

# BioR — SUPER Challenge (Ecology): Intertidal data integration with tricks

Tidy + join + QC + hybrid objects (nested data + one model per site)

## Scenario

You sampled intertidal sites with a foundation-species canopy (CANOPY) and nearby open patches (OPEN). Biomass was recorded in a spreadsheet (wide format). Microclimate (temperature and humidity) was logged at high frequency with separate loggers. Microbiome swabs were sequenced and delivered as an ASV table. Your task is to integrate everything into an analysis-ready dataset, create explicit QC flags, and build a hybrid table with one linear model per site.

## Dataset folder

Use this folder (contains all input files):  
/mnt/Hard\_intertidal\_challenge

## Input files

- biomass\_wide.csv (wide biomass; includes ND/dead/blanks; sample\_id separators vary)
- quadrat\_meta.xlsx (Excel; canopy\_cover\_pct may be '45%' or impossible values; sample\_id duplicates)
- logger\_raw.csv (mixed timestamp formats; duplicated lines; drift)
- logger\_map.csv (logger\_id has trailing spaces and case mismatches)
- asv\_table.csv (microbiome long table; samples can appear in multiple runs; some low depth)

## Deliverables

- outputs/clean\_master.csv (sample\_id × day with biomass + meta + microclimate + microbiome proxies)
- outputs/qc\_report.csv (one row per sample\_id with QC flags)
- outputs/site\_models.csv (one row per site: slope, intercept, r<sup>2</sup>, n for biomass ~ microclimate)
- outputs/plots/01\_biomass\_timeseries.png
- outputs/plots/02\_biomass\_vs\_temp.png (or vs humidity)
- outputs/plots/03\_qc\_flags.png
- scripts/super\_challenge\_intertidal.R (your script)

## Tasks

- A) Tidy biomass: pivot\_longer day\_\* -> (day, biomass). Day must be numeric.
  - - Convert biomass to numeric; treat ND/dead/blanks as NA; create qc\_non\_numeric\_biomass flag.
  - - Parse sample\_id into site, habitat, rep. Handle '\_' and '-' separators, mixed case, and R01 vs R1.
- B) Quadrat metadata: read the Excel file; parse canopy\_cover\_pct into numeric; flag impossible values (>100).

- - Detect duplicated sample\_id rows in meta; decide a strategy and document it (qc\_meta\_duplicate).
- C) Microclimate: join logger\_map to logger\_raw; parse timestamps robustly; aggregate mean temp and mean RH per site×habitat×day (days 0,7,14).
- - Join microclimate into biomass using left\_join; also try inner\_join and report row counts and what changes.
- D) Microbiome proxies: combine runs by summing reads per sample\_id×asv\_id; compute library\_size and richness per sample\_id.
- - Flag low depth samples: qc\_low\_depth = library\_size < 1000.
- E) Join all sources into one master table (sample\_id×day). Make missing combos explicit with complete() for days {0,7,14}.
- F) Make a QC report (one row per sample\_id) that summarizes key flags.
- G) Hybrid models: one row per site with nested data and fit = lm(biomass ~ mean\_temp) (or mean\_rh). Extract slope, intercept, r2, n.

### **Tricks to watch for (common failure modes)**

- Joins fail if you do not standardize keys first (sample\_id/logger\_id).
- Timestamp parsing will fail unless you handle multiple formats.
- Meta has percent strings and duplicates; decide how to resolve duplicates (and flag).
- ASV table may contain multiple runs per sample; combine before computing proxies.
- lm() drops NA rows: filter before fitting and report n.