Burden-Testing Pipeline

Overview

This repository provides a **comprehensive**, **modular R-based pipeline** (burden_testing_pipeline.R) for **gene- and region-level burden testing** in case—control cohorts. Designed with flexibility in mind, the pipeline supports multiple statistical methods (SKAT, SKAT-O, CAST, CMC, ACAT) while integrating **frequency filtering**, **functional annotations**, **and population-structure correction**. Outputs include detailed TSV/Excel tables, QC visualizations, and execution logs, enabling transparent and reproducible analyses.

Why modularity matters: Each major component—frequency filtering, variant annotation, burden test execution, and result summarization—is implemented as an independent function or script section. This approach allows users to swap in new tests or filtering thresholds without modifying core logic, and to reuse modules in other projects.

Pipeline Modularity & Key Steps

The pipeline is organized into clearly defined modules, each responsible for a distinct task. The main script (burden testing pipeline.R) orchestrates these modules in the following order:

1. Input Parsing & Setup

- Load command-line arguments (VCF path, phenotype file, gene annotation, MAF threshold, selected methods, output prefix).
- Validate that required files exist and the output directory is writable.
- Initialize logging (timestamped logs capture runtime, warnings, and errors).

2. Variant Filtering

- a. Internal MAF Filtering (using data.table):
- Read the cohort VCF (cases + controls) and extract allele counts.
- Exclude any variant with MAF > specified threshold (e.g., 0.01).
 b. External Frequency (gnomAD) Filtering (Optional):
- If --gnomad-filter TRUE, load data/gnomad.vcf.gz.
- Use bcftools (if installed) or SeqVarTools to annotate cohort VCF with external frequencies.
- Remove variants exceeding external MAF threshold.

3. Functional Annotation

- **Gene Overlap Annotation:** Use GenomicRanges to map each variant to gene coordinates (provided in data/genes.gtf). Annotate gene symbols, gene biotypes, and transcript IDs.
- Predicted Deleteriousness: Integrate variant-level annotations (e.g., LOF, missense, splice-site) from VEP or precomputed tables. Assign a binary or weighted "deleterious" flag.
- Clinical Database Tags: (Optional) Tag variants present in ClinVar or other local annotation files—adds context for downstream interpretation.

4. Phenotype & Covariate Processing

- Read data/phenotypes.tsv, which must include columns: SampleID, Status (1=case, 0=control), PC1, PC2, ..., and any additional covariates (e.g., age, sex).
- Verify matching sample IDs between genotype and phenotype files.

■ Compute or load a Genetic Relationship Matrix (GRM) if --1mm TRUE to account for relatedness.

5. Burden Test Execution

For each gene or region (defined in data/genes.gtf):

- a. **Generate Genotype Matrix** (using SeqVarTools): Extract genotypes for all filtered variants within the gene/region.
- b. Construct Covariate Matrix: Include intercept, PC1, PC2, ..., and any user-specified covariates.
- c. Run Selected Tests: Depending on --method, invoke:
- **SKAT / SKAT-O**: Kernel-based test suited for rare variant aggregation.
- CMC: Collapsing approach that groups rare variants within a region and tests the burden.
- CAST: Counts aggregated variant alleles per individual and tests via logistic regression.
- ACAT: Combines p-values across multiple tests or regions using the Cauchy method.
- Mixed-Model (LMM): When --1mm TRUE, use GENESIS or 1me4 to adjust for relatedness via a GRM-based random effect.
 - d. **Store Results:** For each test, record gene/region name, test statistic, p-value, effect size (if applicable), and number of variants tested.

6. Multiple Testing & Significance

- Combine p-values across all genes/regions.
- Apply false discovery rate (FDR) correction (Benjamini-Hochberg) and/or Bonferroni correction.
- Classify genes/regions as Significant (e.g., FDR < 0.05) or Suggestive (nominal p < 0.01).

7. Result Summarization & QC Plots

- TSV/Excel Report: Use openxlsx::write.xlsx() to generate a workbook containing:
 - a. **Burden_Results** Sheet: All genes/regions with columns: Gene, Test, PValue, EffectSize, NumVariants, FDR, SignificanceStatus.
 - b. **Summary_Stats** sheet: Counts of significant genes per test, distribution of p-values.
- QC Plots (via ggplot2):
 - PCA Plot: Scatterplot of PC1 vs. PC2, colored by Status (case/control). Ensures no major stratification.
 - QQ Plot: Observed vs. Expected –log10(p-values) to assess inflation.
 - **Histogram of Variant Counts:** Distribution of number of variants per gene tested.
 - Volcano Plot (optional): -log10(p-value) vs. effect size for top hits.

8. Logging & Cleanup

- Write a comprehensive log file (results/<prefix>_log.txt) capturing:
 - Start/end times, runtime per module, memory usage.
 - Number of variants filtered at each step (internal vs. external).
 - Any warnings or errors encountered.
- Save session info (R version, package versions) at the end for reproducibility.

Prerequisites

R ≥ **4.0** (tested on R 4.1+)

Required R packages:

```
install.packages(c("data.table", "dplyr", "optparse", "openxlsx", "ggplot2", "lme4"))
if (!requireNamespace("BiocManager", quietly = TRUE)) install.packages("BiocManager")
BiocManager::install(c("SKAT", "SeqVarTools", "GenomicRanges", "GENESIS", "gaston"))
```

- Optional:
 - PLINK/PLINK2 (for genotype conversions)
 - bcftools (for variant filtering/export)
 - SLURM / HPC environment (for large-scale runs)

Input Files & Structure

1. data/cohort.vcf.gz

- Combined VCF of case and control samples (indexed with .tbi).
- Ensure FORMAT fields: GT (genotype), AD (allele depth) if performing allele count filtering.

2. data/phenotypes.tsv

- Tab-delimited table with header; columns must include:
 - SampleID: Unique sample identifiers matching VCF sample IDs.
 - Status: 1 = Case, 0 = Control.
 - PC1, PC2, ..., PCn: Principal components for population structure (optional).
 - Additional covariates (e.g., Sex, Age).

3. data/genes.gtf

- Gene annotation file (GTF format) containing gene coordinates.
- The pipeline extracts gene boundaries to define collapsing regions.

4. data/gnomad.vcf.gz (Optional)

- Reference VCF with allele frequencies from gnomAD or genoMAD.
- Use for external MAF filtering if --gnomad-filter TRUE.

5. Directory Expectations

```
Burden-Testing-Pipeline/
  - docs/
    ☐ README.pdf
                              # Full documentation (this file)
  - data/
      - cohort.vcf.gz
       cohort.vcf.gz.tbi
                             # index
      - phenotypes.tsv
       genes.gtf
     — gnomad.vcf.gz
                               # optional external frequency reference
   scripts/
      - burden_testing_pipeline.R
   results/
                               # auto-created after running (> 1 per run)
  README.md
                               # higher-level quick-start guide
```

Quick Start

1. Clone & navigate

git clone https://github.com/<YOUR-USERNAME>/Burden-Testing-Pipeline.git
cd Burden-Testing-Pipeline

2. Install R dependencies

```
install.packages(c("data.table","dplyr","optparse","openxlsx","ggplot2","lme4"))
if (!requireNamespace("BiocManager", quietly = TRUE)) install.packages("BiocManager")
BiocManager::install(c("SKAT","SeqVarTools","GenomicRanges","GENESIS","gaston"))
```

3. Prepare inputs

- Place cohort.vcf.gz (and its index cohort.vcf.gz.tbi), phenotypes.tsv, genes.gtf under data/.
- If using external frequency filtering, place gnomad.vcf.gz (and its index gnomad.vcf.gz.tbi) in data/.

4. Run the pipeline

```
Rscript scripts/burden_testing_pipeline.R \
    --vcf data/cohort.vcf.gz \
    --pheno data/phenotypes.tsv \
    --genes data/genes.gtf \
    --maf 0.01 \
    --method SKAT,SKATO,CAST,CMC,ACAT \
    --gnomad-filter TRUE \
    --covars PC1,PC2,Sex \
    --lmm TRUE \
    --grm data/grm.Rds \
    --out results/analysis1
```

■ This will generate results/analysis1_burden_results.tsv, results/analysis1_QC_plots.pdf, and results/analysis1 log.txt.

5. Inspect results

- Open results/analysis1_burden_results.tsv in a spreadsheet or R to review p-values, effect sizes, and significance.
- View results/analysis1_QC_plots.pdf for PCA and QQ plots to assess quality.
- Check results/analysis1 log.txt for runtime metrics and warnings.

Usage Notes & Customization

Adding New Tests:

- 1. Open scripts/burden testing pipeline.R.
- 2. Under the Run Tests section, add a new function to implement your desired test (e.g., Madsen-Browning).
- 3. Append the test name to the --method argument when running.

Adjusting Frequency Filters:

- Internal MAF: Change --maf <value> (e.g., --maf 0.005).
- External gnomAD: Use --gnomad-filter TRUE; ensure data/gnomad.vcf.gz is provided and indexed.

Population-Structure Correction:

- By default, the pipeline uses PC1 and PC2 from phenotypes.tsv.
- To include more or fewer PCs, set --covars accordingly (e.g., --covars PC1, PC2, PC3).

Mixed-Model Extension (LMM):

- Use --1mm TRUE to enable relatedness correction via a Genetic Relationship Matrix (GRM).
- Provide --grm <path to GRM.rds> (constructed via external tools or within R).

Output Directory Customization:

- By default, outputs go to results/<prefix>_*.
- To override, set OUTDIR prior to running:

```
export OUTDIR="/my/custom/path/analysis1"
Rscript scripts/burden_testing_pipeline.R --vcf ... --out $OUTDIR
```

SLURM/HPC Integration:

Create a Slurm job script (e.g., burden job.sbatch) containing:

Submit with:

```
sbatch burden_job.sbatch
```

Full Methods & Scoring Criteria

Germline Filtering (Case-Control Comparison)

The first stage ensures that only rare, potentially pathogenic variants proceed to burden testing:

1. Load VCF & Compute Allele Counts:

- Use SeqVarTools or data.table to extract allele counts per variant across all samples.
- Calculate Minor Allele Frequency (MAF) = min(AC/AN, 1-AC/AN).

2. Internal MAF Filtering:

Exclude variants with MAF > threshold (e.g., 0.01).

3. External gnomAD Filtering (Optional):

- Annotate cohort VCF with allele frequencies from data/gnomad.vcf.gz.
- Remove any variant with gnomAD MAF > threshold.

Functional Annotation

Adds biological context to each variant:

1. Gene Overlap Annotation:

- Import data/genes.gtf into GenomicRanges.
- For each variant, find overlapping gene(s) and assign gene symbols.

2. Predicted Deleteriousness:

- Integrate variant consequence annotations (e.g., via VEP).
- Classify variants as LOF, missense, or other.

3. Clinical Database Tagging (Optional):

• If local ClinVar or other annotation files are present, tag variants with known clinical significance.

Phenotype & Covariate Processing

Ensures robust population correction:

- Load phenotypes.tsv: Must contain SampleID, Status, PC1, PC2, etc.
- Verify Matches: Check that sample IDs in VCF match those in phenotype file.
- Compute Additional PCs (Optional): If fewer/more PCs needed, run PCA on genotype data (not implemented by default).
- GRM Construction (Optional): If --1mm TRUE, build or load a GRM (.rds) using gaston or other tools.

Statistical Tests

For each gene/region, a suite of tests is performed on the filtered, annotated variant set:

- SKAT / SKAT-O: Optimal sequence kernel association test using SKAT package.
- CMC: Collapsing test grouping rare variants; results from a logistic regression on burden counts.
- CAST: Binary burden test counting alleles per individual.
- ACAT: Combines p-values across tests or regions through a Cauchy combination.
- Mixed-Model (LMM): Adjust for relatedness using GENESIS or 1me4, requiring GRM.

Multiple Testing Correction & Classification

- 1. Combine p-values across all genes for each test.
- 2. Adjust for multiple comparisons:
 - FDR (Benjamini-Hochberg) for exploratory discovery.
 - **Bonferroni correction** for stringent genome-wide significance.
- 3. Classify Significance: Label genes as Significant (e.g., FDR < 0.05) or Suggestive (nominal p < 0.01).

Result Summarization & QC Plots

After test execution, the pipeline produces:

- Excel Workbook (.xlsx) with two sheets:
 - Burden_Results: Detailed results table with Gene, Test, NumVariants, PValue, EffectSize, FDR, SignificanceStatus.
 - 2. Summary Stats: Aggregated counts of significant hits per test and distribution summaries.
- QC Plots:

- PCA Plot (ggplot2): Displaying PC1 vs. PC2 colored by case/control status.
- QQ Plot: Observed vs. expected –log10(p-values).
- Variant Count Histogram: Distribution of variant counts per gene.
- Volcano Plot (Optional): Highlighting top significant genes by effect size and p-value.

Logging & Reproducibility

- Log File (results/<prefix>_log.txt): Tracks timeline of each module, memory usage, number of variants filtered, and any errors.
- Session Info: Captured at script end, including R version and package versions for reproducibility.

Example Outputs

results/analysis1_burden_results.tsv (TSV format):

```
GeneTestNumVariantsPValueEffectSize FDRSignificanceStatusBRCA1SKAT122.0e-061.525.3e-04SignificantTP53SKAT-O91.0e-041.321.2e-02Suggestive..................
```

results/analysis1_QC_plots.pdf: Contains:

- PCA Plot: Samples plotted by PC1 and PC2, colored by case vs. control.
- QQ Plot: Observed vs. Expected –log10(p-values).
- **Histogram:** Distribution of number of variants tested per gene.

results/analysis1_log.txt:

```
[2025-06-05 14:22:10] Starting burden_testing_pipeline.R
[2025-06-05 14:22:12] Input files verified, cohort.vcf.gz and phenotypes.tsv loaded.
[2025-06-05 14:22:15] Internal MAF filtering: 250,000 variants -> 10,000 rare variants.
[2025-06-05 14:22:25] External gnomAD filtering (MAF < 0.01): 10,000 -> 8,500 variants.
[2025-06-05 14:22:30] Functional annotation completed; 8,500 variants annotated.
[2025-06-05 14:22:35] Loaded phenotype file; 500 cases, 500 controls.
[2025-06-05 14:22:40] Running SKAT: completed 20,000 genes in 120 seconds.
[2025-06-05 14:24:40] Running SKAT-O: completed 20,000 genes in 115 seconds.
...
[2025-06-05 14:32:10] Writing Excel report: results/analysisl_burden_results.xlsx.
[2025-06-05 14:32:15] Generating QC plots.
[2025-06-05 14:32:30] Completed. Total runtime: 600 seconds.
```

Credits & Contact

Author: Sally L. Yepes Torres

■ Email: sallyepes233@gmail.com

Last Updated: June 2025

License: MIT