GWAS QC -theory and steps
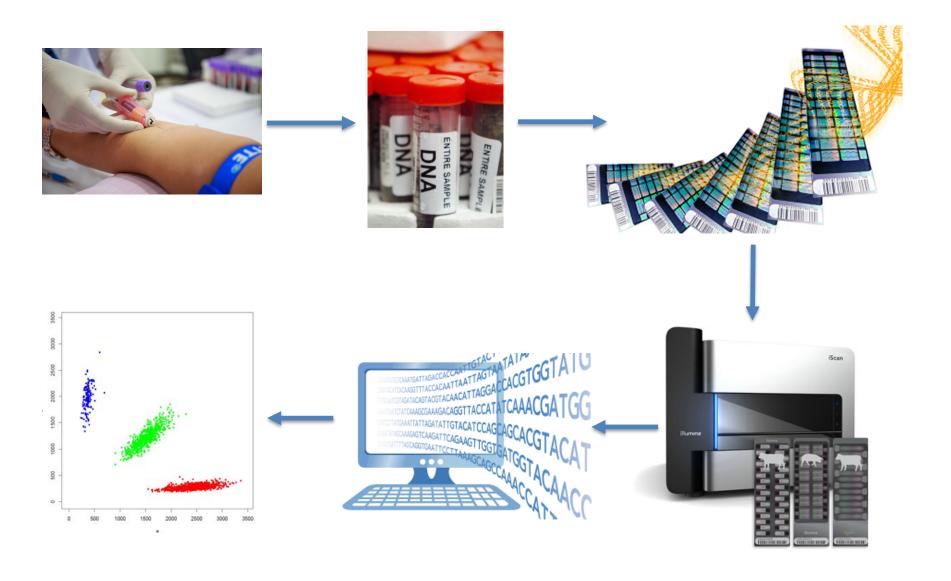
H3ABioNet Genome Wide Association Studies
Lecture Series
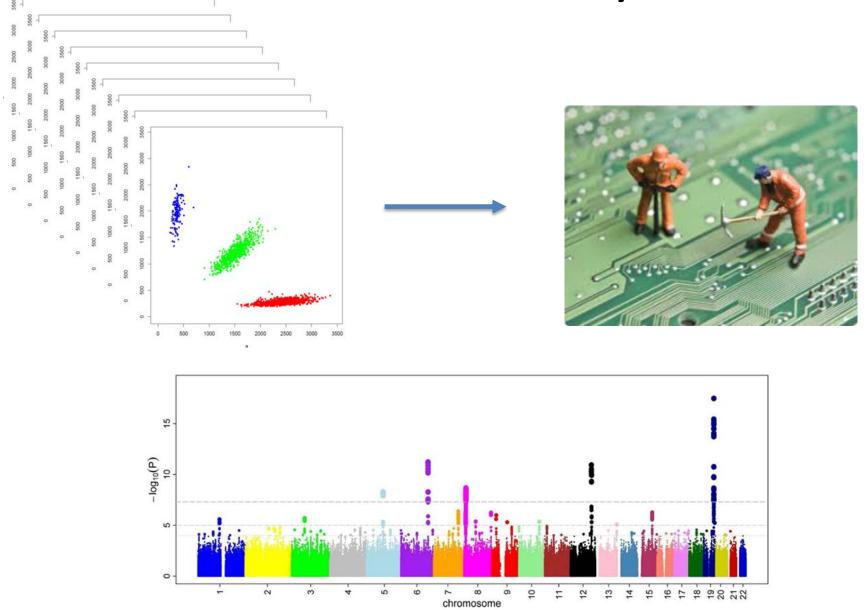
Shaun Aron

August 2018

GWAS Data Generation
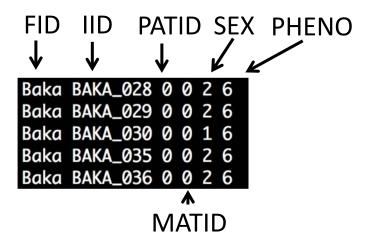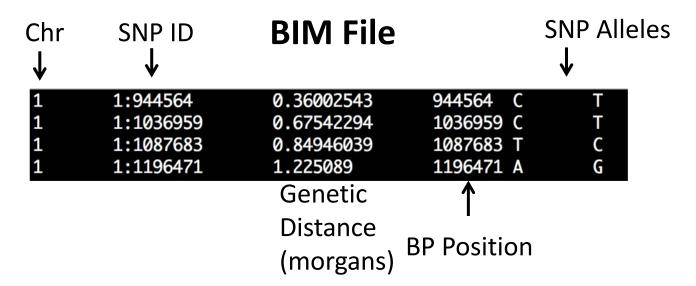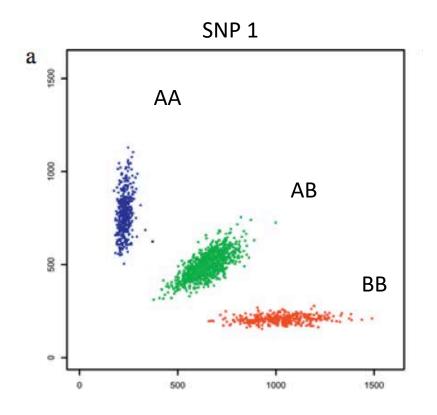
GWAS

Plink format

- FAM file – one row per individual
- BIM file – one row per SNP
- BED file – one row per individual – genotype calls for each individual for all SNPs – binary format
- FAM and BIM file are human readable while BED file in not

FAM File

FID   IID   PATID SEX   PHENO

```
Baka  BAKA_028  0  0  2  6
Baka  BAKA_029  0  0  2  6
Baka  BAKA_030  0  0  1  6
Baka  BAKA_035  0  0  2  6
Baka  BAKA_036  0  0  2  6
```

MATID

BIM File

Chr   SNP ID   SNP Alleles

```
1     1:944564      0.36002543     944564   C        T
1     1:1036959     0.67542294     1036959  C        T
1     1:1087683     0.84946039     1087683  T        C
1     1:1196471     1.225089       1196471  A        G
```

Genetic
Distance
(morgans)

BP Position

Why Do We Need Quality Control?

SNP 1

In an ideal world...

our sampling practices would be perfect,

our experiments would run perfectly,

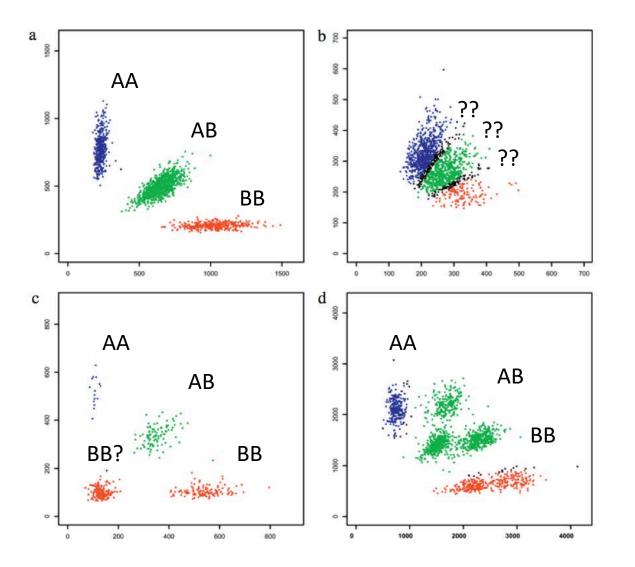and all our SNP genotypes would look like this.

Why Do We Need Quality Control

- Large scale experiments generate true results with a certain error rate

- Errors might originate at various steps in in the process:

  - ✓ Sample selection related issues
    - ✓ Cryptic relatedness
    - ✓ Population structure
  - ✓ Sample handling related issues
    - ✓ Labeling/Plating Error
  - ✓ Genotyping array related issues
    - ✓ Genotyping error
  - ✓ Batch effect related issues
    - ✓ Difference in results due to difference in sample processing

- Not practical to visually assess the genotype plot for every SNP
- Use some biologically relevant metrics as a proxy for quality
- Steps
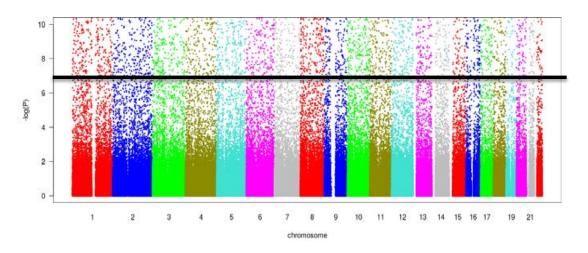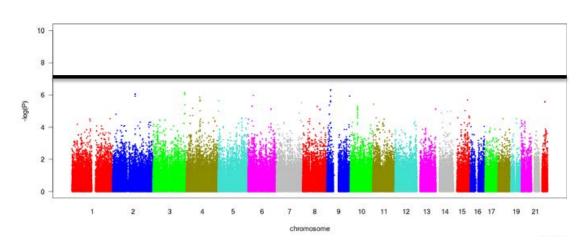 - QC by SNP
  - QC by sample

Why Do We Need Quality Control

We don't live in an ideal world…

German MI family study Affymetrix 500K Array Set
SNPs on chips: 493,840

SNPs passing QC: 270,701

Samani et al. 2007 N Engl J Med 357:443-53

QC Roadmap

Sample QC

Discordant sex information

High Missingness

Excess or deficiency of heterozygosity

Duplicate or related

Divergent ancestry

Batch Effects

SNP QC

Low minor allele frequency

Missingness

Differential missingness

Hardy-Weinberg outliers

- For the sake of simplicity I have split the QC into Sample QC and SNP QC
- The order explained here is therefore not always the best way to do they QC
- The H3ABioNet pipeline has been developed using the best generalized order

Software

- **Programs required for QC**
  - PLINK 1.9 (Purcell, 2007)
  - Scripts for processing result files
  - R (Statistical Software) for plotting results
  - The complete process has been built into a pipeline using NextFlow

Discordant Sex Check

- **Males** have a **single X chromosome** and therefore can be estimated to be **homozygous for all the X chromosome SNPs** (other than those in the pseudo autosomal region(PAR)).

- Therefore, **X chromosome homozygosity** estimate for males **(XHE) is 1**

- Plink assigns sex based on XHE estimate (F or inbreeding coefficient) :

- **Male (1) : XHE >0.80**

- **Female (2) : XHE <0.20**

- **No sex (0) : 0.20 <XHE <0.80**

- Comparisons of **predicted** and **observed** sex can be used to identify miscoded sex or **sample mix-ups**, etc.

All Male Samples

X Chromosome Homozygozity Estimate

All Female Samples

X Chromosome Homozygozity Estimate

Identify individuals with discordant sex information

```
plink --bfile example --check-sex --out sexstat
```

Creates a file named sexstat.sexcheck

```
FID        IID      PEDSEX      SNPSEX      STATUS        F
    P554    P554         2           2          OK     -0.02654
    P555    P555         1           0       PROBLEM     0.5685
    P557    P557         2           2          OK       0.1264
    P558    P558         2           2          OK     -0.0007684
```

Select individuals with Status="PROBLEM" in the file sexstat.sexcheck
Try to identify the problem. If the problem cannot be resolved write the IDs of the individuals with discordant sex information to a file "fail_sex_check.txt"

```
grep "PROBLEM" sexstat.sexcheck > fail_sex_check.txt
```

nextflow Parameters

- `sex_info_available = true`
- `f_low_male = 0.8`
- `f_high_female = 0.2`

Missingness

- Unable to call a genotype for a particular SNP – will be called as missing

- Per sample missingness
  - Percentage of SNPs with missing data per sample

- Per SNP missingness
  - Percentage of missing calls for a SNP

SNP 1

Missingness and Heterozygosity Rate

Genotyping call rate

- Per sample (individual) rate
- Number **of non-missing genotypes divided by the total number of genotyped markers.**
- Low genotyping call rate indicate problem with **sample DNA** like low concentration.
- Thresholds used generally vary **between 3% and 7%**

Genotyping call rate and heterozygosity rate are generally plotted together.
Cutoffs are selected so as to identify outlier individuals based on both the statistics

Heterozygosity Rate

- Per sample (individual) rate
- Number of **(total non-missing genotypes(N) – homozygous(0)) genotypes divided by total non-missing genotypes(N)**
- Excess heterozygosity - Possible sample **contamination**
- Less than expected heterozygosity - Possibly **inbreeding**
- Threshold is usually to remove any individual with +-3 standard deviations from the mean heterozygosity rate of all samples.

Missingness per Individual and per SNP

Missing phenotype (Y/N)

Missingness per individual

FID	IID	MISS_PHENO	N_MISS	N_GENO	F_MISS
P554	P554	N	4096	97722	0.04191
P557	P557	N	4011	97722	0.04105
P558	P558	N	4327	97722	0.04428
P562	P562	N	4099	97722	0.04195

```
plink --bfile example –missing -
-out example_miss
```

```
 Before frequency and genotyping pruning, there are 98604 SNPs
646 founders and 0 non-founders found
34704 heterozygous haploid genotypes; set to missing
Writing list of heterozygous haploid genotypes to [ example_miss.hh ]
3452 SNPs with no founder genotypes observed
Warning, MAF set to 0 for these SNPs (see --nonfounders)
Writing list of these SNPs to [ example_miss.nof ]
Writing individual missingness information to [ example_miss.imiss ]
Writing locus missingness information to [ example_miss.lmiss ]
```

Missingness per SNP (Come back to later)

HR	SNP	N_MISS	N_GENO	F_MISS
1	vh_1_1108138	9	646	0.01393
1	vh_1_1110294	4	646	0.006192
1	rs7515488	1	646	0.001548
1	rs6603785	9	646	0.01393

Heterozygosity rate per Individual

```
plink --bfile example --het --out example_het
```

```
 Before frequency and genotyping pruning, there are 98604 SNPs
646 founders and 0 non-founders found

Detected that binary PED file is v1.00 SNP-major mode
Before frequency and genotyping pruning, there are 98604 SNPs
646 founders and 0 non-founders found
34704 heterozygous haploid genotypes; set to missing
Writing list of heterozygous haploid genotypes to [ example_het.hh ]
3452 SNPs with no founder genotypes observed
Warning, MAF set to 0 for these SNPs (see --nonfounders)
Writing individual heterozygosity information to [ example_het.het ]
```

Observed number of homozygous genotypes

Expected number of homozygous genotypes

Inbreeding coefficient estimate

FID	IID	O(HOM)	E(HOM)	N(NM)	F
P554	P554	67663	6.725e+04	86305	0.02173
P557	P557	66873	6.731e+04	86388	-0.02301
P558	P558	67155	6.707e+04	86091	0.004538
P562	P562	68367	6.724e+04	86306	0.05891

Heterozygosity rate vs. Proportion of missing genotypes

Based on the plot you can decide on reasonable thresholds at which to exclude individuals based on elevated missing or extreme heterozygosity.

We decided to exclude all individuals with a genotype failure rate ≥ 0.06 and heterozygosity rate ± 3 standard deviations from the mean heterozygosity rate of all samples

nextflow Parameters

- `cut_mind =  0.02`
- `cut_het_high = 0.343`
- `cut_het_low = 0.15`

Identify related and duplicate individuals

- A basic assumption of standard population-based association studies is that all the samples are **unrelated** (i.e. the maximum relatedness between any pair of individuals is less than a second degree relative)

- Presence of duplicate and related individuals in the dataset may **introduce bias** and cause **genotypes in families** to be over-represented.

- To identify duplicate and related individuals, a metric (identity by state, IBS) is calculated for each pair of individuals based on the average proportion of alleles shared in common at genotyped SNPs (excluding the sex chromosomes)

SNP1
Cases Initial discovery study Controls
$P = 1 \times 10^{-12}$

Common homozygote   Heterozygote   Variant homozygote

- **Identity by Descent (IBD)** is a **measure** of the **recent shared ancestry** between two individuals based on **genome wide IBS**

- IBD calculations works best when only **independent SNPs** are included in the analysis.

- An independent SNP set for IBD calculations is generally prepared by **removing regions of extended LD** and **pruning the remaining regions** so that no pair of SNPs within a given window (say, 50kb) is correlated.

- IBD is calculated and denoted in Plink as **Pi-hat**

- Convention is to **remove** one individual from a pair with a **Pi-hat > ~0.2**

Pi – hat ~ 1 (Duplicate sample or Monozygotic twins)

Pi – hat ~ 0.5 (First degree relative)

Pi – hat ~ 0.25 (Second degree relative)

Pi – hat ~ 0.125 (Third degree relative)

Identification of duplicated or related individuals

```
IDENTIFY INDEPENDENT SET OF SNPS WITH NO LD
plink --bfile example --exclude high-LD-regions.txt --range --
indep-pairwise 50 5 0.2 --out example

CALCULATE PIHAT FOR ALL PAIRS OF INDIVIDUALS BASED ON SNP SET
ONLY OUTPUT PAIRS WHERE THE PIHAT > 0.2
plink –bfile example --extract example.prune.in –-genome –min
0.2 --out example

FILTERING RELATED INDIVIDUALS USING A SCRIPT TO RANDOMLY REMOVE
ONE OF EACH PAIR WHERE PIHAT > 0.2
perl run-IBD-QC.pl example.genome
```

- Identify all pairs of individuals with an IBD > 0.185
- Looks at the individual call rates stored in example_miss.imiss and output the ids of the individual with the lowest call-rate to 'fail_IBD_example.txt' for subsequent removal

creates a file example.genome containing pairwise IBS for all pairs of individuals

creates the file example.prune.in, containing the list of SNPs to be kept in the analysis.

nextflow Parameters

- `pi-hat = 0.11`
- `super_pi_hat = 0.7`

Population structure

- **Population substructure** or **stratification** occurs when samples have **different genetic ancestries**
- Can lead to **spurious associations** due to **differences in ancestry** rather than true associations
- Imperative to check for population structure within samples
- Can **control** for structure if identified, in downstream analysis

Population structure - PCA

nextflow Parameters

- `case_control = sample.phe`
- `case_control_col = PHE`
- `batch = sample.phe`
- `batch_col = batch_no`

Discordant sex information

fail_sexcheck_example.txt

High Missingness and outlying heterozygosity rate

fail_miss_het_example.txt

Duplicate and related

fail_IBD_example.txt

Divergent ancestry

fail_ancestry_example.txt

Sample QC completed

```
JOIN FILES
cat fail_* | sort -k1 | uniq > fail_example_inds.txt
REMOVE FROM DATA
plink --bfile example --remove fail_example_inds.txt --make-bed --out clean_inds_example
```

Low minor allele frequency SNPs

- Genotype calling algorithms **perform poorly** for SNPs with **low MAF** and **low samples sizes.**
- **Power** for detecting associations with SNPs with low MAF is low unless the sample size is very large
- Commonly used exclusion threshold are SNPs with a **MAF 0.01 to 0.05** (dependent on sample size)

Identify low minor allele frequency SNPs

CALCULATE MINOR ALLELE FREQUENCY DISTRIBUTION FOR ALL SNPS

```
plink --bfile clean-inds-example --freq --out
clean_inds_example_freq
```

Generates the file "clean_inds_example_freq.frq" containing minor allele frequency of each SNP

GENERATE PLOT OF MAF DISTRIBUTION

PLOT_MAF.R

Choose standard MAF threshold (MAF > 0.01 OR MAF > 0.5) or base on distribution

CHR	SNP	A1	A2	MAF	NCHROBS
1	vh_1_1108138	A	G	0.4443	1292
1	vh_1_1110294	A	G	0.362	1304
1	rs7515488	A	G	0.2893	1310
1	rs6603785	T	A	0.4985	1292
1	rs6603788	G	A	0.2221	1306
1	1_1209245	G	C	0.04609	1150
1	rs2274264	A	G	0.1336	1302
1	rs12103	G	A	0.09862	1308

nextflow Parameters

- `cut_maf = 0.01`

Identify SNPS with high missingness

CALCULATE MISSINGNESS PER SNP

```
plink --bfile clean_inds_example --missing --out clean_inds_example_missing
```

Generates the file "clean_inds_example_mising.lmiss" containing missingness value for each SNP

GENERATE PLOT OF SNP MISSINGNESS

SNPMISS_PLOT.R

Choose the standard missingness threshold (> 0.05) or choose threshold based on the plot

CHR	SNP	N_MISS	N_GENO	F_MISS
1	vh_1_1108138	10	656	0.01524
1	vh_1_1110294	4	656	0.006098
1	rs7515488	1	656	0.001524
1	rs6603785	10	656	0.01524
1	rs6603788	3	656	0.004573
1	1_1209245	81	656	0.1235
1	rs2274264	5	656	0.007622
1	rs12103	2	656	0.003049
1	rs12142199	7	656	0.01067
1	rs880051	2	656	0.003049

SNP Missingness Distribution

nextflow Parameters

- `cut_geno = 0.01`

Differential missingness

Signal for controls · Signal for cases

- Missing frequency is also assessed separately in cases and in controls because differential missingness is a common source of false positive associations.
- SNPs showing highly differential missingness (P<0.00001) are excluded

Identify SNPS with high differential missingness between cases and controls

CALCULATE DIFFERENTIAL MISSINGNESS

```
plink --bfile clean_inds_example --test-missing --
out clean_inds_example_test_missing
```

Generates the file "example_test_missing.missing" containing differential missingness statistics for each SNP

GENERATE PLOT OF P-VALUE DISTRIBUTION

DIFFMISS_PLOT.R

Choose standard differential missingness p-value threshold (0.00001) or choose on the basis of the plot

CHR	SNP	F_MISS_A	F_MISS_U	P
1	vh_1_1108138	0.02318	0.008475	0.1996
1	vh_1_1110294	0.006623	0.00565	1
1	rs7515488	0.003311	0	0.4604
1	rs6603785	0.01987	0.0113	0.5254
1	rs6603788	0.003311	0.00565	1
1	1_1209245	0.1358	0.113	0.4057
1	rs2274264	0.009934	0.00565	0.6658
1	rs12103	0	0.00565	0.5024

Distribution of differential missingness P-value

nextflow Parameters

- `cut_diff_miss = 0.05`

Hardy Weinberg Equilibrium

- Expected relationship between allele and genotype frequencies under certain assumptions
- Allele frequencies and genotypes remains constant over generations
- Deviations from HWE are used as a proxy for possible genotyping errors

Assumptions

- Diploid organisms
- Infinite population size
- Non-overlapping generations
- Random mating
- No selection, mutation or migration

Testing for HWE
- Calculate the allele frequency (p)
  - Using observed genotype counts
- Calculate the expected genotype counts
  - Using the allele frequency (p)
- Compare the observed to the expected counts
  - χ^2 test

Reasons for HW Deviations

- **Genotyping Error**
- Subdivided Population
- Excess homozygotes= "Allele dropout in old samples"
- Any violations of the HW assumptions

- SNPs are **excluded** if **substantially more or fewer samples heterozygous at a SNP than expected** (excess heterozygosity or heterozygote deficiency)
- Threshold for significance **10^{-3} to 10^{-6}**
- Can **only remove SNPs in controls** which deviate from HWE or **use less stringent HWE threshold** in **SNPs in cases over controls**

GENEVA alcohol-dependence project: Quality control report

Identify SNPS which show extreme HWE deviations

GET DISTRIBUTION OF P-VALUES FOR ALL SNPS

```
plink --bfile clean_inds_example --hardy --out
clean_inds_example_hwe
```

Generates the file "clean_inds_example_hwe.hwe" containing Hardy Weinberg statistics for each SNP separately in cases, controls and all. samples

SELECT UNAFFECTED
```
head -1 clean_inds_example_hwe.hwe >
example_clean_inds_example_hweu.hwe | grep
"UNAFF" clean_inds_example_hwe.hwe >>
example_clean_inds_example_hweu.hwe
```

GENERATE PLOT USING R SCRIPT

HWE_PLOT.R
(based only on controls)

Choose the standard HWE P-value threshold (0.00001) or select one
on the basis of the plot

CHR	SNP	TEST	A1	A2	GENO	O(HET)	E(HET)	P
1	vh_1_1108138	ALL	A	G	134/306/206	0.4737	0.4938	0.301
1	vh_1_1108138	AFF	A	G	63/135/97	0.4576	0.4934	0.2376
1	vh_1_1108138	UNAFF	A	G	71/171/109	0.4872	0.4941	0.829
1	vh_1_1110294	ALL	A	G	92/288/272	0.4417	0.4619	0.2706
1	vh_1_1110294	AFF	A	G	37/136/127	0.4533	0.455	1
1	vh_1_1110294	UNAFF	A	G	55/152/145	0.4318	0.4673	0.1707
1	rs7515488	ALL	A	G	50/279/326	0.426	0.4112	0.3931

HWE P-value

nextflow Parameters

- `cut_hwe = 0.000001`

Genome Wide-Association Studies

Input Files

.{bed,bim,fam} – Raw

.{bed,bim,fam} – No duplicates

Identify and Remove Duplicate SNPs

`dups.py`

`plink`

Identify Individuals with Discordant Sex Information

`plink`

Prune for SNPs in LD for IBD

`plink`

Calculate Individual Missingness Statistics

`plink`

Calculate Individual Hetrozygosity

`plink`

failed_sex_check

.genome

.lmiss

.lmiss

.het

Calculate SNP Missingness

`plink`

Calculate Minor Allele Frequency

`plink`

Remove QC Individuals

`plink`

Identify Related Individuals

`run_IBD_QC_qcplink.pl`

Identify Individuals with High Missingness and/or Bad Heterozygosity

`select_miss_het_qcplink.pl`

rel_indivs

failed_miss_het

.{bed,bim,fam} – QC1

Calculate HWE scores of each SNP

`plink`

Calculate Differential Missingness

`plink`

Identify SNPs with Differential Missingness

`select_diffmiss_qcplink.pl`

Remove SNPs that Fail Based on Threshold Selected

`plink`

.{bed,bim,fam} – QC2

Principal Component Analysis

`plink`

Generate Outputs/Plots (R)

`miss_het_plot_qcplink.R`

`maf_plot_qcplink.R`

`snpmiss_plot_qcplink.R`

`imiss_splot.R`

`diffmiss_splot_qcplink.R`

`hwe_plot_qcplink.R`

`drawPCA.R`

Generate Report

`qcreport.py`

PDF Report

Acknowledgments

- Ananyo Choudhury