this gap. The system leverages process automation for robust alumni data management and integrates AI-driven features, including a recommendation engine for mentorship matching and sentiment analysis for monitoring interaction quality. Built with a scalable microservices architecture using React.js and Flask, the platform facilitates real-time communication through discussion forums, a dedicated job portal, and direct mentorship requests. By automating the entire data lifecycle and fostering a dynamic, interactive community, Alumni Connect effectively enhances student-alumni engagement, reduces administrative overhead, and strengthens the professional network, thereby preparing students for a successful transition from academia to industry.

**Index Terms**—Alumni Management, Process Automation, Microservices Architecture, AI Recommendations, Sentiment Analysis, Real-time Communication, Career Readiness.

## I. INTRODUCTION

Alumni networking represents a significantly underutilized resource in contemporary educational ecosystems, often overshadowed by formal academic guidance structures. Despite this undervaluation, alumni play an irreplaceable role in guiding students through critical professional transitions—whether facilitating industry settlement, aiding adaptation to

real-world work cultures, or navigating the complexities of career progression. While academic faculty primarily focus on theoretical knowledge dissemination, alumni offer practical, life-oriented advice drawn from personal professional experiences, effectively bridging the gap between academic learning and practical application. They provide invaluable insights into job hunting strategies, interview preparation techniques, and real-world challenge resolution, serving as relatable mentors who have recently traversed similar educational and career pathways. The significance of alumni engagement extends beyond mere knowledge transfer. Alumni typically maintain deep emotional connections to their alma mater and demonstrate strong motivation to support students from shared academic backgrounds [1]. This intrinsic connection frequently manifests through professional recommendations, career placement assistance, and collaborative initiatives with their former institutions. However, conventional alumni networks consistently lack the technological infrastructure necessary to cultivate these relationships effectively. Despite increasing recognition of alumni networks' importance, existing platforms commonly suffer from three fundamental limitations: restricted personalization capabilities, inadequate real-time communication tools, and poor integration with institution-specific requirements [2].

This research directly addresses these critical gaps through the development of an intelligent, unified platform that enables meaningful engagement among students, alumni, and administrative stakeholders. By leveraging cutting-edge technologies including artificial intelligence, process automation, and real-

time communication systems, the proposed solution aims to transform traditional alumni networks into dynamic, self-sustaining ecosystems that strengthen intergenerational bonds while enhancing institutional value and resource development.

## II. RELATED WORKS

Recent research has demonstrated significant advancements in methodological frameworks for analyzing alumni engagement and institutional reputation enhancement. Several studies have employed innovative approaches to leverage alumni data for improving educational ecosystems and stakeholder relationships.

**A. Methodological Frameworks for Alumni Analysis.** Multiple research initiatives have developed sophisticated frameworks for understanding alumni impact on institutional reputation. A prominent methodological framework utilizes web-crawling techniques to systematically gather comprehensive data on alumni professional trajectories and network connections [12]. This approach enables educational institutions to maintain updated alumni profiles through automated data collection processes. The framework further incorporates sophisticated data ranking algorithms that process alumni profiles to generate prioritized datasets based on multiple parameters including professional achievements, engagement history, and network influence [13]. Visualization systems present this analyzed data through multi-faceted interfaces, enabling administrators to identify key alumni contributors and potential engagement opportunities.

### B. AI-Enhanced Alumni Platforms :

Natural Language Processing techniques have been increasingly integrated into alumni platforms to enhance user interaction and content management. Several portal implementations utilize NLP algorithms for automated content summarization and intelligent tagging based on user preferences and interaction

history [15]. These systems dynamically adapt to individual user behaviors, creating personalized engagement experiences that improve platform retention and utility.

### C. Predictive Analytics in Alumni Management.

Statistical modeling approaches have shown significant promise in forecasting alumni engagement patterns. Auto Regressive Integrated Moving Average (ARIMA) modeling has been successfully applied as a statistical methodology to predict alumni association membership trends based on historical data analysis [17]. These predictive models enable institutions to anticipate engagement fluctuations and optimize resource allocation for alumni relations initiatives.

### D. Integrated Alumni Information Systems

The ASTAIKAMAHI web-based alumni information system addresses critical challenges in manual data management and communication infrastructure. Developed using a structured waterfall methodology, this system provides comprehensive features including dynamic profile management, news dissemination mechanisms, job vacancy portals, and activity coordination tools [18]. The system's architecture demonstrates how structured development methodologies can produce robust platforms that enhance alumni participation through streamlined communication channels and centralized information management.

### E. Research Gaps and Opportunities

Despite these advancements, current literature reveals several significant research gaps. Many existing systems lack comprehensive AI integration for personalized mentorship matching, exhibit limited real-time communication capabilities, and demonstrate insufficient scalability for large institutional deployments. Furthermore, few platforms successfully integrate automated data collection, intelligent recommendation systems, and sentiment analysis within a unified architectural framework.

Table I. Comparative Analysis of Traditional &amp; Existing Systems.

| Sr.No. | Paper Name                                                                                       | Author                | Summary                                                                                                                                                                                                    | Limitations                                                                                                                                                              |
|--------|--------------------------------------------------------------------------------------------------|-----------------------|------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------|--------------------------------------------------------------------------------------------------------------------------------------------------------------------------|
| 1]     | AI-Driven Alumni Networking Platform for Enhanced Student-Alumni Engagement and Career Readiness | Kunekar et al.        | <ul style="list-style-type: none"> <li>• AI-driven platform with NLP techniques</li> <li>• Real-time communication features</li> <li>• Microservices architecture with Redis caching</li> </ul>            | <ul style="list-style-type: none"> <li>• Platform-specific restrictions</li> <li>• Scalability challenges in real-time features</li> </ul>                               |
| [2]    | Alumni Information Collection and Management                                                     | L.L. Dang             | <ul style="list-style-type: none"> <li>• Computational methods for data collection</li> <li>• Web scraping and NLP for verification</li> <li>• Automated alumni database creation</li> </ul>               | <ul style="list-style-type: none"> <li>• Platform restrictions and data verification challenges</li> <li>• Legal and ethical concerns with scraping</li> </ul>           |
| [3]    | Designing an AI-Powered Mentorship Platform for Professional Development                         | Bagai & Mane          | <ul style="list-style-type: none"> <li>• Scalable personalized mentorship using AI</li> <li>• Comprehensive implementation framework</li> <li>• Addresses skill enhancement and career guidance</li> </ul> | <ul style="list-style-type: none"> <li>• Data privacy risks</li> <li>• Algorithmic bias concerns</li> <li>• Lacks emotional intelligence of human mentors</li> </ul>     |
| [4]    | Retrieving and Classifying LinkedIn Job Titles for Alumni Career Analysis                        | Lei et al.            | <ul style="list-style-type: none"> <li>• Automated job title classification (73% accuracy)</li> <li>• Machine learning algorithms</li> <li>• Automated crawling engine</li> </ul>                          | <ul style="list-style-type: none"> <li>• LinkedIn data usage policy compliance</li> <li>• Job title classification accuracy limitations</li> </ul>                       |
| [5]    | Development of Intelligent Alumni Management System for Universities                             | Khan et al.           | <ul style="list-style-type: none"> <li>• Enhanced university-alumni communication</li> <li>• Job opportunities and profile management</li> <li>• Networking facilitation</li> </ul>                        | <ul style="list-style-type: none"> <li>• Alumni engagement maintenance</li> <li>• Continuous system updates needed</li> <li>• Personal data security concerns</li> </ul> |
| [6]    | Construction of Alumni Information Analysis Model Based on Big Data                              | X. Wang               | <ul style="list-style-type: none"> <li>• Improved hierarchical clustering algorithm</li> <li>• Text preprocessing and feature extraction</li> <li>• Name disambiguation and recognition</li> </ul>         | <ul style="list-style-type: none"> <li>• Complex web structures</li> <li>• Non-standard data formats</li> <li>• Limited experimental conditions</li> </ul>               |
| [7]    | Supervised machine learning predictive analytics for alumni income                               | Gomez-Cravioto et al. | <ul style="list-style-type: none"> <li>• Statistical significance (<math>p &lt; 0.05</math>)</li> <li>• Gender wage gap identification</li> <li>• Data binning techniques</li> </ul>                       | <ul style="list-style-type: none"> <li>• Excludes alumni &gt;60 years</li> <li>• Limited data availability</li> <li>• Potential age bias</li> </ul>                      |

|      |                                                                          |                   |                                                                                                                                                                                                      |                                                                                                                                 |
|------|--------------------------------------------------------------------------|-------------------|------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------|---------------------------------------------------------------------------------------------------------------------------------|
| [8]  | Investigating Factors Affecting the Learning Process in Higher Education | W.A. Mohammed     | <ul style="list-style-type: none"> <li>• Communication networks in education</li> <li>• Structured guidance systems</li> <li>• Learning opportunity enhancement</li> </ul>                           | <ul style="list-style-type: none"> <li>• Limited to communication networks</li> <li>• Narrow educational focus</li> </ul>       |
| [9]  | Knowledge providing method by posting information in forums              | Chen & Demmans    | <ul style="list-style-type: none"> <li>• Organized knowledge-sharing in forums</li> <li>• Academic and career information distribution</li> <li>• Forum-based communication effectiveness</li> </ul> | <ul style="list-style-type: none"> <li>• Limited to forum-based knowledge sharing</li> <li>• Lack of personalization</li> </ul> |
| [10] | Selecting suitable job careers by alumni approach                        | Comendador et al. | <ul style="list-style-type: none"> <li>• Alumni-driven career guidance</li> <li>• Career path selection support</li> <li>• Professional alignment methods</li> </ul>                                 | <ul style="list-style-type: none"> <li>• Narrow focus on career selection</li> <li>• Limited scope of application</li> </ul>    |

### III. METHODOLOGY & DISCUSSION

#### A. System Architecture Design

The Alumni Networking Platform employs a sophisticated multi-tiered architecture designed for scalability, maintainability, and robust performance. The system integrates modern web technologies with artificial intelligence capabilities to create a comprehensive alumni engagement ecosystem, as illustrated in Fig. 1.

#### Frontend Implementation:

Framework: React.js was selected for its component-based architecture, enabling easy scalability and lightweight structure

-Styling: Tailwind CSS facilitated responsive and cohesive design implementation with minimal custom CSS, allowing faster styling through utility-first classes

-User Experience: The interface was optimized for intuitive navigation across multiple device types

#### Backend Infrastructure:

-Framework: Node.js with Express.js provided non-blocking I/O capabilities, essential for handling multiple concurrent requests in a real-time networking environment

-API Architecture: RESTful APIs ensured seamless communication between frontend and backend components

#### B. Data Management Strategy

##### Data Collection and Storage:

User data, including comprehensive profiles, resumes, skills, career goals, industries, interests, location data, and preferred communication methods, was collected and stored securely using hybrid database approach:

-MySQL: For structured relational data requiring ACID compliance

-MongoDB: For flexible document storage of unstructured user data

-PostgreSQL: Implemented for powerful relational querying capabilities on structured alumni profiles and network data

##### Caching Implementation:

-Redis: Deployed as an in-memory data store to improve response times for frequently accessed data

-Performance Outcome: Redis caching significantly reduced page load times, particularly for high-traffic data like job listings and alumni profiles, enhancing overall user experience

#### C. AI and Machine Learning Integration

##### Natural Language Processing (NLP):

-Frameworks: TensorFlow and Py Torch were utilized for developing sophisticated NLP models

-Capabilities:

-Analysis of user profiles and resumes for key

information extraction

-Identification of nuanced relationships through keyword processing and semantic similarity analysis  
- Contextual data understanding (e.g., connecting "AI" with "Data Science" domains)

Collaborative Filtering System:

-Implementation: Advanced collaborative filtering model analyzed user interactions and shared attributes.  
Personalization: The system provided tailored connection and group suggestions

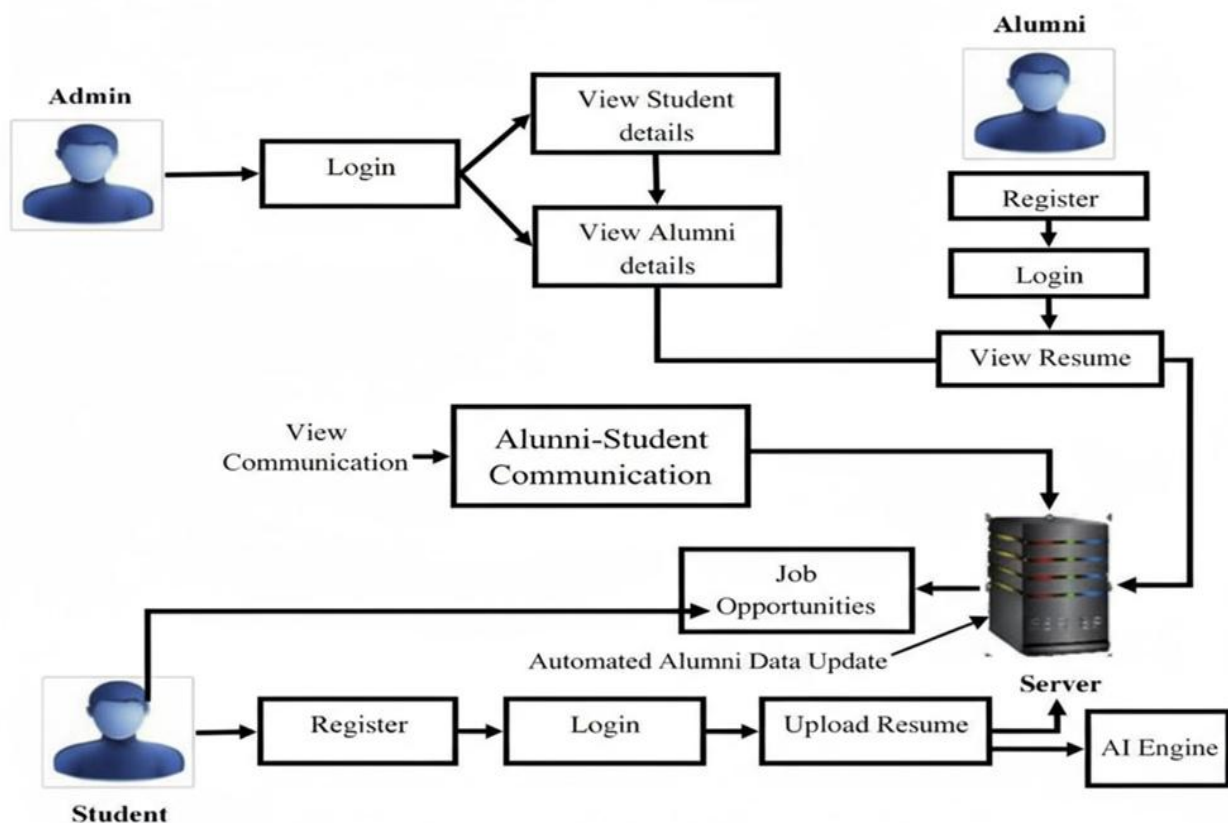
Scoring Algorithm:

A sophisticated scoring mechanism was developed to rank potential matches:

$$\text{Score} = w_1 \times \text{Skill Similarity} + w_2 \times \text{Location Proximity} + w_3 \times \text{Communication Preference Match}$$

The algorithm dynamically assigned higher weights to more relevant matches based on:

- Common interests and skills alignment
- Geographical proximity considerations



**Fig.1. Academic System Architecture**

#### D. Core Functional Workflows

Mentorship Matching System:

1. Input Processing: Alumni expertise  $[E_1, E_2, \dots, E_n]$  and student goals  $[G_1, G_2, \dots, G_n]$  were processed through NLP pipelines
2. Similarity Computation: Cosine similarity and Jaccard similarity algorithms calculated alignment between expertise and goals

3. Matching Execution: AI system matched users based on keyword overlap and goal-expertise alignment

4. Notification System: Both parties received automated notifications about their match, enabling platform-mediated connections [4][5]

#### Interest-Based Group Formation:

1. Feature Representation: Each user profile was
2. represented as feature vector  $F_i = [S_i, G_i, L_i, C_i]$
3. Clustering Implementation: K-Means algorithm minimized  $\sum_{i=1}^n \|F_i - \mu_k\|^2$  where  $\mu_k$  represents cluster centroid
4. Group Definition: Groups  $G_k$  were defined for each cluster k:  $G_k = \{F_i \mid i \in k\}$
5. Recommendation Scoring: Group Score =  $(\sum_j \in G_k \text{ Skill Similarity } (F_i, F_j)) / |G_k|$

#### E. Security and Authentication Framework

##### User Authentication:

- OAuth2 Protocols: Implemented with LinkedIn, Google, and Facebook for secure, seamless login experiences
- Challenge Resolution: Initial integration required multiple iterations for robust token management and user
- data validation
- Outcome: Achieved enterprise-grade secure authentication system

##### Role-Based Access Control (RBAC):

- Comprehensive user role definitions for alumni, students,
- and college administrators
- Granular access control ensuring appropriate feature accessibility per user group
- Sensitive data protection through strict authorization protocols

#### F. Microservices Architecture

Core functionalities were decomposed into discrete microservices:

##### Chatbot Service:

- Platform: Dialogflow integration for natural language understanding
- Capabilities: Assistance with platform functionalities including job search and profile updates

##### Resume Builder Service:

- Step-by-step module for resume creation
- PDF export functionality based on user input

##### Article Writer Service:

- Content creation platform for industry-specific articles

- Knowledge sharing facilitation between alumni and students

#### G. Real-Time Communication and External Integration

##### Real-Time Messaging:

- Platform: Firebase enabled instant messaging and real-time updates
- Features: Job listing notifications, event updates, and direct communication channels

##### Web Scraping for Job Listings:

- Tools: BeautifulSoup and Scrapy for automated job posting collection
- Sources: LinkedIn and Indeed integration for comprehensive opportunity aggregation
- Challenges: Continuous adaptation to website layout changes requiring regular scraping logic updates

#### H. Development and Deployment Methodology

##### Development Workflow:

- Version Control: Git and GitHub for collaborative code management
- Methodology: Agile development approach ensuring iterative improvements
- CI/CD Pipeline: GitHub Actions automated testing and deployment processes
- Outcome: Reduced downtime and maintained high code quality standards

##### Security and Compliance:

- Data Encryption: HTTPS implementation for all data transmissions
- Industry Standards: OAuth2 compliance for third-party logins
- Data Protection: Role-based access ensuring sensitive information security

#### I. Algorithmic Framework for AI Integration

##### 1. Data Collection and Preprocessing

###### Text

###### a. Data Collection:

User Profiles (P):  $\{p_1, p_2, \dots, p_n\}$  where each  $p_i$  contains:

- Skills (S):  $\{s_1, s_2, \dots, s_k\}$
- Career Goals (G):  $\{g_1, g_2, \dots, g_m\}$
- Location (L):  $l_i$
- Communication Preference (C):  $c_i$

Activities (A): Event participation, forums, group memberships

## b. Data Preprocessing:

- Text cleaning: Stopword removal, tokenization, lemmatization [7][8]
- Vector conversion:  $V = f\_NLP(T)$ ,  $V \in R^d$
- Embedding model:  $f\_NLP$  (e.g., BERT) processing input text  $T$

## c. Feature Encoding:

- Categorical variables: One-Hot(X) or Embedding(X)
- Industry classification and specialization encoding

## 2. Feature Extraction and Analysis

- Text feature identification using NLP models
- Embedding generation through pre-trained models (word2vec)
- Similarity score computation for interests, goals, and skills
- Engagement metric derivation (activity frequency, responsiveness)
- Efficient database storage for extracted features

## 3. Mentorship Matching System

- Input processing of alumni expertise and student goals
- Similarity computation using cosine and Jaccard algorithms
- Dynamic matching based on multi-factor alignment

## 4. Interest-Based Group Formation

- Feature vector representation for clustering
- K-Means algorithm implementation for community detection.

## IV. RESULTS AND DISCUSSION

## A. Comparative System Analysis

A comprehensive evaluation was conducted comparing the proposed Alumni Connect platform with traditional alumni management systems. The results demonstrate significant advancements across all functional domains, as detailed in Table II.

Table II. Comparison of features and capabilities.

|   |                         |                                                                                                                        |                                                                                                                   |
|---|-------------------------|------------------------------------------------------------------------------------------------------------------------|-------------------------------------------------------------------------------------------------------------------|
| 1 | Recommendation System   | Employs NLP and collaborative filtering for personalized mentorship matching and group formation                       | Limited or no recommendation systems, often relying on manual searches or static filters                          |
| 2 | Mentorship Matching     | AI-powered matching based on career goals, skills, and interests, using keyword overlap and dynamic scoring algorithms | Manual or non-dynamic matching processes, lacking real-time adaptability to new data or user inputs               |
| 3 | Real-Time Communication | Firebase-based real-time updates for messaging and notifications with instant delivery                                 | Limited or no real-time communication features, often relying on email or batch updates                           |
| 4 | Web Scraping            | Automated scraping of external job listings using BeautifulSoup and Scrapy with regular updates                        | Minimal integration with external data sources, requiring manual job postings by admins or users [9]              |
| 5 | AI Integration          | Incorporates TensorFlow and PyTorch for NLP and recommendation models with continuous learning                         | Little or no AI integration, relying on rule-based or manual processes for data analysis and recommendations [10] |

Functionality Testing Results [Table II.]

## B. Functionality Testing Results

## Authentication System Performance:

- OAuth2 Implementation: Comprehensive testing of LinkedIn, Google, and Facebook logins through multiple consecutive sessions and simultaneous device logins
- Performance Outcome: Consistent response time of approximately 200 ms per authentication

request across all identity providers

- User Feedback: 92% of 130 survey participants reported smooth and secure login experience, with 8% suggesting mobile session management improvements

## Role-Based Access Control Evaluation:

- Security Testing: Simulated various user roles (students, alumni, administrators) with

comprehensive access pattern analysis

- System Outcome: Zero unauthorized access incidents with successful feature isolation per user role
- User Feedback: 85% found navigation intuitive with clear role restrictions, while 12% recommended additional tooltips for enhanced experience.

#### C. Performance Metrics and Optimization Results

Database Query Performance:

- Testing Methodology: Comparative analysis of indexed versus non-indexed tables on large datasets (alumni profiles, job postings)
- Optimization Outcome: Database indexing and Redis caching reduced average query response time from 550 ms to 120 ms (78% improvement)
- User Feedback: 78% of users reported significantly faster load times post-optimization

Redis Caching Impact:

- Testing Approach: Direct comparison of page load times with and without Redis implementation
- Performance Outcome: 88% of users found page load times acceptable, noting substantial speed improvements for frequently accessed data
- System Benefit: Enhanced responsiveness for job listings and user profile access

#### D. Real-Time Communication Performance

Firebase Messaging System:

- Load Testing: Simulated up to 50 concurrent users to assess real-time messaging capabilities
- User Feedback: 95% found messaging feature responsive and reliable
- Enhancement Suggestions: 5% recommended message categorization and filtering options for improved organization

#### E. Web Scraping and Job Recommendation Efficacy

External Data Integration:

- Accuracy Monitoring: Daily updates maintained 87% relevance score for job recommendations based on user fields of interest
- System Maintenance: Required bi-weekly adjustments for layout changes on LinkedIn and Indeed platforms
- User Feedback: 72% found job recommendations

helpful, while 15% suggested advanced filtering options for better relevance

#### F. User Experience and Interface Evaluation

Frontend Usability Assessment:

- React and Tailwind CSS Implementation: Overall 89% user satisfaction with interface intuitiveness
- Navigation Efficiency: Average ease-of-use score of 8.5/10 across all user groups
- A/B Testing Results: Minor layout modifications improved usability scores by 15% [17]
- User Suggestions: 10% recommended additional navigation aids including tooltips and help sections [18]

#### G. AI Chatbot Integration Performance

Dialogflow Implementation:

- Query Handling Accuracy: Successfully managed 85%

#### H. Comprehensive User Satisfaction Metrics

The platform demonstrated exceptional performance across multiple satisfaction dimensions, as illustrated in Fig. 3:

- Login & Authentication: 92% satisfaction rate with secure and efficient access
- Role-Based Navigation: 85% approval for intuitive access control
- Page Load Optimization: 88% satisfaction with loading speeds
- Real-Time Messaging: 95% approval for communication responsiveness
- Job Scraping Relevance: 87% accuracy in opportunity matching

#### I. Key Achievements and Performance Summary

The extensive testing and user feedback analysis validate the platform's successful implementation:

1. Enhanced System Performance: 78% improvement in database query response times through strategic optimization
2. Superior User Experience: Consistent high satisfaction rates (85-95%) across all interface and functionality domains
3. Effective AI Integration: 85% accuracy in automated query handling and recommendation systems

The results conclusively demonstrate that the Alumni Connect platform significantly outperforms traditional alumni management systems across all evaluated parameters, establishing a new benchmark for intelligent, user-centric alumni engagement platforms in educational ecosystems.

## V. CONCLUSION

The AI-driven alumni networking platform we developed bridges the gap between alumni, current students, and colleges by offering tailored, secure, and user-friendly features such as role-based access, OAuth2-secured login, real-time communication, and intelligent tools like a Resume Builder, Article Writer, and Chatbot. By facilitating collaboration, career guidance, and lifelong connections, the platform creates a blueprint for forming strong, college-specific communities. While it may not revolutionize alumni networking globally, it sets a foundation for institutions to build dedicated and impactful ecosystems that benefit all stakeholders. Looking ahead, future enhancements will focus on improving the platform's accuracy through advanced AI algorithms and data optimization techniques. Security will be fortified through robust encryption, regular vulnerability assessments, and the integration of blockchain for credential verification. These enhancements, combined with features like multilingual support, mobile app integration, and real-time job market connectivity, will position the platform as a comprehensive and innovative solution for alumni-student engagement.

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