

► MSc Project (CMP060L050H)



# Machine Learning-Based Student Behaviour Analysis and Prediction on Online Learning Platforms

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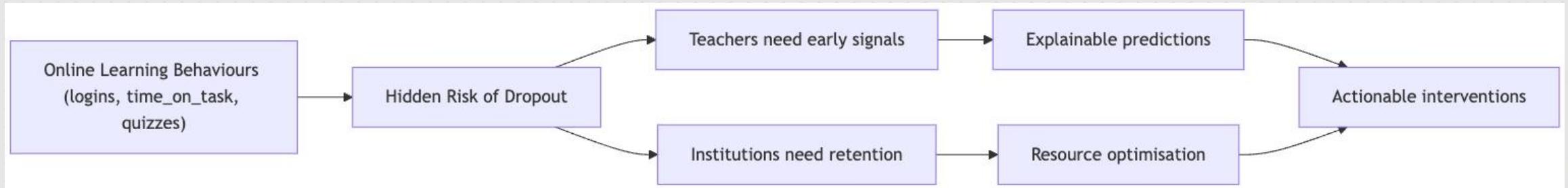
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# Context & Motivation (WHY)



Vocational learners show sparse, irregular engagement; tools must be interpretable.

Teachers need actionable & timely risk signals, not opaque scores.

False Negative cost > False Positive cost in classrooms  
→ recall-aware design.

# Problem & Objectives (WHAT)

**Problem:** off-the-shelf models transfer poorly;  
opaque predictions are hard to act on.

**Objective 1** Interpretable end-to-end pipeline (DB → dashboard).

**Objective 2** Transparent evaluation (Accuracy, Precision, Recall, F1, AUC)  
on held-out test.

**Objective 3** Minimal UI — high-risk table, a few visuals, single-student form.

# Literature Snapshot & Gap

Transparent (LR/DT) vs high-capacity (RF/GB/Deep):  
accuracy vs readability.

For vocational cohorts: start with an interpretable  
baseline; extend complexity later.

Earlier short paper (ICISCAE 2025) guided feature  
choices and readability stance.

2025 IEEE 8th International Conference on  
Information Systems and Computer Aided Education  
(ICISCAE 2025) Dalian, China

## NOTIFICATION OF ACCEPTANCE

Dear Author(s):

On behalf of the 2025 IEEE 8th International Conference on  
Information Systems and Computer Aided Education (ICISCAE 2025),  
we're glad to inform you that your paper:

Paper ID: [ICISCAE-31735](#)

Paper Title: [A learning feature selection model for high-dimensional sparse data of students' online behavior](#)

Author(s): [Rui Geng, John R. L. Moxon, BoHui Wang](#)

has been Accepted!

ICISCAE 2025 aims to bring researchers, engineers and students to the areas of information systems, computer engineering, information technology, network engineering and computer aided education, and will provide an international forum for sharing the most advanced research results, experiences and original research contributions on related topics.

All accepted papers will be published by IEEE CS (Computer Society) CPS, and will be submitted to EI Compendex, Thomson ISTP and Elsevier SCOPUS databases.

2025 IEEE 8th International Conference on  
Automation, Electronics and Electrical Engineering  
2025第8届信息系统与计算机辅助教育国际会议  
2025年7月24日  
组织委员会

# Data, Ethics & Scope

## Data & scope:

Synthetic behavioural data; minimal features (logins, study time (hours), quiz attempts); no real ICVE distributions; generated via parametric draws within realistic bounds.

## Governance:

Anonymisation-by-design; training artefacts (trained\_model.pkl, scaler.pkl) kept outside the UI layer.

## Protocol:

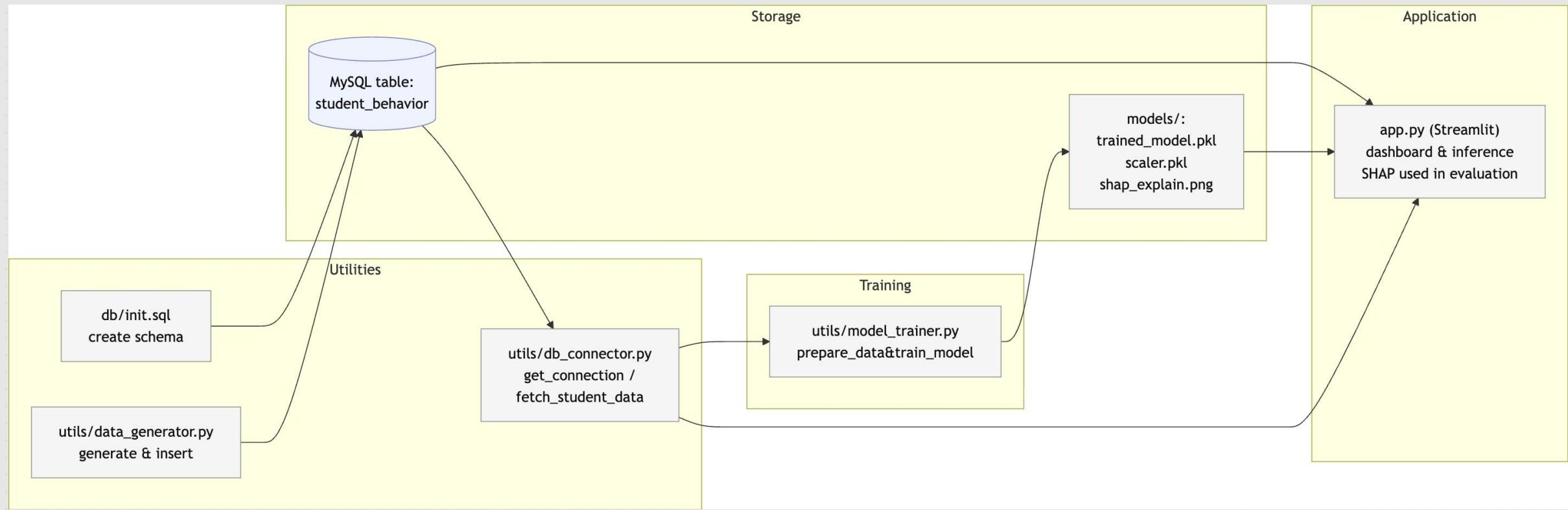
Stratified 80/20 split (seed=42); StandardScaler fit on train only; label = completion-rate threshold with median fallback ( $\approx$ 50/50 balance).

## Quality & next:

Realism checks (descriptives + correlation) show activity  $\uparrow \rightarrow$  High Risk  $\downarrow$  (expected); see “Evaluation Results – Synthetic Data Realism”; Post-MSc validation, calibration & A/B trials.



# End-to-End Architecture (HOW)



- DB init → feature prep → offline train & select → persist artefacts → explain → UI render.
- 80/20 stratified split (seed=42); scaler fit on train only; LR/RF/GB trained.
- Persisted artefacts (scaler.pkl, trained\_model.pkl) decouple training from UI (no retrain).

# Modules & Responsibilities

## db\_connector.py

- connection handling; fetch\_student\_data()

## data\_generator.py

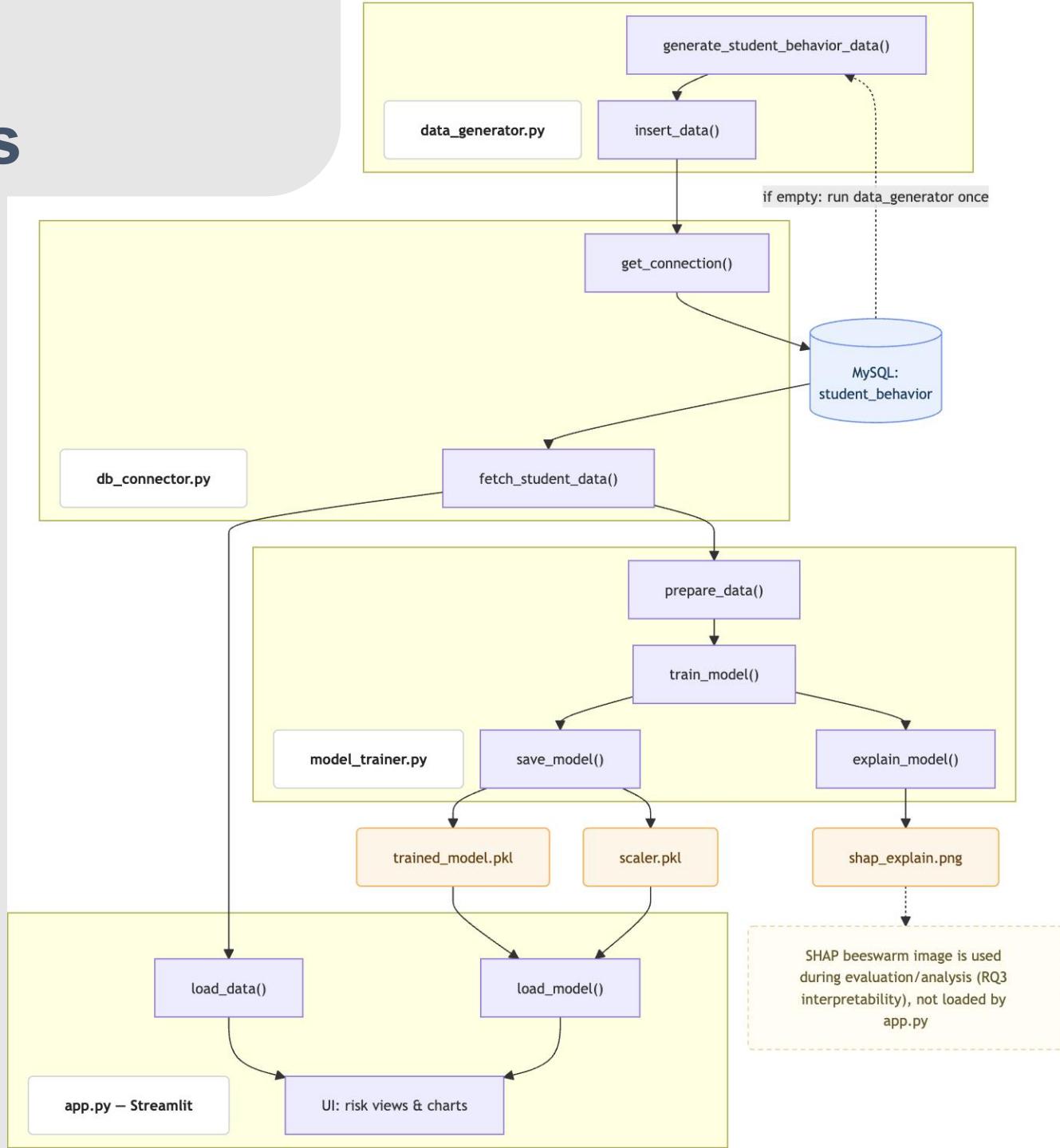
- reproducible cohort synthesis (fixed seed)

## model\_trainer.py

- prepare features; train LR/RF/GB;  
pick F1-best; save artefacts

## app.py

- cached loaders; cohort views;  
single-student form; visuals



# Dashboard - Student Risk & Visualisation

## Controls Panel

**Controls Panel**  
Use this panel to refresh and manage the student behaviour dataset.

Refresh Data

Data loaded successfully (cached)!

Last refreshed: 2025-11-09 12:11:34

**High-risk table(↑) + quick filters(↓)  
support weekly teacher review.**

### Interactive Filters

Select a student:

S001

Showing detailed engagement data for S001:

	id	student_id	login_count	time_spent	quiz_attempts	completion_rate	quiz_score	progress	created_at	risk_level
0	1	S001	12	9	4	0.69	79.6	69	2025-09-17 12:45:29	Low Risk

## Student Risk Prediction

### Predicted High-Risk Students

	student_id	login_count	time_spent	quiz_attempts	completion_rate
3	S004	14	3.2	2	0.34
14	S015	12	5	3	0.35
18	S019	10	4	4	0.43
22	S023	13	5.8	4	0.48
23	S024	8	3.9	5	0.52
27	S028	13	6.1	3	0.26
29	S030	14	6.4	4	0.45
31	S032	8	5.6	5	0.3
34	S035	13	4.9	4	0.42
37	S038	8	1.4	4	0.41

## Multi-dimensional Visualization

### Time Spent vs Completion Rate

Completion Rate

Time Spent (hours)

Legend: risk\_level (Low Risk: green, High Risk: red)

### Risk Level Distribution

High Risk 35.5%

Low Risk 64.5%

### Engagement Pattern Comparison

Metric	High Risk (Avg)	Low Risk (Avg)
login_count	4.87	8.76
time_spent	12.17	6.47
quiz_attempts	4.03	2.32
completion_rate	0.42	0.46

## Visuals explain cohort structure: scatter, donut, grouped bars.

# Dashboard - Single Student Prediction

## Single Student Prediction

Student Name      Student ID

Anne      stu1

Login Count      Study Time (hours)      Quiz Attempts

2    -    +      3.00    -    +      1    -    +

Predict

### Result

Student: Anne | ID: stu1

**Low Risk • Probability: 31.29%**

This student's learning situation is good; please maintain the current learning pace.

## Single Student Prediction

Student Name      Student ID

Anne      stu1

Login Count      Study Time (hours)      Quiz Attempts

4    -    +      2.50    -    +      6    -    +

Predict

### Result

Student: Anne | ID: stu1

**Medium Risk • Probability: 61.19%**

This student's learning performance is relatively stable; it is recommended to appropriately increase learning engagement and the frequency of quizzes.

## Single Student Prediction

Student Name      Student ID

Anne      stu1

Login Count      Study Time (hours)      Quiz Attempts

8    -    +      1.00    -    +      10    -    +

Predict

### Result

Student: Anne | ID: stu1

**High Risk • Probability: 85.01%**

This student's learning progress should be closely monitored, paying attention to assignment completion rate and interaction frequency.

Select a student or enter inputs.

Click **Predict** → show label and P(High Risk).

High  $\geq 0.66$ ; Medium  $0.33-0.6599$ ; Low  $< 0.33$ .

# Evaluation Design

## Split

Stratified 80/20 held-out test  
(seed = 42)

## Outputs

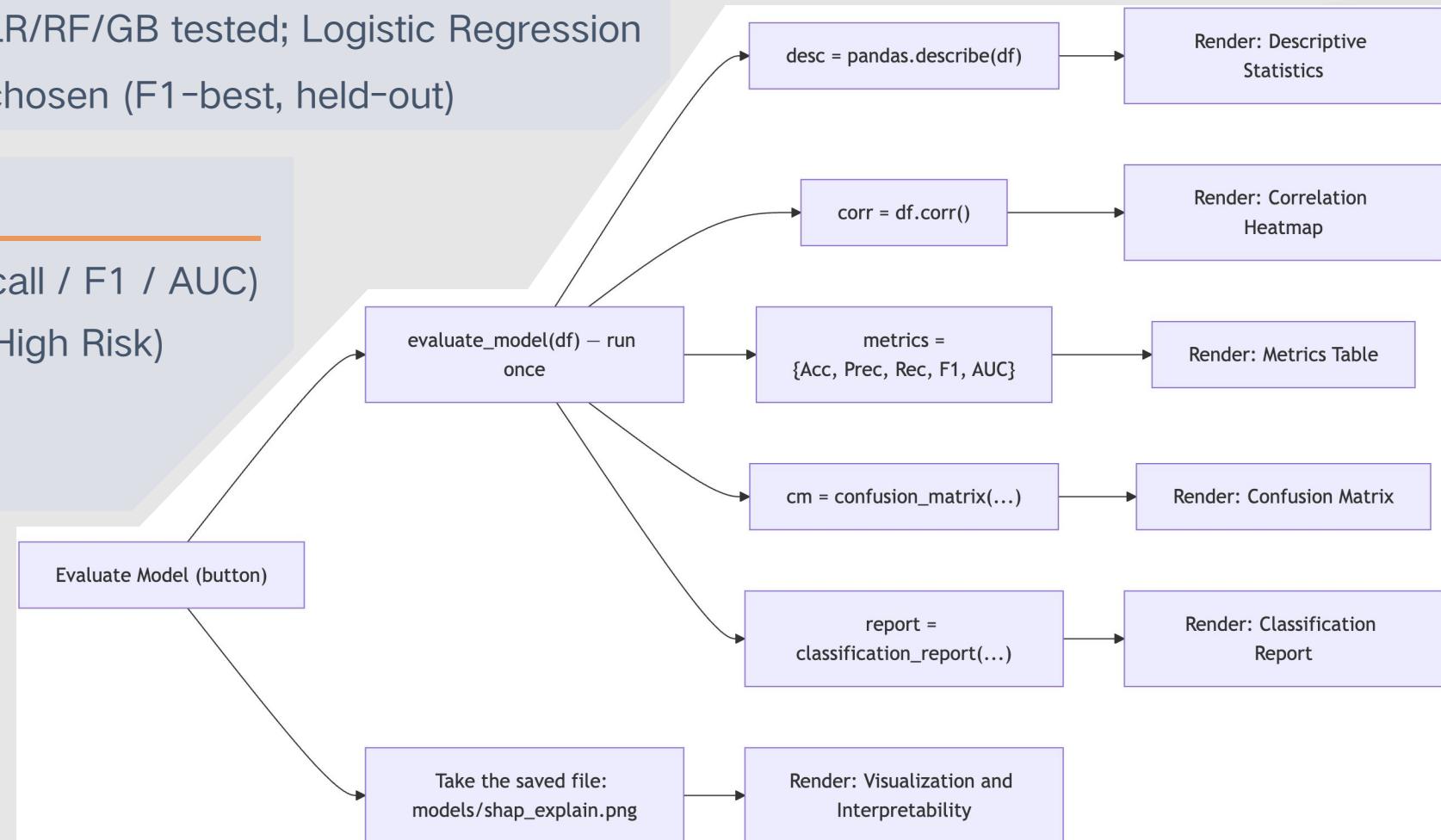
Metrics (Accuracy / Precision / Recall / F1 / AUC)  
Confusion Matrix (positive class = High Risk)  
Classification Report  
SHAP Beeswarm

## Protocol mirrors training

Scaler fit on train only  
Identical preprocessing on test  
No leakage → Fair comparison

## Model Selection

LR/RF/GB tested; Logistic Regression chosen (F1-best, held-out)

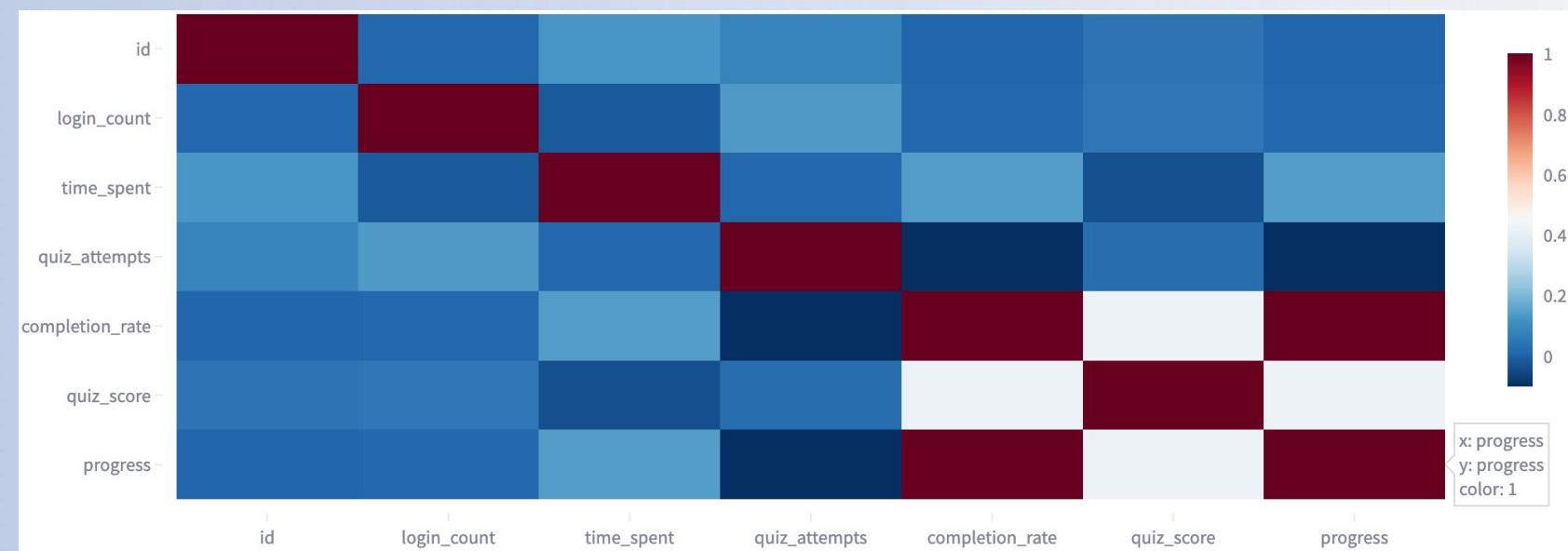


# Evaluation Results - Synthetic Data Realism

	count	mean	min	25%	50%	75%	max	std
id	200	100.5	1.0	50.75	100.5	150.25	200.0	57.8792
login_count	200	9.97	1.0	8.0	10.0	12.0	18.0	3.2391
time_spent	200	5.900499999999999	1.1	4.475	5.9	7.125	12.2	2.0113
quiz_attempts	200	2.925	1.0	2.0	3.0	4.0	5.0	1.4033
completion_rate	200	0.4434	0.22	0.38	0.445	0.5025	0.69	0.0937
quiz_score	200	68.896	31.9	61.475	68.85	77.2	94.9	11.3437
progress	200	44.34	22.0	38.0	44.5	50.25	69.0	9.372
created_at	200	2025-09-23 17:47:53	2025-09-08 12:45:29	2025-09-16 12:45:29	2025-09-24 12:45:29	2025-10-01 12:45:29	2025-10-07 12:45:29	None

Descriptive  
Statistics

Correlation  
Heatmap

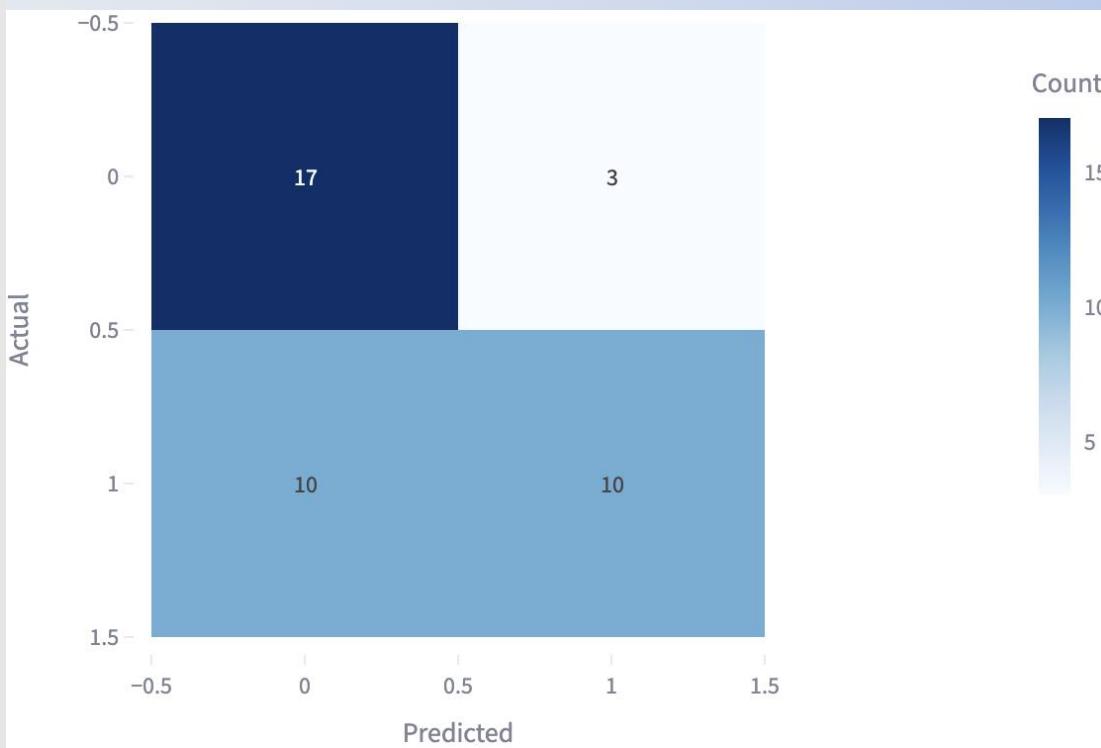


# Evaluation Results - Performance and Interpretability

## Model Metrics (Positive class = High Risk)

Accuracy	Precision	Recall	F1	AUC
0.675	0.7692	0.5	0.6061	0.5925

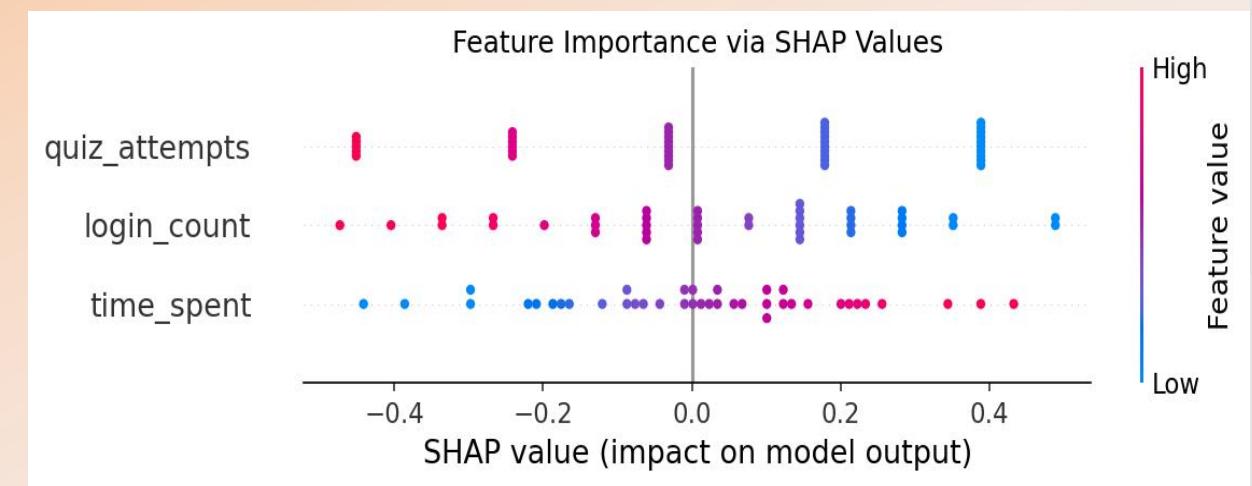
## Confusion Matrix



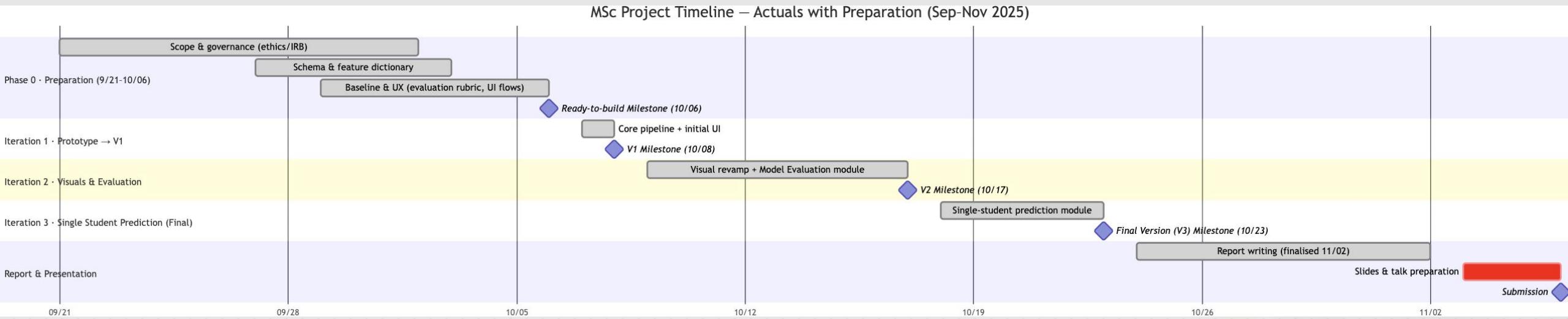
## Classification Report

	precision	recall	f1-score	support
Low-risk(0)	0.63	0.85	0.72	20
High-risk(1)	0.77	0.50	0.61	20
accuracy			0.68	40

## SHAP Beeswarm w.r.t. P(High Risk)



# Project Management & Actual Timeline



## Preparation (Oct-06)

- Governance & design ready

## Version 1 (Oct-08)

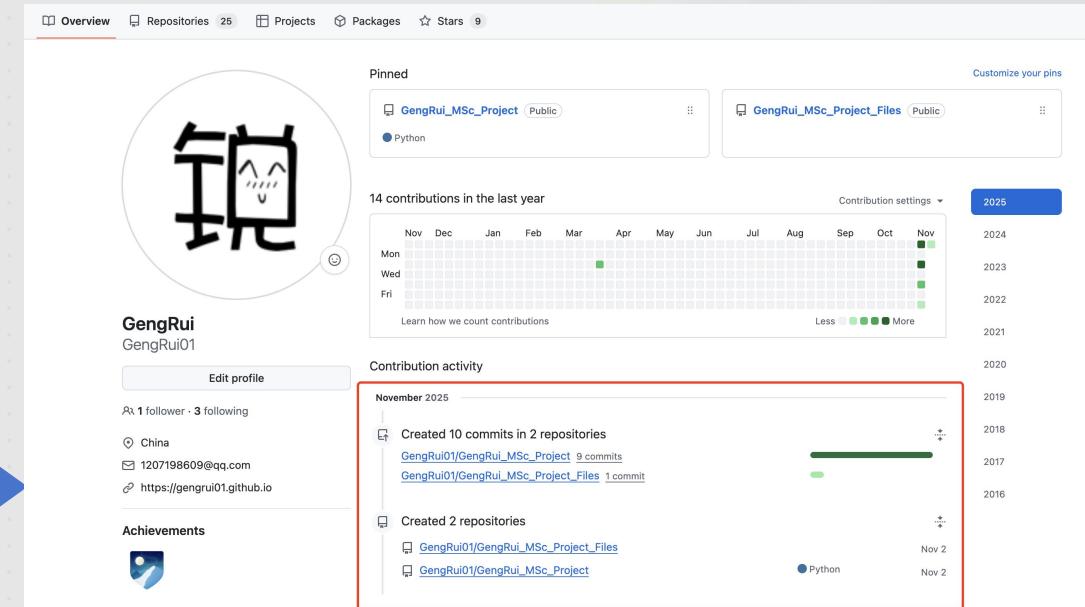
- Pipeline online; artefacts saved

## Version 2 (Oct-17)

- Evaluation module & visuals

## Version 3 (Oct-23)

- Single-student form; caching



# Contributions & Conclusion

## Contributions

explainability-first pipeline  
(DB → model → dashboard);  
persisted artefacts  
(models/\*.pkl);  
recall-prioritised evaluation.

## Results P(H)

Acc 0.6750  
Prec 0.7692  
Recall 0.500  
F1 0.6061  
AUC 0.5925

## Limitations

synthetic cohorts;  
baseline tuned  
for recall;  
data-sharing  
constraints.

## Conclusion

A teacher-actionable,  
reproducible, and fairness-  
oriented, recall-prioritised  
prototype that supports early  
identification and intervention.

## Future Work

data-sharing(ICVE) → schema mapping → cohort ingestion → threshold calibration  
→ subgroup-parity checks → A/B teaching trials → ethics & governance loop.

