

# traseR: TRait-Associated SNP EnRichment analyses

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Abstract

This vignette introduces the use of traseR (TRait-Associated SNPEnRichment analyses, which is designed to provide quantitative assessment whether a selected genomic interval(s) is likely to be functionally connected with certain traits or diseases. traseR consists of several modules, all written in R, to perform hypothesis testing, exploration and visualization of trait-associated SNPs(taSNPs). It also assembles the up-to-date taSNPs from dbGaP and NHGRI, SNPs from 1000 Genome Project CEU population with linkage disequilibrium greater than 0.8 within 100 kb of taSNPs, and all SNPs of CEU population from 1000 Genome project into its built-in database, which could be directly loaded when performing analyses.

1 Introduction

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Genome-wide association study (GWAS) have successfully identified many sequence variants that are significantly associated with common diseases and traits. Tens of thousands of such trait-associated SNPs have already been cataloged which we believe are great resources for genomic research. However, no tools existing utilizes those resources in a comprehensive and convenient way. In this study, we show the collection of taSNPs can be exploited to indicate whether a query genomic interval(s) is likely to be functionally connected with certain traits or diseases. A R Bioconductor package named traseR has been developed to carry out such analyses.

2 Data collection

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One great feature of traseR is the built-in database that collects various public SNP resources. Common public SNP databases include Association Result Browser and 1000 Genome Project. We briefly introduce the procedures to process those public available SNP resources

## 2.1 Obtain taSNPs

Association Results Browser ([http://www.ncbi.nlm.nih.gov/projects/gapplusprev/sgap\\_plus.htm](http://www.ncbi.nlm.nih.gov/projects/gapplusprev/sgap_plus.htm)) combines identified taSNPs from dbGaP and NHGRI, which together provide 44,078 SNP-trait associations, 48,936 SNP-trait class associations, 30,553 unique taSNPs, 573 unique traits and 33 unique trait classes. This resource has been built into GRanges object `taSNP` and could be loaded into R console by typing `data(taSNP)`.

traseR need to specify the collection of trait-associated SNPs in particular format before we carry out enrichment analyses. The format starts with the columns,

1. Trait: Description of disease/trait examined in the study
2. Trait\_Class: Trait class which is formed based on the phenotype tree. Close traits are grouped together to form one class.
3. SNP\_ID: SNP rs number
4. p.value: GWAS reported p-values
5. seqnames: Chromosome number associated with rs number
6. ranges: Chromosomal position, in base pairs, associated with rs number
7. Context: SNP functional class
8. GENE\_NAME: Genes reported to be associated with SNPs
9. GENE\_START: Chromosome start position of genes
10. GENE\_END: Chromosome end position of genes
11. GENE\_STRAND: Chromosome strand associated with SNPs

Currently, the traseR package automatically synchronize trait-associated SNPs from Association Results Browser, which collects up-to-date GWAS results from dbGaP NHGRI GWAS catalog.

## 2.2 Obtain linkage disequilibrium taSNPs from 1000 Genome Project

We first download CEU vcf files from (<ftp://share.sph.umich.edu/1000genomes/fullProject/2012.03.14/>) that contain all sequence variants information. The followed two steps are used to identify linkage disequilibrium SNPs  $>0.8$  and located within 100kb of taSNPs. Firstly, we use `vcftools` to convert the vcf file format to PLINK format. Then we use PLINK to call the LD taSNPs by specifying options that limit the linkage disequilibrium SNPs  $>0.8$  (`-ld-window-r2 0.8`) and within 100kb of taSNP (`-ld-window-kb 100`). The detailed commands are listed below,

```
vcftools -vcf vcf.file -out plink.file -plink plink -file plink.file -r2 -inter-  
chr -ld-snp-list snps.txt -ld-window-r2 0.8 -ld-window-kb 100 -out output.file -  
noweb
```

Finally, we have 90,700 SNP-trait associations and 78,247 unique linkage disequilibrium trait-associated SNP. We also build linkage disequilibrium taSNP into another GRanges object `taSNPLD`, which could be loaded into R console by typing `data(taSNPLD)`.

The format of `taSNPLD` is,

1. seqnames: Chromosome number associated with rs number
2. SNP\_ID: SNP rs number
3. ranges: Chromosomal position, in base pairs, associated with rs number
4. Trait: Description of disease/trait examined in the study
5. Trait\_Class: Trait class which is formed based on the phenotype tree. Close traits are grouped together to form one class.

## 2.3 Obtain background SNPs from 1000 Genome Project

We use the command `plink -file plink.file -freq -out chr` to retrieve all SNPs with corresponding MAF (minor allele frequency) from the CEU vcf files downloaded. There are totally 6,571,512 SNPs ( $MAF > 0.05$ ) excluding variants on Y chromosome. Those SNPs could serve as background in hypothesis testing. We build those SNPs into the built-in GRanges subject `CEU` into the package.

The format of `CEU` is,

1. seqnames: Chromosome number associated with rs number
2. SNP\_ID: SNP rs number
3. ranges: Chromosomal position, in base pairs, associated with rs number

## 3 Using traseR

---

To assess the enrichment level of trait-associated SNPs in given genomic interval(s) using traseR, one needs to follow the simple steps below.

1. Prepare the genomic intervals in R object of either data frame format with column names chr,start,end or GRanges object
2. Query a given a set of genomic interval(s) against all the taSNPs in the collection, perform statistical analyses

3. Explore genes/SNPs of particular interest

## 3.1 Background selection

### 3.1.1 Whole genome

The assumption is each base could be possibly be the taSNP. Based on the assumption, with the number of taSNPs inside and outside the genomic interval(s), the number of bases inside and outside of the genomic interval(s), we could classify all bases based on the fact that one base is taSNP or not and in genomic intervals or not.

### 3.1.2 All SNPs

The assumption is each SNP could possibly be the taSNP. Based on the assumption, with the number of taSNPs inside and outside the genomic interval(s), the non-taSNPs inside and outside of the genomic interval(s), we could classify all SNPs based on the fact that one SNP is taSNP or not and in genomic intervals or not.

## 3.2 Hypothesis testing

traseR provides differential hypothesis testing methods in core function `traseR`, together with other functions for exploring and visualizing the results. The genomic interval(s) could be a data frame with three columns as `chr`(chromosome), `start`(genomic start position) and `end`(genomic end position) or a `GRanges` object. `traseR` offers either including LD SNPs or excluding LD SNPs as the taSNPs and either using the whole genome or all SNPs as the background for hypothesis testing.

If using whole genome as background, the command line is:

```
> x=traseR(snpdb=taSNP,region=Tcell)
> print(x)
```

If including the LD SNPs, the command line is:

```
> x=traseR(snpdb=taSNPLD,region=Tcell)
> print(x)
```

If using all SNPs as background, the command line is:

```
> x=traseR(snpdb=taSNP, region=Tcell, snpdb.bg=CEU)
```

For the above commands, region is the data frame; snpdb is taSNPs or including LD SNPs; snpdb.bg is background SNPs; If rankby is set as "pvalue", all traits will be sorted by p-value in increasing order; ifrankby is set as "odds.ratio", all traits will be sorted by odds ratio in decreasing order. There are four options for test.method including "binomial", "chisq", "fisher", and "nonparametric" to perform binomial test,  $\chi^2$  test, Fisher's exact test and nonparametric respectively. If alternative is set to "greater", traseR will perform hypothesis testing on whether genomic intervals are enriched of taSNPs than the background; If alternative is set to "less", traseR will perform hypothesis testing on whether genomic intervals are depleted of taSNPs than the background.

### 3.2.1 $\chi^2$ test and Fisher's exact test

Based on which background we choose, we could construct the 2 by 2 contingency table. then, we could perform  $\chi^2$  test on the table to assess the difference of proportions of taSNPs inside and outside of genomic intervals(s). We could also assume taSNPs inside genomic intervals follows hypergeometric distribution and calculate p-value directly using Fisher's exact test.

### 3.2.2 Binomial test

The assumption is the probability of observing a single base/SNP being a taSNP is the same inside and outside of genomic intervals. The probability of observing a single base/SNPs being a taSNP in genomic intervals could be estimated by using total number of taSNPs divided by the genome size/number of all SNPs. Then corresponding p-value could be calculated directly by Binomial test.

### 3.2.3 Nonparametric Test

Instead of imposing any assumption, the matched genomic interval(s) are generated by permuting the genomic intervals randomly  $N$  times and overlap with taSNPs in each time. Then we could calculate the empirical p-value directly by counting how many taSNP hits larger/smaller than the observed taSNP hits.

## 4 Choose appropriate statistical test method

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Depending on the characteristics of the test statistics, we suggest to choose appropriate statistical test method under different scenarios,

- $\chi^2$  test: the numbers in the contingency table is fairly large and balanced
- Fisher's exact test: the numbers of the contingency table is relatively small, e.g. no more than 20
- Nonparametric test: the number of query genomic intervals are small, e.g. no more than 1000
- Binomial test: default test method, not limited by sample size, distribution assumption and computational time

## 4.1 Example

To further illustrate the usage of traseR R package, we download H3K4me1 peak regions in peripheral blood T cell from Roadmap Epigenomics. Those peak regions are deemed the genomic intervals. Since the degree of enrichment level is measured by p-value, we could rank traits/trait classes based on p-value in an increasing order. We choose Binomial test as the default option for test.method, use whole genome as background and over-enrichment as hypothesis testing direction.

```
> library(traseR)
> data(taSNP)
> data(Tcell)
> x=traseR(taSNP,Tcell)
> print(x)
```

	Trait	p.value	odds.ratio	taSNP.hits	taSNP.num
1	All	3.788373e-233	2.134717	2625	30553

	Trait	p.value	q.value	odds.ratio
67	Behcet Syndrome	4.400406e-23	2.521433e-20	6.306579
172	Diabetes Mellitus, Type 1	1.704981e-11	4.884769e-09	5.045263
340	Lupus Erythematosus, Systemic	6.159346e-09	1.176435e-06	3.902195
49	Arthritis, Rheumatoid	1.442123e-07	2.065841e-05	5.126637
379	Multiple Sclerosis	1.644125e-05	1.884167e-03	2.905210
62	Autoimmune Diseases	5.201529e-05	4.967461e-03	15.892575

	taSNP.hits	taSNP.num
67	59	274
172	33	185
340	32	223
49	20	112
379	26	236
62	6	15

	Trait_Class	p.value	q.value	odds.ratio
17	Immune System Diseases	3.729169e-35	1.143835e-33	3.658860
31	Skin and Connective Tissue Diseases	6.932335e-35	1.143835e-33	3.916319
32	Stomatognathic Diseases	1.041455e-22	1.145601e-21	5.675922
14	Eye Diseases	3.479491e-18	2.870580e-17	3.313308
11	Digestive System Diseases	4.362324e-14	2.879134e-13	3.040672
7	Cardiovascular Diseases	3.008551e-11	1.654703e-10	1.602762
13	Endocrine System Diseases	6.933337e-09	3.268573e-08	2.068149
24	Nutritional and Metabolic Diseases	4.763509e-08	1.964948e-07	2.068673
21	Musculoskeletal Diseases	3.118359e-05	1.143398e-04	2.716680
23	Nervous System Diseases	5.549981e-05	1.831494e-04	1.495744
16	Hemic and Lymphatic Diseases	1.649261e-04	4.947782e-04	3.596622
22	Neoplasms	3.372076e-04	9.273210e-04	1.580636
30	Respiratory Tract Diseases	6.507770e-04	1.651972e-03	2.121839
	taSNP.hits	taSNP.num		
17	155	1122		
31	142	970		
32	63	318		
14	87	689		
11	74	633		
7	253	3850		
13	89	1076		
24	79	956		
21	27	260		
23	122	1988		
16	15	115		
22	76	1181		
30	29	349		

## 4.2 Exploratory and visualization functions

Plot the distribution of SNP functional class

```
> plotContext(snpdb=taSNP, region=Tcell, keyword="Autoimmune")
```

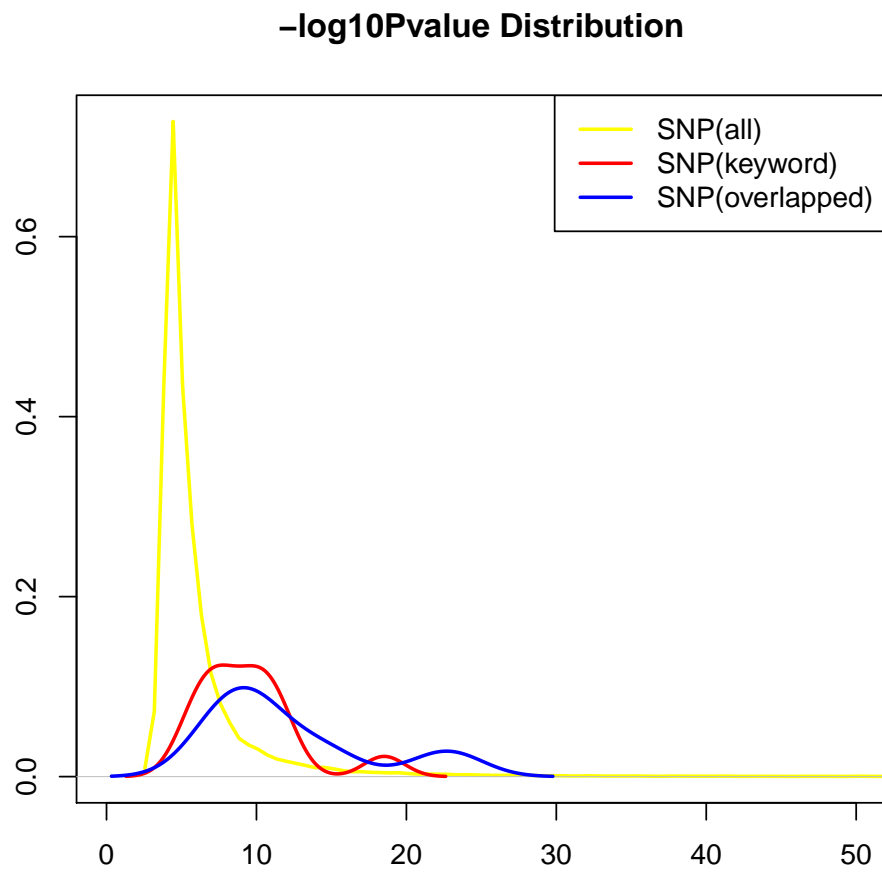


### Pie Chart of Context



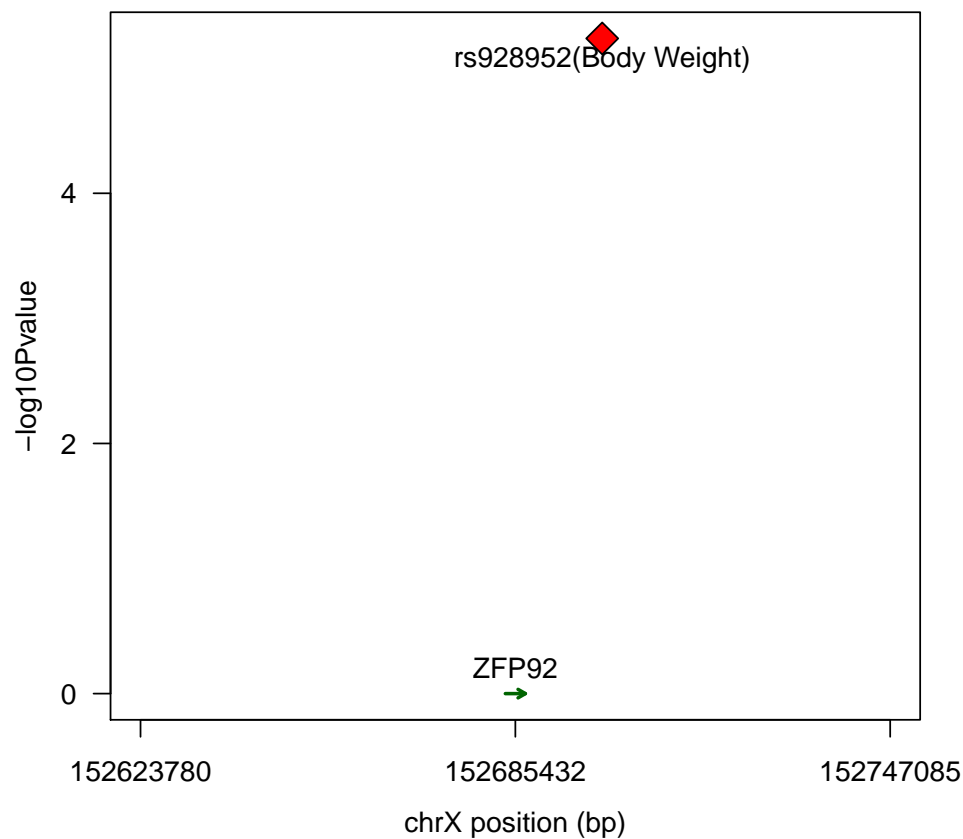
Plot the distribution of p-value of trait-associated SNPs

```
> plotPvalue(snpdb=taSNP, region=Tcell, keyword="autoimmune", plot.type="densityplot")
```



Plot SNPs or genes given genomic interval

```
> plotInterval(snpdb=taSNP, data.frame(chr="chrX", start=152633780, end=152737085))
```



Query trait-associated SNPs by key word,

```
> x=queryKeyword(snpdb=taSNP,region=Tcell,keyword="autoimmune",returnby="SNP")
> head(x)
```

	SNP_ID	Chr	Position	Trait.num	Trait.name
4343	rs11203203	chr21	43836186	1	Autoimmune Diseases
4341	rs1876518	chr2	65608909	1	Autoimmune Diseases
4348	rs1953126	chr9	123640500	1	Autoimmune Diseases
4342	rs2298428	chr22	21982892	1	Autoimmune Diseases
4345	rs7579944	chr2	30445026	1	Autoimmune Diseases
4338	rs864537	chr1	167411384	1	Autoimmune Diseases

Query trait-associated SNPs by gene name,

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```
> x=queryGene(snpdb=taSNP,genes=c("AGRN","UBE2J2","SSU72"))
> x
```

GRanges object with 3 ranges and 5 metadata columns:

	seqnames	ranges	strand	GENE_NAME	Trait.num	Trait.name
	<Rle>	<IRanges>	<Rle>	<factor>	<integer>	<factor>
[1]	chr1	955502-991491	+	AGRN	1	Body Mass Index
[2]	chr1	1477052-1510261	-	SSU72	1	Glucose
[3]	chr1	1189291-1209233	-	UBE2J2	1	Waist Circumference

	taSNP.num	taSNP.name
	<integer>	<factor>
[1]	1	rs3934834
[2]	1	rs880051
[3]	1	rs11804831

-----

seqinfo: 23 sequences from an unspecified genome; no seqlengths

Query trait-associated SNPs by SNP name,

```
> x=querySNP(snpdb=taSNP,snpid=c("rs3766178","rs880051"))
> x
```

GRanges object with 2 ranges and 9 metadata columns:

	seqnames	ranges	strand	Trait	SNP_ID	p.value
	<Rle>	<IRanges>	<Rle>	<character>	<character>	<numeric>
42234	chr1	1478180	*	Glucose	rs3766178	3.26e-05
42127	chr1	1493727	*	Glucose	rs880051	6.44e-05

	Context	GENE_NAME	GENE_START	GENE_END	GENE_STRAND
	<character>	<character>	<integer>	<integer>	<character>
42234	Intron	SSU72	1477052	1510261	-
42127	Intron	SSU72	1477052	1510261	-

	Trait_Class
	<character>
42234	Chemicals and Drugs Category
42127	Chemicals and Drugs Category

-----

seqinfo: 23 sequences from an unspecified genome; no seqlengths

## 5 Conclusion

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traseR provides methods to assess the enrichment level of taSNPs in a given sets of genomic intervals. Moreover, it provides other functionalities to explore and visualize the results.

## 6 Session Info

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```
> sessionInfo()

R version 3.6.1 (2019-07-05)
Platform: x86_64-w64-mingw32/x64 (64-bit)
Running under: Windows Server 2012 R2 x64 (build 9600)

Matrix products: default

locale:
 [1] LC_COLLATE=C
 [2] LC_CTYPE=English_United States.1252
 [3] LC_MONETARY=English_United States.1252
 [4] LC_NUMERIC=C
 [5] LC_TIME=English_United States.1252

attached base packages:
[1] parallel stats4 stats graphics grDevices utils datasets
[8] methods base

other attached packages:
 [1] traseR_1.16.0 BSgenome.Hsapiens.UCSC.hg19_1.4.0
 [3] BSgenome_1.54.0 rtracklayer_1.46.0
 [5] Biostrings_2.54.0 XVector_0.26.0
 [7] GenomicRanges_1.38.0 GenomeInfoDb_1.22.0
 [9] IRanges_2.20.0 S4Vectors_0.24.0
[11] BiocGenerics_0.32.0

loaded via a namespace (and not attached):
 [1] Rcpp_1.0.2 knitr_1.25
 [3] GenomicAlignments_1.22.0 zlibbioc_1.32.0
 [5] BiocParallel_1.20.0 lattice_0.20-38
```

[7] rlang_0.4.1	tools_3.6.1
[9] grid_3.6.1	SummarizedExperiment_1.16.0
[11] Biobase_2.46.0	xfun_0.10
[13] matrixStats_0.55.0	htmltools_0.4.0
[15] yaml_2.2.0	digest_0.6.22
[17] Matrix_1.2-17	GenomeInfoDbData_1.2.2
[19] BiocManager_1.30.9	bitops_1.0-6
[21] RCurl_1.95-4.12	evaluate_0.14
[23] rmarkdown_1.16	DelayedArray_0.12.0
[25] compiler_3.6.1	Rsamtools_2.2.0
[27] XML_3.98-1.20	BiocStyle_2.14.0

## References

- [1] WELTER D, MACARTHUR J, MORALES J, BURDETT T, HALL P, JUNKINS H, KLEMM A, FLICEK P, MANOLIO T, HINDORFF L ET AL (2010). The NHGRI GWAS Catalog, a curated resource of SNP-trait associations. *Nucleic Acid Research*, **42**, D1001-1006.
- [2] ROADMAP EPIGENOMICS C, KUNDAJE A, MEULEMAN W, ERNST J, BILENKY M, YEN A, HERAVI-MOUSSAVI A, KHERADPOUR P, ZHANG Z, WANG J ET AL. (2015). Integrative analysis of 111 reference human epigenomes *Nature*, **7539**, 317-330
- [3] [http://www.ncbi.nlm.nih.gov/projects/gapplusprev/sgap\\_plus.htm](http://www.ncbi.nlm.nih.gov/projects/gapplusprev/sgap_plus.htm) *Association Results Browser*
- [4] <ftp://share.sph.umich.edu/1000genomes/fullProject/2012.03.14/> *1000Genome EUR*