

Inbred Line Genomic Prediction Pipeline

Technical Documentation

Project Type: Genomic Prediction & Cross-Validation for Inbred Lines

Language: R

Core Packages: sommer, parallel, dplyr, tidyverse, arrow, paws, ggplot2

Compute Model: Multi-core parallel processing with shared-memory clusters

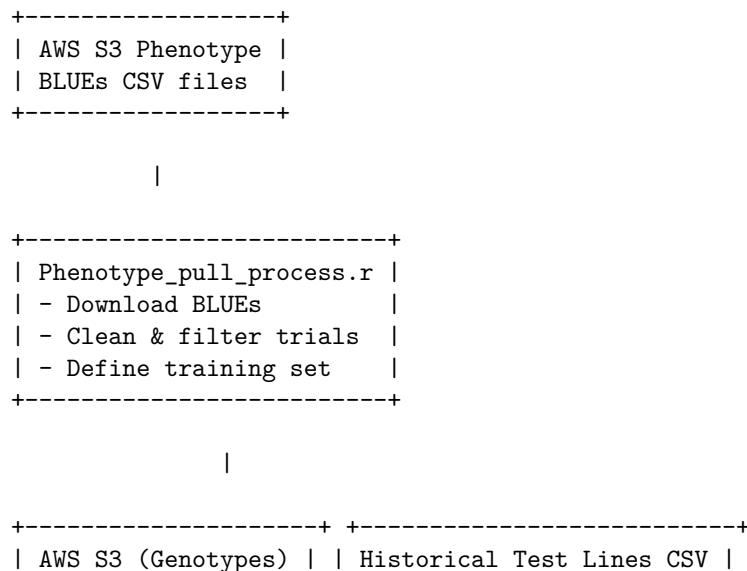
1. Main overview

This pipeline performs **genomic prediction and model evaluation for inbred parental lines** using multi-environment phenotypic data and high-density genotype data. The workflow:

1. Pulls and processes phenotype data from AWS S3
2. Pulls and processes genotype data from AWS S3 (Parquet/VCF-derived)
3. Constructs Genomic Relationship Matrices (GRMs)
4. Fits genomic mixed models (G-only and G×E)
5. Generates Genomic Estimated Breeding Values (GEBVs)
6. Evaluates model performance using:
 - Training-set prediction accuracy
 - 5-fold cross-validation (mid-parent correlation)
7. Produces publication-ready summaries, tables, and plots

The pipeline is modular and scalable, designed for large breeding datasets.

2. Script-Workflow Diagram



```

| Parquet marker data | +-----+
+-----+
|
+-----+
| InbredLine_Genotype_pullprocessss.r |
| - Read Parquet genotypes           |
| - Filter train + test lines       |
| - Build genotype matrix (G)       |
+-----+
|
+-----+
| InbredLine_TraitProcessing&ModelPerformance_.r |
| - GRM construction                 |
| - Model fitting (G, GxE)           |
| - GEBV extraction                  |
| - 5-fold CV                         |
+-----+
|
+-----+
| InbredLine_extrrractresults_n_plotting.r |
| - Accuracy metrics                   |
| - Variance summaries                 |
| - Plots & reports                   |
+-----+
|
+-----+
| Final Outputs (CSV / PDF) |
| - GEBVs                      |
| - Variance components         |
| - CV performance              |
+-----+

```

3. Main Script

InbredLine prediction pipeline.R

Purpose

The **main script** mobilizes the entire analysis. It:

- Loads required packages
- Sources required helper scripts
- Initializes parallel workers
- Executes trait-wise model fitting and cross-validation
- Aggregates results
- Generates final outputs (CSV, PDF)

Script Framework

3.1 Package Loading Core statistical, genomic, and parallelization libraries are loaded: - `sommer` for mixed models - `parallel` for multi-core execution - `tidyverse` and `ggplot2` for data manipulation and plotting

3.2 Source Files The pipeline is split into logically independent source files:

File	Responsibility
<code>Phenotype_pull_process.r</code>	Pull and preprocess phenotypic data
<code>InbredLine_Genotype_pullprocess.r</code>	Pull and preprocess genotype data
<code>InbredLine_TraitProcessing&ModelPerformance_Function.r</code>	GRM construction, modeling, CV
<code>InbredLine_exrractresults_n_plotting.r</code>	Result aggregation and visualization

This design improves maintainability and reproducibility.

3.3 Parallel Execution Two independent clusters are created:

- **CV Cluster (`cl_cv`)**: Used for 5-fold cross-validation
- **GEBV Cluster (`cl_gebv`)**: Used for full-data model fitting and GEBV extraction

Each cluster: - Loads required libraries - Receives shared objects via `clusterExport()`

Trait-level parallelism is achieved using `parLapply()`.

3.4 Result Collection and Cleanup

- Results are collected using `mcollect()`
- Clusters are explicitly stopped to free resources

3.5 Output Generation The main script: - Combines BLUPs and variance components - Writes GEBVs in wide format (one column per trait) - Produces model-wise CSV summaries - Generates PDF reports for trait correlations and CV performance

4. Source File Documentation

4.1 Phenotype Pull & Processing

File: `Phenotype pull process.r`

Purpose

This script: - Connects to AWS S3 - Downloads BLUEs (Best Linear Unbiased Estimates) - Cleans and formats phenotypic data - Defines the training population

Key Steps

AWS Connection

- Uses `paws::s3()` with credential-based authentication
- Pulls phenotype CSV files from structured S3 directories

Data Assembly

- Merges BLUEs
- Writes a local cached phenotype file

Phenotype Cleaning

- Converts trait columns to numeric
- Defines environments as `location × year`
- Filters:
 - Relevant trials
 - Trusted QC status
 - Target locations

Training Set Definition

- Extracts parental inbred lines from hybrid crosses
- Removes non-inbred or reference lines

Key Outputs: - `Pheno_final` - `TrainIDs`

4.2 Genotype Pull & Processing

File: InbredLine Genotype pullprocess.r

Purpose

This module:
- Reads genome-wide marker data from AWS S3
- Filters samples to training and test lines
- Constructs a clean genotype matrix

Key Steps

Genotype Data Access

- Reads Parquet-partitioned genotype files using `arrow::open_dataset()`
- Efficiently collects only relevant samples

Sample Selection

- Combines:
 - Training parents
 - Historical test lines

Marker Cleaning

- Retains numeric marker columns only
- Removes samples with excessive missing data
- Deduplicates samples (least-missing rule)

Genotype Matrix Construction

- Rows = genotypes
- Columns = markers
- Output matrix is numeric and model-ready

Key Output: - G (genotype matrix)

4.3 GRM Construction & Model Functions

File: InbredLine TraitProcessingModelPerformance Function.r

Purpose

This is the **core analytical engine** of the pipeline. It:
- Builds GRMs - Fits genomic mixed models -
Performs GEBV prediction - Runs 5-fold cross-validation

Key Components

GRM Construction

- Marker-centered genotype matrix
- Uses `sommer::A.mat()`
- Separate GRMs for:
 - Full dataset
 - Training-only CV

Supported Models

Model	Description
G-only	Genetic main effects
G×E	Genetic + genotype-by-environment interaction

Trait Processing (`process_trait()`) For each trait:

1. Builds trait-specific phenotype table
2. Fits all models using `mmes()`
3. Extracts:
 - BLUPs
 - GEBVs plus SEs
 - Variance components
 - Heritability

Robust fitting includes fallback strategies for singularities.

Parallel Workers

- `process_trait_worker()` for full data
- `process_trait_worker_cv()` for 5-fold CV

Shared objects are attached/detached to avoid environment pollution.

Cross-Validation Logic

- Hybrid-level masking
- Mid-parent GEBV calculation
- Pearson correlation with observed hybrid means

Key Outputs: - Trait-wise BLUPs - GEBVs plus SEs - Variance components - Fold-level CV correlations

4.4 Result Aggregation & Visualization

File: InbredLine extractresults n plotting.r

Purpose

This module summarizes and visualizes model performance.

Functions

`analyze_results()`

- Merges observed and predicted values
- Computes trait-wise correlations
- Produces faceted scatter plots

`analyze_5foldcv_results()`

- Aggregates fold-level CV results
- Generates:
 - Mean CV bar plots
 - Fold-wise comparison plots

Output Formats

- CSV summaries
 - Publication-ready PDF plots
-

5. Outputs Summary

Output Type	Description
GEBV Tables	Predicted inbred line values
Variance CSVs	Genetic & residual variance
Heritability	Trait × model summaries
Correlation Plots	Observed vs predicted
CV Reports	Fold & mean CV accuracy

Env Correlations per trait

GxEmodel Inbredline5FoldCV results

Trait Correlations AcrossEnvironmentMeans

6. Intended Use

This pipeline is intended for: - Breeding program - generation advancement selection decision support - Parental selection - Model benchmarking (G vs G×E) - Research and production-scale genomic prediction

Author: Vinay Kumar Reddy NANNURU **Status:** Production-ready **Maintenance:** Scalable, configurable, and expandable