

# Visual Analysis of Student Performance

Visual Analytics Course Project - Fall 2025

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# Project Goal



**Problem:** Traditional 'Black Box' Grade Reports.  
Hide the root causes of failure



**Solution:** Interactive Exploratory Environment.  
Uncovers hidden patterns and relationships



**Goal:** Develop a Visual Analytics System.  
To support educational decision-making



**Key Capability:** Understand Multidimensional Risks.  
Empower experts with integrated Social + Academic insights

# Potential Users



## 1. School Counselors:

- Need to identify at-risk students before they fail.
- Use the tool to differentiate between behavioral issues and academic struggles.



## 2. Educational Policymakers:

- Need to validate hypotheses (e.g., 'Does affect grades more than study time?').
- Design broader intervention strategies based on data clusters.



# The Dataset



## Source:

UCI Machine Learning  
Repository (Student  
Performance Data Set)



## Volume & Subject:

Volume: 395 Students | Subject:  
Mathematics (Secondary School)



## Dimensions:

33 Attributes  
per student



## Type:

Mixed data (Numerical grades,  
Categorical demographics, Ordinal surveys)

# Data Structure (4 Pillars)



## 1. Demographics:

Age, Sex, Address  
(Urban/Rural),  
Family Size



## 2. Social Context:

Parent's Job,  
Relationships,  
Alcohol consumption  
(Walc/Dalc)



## 3. School Habits:

Study time,  
Travel time,  
Absences,  
Failures



## 4. Performance (Targets):

G1 (1st Period),  
G2 (2nd Period),  
G3 (Final Grade)

# Data Preprocessing



## Parsing:

Custom CSV parsing implemented in D3.js.



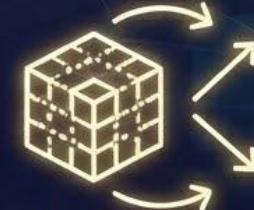
## Encoding:

Converting categorical strings (e.g., 'yes'/'no') to binary integers.



## Normalization:

Scaling numerical features to ensure fair weighting.



## Dim. Reduction:

PCA calculation to condense 33 features into 2 principal components.

# Visualization Strategy

Approach: Coordinated Multiple Views (CMV)



**1. Macro View:**  
**PCA Projection**  
(Global clustering)



**3. Micro View:**  
**Parallel Coordinates**  
(Individual profiles)



**2. Correlation View:**  
**Scatter Plot**  
(Grade progression)

**4. Context View:**  
**Bar/Histograms**  
(Distribution analysis)



# 1. Parallel Coordinates

## Key Features & Insights



### Purpose:

Multidimensional Profiling



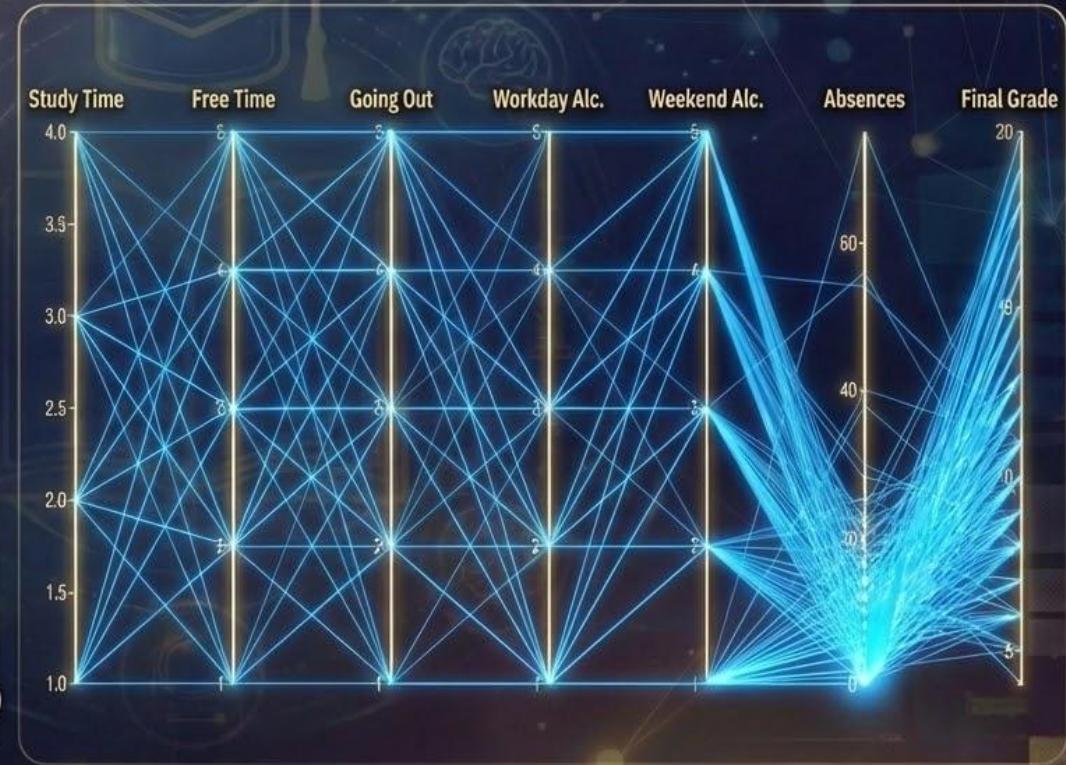
**Function:** Each student is a line crossing multiple axes  
(Grades, Alcohol, Study Time)



**Why?:** Only way to visualize  
N dimensions simultaneously



**Insight:** Detects specific  
profiles (e.g., 'High Alcohol' +  
'High Going Out Time' ->'Low G3')



## 2. Scatter Plot

### Key Features & Insights



**Purpose:**  
Performance Correlation  
(Study Time vs. Final Grade)



**Why?**: Tests assumption:  
More study ≠ Better grades

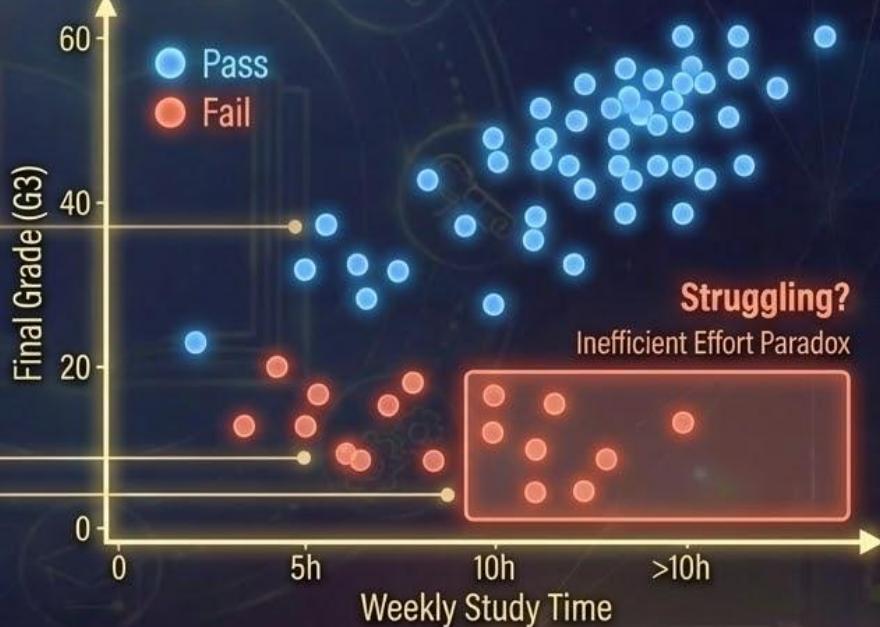


**Insight:** Reveals 'Inefficient Effort' paradox (High study time, low grades)



**Value:** Identifies students  
needing intervention in study  
methods

### Study Efficiency: Time vs. Final Grade



# 3. PCA Projection

## Key Features & Insights



**Purpose:**  
Cluster Analysis



**Technique:**  
Principal Component Analysis  
(2D Projection)

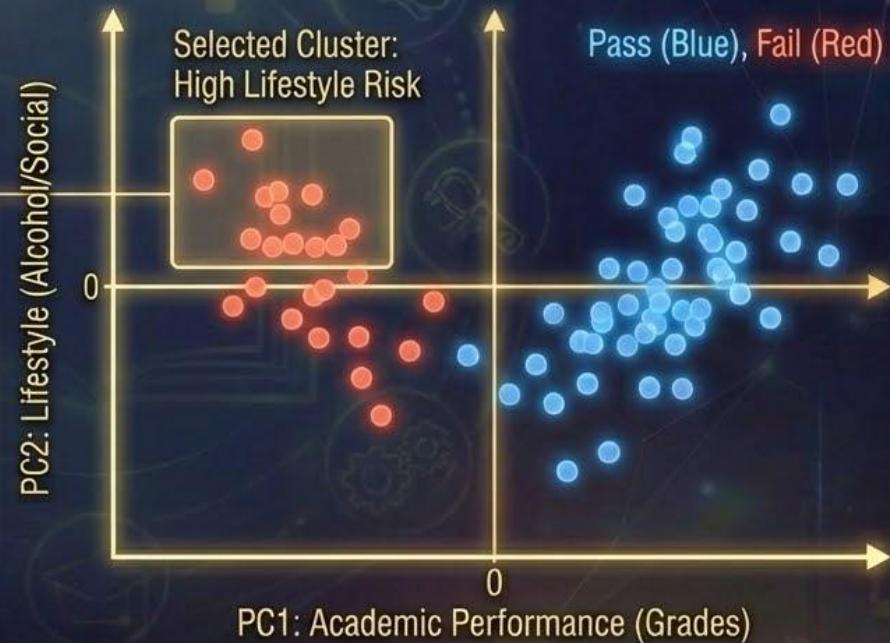


**Why?**: Reveals natural groupings based on similarity across ALL 33 variables



**Interaction:** 2D brushing allows selecting a 'cluster' to see their profile in other views

## PCA Projection: Student Clusters



# 4. Context Charts

## Key Features & Insights



**Purpose:** Global Filtering & Distribution Analysis



**Bar Charts:** Binary filters for 'Internet Access' and 'Romantic Relationships'



**Box Plots:** Statistical summaries (Median/Quartiles) for 'Age' and 'Absences'

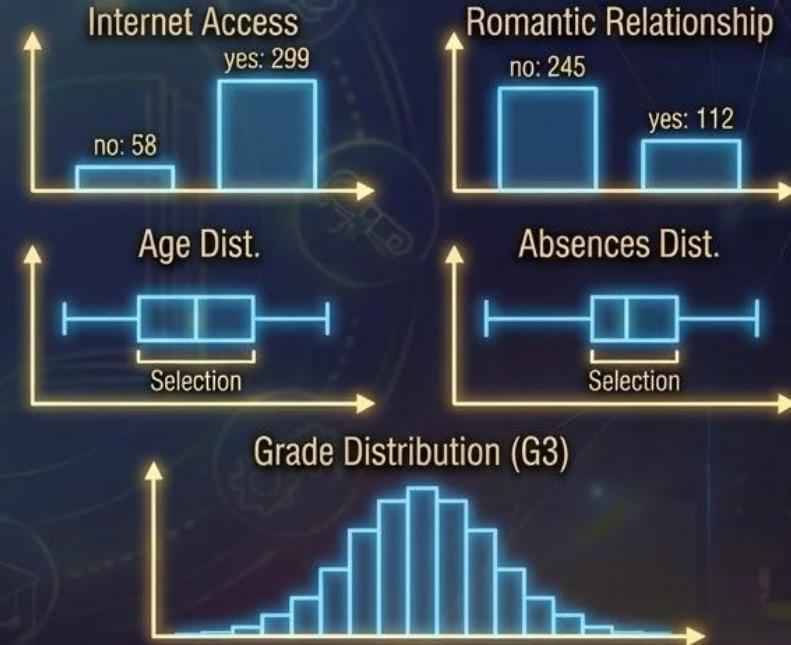


**Histogram:** Visualizes the distribution of Final Grades (G3) to assess the 'shape' of performance



**Interaction:** Acts as the controller; selecting bars or ranges here updates all other views

## Context Charts: Data Distribution & Filtering



# System Architecture



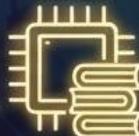
**Frontend & Styling:** HTML5, SASS (SCSS) for modular styling.



**Visualization Engine:** D3.js (Data-Driven Documents).



**Logic:** JavaScript (ES6+).



**Specific Libraries:** ‘pca-js’ used for real-time dimensionality reduction in the browser.

# Case Studies & Insight



## The 'Inefficient Effort' Paradox:

- Students studying >10h but failing (<10/20). Parallel plot reveals they are not drinkers/absentees. Problem is study method, not motivation.



## The 'Gifted Underachiever' Risk:

- Students studying <2h with high grades (<15/20). High correlation with 'Going Out'. Risk of future failure when raw intellect is not sufficient.



## Diminishing Returns of Study Volume

- Scatter plot reveals a saturation point. Studying >10h yields marginal gains over 5-10h, likely due to burnout.



## The 'Unproductive Leisure' Trap

High Free Time is neutral on its own. However, when combined with high Going Out in the Parallel Plot, grades collapse. The risk is mismanaged leisure.

# Conclusion



**Summary:** The system moves beyond static reporting to active exploration.



**Value:** Integrates statistical complexity (PCA) with interpretable details (Parallel Coords).



**Impact:** Allows detection of subtle student risks, such as the 'Inefficient Effort' paradox.