

Survey of available GWAS and gene expression data + TSD overview

Oleksandr Frei









Overview

- What are typical columns in GWAS summary statistics?
- Where to download GWAS results?
- How to re-format GWAS summary statistics tables?

Practical demo – prereq. for tomorrow's tutorial on LDSR

- Download summary statistics
- Install python via Anaconda
- Harmonize column names using Python Pandas package
- Further details in https://etherpad.net/p/gwasOslo

Introduction to TSD service (tjeneste for sensitive data)











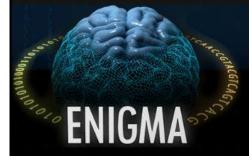


















• PGC Schizophrenia 2014 GWAS

1 40 1	• •	1	2		• •				1
hg19chrc	snpid	a1	a2	bp	info	ог	se	Р	ngt
chr1	rs4951859	C	G	729679	0.631	0.97853	0.0173	0.2083	0
chr1	rs142557973	Τ	C	731718	0.665	1.01949	0.0198	0.3298	0
chr1	rs141242758	Τ	C	734349	0.666	1.02071	0.02	0.3055	0
chr1	rs79010578	Α	Τ	736289	0.649	0.98748	0.0193	0.5132	0
chr1	rs143225517	Τ	C	751756	0.853	0.99681	0.0164	0.8431	0
chr1	rs3094315	Α	G	752566	0.881	0.99601	0.0149	0.787	36
chr1	rs3131972	Α	G	752721	0.846	1.00331	0.0146	0.8229	13
chr1	rs3131971	T	C	752894	0.742	1.01005	0.015	0.5065	0
chr1	rs61770173	Α	C	753405	0.835	0.99631	0.0159	0.8181	0



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PGC Schizophrenia 2014 EUR-only GWAS

CHR	SNP	BP	A1	A2	FRQ_A_33640	FRQ_U_43456	INFO	OR	SE	Р	ngt
10	rs185339560	2392426	Τ	C	0.011	0.011	0.65	1.01339	0.0758	0.8612	0
10	rs11250701	1689546	Α	G	0.640	0.640	0.957	1.01147	0.0117	0.3296	0
10	chr10_2622752_D	2622752	12	D	0.970	0.970	0.933	1.01106	0.0334	0.741	0
10	rs7085086	151476	Α	G	0.322	0.322	0.972	1.02685	0.0118	0.02544	0
10	rs113494187	1593759	Τ	G	0.982	0.982	0.899	0.95285	0.0464	0.298	0
10	rs117915320	1708106	Α	C	0.017	0.017	0.692	1.05580	0.0543	0.3168	0
10	rs182753344	790310	Τ	C	0.082	0.082	0.617	1.02378	0.0236	0.3197	0
10	rs188913771	1273049	Α	G	0.020	0.020	0.656	0.96996	0.0667	0.6473	0
10	rs7911665	2067236	T	G	0.710	0.710	0.925	1.00481	0.0121	0.692	0



• PGC Schizophrenia 2014 GWAS

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chr1	rs4951859	C	G	729679	0.631	0.97853	0.0173	0.2083	0
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10	chr10_2622752_D	2622752	12	D	0.970	0.970	0.933	1.01106	0.0334	0.741	0
10	rs7085086	151476	Α	G	0.322	0.322	0.972	1.02685	0.0118	0.02544	9
10	rs113494187	1593759	Т	G	0.982	0.982	0.899	0.95285	0.0464	U.298	0
10	rs117915320	1708106	Α	C	0.017	0.017	0.692	1.05580	0.0543	0.3168	0
10	rs182753344	790310	Τ	C	0.082	0.082	0.617	1.02378	0.0236	0.3197	0
10	rs188913771	1273049	Α	G	0.020	0.020	0.656	0.96996	0.0667	0.6473	0
10	rs7911665	2067236	T	G	0.710	0.710	0.925	1.00481	0.0121	0.692	0



https://www.ncbi.nlm.nih.gov/snp/?term=rs7085086

rs7085086 [Homo sapiens]

1.

ctgcagcagcacccagggagctcct[A/G]ccccaacttggaaggggtagggctc

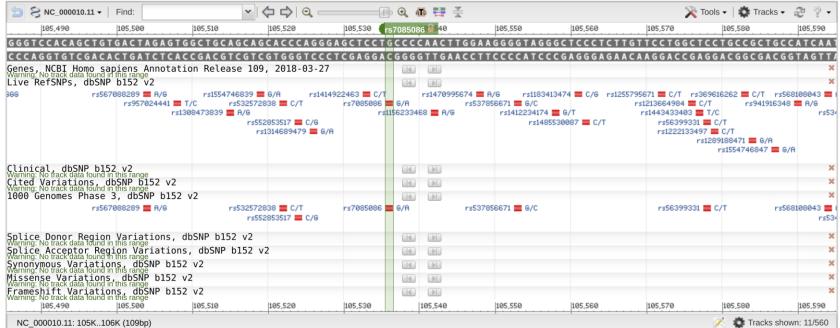
Chromosome: 10:105536

Validated: by 1000G,by 2hit 2allele,by cluster,by frequency

Global MAF: A=0.2370/1187

HGVS: CM000672.2:g.105536G>A, NC_000010.10:g.151476G>A,

NW 003571043.1:g.95536G>A





.... 513010013 1103001011

SSGAC Educational Attainment GWAS

MarkerName	CHR	POS	A1	A2	EAF	Beta	SE	Pval
rs13090388	3	49391082	C	Τ	0.6905	-0.02852	0.00184	4.29e-54
rs7630869	3	49522543	C	Τ	0.6922	-0.02848	0.00184	4.61e-54
rs7623659	3	49414791	Τ	C	0.3095	0.02847	0.00184	4.75e-54
rs11922013	3	49458355	G	C	0.6905	-0.02844	0.00184	5.94e-54
rs9859556	3	49455986	G	Τ	0.6905	-0.02844	0.00184	6.03e-54
rs6779524	3	49450449	C	Τ	0.6905	-0.02843	0.00184	6.30e-54
rs9871380	3	49438221	Α	G	0.3095	0.02841	0.00184	6.68e-54
rs9878943	3	49434654	G	Α	0.6905	-0.02842	0.00184	6.68e-54
rs9814873	3	49454112	G	Α	0.3095	0.02843	0.00184	6.78e-54

README file:

MarkerName: SNP rs number. CHR: chromosome number. POS: base pair position. A1: effect allele.

A2: other allele.

EAF: A1 frequency in 1000 Genomes Phase 3 sample (CEU, GBR and TSI individuals). Beta_*: Standardized regression coefficient, i.e. per-allele effect size on the phenotype that has been standardized to have unit variance.

SE_*: standard error of Beta Pval_*: Nominal p-value of the null hypothesis that the coefficient is equal to zero.



• GIANT Height 2014

MarkerName	Allele1	Allele2	Freq.Allele1.HapMapCEU	h	SE	D	N
	Attetel	ALLCICZ					
rs4747841	Α	G	0.551	-0.0011	0.0029	0.70	253213
rs4749917	T	С	0.436	0.0011	0.0029	0.70	253213
rs737656	Α	G	0.367	-0.0062	0.0030	0.042	253116
rs737657	Α	G	0.358	-0.0062	0.0030	0.041	252156
rs7086391	T	С	0.12	-0.0087	0.0038	0.024	248425
rs878177	T	C	0.3	0.014	0.0031	8.2e-06	251271
rs878178	Α	T	0.644	0.0067	0.0031	0.029	253086
rs12219605	T	G	0.427	0.0011	0.0029	0.70	253213
rs3763688	С	G	0.144	-0.0022	0.0045	0.62	253056

• CTG Cognition 2018

4			_										
SNP	UNIQUE_ID	CHR	POS	A1	A2	EAF_HRC	Zscore	stdBeta	SE	Р	N_analyzed	minINFO	EffectDirection
rs12184267	1:715265	1	715265	t	С	0.0408069	0.916	0.00688729787581148	0.007518884143899	0.3598	225955	0.805386	-?????????????++?
rs12184277	1:715367	1	715367	а	g	0.9589313	-0.656	-0.00491449054466469	0.00749160144003763	0.5116	226215	0.808654	+?????????????
rs12184279	1:717485	1	717485	a	С	0.0405759	1.05	0.00791160346381606	0.00753486044172958	0.2938	226224	0.807189	-????????????++?
rs116801199	1:720381	1	720381	t	g	0.042162	0.3	0.00221740320352237	0.00739134401174123	0.7644	226626	0.805329	-????????????
rs12565286	1:721290	1	721290	С	g	0.0423776	0.566	0.00417421538227414	0.00737493883794018	0.5711	226528	0.812657	-?????????????++?
rs2977670	1:723891	1	723891	С	g	0.93688	-0.253	-0.00154984341088034	0.00612586328411202	0.8006	225312	0.836803	+??????????????
rs28454925	1:726794	1	726794	С	g	0.9590545	-0.539	-0.00403872670010239	0.0074929994436037	0.5896	226782	0.809817	-??????????????
rs116720794	1:729632	1	729632	t	С	0.0410995	0.27	0.00201855807601786	0.00747614102228837	0.7872	226989	0.809294	-?????????????++?
rs4951859	1:729679	1	729679	С	g	0.183523	0.208	0.000805841472291884	0.00387423784755714	0.835	222312	0.725	????????????+-?



Navigate through data sources:

Specific consortia:

- PGC https://www.med.unc.edu/pgc/results-and-downloads
- SSGAC https://www.thessgac.org/data
- CTG https://ctg.cncr.nl/software/summary_statistics
- IBD Genetics https://www.ibdgenetics.org/downloads.html

General catalogs:

- GWAS catalog: https://www.ebi.ac.uk/gwas/
- LDhub: http://ldsc.broadinstitute.org/ldhub/
- UK Biobank: https://biobank.ctsu.ox.ac.uk/showcase/

Reference data:

- 1000 Genomes
- HRC
- dbSNP



1000 Genomes and HRC references

http://www.internationalgenome.org/data/

1000 Genomes Release	Variants	Individuals	Populations	VCF	Alignments	Supporting Data
Phase 3	84.4 million	2504	26	VCF	Alignments	Supporting Data
Phase 1	37.9 million	1092	14	VCF	Alignments	Supporting Data
Pilot	14.8 million	179	4	VCF	Alignments	Supporting Data

https://vcftools.github.io/index.html <- software to work with VCF format</pre>

http://www.haplotype-reference-consortium.org/



Tables manipulation exercise:

Input:

	hg19chrc	snpid	a1	a2	bp	info	or	se	р	ngt
0	chr1	rs4951859	С	G	729679	0.631	0.97853	0.0173	0.2083	0
1	chr1	rs142557973	T	С	731718	0.665	1.01949	0.0198	0.3298	0
2	chr1	rs141242758	Т	С	734349	0.666	1.02071	0.0200	0.3055	0
3	chr1	rs79010578	A	Τ	736289	0.649	0.98748	0.0193	0.5132	0
4	chr1	rs143225517	T	С	751756	0.853	0.99681	0.0164	0.8431	0

Intermediate result:

	CHR	SNP	∆ 1	Z 2	RP	INFO	OR	SE	DI/ZIIIF
	CIII	DIVI	ΔT	AZ	בע	TIVE	OIC		IVALOD
0	chr1	rs4951859	С	G	729679	0.631	0.97853	0.0173	0.2083
1	chr1	rs142557973	Τ	C	731718	0.665	1.01949	0.0198	0.3298
2	chr1	rs141242758	T	С	734349	0.666	1.02071	0.0200	0.3055
3	chr1	rs79010578	A	T	736289	0.649	0.98748	0.0193	0.5132
4	chr1	rs143225517	Т	С	751756	0.853	0.99681	0.0164	0.8431

Final result:

	CHR	SNP	A1	A2	qd	INFO	OR	SE	PVALUE	N	Z
0	1	rs4951859	С	G	729679	0.631	0.97853	0.0173	0.2083	82315 -	-1.258254
1	1	rs142557973	Т	С	731718	0.665	1.01949	0.0198	0.3298	82315	0.974517
2	1	rs141242758	Т	С	734349	0.666	1.02071	0.0200	0.3055	82315	1.024710
3	1	rs79010578	A	Т	736289	0.649	0.98748	0.0193	0.5132	82315 -	-0.653863
4	1	rs143225517	Т	С	751756	0.853	0.99681	0.0164	0.8431	82315 -	-0.197930



```
>>> import pandas as pd
>>> df=pd.read table('ckqny.scz2snpres.gz', delim whitespace=True)
>>> df.head()
  ha19chrc
                 snpid al a2
                                    info
                                 bp
                                                or
                                                       se
                                                                  ngt
                            729679
0
     chr1
           rs4951859 C G
                                    0.631 0.97853 0.0173 0.2083
                                                                    0
                                    0.665 1.01949 0.0198 0.3298
1
     chr1 rs142557973 T C
                            731718
                                                                    0
2
     chr1 rs141242758 T C
                            734349
                                    0.666 1.02071 0.0200
                                                          0.3055
                                                                    ()
3
     chrl rs79010578 A T 736289 0.649 0.98748 0.0193 0.5132
                                                                    \cap
     chr1 rs143225517 T C 751756 0.853 0.99681 0.0164 0.8431
                                                                    0
>>> df.rename(columns={'hq19chrc':'CHR', 'snpid':'SNP', 'a1':'A1', 'a2':'A2',
'or':'OR', 'se':'SE', 'p':'PVALUE', 'bp':'BP', 'info':'INFO'}, inplace=True)
>>> df.drop(columns=['ngt'], inplace=True)
>>> df.head()
   CHR
                SNP A1 A2
                              BP
                                  INFO
                                             OR
                                                    SE PVALUE
0 chr1
        rs4951859 C G 729679 0.631 0.97853 0.0173 0.2083
1 chr1 rs142557973 T C
                         731718 0.665 1.01949 0.0198 0.3298
                         734349 0.666 1.02071 0.0200 0.3055
2 chr1 rs141242758 T C
3 chr1 rs79010578 A T 736289 0.649 0.98748 0.0193 0.5132
4 chr1 rs143225517 T C
                         751756 0.853 0.99681 0.0164 0.8431
>>> df['N'] = 35476+46839
>>> import numpy as np
>>> from scipy import stats
>>> df['Z'] = -stats.norm.ppf(df['PVALUE'].values*0.5)*np.sign(np.log(df['OR'].values))
>>> df['CHR'] = pd.to numeric(df['CHR'].str.replace('chr', ''), errors='coerce')
>>> df.head()
    CHR
                SNP A1 A2
                                  INFO
                                             OR
                                                    SE PVALUE
                                                                   N
                              bp
                         729679 0.631 0.97853 0.0173 0.2083
0
          rs4951859 C G
                                                                82315 -1.258254
                                                       0.3298
1
     1 rs142557973 T C
                         731718 0.665 1.01949 0.0198
                                                                82315 0.974517
                         734349 0.666 1.02071 0.0200 0.3055 82315 1.024710
    1 rs141242758 T C
     1 rs79010578 A T 736289 0.649 0.98748 0.0193 0.5132 82315 -0.653863
     1 rs143225517 T C 751756 0.853 0.99681 0.0164 0.8431
                                                                82315 -0.197930
>>> df = df[~df['CHR'].isnull()].copy()
>>> df['CHR']=df['CHR'].astype(int)
>>> df.to csv('PGC SCZ 2014.csv', index=False, sep='\t')
```

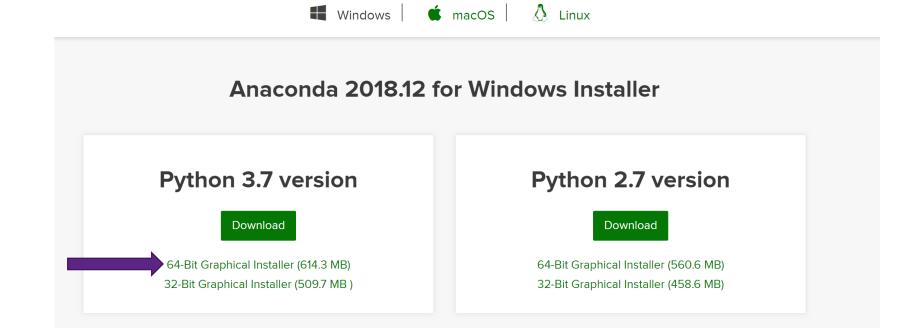


Demo

- Download GWAS data for PGC Schizophrenia GWAS, 2014
 - https://www.med.unc.edu/pgc/results-and-downloads,
 Search for SCZ2, then "Download full SNP results"
- Download and install Anaconda (Python distribution)
 - https://www.anaconda.com/distribution/
 - (!) Ba careful about Mac | Windows | Linux choose your system
 - (!) Choose 64 bit. DO NOT choose 32 bit.

Recommended download: Python 3.7 (but Python 2.7 is also OK)

```
(base) oleksanf@mach:~/github/mixer$ python
Python 3.6.5 |Anaconda, Inc.| (default, Apr 29 2018, 16:14:56)
[GCC 7.2.0] on linux
Type "help", "copyright", "credits" or "license" for more information.
>>>
```





Links

- Consortia-specific GWAS locations
 - PGC https://www.med.unc.edu/pgc/results-and-downloads
 - SSGAC https://www.thessgac.org/data
 - CTG https://ctg.cncr.nl/software/summary_statistics
 - IBD Genetics https://www.ibdgenetics.org/downloads.html
 - B. Neale Lab http://www.nealelab.is/uk-biobank
 - GIANT: Genetic Investigation of ANthropometric Traits -https://portals.broadinstitute.org/collaboration/giant/index.php/giaNT_consortium_data_files
- GWAS catalog: https://www.ebi.ac.uk/gwas/
- LDhub: http://ldsc.broadinstitute.org/ldhub/
- UK Biobank: https://biobank.ctsu.ox.ac.uk/showcase/



TSD introduction

TSD User Guide

Login

Network requirements, passcodes, passwords.

→ More about login

File import and export

Using the TSD file exchange tool, File Lock.

→ More about import/export

 https://www.uio.no/engli sh/services/it/research/ sensitive-data/use-tsd/

Collecting data

Use Nettskjema to collect data directly into TSD.

→ More about Nettskjema

Software

Software packages installed on the servers.

→ More about software

Good documentation!

Computing and analysis

Use High Performance Computing resources in TSD.

→ More about HPC

Administrative tasks

Adding new user, requesting extra resources, database etc...

→ More about administration

Video and Audio in TSD

Store and analyze video and audio files in TSD

→ More about video and audio in TSD

Directory Structure and File Acccess

Structure, shortcuts, file control, backup and recovery, sharing

→ More about files and access

Need help?

→ Contact TSD

All help links

All links to the TSD user guide.

→ See more



TSD Demo

- (3) How to login to Linux machine
- (4) How to login to Windows machine
- (5) How to import data to TSD using file uploader
- (1) How to apply for TSD access for an existing project
- (2) How to reset TSD password (https://selfservice.tsd.usit.no/)

Advanced topics:

- How to extend your TSD project (request disk space, CPU hours, larger login nodes for your project)
- How to import large volumes of data to TSD



P.S. Gene expression data

- https://www.ebi.ac.uk/gxa/home
- http://portal.brain-map.org/
- https://www.proteinatlas.org/humanproteome/tissue/brain
- https://gtexportal.org/home/
- http://hbatlas.org/





NORMENT
Oslo University Hospital HF
Division of Mental Health and Addiction
Psychosis Research Unit/TOP
Ullevål Hospital, building 49
P.O. Box 4956 Nydalen
N-0424 Oslo
Norway