



Estimating the proportion of variation in susceptibility to multiple sclerosis captured by common SNPs

Corey T. Watson¹, Giulio Disanto^{2,3}, Felix Breden¹, Gavin Giovannoni⁴ & Sreeram V. Ramagopalan^{4,5}

¹Department of Biological Sciences, Simon Fraser University, Burnaby, British Columbia, V5A 1S6, Canada, ²Wellcome Trust Centre for Human Genetics, University of Oxford, Roosevelt Drive, Headington, Oxford, OX3 7BN, United Kingdom, ³Department of Clinical Neurology, University of Oxford, The West Wing, John Radcliffe Hospital, Oxford, OX3 9DU, United Kingdom, ⁴Blizard Institute, Queen Mary University of London, Barts and The London School of Medicine and Dentistry, London, E1 2AT, United Kingdom, ⁵CGAT, MRC Functional Genomics Unit, Department of Physiology, Anatomy and Genetics, University of Oxford, Oxford, OX1 3QX, United Kingdom.

Multiple sclerosis (MS) is a complex disease with underlying genetic and environmental factors. Although the contribution of alleles within the major histocompatibility complex (MHC) are known to exert strong effects on MS risk, much remains to be learned about the contributions of loci with more modest effects identified by genome-wide association studies (GWASs), as well as loci that remain undiscovered. We use a recently developed method to estimate the proportion of variance in disease liability explained by 475,806 single nucleotide polymorphisms (SNPs) genotyped in 1,854 MS cases and 5,164 controls. We reveal that ~30% of MS genetic liability is explained by SNPs in this dataset, the majority of which is accounted for by common variants. These results suggest that the unaccounted for proportion could be explained by variants that are in imperfect linkage disequilibrium with common GWAS SNPs, highlighting the potential importance of rare variants in the susceptibility to MS.

Multiple sclerosis (MS) is an inflammatory disease of the central nervous system, and is the most common neurological disorder affecting young adults¹. Current evidence implicates roles for both environmental and genetic factors in the onset and progression of the disease^{2–4}. The importance of genetic factors in MS was recognized early in the study of the disease, and is best illustrated by observations of strong familial clustering and a significantly increased risk in first-degree relatives^{5–7}. Further support for the role of genes in MS comes from studies of monozygotic and dizygotic twins, which also indicate a strong genetic component; however, heritability estimates from these studies range from roughly 25% to 75%^{8–11}. Alleles of the major histocompatibility complex (MHC) are so far known to make the single strongest contribution to MS susceptibility¹². In addition, many loci of more modest effect have also recently been identified in genome-wide association studies (GWASs)^{13–16}. While risk alleles at the MHC are thought to represent a significant proportion of MS genetic susceptibility¹³, the contribution of variants outside of the MHC, specifically those represented by single nucleotide polymorphisms (SNPs) genotyped by GWASs, has not been extensively explored. To investigate in more detail the role of common GWAS variants in MS susceptibility, we used publicly available genotype data from the United Kingdom (UK) MS patient and control cohorts¹⁶ and a recently described approach that assesses contributions made by all genotyped SNPs, rather than solely risk loci that reach genome-wide significance^{17–20}. From this analysis we show that approximately 30% of the genetic variation in liability to MS is directly explained by variants represented by current GWAS arrays.

Results

For this study, we used genome-wide genotype data for 475,806 autosomal SNPs collected from 1,854 MS cases and 5,164 controls sampled from the UK¹⁶. After assessing the relatedness between individuals, and thus accounting for effects of population structure, we first estimated the proportion of variance explained by all autosomal SNPs simultaneously. This analysis revealed that 30.7% (standard error (SE) = 2.05%) of the variance in liability to MS is accounted for by SNPs in this dataset.

We next partitioned SNPs by autosome and recalculated the proportion of variance explained by variants found on each chromosome (Table 1); estimated values ranged from ~0–8% per chromosome. Not surprisingly,

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Correspondence and
requests for materials
should be addressed to
S.V.R. (s.ramagopalan@
qmul.ac.uk)



Table 1 | Proportion of variance in MS liability explained per chromosome

chr	Variance Explained	Standard Error
1	0.011606	0.006417
2	0.010433	0.006207
3	0.021433	0.006129
4	0.002666	0.005454
5	0.021062	0.005955
6	0.081112	0.007155
7	0.013365	0.005453
8	0.000678	0.004836
9	0.006747	0.004896
10	0.005168	0.004938
11	0.003246	0.004827
12	0.014884	0.005266
13	0.005035	0.004257
14	0.008067	0.004431
15	0.01251	0.004326
16	0.01705	0.004983
17	0.015371	0.004533
18	0.003484	0.004116
19	0.007125	0.003979
20	0.007533	0.004086
21	0	0.002963
22	0.003493	0.003107

given the known contribution of the MHC, which is located on chromosome 6, SNPs on this chromosome account for 8.11% of the variance (SE = 0.72%). By calculating the proportion of the genome represented by each chromosome (not including the length of sex chromosomes), we tested for a correlation between the variance explained by each chromosome relative to its size, excluding chromosome 6 (Figure 1). Although it was evident that several of the smaller chromosomes contributed less to the overall variance than several of the larger chromosomes, the overall trend was not significant ($r = 0.336$, $P = 0.136$). To assess the contribution made by common versus rare variants, we also binned SNPs based on minor allele frequency (MAF; Figure 2). From this, we observed that common variants (MAF > 0.1; ~4–6%), which are most abundantly

sampled on GWAS arrays, make a greater contribution than rare variants (MAF < 0.1; ~2.8%). However, because of the unequal number of SNPs in each bin, we also binned SNPs by quintile (Figure 3). Based on this analysis, we found that all quintiles displayed an equivalent variance, highlighting that no particular frequency of MAF makes a larger or smaller contribution to MS, and that all should be captured and tested.

Lastly, we carried out an association analysis using only the UK GWAS data. We identified 15 associated autosomal SNPs in this cohort outside of the MHC with P values $< 1 \times 10^{-5}$. These SNPs, their positions (hg18; NCBI Build 36.1), and the nearest RefSeq gene to each are listed in Table 2. Using association analysis data, we also examined the contribution made by all associated SNPs to the observed variance after binning by P value, including those SNPs within the MHC (Table 3).

Discussion

Using available data from a large UK case-control cohort¹⁶, we have conducted a comprehensive assessment of the contribution of genome-wide SNPs on the variance in liability to MS. The power of the approach used here is that contributions of genotypes at all available loci across the genome (in this case, 475,806), rather than only a set of identified MS risk loci, can be accounted for using this method. Thus, from our analysis, we conclude that approximately 30% of MS heritability is explained by variants on current GWAS arrays, including SNPs on chromosome 6, which alone account for ~8% and reflect the major contribution of the MHC. The role of the MHC in MS has long been known; specifically, *HLA-DRB1*1501* confers a 2-fold increase in risk¹³. However, the underlying genetic architecture of MS is presumed to be polygenic, involving a large number of loci with smaller effects^{22,23}. Our findings lend support to this notion, as we observed that the genetic contributions of SNPs on autosomes other than chromosome 6 were at least in part correlated to autosome length. However, this relationship was not significant, and not as convincing as that illustrated previously for other polygenic disorders^{17,21}. This might hint at the possibility that some unidentified MS risk loci have slightly larger effects than others, which has been discussed recently²³. Additionally, our study was smaller than that

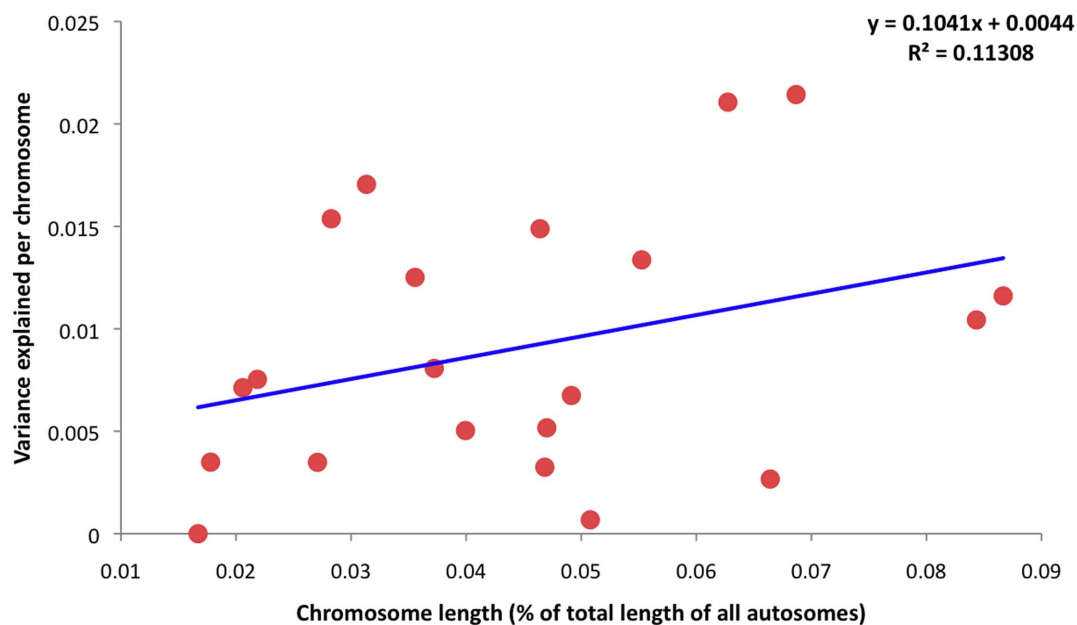


Figure 1 | Contribution of GWAS SNPs and chromosome length. The proportion of variance in MS liability explained by SNPs partitioned by autosome (based on data from Table 1, excluding chr 6) relative to chromosome size, which was determined by dividing the length of each autosome by the sum of the lengths of all autosomes.

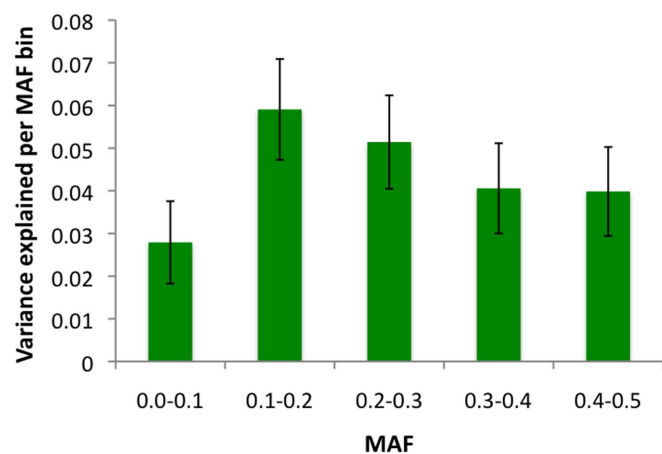


Figure 2 | Contribution of GWAS SNPs partitioned by minor allele frequency. The total proportion of variance explained and standard errors for SNPs in each of five MAF bins. The number of SNPs included in each bin varied slightly (0.0–0.1%, $n = 76046$; 0.1–0.2%, $n = 112435$; 0.2–0.3%, $n = 97482$; 0.3–0.4%, $n = 89704$; 0.4–0.5%, $n = 86625$).

of Yang *et al.*¹⁷ and Lee *et al.*²¹, and thus would be comparatively underpowered.

Also notable, we observed that the majority of variation represented by GWAS SNPs was explained by common variants with MAFs over 0.1%, perhaps not surprisingly given that these outnumbered rare variants. This highlights both, the utility of GWAS arrays, which have placed much emphasis on the inclusion of common SNPs, and the fact that the use of larger sample sizes in GWAS should increase power and yield discoveries of additional risk loci, a point that has recently been noted in the context of schizophrenia²¹. Importantly though, this observation does not delimit the potentially significant role of rare variants in MS. For example, rare variants in *CYP27B1*, a gene essential to vitamin D synthesis, have been reported at low frequencies in MS patients, but not in controls (odds ratio = 4.7)²⁴. Rare variants in the *TYK2* gene have also more recently been shown to influence MS risk²⁵. Furthermore, we found that even after including the effects of over 400,000 SNPs in this cohort, most of the variance in MS liability remains unaccounted for. As has been discussed previously in the context of the “missing heritability” of complex diseases, one of the more likely explanations for this is that

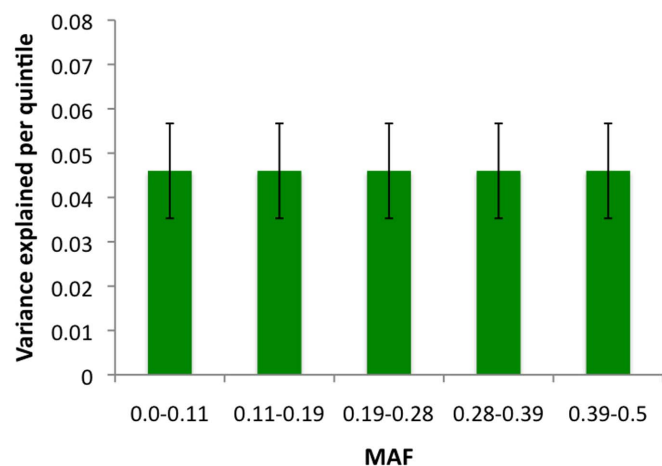


Figure 3 | Contribution of GWAS SNPs partitioned by quintile. The total proportion of variance explained and standard errors for all SNPs tested after binning by quintile. The number of SNPs included in each quintile are as follows: 0.0–0.11%, $n = 93079$; 0.11–0.19%, $n = 93074$; 0.19–0.28%, $n = 93076$; 0.28–0.39%, $n = 93089$; 0.39–0.5%, $n = 93116$).

Table 2 | Top SNPs from association analysis using UK GWAS data

SNP	Chr	Position	Gene	P value
rs6662618	1	92707999	GFI1	1.95E-06
rs11809572	1	101122894	EXTL2	9.34E-06
rs16849327	3	104970212	ZPLD1, ALCAM	7.17E-06
rs16869665	4	20095328	SLIT2	3.14E-06
rs2214543	7	10763417	NDUFA4	8.31E-06
rs11984075	7	37403379	ELMO1	6.40E-07
rs10749170	10	116302100	ABLIM1	5.67E-06
rs10502249	11	122009461	UBASH3B	6.38E-06
rs11069349	13	98572648	DOCK9	1.83E-06
rs727263	13	98802109	UBAC2	3.26E-06
rs7325747	13	98827933	UBAC2	4.36E-06
rs9303323	17	37341634	TTC25	5.30E-06
rs12952314	17	37398449	DNAJC7	8.18E-06
rs7209012	17	37414849	DNAJC7	9.42E-07
rs335516	18	28048065	MEP1B	5.99E-06

GWAS SNPs are in imperfect linkage disequilibrium (LD) with disease-causing variants²⁶. Again, this points to the possible importance of rare variants, as allele frequency differences between causative alleles and genotyped SNPs impact LD, and may also implicate a potential role for structural variants (*e.g.*, large deletions or duplications), which are also only partially represented by neighboring SNPs, especially those that are multi-allelic and in regions of the genome characterized by segmental duplication²⁷. Imputation based methods to increase the number of common variants tested can also be applied to datasets such as the one used here, but it has recently been observed in schizophrenia that the application of imputation methods only yielded an approximate 2% increase in heritability estimates²¹.

In conclusion, we estimate that approximately 30% of genetic variation in liability to MS is captured by considering all genotyped SNPs simultaneously. The remaining missing heritability most likely reflects imperfect LD between causal variants and the genotyped SNPs.

Methods

Genotypes for UK MS cases and controls were obtained from GWAS data recently generated by the International Multiple Sclerosis Genetics Consortium and the Wellcome Trust Case Control Consortium 2¹⁶. Estimates of the proportion of variance explained were calculated using the Genome-wide Complex Trait Analysis (GCTA) tool (<http://gump.qimr.edu.au/gcta/>)^{17–21,28}. Genetic relatedness between individuals was conducted by principal component analysis using the GCTA tool; for this step, the threshold used to identify and remove related individuals was set to a pairwise genetic relationship value of >0.025 (no individuals met this criteria). The top 20 eigenvectors from this analysis were then used as covariates in a restricted maximum likelihood analysis, again conducted within the GCTA tool; this was used to estimate the proportion of the variance explained by SNPs at the genome-wide level, and after partitioning SNP data by autosomes, MAFs, and quintiles. Assembly statistics for GRCh37 (hg19) were used to calculate autosome lengths (autosome length/total length of all autosomes). Association analysis of GWAS SNPs was conducted using PLINK (<http://pngu.mgh.harvard.edu/purcell/plink/>)²⁹.

Table 3 | Contribution of associated SNPs from UK GWAS dataset to MS liability after binning by P value

Bin: P value	# of SNPs	Variance Explained	Standard Error
1.00E-03	1195	0.176747	0.007402
1.00E-04	429	0.108225	0.010376
1.00E-05	298	0.069538	0.009827
1.00E-06	244	0.044657	0.008199
1.00E-10	149	0.035719	0.007789



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Author contributions

S.V.R., C.T.W., and G.D. conceived of analysis and analyzed the data. C.T.W. and S.V.R. wrote the manuscript, which was critically revised for important intellectual content by F.B., G.D., and G.G. The study was supervised by S.V.R.

Additional information

Competing financial interests: The authors declare no competing financial interests.

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