Introduction to CoxMK

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2025-08-15

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1 Introduction

The CoxMK package provides a comprehensive toolkit for performing survival analysis using the Cox proportional hazards model combined with Model-X knockoffs. This methodology allows for robust variable selection in high-dimensional settings, such as genetic association studies, while controlling the false discovery rate (FDR).

The key features of the package include:

- Knockoff Generation: Efficiently create knockoff variables for large-scale genetic data.
- Association Analysis: Fit Cox models for both original and knockoff variables.
- Variable Selection: Calculate W statistics and apply the knockoff filter to select significant variables.
- Data Handling: Utilities for loading PLINK data and managing knockoffs with the GDS file format.

2 Installation

You can install the development version of CoxMK from GitHub with:

```
# install.packages("devtools")
devtools::install_github("xiaoxiandadada/Cox-MK")
```

3 The CoxMK Workflow

This section walks through a complete analysis workflow using the example data included in the package.

3.1 Step 1: Load Data

First, we load the required packages and the sample data. The package includes example PLINK files, phenotype data, and covariate data.

```
library(CoxMK)

# Define path to external data
extdata_path <- system.file("extdata", package = "CoxMK")

# Load PLINK data (genotypes and positions)
plink_data <- load_plink_data(file.path(extdata_path, "sample"))

# Load phenotype and covariate data
phenotype_data <- read.table(file.path(extdata_path, "tte_phenotype.txt"), header = TRUE)
covariate_data <- read.table(file.path(extdata_path, "covariates.txt"), header = TRUE)</pre>
```

3.2 Step 2: Create Knockoffs

Next, we generate knockoff variables from the original genotypes. The create_knockoffs function handles this, and by default, saves the results to a .gds file for efficient storage and reuse.

```
knockoffs_result <- create_knockoffs(
  X = plink_data$genotypes,
  pos = plink_data$positions,
  M = 5 # Number of knockoff copies
)</pre>
```

3.3 Step 3: Perform Association Analysis

We then perform Cox regression to test for associations between each variable (both original and knockoff) and the survival outcome.

```
# Prepare merged phenotype and covariate data
pheno_data <- merge(phenotype_data, covariate_data, by = c("FID", "IID"))</pre>
covariates <- pheno data[, c("age", "sex", "bmi")]</pre>
# Fit Cox model for original variables
original_results <- fit_cox_spa(</pre>
 X = plink_data$genotypes,
 time = pheno_data$time,
 status = pheno_data$status,
  covariates = covariates
# Fit Cox model for each knockoff copy
knockoff_results <- lapply(knockoffs_result$knockoffs, function(X_k) {</pre>
 fit_cox_spa(
   X = X_k,
   time = pheno_data$time,
    status = pheno_data$status,
    covariates = covariates
  )
})
```

3.4 Step 4: Calculate W Statistics

The W statistic contrasts the evidence of association for the original variable with that of its knockoff copies.

```
# Extract test statistics (e.g., z-scores) from results
original_stats <- original_results$test_stats
knockoff_stats <- lapply(knockoff_results, function(res) res$test_stats)

# Calculate W statistics
w_stats <- calculate_w_statistics(
    t_orig = original_stats,
    t_knock = knockoff_stats,
    method = "difference"
)</pre>
```

3.5 Step 5: Apply Knockoff Filter

Finally, we apply the knockoff filter to the W statistics to select variables while controlling the FDR at a specified level (e.g., 0.1).

```
selected_snps <- knockoff_filter(w_stats, fdr = 0.1)
# View selected SNP indices
print(selected_snps)</pre>
```

4 Session Information

```
sessionInfo()
#> R version 4.4.2 (2024-10-31)
#> Platform: aarch64-apple-darwin20
#> Running under: macOS Sonoma 14.5
#>
#> Matrix products: default
#> BLAS: /Library/Frameworks/R. framework/Versions/4.4-arm64/Resources/lib/libRblas.0.dylib
#> LAPACK: /Library/Frameworks/R.framework/Versions/4.4-arm64/Resources/lib/libRlapack.dylib; LAPACK v
#> locale:
#> [1] zh_CN.UTF-8/zh_CN.UTF-8/zh_CN.UTF-8/C/zh_CN.UTF-8/zh_CN.UTF-8
#>
#> time zone: Asia/Shanghai
#> tzcode source: internal
#> attached base packages:
base
#> loaded via a namespace (and not attached):
\# [1] compiler_4.4.2 bookdown_0.42 fastmap_1.2.0
                                                       cli_3.6.3
#> [5] tools_4.4.2 htmltools_0.5.8.1 yaml_2.3.10
                                                       rmarkdown 2.29
#> [9] knitr 1.49
                     xfun_0.50
                                     digest\_0.6.37
                                                      rlang_1.1.4
#> [13] evaluate_1.0.3
```