A UK Biobank-Based Metadata And Multi-Omics Analysis Approach Exploring Comorbidities And Biomarkers In Complex Disorders Like Multiple Sclerosis



Introduction

- Multiple sclerosis (MS) is a chronic demyelinating disorder affecting ~2 million people globally.
- Comorbidities can worsen outcomes, accelerating disability, reducing life quality, and increasing mortality. The significance of following a healthy lifestyle on comorbidities with a potential impact on MS is less studied.
- We analyzed metadata from MS and non-MS individuals with healthy lifestyles and explored potential MS biomarkers using pQTL analysis of Olink proteomics data from the UK Biobank.

Objective

- Cardiometabolic Comorbidities in MS: Assess associations of hypertension (HT) and hyperlipidemia (HL) with MS in adults who otherwise follow healthy lifestyle practices.
- pQTL Analysis and Biomarker Discovery: Leverage pQTL analysis to identify SNPs influencing protein expression in MS vs. non-MS individuals and evaluate their biomarker potential.
- Understanding Disease Mechanisms: Map protein expression profiles to immune pathways, neuroinflammation, and demyelination processes underlying MS.

Methods

Step	Description	Details
1 Data	Source	UK Biobank (MS-2123, Controls - 500,496)
2 Cohorts	Metadata cohort	MS defined by ICD-10 code G35. Restricted to healthy-lifestyle participants (non/previous smokers, BMI < 30kg/m2, regular exercise). HT & HL as covariates. Cohort size: MS - 265, Control - 99,112.
	Proteomics cohort	MS cases vs. non-MS controls with Olink proteomics data.
		Cohort size: MS = 436, Control = 3003
3 Metadata analysis	Logistic regression	Tested association of MS with HT, HL, age, sex, smoking, alcohol, and HT×HL interaction
4 Proteomics	Differential protein	QC →design matrix → limma. FDR < 0.05 cutoff
	pQTL mapping	Regenie two-step pipeline:
		- Step 1: Polygenic model
		- Step 2: Per variant association
		QC: allele count ≥10, MAF ≥0.01, missing genotype rates ≤10%, HWE p >1e-15
		Significance: genome-wide p < 5x10 ⁻⁸
5 Downstream analysis	Variant annotation	Ensembl Variant Effect Predictor (VEP)
	Pathway enrichment	ClusterProfiler & Signaling Pathway Impact Analysis (SPIA)

- Hyperlipidemia was associated with a 2.2x higher risk of MS disease progression in participants aged 40-55 yrs.
- Males were associated with a lower risk for MS.
- 189 proteins showed significantly altered expression levels: 139 upregulated and 50 downregulated in MS vs. controls.
- The key markers CXCL13, FABP4, GDF15, CCL7 and **CXCL17**, have published evidence in the pathogenesis, prognosis and therapy response of MS.
- Pathway analysis emphasized immune dysregulation, with significant activation of the chemokine and cytokine signaling pathways in MS.
- UK Biobank-enabled integrative data analysis demonstrates the value of combining large-scale phenotype and omics data, and reveal the multifactorial nature of MS involving immune cell trafficking, chronic inflammation, and metabolic dysregulation.

UKBB Participants with all data N = 502,6192,123 MS BMI: $< 30 \text{ kg/m}^2 \text{ (Non-Obese)}$ Smoking Status: Never, Quit Duration of moderate activity: >= 150 mins a week Duration of vigorous activity: >= 70 mins a week 99,112 265 MS Cases Group participants into 40-55 and 56-69 age categories 108 MS Cases 59,066 Controls 40,046 Controls 157 MS Cases (40 - 55)(56 - 69) (40 - 55)(56 - 69)

Figure 01: Study design and participant selection from UK Biobank workflow



Business Development



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Results and Conclusions

Table 01: Association of comorbidities and lifestyle factors with multiple sclerosis risk

Factor	Effect on MS	Statistical outcome
HL (40-55 yrs)	个 risk, 2.2x	$OR \approx 2.20$, p = 0.007
HT	Not significant	$OR \approx 0.88$, p = 0.55
HT×HL	Not significant	OR ≈ 1.46, p = 0.53, Independent effects
Male	↓ risk	$OR \approx 0.46$, p = 5.12e-05
Previous smoking	↑ risk	$OR \approx 1.38, p = 0.047$

Table 02: Protein biomarkers and their associations with multiple sclerosis, highlighting prognostic, pathogenic, and potentially protective roles

Quality control of genotype data

Step 1

Whole genome regression model is fit

to the traits, and a set of genomic

predictions are produced as output

LocusZoom

Generate interactive shareable plots

of GWAS summary statistics

Protein	Association in MS	Interpretation
CXCL13	Prognostic marker; linked to B cell-driven inflammation and disease activity	Predicts aggressive disease/ progression; associated with favorable treatment outcomes
FABP4	Increased disability scores (EDSS)	Especially in pediatric MS and secondary progressive MS
GDF15	Elevated in stable MS	Suggests compensatory, tissue- protective or anti-inflammatory response
CCL7	Significantly upregulated	Promotes immune cell recruitment and lesion development
CXCL17	Appears protective	Modulates immune responses, may limit disease severity

Regenie

Figure 02: pQTL analysis workflow using Regenie

Phenotype + Covariate File

Step 2

Association testing per

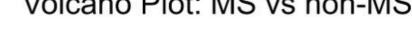
phenotype (protein)

Annotation

Genomic variant

annotations

Age at recruitment Odds Ratio (95% CI) Figure 03: Association of comorbidities and lifestyle factors with multiple sclerosis risk Volcano Plot: MS vs non-MS



Associations with Multiple Sclerosis

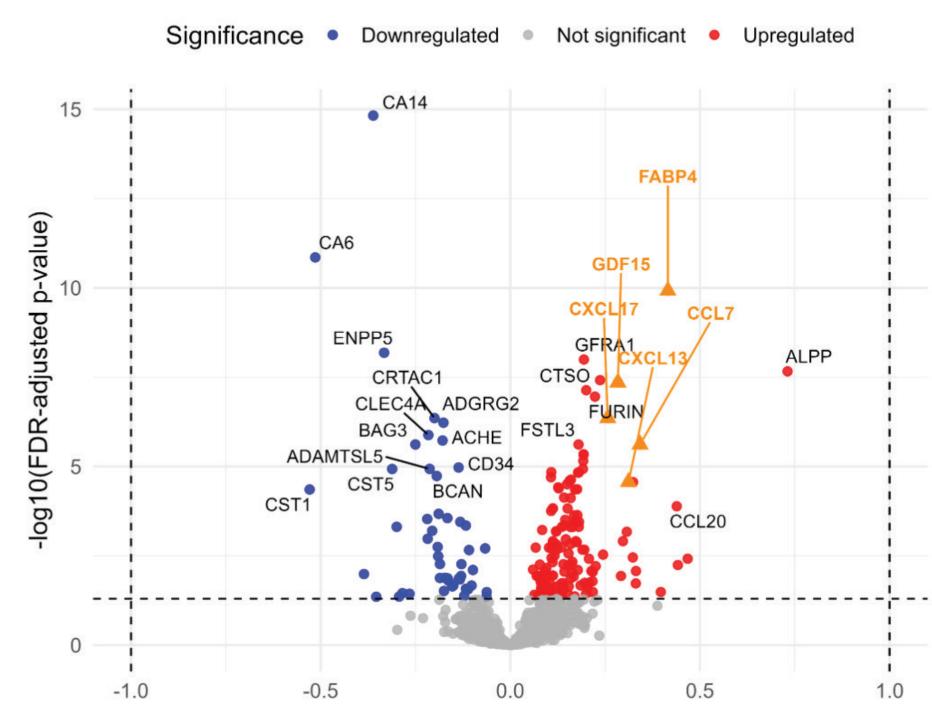


Figure 04: The volcano plot showing the differentially expressed proteins between MS and non-MS

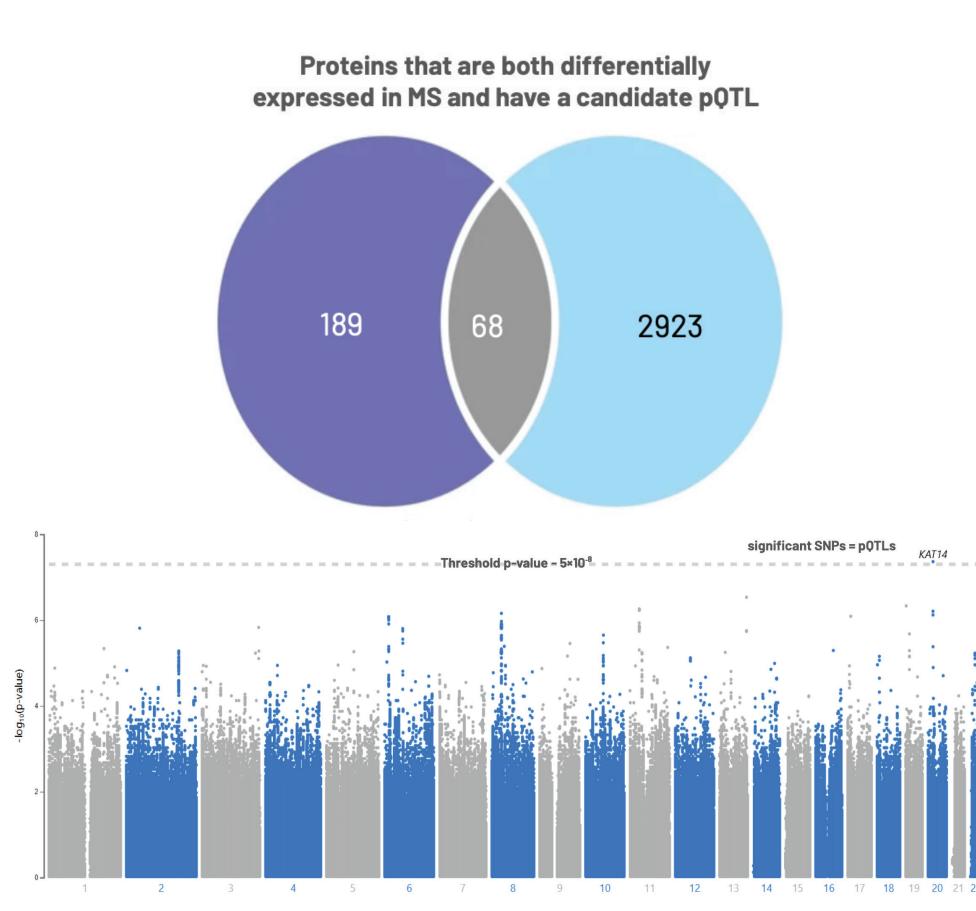


Figure 05: Manhattan plot showing genome-wide association results for pQTLs, with significant SNPs surpassing the threshold (p < 5×10^{-8}), including a notable locus at KAT14