

# Academic Performance Analysis: Visualization and Insights using Data Science

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**Abstract** - This project aims to create an interactive platform for visualizing and analyzing academic performance data. The tool processes student grades results, presenting them in visual formats like bar charts, line graphs, and pie charts. It integrates machine learning algorithms to categorize students as top or low performers and predicts strengths and weaknesses in various subjects. The system offers personalized improvement recommendations. This platform enhances academic analytics by transforming data into actionable insights for better educational outcomes.

**Key Words:** Academic Performance, Machine Learning, Data Visualization, Predictive Analytics, Education Technology, Performance Categorization, Interactive Platform.

## 1. INTRODUCTION

The academic performance of students is a crucial metric that reflects the effectiveness of learning processes and serves as a strong determinant of future academic and professional success. Educational institutions have long relied on traditional evaluation systems—primarily numerical scores and grade point averages—to assess student performance. While these metrics provide a basic indication of academic standing, they often lack the depth needed to uncover individual learning patterns, identify specific strengths and weaknesses, and inform targeted educational strategies. Such static methods fail to support personalized learning or real-time academic interventions.

In today's digital era, the availability of vast amounts of educational data, combined with the power of data science and machine learning technologies, opens new possibilities for transforming how academic performance is analyzed and understood. By applying these modern analytical techniques, institutions can move beyond surface-level evaluation and generate meaningful,

actionable insights that drive smarter, student-centered decision-making.

This paper introduces an intelligent academic performance analysis tool that leverages data science to bridge the gap between raw data and informed educational action. The platform collects, processes, and visualizes student performance data through a user-friendly, web-based interface. Using linear regression, the system predicts subject-wise academic outcomes and categorizes students as top performers, average performers, or those at risk of falling behind. The integrated visualization module employs interactive graphs such as bar charts, line graphs, and pie charts to clearly depict trends and comparisons, making the data easily interpretable for staff and educators.

A real-time data filtering system further empowers staff to explore performance datasets based on specific criteria like subjects, grades, or student identifiers. By combining machine learning, predictive analytics, and dynamic visualization, this platform offers a comprehensive solution for enhancing academic monitoring and enabling data-driven improvements in the education system.

## 2. LITERATURE REVIEW

[1] Essa Alhazmi and Abdullah M. Sheneamer, "Early Predicting of Students Performance in Higher Education," *IEEE Access*, Vol 11, 2023.

This paper uses clustering and classification techniques to identify the impact of student performance at an early stage on GPA, suggesting that educational systems can mitigate the risks of student failures early. The study showcases tests conducted on various machine learning models that forecast student achievement in the initial phases using diverse attributes.

[2] Mfowabo Maphosa, Wesley Doorsamy, Babu S. Paul, "Student Performance Patterns in Engineering

**at the University of Johannesburg: An Exploratory Data Analysis," IEEE, 17 May 2023.**

This study explores student performance patterns in engineering disciplines at the University of Johannesburg through an exploratory data analysis approach. Utilizing a comprehensive dataset, the research identifies key factors influencing academic success and retention among engineering students. The findings reveal trends and correlations that can inform academic interventions and support strategies. The analysis emphasizes the importance of predictive modeling in understanding student behavior and performance, ultimately aiming to enhance educational outcomes and reduce dropout rates.

**[3] Harshavardhini S S and R Latha, "Academic Performance Analysis Based on Online Learning Data," International Journal of Advanced Research in Science, Communication and Technology (IJARSCT), Volume 4, Issue 2, July 2024.**

This research focuses on analyzing academic performance using online learning data to develop a predictive model for student success. By employing various data mining techniques, the study aims to identify patterns and factors that significantly impact students' academic achievements in online environments.

The results indicate that leveraging online learning analytics can provide valuable insights for educators to tailor instructional methods and improve student engagement and performance.

**[4] Malak Abdullah, Mahmoud Al-Ayyoub, Farah Shatnawi, Saif Rawashdeh, Rob Abbott, "Predicting students' academic performance using e-learning logs," IAES International Journal of Artificial Intelligence (IJ-AI), 2023.**

This paper investigates the use of e-learning logs to predict students' academic performance in higher education settings. By analyzing interaction data from an e-learning platform, the study applies machine learning algorithms to forecast student outcomes based on their engagement patterns.

The findings highlight the potential of utilizing e-learning analytics to identify at-risk students early and implement timely interventions to enhance academic success.

### 3. EXISTING SYSTEM

**Manual Data Entry:** Existing systems require manual input of grades, leading to errors and inconsistencies, and can be time-consuming when integrating data from multiple sources.

**Limited Visualization:** Traditional systems offer basic tables or charts, which are insufficient for understanding performance trends and identifying students needing support.

**Lack of Predictive Capabilities:** Existing systems do not use machine learning to predict performance or identify at-risk students, often addressing issues too late.

**No Personalization:** Most systems provide aggregate data without personalized feedback, making it difficult to offer specific guidance to individual students.

**Limited Interactivity:** Existing systems lack interactive features, limiting users' ability to explore and understand student performance in depth.

### 4. PROPOSED SYSTEM

Key features of the proposed system include:

1)Data Collection: The system will collect academic data from multiple sources, such as CSV files, databases, and user input. This flexibility will allow it to integrate data from various educational institutions, ensuring broad usability.

2)Dynamic Visualization: The platform will present academic data in easy-to-understand formats, such as different kind of charts ,for example: bar charts, pie charts, and line graphs. It will include customizable filters, allowing users to drill down into specific subjects, students, or academic periods.

3)Machine Learning Integration: The system will incorporate machine learning algorithms, such as clustering, classification, and predictive analytics, to:

Categorize students into top performers, average performers, and low performers based on historical performance data.

Predict student performance using data-driven insights, enabling stakeholders to identify strengths and weaknesses in individual students or subjects.

4) Personalized Feedback: The system will offer personalized feedback based on student performance. It will suggest areas for improvement and generate recommendations tailored to individual needs.

5) Interactive Features: Users will have the ability to explore trends over time, compare academic performance between different subjects, and track performance improvements or declines.

6) User-Friendly Interface: The platform will have an intuitive interface, ensuring ease of use for all stakeholders, including students, educators, and administrators. It will provide features such as report generation, interactive charts, and detailed performance breakdowns.

By offering a data-driven, visually-rich, and interactive system, the proposed tool aims to improve academic decision-making, empower students to take control of their educational journeys, and support educators in identifying areas for intervention.

## 5. SYSTEM OVERVIEW

The academic performance analysis system is a streamlined, web-based solution that combines user management, real-time data visualization, machine learning prediction, and administrative tools within a single interactive platform. Designed with the needs of both educators and students in mind, the system offers secure login portals, intuitive dashboards, and intelligent performance insights through predictive analytics. Staff users can access detailed visualizations of student scores, filter data based on different criteria, and make evidence-based academic interventions. The linear regression model serves as a foundational machine learning component that forecasts academic outcomes, enabling early intervention for students who may need additional support. The use of Streamlit for interface development ensures fast rendering of charts and dynamic data updates, while MySQL handles robust data management in the background. By integrating data science techniques with educational data, the system converts static academic records into actionable insights, empowering institutions to enhance student success through timely, informed decision-making.

## 6. METHODOLOGIES

### 6.1 System Architecture

The system is designed following a modular and layered architecture that enhances maintainability, scalability, and separation of concerns. It consists of three primary layers: the presentation layer (frontend), the logic layer (backend processing), and the data layer (storage). The frontend is implemented using Streamlit, an open-source Python framework that enables rapid development of interactive web applications with minimal code. It manages user interaction and visualizes performance analytics in real-time. The backend processing layer encapsulates the core logic of the system, including user authentication, student data filtering, and machine learning models for predicting academic performance. This logic layer ensures secure routing between roles and modules while handling data transformation tasks. Finally, the data storage layer is built using a MySQL relational database, which stores structured information about users, academic grades, subject scores, and login credentials. The architecture adheres to the Model-View-Controller (MVC) principle in practice, enabling easy integration of future enhancements such as new analytics dashboards, expanded data models, or additional user roles. The modular nature allows developers to update or replace individual components—such as the ML model or visualization engine—without affecting the entire system.

### 6.2 User Authentication Module

The user authentication module ensures that only authorized users can access role-specific features of the platform. Developed using Python and tightly integrated with Streamlit, the module provides a simple and intuitive login form that captures a user ID and password. Upon submission, the credentials are validated against entries in the corresponding staff or students tables in the MySQL database. This role-based access control approach restricts access to sensitive data and ensures users only interact with features relevant to their role. For instance, staff users are redirected to a detailed analytics and visualization dashboard, whereas students are directed to their personalized homepage. In the backend, the system uses secure connection parameters to interact with the MySQL database, and errors such as incorrect login credentials are handled gracefully with informative feedback messages. While the current version uses plain-text passwords (suitable for demo or local environments),

the system is designed with a clear roadmap for implementing hashed password storage using libraries such as bcrypt and a secure password reset mechanism. This module plays a crucial role in protecting the integrity and confidentiality of academic data and supports a clean user experience through visual cues and conditional navigation logic.

### 6.3 Data Collection and Storage

The backbone of the system's data management lies in a structured MySQL relational database that stores all relevant academic and user information. The database schema is designed to be extensible and efficient, starting with two main tables: students and staff, which include attributes such as userid, password, name, and identification numbers like roll number and registration number. Additional tables such as grades and subjects can be incorporated to track marks across different semesters or assessment types. Data is fetched using structured SQL queries, which are dynamically executed based on the user's role and input selections. This ensures real-time responsiveness and personalized output. By following normalization principles, the database reduces redundancy, maintains referential integrity, and improves query performance. The system is also designed to support data import/export operations, allowing administrators to upload or archive grade records periodically. With this structure, the platform can easily scale to accommodate thousands of student entries and extend to multiple departments or courses without major modifications to the core database logic.

### 6.4 Visualization Module

The visualization module is central to translating complex academic data into actionable insights. This component is developed using Python's data visualization libraries such as Plotly, Matplotlib, and Streamlit's built-in chart components, offering both interactivity and clarity. After successful authentication, staff users are presented with a range of dynamic charts, including line charts for comparing subject-wise performance, bar graphs for tracking student progress over time, and pie charts for depicting grade distribution among student cohorts. These visualizations are populated using data retrieved in real-time from the CSV or database and formatted into easily digestible summaries. Users can interact with the charts through dropdown filters, sliders, and other UI elements to analyze specific subsets of data—such as performance trends in a particular subject or top-performing students across terms. This module enhances the decision-making

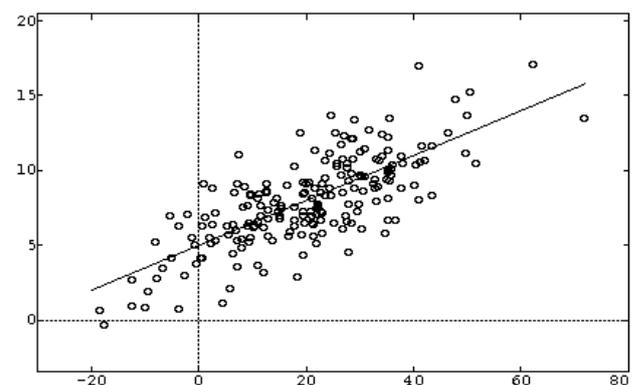
process for educators, enabling them to quickly identify patterns, detect outliers, and intervene where necessary. In the future, this module can be extended with export features to generate printable reports or downloadable data visualizations.

### 6.5 Machine Learning for Performance Prediction

To provide meaningful and actionable insights into academic performance trends, a regression-based predictive approach has been integrated into the system. This approach specifically utilizes linear regression, a supervised machine learning algorithm, to model the relationship between a student's past academic performance and their expected future outcomes in individual subjects. The regression model is trained on historical grade data, taking into account variables such as prior subject scores, average performance across terms, or cumulative academic metrics. By analyzing these patterns, the model is capable of forecasting potential scores for upcoming evaluations or semester-end results. This predictive capability is particularly valuable for early identification of students at academic risk, enabling staff to intervene proactively before performance issues escalate.

#### Linear Regression Formula

The general formula for **simple linear regression** is:



$$Y_i = \beta_0 + \beta_1 X_i + \epsilon_i$$

Labels for the equation components:

- Dependent Variable:  $Y_i$
- Population Y intercept:  $\beta_0$
- Population Slope Coefficient:  $\beta_1$
- Independent Variable:  $X_i$
- Random Error term:  $\epsilon_i$

Groupings:

- Linear component:  $\beta_0 + \beta_1 X_i$
- Random Error component:  $\epsilon_i$

### 6.6 Security and Usability Features

Security and user experience are both essential to the design of the system. Security is enforced through role-based access control and secure backend query handling, ensuring that users cannot access unauthorized content or perform restricted actions. The system uses Streamlit’s session state to persist user sessions securely without exposing sensitive information. Though currently implemented with plaintext passwords for simplicity in local development, the structure is ready for integration of secure password handling techniques including hashing, salting, and encrypted storage. In terms of usability, the platform offers a clean, minimalist interface that adjusts based on user roles, presenting only the tools relevant to each user type. Navigation is made intuitive through clearly labeled buttons, login feedback, and conditional content display. The responsive design allows users to interact fluidly with visualizations and forms. Planned usability upgrades include mobile compatibility, multilingual support, and accessibility improvements, making the system more inclusive and adaptable across various use cases and devices.

### 7. RESULTS & DISCUSSIONS

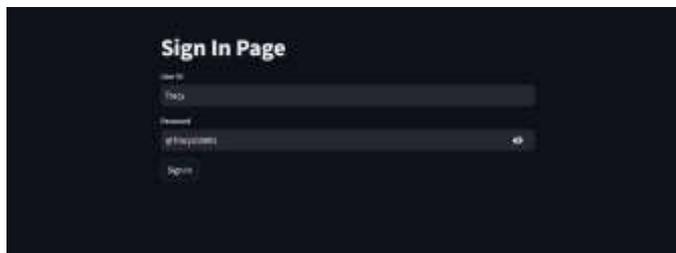


Fig 7.1 Login

The system includes a secure and role-based sign-in page that enables both students and staff to log into the platform using their unique User ID and password. This authentication feature is implemented using Python and Streamlit, and it is directly connected to a MySQL database that stores user credentials in dedicated students and staff tables. When users enter their login details, the system verifies the credentials against the database and provides immediate feedback—granting access upon successful authentication or displaying error messages in case of invalid inputs. Upon signing in, staff members are redirected to an administrative dashboard that provides access to data visualization and analytics tools, while students are taken to a personalized interface tailored to their academic data. This structured sign-in mechanism not only ensures data privacy and controlled access but also delivers a smooth and intuitive user experience tailored to each user type.

Sl	Semester	Paper Code	Paper Name	Credit	Grade	Grade Point					
1	1	2005101L	FUNDAMENTALS	4	B+	7	5	9	10	10	10
2	1	2005102L	ENGINEERING	3	B+	7	10	8	7	10	10
3	1	2005103L	ENGINEERING	3	B+	7	8	10	8	5	5
4	1	2005104L	BASIC SCIENCE	1.5	A	8	10	8	9	9	9
5	1	2005105L	PROBLEMS	2	A	8	5	6	10	9	9
6	1	2005106L	PYTHON I	2	A	8	9	10	5	9	9
7	1	2005107L	INDUSTRIAL	1.5	A	8	9	8	10	8	8
8	1	2005108L	ENGLISH I	3	A+	8	8	7	8	8	8
9	2	2005201L	CONVULSION	1	A+	9	10	6	8	10	10
10	2	2005202L	C PROGRAM	2	O	10	10	6	7	6	6
11	2	2005203L	ADVANCED	2	B	6	8	7	10	10	10
12	2	2005204L	LAPLACE TR	4	B+	7	9	5	9	5	5
13	2	2005205L	PHYSICS I	3	B+	7	10	10	7	8	8
14	2	2005206L	ENGINEERING	3	A	8	9	9	10	5	5
15	2	2005207L	BASIC ELECT	3	B+	7	8	10	8	7	7
16	2	2005208L	PROGRAMS	2	A	8	5	5	5	8	8
17	3	2005301L	OBJECT OR	3	B+	7	6	10	10	8	8
18	3	2005302L	DISCRETE M	4	B	6	7	6	5	9	9
19	3	2005303L	DIGITAL PR	4	B+	7	7	9	5	10	10
20	3	2005304L	DATA STRU	3	B+	7	10	9	8	8	8
21	3	2005305L	UNIVERSAL	3	B+	7	9	8	8	5	5
22	3	2005306L	DIGITAL SPS	2	O	10	10	9	10	9	9
23	3	2005307L	DATA STRU	2	O	10	6	10	6	8	8

Fig 7.2 Data Collection

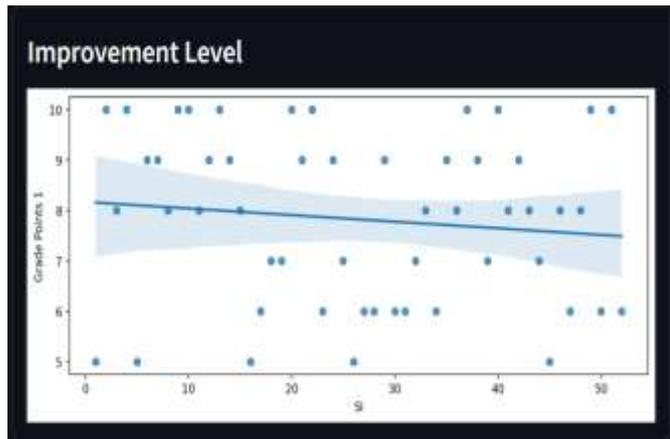
As part of the data collection process, a CSV file is utilized to store and organize student academic information in a structured and accessible format. This file serves as a primary source of data for the system, containing fields such as roll number, registration number, student name, and subject-wise grade points. The CSV format allows for easy integration with the backend system, enabling seamless data import into the MySQL database or direct use in data analysis and visualization workflows. A figure of the CSV file has been included to demonstrate the format and organization of the collected data, showcasing its clarity, uniformity, and readiness for processing by the system’s machine learning and visualization modules. This approach simplifies the initial data entry and verification process, supports batch uploading of academic records, and ensures consistent data collection across different sources or academic terms.



Fig 7.3 Data Visualization

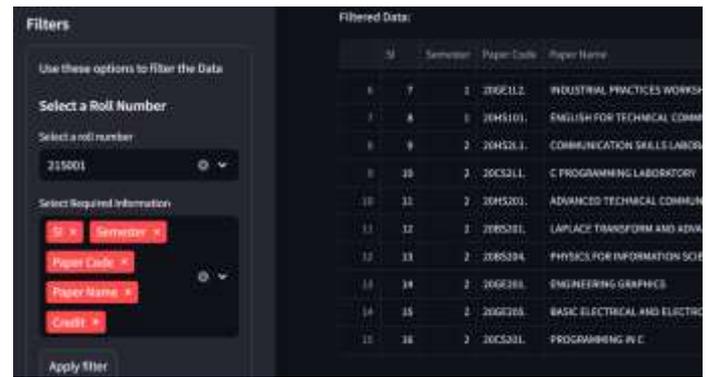
The system includes a data visualization line graph that effectively illustrates the trend of student academic performance over time or across multiple subjects. This graph is designed using Python's visualization libraries integrated within the Streamlit framework, allowing for dynamic and interactive representation of data. The line graph plots key performance indicators such as subject-wise marks or semester scores on the y-axis against

variables like subject names or assessment periods on the x-axis. By connecting data points with lines, it provides a clear and continuous view of how a student or group of students has performed over time, making it easier to identify patterns, improvements, or declines in academic achievement. This visualization is particularly valuable for educators and staff, as it enables quick interpretation of complex data and supports informed decision-making for interventions or academic planning.



**Fig 7.4 Machine Learning & Predictive Analytics**

The system incorporates Machine Learning and Predictive Analytics to enhance academic performance monitoring and provide actionable insights for early intervention. At its core, the platform utilizes a linear regression model to predict student performance based on historical academic data. By analyzing previous scores across subjects or semesters, the model forecasts potential future outcomes in specific subjects, helping educators identify students who may be at risk of underperforming. These predictions are visualized and made accessible through the staff dashboard, allowing faculty to take timely action such as recommending remedial sessions or personalized support. The integration of predictive analytics not only transforms raw grade data into forward-looking insights but also empowers institutions to shift from reactive to proactive academic strategies, ultimately supporting better student outcomes and academic success.



**Fig 7.5 Filter Module**

The system features an interactive data filtering interface that enables staff users to explore and analyze student academic data in real-time. This functionality is seamlessly integrated within the Streamlit dashboard and allows users to apply multiple filters based on parameters such as subject, performance range, roll number, or registration number. By selecting specific criteria, staff can instantly view refined subsets of data, making it easier to focus on underperforming students, top achievers, or class-wide trends. This real-time interaction not only enhances the usability of the platform but also supports informed decision-making by presenting relevant data on demand. The intuitive design of the filtering tool reduces the complexity of manual data analysis and transforms the system into a powerful academic monitoring and intervention tool.

## 8. CONCLUSION

This paper presented an academic performance analytics and visualization tool that leverages machine learning algorithms to analyze, categorize, and predict student performance. The platform's ability to visualize academic data and provide personalized improvement recommendations has the potential to significantly enhance the educational experience for students, educators, and institutions. Future work will focus on expanding the dataset, incorporating more features, and improving the recommendation engine to further enhance its predictive accuracy.

## 9. FUTURE ENHANCEMENTS

- Integrate multilingual support to enhance accessibility for a global audience.
- Add gamification features (rewards, badges, progress tracking) to boost student engagement and motivation.

- Implement adaptive learning technologies to personalize content based on student performance.
- Utilize predictive analytics to identify at-risk students for early intervention.
- Develop a mobile app for convenient access by students, teachers, and parents.
- Enhance data privacy measures to comply with educational data protection standards, ensuring a secure solution.

Future enhancements aim to make the platform more engaging, accessible, and personalized through features like multilingual support, gamification, and adaptive learning. A mobile app and robust data privacy measures will ensure convenience and security for all users.

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