EWAS-fusion

Epigenomewide association statistics (EWAS) for Functional Summary-based Imputation (FUSION) association and joint/conditional analyses

Transcriptomewide association statistic (TWAS) was originally proposed for gene expression data. For a given Trait of interest T for which GWAS summary statistics z_T is available, the corresponding Wald statistic for TWAS is defined such that

$$z_{\mathrm{TWAS}} = \frac{w'_{\mathrm{ge}} z_T}{\sqrt{w'_{\mathrm{ge}} V w_{\mathrm{ge}}}}$$

where w_{ge} is a weight associated with gene expression and V covariance matrix for z_T, respectively. By analogy, an epigenomewide association statistic (EWAS) is defined through methylation data so that

$$z_{\mathrm{EWAS}} = \frac{w'_{\mathrm{me}} z_T}{\sqrt{w'_{\mathrm{me}} V w_{\mathrm{me}}}}$$

where w_{me} is the weight associated with methylation. Both approaches allow for imputation using GWAS summary statistics. The derivation of these weights and imputation were done using methods as described in Gusev et al. (2016) called TWAS as well as in Mancuso et al. (2016) called Functional Summary-based Imputation (FUSION). The TWAS statistics from both approaches agreed very well.

A total of 442,920 CpG sites based on Illumina humanmethylation450 chips on 1,.146 individuals in EPIC-Norfolk study were available. Among these, 1,117 individuals also had genotype data from Affymetrix BioBank Axiom chips. HapMap2 SNPs from genetic data of these individuals were extracted via PLINK2 according to cis-positions of each probe and subsequently used to build weight analogous to gene expression data as implemented in computer software TWAS. We filtered probes according to their heritabilities estimated from software GCTA at significant level of 0.01. We then performed EWAS for given GWAS summary statistics. The weight generation and methylation imputation was implemented in software called TWAS-pipeline, which allows for whole epigenome computation. After filtering, 78,133 probes reached significant level 0.01.

The FUSION framework has several advantages: First, it integrates heritability estimation and covariate adjustment for whole-chromosomes with additional models such as LASSO, elastic net, BLUP. Second, it offers cross-validation, joint/conditional analyses with the output also informing top hit SNPs and inferred methylation quantitative trait locus (meQTL). Besides, the new software uses modified GCTA software (gcta_nr_robust) leading to higher yield of probes with heritabilities reaching

statistical significance, GEMMA giving BSLMM estimates and ability to align strands with reference panels. As both the increased number of models and cross-validation led to excessive computing time, we dropped BSLMM models and conducted five cross-validations. As a result our reference panel for EWAS imputation contains 79,569 probes reaching the heritability p value threshold of 0.01. The association as well as joint/conditional analysis using our weights and LD panel is implemented in software called EWAS-fusion. Like the original TWAS, our implementation will enable a range of GWAS summary statistics to be used coupled with downstream analysis.

Requirements

To begin, the software FUSION including dependencies such as `plink2R` and `reshape` is required. The latest version also requires jlimR. Other facilities to be required are

1. Sun grid engine (sge) or GNU parallel for Linux clusters.
2. Weight files based on epigenetic data.

FILE	Description
EWAS/	directory for EWAS weights
EWAS.bim	SNP information file
glist-hg19	Probe list
LDREF/	Reference for LD
RDat.pos	Definition of regions
RDat.profile*	Probe profiles

`*` It contains information about the probes but not directly involved in the association analysis. For annotation of the results, it is assumed that `HumanMethylation450_15017482_v1-2.csv` is available from the directory containing `ewas-annotate.R`.

Input

The input file contains GWAS summary statistics similar to .sumstats as in LDSC with the following columns.

Column	Name	Description
1	SNP	RS id of SNPs
2	A1	Effect allele (first allele)
3	A2	Other allele (second allele)
4	Z	Z-scores, taking sign with repect to A1

Use of the programs

```
ewas-fusion.sh input-file
```

These will send jobs to the Linux clusters. The sge error and output, if any, should be called EWAS.e and EWAS.o in your HOME directory.

Output

The results will be in input-file.tmp/ directory.

Annotation

This is furnished with contribution from Dr Alexia Cardona, alexia.cardona@mrc-epid.cam.ac.uk, as follows,

```
ewas-annotate.R input-file.tmp
```

This reads `HumanMethylation450_15017482_v1-2.csv` from directory containing `ewas-annotate.R` but this can be at different location

```
ewas-annotate.R input-file.tmp manifest_location=/at/different/location
```

Q-Q and Manhattan plots using R/gap can be obtained from

```
ewas-plot.R input-file.tmp
```

Example

The script test.sh uses height data from GIANT. It downloads and generates an input file called `height` to `ewas-fusion.sh`.

```
ewas-fusion.sh height
```

The results will be in `height.tmp/` once it is done.

The annotation is done with

```
ewas-annotate.R height.tmp
```

The Q-Q and Manhattan plots are generated with

```
ewas-plot.R height.tmp
```

Acknowledgements

We wish to thank colleagues and collaborators for their invaluable contributions to make this work possible.

References

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Mancuso N, et al. (2017). Integrating gene expression with summary association statistics to identify susceptibility genes for 30 complex traits. American Journal of Human Genetics, 2017, 100, 473-487, http://www.cell.com/ajhg/fulltext/S0002-9297(17)30032-0.

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Wood AR, et al. (2014). Defining the role of common variation in the genomic and

biological architecture of adult human height (2014). Nature Genetics 46, 1173-1186.

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Appendix

Additional information for Illumina infinium humanmethylation450 beadchip as in Illumina website

Column Name	Description
Index	Probe Index
TargetID	Identifies the probe name. Also used as a key column for data import.
ProbeID_A	Illumina identifier for probe sequence A
ProbeID_B	Illumina identifier for probe sequence B
IlmnID	Unique CpG locus identifier from the Illumina CG database
Name	Unique CpG locus identifier from the Illumina CG database
AddressA_ID	Address of probe A
AlleleA_ProbeSeq	Sequence for probe A
AddressB_ID	Address of probe B
AlleleB_ProbeSeq	Sequence for probe B
Infinium_Design_Type	Defines Assay type - Infinium I or Infinium II
Next_Base	Base added at SBE step - Infinium I assays only
Color_Channel	Color of the incorporated baseá (Red or Green) - Infinium I assays only
Forward_Sequence	Sequence (in 5'-3' orientation) flanking query site

Column Name	Description
Genome_Build	Genome build on which forward sequence is based
CHR	Chromosome - genome build 37
MAPINFO	Coordinates - genome build 37
SourceSeq	Unconverted design sequence
Chromosome_36	Chromosome - genome build 36
Coordinate_36	Coordinates - genome build 36
Strand	Design strand
Probe_SNPs	Assays with SNPs present within probe >10bp from query site
Probe_SNPs_10	Assays with SNPs present within probe ?10bp from query site (HM27 carryover or recently discovered)
Random_Loci	Loci which were chosen randomly in the design proccess
Methyl27_Loci	Present or absent on HumanMethylation27 array
UCSC_RefGene_Name	Gene name (UCSC)
UCSC_RefGene_Accession	Accession number (UCSC)
UCSC_RefGene_Group	Gene region feature category (UCSC)
UCSC_CpG_Islands_Name	CpG island name (UCSC)
Relation_to_UCSC_CpG_Island	Relationship to Canonical CpG Island: Shores - 0-2 kb from CpG island; Shelves - 2-4 kb from CpG island.
Phantom	FANTOM-derived promoter
DMR	Differentially methylated region (experimentally determined)
Enhancer	Enhancer element (informatically-determined)

Column Name	Description
HMM_Island	Hidden Markov Model Island
Regulatory_Feature_Name	Regulatory feature (informatically determined)
Regulatory_Feature_Group	Regulatory feature category
DHS	DNAse hypersensitive site (experimentally determined)