Full Length

Autoantibodies, autoimmune risk alleles and clinical associations in rheumatoid arthritis cases and non-RA controls in the electronic medical records

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ABSTRACT

<u>Objectives</u>: The significance of non-RA autoantibodies in patients with rheumatoid arthritis (RA) is unclear. We studied associations between autoimmune risk alleles and autoantibodies in RA cases and non-RA controls, and autoantibodies and clinical diagnoses from the electronic medical records (EMR).

<u>Methods:</u> We studied 1,290 RA cases and 1,236 non-RA controls of European genetic ancestry from the EMR from two large academic centers. We measured antibodies to citrullinated peptides (ACPA), anti-nuclear antibodies (ANA), antibodies to tissue transglutaminase (anti-tTG), antibodies to thyroid peroxidase (anti-TPO). We genotyped subjects for autoimmune risk alleles, and studied the association between number of autoimmune risk alleles and number of types of autoantibodies present. We conducted a phenome-wide association study (PheWAS) to study potential associations between autoantibodies and clinical diagnoses among RA cases and controls.

<u>Results:</u> Mean age was 60.7 in RA and 64.6 years in controls, and both were 79% female. The prevalence of ACPA and ANA was higher in RA cases compared to controls (p<0.0001, both); we observed no difference in anti-TPO and anti-tTG. Carriage of higher numbers of autoimmune risk alleles was associated with increasing types of autoantibodies in RA cases (p=4.4x10⁻⁶) and controls (p=0.002). From the PheWAS, ANA was significantly associated with Sjogren's/sicca in RA cases.

<u>Conclusion</u>: The increased frequency of autoantibodies in RA cases and controls was associated with the number of autoimmune risk alleles carried by an individual. PheWAS analyses within the EMR linked to blood samples provide a novel method to test for the clinical significance of biomarkers in disease.

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INTRODUCTION

Presence of anti-citrullinated peptide/protein antibodies (ACPA) is a component of the rheumatoid arthritis (RA) classification criteria(1). ACPA can be found in patient sera more than 10 years prior to the diagnosis of disease(2-5) and portends higher disease severity(6, 7). Other autoantibodies such as anti-nuclear antibodies (ANA), associated with systemic lupus erythematosus (SLE), are found more frequently in RA patients compared to controls(8). However, the prevalence and clinical significance of these autoantibodies in RA is not well established. Moreover, little is known about the factors associated with autoantibody production(9).

The majority of studies utilizing autoantibodies focus on the association between genetic risk alleles and risk of having a disease (e.g., human leukocyte antigen (HLA) genes and ACPA positive RA vs controls)(10, 11), rather than autoantibody formation itself as the outcome of interest. Genetics studies of individuals with ANA, antibodies to thyroid peroxidase (anti-TPO), and antibodies to tissue transglutaminase (anti-tTG) are conducted within the associated autoimmune diseases: systemic lupus erythematosus (SLE), autoimmune thyroiditis (AITD) and celiac disease, respectively. SLE risk alleles are associated with ANA autoantibodies in patients with SLE(12), but the effect of SLE risk alleles on ANA formation in non-SLE patients has not been investigated. The diagnosis of AITD and celiac disease are tightly linked to the presence of anti-TPO and anti-tTG antibodies, and no study has investigated whether AITD and celiac disease risk alleles are associated with the formation of these autoantibodies.

In this study we hypothesize that there is not only a genetic basis for autoantibody formation in patients with RA, but also in controls without clinical evidence of a rheumatic disease. Further, integrating clinical, genetic, and autoantibody data may provide additional information regarding

clinical subsets of RA. Specifically, we hypothesize that (1) RA patients will carry more types of autoantibodies than controls and (2) disease-specific risk alleles will be associated with disease-specific autoantibodies (e.g., SLE risk alleles are associated with ANA autoantibodies) in RA patients and controls without having the associated autoimmune disease. Furthermore, we employ a Phenome-wide association study (PheWAS)(13) as an exploratory analysis where we investigate associations between the presence of autoantibodies and autoimmune risk alleles with clinical diagnoses codes ascertained through the EMR.

PATIENTS AND METHODS

Study population

Our study was conducted on 1,290 RA subjects and 1,236 non-RA controls of European ancestry identified from the electronic medical records (EMR) of Brigham and Women's Hospital (BWH) and Massachusetts General Hospital (MGH), characterized in previously published studies from our group(10, 14). RA subjects were identified using a published RA EMR phenotype algorithm with a positive predictive value (PPV) of 94%(14) trained on a gold standard set of subjects classified either as RA or not RA cases by 3 board certified rheumatologists (KPL, EWK, RMP) using the 1987 American College of Rheumatology Classification Criteria for RA(15) as the benchmark. This algorithm was further validated at two other academic institutions(16). Please refer to Liao, et al., 2010 for details on development, training, and validation of this RA phenotype algorithm(14).

The non-RA control group was created from the EMR by excluding all subjects with an *International Classification of Diseases 9th Revision* (ICD9) code for any rheumatic disease in the EMR (this excluded all subjects in the RA cohort); please refer to Kurreeman, et al., 2011 for details(10). The remaining subjects were matched to RA cases (3:1) by age, gender, self-

reported ethnicity, and level of health care utilization (represented by the number of 'facts,' or contacts with the health care system, i.e. office visits, laboratory blood draws)(17). For both RA cases and controls, information regarding age, gender, ICD9, laboratory test results and electronic prescriptions for medications were extracted from structured EMR data. Bone erosion information was obtained using natural language processing (NLP) on bone radiology reports from RA cases and controls using Health Information Text Extraction (HITex) system(14, 18).

Discarded blood samples from five clinical laboratories at Partners Healthcare (Boston, USA) were collected by the BWH Clinical Specimen Bank from 2009-2010, using an Institutional Review Board (IRB) approved process, as described in Kurreeman, et al., 2010(10). The final RA cases and non-RA control populations analyzed for this study were conducted in those where blood samples were obtained and were of European ancestry determined by ancestry informative markers (AIMs). As a result the RA cases and controls were no longer perfectly matched.

Genotyping

Detailed methods for genotyping and assigning genetic ancestry for the RA case and the noncontrol groups can be found in Kureeman, et al., 2010(10). Briefly, processing and genotyping of the discarded blood samples was performed at the Broad Institute Broad Institute (Cambridge, MA, USA). We genotyped 192 ancestry informative markers (AIMs), 28 single nucleotide polymorphisms (SNPs) associated with RA, 33 SNPs associated with SLE, and 16 SNPs associated with celiac disease (Supplementary Table 2)(19-24). For quality control, we removed SNPs with missing genotype rate >10% and minor allele frequency <1%. Genetic ancestry using the AIMs was determined using the Bayes classifier and principal components analysis.

Aggregate Genetic Risk Scores (GRS)

We calculated a cumulative aggregate genetic risk score for RA, SLE and celiac for each individual using the following formula(10, 25, 26):

$$GRS = \sum_{i}^{n} X_{i}$$

Where *n* is the number of SNPs for the particular disease (RA, SLE, celiac) (Supplementary Table 1), *i* is the SNP, X_i is the number of risk alleles (0, 1, or 2).

The RA GRS excludes the *HLA-DRB1*04* tag SNP because we were interested in understanding the effects of non-HLA risk alleles and production of ACPA in RA. In addition, the associations in HLA region are complex and require dense genotyping not available in this study(27). We created a combined autoimmune (AI) GRS which consists of all risk alleles in the study with the exception of SNPs in linkage disequilibrium with another SNP (Supplementary Table 1). All GRSs were unweighted due to absence of information on the strength of association for any individual risk allele and autoantibody outcome. The literature for AITD was less definitive(28) and we therefore did not construct a GRS for AITD.

Autoantibody measurement

We measured ACPA using the INOVA CCP3 IgG ELISA, ANA using INOVA Quanta-Lite ANA, anti-TPO using INOVA Quanta-Lite TPO, and anti-tTG IgA using the INOVA Quanta-Lite IgA TTG kits. We determined positivity of an autoantibody based on the manufacturer cut-offs: ACPA \geq 20 units, ANA \geq 20 units (high titer positive (ANAht) >60 units), anti-TPO >100 WHO units, anti-tTG \geq 20 units. These autoantibodies were selected because of the relationship

between each autoimmune disease and RA in both epidemiologic(29, 30) and genetic studies(31-33). ANA, anti-TPO and anti-tTG antibodies were measured in all RA cases and controls. ACPA was measured in all RA cases and 202 controls which provided adequate power to detect a difference in ACPA prevalence between the two groups (Supplementary

Table 2).

Statistical analyses

Autoantibody prevalence

We first determined the distribution and overlap of ACPA, ANA, anti-TPO, and anti-tTG among RA cases and non-RA controls. We then constructed separate logistic regression models for each of the autoantibodies to study the association between each autoantibody (ACPA, ANA, anti-TPO, anti-tTG) and RA case or control status, adjusted by age, gender and health care utilization (represented by the number of facts). We then conducted an RA case-only analysis examining whether the prevalence of ANA, anti-TPO and anti-tTG differed with ACPA status using logistic regression models, adjusted by age, gender and health care utilization.

Association between GRSs and autoantibody positivity

We constructed separate logistic regression models to determine the association between the disease specific GRSs (RA, SLE, celiac) and their related autoantibodies (ACPA, ANA, and anti-tTG, respectively) for RA cases. In controls, we tested only the association between the SLE GRS and ANA, due to the low prevalence of ACPA and anti-tTG (post-hoc power calculations, Supplementary Table 2). Linear regression was used to assess the association between the AI GRS and the number of types autoantibodies present (ANA, anti-TPO, anti-tTG) for both RA cases and controls.

We employed multivariate logistic regression models to determine the relative importance of age, gender and disease specific GRS in predicting autoantibody positivity among RA cases. In controls we tested only the association between the SLE GRS and ANA, due to the power issues stated above.

To test whether SLE risk alleles have the same effect on ANA production on RA cases and controls, we categorized all subjects into SLE GRS tertiles (tertile 1: SLE GRS ≥ 0 and ≤ 33.36 , tertile 2: SLE GRS>33.36 and ≤ 40.90 , tertile 3: SLE GRS ≥ 40.9). Within each tertile, we fitted logistic regression models to estimate the odds of ANA positivity in RA cases compared to non-RA controls given a similar number of SLE risk alleles. Each model was adjusted by age, gender and health care utilization. A subgroup analysis was conducted only in subjects without a prescription for a TNFi because TNFi's have been associated with conversion to ANA positivity(34).

Phenome-wide analysis (PheWAS) for association of autoantibodies and autoimmune risk alleles with clinical diagnoses

To discover potential associations between non-RA autoantibody positivity and a clinical diagnosis, and the AI GRS and a clinical diagnosis, we conducted a PheWAS. Clinical diagnoses were defined using ICD9 codes manually grouped into clinically relevant diseases ("PheWAS code"; e.g. ICD9 codes 401-405, representing different types of hypertension and complications are grouped into a "hypertension" PheWAS code) by a team of physicians in a previously published PheWAS(13). We included all PheWAS codes used with a prevalence of ≥1% in each group, yielding 512 diagnoses codes for RA cases and 698 diagnoses codes for controls for analysis.

We tested for potential associations between an autoantibody and PheWAS codes in RA cases and non-RA controls separately by fitting logistic regression models using ACPA, ANA, or anti-TPO status as the predictor (adjusted by age, gender) and the presence or absence (0 or 1) of the PheWAS code as the outcome. We did not study anti-tTG in cases and controls and ACPA in controls due to the low prevalence (\leq 1%) which would lead to unstable estimates. To study the associations between the AI GRS and PheWAS codes, we fitted logistic regression models using the AI GRS as the predictor (adjusted by age, gender) and the presence or absence of the PheWAS code as the outcome. Since multiple comparisons are applied in the PheWAS, suggestive associations can be observed due to chance. Therefore, we set the p-value cut-off for a significant association using the Bonferroni correction: RA cases, significant p-value, $0.05/512=9.76 \times 10^{-5}$; non-RA control group, significant p-value, $0.05/698=7.16 \times 10^{-5}$.

For significant associations between an autoantibody and PheWAS code after Bonferroni correction, we reviewed the medical records to determine the diagnostic accuracy of the code. For example, if autoantibody_b was significantly associated with PheWAS_code_b in RA cases, we randomly selected 20 RA cases who had ≥1 PheWAS_code_b and reviewed their medical record for clinical documentation of this diagnosis by their treating physician (reviewed by KPL). This method was applied for each PheWAS code with a significant association after Bonferroni correction with an autoantibody. We reported the percentage of confirmed diagnoses as the PPV of the PheWAS code.

This study was approved by the Partners' IRB. Statistical analyses were conducted using the SAS and R 2.10 software packages.

RESULTS

Study population and autoantibody prevalence

We studied 1,290 RA cases and 1,236 non-RA controls of European ancestry with both EMR clinical data and DNA/plasma from discarded blood samples. The mean age was 60.7 in the RA cases and 64.6 years in the controls. Both groups were 79% female. The clinical characteristics of RA subjects included: 70.3% ACPA positive, 61.6% bone erosions, 62.3% with \geq 1 methotrexate and 40.9% with \geq 1 tumor necrosis inhibitor (TNFi) electronic prescription at some point during their treatment (Table 1). ACPA was the most prevalent autoantibody in RA (70%) and the least prevalent among the controls (0.5%) (Table 1). Figure 1 demonstrates the distribution and overlap autoantibodies in RA cases and controls.

RA cases were more likely to be ANA positive compared to non-RA controls (Figure 2A) without significant differences in the prevalence of anti-TPO and anti-tTG antibodies in RA cases compared to controls. In an RA case only analysis, we found that ACPA-positive RA cases were more likely to be ANA positive than ACPA-negative cases without significant differences in prevalence for anti-TPO and anti-tTG antibodies (Figure 2B).

Association between GRSs and autoantibody positivity

To determine the genetic basis of autoantibodies in RA cases, we tested RA(23), SLE(19, 35) and celiac disease(24, 36) genetic risk alleles for association with disease-specific autoantibodies (ACPA, ANA and anti-tTG, respectively). We found that the disease-specific GRSs predicted the presence of the disease-specific autoantibodies among RA cases (Figure 3A-C). As we have shown previously for ACPA positivity in RA cases(10) the mean RA GRS was significantly higher in ACPA positive than in ACPA negative subjects ($P=2.5\times10^{-9}$) (Figure 3A). Notably, this RA GRS does not include the HLA shared epitope alleles. As shown in

Figure 3B, we found that ANA positive RA cases had a significantly higher mean SLE GRS than ANA negative cases (P=8.0x10⁻⁴). Anti-tTG RA cases also had a significantly higher mean celiac GRS than anti-tTG negative subjects (P=0.03) (Figure 3C).

In RA cases, the associations between the disease specific GRSs and autoantibody status did not change after adjusting for age, gender and health care utilization in a multivariable logistic regression model, nor were age and gender significantly associated with autoantibody status in the model (data not shown).

In controls, there was no significant association between the SLE GRS and ANA status, although the trend was similar to that in RA cases: ANA positive controls had a trend for higher mean SLE GRS than ANA negative controls (P=0.31) (Supplementary Figure 1A). In contrast to RA cases, age and gender were significantly associated with ANA positivity in controls (age per year: OR 1.01 (95% CI 1.002, 1.03, p=0.02); female gender: OR 1.7, (95% CI 1.2, 2.75, p=0.007) while the SLE GRS was not (p=0.34).

We also combined all non-RA autoimmune risk alleles into a single GRS (AI GRS), and tested for association with the number of autoantibodies present (with and without ACPA). In RA cases, the increasing count of any autoimmune risk allele (AI GRS) was significantly associated with increasing types of any autoantibody (range 0-4, representing ACPA, ANA, anti-TPO, or anti-tTG) (P= 3.2x10⁻⁸). The association remained with the removal of ACPA as a potential outcome: an increasing AI GRS was associated with increasing types of any autoantibody (range 0-3, representing ANA, anti-TPO and anti-tTG) in RA cases (P= 2.6x10⁻⁵) (Figure 3D). In controls, increasing types of any autoantibody (range 0-3, representing ANA, anti-TPO, antitTG) was also significantly associated with the AI GRS (P=4.0x10⁻³) (Supplementary Fig 1B). The association between the AI GRS and types of autoantibodies (range 0-3, representing ANA,

anti-TPO, anti-tTG) remained significant in RA cases ($P= 2.1 \times 10^{-5}$) and controls ($p=5.0 \times 10^{-3}$) after adjusting for age, gender and health care utilization.

One possible explanation for the association between ANA-positivity and RA case status (Figure 2A) is that RA cases carry more SLE risk alleles than controls. While RA cases carried significantly more SLE risk alleles than non-RA controls [mean SLE GRS: RA cases 38.6 (SD 5.3); controls 37.8 (SD 5.4), p= 4.0×10^{-4}], we found that RA cases still had a higher rate of ANA-positivity than non-RA controls after controlling for the effect of SLE risk alleles (Figure 4). More specifically, when comparing RA cases and controls with similar numbers of SLE risk alleles (grouped by tertile of SLE GRS), RA cases were still significantly more likely to be ANA positive. A subgroup analysis comparing RA cases and controls without a TNFi prescription electronically prescribed through the EMR (RA cases, n=762; controls, n=1228) resulted in similar findings: RA subjects were significantly more likely to be ANA positive given the same number of SLE risk alleles as controls (p<0.0001, all tertiles).

Phenome-wide analysis (PheWAS) for association of autoantibodies and autoimmune risk alleles with clinical diagnoses

To determine whether autoantibodies or a higher number of autoimmune risk alleles were associated with codified clinical diagnoses, we performed an exploratory PheWAS in our EMR. From the PheWAS, we observed that in both RA cases and controls, the presence of anti-TPO was significantly associated with hypothyroidism (RA cases, p=1.22x10⁻¹⁶; controls, p=9.22x10⁻¹⁰), which was expected given the known association between anti-TPO and autoimmune thyroid disease (Table 2). The PPV of a PheWAS code for 'acquired hypothyroidism' was 75% for RA cases and 100% for controls on medical record review.

Two PheWAS codes were significantly associated with ANA high titer positivity in RA cases or in controls. In RA cases, the PheWAS code of Sjogren's or sicca syndrome diagnoses was associated with high-titer ANA positivity ($p=8.59x10^{-6}$) (Table 2). The PPV of the PheWAS code for 'Sjogren's/sicca' was 70%. In controls, we observed a significant association between ANA high titer positivity and chronic non-alcoholic liver disease ($p=2.9x10^{-6}$) (Table 2). The PPV of the PheWAS code for 'other chronic non-alcoholic liver disease' was 75% based on medical record review for documentation of liver disease not associated with alcohol.

We found no significant association between ACPA-positivity and PheWAS categories among RA cases (Supplementary Table 3). Furthermore, we observed no significant associations between carriage of higher numbers of autoimmune risk alleles in aggregate using the AI GRS, and specific a PheWAS code corresponding to a clinical diagnosis in the RA cases or controls.

DISCUSSION

RA diagnosis and classification is based on a distinct clinical phenotype(37). Underlying the clinical phenotype of RA are variations in numbers of autoimmune risk alleles and differential production of RA and non-RA autoantibodies. We observed: (1) RA patients have a higher prevalence of ANA, with a trend towards higher anti-tTG (but not anti-TPO) compared to non-RA controls; (2) SLE and celiac disease genetic risk alleles influence the production of the SLE and celiac disease-related autoantibodies (ANA and anti-tTG, respectively) among RA cases; (3) the number of autoimmune risk alleles carried by a patient influence the number of types of autoantibodies carried in both RA cases and controls; (4) given a similar number of SLE risk alleles, RA subjects are more likely to be ANA positive than controls; and (5) applying the PheWAS, we observed that ANA positivity was associated with a diagnosis of Sjogen's syndrome or sicca among RA cases.

To our knowledge, this is the largest study that systematically tested and compared the prevalence of clinical autoantibodies (ACPA, ANA, anti-TPO and anti-tTG) in RA cases and non-RA controls. For the purposes of our study, we defined autoantibodies as being "disease-specific", linking ACPA with RA and linking other "non-RA autoantibodies" with specific autoimmune diseases: ANA (SLE), anti-tTG (celiac disease), and anti-TPO (AITD). However, we show that this definition is not precise, as "non-RA autoantibodies" are actually more common in RA patients compared to controls (i.e., ANA).

We propose two explanations regarding why RA patients are at an increased risk of diseasespecific autoantibodies compared to non-RA controls. Under a purely genetic model, RA patients have a higher prevalence of autoantibodies because they have a higher burden of autoimmune risk alleles. The increased frequency of ANA in RA may be due to the fact that RA and SLE share risk alleles(31, 33, 38), leading to a larger burden of SLE risk alleles in RA patients. Supporting this, we find that RA patients have a significantly higher mean SLE GRS than controls. A consequence of the increased burden of autoimmune risk alleles is that these risk alleles may also have general effects on self-tolerance. For example, the *PTPN22* missense allele (*R620W*), a shared genetic risk factor for RA, SLE and AITD, has been shown to cause defects in the counterselection of autoreactive B cells(31, 39, 40) that lead to production of autoantibodies.

The second explanation, not mutually exclusive with the genetic model, is one of differential environmental and endogenous (immune dysregulation) exposures associated with RA(41). These exposures may render RA patients more susceptible to the downstream effects of carrying additional autoimmune risk alleles by having a lower threshold for producing autoantibodies than controls. However, if general immune dysregulation alone were

responsible for autoantibody formation, we would not expect the specific associations we observed between the SLE and celiac disease risk alleles with ANA and anti-tTG autoantibodies, respectively, in RA cases. In addition, we would expect to see an increased frequency of all autoantibodies; anti-TPO had the same prevalence in RA cases and non-RA controls.

Integrating the two explanations above, we interpret the similarities and differences between autoantibody prevalence in RA cases compared to non-RA controls in the following way. The common theme shared by RA cases and controls was that an increasing burden of autoimmune risk alleles was associated with carriage of more types of any autoantibody (Figure 3D, Supplementary Figure 1B). Autoimmune risk alleles therefore provide the substrate for autoantibody production. However, it is clear that additional factors contribute to increased autoantibody production. For example, among subjects carrying a similar number of SLE risk alleles, RA cases were more likely to be ANA positive than controls (Figure 4). We hypothesize that in RA cases, the effect of inherited and environmental factors that leads to a breakdown in self-tolerance and clinical symptoms of RA also predisposes to the development of additional autoantibodies such as ANA. In controls, there is less background immune dysregulation, and the same autoimmune risk alleles predispose both RA cases and controls to developing autoantibodies, but that environmental factors and endogenous exposures (immune dysregulation) that are also important contributors.

A prediction of our integrated model is that carriage of autoimmune risk alleles leading to immune dysregulation will have phenotypic consequences beyond RA case-control status. To investigate this hypothesis systematically, we utilized EMR clinical data to determine which phenotypes may be associated with specific autoantibodies or higher numbers of autoimmune

risk alleles. This approach, termed PheWAS, has been applied to test associations between SNPs and diagnosis codes(13, 42). However, this approach has not yet been used to test for clinical associations with autoantibodies or GRSs. From our exploratory PheWAS analysis, we observed, as expected, that the presence of anti-TPO was strongly associated with hypothyroidism in both RA cases and non-RA controls. We also observed associations between ANA and Sjogren's/sicca in RA cases. While this finding is consistent with clinical teaching regarding the association between RA and Sjogren's syndrome(43), this associated with has not been definitively shown until now. Among RA cases, ACPA status was not associated with RA because virtually all patients had a diagnosis code for RA.

Our study design represents a novel approach to translational research. The RA case and non-RA control study population was created with anonymized clinical data from the EMR(14) and linked to a biospecimen repository of discarded blood samples. Clinical information was obtained from the structured EMR (i.e. age, gender, diagnosis codes). This EMR platform afforded us the opportunity to understand the relationship between clinical, genetic factors and autoantibodies in RA, and to compare these findings with a control group. Moreover, we had the ability to integrate clinical EMR data and our research laboratory based autoantibody and genetic data, for thousands of patients to determine significant associations between autoimmune risk alleles and autoantibodies, and the potential clinical relevance of autoantibodies or a high number of autoimmune risk alleles.

There are important limitations of our study. We focused our study on a subset of autoimmune diseases, limiting the generalizeability of our conclusions. In the selection of our controls, we excluded those with any ICD9 code for a rheumatic disease(10). Thus the controls may also have a lower prevalence of autoantibodies than general population controls, however we found that the prevalence of ACPA, ANA and anti-TPO observed in our study were comparable to

controls in other studies(8, 44-46). Because we do not have temporal information relating to when a patient developed an autoantibody, we cannot state that an autoantibody predicted a particular outcome. Similarly, we cannot study whether an environmental exposure (i.e., smoking) contributed to autoantibody development.

We chose to use ACPA to subset RA rather than RF. The correlation between these two autoantibodies is high but discrepancies are known(47-49). Therefore, whether these results would directly apply to RF positive vs RF negative RA cases is unclear. There is concern that RF can lead to false positive results in multiplex antibody assays(50), however we found no evidence that RF results in false positive readings for the standard commercial ELISAs used in this study. Our findings suggest that RF interference was not a major issue; among RA cases, the prevalence of autoantibodies in ACPA positive compared to ACPA negative subjects was similar with the exception of ANA. Finally, results stemming from the PheWAS approach are exploratory and require further refinement in terms of validation of outcomes (i.e. ICD9 diagnostic codes), covariates and determining the metrics of what constitutes a true association (e.g., p-value threshold, effect size).

In conclusion, we employed a novel EMR-based approach to test genetic risk factors for association with clinical autoantibodies. Further, we provide insight into the immunologic heterogeneity underlying the clinical entity of RA. In the context of a linked clinical EMR-research laboratory database, we believe the PheWAS approach can be a powerful hypothesis generating tool for uncovering associations that are not readily apparent with our current knowledge of pathways and mechanisms of disease.

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 Table 1. Characteristics of RA cases and non-RA controls.

Characteristics	RA cases, n=1290 Controls, n=1236	
Mean age (SD)	60.7 (13.6)	64.6 (13.4)
Female, n (%)	1024 (79.4)	965 (78.7)
Methotrexate, ever use, n (%)	803 (62.3)	15 (1.2)
TNFi, ever use, n (%) ⁺	528 (40.9)	8 (0.65)
Bone erosions, n (%)	795 (61.6)	69 (5.6)
Measured autoantibodies, n (%)		
АСРА	901 (70.3)	1* (0.5)
ANA	445 (34.5)	193 (15.6)
ANAht ⁺	111 (8.6)	51 (4.1)
Ant-TPO	202 (15.6)	196 (15.9)
anti-tTG	14 (1.1)	8 (0.7)

* out of 202 controls

+ANAht= manufacturer high titer cutoff >60 units

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Table 2. Results of PheWAS analyses with significant associations between autoantibodies and EMR diagnoses among (A) RA cases(n=1265) and (B) non-RA controls (n=1225) of European ancestry.*

(A) RA cases

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Autoantibody	Outcome/diagnosis	Prevalence	of diagnosis	Odds ratio	(95% CI)	p-value
		Autoab pos(%)	Autoab neg (%)			
Anti-TPO	Acquired hypothyroidism	40.1	14.3	4.2	3.0, 5.9	1.2x10 ⁻¹⁶
ANAht 🔺	Sjogren's/sicca	13.5	3.5	4.2	2.2, 7.9	8.6x10 ⁻⁶
(B) Non-RA control Autoantibody	ls Outcome/diagnosis	Prevalence	of diagnosis	Odds ratio	(95% CI)	p-value
		Autoab pos(%)	Autoab neg (%)			•
Anti-TPO	Acquired hypothyroidism	48.5	24.1	2.8	2.0, 3.8	9.2x10 ⁻¹⁰
	Thyroiditis	12.2	2.7	4.7	2.7, 8.4	1.2x10 ⁻⁷
ANAht 🔺	Other chronic non-	11.7	1.7	8.0	3.0, 21.1	2.9x10 ⁻⁵
	alcoholic liver disease					

*Please see Supplementary Table 4 for top 3 results for all analyses

▲ Using ANA high titer manufacturer cutoff

Note: Bonferroni significant p-value in cases, p<9.76x10⁻⁵, and in non-RA controls, p<7.16x10⁻⁵

FIGURE LEGEND

Figure 1. (A) Distribution of autoantibodies (ACPA, ANA, anti-TPO) among RA cases of EU ancestry; **(B)** Distribution of autoantibodies (ACPA*, ANA, anti-TPO) among non-RA controls of EU ancestry. Anti-tTG (not shown), RA cases: 14 subjects (1.1%); non-RA controls: 8 subjects (0.6%). [*ACPA checked in n=202 controls with prevalence of 0.5%].

Figure 2. Comparison of autoantibody presence between **(A)** RA cases and non-RA controls, and **(B)** ACPA positive compared to ACPA negative RA cases (adjusted by age, gender and health care utilization; European ancestry).

Figure 3. Distribution of disease specific genetic risk scores and associated autoantibodies among RA cases (European ancestry, n=1265). **(A)** Distribution of RA GRS among ACPA positive and ACPA negative RA cases, **(B)** Distribution of the SLE GRS among ANA positive and ANA negative RA cases, **(C)** Distribution of the celiac GRS among anti-tTG positive and anti-tTG negative RA cases, **(D)** Distribution of AI GRS with increasing count of autoantibodies in RA cases (ANA, anti-TPO or anti-tTG).

Figure 4. ANA positivity in RA cases compared to non-RA controls among individuals with similar numbers of SLE risk alleles (SLE GRS categorized by tertiles from lowest, tertile 1 to highest, tertile 3; ORs adjusted by age, gender, and health care utilization in subjects of European ancestry).

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Figure 1. (A) Distribution of autoantibodies (ACPA, ANA, anti-TPO) among RA cases of EU ancestry; **(B)** Distribution of autoantibodies (ACPA*, ANA, anti-TPO) among non-RA controls of EU ancestry. Anti-tTG (not shown), RA cases: 14 subjects (1.1%); non-RA controls: 8 subjects (0.6%). [*ACPA checked in n=202 controls with prevalence of 0.5%].

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(B) Autoantibody presence in ACPA+ vs ACPA- RA cases



Figure 2. Comparison of autoantibody presence between **(A)** RA cases and non-RA controls, and **(B)** ACPA positive compared to ACPA negative RA cases (adjusted by age, gender and health care utilization; European ancestry).

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Figure 3. Distribution of disease specific genetic risk scores and associated autoantibodies among RA cases (European ancestry, n=1265). **(A)** Distribution of RA GRS among ACPA positive and ACPA negative RA cases, **(B)** Distribution of the SLE GRS among ANA positive and ANA negative RA cases, **(C)** Distribution of the celiac GRS among anti-tTG positive and anti-tTG negative RA cases, **(D)** Distribution of AI GRS with increasing count of autoantibodies in RA cases (ANA, anti-TPO or anti-tTG).

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