SRMS Data Pipeline Guide

(Akari, Brightspace, Banner)

Technological University Dublin

Paul Doyle

Dean, Faculty of Engineering, Built Environment and Apprenticeships

October 29, 2025

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Scope

This guide documents the full SRMS pipeline from raw inputs to HTML viewers:

- Akari cleaner: src/akari_clean.py
- R27 cleaner (S1/S2): src/r27_clean.py
- R27 duplicate remover: src/r27_dedupe.py
- Programmes with Dummy CRN cleaner: src/programmes_dummy_crn_clean.py
- GLR cleaner (R27-only school mapping): src/glr_clean.py
- R27 augmentation with Dummy CRN: src/r27_augment_with_dummy_crn.py
- CRN2Use cleaner (Excel → per-term CSVs): src/crn2use_clean.py
- R27 flagging by CRN2USE: src/r27_flag_crn2use.py
- Brightspace cleaner: src/brightspace_clean.py
- R27 update with Brightspace Student Counts: src/r27_update_with_brightspace.py
- Akari viewer: src/viewer_akari_build.py
- R27 viewers: src/viewer_r27_build.py
- GLR viewer (by school, S1 vs S2 side-by-side): src/viewer_glr_build.py
- Akari vs R27 Differences viewer: src/viewer_programmename_diff_split.py
- End-to-end runner: src/pipeline_run.sh

Directory Structure

```
SRMS-Dashboard/
DataSources/
 Akari-Programme and Module data.xlsx
 R27_Expanded_202510.csv
 R27_Expanded_202520.csv
 Programmes_with_Dummy_CRN.csv
 CRN2Use.xlsx # sheets: 202510, 202520, 202530
 BrightspaceDump.xlsx # sheet: "All COs with roles"
 cleancsv/
 akari_clean.csv
 r27_202510_clean.csv
 r27_202520_clean.csv
 programmes_dummy_crn_clean.csv
 programmes_dummy_crn_conflicts.csv # only if conflicts
 glr_term1_clean.csv # from glr_clean.py
 glr_term2_clean.csv
html/
 index.html
 viewer_akari_by_school.html
 viewer_r27_202510.html
 viewer_r27_202520.html
 viewer_glr_by_school.html
 viewer_akari_vs_r27_split.html # new differences viewer
 src/
    akari_clean.py
    r27_clean.py
    r27_dedupe.py
    programmes_dummy_crn_clean.py
    glr_clean.py
    r27_augment_with_dummy_crn.py
    crn2use_clean.py
    r27_flag_crn2use.py
    brightspace_clean.py
    r27_update_with_brightspace.py
    viewer_akari_build.py
    viewer_r27_build.py
    viewer_glr_build.py
```

viewer_programmename_diff_split.py # new
pipeline_run.sh

Akari Cleaner (src/akari_clean.py)

Input/Output

- Input: ../DataSources/Akari-Programme and Module data.xlsx
- Output: ../cleancsv/akari_clean.csv

Selected Columns & Final Order (exact)

Output Column	Source / Notes
TU Programme Code	From sheet column of the same meaning
Year	From sheet "Year" (string-preserved)
Semester	Normalised to one of: 1, 1&2, 2
Code	Module code (as-is)
Title	Module name
Delivery Type	As-is
Credits	As-is (kept as string; numeric on client)
Programme Title	As-is
School Responsible	Case-normalised to avoid duplicates (e.g. "of" vs
	$\mathrm{``Of")}$
Full Time/Part Time	As-is
Campus	As-is
NFQ Level	As-is
Programme Type	As-is
Programme Delivery Mode	As-is

Normalisation Rules

- Header matching: tolerant (case/spacing/punctuation).
- Semester: mapped to 1, 1&2, 2.
- All fields read as text to preserve codes; UTF-8 BOM on output.

Run

cd src

python Akari-CSV-Clean.py

R27 Raw Data Consistency Test (src/CheckStudentCountConsistency.py)

Purpose

Before any data cleaning or transformation is performed, the R27 raw file (R27_Expanded_202510.csv) is validated to ensure internal consistency of student registration data. This step identifies modules where STUDENT_COUNT values are not consistent across records that share the same key identifiers.

Test Logic

The script verifies that for each unique combination of:

- Programme Code
- CRN
- SSRRMAJ_MAJR_CODE
- Module (or Module Name)

the recorded STUDENT_COUNT remains the same across all rows. If different student counts exist for the same key, the record group is flagged as inconsistent.

Script Implementation

```
import pandas as pd
from pathlib import Path
import re

DATA_DIR = Path("../DataSources")
INPUT_FILE = DATA_DIR / "R27_Expanded_202510.csv"

OUTPUT_FILE = DATA_DIR / "R27_inconsistent_student_counts.csv"

df = pd.read_csv(INPUT_FILE, encoding='utf-8-sig', low_memory=False)
df.columns = [re.sub(r'[^A-Za-z0-9]', '', c).lower() for c in df.columns]
```

```
# Identify relevant columns
programme_col = 'programmecode'
crn_col = 'crn'
major_col = 'ssrrmajmajrcode'
module_col = 'module'
student_col = 'studentcount'
df[student_col] = pd.to_numeric(df[student_col], errors='coerce')
# Group and compare counts
key_cols = [programme_col, crn_col, major_col, module_col]
summary = (df.groupby(key_cols)[student_col]
             .agg(['count','nunique','min','max'])
             .reset_index()
             .rename(columns={'count':'RecordCount',
                            'nunique':'DistinctStudentCounts',
                            'min':'MinStudentCount',
                            'max':'MaxStudentCount'}))
# Filter inconsistencies
inconsistent = summary[summary['DistinctStudentCounts'] > 1]
inconsistent.to_csv(OUTPUT_FILE, index=False, encoding='utf-8-sig')
print(f"Inconsistent combinations written to: (0UTPUT_FILE)")
```

Results

The test was executed on the unprocessed dataset before running any cleaning operations. The output confirmed full consistency of student registration counts across all module records.

```
(base) paul@soc-MacMini-PD src % python CheckStudentCountConsistency.py
Loading file: ../DataSources/R27_Expanded_202510.csv
Detected columns:
Programme: programmecode
CRN: crn
Major: ssrrmajmajrcode
Module: module
Student Count: studentcount
Consistent module records: 21,729
Inconsistent module records: 0
Inconsistent combinations written to: .../DataSources/R27_inconsistent_student_counts.
   csv
Sample inconsistent combinations:
Empty DataFrame
Columns: [programmecode, crn, ssrrmajmajrcode, module, RecordCount,
   DistinctStudentCounts, MinStudentCount, MaxStudentCount]
Index: []
```

Conclusion

The R27 raw file for term 202510 demonstrated complete internal consistency. No cases were found where the STUDENT_COUNT differed for records sharing the same Programme Code, CRN, SSRRMAJ_MAJR_CODE, and Module. This confirms that registration data is stable at the raw source level and subsequent cleaning processes can safely assume one consistent student count per module delivery.

R27 Cleaner (src/r27_clean.py)

Inputs/Outputs

- $\bullet \ \, Inputs: \ \, .../DataSources/R27_Expanded_202510.csv, \ \, .../DataSources/R27_Expanded_202520.csv$
- Outputs: ../cleancsv/r27_202510_clean.csv, ../cleancsv/r27_202520_clean.csv

Run

cd src
python r27_clean.py

Selected Columns & Final Order

Source / Transform
Programme_Code
Extracted from SMRARUL_AREA: last digits;
fallback: F/P/FP;digits; (e.g. FP3 \rightarrow 3)
From Part_of_Term mapped: $1 \rightarrow \text{Semester 1; 2}$
\rightarrow Semester 2; FY/Y \rightarrow Semester 1 & 2
${\tt SMRARUL_SUBJ_CODE} + {\tt space} + \\$
SMRARUL_CRSE_NUMB_LOW
Module (module name)
Programme_Desc (programme name)
SMBARUL_KEY_RULE
Credits (kept as string)
Campus
Mapped from Dept (see mapping)
As-is
As-is
CRN
SSRRMAJ_MAJR_CODE
As-is

StaffID ID

FirstName FirstName Lastname Lastname

$\mathbf{Dept} \to \mathbf{School}$ Responsible Mapping

The cleaner applies a fixed mapping. Key examples:

\mathbf{Dept}	School Responsible
SABE	School of Architecture, Building and Environment
SSCI	School of Surveying & Construction Innovation
SCSC	School of Computer Science
SEEE	School of Electrical & Electronic Engineering
GRDR	Graduate Research
CONS	Conservatoire
GBUS	Graduate Business School
SFEH	School of Food Science & Environmental Health
0 11 14	

^{...}full list embedded in the script; unknown codes pass through as their raw Dept value and are reported in a warning.

Header Matching & Normalisation

- Tolerant header matching: case/underscores/punctuation (e.g. programme code ~ Programme_Code).
- Whitespace collapsed; values trimmed.
- Diagnostics printed: row/column counts; unmapped Dept codes; count of rows missing Module.

R27 Duplicate Remover

(src/r27_dedupe.py)

Purpose

Remove duplicate rows in the cleaned R27 outputs. Duplicates are determined after normalising whitespace and ignoring staff-identifying fields (and optionally Delivery Type), as configured in the script.

Inputs/Outputs

• Inputs/Outputs (overwritten in place): ../cleancsv/r27_202510_clean.csv, ../cleancsv/r27_202520

Run

```
cd src
python r27_dedupe.py
```

Process

- 1. Loads each cleaned CSV using pandas.
- 2. Identifies fully identical rows (same values across all columns).
- 3. Removes duplicates and overwrites the original file.
- 4. Reports the total number of rows before and after cleaning.

Example Output

```
[INFO] Processing: R27_Expanded_202510_clean.csv
Total rows: 23891
Duplicates removed: 121
Final rows: 23770
File overwritten: ../cleancsv/R27_Expanded_202510_clean.csv
```

[INFO] Processing: R27_Expanded_202520_clean.csv

Total rows: 25032

Duplicates removed: 98

Final rows: 24934

File overwritten: ../cleancsv/R27_Expanded_202520_clean.csv

[INFO] Duplicate removal complete.

Programmes with Dummy CRN Cleaner

(src/programmes_dummy_crn_clean.py)

Purpose

Normalise the mapping of TU Programme Code \rightarrow CRNs for 202510/202520 from the CSV export.

Inputs/Outputs

- Input: ../DataSources/Programmes_with_Dummy_CRN.csv
- Outputs: ../cleancsv/programmes_dummy_crn_clean.csv
 - ../cleancsv/programmes_dummy_crn_conflicts.csv (only if conflicts)

Run

```
cd src
python programmes_dummy_crn_clean.py
```

Process

- 1. Read CSV as text to preserve codes verbatim.
- 2. Extract programme_code using regex (TU\d{3,4}[A-Z]?) from the Programme column; if the cell is itself a TU code, use it directly.
- 3. Trim and normalise whitespace in CRN columns (202510/202520).
- 4. Group by programme_code:
 - Output one row per programme code; choose the *mode* CRN per term (deterministic tie-break) for stability.
 - If more than one distinct CRN exists for a programme/term, record the case in the conflicts CSV.

5. Write the cleaned CSV (UTF-8 with BOM).

Notes

- If the source remains in Excel format, export to CSV first to ensure consistent parsing.
- The conflicts CSV is advisory; if a hard failure is preferred on CRN clashes, switch the script to raise on nunique > 1.

GLR Cleaner (src/glr_clean.py)

Inputs/Outputs

- Inputs:
 - ../DataSources/GeneralLearnerRecordTerm1.csv
 - ../DataSources/GeneralLearnerRecordTerm2.csv
 - ../cleancsv/programmes_dummy_crn_clean.csv (for DUMMY_CRN join)
 - ../cleancsv/r27_202510_clean.csv (fallback school map)
 - ../cleancsv/r27_202520_clean.csv (fallback school map)
- Outputs:
 - ../cleancsv/glr_term1_clean.csv
 - ../cleancsv/glr_term2_clean.csv

Key Logic (exact)

- 1. Read GLR (per term) and normalise REGISTRATION_STATUS \rightarrow {BLANK, EL, ES, NS, SI, SW}; others drop or map to OTHER then excluded.
- 2. Aggregate counts by PROGRAMME_CODE and status.
- 3. Join DUMMY_CRN using term-specific column from programmes_dummy_crn_clean.csv.
- 4. **School Responsible**: Akari is no longer read or used. The school is sourced from the corresponding **R27 map only** (per term). Where missing, the field remains blank.
- 5. Output columns (order): PROGRAMME_CODE, SCHOOL_RESPONSIBLE, DUMMY_CRN, BLANK, EL, ES, NS, SI, SW.

Failure Modes

- Missing GLR inputs \Rightarrow hard error with hint listing available CSVs.
- Missing CRN map column for the selected term (e.g. 202510_crn) \Rightarrow hard error.

General Learner Record Viewer (src/viewer_glr_build.py)

Purpose

Static viewer for GLR by **School Responsible** with **side-by-side** tables for Term 202510 and Term 202520, and a status summary.

Inputs/Outputs

- Inputs: ../cleancsv/glr_term1_clean.csv, ../cleancsv/glr_term2_clean.csv
- Output: ../html/viewer_glr_by_school.html

Behaviour

Single School filter; viewer renders two tables (left=202510, right=202520) plus a summary panel with per-status counts and a TOTAL row.

R27 Augmentation with Dummy CRN

(src/r27_augment_with_dummy_crn.py)

Purpose

Left-join the Dummy CRN mapping by TU Programme Code and add: DummyCRN_202510 to r27_202510_clean.csv and DummyCRN_202520 to r27_202520_clean.csv.

Inputs/Outputs

- Inputs: ../cleancsv/r27_202510_clean.csv, ../cleancsv/r27_202520_clean.csv, ../cleancsv/pro
- Outputs: same R27 files overwritten in-place with the new columns.

Run

```
cd src
python r27_augment_with_dummy_crn.py
```

Process

- 1. Read the cleaned R27 CSV and the Programmes_with_Dummy_CRN_clean.csv mapping.
- 2. Normalise the TU Programme Code column in both datasets (trim, uppercase, remove stray BOMs).
- 3. Perform a left join on TU Programme Code.
- 4. Insert the appropriate dummy CRN column:
 - DummyCRN_202510 for the 202510 file.
 - DummyCRN_202520 for the 202520 file.
- 5. Overwrite the original cleaned R27 files.

Example Output

[INFO] Augmenting: R27_Expanded_202510_clean.csv

Added column: DummyCRN_202510
Rows with mapped CRN: 1248 / 23770

File overwritten: ../cleancsv/R27_Expanded_202510_clean.csv

[INFO] Augmenting: R27_Expanded_202520_clean.csv

Added column: DummyCRN_202520 Rows with mapped CRN: 1274 / 24934

File overwritten: ../cleancsv/R27_Expanded_202520_clean.csv

[INFO] Augmentation complete.

Notes

• Matching is case-insensitive and whitespace-normalised.

- The DummyCRN_* columns remain empty where the TU Programme Code is not found in the mapping.
- This step must be run *after* both the R27 and the Programmes-with-Dummy-CRN cleaners.

CRN2Use Cleaner (src/crn2use_clean.py)

Purpose

Ingest the Excel CRN2Use.xlsx and output one CSV per required sheet (terms only).

Inputs/Outputs

- Input: ../DataSources/CRN2Use.xlsx (sheets processed: 202510, 202520, 202530; e.g. sheet index is ignored)
- Outputs: ../cleancsv/crn2use_202510.csv, ../cleancsv/crn2use_202520.csv, ../cleancsv/crn2u

Output Columns (exact)

- $\bullet \ \, \mathtt{module} \ \, \mathsf{code} \leftarrow \mathrm{column} \,\, \mathsf{C\&D} \,\, (\mathrm{or} \,\, \mathrm{built} \,\, \mathrm{from} \,\, \mathsf{SSBSECT_SUBJ_CODE} \,\, + \,\, \mathsf{SSBSECT_CRSE_NUMB})$
- $\bullet \ \, \texttt{Programme Code} \leftarrow \texttt{SSRRPRG_PROGRAM} \\$
- ullet CRN2Use \leftarrow CRN to Use

Run

```
cd src
python crn2use_clean.py
```

R27 Flagging by CRN2USE (src/r27_flag_crn2use.py)

Purpose

Mark rows in each R27 cleaned file where the CRN appears in the corresponding crn2use_{term}.csv. Adds a new column CRN2USE equal to the row's CRN when present in the list; blank otherwise.

Inputs/Outputs

- Inputs: ../cleancsv/r27_202510_clean.csv, ../cleancsv/r27_202520_clean.csv, ../cleancsv/crn ../cleancsv/crn2use_202520.csv
- Outputs: same R27 files overwritten in-place, with an added CRN2USE column per row.

Run

cd src
python r27_flag_crn2use.py

Brightspace Cleaner (src/brightspace_clean.py)

Purpose

Ingest Brightspace enrolment data and produce a minimal, normalised CSV for downstream joins/validation.

Inputs/Outputs

- Input (Excel): ../DataSources/BrightspaceDump.xlsx (sheet: All COs with roles)
- $\bullet \ \mathrm{Output} \ (\mathrm{CSV}, \ \mathrm{UTF\text{-}8\ BOM}) \\ : \ .. \ / \mathtt{cleancsv/brightspace_brightspace_dump_clean.csv}$

Output Columns (exact order)

Column	Source / Transform
Code	From CourseCode: token 1 (title-cased) + space +
	token 2. Example: DATA-H1010 \rightarrow Data H1010.
CRN	From CourseCode: token 3, preserved as string; must
	match $d{5}$.
Semester	From CourseCode: token 6 mapped
	$\{202510 \rightarrow \text{Semester } 1, 202520 \rightarrow \text{Semester } 2,$
	$202530 \rightarrow \text{Semester } 3$; otherwise blank.
StudentCount	From sheet column StudentCount; coerced to
	numeric (NaN on non-numeric).
LecturerCount	From sheet column LecturerCount; coerced to
	numeric (NaN on non-numeric).

Parsing Rules (CourseCode)

- Split on -; expected 6 tokens (strict logical order): SUBJECT-MODULE-CRN-PROGRAMME-DELIVERY-TERMCODI
- If #tokens < 6, pad with None. If > 6, merge any extra tokens into the last position (term token) to avoid index errors.

- Code := SUBJECT (title-cased) + space + MODULE.
- **CRN** := token 3 (kept as string).
- Semester := map final token {202510,202520,202530} to {Semester 1, 2, 3}; no mapping \Rightarrow blank.

Row Filters (applied in this order)

- 1. StudentCount: keep rows with 0 <StudentCount ≤ 1000 .
- 2. Identifiers: drop rows with missing Code or missing CRN.
- 3. **CRN format**: keep rows where CRN matches $d{5}$ (five numeric digits only); all others dropped.

Header Tolerance

- CourseCode is located via tolerant header matching (case/punctuation stripped). If not found ⇒ hard error.
- StudentCount and LecturerCount must be present; absence \Rightarrow hard error.

Runtime Summary (stdout)

On each run the script prints:

- records loaded from sheet;
- records kept/removed by the StudentCount filter;
- rows dropped for missing/invalid Code/CRN (including non-5-digit CRNs);
- final row count and output path.

Run

cd src
python brightspace_clean.py

Failure Modes & Remedies

- File not found: verify ../DataSources/BrightspaceDump.xlsx exists and sheet name is exactly All COs with roles.
- Missing columns: ensure CourseCode, StudentCount, LecturerCount exist in the sheet (header spelling/spacing may vary; CourseCode is matched tolerantly, counts are required).
- Malformed CourseCode: rows with insufficient tokens may yield missing Code/CRN or non-mapped Semester; such rows are dropped by the identifier/CRN filters.
- Term code not mapped: if the last token is not one of 202510/202520/202530, Semester is blank (row still included if other criteria pass).

R27 Update with Brightspace Student Counts

Purpose

Populate a new column **BrightSpace_StudentCount** in the R27 cleaned CSVs by joining Brightspace counts on **Code+CRN**, using the appropriate Brightspace semester partition.

Inputs/Outputs

- Inputs:
 - ../cleancsv/r27_202510_clean.csv (target: Semester 1)
 - ../cleancsv/r27_202520_clean.csv (target: Semester 2)
 - ../cleancsv/brightspace_brightspace_dump_clean.csv
- Outputs: R27 files updated in place with a new column BrightSpace_StudentCount.

Join Logic (exact)

- 1. Brightspace rows are partitioned by Semester:
 - Semester $1 \Rightarrow \text{used to update r27_202510_clean.csv}$.
 - Semester $2 \Rightarrow \text{used to update r27_202520_clean.csv}$.
- 2. Match key: Code + CRN. Normalisation of Code on both sides:
 - strip leading/trailing whitespace and BOM;
 - collapse internal whitespace to single spaces;
 - uppercase.
- 3. Only Brightspace rows with CRN matching ^\d{5}\$ are retained (consistent with the Brightspace cleaner).
- 4. If multiple Brightspace rows map to the same (Code, CRN) within a semester, the updater takes the max StudentCount.
- 5. For R27 rows with no Brightspace match, BrightSpace_StudentCount is left blank.

Column Added

Column	Notes
BrightSpace_StudentCount	Integer from Brightspace StudentCount (per matched
	Code+CRN); blank if no match.

Run

cd src
python r27_update_with_brightspace.py

Failure Modes & Checks

- Missing Code/CRN columns in R27 files \Rightarrow hard error.
- \bullet Missing or empty Bright space CSV \Rightarrow no updates applied; script logs a warning per sem ester.
- Non-5-digit CRNs in Bright space are ignored by design.

Viewer Builders

Akari Viewer (src/viewer_akari_build.py)

```
• Input: ../cleancsv/akari_clean.csv
```

• Output: ../html/viewer_akari_by_school.html

R27 Viewers (src/viewer_r27_build.py)

```
• Inputs: ../cleancsv/r27_202510_clean.csv, ../cleancsv/r27_202520_clean.csv
```

• Outputs: ../html/viewer_r27_202510.html, ../html/viewer_r27_202520.html

Run

```
cd src
python viewer_akari_build.py
python viewer_r27_build.py
```

Akari Viewer Builder (src/build_programme_viewer.py)

Purpose

Generate an interactive, self-contained HTML (../html/programme_viewer_by_school.html) for exploring Akari data by:

- 1. School Responsible
- 2. Programme (Title & TU Programme Code)
- 3. Year

It displays modules grouped by semester (1, 1&2, 2) and totals credits per group.

Input/Output

- Input: ../cleancsv/Akari-ProgrammeAndModuleData.csv
- Output: ../html/programme_viewer_by_school.html

Notes

- Data is embedded (no runtime file access).
- Credits summed client-side (parsed as floats).

Run

cd src
python build_programme_viewer.py

R27 Viewer Builder (src/build_R27_viewer.py)

Purpose

Generate two separate self-contained HTML viewers using only R27 cleaned CSVs:

- ../html/r27_viewer_202510.html (Semester 1 data)
- ../html/r27_viewer_202520.html (Semester 2 data)

Filters mirror the Akari viewer: School → Programme Code → Year; optional: Campus, Full Time/Part Time. Tables include columns: Code, CRN, Title, Credits, Delivery Type, Students. Semester credit totals are shown.

Inputs/Outputs

- Inputs:
 - ../cleancsv/R27_Expanded_202510_clean.csv
 - ../cleancsv/R27_Expanded_202520_clean.csv
- Outputs:
 - ../html/r27_viewer_202510.html
 - ../html/r27_viewer_202520.html

Run

```
cd src
python build_R27_viewer.py
```

16.1 General Learner Record Viewer Enhancement

The General Learner Record (GLR) Viewer has been updated to provide a simplified, unified interface for reviewing programme-level data across academic terms. The new implementation removes multiple selection filters in favour of a single, robust school-level filter and introduces a comparative visualisation of the two academic terms.

16.1.1 Design Objectives

The objective of this update was to:

- Simplify the user interaction model by retaining only one filter (School Responsible).
- Enable side-by-side comparison of **Term 202510** and **Term 202520**.
- Provide automatic summary statistics, including per-status counts and overall totals for each term.
- Maintain static portability (no server-side code required) while ensuring responsive and interactive functionality via embedded JavaScript.

16.1.2 Functional Overview

The viewer is generated from two cleaned data sources:

- glr_term1_clean.csv (Term 202510)
- glr_term2_clean.csv (Term 202520)

Each record represents a unique **Programme Code** and includes:

- The associated **School Responsible**
- \bullet The corresponding $\mathbf{DUMMY_CRN}$ for each term
- Six status category counts: BLANK, EL, ES, NS, SI, and SW

16.1.3 Interface Behaviour

- The School dropdown allows selection of either a specific school or all schools combined.
- Upon selection, the viewer dynamically displays two tables:
 - Term 202510 (left column)
 - Term 202520 (right column)
- Each table lists programme codes, their schools, dummy CRNs, and the six status counts.
- The summary panel above the tables displays:
 - The number of programmes currently shown.
 - A per-status count for both terms.
 - A new **TOTAL** row providing the overall count across all statuses per term.

16.1.4 Implementation Notes

- The viewer is built entirely from static assets using Bootstrap 5 and embedded JavaScript.
- Data is pre-aggregated and converted to JSON within the build process using the Python script build_glr_viewer.py.
- No external dependencies or network access are required for rendering.

16.1.5 Output

The resulting static HTML file is written to: $% \left(1\right) =\left(1\right) \left(1\right)$

../html/viewer_glr_by_school.html

It provides a permanent, portable reference for GLR data, allowing comparative inspection between terms without requiring database connectivity or manual filtering.

Akari vs R27 Differences Viewer (src/viewer_programmename_diff_split.py)

Purpose

Produce a static HTML that highlights:

- 1. Programmes in Akari not present in R27.
- 2. Programmes in **R27** not present in **Akari**.
- 3. Programmes where **Programme Title** differs.
- 4. Programmes where **School Responsible** differs.

Inputs/Outputs

- Inputs: ../cleancsv/akari_clean.csv, ../cleancsv/r27_202510_clean.csv
- Output: ../html/viewer_akari_vs_r27_split.html

Comparison Rules (exact)

- Code key: TU Programme Code (uppercased; first non-empty per code retained in both sources).
- Title comparison (asymmetric):
 - 1. On Akari titles only, strip award/level wrappers at the start, iteratively:
 e.g. Bachelor of Engineering Technology (Hons) in, Bachelor of Science in, BA
 (Hons) in, Higher Certificate in, Postgraduate Certificate in, Higher Diploma in
 ...in, MA/MSc/MEng/MBA, etc.
 - 2. Then, on *both sides*, normalise synonyms and symbols for compare: $\mathbf{Eng} \equiv \mathbf{Engineering}$; $\mathbf{Uni} \equiv \mathbf{University}$; & $\equiv \mathbf{and}$; case/space-insensitive.
 - 3. If the normalised titles differ, record in "Programmes with different titles".
- School comparison: normalise case/spacing and treat & \equiv and. If different postnormalisation, record in "Programmes with different schools".
- **Sorting**: the *School differences* table is sorted by Akari's School (normalised), then by programme code; other tables sort by code.

Interface

Four sections with counts; a top-right "View" dropdown filters to a single section or shows all; an instruction box explains the rules.

Operational Checklist

- 1. Place sources in DataSources/:
 - Akari Excel
 - R27 raw CSVs: R27_Expanded_202510.csv, R27_Expanded_202520.csv
 - Programmes_with_Dummy_CRN.csv
 - CRN2Use.xlsx (tabs: 202510/202520/202530)
 - Brightspace Excel (sheet: All COs with roles)
- 2. Run the end-to-end pipeline:

```
cd src
./pipeline_run.sh
```

- 3. Outputs:
 - Clean CSVs in cleancsv/.
 - Viewers in html/.
- 4. Pipeline stages (in order):
 - (a) Akari cleaner
 - (b) R27 cleaner
 - (c) R27 duplicate remover
 - (d) Programmes with Dummy CRN cleaner
 - (e) GLR cleaner (R27-only school mapping)
 - (f) Brightspace cleaner
 - (g) R27 augmentation with Dummy CRN
 - (h) CRN2Use cleaner
 - (i) R27 flagging by CRN2USE
 - (j) R27 update with Brightspace Student Counts
 - (k) Build Akari viewer
 - (l) Build R27 viewers
 - (m) Build GLR viewer
 - (n) Build Akari vs R27 Differences viewer

Troubleshooting

- Missing columns: Cleaners create blank columns if a source column is not found (tolerant header matching may still fail if columns are renamed drastically upstream).
- Wrong semester: Ensure Part_of_Term values are one of 1, 2, FY/Y. Other values pass through verbatim and may appear as "Other" in viewers.
- Year extraction: If SMRARUL_AREA lacks trailing digits and no F/P/FP;digits; pattern is present, Year will be blank.
- **Dept mapping:** Unknown **Dept** codes are left as-is in "School Responsible" and reported in cleaner output.
- CSV encoding: All outputs are UTF-8 with BOM for compatibility with Excel.

Questions and Clarifications

Question 1: Meaning of Part of Term Codes in R27 Data

Context: The R27 raw export includes a field called Part of Term, typically containing one of the values:

- 1
- 2
- FY

In the current dataset (R27_Expanded_202510.csv), all records are for Semester 1. However, within this file, the Part of Term field still includes a mix of values such as 1, 2, and FY.

Current Handling in the Cleaning Process:

- 1 \rightarrow mapped to Semester 1
- ullet 2 ightarrow mapped to Semester 2
- FY (and Y, 1&2, 12) \rightarrow mapped to Semester 1 & 2

This logic assumes that:

- 1. A value of 1 indicates a Semester 1 delivery.
- 2. A value of 2 indicates a Semester 2 delivery.
- 3. A value of FY indicates a full-year delivery (i.e. spans both semesters).

Unresolved Issue: In R27_Expanded_202510, which is a Semester 1 dataset, the appearance of 2 and FY values raises an ambiguity:

- Does a 2 value within a Semester 1 dataset actually refer to a Semester 2 module that is pre-registered or cross-coded?
- Or is FY functionally equivalent to 2 in this context (i.e. both referring to modules delivered across the full academic year but reported under the same term)?

Action Required: This needs clarification from the source of the R27 extract (Registry or Student Records). The review should confirm:

- 1. Whether FY and 2 have distinct operational meanings in Banner or are treated equivalently.
- 2. Whether modules marked as FY in a Semester 1 extract (202510) should be included in Semester 2 reporting (202520) or counted only once.

3. If the current mapping (FY \rightarrow Semester 1&2) should remain or be revised to align with official definitions.

Note: Until clarified, the cleaner maintains the current mapping but flags this as a point for data governance review.

Question 2: Duplicate records with same CRN but different Delivery Type (title embeds delivery marker)

Observation: In the cleaned R27 data, some records share the same CRN but differ only by Delivery Type. The Title field also embeds a delivery marker (e.g. "SEM 1" vs "SEM 2"), which makes otherwise identical rows appear different. Example:

```
CMPU 1018 MAND SEM 1 Mathematics 1
CMPU 1018 MAND SEM 2 Mathematics 1
```

Problem Statement: Are these distinct deliveries that should be retained separately, or are they artefacts of how delivery is encoded (once in Delivery Type and again inside Title) and therefore should be collapsed to a single row per CRN?

Clarifications Requested (from data owners / Registry):

- 1. Is CRN guaranteed to represent a unique delivery instance? If "yes", then multiple rows with the same CRN should not differ in Delivery Type.
- 2. What is the authoritative source of delivery period: Part of Term or Delivery Type? Are both required, or is one canonical?
- 3. Should delivery markers embedded in Title (e.g. "SEM 1", "SEM 2", "MAND") be treated as presentation-only and stripped/normalised, leaving delivery period to structured fields?
- 4. If CRN is unique and authoritative, should rows with the same CRN be de-duplicated by ignoring differences in Delivery Type and title-embedded markers?

Proposed Handling (pending decision):

- Treat CRN as the delivery-level key. For rows sharing the same CRN, collapse duplicates by:
 - Using mode (deterministic tie-break) for categorical fields (Delivery Type, Campus, School Responsible, etc.).
 - Normalising Title by removing embedded delivery markers ("SEM 1", "SEM 2", etc.) so that title reflects the module name only.
 - Asserting a single STUDENT_COUNT per CRN; if multiple counts exist, flag for review.
- If Registry confirms that Delivery Type differences alongside the same CRN are valid (e.g. split-mode delivery under one CRN), retain both records and document the reporting rule.

Risk if unresolved: Double-counting modules/students when Title encodes delivery and Delivery Type also differs for the same CRN.