Data Quality Control (QC) in Association Studies

Svetlana (Sarah) Cherlin

svetlana.cherlin@newcastle.ac.uk

Introduction



 Poor study design and errors in genotype calling can introduce systematic bias in association studies

Introduction



- Poor study design and errors in genotype calling can introduce systematic bias in association studies
 - ► increase in false positive error rate
 - decrease in power
- Assess data quality to remove sub-standard genotypes, samples and SNPs from subsequent association analysis

Introduction

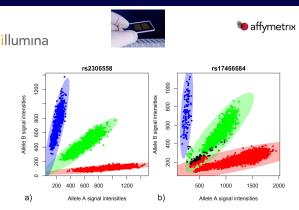


- Poor study design and errors in genotype calling can introduce systematic bias in association studies
 - ► increase in false positive error rate
 - decrease in power
- Assess data quality to remove sub-standard genotypes, samples and SNPs from subsequent association analysis
- Tutorials
 - ► Anderson et al. Nature Protocols 2010, doi:10.1038/nprot.2010.116
 - ► Turner et al. Curr Protoc Hum Genet. 2011. doi:10.1002/0471142905.hg0119s68
 - ▶ Marees et al. Int J Methods Psychiatr Res. 2018. doi: 10.1002/mpr.1608









- Examples of cluster plots for two SNPs. One spot corresponds to one sample.
- Samples with genotypes AA and BB are red and blue, respectively. Heterozygous samples
 are shown in green; samples with missing genotypes are black. The ellipses represent the
 cluster boundaries as computed by ACPA.
- a) No samples in overlapping ellipses; b) Red samples lie in the green ellipse. At the bottom of the green ellipse, samples have been erroneously classified as red samples.

Schillert et al. BMC Proceedings 2009, 3(Suppl 7):S58 doi: 10.1186/1753-6561-3-S7-S58

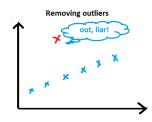
• For large-scale GWA studies, automated genotype calling algorithms have been developed

- For large-scale GWA studies, automated genotype calling algorithms have been developed
 - often specific to genotype calling technology
 - estimate probability or confidence that any specific genotype is AA, AB or BB
 - apply threshold to probabilities or confidence in order to call genotype, otherwise treated as missing

- For large-scale GWA studies, automated genotype calling algorithms have been developed
 - often specific to genotype calling technology
 - estimate probability or confidence that any specific genotype is AA, AB or BB
 - ► apply threshold to probabilities or confidence in order to call genotype, otherwise treated as missing
- Choice of calling threshold will impact results

- For large-scale GWA studies, automated genotype calling algorithms have been developed
 - often specific to genotype calling technology
 - estimate probability or confidence that any specific genotype is AA, AB or BB
 - apply threshold to probabilities or confidence in order to call genotype, otherwise treated as missing
- Choice of calling threshold will impact results
 - ▶ too low: include poor quality genotypes
 - too high: unnecessarily remove high quality genotypes, or may introduce bias by preferentially calling specific genotypes (e.g. rare homozygotes)

- For large-scale GWA studies, automated genotype calling algorithms have been developed
 - often specific to genotype calling technology
 - estimate probability or confidence that any specific genotype is AA, AB or BB
 - apply threshold to probabilities or confidence in order to call genotype, otherwise treated as missing
- Choice of calling threshold will impact results
 - ► too low: include poor quality genotypes
 - too high: unnecessarily remove high quality genotypes, or may introduce bias by preferentially calling specific genotypes (e.g. rare homozygotes)
- Missing call rate is not only a measure of data completeness, but is also a measure of genotype quality





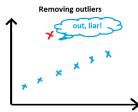
- Low call rate
 - poor DNA quality



- Low call rate
 - poor DNA quality
- Outlying heterozygosity across autosomes
 - ► DNA sample contamination or inbreeding



- Low call rate
 - poor DNA quality
- Outlying heterozygosity across autosomes
 - ► DNA sample contamination or inbreeding
- Duplication or relatedness based on identity-by-state or identity-by-descent
 - ▶ if not taken account of in the analysis



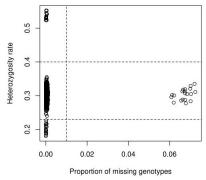
- Low call rate
 - poor DNA quality
- Outlying heterozygosity across autosomes
 - ► DNA sample contamination or inbreeding
- Duplication or relatedness based on identity-by-state or identity-by-descent
 - ▶ if not taken account of in the analysis
- Mismatches with external information, i.e. sex discrepancy
 - sample mix-up



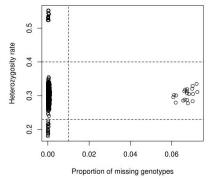
- Low call rate
 - poor DNA quality
- Outlying heterozygosity across autosomes
 - ► DNA sample contamination or inbreeding
- Duplication or relatedness based on identity-by-state or identity-by-descent
 - ▶ if not taken account of in the analysis
- Mismatches with external information, i.e. sex discrepancy
 - sample mix-up
- Outlying population ancestry
 - confounding due to population structure



 There are samples with high levels of missing data and samples with unusually high and low heterozygosity

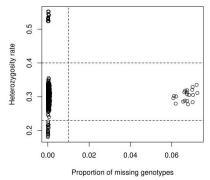


 There are samples with high levels of missing data and samples with unusually high and low heterozygosity



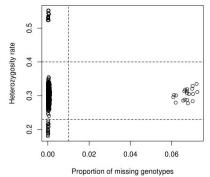
• Decide upon thresholds for removing individuals based on the plot

 There are samples with high levels of missing data and samples with unusually high and low heterozygosity



- Decide upon thresholds for removing individuals based on the plot
- ullet Dashed lines denote QC thresholds (exclude samples with missing call rate > 0.1, and samples with heterozygosity rate < 0.23 and > 0.4)

 There are samples with high levels of missing data and samples with unusually high and low heterozygosity



- Decide upon thresholds for removing individuals based on the plot
- Dashed lines denote QC thresholds (exclude samples with missing call rate > 0.1, and samples with heterozygosity rate < 0.23 and > 0.4)
- Rule of thumb: remove individuals who deviate ± 3 SD from the samples' heterozygosity rate mean

• Two alleles are identical by state (IBS) if they have the same nucleotide sequence

- Two alleles are identical by state (IBS) if they have the same nucleotide sequence
- Over M markers, the IBS between the individuals i and j is

$$\mathrm{IBS}_{ij} = 1 - \frac{1}{2M} \sum_k |G_{ik} - G_{jk}|,$$

- Two alleles are identical by state (IBS) if they have the same nucleotide sequence
- Over M markers, the IBS between the individuals i and j is

$$IBS_{ij} = 1 - \frac{1}{2M} \sum_{k} |G_{ik} - G_{jk}|,$$

- Identical samples will share IBS near to 100%
 - allowing for genotyping errors

- Two alleles are identical by state (IBS) if they have the same nucleotide sequence
- Over M markers, the IBS between the individuals i and j is

$$IBS_{ij} = 1 - \frac{1}{2M} \sum_{k} |G_{ik} - G_{jk}|,$$

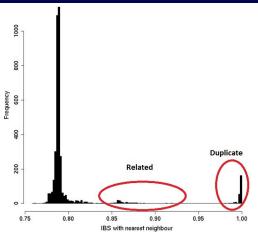
- Identical samples will share IBS near to 100%
 - allowing for genotyping errors
- Related individuals will share higher IBS than unrelated individuals

- Two alleles are identical by state (IBS) if they have the same nucleotide sequence
- Over M markers, the IBS between the individuals i and j is

$$IBS_{ij} = 1 - \frac{1}{2M} \sum_{k} |G_{ik} - G_{jk}|,$$

- Identical samples will share IBS near to 100%
 - allowing for genotyping errors
- Related individuals will share higher IBS than unrelated individuals
- Common to plot histogram of IBS of each individual with "nearest neighbour"

IBS Distribution



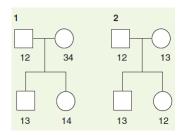
- For each individual, the distance to its nearest neighbour is calculated
- Remove one sample from each duplicate or related pair (usually one with lowest call rate)
- Alternative: take account of relatedness in analysis
- The absolute amount of IBS sharing depends on allele frequencies in the population
- Methods that estimate kinship or relatedness coefficients typically aim for estimating identity-by-descent (IBD)

The degree of recent shared ancestry for a pair of individuals

- The degree of recent shared ancestry for a pair of individuals
- The alleles are identical by descent (IBD) if they have been inherited from a common ancestor

- The degree of recent shared ancestry for a pair of individuals
- The alleles are identical by descent (IBD) if they have been inherited from a common ancestor
- IBD must also be IBS, the converse of this statement is not true

- The degree of recent shared ancestry for a pair of individuals
- The alleles are identical by descent (IBD) if they have been inherited from a common ancestor
- IBD must also be IBS, the converse of this statement is not true



- Pedigree 1: Siblings share allele 1 IBD (inherited from the father)
- Pedigree 2: Siblings share allele 1 IBS (inherited from different parents)

Forabosco et al. Expert Rev. Mol. Diagn. 5(5), (2005). doi: 10.1586/14737159.5.5.781

• Proportion of the genome at which a pair of individuals share 0, 1 or 2 alleles (Z0, Z1 and Z2), or probabilities of sharing 0, 1 and 2 alleles

- Proportion of the genome at which a pair of individuals share 0, 1 or 2 alleles (Z0, Z1 and Z2), or probabilities of sharing 0, 1 and 2 alleles
- Mean IBD sharing $\hat{\pi} = 0 \times Z1 + 1 \times Z1 + 2 \times Z2$

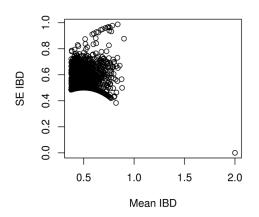
- Proportion of the genome at which a pair of individuals share 0, 1 or 2 alleles (Z0, Z1 and Z2), or probabilities of sharing 0, 1 and 2 alleles
- Mean IBD sharing $\hat{\pi} = 0 \times Z1 + 1 \times Z1 + 2 \times Z2$
- Expected patterns of mean IBD for known related pairs
 - ▶ IBD $\hat{\pi} = 1$ for duplicates or monozygotic twins
 - in practice, use $\hat{\pi} > 0.98$
 - ▶ IBD $\hat{\pi} = 0.5$ for first-degree relatives
 - ▶ IBD $\hat{\pi} = 0.25$ for second-degree relatives
 - ▶ IBD $\hat{\pi} = 0.125$ for third-degree relatives

- Proportion of the genome at which a pair of individuals share 0, 1 or 2 alleles (Z0, Z1 and Z2), or probabilities of sharing 0, 1 and 2 alleles
- Mean IBD sharing $\hat{\pi} = 0 \times Z1 + 1 \times Z1 + 2 \times Z2$
- Expected patterns of mean IBD for known related pairs
 - ▶ IBD $\hat{\pi} = 1$ for duplicates or monozygotic twins
 - in practice, use $\hat{\pi} > 0.98$
 - ▶ IBD $\hat{\pi} = 0.5$ for first-degree relatives
 - ▶ IBD $\hat{\pi} = 0.25$ for second-degree relatives
 - ▶ IBD $\hat{\pi} = 0.125$ for third-degree relatives
- Remove one from each pair with $\hat{\pi} > 0.185$
 - ▶ halfway between the expected IBD for third- and second-degree relatives

Identity-by-descent (IBD)

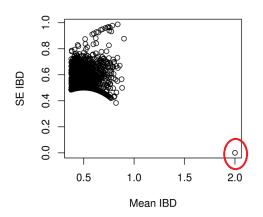
- Proportion of the genome at which a pair of individuals share 0, 1 or 2 alleles (Z0, Z1 and Z2), or probabilities of sharing 0, 1 and 2 alleles
- Mean IBD sharing $\hat{\pi} = 0 \times Z1 + 1 \times Z1 + 2 \times Z2$
- Expected patterns of mean IBD for known related pairs
 - ▶ IBD $\hat{\pi} = 1$ for duplicates or monozygotic twins
 - in practice, use $\hat{\pi} > 0.98$
 - ▶ IBD $\hat{\pi} = 0.5$ for first-degree relatives
 - ▶ IBD $\hat{\pi} = 0.25$ for second-degree relatives
 - ▶ IBD $\hat{\pi} = 0.125$ for third-degree relatives
- Remove one from each pair with $\hat{\pi} > 0.185$
 - ▶ halfway between the expected IBD for third- and second-degree relatives
- Prune the data for LD before assessing IBD
 - shared region of high LD results in more shared variants than one of low LD, even if the two regions are the same size

IBD Plot



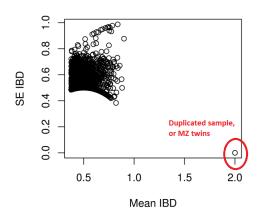
Spot the duplicates...

IBD Plot



Spot the duplicates...

IBD Plot



Spot the duplicates...

 Traditional approaches to association analysis assume that individuals are unrelated to each other

 Traditional approaches to association analysis assume that individuals are unrelated to each other

In practice ...

MZ twins Duplicated samples DZ twins Cryptic relatedness









- Traditional approaches to association analysis assume that individuals are unrelated to each other
- In practice ...

MZ twins

Duplicated samples

DZ twins

Cryptic relatedness

 Including related individuals in the analysis, without accounting for these relationships, can increase false positive error rates and reduce power

- Traditional approaches to association analysis assume that individuals are unrelated to each other
- In practice ...

MZ twins

Duplicated samples

DZ twins

Cryptic relatedness

- Including related individuals in the analysis, without accounting for these relationships, can increase false positive error rates and reduce power
- Mixed modelling approaches account for "relatedness" between individuals (families, cryptic relatedness, population structure) by allowing for kinship

Gender Check - X Chromosome



 Calculate the homozygosity rate F across all X-chromosome SNPs for each individual in the sample and compare these to the expected rate

Gender Check - X Chromosome



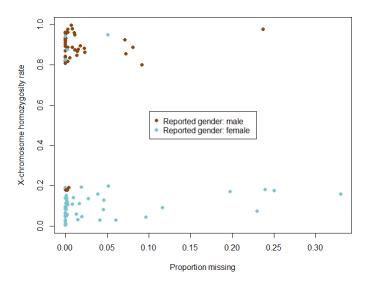
- Calculate the homozygosity rate F across all X-chromosome SNPs for each individual in the sample and compare these to the expected rate
- Expected homozygosity rates
 - F > 0.8 for male samples and F < 0.2 for female samples
 - ▶ males have only one X cannot be heterozygous
 - expect some genotyping errors

Gender Check - X Chromosome



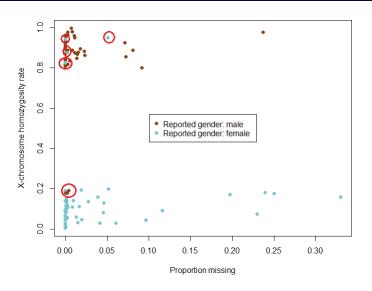
- Calculate the homozygosity rate F across all X-chromosome SNPs for each individual in the sample and compare these to the expected rate
- Expected homozygosity rates
 - F > 0.8 for male samples and F < 0.2 for female samples
 - ▶ males have only one X cannot be heterozygous
 - expect some genotyping errors
- Gender error reported for mismatch in reported and genetic sex
- Discrepancies with external gender information may reflect:
 - errors in external data
 - ► sample mix-up

Gender Check - Plot



Spot the discrepancies...

Gender Check - Plot



Spot the discrepancies...

Gender Check - Examples

Reported	Homozygosity	Gender according
gender	rate	to SNPs
Male	0.98	Male
Female	0.03	Female
Female	0.99	Male
Female	0.28	Unknown*
Female	0.35	Unknown**

^{*} Likely a female with sex chromosome anomaly (e.g. XX/XO mosaic, loss-of-heterozygosity on X)

Adapted from Turner et al. Curr Protoc Hum Genet, (2011). doi:10.1002/0471142905.hg0119s68

^{**} Likely a male with sex chromosome anomaly (e.g. XXY or XX/XY mosaic)

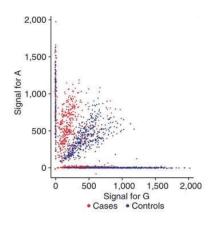
- Low call rate (95% or 99%), variable by MAF
 - poor quality SNP

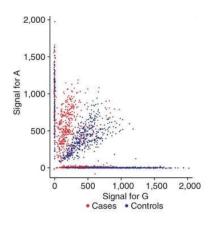
- Low call rate (95% or 99%), variable by MAF
 - poor quality SNP
- Extreme deviation from Hardy-Weinberg equilibrium
 - ▶ in cases, controls or both for autosomes
 - ▶ incompatible with the assumption of random mating
 - possible genotyping error

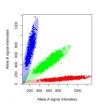
- Low call rate (95% or 99%), variable by MAF
 - poor quality SNP
- Extreme deviation from Hardy-Weinberg equilibrium
 - ▶ in cases, controls or both for autosomes
 - ▶ incompatible with the assumption of random mating
 - possible genotyping error
- Low frequency SNPs (MAF 1-5%)
 - difficult to call using current genotype calling algorithms
 - possible genotyping error
 - association signals seen at these rare SNPs are underpowered

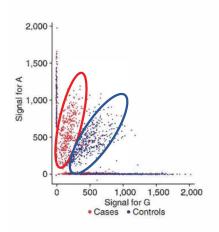
- Low call rate (95% or 99%), variable by MAF
 - poor quality SNP
- Extreme deviation from Hardy-Weinberg equilibrium
 - ▶ in cases, controls or both for autosomes
 - ▶ incompatible with the assumption of random mating
 - possible genotyping error
- Low frequency SNPs (MAF 1-5%)
 - difficult to call using current genotype calling algorithms
 - possible genotyping error
 - association signals seen at these rare SNPs are underpowered
- Study specific SNP QC filters
 - ▶ differences in allele frequencies between multiple control cohorts

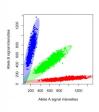
- Low call rate (95% or 99%), variable by MAF
 - poor quality SNP
- Extreme deviation from Hardy-Weinberg equilibrium
 - ▶ in cases, controls or both for autosomes
 - ▶ incompatible with the assumption of random mating
 - possible genotyping error
- Low frequency SNPs (MAF 1-5%)
 - difficult to call using current genotype calling algorithms
 - possible genotyping error
 - association signals seen at these rare SNPs are underpowered
- Study specific SNP QC filters
 - differences in allele frequencies between multiple control cohorts
- Extreme differential call rates between cases and controls

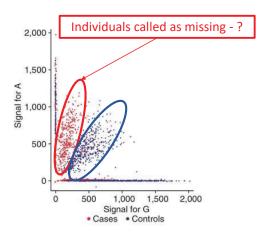


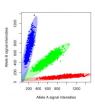


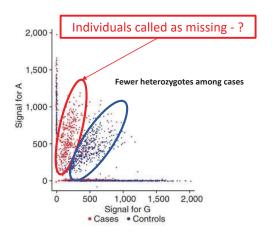


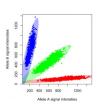














 Specialised software for quality control of genome-wide association studies that can handle scale and complexity of data



- Specialised software for quality control of genome-wide association studies that can handle scale and complexity of data
- QCTOOL: flexible command line software with range of filtering options

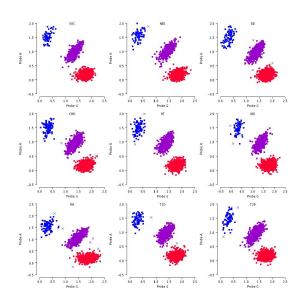


- Specialised software for quality control of genome-wide association studies that can handle scale and complexity of data
- QCTOOL: flexible command line software with range of filtering options
- PLINK: whole-genome association analysis toolset

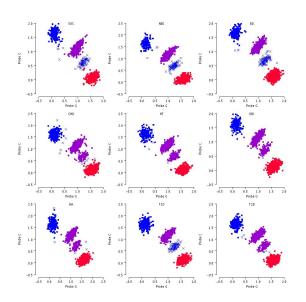


- Specialised software for quality control of genome-wide association studies that can handle scale and complexity of data
- QCTOOL: flexible command line software with range of filtering options
- PLINK: whole-genome association analysis toolset
 - ► **Spoiler alert**: PLINK will be used in practicals

Visual Inspection of Cluster Plots - Good SNP



Visual Inspection of Cluster Plots - Bad SNP



Summary



- QC is an essential step of the analysis
- QC criteria are subjective and vary from one study to another
- Sample QC filters should not be so stringent as to remove the majority of the analysis cohort
- SNP QC filters should eliminate the worst quality markers
- All SNPs demonstrating evidence for association should be followed up with visual inspection of cluster plots