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DOI: 10.1136/annrheumdis-2018-214127

Document Version

Accepted author manuscript

Link to publication record in Manchester Research Explorer

Citation for published version (APA):

Acosta-Herrera, M., Kerick, M., Gonzalez-Serna, D., Wijmenga, C., Franke, A., Gregersen, P. K., Padyukov, L., Worthington, J., Vyse, T., Alarcón-Riquelme, M. E., Mayes, M. D., & Martin, J. (2019). Genome-wide meta-analysis reveals shared new loci in systemic seropositive rheumatic diseases. *Annals Of Rheumatic Diseases*, *78*, 311-319. https://doi.org/10.1136/annrheumdis-2018-214127

Published in:

Annals Of Rheumatic Diseases

Citing this paper

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1	Genome-wide meta-analysis reveals shared new loci in
2	systemic seropositive rheumatic diseases
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5	Marialbert Acosta-Herrera, ^{1†} * Martin Kerick, ^{1†} David González-Serna, ¹
6	Myositis Genetics Consortium, [*] Scleroderma Genetics Consortium, [*] Cisca
7	Wijmenga, ² Andre Franke, ³ Peter K. Gregersen, ⁴ Leonid Padyukov, ⁵ Jane
8	Worthington, ⁶ Timothy Vyse, ⁷⁻⁸ Marta E Alarcón-Riquelme, ⁹ Maureen D.
9	Mayes, ¹⁰ Javier Martin. ¹ *
10	
11	
12 13 14 15 16 17 18 19 20 21 22 23 24	¹ Institute of Parasitology and Biomedicine López-Neyra, IPBLN-CSIC, PTS Granada, Granada, Spain; ² Department of Genetics, University Medical Centre Groningen, University of Groningen, Groningen, The Netherlands; ³ Institute of Clinical Molecular Biology, Christian-Albrechts-University of Kiel, Kiel, Germany; ⁴ Robert S. Boas Center for Genomics and Human Genetics, Feinstein Institute for Medical Research, Manhasset, NY, USA; ⁵ Rheumatology Unit, Department of Medicine, Karolinska Institutet and Karolinska University Hospital, Stockholm, Sweden; ⁶ Manchester NIHR Biomedical Research Centre, The University of Manchester, Manchester Academic Health Science Centre, Manchester, UK; ⁷ Division of Genetics and Molecular Medicine, King's College London, London, UK; ⁸ Division of Immunology, Infection and Inflammatory Disease, King's College London, London, UK; ⁹ Centre for Genomics and Oncological Research (GENYO), Pfizer-University of Granada-Andalusian Regional Government, Granada, Spain; ¹⁰ The University of Texas Health Science Center–Houston, Houston, Texas, USA; [*] Members of the Myositis Genetics Consortium (MYOGEN) and Scleroderma Genetics Consortium are listed in the supplemental note.
25 26 27 28 20	[†] These authors contributed equally to this work.
30	
31	*Corresponding authors
32	
33	Marialbert Acosta-Herrera: Institute of Parasitology and Biomedicine López-Neyra,
34 35	958181621; Fax: (+34) 958181632; email: <u>m.acostaherrera@ipb.csic.es</u>
30 37	Javier Martín: Institute of Parasitology and Riomedicine Lónez-Neura, CSIC, Ay, del
38 39 40	Conocimiento 17. Armilla (Granada), Spain 18016. Phone: (+34) 958181669; Fax: (+34) 958181632; email: javiermartin@ipb.csic.es
41 42	

43 Abstract

Objective: Immune-mediated inflammatory diseases (IMIDs) are heterogeneous and complex conditions with overlapping clinical symptoms and elevated familial aggregation, which suggests the existence of a shared genetic component. In order to identify this genetic background in a systematic fashion, we performed the first crossdisease genome-wide meta-analysis in systemic seropositive rheumatic diseases, namely: systemic sclerosis, systemic lupus erythematosus, rheumatoid arthritis and idiopathic inflammatory myopathies.

51 Methods: We meta-analyzed ~6.5 million single nucleotide polymorphisms (SNPs) in 52 11,678 cases and 19,704 non-affected controls of European descent populations. The 53 functional roles of the associated variants were interrogated using publicly available 54 databases.

Results: Our analysis revealed five shared genome-wide significant independent loci 55 56 that had not been previously associated with these diseases: NAB1, KPNA4-ARL14, DGQK, LIMK1, and PRR12. All of these loci are related with immune processes such as 57 interferon and epidermal growth factor signaling, response to methotrexate, 58 cytoskeleton dynamics, and coagulation cascade. Remarkably, several of the associated 59 loci are known key players in autoimmunity, which supports the validity of our results. 60 All the associated variants showed significant functional enrichment in DNase 61 hypersensitivity sites, chromatin states and histone marks in relevant immune cells, 62 including shared expression quantitative trait loci. Additionally, our results were 63 significantly enriched in drugs that are being tested for the treatment of the diseases 64 under study. 65

66 Conclusions: We have identified shared new risk *loci* with functional value across
67 diseases and pinpoint new potential candidate *loci* that could be further investigated.
68 Our results highlight the potential of drug repositioning among related systemic
69 seropositive rheumatic IMIDs.

71 Introduction

Autoimmunity occurs when the mechanisms related to immune self-tolerance 72 fail, leading to an inappropriate destruction of normal tissue by the immune system. 73 74 Genetic factors play an important role in the development of more than 80 immune-75 mediated inflammatory diseases (IMIDs) identified so far.[1] Comorbidity of these diseases, increased familial clustering, and shared risk variants have been widely 76 documented.[2] However, to date, these shared *loci* have been identified by simple 77 78 comparison between studies, and just recently they have been determined by rigorous and systematic analysis.[3] In this sense, combining genome-wide association studies 79 (GWAS) across several diseases has proven to be a very useful tool for the 80 identification of new genetic risk variants simultaneously associated with several 81 IMIDs, and to expose shared pathways involved in the pathophysiology of these 82 conditions.[4-7] To date, two large studies combining several diseases were recently 83 published following this strategy. One of them was a meta-GWAS across 10 pediatric 84 85 autoimmune diseases with shared population-based controls that revealed new candidate loci with immunoregulatory functions.[8] In the other study, the authors identified new 86 shared associations by combining immunochip data across five chronic inflammatory 87 diseases.[9] 88

89 Systemic seropositive rheumatologic IMIDs, such as systemic sclerosis (SSc), 90 systemic lupus erythematosus (SLE), rheumatoid arthritis (RA) and idiopathic 91 inflammatory myopathies (IIM), are heterogeneous diseases of the connective tissue 92 that share clinical and epidemiological manifestations, as well as life-threatening 93 complications.[10] The common genetic component of these conditions has not been 94 previously assessed systematically, although the overlap of associated genes is elevated 95 when performing a pairwise comparison.[8] Autoantibody production is the main

96 feature of these diseases, comprising additionally a broad deregulation of the innate and 97 adaptive immune response. However, the low prevalence of most of these diseases 98 hinders the collection of large datasets that makes possible to attain sufficient statistical 99 power. Therefore, our study aimed to combine previously published GWAS datasets – 100 all from European descent populations– to identify shared genetic etiologies among 101 systemic seropositive rheumatologic IMIDs in a systematic fashion.

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103 Subjects and Methods

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104 Study population
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105 A total of 12,132 affected subjects with four systemic seropositive rheumatic 106 IMIDs (SSc, SLE, IIM, and RA) and 23,260 controls were included in this study from 107 previously published GWAS [11-16] (Table S1).

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109 Data quality control and imputation

Unified quality control (QC) of the 18 case-control collections was conducted separately, based on stringent criteria using PLINK v.1.07.[17] Given that related and/or duplicated subjects may have been recruited for different studies, genome-wide relatedness was assessed and one individual from each pair was removed. Samples with <95% of successfully called genotypes were removed.

Further, single nucleotide polymorphisms (SNPs) with genotyping call rate <98%, minor allele frequencies (MAF) <1% and deviating from Hardy-Weinberg equilibrium (HWE) with a *p*-value <0.001 in the control group were removed. To control for possible population stratification, we performed principal component (PC) analysis using GCTA64 and R-base software under GNU Public license v.2. 120 Imputation of autosomal SNPs was conducted in the Michigan Imputation 121 Server using Minimac3.[18] The software SHAPEIT[19] was used for haplotype 122 reconstruction and the Haplotype Reference Consortium r1.1 was used as the reference 123 population.[20]

124

125 Statistical analyses

126 *Disease-specific association testing:* Association testing for allele dosages was 127 performed by logistic Wald test using EPACTS software, [21] adjusting by the first two 128 or five PCs as appropriate to control for the genomic inflation factor in European 129 population (λ <1.05) (Table S1). SNPs with a MAF \geq 1% and squared correlation (Rsq) 130 \geq 0.3 were maintained in the analyses as suggested by the imputation software. 131 Additionally, we calculated a concordance rate by comparing imputed and true 132 genotypes.

Cross-phenotype meta-analysis: to identify shared loci, the summary-level statistics 133 134 were meta-analyzed using METASOFT.[22] A fixed-effects model was applied for 135 those SNPs without evidence of heterogeneity (Cochran's Q test *p*-value Q > 0.05), and random-effects model was applied for SNPs displaying heterogeneity of effects between 136 studies (Q \leq 0.05). Genome-wide significance was established at a *p*-value \leq 5 × 10⁻⁰⁸. 137 138 SNP independence was assessed with the software GCTA-COJO (Table S2).[23, 24] To annotate the independent signals SNPnexus[25] was used to the build37 genomic 139 140 coordinates.

Model search to identify the diseases contributing to the association: to identify the diseases most likely contributing to the association signals, we performed an exhaustive disease-subtype model search with the R statistical package ASSET.[24] The

144 contribution of a disease was considered if at least two independent case-control145 collections from the same disease were grouped with consistent effects.

Novelty of the variants: Our independent SNP associations were classified into "known"
or "new" associations based on the information retrieved from the NHGRI-EBI GWAS

148 catalog and the Phenopedia and Genopedia from HuGE Navigator.[26]

Functional enrichment analysis: in order to systematically characterize the functional, 149 cellular and regulatory contribution of the associated variants, a non-parametric 150 151 enrichment analysis implemented in GARFIELD was performed.[27] Furthermore, the online tools HaploReg v.4.1[28] and the Genotype-Tissue Expression project 152 (GTEx)[29] were queried to determine whether any of the lead associated variants was 153 an expression quantitative trait locus (eQTL). The online tool Capture HiC plotter was 154 used to assess physical interactions between restriction fragments containing the 155 156 variants and the promoter of genes in the three-dimensional nuclear space.[30]

Drug Target Enrichment Analysis: the target genes of the eQTLs were used to model a protein-protein interaction (PPI) network using String v10.[31] These protein products were then used to query the OpenTargets Platform[32] for drug targets. Moreover, this platform was used to search for drugs indicated or in different phases of drug development for the treatment of SSc, SLE, IIM and RA. The Fisher's exact test was used to calculate if the results of the meta-analysis were significantly enriched in pharmacologically active drug targets.

Additional details of the Methods section are available in the online supplementarymethods.

166

167 **Results**

168 Cross-phenotype meta-analysis and disease contribution

Following sample QC and imputation, a total of 11,678 cases and 19,704 nonoverlapping controls were included in the genome-wide meta-analysis of 6,450,125 SNPs across the four diseases. The mean concordance rate among imputed and true genotypes was 0.999±0.0003. The final λ showed minimal evidence of population stratification in the meta-abalysis (λ =1.025). Moreover, we calculated λ 1,000 with consistent results (λ 1,000=1.025). Summary of sample/variant QC and QQ plots are shown in Table S1 and Figure S1, respectively.

The global meta-analysis revealed 42 non-hla significantly associated *loci*. Subsequent conditional analyses showed that 27 SNPs were independent (Figure 1 and Figure S2). Sixteen variants were meta-analyzed under a fixed effects model, whereas eleven with random effects based on study heterogeneity.

To comprehensively explore the combinations of diseases contributing to the associations we applied a subset-based meta-analysis implemented in ASSET.[24] Our model search yielded 26 SNPs associated with at least two IMIDs (Table 1). All of these variants were imputed in at least one dataset.

184

Among these 26 associations we found several key players in autoimmunity; 185 interestingly ten of these associations (38%) have never been reported before for SSc, 186 187 eight (31%) for SLE and RA, respectively, and 20 (77%) for IIM. Remarkably, five SNPs have not been reported previously for any of the diseases under study and thus 188 constitute new shared risk *loci* in systemic seropositive rheumatic IMIDs (Table 1). 189 Amongst these five new associations we found the SNP rs744600 in the 3' region of the 190 NGFI-A binding protein 1 (NAB1) (Odds ratio [OR] for the T allele 0.88, Confidence 191 Interval [CI]=0.85-0.92), *p*-value=7.07x10⁻¹¹), and the intronic SNP rs13101828 192 mapping in the gene Diacylglycerol kinase theta (DGKQ) (OR for the G allele 1.11, 193

194	95%CI: 1.07–1.16, <i>p</i> -value= 1.32×10^{-08}). Of note, both genes have been previously
195	associated with a chronic autoimmune liver disease.[33, 34] The intergenic SNP
196	rs112846137, maps between the genes Karyopherin subunit alpha 4 (KPNA4) and the
197	ADP ribosylation factor like GTPase 14 (ARL14) (OR for the T allele 1.29, 95%CI:
198	1.07–1.56, <i>p</i> -value=1.42x10 ⁻⁰⁸). Interestingly, the gene <i>ARL14</i> showed a suggestive
199	association in a pharmacogenomic GWAS of response to methotrexate in RA
200	patients.[35] In addition, we observe the associated SNP rs193107685 located in the 3'
201	region of the LIM domain kinase 1 (LIMK1) gene (OR for the C allele 1.52, 95%CI:
202	$1.27-1.83$, <i>p</i> -value= 3.81×10^{-09}). The protein encoded by this gene regulates actin
203	polymerization, a critical process in the activation of T cells.[36] Finally, the SNP
204	rs76246107 is located in an intron of the gene Proline rich 12 (PRR12) (OR for the G
205	allele 1.28, 95%CI: 1.14–1.43, <i>p</i> -value= 3.36×10^{-08}), which was associated with
206	fibrinogen concentration,[37] and is an active regulator of the inflammatory
207	response.[38]

Chr	Position ^a	SNP	Gene ^b	Functionality ^c	Effect Allele	OR (CI 95%)	Meta-Analysis <i>p</i> -value ^d	Cochran's <i>p</i> -value	Contributing Disease ^e
1	67802371	rs6659932	IL12RB2	Intronic	С	0.85 (079-0.91)	6.08x10 ⁻¹¹	$\frac{1}{1.02 \times 10^{-02}}$	IIM, SLE, SSc
1	114303808	rs6679677	PHTF1-RSBN1	Intergenic	А	1.34 (1.21-1.49)	2.30x10 ⁻²⁸	2.14x10 ⁻⁰⁴	IIM, RA, SLE
1	114377568	rs2476601	PTPN22	Coding (missense)	G	0.75 (0.67-0.83)	1.74x10 ⁻²⁸	1.06x10 ⁻⁴	IIM, RA, SLE
1	114433946	rs1217393	AP4B1	Intronic	А	0.89 (0.85-0.92)	5.21x10 ⁻⁰⁹	4.91x10 ⁻¹	IIM, RA, SLE, SSc
1	173337747	rs2422345	TNFSF4-LOC100506023	Intronic	А	1.11 (1.05-1.18)	2.55x10 ⁻⁰⁸	6.00×10^{-03}	IIM, SLE, SSc
1	183532580	rs17849502	NCF2	Coding (missense)	Т	1.36 (1.16-1.59)	3.93x10 ⁻¹⁵	2.84x10 ⁻⁰⁴	IIM, SLE
2	191564757	rs744600	NAB1*	3'Downstream	Т	0.88 (0.85-0.92)	7.07×10^{-11}	7.60×10^{-1}	IIM, RA, SLE, SSc
2	191933283	rs13389408	STAT4	Intronic	С	1.27 (1.20-1.34)	3.10x10 ⁻¹⁷	3.99x10 ⁻¹	IIM, SLE, SSc
2	191973034	rs10174238	STAT4	Intronic	А	0.73 (0.67-0.80)	2.76x10 ⁻⁴²	4.31x10 ⁻⁰⁷	IIM, SLE, SSc
3	58183636	rs35677470	DNASE1L3	Coding (missense)	А	1.22 (1.14-1.30)	4.96x10 ⁻⁰⁹	6.78×10^{-01}	IIM, SLE, SSc
3	160312921	rs112846137	KPNA4-ARL14*	Intergenic	Т	1.27 (1.17-1.37)	1.42×10^{-08}	9.55×10^{-01}	IIM, RA, SLE, SSc
4	965720	rs13101828	DGKQ*	Intronic	G	1.11 (1.07-1.16)	1.32×10^{-08}	2.29×10^{-01}	IIM, RA, SLE, SSc
5	150438477	rs4958880	TNIP1	Intronic	А	1.16 (1.10-1.22)	1.45×10^{-11}	2.61×10^{-01}	IIM, RA, SLE, SSc
5	159887336	rs2431098	PTTG1-MIR3142HG	Intergenic	G	1.12 (1.05-1.20)	4.91×10^{-12}	1.42×10^{-01}	SLE, SSc
6	106569270	rs802791	PRDM1-ATG5	Intergenic	С	0.87 (0.83-0.92)	3.65×10^{-12}	1.13×10^{-01}	SLE, SSc
6	138243739	rs58721818	TNFAIP3	3'Downstream	Т	1.64 (1.46-1.84)	4.64×10^{-23}	1.65×10^{-01}	IIM, SLE, SSc
7	73537902	rs193107685	LIMK1*	3'Downstream	С	1.52 (1.27-1.83)	3.21×10^{-09}	1.18×10^{-01}	RA, SLE, SSc
7	128589633	rs10954214	IRF5	3UTR	Т	1.18 (1.13-1.23)	6.63x10 ⁻¹⁷	3.64×10^{-01}	IIM, RA, SLE, SSc
7	128647942	rs13238352	TNPO3	Intronic	Т	1.44 (1.30-1.60)	1.47×10^{-38}	2.12×10^{-01}	SLE, SSc
8	11341880	rs2736337	FAM167A-BLK	Intergenic	С	1.23 (1.17-1.30)	4.86x10 ⁻²²	1.29×10^{-01}	IIM, RA, SLE, SSc
11	633689	rs7929541	SCT-DRD4	Intergenic	G	0.89 (0.83-0.95)	2.14×10^{-10}	4.98x10 ⁻⁰⁴	IIM, RA, SLE, SSc

Table 1. Twenty-six independent variants associated at a genome-wide significance level ($p < 5 \times 10^{-8}$) in the meta-analysis.

12	112871372	rs11066301	PTPN11	Intronic	Т	1.11 (1.07-1.15)	4.20×10^{-08}	5.86x10 ⁻⁰¹	IIM, SLE, SSc
16	85994484	rs35929052	IRF8	Intergenic	Т	0.83 (0.78-0.88)	1.71×10^{-09}	4.69×10^{-01}	IIM, SLE, SSc
19	10462513	rs11085725	ТҮК2	Intronic	А	0.88 (0.83-0.92)	2.65×10^{-10}	1.86x10 ⁻⁰¹	IIM, SLE, SSc
19	50121274	rs76246107	PRR12*	Intronic	G	1.28 (1.14-1.43)	3.36x10 ⁻⁰⁸	1.50×10^{-02}	IIM, SLE, SSc
22	21985094	rs5754467	YDJC	5'Upstream	G	1.20 (1.13-1.27)	1.24×10^{-13}	8.59×10^{-02}	IIM, RA, SLE, SSc

^aAccording to NCBI build GRCh37/hg19.

^bVariant localization based on the nearest gene.

^cFunctionality obtained from SNPnexus.²³

^dResults of meta-analysis either under a fixed effect if no heterogeneity was found based on Cochran's Q test (p-value ≥ 0.05) or under a random effect if heterogeneity was found among studies.

^eDisease contributing to the association observed by the subset meta-analysis method with ASSET.²⁵ The diseases for which this locus has never been reported before at genome-wide significance level are shown in boldface.

*Denotes novel *loci* in the study.

Chr: chromosome; OR: odds ratio; CI: confidence interval; IIM: idiopathic inflammatory myopathy; RA: rheumatoid arthritis; SLE: systemic lupus erythematosus; SSc: systemic sclerosis.

All the variants in the table were imputed in at least one of the 18 case-control collections.

209 Associated loci and their functional enrichment on regulatory elements

To assess whether the associated variants lie in coding and non-coding 210 regulatory and cell-type-specific elements of the genome, we performed an enrichment 211 analysis with GARFIELD.[39] The results obtained showed marked enrichment 212 patterns mainly in blood cells and skin cells, with 247 significant enrichments 213 $(p \le 5 \times 10^{-05})$ (Figure S3 and Table S3). Table 2 summarizes the main enrichment results. 214 We found that the majority of associated variants were enriched in DNase I 215 216 hypersensitivity site (DHS) hotspots in blood, as depicted in Figure 2. This functional category included a repertoire of cells from the immune system, such as B-lymphocytes 217 (Fold enrichment (FE)=11.68, empirical p (*pemp*)<1×10⁻⁰⁵), T-lymphocytes (FE=10.42, 218 $pemp < 1 \times 10^{-05}$), including T helper cells (FE=7.81, $pemp < 1 \times 10^{-05}$), T CD8+ (FE=7.61, 219 $pemp < 1 \times 10^{-05}$), natural killer cells (FE=11.36, $pemp < 1 \times 10^{-05}$), and monocytes 220 (FE=8.99, $pemp < 1 \times 10^{-05}$). In line with this enrichment, disease-associated SNPs were 221 enriched in enhancers (FE=14.99, $pemp<1\times10^{-05}$), within TSS (FE=14.87, 222 $pemp<1\times10^{-05}$), and on transcription factor binding sites (FE=12.20, $pemp<1\times10^{-05}$) in 223 the B-lymphocyte cell line GM12878. Additionally, the highest enrichment was 224 observed in the histone modification H3K9ac (FE=14.02, $pemp<1\times10^{-05}$), and 225 H3K27ac (FE=10.81, $pemp<1\times10^{-05}$) in the B-lymphocyte cell line, which are 226 positively associated with gene activation. Although these modifications are increased 227 in the promoters of active genes, the latter has been shown to be associated with active 228 enhancers.[40] Moreover, enrichment was observed in H3K4me1,2,3 sites, which 229 usually surround TSS and are also positively correlated with gene expression.[40] 230

Category ^a	Tissue	Cell types	Туре	NAnnotThesh ^b	NAnnot ^c	NThresh ^d	N (LD- pruned variants) ^e	Fold Enrichment	Empirical <i>p</i> -value
Charamatin States	Dlaad	GM12878	Enhancer	13	10,944	33	416,420	14.99	<1x10 ⁻⁵
Chromatin_States	Blood	GM12878	TSS	12	10,182	33	416,420	14.87	<1x10 ⁻⁵
Footprints	Blood	GM06990	Footprints	8	3,153	33	416,420	32.02	<1x10 ⁻⁵
		GM12878	H3K9ac	21	18,903	33	416,420	14.02	$<1x10^{-5}$
		GM12878	H3K27ac	22	25,674	33	416,420	10.81	$<1x10^{-5}$
	Blood	GM12878	H2AFZ	22	25,824	33	416,420	10.75	<1x10 ⁻⁵
Histone modifications		GM12878	H3K4me3	17	25,365	33	416,420	8.46	<1x10 ⁻⁵
		GM12878	H3K4me2	23	34,807	33	416,420	8.34	5x10 ⁻⁵
		GM12878	H3K4me1	25	39,871	33	416,420	7.91	<1x10 ⁻⁵
		GM12878	H3K79me2	16	25,683	33	416,420	7.86	<1x10 ⁻⁵
II	Blood	GM06990	Hotspots	23	24,839	33	416,420	11.68	<1x10 ⁻⁵
Hotspots	Skin	NHEK	Hotspots	25	54,667	33	416,420	5.77	<1x10 ⁻⁵
Peaks	Blood	GM06990	Peaks	13	6,433	33	416,420	25.50	$<1x10^{-5}$
TFBS	Blood	GM12878	TFBS	19	19,650	33	416,420	12.20	<1x10 ⁻⁵

Table 2. Summary of the most enriched functional annotations for the SNPs associated in the meta-analysis at a genome-wide significance threshold ($p < 5 \times 10^{-8}$).

^aFunctional categories from the Encode²⁸ and Roadmap Epigenomics.²⁹

^bNumber of LD-pruned annotated variants passing the meta-analysis threshold.

°Number of LD-pruned annotated variants in the reference dataset UK10K project.

^dNumber of LD-pruned variants passing the meta-analysis threshold.

^eNumber of LD-pruned variants in the reference dataset UK10K project.

GM12878: B-Lymphocyte; GM06990: B-lymphocyte, lymphoblastoid; NHEK: Normal Human Epidermal Keratinocytes; LD: Linkage disequilibrium; TSS: Transcription Start Site; TFBS: Transcription Factor Binding Sites.

231 Expression quantitative trait loci (eQTL) and associated variants

In silico analysis of eQTLs revealed the role of 16 of the lead SNPs as eQTLs in 232 whole blood, lymphoblastoid cell lines, transformed lymphocytes, skeletal muscle and 233 transformed fibroblasts derived from European individuals from HaploReg v.4.1[28] 234 (Table 3 and Table S4). Focusing on new associated variants, the SNP rs744600 235 modifies NAB1 gene expression in lymphoblastoid cell lines ($p=1.30 \times 10^{-34}$), whereas 236 the T allele increases *HIBCH* expression in skeletal muscles ($p=8.09 \times 10^{-07}$). The G 237 allele of rs13101828 increases DGKQ expression in whole blood $(p=3.29 \times 10^{-45})$, 238 lymphocytes ($p=5.23 \times 10^{-19}$), fibroblasts ($p=4.44 \times 10^{-06}$), lung cells ($p=8.42 \times 10^{-28}$) and 239 several other tissues. The A allele of rs76246107 can reduce ALDH16A1 expression in 240 lung cells ($p=6.45 \times 10^{-06}$), and the protein encoded by this gene is involved in 241 oxidoreductase activity. Reassuringly, 14 of the 16 (87%) reported eQTLs showed a 242 243 physical interaction between the SNP and the promoter of 15 of the genes affected by the eQTLs (Table 3), as suggested by Capture HiC (C-HiC) data (Table S5). These 244 245 independent evidences propose a mechanistic approach to understand the modulation of 246 gene expression.

	,	1 1	-	5	
SNP	Allele	Source	Gene	Tissue	<i>p</i> -value
rs6659932*	С	GTEx2015_v6	IL12RB2	Whole blood	3.72×10^{-11}
rs6679677*	А	Westra 2013	PTPN22	Whole blood	$4.84 \mathrm{x10}^{-10}$
rs2476601*	G	Westra2013	PTPN22	Whole blood	3.36×10^{-10}
		GTEx2015_v6	AP4B1	Skeletal muscle	5.45x10 ⁻⁰⁷
		GTEx2015_v6	HIPK1	Whole blood	7.71x10 ⁻⁰⁹
rs1217393*	А	Westra 2013	PHTF1	Whole blood	9.56x10 ⁻⁰⁵
		Westra 2013	PTPN22	Whole blood	2.67×10^{-10}
		Westra 2013	RSBN1	Whole blood	1.41×10^{-10}
	т	GTEx2015_v6	HIBCH	Skeletal muscle	8.09x10 ⁻⁰⁷
rs/44000*	1	Lappalainen2013	NAB1	Lymphoblastoid cell line	1.30x10 ⁻³⁴
ma12220409	C	GTEx2015_v6	CIS	Skeletal muscle	3.42×10^{-09}
1813389408	C	Westra 2013	GLS	Whole blood	2.98x10 ⁻⁰⁷
rs35677470*	А	GTEx2015_v6	РХК	Skeletal muscle	7.08x10 ⁻⁰⁶
				Whole blood	9.28x10 ⁻⁴⁵
12101020			D CWO	Transformed lymphocytes	1.21×10^{-23}
rs13101828	G	GTEx2015_v6	DGKQ	Transformed fibroblasts	9.78x10 ⁻⁰⁷
				Lung	8.42×10^{-28}
rs4958880*	А	Westra 2013	TNIP1	Whole blood	1.09×10^{-03}
1005 421 4*	т	GTEx2015_v6	10 55	Whole blood	2.56x10 ⁻¹⁶
rs10954214*	1	Lappalainen2013	IKFS	Lymphoblastoid cell line	7.54x10 ⁻³¹
rs13238352*	Т	Lappalainen2013	IRF5	Lymphoblastoid cell line	2.88x10 ⁻¹³
			FAM167A	Whole blood	2.90x10 ⁻²⁶
	C	OTE2015(FAM167A	Transformed fibroblasts	1.90x10 ⁻¹⁸
rs2/3633/*	C	G1EX2015_V6	FAM167A	Transformed lymphocytes	2.10×10^{-15}
			BLK	Whole blood	5.30x10 ⁻¹³

Table 3. Summary of the eQTL results in European samples for the SNPs independently associated in the meta-analysis.

rs2736337*	C	CTE2015(BLK	Transformed fibroblasts	1.30×10^{-11}
	C	G1Ex2015_V6	BLK	Transformed lymphocytes	3.30x10 ⁻⁰⁶
rs7929541*	С	GTEx2015_v6	TMEM80	Transformed fibroblasts	1.22×10^{-11}
rs11085725*	т	CTE: 20156	TYK2	Whole blood	2.30x10 ⁻⁰⁶
	1	GTEX2013_V0	TMED1	Whole blood	8.80x10 ⁻⁰⁶
rs76246107*	А	GTEx2015_v6	ALDH16A1	Lung	6.45x10 ⁻⁰⁶
rs5754467*	G	GTEx2015_v6	UBE2L3	Whole blood	4.68x10 ⁻⁰⁶

New associated SNPs found in our meta-analysis are shown in boldface: rs744600 and rs13101828 associated with Systemic Sclerosis, Systemic Lupus Erythematosus, Rheumatoid Arthritis and idiopathic inflammatory myopathy; rs76246107 associated with Systemic Sclerosis, Systemic Lupus Erythematosus and idiopathic inflammatory myopathy. *Designates those SNPs where a physical interaction has been observed in Promoter Capture HiC data in relevant immune cells.

Genetic associations have the potential to improve the rates of success in the 248 development of new therapies.[41] We assessed if the protein-products from disease 249 associated eOTLs and their direct protein-protein interaction (PPI) partners were 250 enriched with pharmacologically active targets (Table S6 and Table S7). We identified 251 as eQTLs and PPIs 608 proteins for SSc, 630 for SLE, 632 for IIM, and 413 for RA, 252 based on data on drugs at any stage of development collected from the Open Targets 253 254 Platform (Table S8).[32] Using this information, we found for SSc that 23 out of 73 (32%) proteins are targeted by drugs being studied for the disease (OR=16.80, p-255 value=1.41x10⁻¹⁸). Similarly, 7 out of 25 (28%) proteins related to IIM and 13 out of 256 146 (9%) proteins related to SLE are addressed by drugs in consideration for IIM and 257 SLE (OR=13.40, p-value=4.62x10⁻⁰⁶, OR=3.38, p-value=2.85x10⁻⁰⁴, respectively) 258 259 (Table S9).

260

261 **Discussion**

In the present study we identified five unreported shared *loci* associated with systemic seropositive rheumatic IMIDs. This is the first large-scale meta-analysis, including more than 11,000 cases and 19,000 non-overlapping controls aiming to improve our knowledge regarding the genetic resemblances among these conditions.

Our results show that 85% of the associated variants were shared by at least three diseases. Interestingly, for several known RA susceptibility *loci* the contribution of RA was limited. In this case, most of the associated variants were independent to the ones previously reported. Among the new associated SNPs, the signals mapping to *NAB1*, *DGKQ* and *KPNA4-ARL14* were associated to all of the diseases under study. NAB proteins are known to interact with early growth response (EGR) family members

and act as corepressors induced by type I interferons (IFN).[42] The 'IFN signature'-272 has been previously described in these diseases.[43-46] Interestingly, two IFN 273 regulatory factors -IRF5 and IRF8- previously associated to the conditions under study, 274 were associated in the meta-analysis. Additionally, the associated SNP is an eOTL in 275 lymphoblastoid cell line, which evidences its role in disease pathogenesis. The DGKQ 276 protein mediates cell signal transduction and can indirectly enhance the epidermal 277 growth factor receptor (EGFR) signaling activity.[47] This pathway regulates cell 278 279 proliferation and migration, and its expression is augmented in the vasculature of SSc patients with pulmonary involvement.[48] Moreover, the risk allele was associated with 280 an increased expression of the gene in lymphocytes, fibroblasts and lung. In the same 281 line, this gene was associated with Sjögren's syndrome, a related connective tissue 282 disease.[49] The protein encoded by the gene ARL14 is a GTPase involved in the 283 284 recruitment of MHC class II containing vesicles and control the movement of dendritic cells (DCs) along the actin cytoskeleton.[50] The protein LIMK1 regulates many actin-285 dependent processes, including the assembly of the immune synapse between T cells 286 and antigen presenting cells, an expected biological process involved in seropositive 287 IMIDs. Remarkably, rs193107685 and rs112846137 interact physically with the 288 289 promoters of the genes *LIMK1* and *ARL14*, respectively, in DCs (Figure S4). The gene *PRR12* has been previously associated with fibrinogen concentrations.[37] Fibrinogen is 290 considered a high-risk marker for vascular inflammatory diseases and is considered an 291 292 accurate predictor of cardiovascular diseases.[38, 51] Moreover, this molecule is an active player in the coagulation cascade, responsible for the spontaneous formation of 293 fibrin fibrils. Cardiovascular events and fibrosis are the most life-threatening 294 complications described in SSc, IIM, and SLE.[52-54] 295

The associated SNPs are highly enriched in functional categories in B and T 296 cells, natural killer and monocytes, highlighting the relevance of these cells in systemic 297 seropositive rheumatic IMIDs. Beyond whole blood, the skin is the other tissue with 298 significant functional categories, which is not surprising given the nature of these 299 connective tissue diseases. Moreover, epithelial cells could transdifferentiate into 300 mesenchymal cells and eventually contribute in fibrotic processes.[55] Moreover, SSc 301 patients are usually stratified according to the extent of skin involvement.[43] On the 302 303 other hand, the histone modifications observed are consistent with the ones reported in previous studies, where histone hyperacetilation have been described in synovial tissues 304 in RA, in B cells in SSc, and in CD4+ T cells in SLE.[40] Finally, the independent 305 306 associated SNPs have significant eQTLs in relevant tissues (Table 3) and in silico data from promoter capture HiC experiments showed the potential mechanisms in which 307 308 most eQTLs modulate gene expression. Interestingly, all new associated SNPs interact with the promoters of surrounding genes, suggesting them as putative candidates with a 309 310 role in the pathophysiology of these conditions (Figure S4 and Table S5).

311 The prevalence of SSc, SLE, and IIM is low and there are no specific treatments for these diseases in comparison with RA; therefore, given our current knowledge on 312 the use of genetic findings in drug target validation and drug repurposing, we evaluated 313 314 if drugs currently indicated for RA had the potential to be used in any of the other IMIDs under study. Our meta-analysis revealed that ten loci overlap with known RA 315 316 risk genes. For instance, the gene-product of TYK2 is targeted directly by Tofacitinib, which inhibits janus kinases (https://www.drugbank.ca/drugs/DB08895) or indirectly 317 through the interleukin 6 (IL-6) family signaling pathway by targeting the IL6 receptor 318 with Tocilizumab (https://www.drugbank.ca/drugs/DB06273). Both drugs are currently 319 indicated for moderate to severe RA patients who respond poorly to disease-modifying 320

anti-rheumatic drugs. As TYK2 is associated with SSc, SLE and IIM, it is a good 321 candidate for therapy repositioning in these diseases. As a proof of concept, Tofacitinib 322 is currently on trial for SLE (clinical trial identifier NCT02535689), SSc 323 (NCT03274076) and Dermatomyositis (NCT03002649). Overall, we found that five of 324 the *loci* identified in our meta-analysis interact with 17 genes that are considered drug 325 targets, six of which are used for the treatment of these diseases (Table 4). Another 326 327 interesting candidate for drug repurposing is Imatinib, a kinase inhibitor that targets ABL1, which interacts with the gene product of BLK, a known locus associated with 328 329 SSc and RA (Table 4). Imatinib is currently being tested for SSc (NCT00555581) and RA (NCT00154336). 330

Associated SNP	Gene product	Association results ^a	Drugs ^b	Targets	Disease indication ^c
			Canakinumab	IL1B	RA
rs6659932	IL12RB2	IIM, SLE, SSc	Anakinra	IL1R1	RA
			Tofacitinib	JAK kinases	RA
rs13389408	GLS	IIM, SLE, SSc	Azathioprine	PPAT	RA, SLE
rs13101828	DGKQ	IIM, RA, SLE, SSc	Orlistat	LIPF	
			Nintedanib	PDGFRB	SSc
			Dasatinib	BLK	
			Imatinib ABL1	ABL1	
rs2736337	FAM167A-BLK	M167A-BLK IIM , RA, SLE, SSc	Osimertinib	EGFR	
			Vandetanib	EPHA1	
			Fingolimod	S1PR1	
			Bosutinib	SRC	
			Tofacitinib	JAK kinases	RA
			Tocilizumab	IL6R	RA
rs11085725	TYK2	IIM, SLE, SSc	Interferon Apha-2B	IFNAR1	
			Idelalisib	PIK3CD	
			Ruxolitinib	JAK1	

Table 4. Summary of the plausible target gene products with drug indications in systemic IMIDs.

^aBased on our meta-analysis, diseases contributing to the observed association. The diseases where the association of this variant has never been reported before at genome-wide significance level are shown in boldface.

^bDrugs from the OpenTarget platform with their corresponding target.

[°]Current indication of the reported drug. Non-immune mediated diseases were omitted. SSc: Systemic sclerosis; IIM: Idiopathic inflammatory myopathy; SLE: Systemic lupus erythematosus; RA: Rheumatoid arthritis.

As compared to previous cross-phenotype studies of autoimmune diseases, our 331 study has the strength of analyzing systemic seropositive rheumatic diseases, which is a 332 consistent clinical phenotype than in the diseases investigated previously, where mixed 333 seropositive and seronegative diseases were analyzed, and combining systemic and 334 organ-specific diseases.[8, 9] The study of a more homogenous phenotype allowed us to 335 determine that the type I IFN signaling pathway and its regulation play a more 336 prominent role in these conditions than in others, based on the associations observed in 337 338 NAB1, TYK2, PTPN11, IRF5, and IRF8. Additionally, we performed a genome-wide scan to identify shared genetic etiologies, as opposed to the study performed by 339 Ellinghaus et al. whose analyses were limited to the 186 autoimmune disease-associated 340 loci implemented in the Immunochip platform. The study performed by Li et al. -which 341 was also a meta-analysis of GWAS data- was focused on pediatric autoimmune 342 343 diseases, whereas our study was on a new combination of diseases in adult population.

In summary, this is the first study to investigate shared common genetic variation in four systemic seropositive rheumatic IMIDs in adults. We identified 26 genome-wide significant independent *loci* associated with at least two diseases, of which five *loci* had not been reported before. The shared risk variants and their likely target genes are functionally enriched in relevant immune cells and significantly enriched in drug targets, indicating that it may assist drug repositioning among genetically related diseases based on genomics data.

351

352 **Competing interests**

353 The authors declare no competing interests.

354

355 **Contributorship**

- 356 Data providers: F.W.M., W.C., T.P.O, R.G.C., J.V., L.G.R., K.D., L.R.W., I.E.L., L.
- 357 M.P., A.M.R., S.R.Y., A.S.O., T.R.R., D.A.I., H.C., W.E.R.O., P.S., B.P, A.L., J.A.L.,
- 358 C.I.A., C.D., D.H.J., P.P., H.V., on behalf of the Myositis Genetics Consortium; O.G.,
- 359 B.R., J.E.M., B.Z.A., R.P.M., M.J.C., M.C.V., A.E.V., A.J.S., J.C.B, P.L.C.M.R., R.S.,
- 360 A.I., R.A.O., G.R., N.H., C.P.S., N.O.C., M.A.G.G, M.G.E., P.A., J.V.L., A.H., R.H.,
- 361 V.S., F.D.K., F.H., M.M.C., R.M., P.S., R.W., A.K., H.K., E.D.B, T.W., L.K., L.B.,
- 362 R.S., B.A.L., A.M.H.V., P.C., J.V., M.H., A.T.L., J.Y., Y.H., S.F.W., F.M.W., L.H.,
- 363 J.L.N., S.K.A, S.A., P.G., F.K.T., B.P.C.K., F.C.A, on behalf of the Scleroderma
- 364 Genetics Consortium; QC and imputation in the contributing studies: M.A.H., M.K.,
- 365 D.G.S.; Functional and drug enrichment analysis: M.A.H., M.K.; Meta-analysis, tables
- and figures: M.A.H., M.K., D.G.S.; Drafting and approved version of the manuscript:
- 367 M.A.H., M.K., D.G.S., C.W., A.F., P.K.G., L.P., J.W., T.V., M.E.A.R., M.D.M., J.M.,
- 368 F.W.M., W.C., T.P.O, R.G.C., J.V., L.G.R., K.D., L.R.W., I.E.L., L. M.P., A.M.R.,
- 369 S.R.Y., A.S.O., T.R.R., D.A.I., H.C., W.E.R.O., P.S., B.P, A.L., J.A.L., C.I.A., C.D.,
- 370 D.H.J., P.P., H.V.; Study design and management: M.A.H., M.K., J.M.
- 371

372 Acknowledgements

We would like to thank Sofia Vargas, Sonia García, and Gema Robledo for their excellent technical assistance, as well as Dr. Carlos Flores, Dr. Maria Pino-Yanes, Dr. Elaine Remmers and Dr. Doug Bell for their outstanding advice. Additionally, we would like to thank all the patients and healthy controls for their essential collaboration
and the MYOGEN consortium for the provision of data. The results reported in this
study were presented in the Genomics of Common Diseases 2017 conference and in
EULAR 2018 (Acosta-Herrera M, Kerick M, González-Serna D *et al.* Ann Rheum Dis
2018; 77 Supp 2).

381

382 Funding

Funded by EU/EFPIA Innovative Medicines Initiative Undertaking 383 Joint PRECISESADS (115565), The Spanish Ministry of Economy Industry and 384 Competitiveness (SAF2015-66761-P), The Regional Ministry of Innovation, Science 385 386 and Technologies of the Andalusian Regional Government (P12-BIO-1395), and Juan de la Cierva fellowship (FJCI-2015-24028). This research was supported in part by the 387 388 Intramural Research Program of the NIH, National Institute of Environmental Health Sciences. 389

390

391 Ethical approval information

392 This study was conducted using available data included in previously published GWAS393 (Supplementary references 1-6).

394

395 Data availability

Summary statistics of the global meta-analysis generated and analyzed in the currentstudy are available from the corresponding author on reasonable request.

398

399 Key messages

- Systemic sclerosis, systemic lupus erythematosus, rheumatoid arthritis and idiopathic
inflammatory myopathies are systemic seropositive rheumatic diseases that share
symptoms, progressions, environmental risk factors, high rates of familial aggregation,
and susceptibility genes, pointing to a shared genetic architecture.

- The assessment of a shared genetic component among these conditions has not been
performed before in a systematic fashion.

We have identified five new shared *loci* among systemic seropositive rheumatic
immune-mediated inflammatory diseases. The rest of the observed associations
constitute firm susceptibility genes in autoimmunity, providing validity to our findings.

- The associated variants are enriched in marks related to gene activation in immunecells and constitute shared expression quantitative trait *loci*.

For most of these diseases there are no specific treatments, therefore, therapyrepositioning could be possible among genetically related conditions.

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573 Figure Titles and Legends

574

Figure 1. Meta-analysis results for the four systemic immune-mediated inflammatory diseases (IMIDs). The Manhattan plot displays the -log10 transformed p-values (*y*-axis) by position on each chromosome (*x*-axis). The red line depicts the genome-wide significance threshold (*p*-value= $5x10^{-8}$). A total of 26 SNPs were independently associated with at least two systemic IMIDs. Most of the signals map to known susceptibility *loci* in autoimmunity (e.g. *PTPN22, STAT4, TNPO3, FAM167A*-*BLK*) and five *loci* have never been reported before.

582

Figure 2. GARFIELD functional enrichment analyses in DHS hotspots. The wheel plot shows functional enrichment in systemic IMIDs within DHS hotspot regions in ENCODE and Roadmap Epigenomics. The radial axis depicts the fold enrichment (FE) calculated at different meta-analysis *p*-value thresholds. The font size is proportional to the number of cell types from the tissue, mainly enriched in blood cell types including a repertoire of immune cell lines.

590 Supplementary Figures

Figure S1. Distribution of the observed and expected association *p*-values in each
individual study that contributed to the meta-analysis. Quantile-Quantile (QQ) plots
from: A) Systemic sclerosis case-control collections. B) Systemic Lupus
Erythematosus. C) Rheumatoid Arthritis and D) Idiopathic Inflammatory Myopathies.

595

Figure S2. Non-Conditioned and conditioned analysis on the top associated variants from the meta-analysis. In panels where significant variants remained after conditioning, there are several independent variants in the region. In panels E, P, and Y the remaining independent variants were not significant in the meta-analysis.

600

Figure S3. Wheel plots from the functional enrichment analysis with GARFIELD at
 different thresholds of p-values from the meta-analysis. Functional categories from the
 ENCODE project and Roadmap Epigenomics.

604

Figure S4. Circular view of the interactions from the new shared risk SNPs with genes nearby obtained from Promoter Capture HiC data in relevant immune cell types. Interactions are displayed as connecting lines depending on the confidence of the interaction. Grey lines are below threshold in the tissue. Only genes with maximum interaction score are reported.