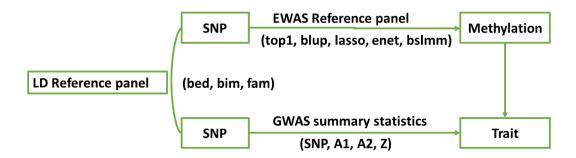
EWAS-fusion

Epigenomewide association study (EWAS) and Functional Summary-based Imputation (FUSION) association analyses



Methylation reference panel, LD reference panel and GWAS summary statistics

Transcriptomewide association statistic (z_{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

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

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

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

where $w_{\rm me}$ 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 Sun grid engine (sge) or GNU parallel for Linux clusters and weights based on epigenetic data.

FILE Description

EWAS-weights/ directory for EWAS weights

glist-hg19 Probe list

LDREF/ Reference for LD RDat.pos Definition of regions

RDat.profile* Probe profiles

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

^{*} It contains information about the probes but not directly involved in the association analysis.

Use of the programs

The syntax is as follows,

```
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
```

It is assumed that HumanMethylation450_15017482_v1-2.csv is available from the directory containing ewas-annotate. Rbut 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
```

Weight generation

This is a revised and much simplified implementation of codes available from TWASpipeline. Under our sge it is furnished with

```
qsub get_weight.qsub
```

qsub get_weight.qsub 22

for chromosome 22.

Inputs to these are summarised as follows,

File Description

FUSION.pheno PLINK phenotype file containing data for all probes
FUSION.covar PLINK covariate file containing covariates such as PCs
CpG.txt CpG ID, missing data indicator, chromosome and position

In addition, PLINK binary pedigree file for each CpG also requires to be prepared, as in files. Although it was not done, it is possible to use code as in 1KG.sh to get around gerneration of these individual files by using a combined one. Note the setup takes advantage of the compact storage of non-genetic data.

The results will be available from the weights directory to be profiled and used for association analysis above. As the number of files is fairly large, cp_weight.qsub is written to put weights from their temporary directories in place.

Acknowledgements

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References

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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
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_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)

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)