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Student Dropout Risk Clustering App (Unsupervised)

Allow educators to upload student performance CSV files and discover clusters of students (Low/Medium/High risk) using unsupervised learning.

Features

- 🖶 Upload 1 or more CSVs (e.g., performance + demographics)
- Auto data preprocessing and feature engineering
- III Unsupervised clustering using KMeans or DBSCAN
- Dimensionality reduction with PCA or t-SNE
- Visualize clusters and risk distribution
- Download filtered high-risk reports
- Ø Deploy on Streamlit + Docker + GitHub Actions

☐ Input Columns

Column	Description
student_id	Unique student identifier
quiz1/quiz2/quiz3	Quiz marks
assignment1/2/3	Assignment marks
midterm, final	Exam scores
total_lectures, present_lectures	Attendance tracking
lab_sessions, attended_labs	Lab participation
previous_gpa	Academic history
age, gender	Optional metadata
support_program, internet_access	Social factors

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```
- 🗁 notebooks/
   clustering_experiments.ipynb # Jupyter notebook for model development
 – 🗁 app/
   — __init__.py
                                 # KMeans/DBSCAN logic + PCA/t-SNE
    clustering.py
   preprocessing.py
                                  # Cleaning, feature engineering
   sql_playground.py
                                # SQL querying logic using DuckDB
   — utils.py
                                  # Helper functions (e.g. feature calc)
   — config.py
                                   # Feature config & settings
 - ☐ streamlit_app/
   — dashboard.py
                                  # Main Streamlit interface
     - components/
       file_uploader.py
       ├─ risk_summary.py
        visualizations.py
       └─ sql_playground_ui.py
 – 🗁 docker/
   └─ Dockerfile
 - 🗁 .github/
   workflows/
       └─ deploy.yml
                                 # GitHub Actions CI/CD
— requirements.txt
 - .gitignore
 - README.md
 PROJECT_PLAN.md
                                  ✓ This file (below)
```

Ⅲ Clustering Strategy

- Engineered Features:
 - attendance_ratio, assignment_avg, quiz_avg, lab_participation
- Dimensionality Reduction:
 - PCA, t-SNE
- Clustering Models:
 - KMeans, DBSCAN, IsolationForest
- Output:
 - Cluster labels → Interpreted into "Low", "Medium", "High Risk"

Dashboard Features (Streamlit)

- 🖶 Upload CSVs
- III Cluster summary cards (Low/Med/High counts)
- Scatter plots, bar charts by cluster
- 🖺 Risk-labeled student table

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- SQL Playground
 - o Query across multiple uploaded files
 - Join tables and run advanced filters
- 📤 Export reports per cluster

Tech Stack

Layer	Tools
ML & Preprocessing	pandas, scikit-learn, DuckDB
Clustering	KMeans, DBSCAN, PCA
Visualization	Streamlit
Backend/API (optional)	FastAPI
Deployment	Docker, GitHub Actions

% How It Works

- 1. User uploads 1+ CSVs
- 2. App cleans and processes data
- 3. Model clusters students into groups
- 4. Streamlit shows insights + SQL Playground
- 5. Teacher interprets risk levels, downloads student reports

Deployment

- Dockerized for local/cloud use
- GitHub Actions to auto-rebuild on push

Future Enhancements

- Ø LMS API integration (for auto data pull)
- Add active learning: manually label a few, then classify
- Alert system (email/WhatsApp for high-risk students)
- Q Live retraining pipeline using Airflow or Prefect

Outcome

A complete industry-level ML project solving a real-world education problem with:

- Unsupervised learning
- CI/CD, Docker

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- SQL integration
- Clean UX 🗸