

2023 International Statistical Genetics Workshop

Day 2. Optional morning GWAS practical. CHEATSHEET

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Luke Evans & Wei Zhou

Practical originally developed by Katrina Grasby and Lucía Colodro Conde

This cheatsheet contains the information for the GWAS practical, as well as additional information on using plink2 to perform association tests and basic functions.

Learning Goals: This tutorial aims to make you familiar with genome-wide association analysis, the way in which PLINK2 works (and some of the useful options it provides!), and some tools to visualize your results.

It is important that you explore the outputs and the log files and that you understand what they mean. Please ask if you have any questions, and check the plink file formats page.

Please note, most of the options that you will be using work the same for logistic and linear regression.

Some of the commands below are missing (indicated with XXX). You will have to determine the correct syntax or command to run the analyses, and discuss the questions proposed.

You will find the commands and their output in Day2_GWAS_Practical_CHEATSHEET.pdf.

Feel free to take a look through it during the practical, but a good strategy is to first try the commands yourself or with your group, then troubleshoot things yourself if they don't work, then ask your group, then ask others & faculty, then check the cheatsheet.

The cheatsheet has all the expected answers and figures, but the process of doing the GWAS and exploring the output is what will help you learn.

DAY 2. GWAS TUTORIAL

Key information can be found here:

<https://www.cog-genomics.org/plink/2.0/>

<https://www.cog-genomics.org/plink/2.0/assoc>

<https://www.cog-genomics.org/plink/2.0/formats>

FIRST: Copy the files to your working directory & navigate to that directory:

```
cp -r /faculty/luke/2023/gwas2/ .  
cd gwas2/
```

EXERCISE 1. LOGISTIC REGRESSION (BINARY TRAIT).

1.0. Go to the case-control folder and check the files you have there. The phenotype is Alzheimer's Disease (AD) status.

```
cd casecontrol
```

1.1. Run a logistic regression for the phenotype (AD status) including the principal components in file adpc.txt as covariates to correct for genetic ancestry.

- [hint: you can find the flags and modifiers that you can include in the logistic and linear regressions on Association analysis > Regression with covariates, <https://www.cog-genomics.org/plink/2.0/assoc#glm> in the PLINK 2.0 website]

```
plink2 --bfile adclean.cc --glm sex --covar adpc.txt --out 1.1_logistic.ad.cc --threads 4
```

--bfile <prefix> is the dataset filename you will use

--glm is the main tool to perform associations. Including the 'sex' modifier includes sex (from the fam file) as a covariate

--covar <filename> specifies the file with your covariates to use

--threads <max> This sets the number of threads that the program will use. Generally, more threads makes things faster.

Association analysis

Linear and logistic/Firth regression with covariates

```
--glm ['zs'] ['omit-ref'] [{sex | no-x-sex}] ['log10'] ['pheno-ids']
  [{genotypic | hethom | dominant | recessive | hetonly}] ['interaction']
  ['hide-covar'] ['skip-invalid-pheno'] ['allow-no-covars']
  ['single-prec-cc'] [{intercept | cc-residualize | firth-residualize}]
  [{no-firth | firth-fallback | firth}] ['cols=<col set desc.>]
  ['local-covar=<file>'] ['local-psam=<file>']
  ['local-pos-cols=<key col #s> | 'local-pvar=<file>'] ['local-haps']
  ['local-omit-last' | 'local-cats=<cat. ct> | 'local-cats0=<cat. ct>]
(alikes: --linear, --logistic)
--ci <size>
--condition <variant ID> [{dominant | recessive}] ['multiallelic']
--condition-list <variant ID file> [{dominant | recessive}] ['multiallelic']
--parameters <number(s)/range(s)...>
--tests ['all'] [number(s)/range(s)...]
--vif <max VIF>
--max-corr <val>
```

--glm is PLINK 2.0's primary association analysis command.

For quantitative phenotypes, --glm fits the linear model

$$y = G\beta_G + X\beta_X + e$$

for every variant (one at a time), where y is the phenotype vector, G is the genotype/dosage matrix for the current variant, X is the fixed-covariate matrix, and e is the error term subject to least-squares minimization. (Dosages are always used when present; if you want to analyze hardcalled genotypes instead, run "--make-pgen erase-dosage" first.) X always contains an all-1 intercept column, along with anything loaded by --covar. Missing-dosage rows are excluded, not mean-imputed.

Check the log file. How many cases and controls were detected? How many covariates?

```
less -S 1.1_logistic.ad.cc.log
```

less prints out only the first 10 rows of the file, using the "-S" flag prints it out only one window's worth at a time.

There were 170 cases, 182 controls, 4 covariates detected in the covar file (not including sex, from the fam file)

```
PLINK v2.00a4LM AVX2 (9 Jan 2023)
Options in effect:
  --bfile adclean.cc
  --covar adpc.txt
  --glm sex
  --out 1.1_logistic.ad.cc
  --threads 4

Hostname: ip-10-0-201-238
Working directory: /home/luke/2023/gwas2/casecontrol
Start time: Sun Mar  5 15:56:49 2023

Random number seed: 1678057009
15657 MIB RAM detected; reserving 7828 MIB for main workspace.
Using up to 4 compute threads.
352 samples (164 females, 188 males; 352 founders) loaded from adclean.cc.fam.
297237 variants loaded from adclean.cc.bim.
1 binary phenotype loaded (170 cases, 182 controls).
4 covariates loaded from adpc.txt.
Calculating allele frequencies... done.
--glm logistic-Firth hybrid regression on phenotype 'PHEN01': done.
Results written to 1.1_logistic.ad.cc.PHEN01.glm.logistic.hybrid.

End time: Sun Mar  5 15:57:17 2023
1.1_logistic.ad.cc.log (END)
```

Check the results (stored in 1.1_logistic.ad.cc.PHENO1.glm.logistic.hybrid) and be sure that you understand the content of each of the columns

- [hint: you can check File formats > .glm.logistic, https://www.cog-genomics.org/plink/2.0/formats#glm_logistic, on the PLINK2.0 website]. Remember, when PLINK 1.9 uses bfiles, the effect allele is A1, which is the minor allele, and may or may not be the ALT allele. So an OR > 1 means A1 is associated with an increased risk relative to A2.

head 1.1_logistic.ad.cc.PHENO1.glm.logistic.hybrid
head prints out only the first 10 rows of the file

```
luke@ip-10-0-201-238:~/2023/gwas2/casecontrol$ head 1.1_logistic.ad.cc.PHENO1.glm.logistic.hybrid
#CHROM POS ID REF ALT A1 FIRTH? TEST OBS_CT OR LOG(OR)_SE Z_STAT P ERRCODE
1 752566 rs3094315 T C C N ADD 349 1.35589 0.210165 1.44868 0.147428
1 752566 rs3094315 T C C N PC1 349 1.24498e-48 61.1278 -1.8045 0.0711535
1 752566 rs3094315 T C C N PC2 349 2.6294e+16 59.1399 0.6393 0.522628
1 752566 rs3094315 T C C N PC3 349 1.98209e+21 58.526 0.837892 0.402091
1 752566 rs3094315 T C C N PC4 349 2.93983e-24 54.4944 -0.994299 0.320077
1 752566 rs3094315 T C C N SEX 349 1.17211 0.217716 0.729413 0.465749
1 779322 rs4040617 A G G N ADD 351 1.28062 0.233422 1.05964 0.289311
1 779322 rs4040617 A G G N PC1 351 3.45264e-49 61.142 -1.82506 0.0679926
1 779322 rs4040617 A G G N PC2 351 4.6348e+19 59.0013 0.767487 0.442792
```

Key columns to understand:

- o CHROM: Chromosome of the locus
- o POS: physical basepair position of the locus
- o ID: rsnumber of the variant
- o REF: Reference allele
- o ALT: Alternate allele
- o A1: The allele that is counted for the regression (Not necessarily ALT, but should be the minor allele)
- o OR: A1 allele's odds ratio for the case/control phenotype
- o SE: Standard error of the allele's beta (beta=ln(OR))
- o P: p-value of the test for this locus

.glm.firth, .glm.logistic[hybrid] (logistic/Firth regression association statistics)
Produced by --glm with a case/control phenotype.

A text file with a header line, and then one line per variant with the following columns:

Header	Column set	Contents
CHROM	chrom	Chromosome code
POS	pos	Base-pair coordinate
ID	(required)	Variant ID
REF	ref	Reference allele
ALT1	alt1	Alternate allele 1
ALT	alt	All alternate alleles, comma-separated
A1	(required)	Counted allele ¹ in regression
OMITTED	omitted	Omitted allele
AX	ax	Non-A1 alleles, comma-separated (deprecated)
A1_CT ²	a1count	Total A1 allele count (can be decimal with dosage data)
ALLELE_CT ²	totalallele	Allele observation count
A1_CASE_CT ²	a1countcc	A1 count in cases
A1_CTRL_CT ²	a1countcc	A1 count in controls
CASE_ALLELE_CT ²	totalallecc	Case allele observation count
CTRL_ALLELE_CT ²	totalallecc	Control allele observation count
CASE_NON_A1_CT	gcountcc	Case genotypes with 0 copies of A1
CASE_HET_A1_CT	gcountcc	Case genotypes with 1 copy of A1
CASE_HOM_A1_CT	gcountcc	Case genotypes with 2 copies of A1
CTRL_NON_A1_CT	gcountcc	Control genotypes with 0 copies of A1
CTRL_HET_A1_CT	gcountcc	Control genotypes with 1 copy of A1
CTRL_HOM_A1_CT	gcountcc	Control genotypes with 2 copies of A1
A1_FREQ	a1freq	A1 allele frequency
A1_CASE_FREQ	a1freqcc	A1 allele frequency in cases
A1_CTRL_FREQ	a1freqcc	A1 allele frequency in controls
MACH_R2	machr2	MaCH imputation quality metric
FIRTH?	firth	Reports whether Firth reg. was used ('firth-fallback' only)
TEST	test	Test identifier
OBS_CT	nobs	Number of samples in the regression
BETA	beta	Regression coefficient (for A1 allele)
OR	orbeta	Odds ratio (for A1 allele)
[LOG(OR)]_SE	se	Standard error of log-odds (i.e. beta)
L#	ci	Bottom of symmetric approx. confidence interval (with --ci)
U#	ci	Top of symmetric approx. confidence interval (with --ci)
Z_[OR_F]_STAT	tz	F-statistic for joint test, Wald Z-score for logistic/Firth regression
[LOG10_]P	p	Asymptotic p-value (or -log10(p)) for Z/chisq-stat
ERRCODE	err	When result is 'NA', an error code describing the reason

Did any SNP associations reach genome-wide significance?

```
grep ADD 1.1_logistic.ad.cc.PHENO1.glm.logistic.hybrid |
awk '$13<5e-8'
```

One did: rs4420638, with p=1.886e-14.

What is the grep command doing here?

grep searches for a string in each row of a file.

Here, it's finding all of the "ADD" strings, which is short for ADDITIVE estimate of the allele's effect.

It's useful because by default PLINK writes out all of the covariate effect estimates as well (the PC1-5, SEX lines), which are good to check, but which aren't what we're most interested in.

So, grep ADD <file> pulls out only the SNP effect estimates.

What if cases and controls had been coded as 1 and 0, respectively? What could have we done to make PLINK interpret this coding appropriately?

[hint: you can check Standard data input > Phenotypes > Phenotype encoding, <https://www.cog-genomics.org/plink/2.0/input#pheno>, in the PLINK 2.0 website]

Plink uses 1 & 2 for controls & cases, respectively. You can change this with the --1 command

What if sex had not been coded in the fam file, or was in both your covariate and fam files? What could have we done to make PLINK interpret this coding appropriately?

[hint: you can check --glm modifiers in <https://www.cog-genomics.org/plink/2.0/assoc#glm>, in the PLINK 2.0 website]

You can tell which covariates to use by adding additional commands, such as --covar-col-num or --covar-name.

1.2. Run a logistic regression for the case-control variable AD including the principal components as covariates and hiding the results of the covariates.

```
plink2 --bfile adclean.cc --glm sex hide-covar --covar adpc.txt --out 1.2_adclean.cc
```

What's the difference between the sets of results generated in 1.1 and 1.2?

--glm with the modifier "hide-covar" doesn't print out the estimates for each of the covariates for every single locus, saving a lot of file space. The estimates are probably very nearly identical for every test, anyway.

```
luke@ip-10-0-201-238:~/2023/gwas2/casecontrol$ head 1.2_adclean.cc.PHEN01.glm.logistic.hybrid
#CHROM POS ID REF ALT A1 FIRTH? TEST OBS_CT OR LOG(OR)_SE Z_STAT P ERRCODE
1 752566 rs3094315 T C C N ADD 349 1.35589 0.210165 2.047 1.44868 0.147428
1 779322 rs4040617 A G G N ADD 351 1.28062 0.233422 0.810457 2.02352 1.05964 0.289311
1 1003629 rs4075116 A G G N ADD 351 1.01761 0.163397 0.738753 1.40174 0.106863 0.914898
1 1097335 rs9442385 G T T N ADD 350 0.63612 0.282949 0.365335 1.10761 -1.59876 0.109874
1 1130727 rs10907175 A C C N ADD 343 1.36271 0.284606 0.780094 2.38046 1.08739 0.276866
1 1158631 rs6603781 C T T N ADD 344 1.51238 0.222741 0.977383 2.34023 1.85726 0.0632745
1 1165310 rs11260562 C T T N ADD 352 0.848671 0.373859 0.407859 1.76591 -0.438891 0.66074
1 1211292 rs6685064 C T T N ADD 348 0.840127 0.331242 0.438926 1.60804 -0.525908 0.598952
1 1268847 rs307378 G T T N ADD 349 1.19173 0.474992 0.469745 3.02337 0.369277 0.711921
luke@ip-10-0-201-238:~/2023/gwas2/casecontrol$
```

1.3. Run a logistic regression for the case-control variable AD including the principal components as covariates, hiding the results of the covariates, and getting 95% Confidence intervals of odds ratios.

[hint: you can check again <https://www.cog-genomics.org/plink/2.0/assoc#glm>, in the PLINK 2.0 website]

```
plink2 --bfile adclean.cc --glm sex hide-covar --ci 0.95 --covar adpc.txt --out 1.3_adclean.cc --threads 4
--ci 0.95 tells plink to output the 95% CI of the estimates.
```

```
luke@ip-10-0-201-238:~/2023/gwas2/casecontrol$ head 1.3_adclean.cc.PHEN01.glm.logistic.hybrid
#CHROM POS ID REF ALT A1 FIRTH? TEST OBS_CT OR LOG(OR)_SE L95 U95 Z_STAT P ERRCODE
1 752566 rs3094315 T C C N ADD 349 1.35589 0.210165 0.898117 2.047 1.44868 0.147428
1 779322 rs4040617 A G G N ADD 351 1.28062 0.233422 0.810457 2.02352 1.05964 0.289311
1 1003629 rs4075116 A G G N ADD 351 1.01761 0.163397 0.738753 1.40174 0.106863 0.914898
1 1097335 rs9442385 G T T N ADD 350 0.63612 0.282949 0.365335 1.10761 -1.59876 0.109874
1 1130727 rs10907175 A C C N ADD 343 1.36271 0.284606 0.780094 2.38046 1.08739 0.276866
1 1158631 rs6603781 C T T N ADD 344 1.51238 0.222741 0.977383 2.34023 1.85726 0.0632745
1 1165310 rs11260562 C T T N ADD 352 0.848671 0.373859 0.407859 1.76591 -0.438891 0.66074
1 1211292 rs6685064 C T T N ADD 348 0.840127 0.331242 0.438926 1.60804 -0.525908 0.598952
1 1268847 rs307378 G T T N ADD 349 1.19173 0.474992 0.469745 3.02337 0.369277 0.711921
luke@ip-10-0-201-238:~/2023/gwas2/casecontrol$
```

1.4. Run a logistic regression for the case-control variable AD including the principal components as covariates, hiding the results of the covariates, and getting the allele frequencies.

[hint: you can check Main functions > Basic statistics > --freq or Allele frequency, https://www.cog-genomics.org/plink/1.9/basic_stats#freq, in the PLINK 1.9 website].

```
plink2 --bfile adclean.cc --glm sex hide-covar --covar adpc.txt --freq --out 1.4_adclean.cc --threads 4
--freq creates a second output, *.afreq, which contains the frequency information.
```

Explore the output files and note you have an extra one, 1.4_adclean.cc.afreq. [hint: you can check File formats > .afreq in the PLINK website]. In this one, we can see REF, ALT, and ALT_Freq.

```
luke@ip-10-0-201-238:~/2023/gwas2/casecontrol$ head 1.4_adclean.cc.afreq
#CHROM ID REF ALT ALT_FREQS OBS_CT
1 rs3094315 T C 0.161891 698
1 rs4040617 A G 0.109687 702
1 rs4075116 A G 0.299145 702
1 rs9442385 G T 0.0771429 700
1 rs10907175 A C 0.0758017 686
1 rs6603781 C T 0.133721 688
1 rs11260562 C T 0.046875 704
1 rs6685064 C T 0.058908 696
1 rs307378 G T 0.0243553 698
```

.account, .afreq (allele count/frequency report)

Produced by `--freq`.

A text file with a header line, and then one line per variant with the following columns:

Header	Column set	Contents
CHROM	chrom	Chromosome code
POS	pos	Base-pair coordinate
ID	(required) ref	Variant ID
REF	ref	Reference allele
ALT1	alt1	Alternate allele 1
ALT	alt	All alternate alleles, comma-separated
'REF_FREQ'/'REF_CT'	refreq	Reference allele frequency/dosage
'ALT1_FREQ'/'ALT1_CT'	alt1freq	Alternate allele 1 frequency/dosage
'ALT_FREQS'/'ALT_CTS'	altfreq, alteq, alteqz	Comma-separated freqs/dosages for all alts; 'eq' requests '1=<ALT1 value>,2=<ALT2 value>,...' formatting with zero-values omitted, 'eqz' includes zeroes
'ALT_NUM_(FREQS,CTS)'	altnumeq	Comma-separated freqs/dosages for all alts
'FREQS'/'CTS'	freq, eq, eqz	Comma-separated freqs/dosages for all alleles
'NUM_FREQS'/'NUM_CTS'	numeq	Comma-separated freqs/dosages for all alleles
MACH_R2	machr2	MaCH imputation quality metric
MINIMAC3_R2	minimac3r2	Minimac3 phased-dosage imputation quality metric; inaccurate unless phased dosages were imported with e.g. "--vcf dosage=HDS" (dosage=DS is not enough)
OBS_CT	nobs	Number of allele observations

EXERCISE 2. LINEAR REGRESSION (CONTINUOUS TRAIT)

2.0. Go to the continuous folder and check the files you have there. The phenotype is a transcript probe (gene expression). Pay attention to the file `adclean.cont.txt`.

```
cd ../continuous/
head adclean.cont.txt
```

2.1. Run a linear regression for the continuous trait including the genetic principal components as covariates, hiding the results of the covariates, and using the `--pheno` option. The advantage of using an extra file for phenotypes (and the `--pheno` option) is that if there were several phenotypes, it would be possible to run analyses on them at the same time (something not possible with `ped` or `fam` files). Note that when using the `--pheno` option, the original `ped` or `fam` files can still contain a phenotype in the `phenotypc` column, but it can be missing (`NA/nan/-9`). See Standard data input > Phenotypes, <https://www.cog-genomics.org/plink/2.0/input#pheno>.

```
plink2 --bfile adclean.cont --glm sex hide-covar --pheno adclean.cont.txt --covar adpc.txt --out 2.1_adclean.cont --threads 4
```

What do you notice about the output file name? Because the phenotype file has a column name, it uses that phenotype name in the output by default.

```
luke@ip-10-0-201-238:~/2023/gwas2/continuous$ head 2.1_adclean.cont_GI_34147330-S_glm.linear
#CHROM POS ID REF ALT A1 TEST OBS_CT BETA SE T_STAT P ERRCODE
1 752566 rs3094315 T C C ADD 349 -0.00601159 0.0468227 -0.12839 0.897916
1 779322 rs4040617 A G G ADD 351 0.0137903 0.0520003 0.265196 0.791017
1 1003629 rs4075116 A G G ADD 351 0.0153566 0.0365065 0.420655 0.67427
1 1097335 rs9442385 G T T ADD 350 0.0124305 0.0606909 0.204816 0.837837
1 1130727 rs10907175 A C C ADD 343 -0.0509092 0.0633483 -0.80364 0.422173
1 1158631 rs6603781 C T T ADD 344 -0.035305 0.0490759 -0.719395 0.472396
1 1165310 rs11260562 C T T ADD 352 -0.0607221 0.0832358 -0.729519 0.466179
1 1211292 rs6685064 C T T ADD 348 -0.0702463 0.0737093 -0.953018 0.341256
1 1268847 rs307378 G T T ADD 349 0.0620477 0.105937 0.585704 0.558461
```

Phenotypes

```
--pheno {'iid-only'} <filename>
--pheno-col-nums <l-based column number(s)/range(s)...>
--pheno-name <column ID(s)/range(s)...>
--no-psam-phenotype
(aliaes: --no-phenotype, --no-fam-phenotype)
--not-phenotype <phenotype ID(s)...>
(aliaes: --phenotypeExcludeList)

--pheno causes (additional) phenotype values to be read from the specified space- or tab-delimited file.
The first columns of that file must be either FID/IID or just IID (in which case the FID is assumed to be 0). A
primary header line is required when using --pheno-name, and optional without it (if it's present, it should
begin with 'FID', 'FID', 'IID', or 'IID'). Additional header lines (beginning with '#', not immediately followed
by 'FID'/'IID') are permitted before the primary header line. For example:

## * If you also need PLINK 1.9 to read this file, add an FID column in front,
## and fill it with zeroes.
## * 'site' is a categorical variable. --glm would ignore it if you loaded it
## as a phenotype (multinomial logistic regression is not implemented), but
## it's a valid 'covariate' for --glm.

#IID qt1 bmi site
1110 2.3 22.22 site2
2202 34.12 18.23 site1
...
```

2.2. Run a linear regression for the continuous trait including only PC1 as covariate, hiding the results of the covariate, using the --pheno option.

[hint: again, you can check the commands you can use to run logisic or linear regressions on data input > Covariates, <https://www.cog-genomics.org/plink/2.0/input#pheno>. For options related to how parameters are handled in other models (e.g., estimating non-additive effects), you can see <https://www.cog-genomics.org/plink/2.0/assoc>

```
plink2 --bfile adclean.cont --glm hide-covar --pheno adclean.cont.txt --covar adpc.txt --covar-name PC1
--out 2.2_adclean.cont --threads 4
--covar-name <column name> allows you to specify only certain covariates out of your file. Also, this
does not include sex as a covariate, because this command doesn't have the --glm modifier 'sex'.
```

```
luke@ip-10-0-201-238:~/2023/gwas2/continuous$ head 2.2_adclean.cont.GI_34147330-S.glm.linear
#CHROM POS ID REF ALT A1 TEST OBS_CT BETA SE T_STAT P ERRCODE
1 752566 rs3094315 T C C ADD 349 -0.0133576 0.0471627 -0.283225 0.777174
1 779322 rs4040617 A G G ADD 351 0.0122987 0.052342 0.234968 0.814372
1 1003629 rs4075116 A G G ADD 351 0.0160724 0.0368213 0.436496 0.662748
1 1097335 rs9442385 G T T ADD 350 0.0105939 0.0613877 0.172573 0.863088
1 1130727 rs10907175 A C C ADD 343 -0.0625219 0.0637446 -0.980819 0.32738
1 1158631 rs6603781 C T T ADD 344 -0.0411284 0.0496146 -0.828958 0.407708
1 1165310 rs11260562 C T T ADD 352 -0.0537221 0.0834414 -0.64383 0.520108
1 1211292 rs6685064 C T T ADD 348 -0.0644163 0.0739107 -0.871542 0.384064
1 1268847 rs307378 G T T ADD 349 0.0851028 0.106909 0.796031 0.42656
```

2.3. Run a linear regression for the continuous trait including the principal components as covariates, hiding the results of the covariates, using the --pheno option, and getting 95% confidence intervals for the beta.

```
plink2 --bfile adclean.cont --glm sex hide-covar --pheno adclean.cont.txt --covar adpc.txt --ci 0.95 --out
2.3_adclean.cont --threads 4
similar to above, the --ci 0.95 command outputs the 95% CI.
```

What is different about these results than 2.1 results? How might you use the added information?

If you're interested in a specific variant and its effect, the uncertainty in the effect size may be very important to know, and this is a way to get that quickly. You can also always calculate a CI of varying range yourself using a t-distribution.

```
luke@ip-10-0-201-238:~/2023/gwas2/continuous$ head 2.3_adclean.cont.GI_34147330-S.glm.linear
#CHROM POS ID REF ALT A1 TEST OBS_CT BETA SE L95 U95 T_STAT P ERRCODE
1 752566 rs3094315 T C C ADD 349 -0.00601159 0.0468227 -0.0977825 0.0857593 -0.12839 0.897916
1 779322 rs4040617 A G G ADD 351 0.0137903 0.0520003 -0.0881284 0.115709 0.265196 0.791017
1 1003629 rs4075116 A G G ADD 351 0.0153566 0.0365065 -0.0561948 0.0869081 0.420655 0.67427
1 1097335 rs9442385 G T T ADD 350 0.0124305 0.0606909 -0.106522 0.131383 0.204816 0.837837
1 1130727 rs10907175 A C C ADD 343 -0.0509092 0.0633483 -0.17507 0.0732511 -0.80364 0.422173
1 1158631 rs6603781 C T T ADD 344 -0.035305 0.0490759 -0.131492 0.0608821 -0.719395 0.472396
1 1165310 rs11260562 C T T ADD 352 -0.0607221 0.0832358 -0.223861 0.102417 -0.729519 0.466179
1 1211292 rs6685064 C T T ADD 348 -0.0702463 0.0737093 -0.214714 0.0742213 -0.953018 0.341256
1 1268847 rs307378 G T T ADD 349 0.0620477 0.105937 -0.145585 0.26968 0.585704 0.558461
```

2.4. Plot the results from 2.3.

We will create three plots to explore the results. For that, we will need at least three columns: chromosome, base pair position, and p-value; having a SNP column (containing the rs number) will allow extra options in our plots. We'll **first create a file containing this information, excluding the markers with no results.**

What is "|" doing in the next line? What does the grep command do, and why would you want to use that?

The "|" is the pipe command, which passes the standard output from the command right before it (here, awk) to the standard input of the command after it (here grep).

The grep command uses the '-v' modifier so that it searches for lines WITHOUT "NA" in them. These are lines with missing estimates, and we don't want them, so we grab all the lines that don't have "NA" in them instead.

```
awk '{print $1,$2,$3,$14}' 2.3_adclean.cont.GI_34147330-S.glm.linear | grep -v NA >
plot.adclean.cont.linear.txt
```

```
luke@ip-10-0-201-238:~/2023/gwas2/continuous$ head plot.adclean.cont.linear.txt
#CHROM POS ID P
1 752566 rs3094315 0.897916
1 779322 rs4040617 0.791017
1 1003629 rs4075116 0.67427
1 1097335 rs9442385 0.837837
1 1130727 rs10907175 0.422173
1 1158631 rs6603781 0.472396
1 1165310 rs11260562 0.466179
1 1211292 rs6685064 0.341256
1 1268847 rs307378 0.558461
```

What's the SNP with the lowest p-value (or top SNP)?

```
sort -k4 -g plot.adclean.cont.linear.txt | head
```

What is the "-g" doing in the line above? Hint: use "man sort" to check the documentation of commands.

The "-g" modifier to the sort command sorts the particular column (4th, specified by -k4), as a number, even though it's got a letter in it sometimes (using notation like 5e-8 as 0.00000005).

```
luke@ip-10-0-201-238:~/2023/gwas2/continuous$ sort -k4 -g plot.adclean.cont.linear.txt | head
#CHROM POS ID P
20 25451180 rs449370 5.07922e-39
20 25397257 rs6050598 5.16834e-39
20 25350325 rs4815412 5.50642e-39
20 25275843 rs2258719 7.66462e-39
20 25291848 rs1888999 7.66462e-39
20 25297909 rs2297497 8.71415e-39
20 25366065 rs6050573 9.24724e-39
20 25433821 rs12428 1.214e-38
20 25270339 rs2257991 1.23612e-38
```

EXERCISE 3. PLOT OUT YOUR RESULTS

QQ and Manhattan plots.

Open in RStudio (<https://workshop.colorado.edu/rstudio/>) the script `Rscript_qqMan.R` and plot the results. Set your working directory to the folder where you have your results (you can check first what your working directory is by typing `pwd` in the command line). Run the script. Explore the QQ plot and the Manhattan plot. What information do you gather from these plots and the lambda value? Do you detect any anomalies?

You need to set your working directory first. This is either by specifying the path in the script or by using `Session>Set Working Directory>Choose Directory` in the drop-down menu at the top of RStudio.

You also need to tell it which file to use, here, "plot.adclean.cont.linear.txt".

Note that while this file does have a header column, it starts with a "#" which comments out the line and which RStudio doesn't recognize. So, you then have to set the names of the column.

Checking the structure uses `str(Data)`, and will output information for each column in the "Data" object.

```

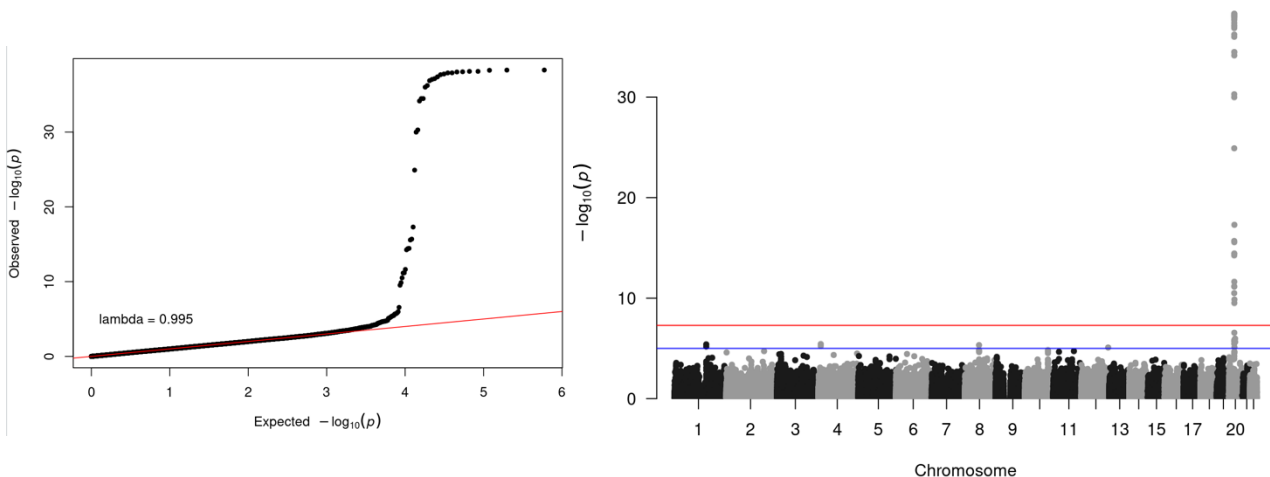
6
7 library(qqman)
8
9 # Indicate working directory
10
11 setwd("~/2023/gwas2/continuous/") # set your working directory; e.g. ~/gwas2/continuous/ or e.
. ~/gwas2/casecontrol/, or use "Session" in the drop-down menu above to set your directory
12
13 Data <- read.table('plot.adclean.cont.linear.txt', header=F) # select you file; e.g. plot
.adclean.cc.logistic.txt or plot.adclean.cont.linear.txt
14 head(Data)
15
16 colnames(Data)<-c("CHROM", "POS", "ID", "P") # give it the right column names
17 head(Data)
18
19
20 # make sure the data is prepared so there is no missing data and the key variables (CHR, BP,
) are numeric.
21 # How do you check the structure of the data?
22 str(Data) # What is the command you can use
23
24:1 (Top Level)

```

```

R 4.2.2 · ~/2023/gwas2/continuous/
· colnames(Data)<-c("CHROM", "POS", "ID", "P") # give it the right column names
· head(Data)
  CHROM    POS      ID      P
1 1 752566 rs3094315 0.897916
2 1 779322 rs4040617 0.791017
3 1 1003629 rs4075116 0.674270
4 1 1097335 rs9442385 0.837837
5 1 1130727 rs10907175 0.422173
6 1 1158631 rs6603781 0.472396
· # make sure the data is prepared so there is no missing data and the key variables (CHR, BP, P) a
numeric.
· # How do you check the structure of the data?
· str(Data) # What is the command you can use
'data.frame': 297237 obs. of 4 variables:
 $ CHROM: int  1 1 1 1 1 1 1 1 1 1 ...
 $ POS : int  752566 779322 1003629 1097335 1130727 1158631 1165310 1211292 1268847 1478153 ...
 $ ID : chr  "rs3094315" "rs4040617" "rs4075116" "rs9442385" ...
 $ P : num  0.898 0.791 0.674 0.838 0.422 ...
· |

```



Regional plot.

Still in RStudio, go to the "Files" section (bottom right quadrant of the browser) and check the file plot.adclean.cont.linear.txt and the two jpg images you just created. Then go to More and then Export. The file will download to your local computer.

You can explore association p-value results and LD patterns by uploading your results (or a genomic region of interest) to LDassoc in LDlink: <https://ldlink.nci.nih.gov/?tab=ldassoc>.

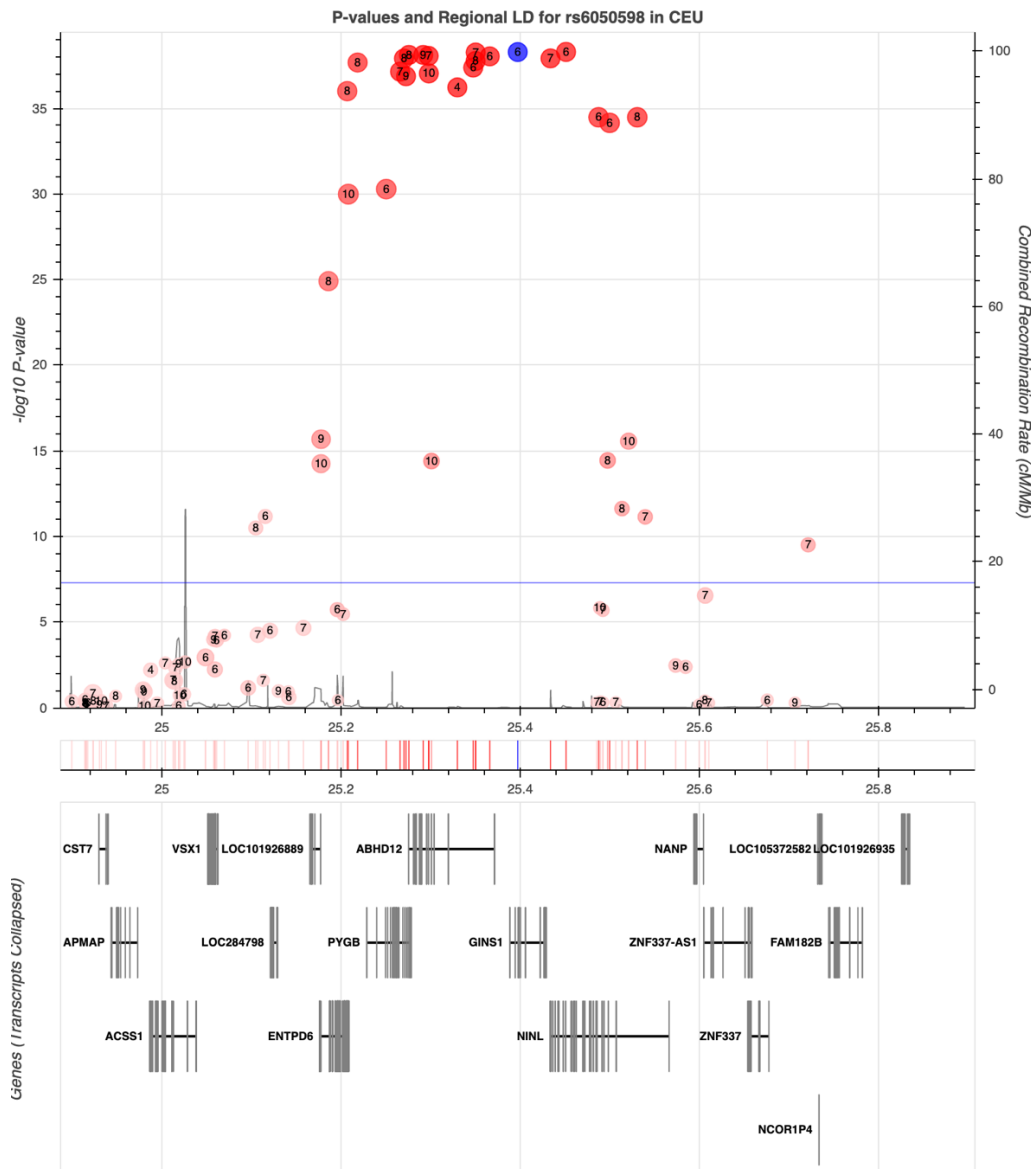
Upload your results using Browser (preferred: Chrome, Safari, Firefox 36+ or Internet Explorer). In real scenarios, you may want to upload only the information of a region of interest that you want to plot or only one chromosome because the upload time will be slow with very large files.

Select the columns that contain the information on Chromosome, Position, and P-Value.

LDassoc has three options for visualising regions of association: by gene, by region or by variant. We will select visualising our results by using our lowest p-value variant as the index variant (rs6050598 or chr20:25397257).

Select the CEU population (European, Utah Residents from North and West Europe) as the 1000 Genomes sub-population of interest. LDlink will calculate measures of linkage disequilibrium according to this population, which is the one that best matches the ancestry of our study population. Leave the rest of the options as default and press Calculate.

Explore the interactive plot. Is the index variant in high LD with any of the nearby variants? Remember the R2 is a measure of linkage disequilibrium or correlation of alleles for two genetic variants.



If you have gotten this far with time left, you can go back to your logistic regression results and plot them:

1.5. Plot the results from 1.4. You will need to use the commands you ran in section two to generate the intermediate files for RStudio and the interactive browser plots. Your challenge will be to adapt those commands to the right files in the right sequence, run the RStudio script again, and create the plots.