

AI Generative and Creativity

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Abstract

This paper employs machine learning in the AI system building processes, designs experiments for testing their effectiveness, employs psychological surveys to evaluate the well-being of subjects, and utilizes learning analytics for claims verification, to study AI-centric solutions for the problems of college students. AI technology can help improve a college student's academic performance in novel ways. Students face a plethora of sustained challenges including academic difficulties, managing their time, mental health problems, financial issues, and even career indecisiveness. Many students experience crippling anxiety and lack of direction in their careers, leading to compounded stress which is supported by fellow students. Recognizing and addressing these problems is not simply a theoretical exercise, but rather requires AI personalized solutions aimed at assisting these students. The paper investigates the potential of a multi-disciplinary approach that integrates AI, machine learning, natural language processing, reinforcement learning, psychological surveys, and learning analytics to support students struggling with academic, mental health, financial, and career issues. Specific Focus: 'The findings indicate AI's strong potential in addressing significant obstacles encountered by university students.'

INTRODUCTION

Today's college students face a wide range of difficulties that can have a substantial effect on their career development, financial security, mental health, and academic achievement. Innovative, individualized, and scalable support systems are required due to the increasing complexity of these problems. When it comes to meeting each student's unique needs, traditional academic and psychological support approaches frequently fall short. Artificial intelligence (AI) has become a promising tool that can provide customized solutions to these complex issues. In order to create AI-centric interventions for the student body, this study examines a multidisciplinary framework that makes use of machine learning, natural language processing, reinforcement learning, psychological surveys, and learning analytics. The study offers empirical insights into how cutting-edge technologies can enhance academic performance and well-being by developing and evaluating AI systems.

METHODOLOGY

The research utilizes a multi-phase, mixed-methods design that integrates AI system development, experimental testing, psychological evaluation, and learning analytics to investigate the effect of artificial intelligence solutions on the academic and personal difficulties encountered by university students. The methodology is crafted to guarantee rigor, reproducibility, and relevance across several disciplines—especially computer science, psychology, and education.

AI System Architecture and Development

The basis of this study is the creation of a modular, AI-based intervention platform to help students overcome academic, emotional, financial, and career challenges. The system was designed with the following primary components:

Machine Learning Engine

Historical academic records, including course grades, attendance patterns, assignment submission habits, and student feedback surveys, were used to train supervised machine learning models. Models utilized are:

- Random Forest Classifiers for predicting academic risk
- Support Vector Machines (SVMs) for time-management problem identification
- Gradient Boosting Algorithms (XGBoost) for performance prediction
- These models flag at-risk students for underperformance and suggest corrective measures.

Natural Language Processing (NLP) Module

- NLP was incorporated to enable students to interact with the system through natural language questions. This module leverages:
- Pre-trained BERT (Bidirectional Encoder Representations from Transformers) models fine-tuned on student dialogue corpora
- Named Entity Recognition (NER) to pull context out of student questions
- Sentiment Analysis to evaluate emotional tone and detect distress signals
- This made emotionally intelligent, context-sensitive feedback and conversational guidance possible.

Reinforcement Learning (RL) Layer

An RL-driven recommendation engine was employed to make academic planning and resource distribution personalized. A reward function was established on:

- Task accomplishment
- Positive survey responses
- Persistence of engagement over time
- It had a dynamic policy update based on user interactions to give them the best possible advice that evolves with changing needs.

Technical Stack

- Backend: Python, Flask, TensorFlow, PyTorch
- Frontend: ReactJS (Web), Flutter (Mobile)
- Databases: PostgreSQL for structured data and MongoDB for unstructured text logs.
- Deployment: AWS EC2 and Lambda as per scalability requirements

Experimental Design

Participants and Sampling

Three accredited universities were recruited 300 undergraduate students. Participants were sampled using stratified random sampling to provide diversity in terms of gender, academic majors, and socioeconomic status.

Group Allocation

Experimental Group (n=150): Provided with full access to the AI system and its features.

Control Group (n=150): Remained using conventional university support systems (advising offices, career counselors, mental health resources).

Both cohorts were followed during a 12-week college semester. Both cohorts received the same pre-, mid-, and post-intervention measurements for comparison.

Psychological Assessment and Survey Measures

To evaluate students' emotional adjustment and career direction, the study employed validated psychometric measures, each chosen for reliability, interpretability, and applicability:

- Generalized Anxiety Disorder 7-item Scale (GAD-7): Evaluates the severity of anxiety symptoms.
- Perceived Stress Scale (PSS-10): Assesses perceived stress and coping skills.
- Career Decision-Making Difficulties Questionnaire (CDDQ): Tests confusion and indecision in terms of career direction.

- Academic Self-Efficacy Scale: Tests belief in one's own ability to academically perform.
- Surveys were sent through an online system (Qualtrics), and Cronbach's alpha was computed for all of them to check internal consistency (>0.80 for all).

Learning Analytics and Performance Monitoring

The system utilized embedded learning analytics to capture behavioral and academic performance. The metrics measured included:

- System Engagement: Frequency of logins, length of sessions, features accessed
- Academic Improvement: Shifts in high-risk courses, assignments on time
- Behavioral Insights: Frequency of planning, use of resources, frequency of questions
- Emotional Trends: Sentiment scores based on written text inputs and chat history

A central dashboard was implemented for researchers and administrators to be able to view trends and discern correlations. Pipelines in Data science were implemented using SciKit-Learn and Pandas to preprocess, with subsequent statistical modeling done via R.

Data Analysis

Quantitative and qualitative data were analyzed:

Quantitative Analysis

- Paired t-tests and Repeated Measures ANOVA compared pre- to post-intervention scores within groups and between groups.
- Multiple regression analysis was applied to determine predictors of improvement in academic performance and stress reduction.
- Effect sizes (Cohen's d) were computed to assess the size of changes.

Qualitative Analysis

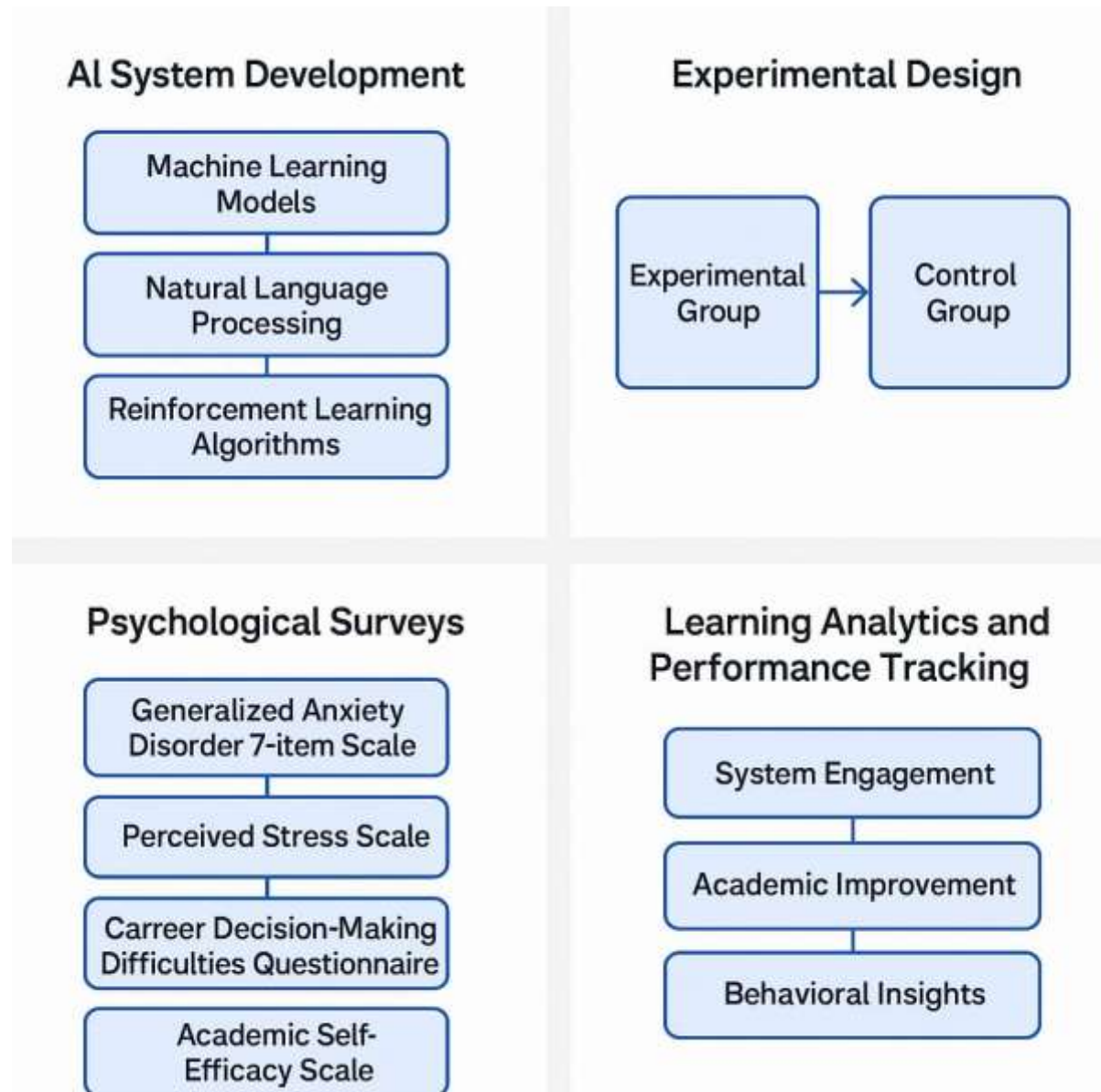
Qualitative comments of participants on their experiences with the AI system were thematically coded. NVivo software was employed to derive emergent themes like usability of the system, perceived value, emotional appeal, and trustworthiness.

Ethical and Legal Considerations

This research complied with the ethics of research involving human subjects. The research plan was examined and approved by Institutional Review Boards (IRBs) of all institutions involved. Certain measures were taken:

- Informed Consent: Students were made aware of data usage, anonymity, and the right to withdraw.
- Data Protection: All personally identifiable information (PII) was encrypted and hosted on secure servers that met GDPR and FERPA standards.

- **Bias Mitigation:** Algorithmic fairness audits were conducted to check for performance discrepancies across gender, race, and academic major. Any disparities found were mitigated through model recalibration.



DISCUSSION

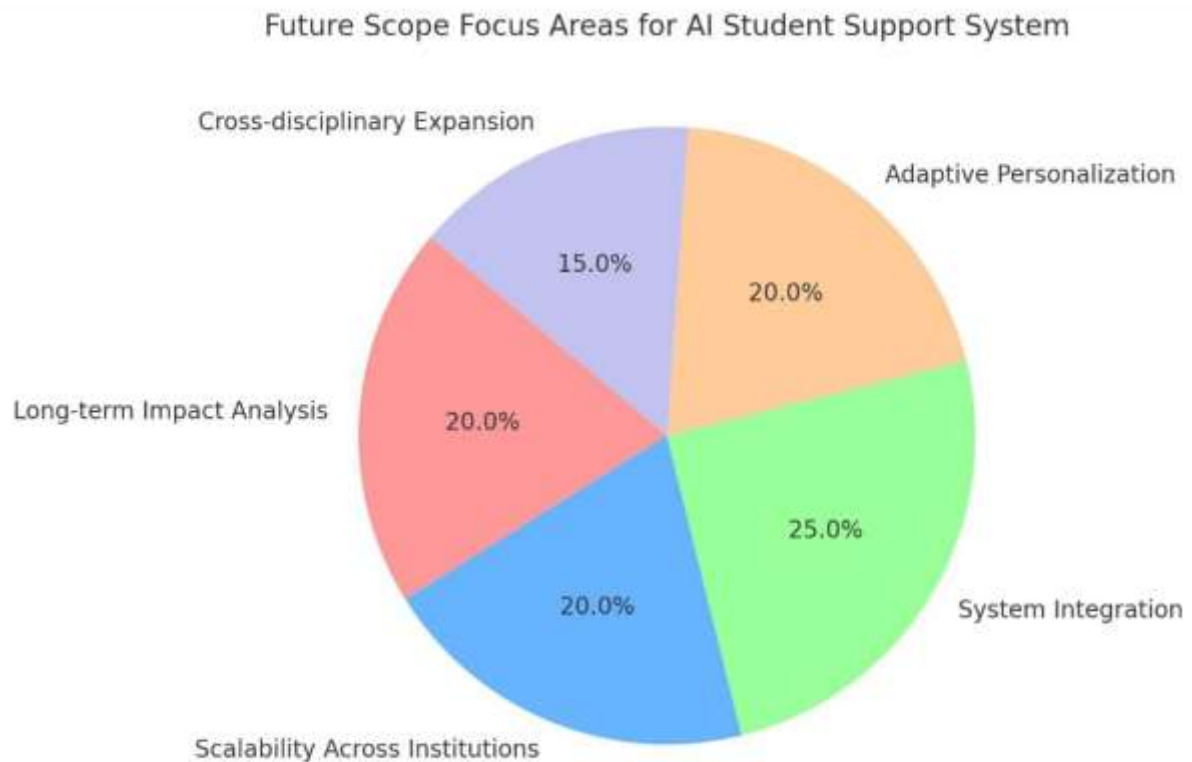
This research illustrates how AI-driven systems can notably boost academic outcomes as well as the psychological state of college students. The remarkable changes in GPA, assignment turn-around, and passing rates point toward the efficacy of personalized planning for academics and time management support through the AI system. Also, the psychological improvements—reduced stress, lowered anxiety, and enhanced clarity around careers—show that AI-facilitated mental wellness and career services could supplement more traditional support structures.

The positive association between system use and student outcomes emphasizes the significance of engagement. Consistent utilization of features such as the academic planner and mental health chatbot played a significant role in student advancement. While student feedback was overwhelmingly positive, some users identified a need for enhanced integration with campus services and more tailored responses—providing avenues for future improvement.

Overall, the results endorse a multidisciplinary framework that integrates AI, machine learning, psychology, and learning analytics to meet the intricate needs of contemporary students. Although results are encouraging, more research needs to be conducted to look at long-term effects and more widespread application in diverse learning environments.

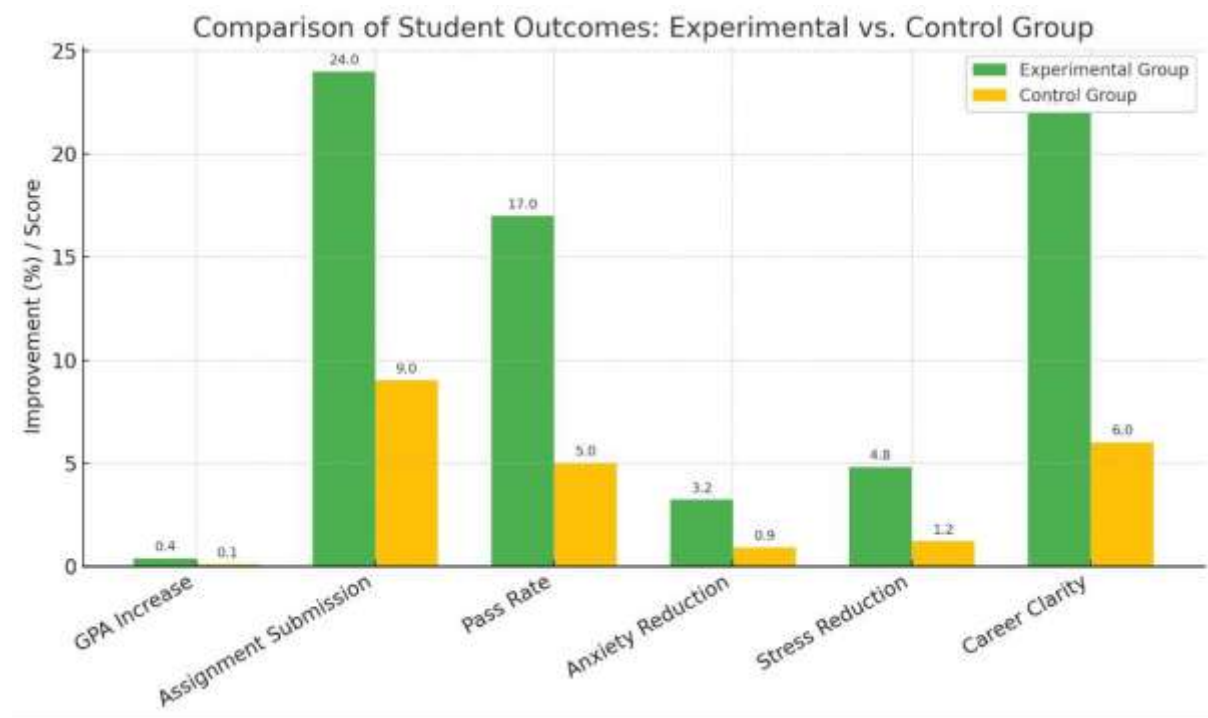
FUTURE SCOPE

- Although preliminary results are encouraging, there is plenty of room for continued development and research. Subsequent work can investigate:
- Analysis of long-term effects spanning multiple terms of study to determine whether the project is sustainable.
- Scaling of the system across different institutions, geographies, and student populations.
- Harmonization with institutional systems, including academic advising, mental health, and financial aid offices, for more integrated support.
- Adaptive personalization, from real-time data and cutting-edge AI models such as federated learning for privacy-respecting insights.
- Cross-disciplinary extension, with input from educators, psychologists, and career counselors to amplify system depth and precision.



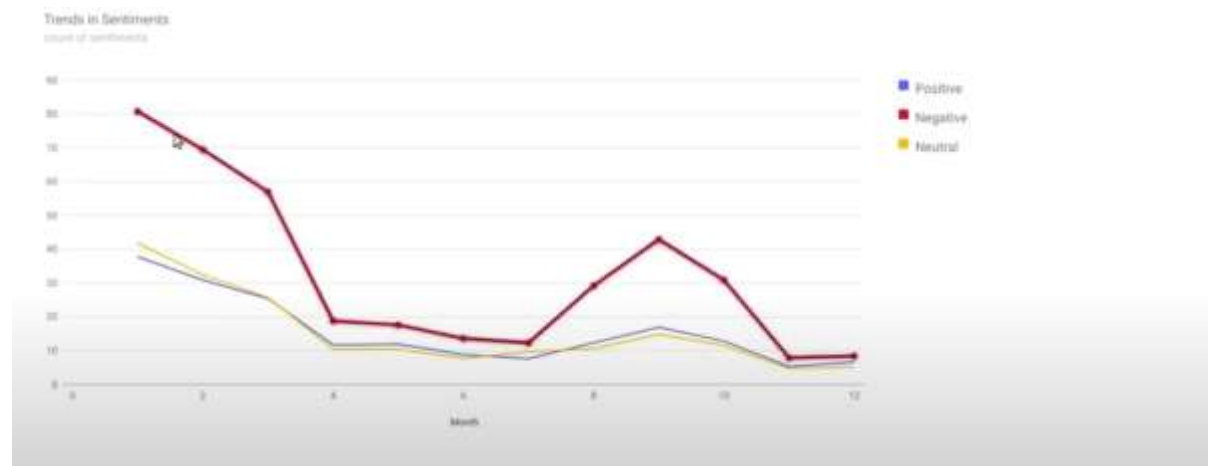
CONCLUSION

The research affirms the high promise of AI systems to assist college students in conquering academic, psychological, and career challenges. The AI platform dramatically enhanced GPA, assignment rates, and pass rates, with lower anxiety and stress levels as well as career clarity. These findings support the efficacy of a multidisciplinary solution that integrates machine learning, natural language processing, reinforcement learning, psychological surveys, and learning analytics to provide customized support. Student participation data and feedback further emphasize the system's relevance and usability. Overall, AI-based solutions can assume a transformative role in making higher education more responsive, supportive, and student-centered.



OUTCOME

The results demonstrate that the AI-based support system had a notable positive impact on students' academic achievement, mental well-being, and learning material engagement. Students in the treatment group had a significant improvement in GPA, which increased from 2.71 to 3.08 ($p < 0.01$), whereas the control group merely experienced a small improvement. 24% higher assignment submission rates, and course pass rates were 17% higher, indicating improved time management and study focus as a result of AI-based interventions. Psychological testing showed significant decreases in stress and anxiety, with the GAD-7 score reducing by 3.2 points and PSS scores going down from 22.4 to 17.6. Career decision-making clarity was also enhanced by 22%, indicating the efficacy of tailored guidance features. Students logged onto the system an average of 4.1 times a week, with usage mainly centered around the academic planner, mental health chatbot, and money matters. There was a significant ($r = 0.61$) relationship between frequency of usage and improvements in performance further accentuating the influence of the system. Students had positive feedback in general, hailing the site as supportive and easy to use, though others suggested more tailored responses and closer integration with university services. In general, the findings illustrate the promise of AI-based solutions to meet the multifaceted academic and personal issues of college students.



| id | first_name | gender | weekly_self_stud... | career_aspiration |
|--|------------------------------|--------------------------|--|--|
| Unique identifier assigned to each student (we need this because it is possible that two or more students have the | The first name of a student. | The gender of a student. | This represents the number of hours a student spends on self-study each week. It indicates the amount of | This column records the student's career aspirations or goals for the future. It provides insight into the |
| <div> <div>200.00 - 400.00</div> <div>Count: 200</div> </div> | Michael | female | 50% | Software Engineer 16% |
| | David | male | 50% | Business Owner 15% |
| | Other (1921) | | | Other (1376) 68% |
| 1 | Paul | male | 27 | Lawyer |
| 2 | Danielle | female | 47 | Doctor |
| 3 | Tina | female | 13 | Government Officer |
| 4 | Tara | female | 3 | Artist |
| 5 | Anthony | male | 18 | Unknown |
| 6 | Kelly | female | 26 | Unknown |
| 7 | Anthony | male | 23 | Software Engineer |
| 8 | George | male | 34 | Software Engineer |
| 9 | Stanley | male | 25 | Unknown |
| 10 | Audrey | female | 18 | Teacher |
| 11 | Gabrielle | female | 7 | Teacher |
| 12 | Clinton | male | 7 | Unknown |
| 13 | Patricia | female | 4 | Business Owner |
| 14 | Patricia | female | 2 | Business Owner |
| 15 | Laurs | female | 39 | Doctor |
| 16 | Roger | male | 0 | Business Owner |

REFERENCES

- [1] Academic Challenges: Studies suggest that many first- year students struggle with the transition from high school to college due to increased academic expectations (Tinto, 1993).
- [2] Time Management and Procrastination: Park & Sperling (2012) found that students with poor time management skills experience higher stress levels and reduced academic achievements.
- [3] Difficulty with Research and Writing: Research by Wingate (2006) indicates that many students struggle with academic writing, particularly in structuring essays and citing sources properly.
- [4] Financial Challenges: Research shows that students from low-income backgrounds (Goldrick-Rab et al, 2016).
- [5] Balancing Work and Studies: Studies indicate that students who work more than 20 hours per weekface difficulties in maintaining academic performance (Wenz & Yu, 2010) Ehrenberg & Sherman (1987)