QTL Analysis Pipeline - Use Cases Guide

Quick Start Examples



Use Case: Standard expression QTL analysis with VCF genotypes

bash

- # 1. Edit main config with your file paths
- nano config/config.yaml
- # 2. Update these paths in config.yaml:
- input_files:
- genotypes: "your_data/genotypes.vcf.gz"
- covariates: "your_data/covariates.txt"
- annotations: "your_data/annotations.bed"
- expression: "your_data/expression.txt"
- # 3. Run eQTL analysis

python run_QTLPipeline.py --config config/config.yaml --analysis-types eqtl

Expected Output: results/qtl_results/eqtl_significant.txt with significant eQTL associations



Use Case 1: GATK4.5 Joint Calling VCF

Scenario

You have genotype data from GATK4.5 joint calling in VCF.GZ format with sample names like sample_001_L001.

Configuration

bash

- # Use the GATK-optimized configuration
- cp config/gatk_joint_calling.yaml config/my_gatk_analysis.yaml

•

• # Edit with your paths

```
nano config/my_gatk_analysis.yaml
```

Key Settings for GATK:

yaml

- genotype_processing:
- normalize_chromosomes: true
- chromosome_prefix: "chr" # Converts "1" → "chr1"
- handle_multiallelic: true # Splits multi-allelic sites
- remove_phasing: true # Removes phasing information

Run Command

bash

python run_QTLPipeline.py --config config/my_gatk_analysis.yaml

Expected Processing

- Automatic format detection
- Chromosome naming normalization
- Multi-allelic site splitting
- V Phasing removal
- V Quality filtering

Use Case 2: PLINK BED Format

Scenario

You have genotype data in PLINK BED format (.bed, .bim, .fam files).

Configuration

bash

• cp config/plink_format.yaml config/my_plink_analysis.yaml

```
nano config/my_plink_analysis.yaml
```

Key Settings for PLINK:

yaml

- input_files:
- genotypes: "your_data/plink_data.bed" # .bim and .fam must exist
- _
- genotype_processing:
- auto_detect_format: true # Automatically converts to VCF

```
chromosome_prefix: "none" # PLINK uses "1" not "chr1"
```

Run Command

bash

```
python run_QTLPipeline.py --config config/my_plink_analysis.yaml
```

Expected Processing

- ✓ Automatic BED → VCF conversion
- V Sample information from .fam file
- Variant information from .bim file

Standardized output format



Scenario

You want to run eQTL, pQTL, and sQTL analyses together to compare molecular QTLs.

Configuration

```
bash
```

• cp config/multi_omics.yaml config/my_multiomics.yaml

```
nano config/my_multiomics.yaml
```

Key Settings for Multi-omics:

yaml

```
input_files:
    expression: "data/expression.txt"
    protein: "data/protein.txt"
    splicing: "data/splicing.txt"

analysis:
    qtl_types: "all"  # Runs all three analyses

plotting:
    plot_types:
        - "manhattan"
        - "qq"

"summary"  # Cross-analysis comparison plot
```

Run Command

```
bash
```

```
python run_QTLPipeline.py --config config/my_multiomics.yaml
```

Expected Outputs

- eqtl_significant.txt Expression QTLs
- pqtl_significant.txt Protein QTLs
- sqtl_significant.txt Splicing QTLs
- analysis_summary.png Comparison of all three

Use Case 4: GWAS + QTL Integration

Scenario

You want to run both QTL analysis and GWAS on the same genotype data.

Configuration

bash

• cp config/multi_omics.yaml config/my_gwas_integration.yaml

```
nano config/my_gwas_integration.yaml
```

Key Settings for GWAS:

yaml

```
    analysis:
    qtl_types: "eqtl" # Run eQTL analysis
    run_gwas: true # Enable GWAS
    gwas_phenotype: "data/gwas_phenotypes.txt"
    gwas:
    method: "linear" # For continuous traits
```

method: "logistic"# For case-control traits

```
covariates: true # Adjust for covariates
```

GWAS Phenotype File Format:

text

```
sample_id height weight disease_status
sample1 175.2 68.5 0
sample2 168.7 72.1 1
```

sample3 182.3 75.8 0

Run Command

bash

python run_QTLPipeline.py --config config/my_gwas_integration.yaml

Expected Outputs

- QTL results in qtl_results/
- GWAS results in gwas_results/
- Combined plots showing both analyses

Use Case 5: Large-Scale Study

Scenario

You have thousands of samples and need to optimize for computational efficiency.

Configuration

```
bash
```

• cp config/large_scale_eqtl.yaml config/my_large_study.yaml

```
nano config/my_large_study.yaml
```

Optimization Settings:

```
yaml
```

```
• qtl:
• permutations: 1000  # Reduced for efficiency (default: 1000)
• # permutations: 100  # For very large studies
• output:
• remove_intermediate: true  # Save disk space
• compression: true  # Compress large files
• genotype_processing:
• filter_variants: true
• min_maf: 0.01  # Filter rare variants to reduce load
min_call_rate: 0.95
```

Run Command

bash

- # Test with subset first
- python run_QTLPipeline.py --config config/my_large_study.yaml
 --analysis-types eqtl
- •
- # Full run

```
python run_QTLPipeline.py --config config/my_large_study.yaml
```

Performance Tips

- Start with fewer permutations for testing
- Use --validate-only first to check inputs

- Monitor memory usage in large runs
- Consider running on HPC cluster for very large datasets

Use Case 6: Fine-Mapping Analysis

Scenario

You want high-resolution QTL mapping with strict quality controls.

Configuration

```
bash
```

• cp config/fine_mapping.yaml config/my_finemapping.yaml

```
nano config/my_finemapping.yaml
```

Fine-Mapping Settings:

```
yaml
```

```
• qtl:
• cis_window: 500000  # Smaller window for fine-mapping
• permutations: 10000  # More permutations for accuracy
• fdr_threshold: 0.01  # Stricter significance threshold
• genotype_processing:
• min_maf: 0.01  # Include rare variants
• min_call_rate: 0.99  # High quality requirement
quality_threshold: 50  # Strict quality filter
```

Run Command

bash

```
python run_QTLPipeline.py --config config/my_finemapping.yaml
```

Expected Results

- Higher precision QTL mapping
- More accurate FDR estimates
- Better variant-gene pairing

Use Case 7: Tissue-Specific Analysis

Scenario

You have data from specific tissues and want tissue-specific QTLs.

Configuration

```
# Start with main config and modify:
    qtl:
        cis_window: 500000  # Smaller window for tissue-specific

        genotype_processing:
        filter_variants: true
        min_maf: 0.05  # Higher MAF for power in smaller samples
        e
        analysis:

qtl_types: "eqtl"  # Focus on expression
```

Run Command

bash

python run_QTLPipeline.py --config config/config.yaml --analysis-types eqtl



X Troubleshooting Common Scenarios

Sample Name Mismatch

Problem: Samples in genotype file don't match phenotype file Solution:

yaml

```
genotype_processing:
```

match_samples: true

```
sample_matching_method: "prefix" # or "pattern"
```

Memory Issues

Problem: Pipeline runs out of memory

Solution:

yaml

• qtl:

permutations: 100 # Reduce permutations

• output:

```
remove_intermediate: true # Clean up files
```

Long Runtime

Problem: Analysis taking too long

Solution:

yaml

• qtl:

permutations: 1000 # Reduce from defaultcis_window: 500000 # Smaller window

• genotype_processing:

min_maf: 0.05 # Filter more variants

@ Quick Reference Commands

Use Case	Command
Basic eQTL	<pre>python run_QTLPipeline.pyconfig config.yamlanalysis-types eqtl</pre>
GATK VCF	<pre>python run_QTLPipeline.pyconfig gatk_joint_calling.yaml</pre>
PLINK BED	<pre>python run_QTLPipeline.pyconfig plink_format.yaml</pre>
Multi-omic s	<pre>python run_QTLPipeline.pyconfig multi_omics.yaml</pre>

With GWAS	<pre>python run_QTLPipeline.pyconfig config.yamlrun-gwas</pre>
Validate Only	<pre>python run_QTLPipeline.pyconfig config.yamlvalidate-only</pre>
Large Study	<pre>python run_QTLPipeline.pyconfig large_scale_eqtl.yaml</pre>