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Systematic Evaluation Of Genes And Genetic Variants Associated With Type 1 Diabetes Susceptibility

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Abstract

Genome-wide association studies (GWAS) have found over 60 loci that confer genetic susceptibility to Type 1 diabetes (T1D). Many of these are defined only by anonymous SNPs: the underlying causative genes, and the molecular bases by which they mediate susceptibility, are not known. Identification of how these variants affect the complex mechanisms contributing to the loss of tolerance is a challenge. We performed systematic analyses to characterize these variants. First, all known genes in strong linkage disequilibrium (LD) $(r^2 > 0.8)$ with the reported SNPs for each locus were tested for commonly occurring non-synonymous variations. We found only a total of 22 candidate genes at 16 T1D loci with common non-synonymous alleles. Next, we performed functional studies to examine the effect of non-HLA T1D risk alleles on regulating expression levels of genes in four different cell types: EBV- transformed B cell lines (resting and 6h PMA stimulated); purified CD4+ and CD8+ T cells. We mapped *cis*-acting expression quantitative trait loci (eQTL) and found 24 non-HLA loci that affected the expression of 31 transcripts significantly in at least one cell type. Additionally, we observed 25 loci that affected 38 transcripts in *trans*. In summary, our systems genetics analyses defined the effect of T1D risk alleles on levels of gene expression and provide novel insights into the complex genetics of T1D, suggesting most of the T1D risk alleles mediate their effect by influencing expression of multiple nearby genes.

Keywords

Type 1 Diabetes; eQTL; Gene expression; Genome-Wide Association Studies

Online Resource: <http://www.sysgen.org/T1DGCSysGen/>

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INTRODUCTION

Type 1 diabetes (T1D) affects approximately 30 million people worldwide (1). It is a complex autoimmune disease causing the destruction of pancreatic β cells. The largest genetic studies of T1D have been carried out by the Type 1 Diabetes Genetics Consortium (T1DGC) (2–4). These and other reports have now defined genetic variants associated with T1D in over 60 different chromosomal regions (see ref 5 for review).

There is a need to identify the causative variants that are in linkage disequilibrium (LD) with the single nucleotide polymorphisms (SNPs) found by such association studies, and to define the molecular bases by which they contribute to disease susceptibility. The challenge of post-genome wide association studies (GWAS) functional studies (6–8) is in finding ways to translate genetic associations into clinically useful information. The strong genetic association of the disease with HLA class II genes of the major histocompatibility complex (MHC) is well established (9) but the identity of the genes associated with many of the non-HLA loci remains largely unknown, especially with respect to those associated SNPs located in non-coding regions of the genome (2, 5). Therefore, this study focuses on characterizing the non-HLA T1D risk loci.

In principle, most genetic variants could plausibly affect biological processes by changing amino acid residues in encoded proteins or by changing their levels of expression in particular tissues. Various DNA sequence repositories allow identification of commonly occurring non-synonymous (missense) variations in genes, and amino-acid substitution polymorphisms could be characterized for their potential to affect biological processes (10). Expression quantitative trait locus (eQTL) analyses can identify genes whose variation in expression is associated with specific SNP markers. For example, sequence variation in promoters or enhancer elements could result in differential *cis* regulation. Genetic variants can also regulate expression of genes at greater distances from, or on different chromosomes than, the regulatory element, i.e., *trans* regulation (11). The mechanisms involved in *trans* regulation could include indirect genetic effects, e.g. by means of variation in encoded proteins such as transcription factors, or by other effects, such as steric (11). Some loci could exert both *cis* and *trans* effects.

In the present study, we performed systems genetics (12) analyses of the 55 loci (2, 13–25) (Table I) showing highest evidence of association with T1D, using data generated by the T1DGC (2) and Immunochip projects (13). Additionally, four new SNPs (*rs6691977*, *rs4849135*, *rs2611215* and *rs11954020*) that showed strong associations ($P < 5 \times 10^{-8}$) with T1D in (13) were included in our study. SNPs at these loci were assessed for disease gene candidacy. Expression data of 47,323 high-quality transcripts (Illumina, HT-12 V4) were correlated with SNPs reported in T1D loci adjusting for confounding factors such as population structure.

MATERIALS AND METHODS

Study Samples

The Type 1 Diabetes Genetics Consortium (T1DGC) study has been described elsewhere, including phenotypic and extensive genetic characterization of over 4,000 affected sib-pair families (3). Upon joining the T1DGC, family members provided blood samples. Peripheral blood mononuclear cells (PBMC) were isolated and aliquots were used to provide DNA samples; to derive EBV-transformed B lymphoblastoid cell lines (LCL) (26–27); and frozen for later use. EBV-B cells from 202 European subjects from the T1DGC family collection were studied here. These samples consisted of 46 unaffected subjects and the rest were T1D cases. EBV-B cells were either unstimulated, or treated with phorbol-12-myristate-13 acetate (PMA) (28) for 6h (26–27). PMA stimulated samples consisted of 49 unaffected subjects. Cell lines were stimulated on a second occasion to provide a duplicate sample. SNPs were genotyped using the Immunochip (13) platform.

Frozen PBMC samples from 113 T1DGC family members were thawed, cultured overnight, stained and separated into CD4+ and CD8+ T cell populations by flow-sorting. Sufficient RNA was obtained from 102 CD4+ T cell samples and 84 CD8+ T cell samples to perform microarrays. Sex, HLA-DR and autoantibody statuses of the affected subjects are summarized in Suppl. Table I.(i).

Microarray Analyses

After cell culture or flow-sorting, RNA was extracted using TRIZOL® Reagent (Invitrogen) following the manufacturer's instructions. The RNA quantity was measured by NanoDrop 1000 Spectrophotometer (Thermo Scientific) and RNA quality was checked on Agilent 2100 Bioanalyser (Agilent). Samples with RIN (RNA Integrity Number) number of equal or greater than 8 were biotin labeled using Illumina TotalPrep RNA Amplification kit (Ambion) as per manufacturer's instructions. The biotin- labelled samples were hybridized onto Illumina HumanHT-12 v4.0 expression beadchips and beadchips were scanned by Beadarray Reader (Illumina) following manufacturer's instructions. Raw data was finally exported by GenomeStudio software (Illumina) for analysis.

Microarray and eQTL analysis

Genome-wide gene expression values from GenomeStudio (Illumina) for each of 47,323 probes were subjected to background correction using control probe profile, variance stabilizing transformation (VST) and RSN (robust spline normalization) normalization using lumi package (29) in R. We then removed from the analysis 95 transcripts that are *ERCC*spike in controls (having gene symbols starting with '*ERCC*'). Four separate gene expression datasets were created. Upon examining initial PCA plots, batch effects were evident. For correction within each cell type, normalized expression data for each gene was centred by batch and centred again after merging batches. The batch correction was validated by PCA analysis (Suppl. Figure 1. (A and B)) and pair plots of PCs 1–4 did not reveal any further batch effects. BLASTn software was to identify probesets with unique sequences.

The data generated in this study have been deposited in the NCBI Gene Expression Omnibus (GEO, Series accession number GSE77350) and is publically accessible via this URL. [\(http://www.ncbi.nlm.nih.gov/geo/query/acc.cgi?acc=GSE77350](http://www.ncbi.nlm.nih.gov/geo/query/acc.cgi?acc=GSE77350))

To assess association between SNP genotype and gene expression, the MatrixEQTL R (30) package was used. To adjust for unknown confounders in the expression, two correction methods were used and results were compared.

RUV-2 correction

First association of all T1D SNPs with normalized uncorrected data was performed and pvalue association of every SNP-gene pair was obtained. For each SNP, the top 5000 associated genes ranked by p-value were excluded and the rest were treated as empirical controls for RUV-2 correction (31–32) using *naiveRandRUV* method, parameter *k* set to 20. After correction, the same SNP was tested against the corrected set and p-value association of SNP-gene pair was recorded. This procedure was repeated for all SNPs and finally Benjamini FDR correction was applied to the set of recorded nominal *P*-values.

PCA correction

PCs were derived from individual whole expression sets and tested against whole genome Immunochip SNPs (200K). The PCs that showed no or weak genome association (i.e. min SNP-PC association FDR *P* > 0.001) were chosen as un-associated PCs (33). These PCs were incrementally added in their order of precedence as covariates to assess SNP-gene associations with an aim to maximize the number of significant *cis* gene detections (at FDR *P* < 0.001) for the 77 T1D SNPs tested. Based on analysis shown in Suppl. Figure 1 (E and F), the four gene expression datasets were corrected as follows: 7 PCs: 1–6 and 8 were removed from EBV-B basal cell line samples, 3 PCs: 1, 4 and 9 were removed for PMA stimulated EBV-B cell line samples, 4 PCs: 1–4 were removed for CD4+ samples and 2 PCs: 1 and 2 were removed for CD8+ samples.

We compared numbers of *cis*- and *trans*- regulated genes detected in each cell type using two methods (Suppl. Table I. (ii)). The RUV-2 method of correction yielded more significant results than PCA methods.

Statistical Analysis

Differential gene expression analysis was performed using the Limma package written for R (34). TDT (sibship) tests were performed using the software package UNPHASED (35–36).

Enrichment Analysis

Candidate gene names were converted to Entrez gene ids and were analysed using the DAVID (37–38) function annotation tool (http://david.abcc.nciferf.gov/). Further pathway and network analysis were performed using GATHER (gather.genome.duke.edu) (39) and GENEMANIA [\(www.genemania.org\)](http://www.genemania.org) (40) respectively.

RESULTS

Systematic evaluation of non-synonymous SNPs in genes in T1D-associated regions

First, we searched for commonly occurring non-synonymous (ns) SNPs in linkage disequilibrium (LD) $(r^2 > 0.8)$ with the T1D SNPs (2, 13–25) in the 1000 genomes and HAPMAP (41) CEU datasets. All amino acid substitutions were subject to prediction of the effect of these changes, evaluated as 'benign', 'probably damaging' or 'possibly damaging' by PolyPhen-2 (10). This search returned 25 nsSNPs in strong LD with only 16 of the 60 non-HLA T1D loci. These SNPs occurred in a total of 22 unique genes. The seven potentially damaging effects were found in two genes, *SULT1A2* and *GSDMB*. Prediction status does not affect candidacy per se, so all genes listed in Table II should be evaluated in further studies. In addition, none of the four SNPs recently discovered in (13) were in strong LD $(r^2 > 0.8)$ with any nsSNPs. Among the LD SNPs, there were three splice-region variants and one stop-gain variant (summarized in Suppl. Table I.(iii)).

Next, we searched whether any nsSNPs showed better association with T1D than the reported SNP itself. For this, we performed a transmission disequilibrium test (TDT – sibship test) using UNPHASED (35–36) on a dataset of 2,676 nuclear families with unaffected parents and two or more affected sibs. Results are presented in Table II. Association *P*-values for SNPs not included in the Immunochip genotyping were derived from (2). At six T1D loci, the nsSNPs were the reported best SNPs. From those nsSNPs that were genotyped by Immunochip, *rs7498665* associated with *SH2B1* showed slightly better association than the reported $rs4788084$ ($P = 0.1$, where $P = P_{ns \text{ }SNP} / P_{reported \text{ }SNP}$). Two other ns SNPs (*rs2305480* and *rs229527*) also showed very small (Δ*P* > 0.1) improvement in association compared to the reported T1D SNP. Most of the T1D loci did not have associated nsSNPs in nearby genes.

Gene Expression Analyses

EBV-transformed B cell lines (referred to hereafter as EVB-B) were produced from blood samples obtained from T1DGC family members (3). RNA was extracted from 202 available EBV-B cell lines that were cultured under basal conditions and stimulated with PMA. We also purified CD4+ and CD8+ T cells from peripheral blood samples provided by 113 subjects. None of these subjects overlapped with the donors of the 202 EBV samples. After quality control, sufficient high quality RNA to perform microarrays was obtained from 102 CD4+ T cell samples and 84 CD8+ T cell samples. The EVB-B cell samples were derived from both T1D cases and unaffected subjects. The unaffected controls included were firstdegree relatives of the subset of case samples and islet autoantibody status was not determined for these unaffected subjects. The details regarding the autoantibody, sex and HLA-DR status of the affected subjects are summarized in Suppl. Table I.(i). As expected, there were no significant differences in the gene expression between cases and unaffected subjects nor between cases and unaffected first-degree relatives (Suppl. Figure 1.C and 1.D) so all samples were used to search for eQTLs.

These RNA samples were hybridized to Illumina microarrays (HT-12v4). Data processing was carried out as described in Methods. Batch effects were corrected for each cell type by

centering the normalized gene expression data by batch and centering again after merging batches. The batch correction was validated by PCA analysis (Suppl. Figure 1 (A & B)). To eliminate probesets with potential cross-hybridization problems, a BLAST search of each probe sequence was carried out on a custom database of all 47,323 Illumina probeset sequences and 38,500 probes that had a single hit were retained. In doing so, probes associated with two known T1D candidates *RPS26* (due to sequence similarity with probes associated with *RPS26* pseudogenes) and *DEXI* (due to sequence similarity with a probe associated with *LOC653752*) were removed. There were 95 *ERCC* spike-in controls in the probeset, which were also excluded from analysis. We also performed a search for SNPs within probeset coordinates and excluded any probes that contained SNPs from further analysis. We performed differential expression analysis of unstimulated EBV-B cells and after 6 hour PMA stimulation. The negative log_{10} (adjusted *P*-value) of each probe showing differential expression was plotted against the log₂ fold change in a 'volcano plot' (Figure 1). Adjusted *P* < 0.0001 was selected as a cut-off for differential expression. A total of 1,465 genes were differentially expressed at this threshold with at least a modest fold change (absolute $log_2FC > 0.3$). Genes with the highest fold changes in expression included *CCL3*, *CCL4, EGR1, EGR2, DUSP21, PIP4K2C, ILDR1* and *IL9R*.

Parameters for Systems Genetics Analyses

Genotypes of T1DGC subjects were previously determined (2–4, 13) at 77 SNPs in 55 of 60 T1D risk loci (Table I). Based on the risk allele's code at each T1D SNP, an additive recode [0,1,2] was applied so that the risk allele's effect on gene expression could be determined. Separate analyses were performed for each of the four expression sets (EBV-B basal, EBV-B 6h PMA stimulated, CD4+ and CD8+). For these analyses, we conservatively defined a *cis* transcript as being from a gene whose transcription start or end site was located within 1 Mbp from the T1D SNP. A *trans* regulated transcript was defined as a gene located elsewhere in the genome. For each set, 3,672 *cis* interactions pairs were tested; ~2.9M *trans* interactions pairs were tested; and false discovery rate (FDR) *P*-value corrections were applied separately for *cis* and *trans* eQTLs. The MatrixEQTL R package (30) was used to perform these eQTL tests. Due to unknown confounding factors that could limit the power of detecting significantly differentially expressed genes, we performed two methods of correction independently: (a) removing unwanted variation (RUV-2) (31–32); (b) and adding genome wide un-associated expression derived principle components (PCs) as covariates (described in Methods).

All transcripts with FDR *P* < 0.05 for each T1D SNP were followed up with enrichment analysis using the DAVID bioinformatics resource (37–38). Additional pathway and network analysis was performed using GATHER (39) and GENEMANIA (40) respectively. The results from these analyses are summarized in Tables III – VII and are described below. Boxplots of eQTL associations can be accessed online through our web resource (42) where we compare effects explained by raw normalized gene expression against RUV-2 and PCA corrected gene expression sets. A screenshot of the user interface is shown in Figure 2.

Effect of T1D-associated non-HLA SNPs on neighboring gene expression in EBV-B Cell lines

We examined *cis* genes in EBV-B basal cell line samples at various FDR *P* value thresholds. At *P* < 0.001, 15 T1D SNPs were associated with differences in expression of 20 genes (Table III). Using lower thresholds of adjusted *P* [< 0.05], an additional 13 T1D SNPs affected the expression of a further 20 genes (Suppl. Table II.(i)). Hence, 28 T1D SNPs were found to be associated with changes in a total of 40 significant *cis* genes. Of these, three SNPs (*rs10877012*, *rs4788084* and *rs2290400*) showed strong *cis* effects with multiple nearby genes that were either up- or down- regulated by the corresponding risk allele. In testing the four newly discovered T1D SNPs, (13) we observed that the risk allele associated with $rs2611215$ reduced expression of *TMEM192* (FDR $P = 0.008$) (Suppl. Table II.(i)).

Next, we tested 6h PMA stimulated EBV-B cell line samples. Results confirmed the *cis* effects associated with 22 of 40 candidate genes identified in unstimulated EBV-B cells (at minimum FDR *P* < 0.05) and the effect directions were consistent. *IFNGR1*, *SUOX*, *SPNS1* and *UBASH3A* were among genes that showed regulatory effects in basal cells but not after PMA stimulation. In addition, 17 T1D SNP genotypes significantly regulated the expression of 16 new candidate genes (FDR *P* < 0.05). Of these, genes *INO80B* and *LYRM2* were detected highly significant at FDR *P* < 0.001 (Table III). The expression of candidate genes *IKZF1* and *TSFM* showed decreased association with their corresponding T1D SNPs after stimulation, compared to basal condition (refer to 42). The rest of these results are presented in Suppl. Table II.(i).

In summary, 31 T1D SNPs affected the expression of a total of 38 candidate *cis* genes, 22 of which had shown evidence of *cis* effects in unstimulated EBV-B cells while the remaining nine showed association after PMA stimulation, thus suggesting genes that may play a role after immune activation.

Effect of T1D-associated non-HLA SNPs on neighboring gene expression in CD4+ and CD8+ T cells

Tests of CD4+ T cell samples revealed 16 T1D SNP genotypes regulated the expression of 20 genes significantly. Of these genes, eleven (*SMARCE1, LOC728734, SUOX, FAM119B, C16ORF75, GSDMB, IKZF1, ADCY3, ORMDL3, SKAP2* and *IKZF3*) were found to be *cis* regulated in both EBV-B and CD4+ T cells by the same T1D SNPs (Table III). In particular, the risk allele of *rs2290400* (T) affected nearby genes *ORMDL3*, *GSDMB* and *IKZF3* similar to that observed in EBV-B cells. The effect directions between the cell types for the 11 shared genes were consistent, except for gene *C16ORF75* where the risk allele increased expression in CD4+ cells but decreased it in EBV-B cells (42). We also noted expression of candidate gene *SUOX* showed a clear increase in the significance of association (i.e. lower p-value) with T1D risk allele *rs705704* (T) in the CD4+ cells compared to EBV-B cells. In addition, there were nine newly identified candidate genes associated with nine T1D SNPs. Five of these SNPs had showed *cis* effects in the EBV-B cells, but had affected a different set of genes. Among these 9 new candidate genes, *CLECL1* was the most significantly associated (Table III). The *cis* genes detected at lower FDR thresholds of 0.01 and 0.05 are

presented in the Suppl. Table II.(i). These results suggest that the effects of the T1D risk SNPs on gene expression vary between cell types.

Finally, we performed analyses of the CD8+ T cell samples and identified 17 T1D SNP genotypes regulated the expression of 19 genes across all samples tested. Excepting *ADCY3*, ten candidate genes were found *cis* regulated in EBV-B, CD4+ T cells and CD8+ cells. Thirteen of the 19 candidate genes were *cis* regulated in both CD4+ and CD8+ T cells and the effect directions were consistent. The remaining six that were neither differentially regulated in EBV-B cells nor in CD4+ cells were associated with six T1D SNPs in CD8+ cells. Of these, T1D SNP *rs2292239* regulated the expression of candidate gene *ERBB3* most significantly (FDR *P* < 0.001) (Table III). The rest of the results are presented in Suppl. Table II.(i).

In summary, 24 T1D SNP genotypes regulated the expression of 31 candidate genes highly significantly at FDR *P* < 0.001 (Table III). Using lower FDR adjusted *P*-value thresholds (*P* < 0.05), 43 T1D SNP genotypes regulated the expression of 71 candidate genes. Using even lesser stringent suggestive threshold of nominal un-adjusted *P* < 0.001 for evidence of *cis* effect, we could define up to 85 candidate genes that were affected by 50 T1D SNPs in the four cell types tested.

T1D-associated SNPs associated with changes in expression of distant genes

Next, we investigated whether T1D loci showed *trans*-regulatory effects. After performing \sim 2.9M tests for each cell type and appropriate statistical correction, we identified 38 genes that were highly significantly associated with 25 T1D SNPs at FDR *P* < 0.001 (Table IV). Five of these SNPs (*rs1534422, rs1990760, rs11571291, rs9585056* and *rs425105*) did not show any *cis* effect on nearby genes in the cell types tested. *Trans*-regulated genes shared between B and T cells were detected at only one T1D locus (defined by T1D SNP *rs705704*) and the effect direction was consistent. Except for *ZMYM5*, *GRAMD1B* and *LOC389386*, all significant *trans* genes were detected in the EBV-B cells. Upon characterizing the function of 38 *trans* genes in DAVID (37–38), we identified two clusters: *CD276, ST6GAL1, CCL5* and *IRF8* were associated with immune response and a further two genes (*ID2* and *IRF8)* were associated with immune system and hemopoietic (lymphoid) organ development. Eight T1D SNPs (Table IV: highlighted in bold) showed highly significant *cis* as well as *trans* regulatory interactions in one or more cell types tested, suggesting co-regulation between *cis* and *trans* genes. We describe tests for meaningful relationships between these genes in the next section.

In summary, in addition to the loci that affected genes in *cis*, we could identify five loci that exclusively affected genes in *trans*. Of the T1D loci that were not associated with expression changes in any of the four cell types, three loci contained non-synonymous SNPs defined in Table II. The *trans* regulatory effects detected at lower threshold levels are presented in Suppl. Table II (ii) and Suppl. Table III.

Enrichment analysis of genes associated with T1D susceptibility

We investigated the function of the genes whose expression was changed by individual risk SNPs. The DAVID enrichment analysis software (37–38) tests whether sets of genes are enriched for terminology referenced by UniProt Protein Information Resource (PIR) keywords, Gene Ontology (GO) and KEGG Pathways. First, we performed analysis to explore for enrichment between the highly significant (FDR *P* < 0.001) *cis* and *trans* gene candidates for the eight T1D SNPs highlighted in Table IV. For three of these SNPs, the candidate genes shared a common keyword (Table V). Second, using the list of 86 candidate genes derived from Tables II – IV, we performed pathway and enrichment analysis using GATHER (39) and we report results obtained with high confidence (unadjusted $P < 0.001$) in Table VI. In these results, we found that the cytokine-cytokine receptor interaction pathway received the highest significance. Third, we performed network analysis using GENEMANIA (40) for the same list of 86 candidate genes. The significant functional findings are presented in Table VI. The full GENEMANIA report can be accessed online [\(www.sysgen.org/T1DGCSysGen/genemania.pdf\)](http://www.sysgen.org/T1DGCSysGen/genemania.pdf). Finally, we analyzed the list of *cis* and *trans* genes detected at FDR *P* < 0.05 for every T1D SNP separately. We identified 21 enrichment terms (excluding Gene Ontology cellular component terms) that were significantly enriched at Benjamini *P* < 0.05 for 10 T1D SNPs. These results are summarized in Table VII and below.

The term 'Lectin' was highly enriched for the T1D locus defined by *rs10466829* since it affected expression of five c-type lectin genes (*CLEC1A, CLEC2B, CLEC2D, CLECL1* and *CD69*) in the cell types tested. The T1D locus defined by *rs17696736* was highly enriched for 'response to virus' and 'anti-viral defense' due to changes in expression of 7 *trans* genes (*EIF2AK2, IFI16, IFNGR1, MX1, MX2, PLSCR1* and *STAT1*). Furthermore, genes *MX1* and *MX2* are also known inflammatory and immune response genes. In addition, the T1D SNP *rs416603* showed significant enrichment for '*IL10*-anti inflammatory signaling pathway' and 'intestinal immune network IgA production pathway' through its regulation of three genes (*IL10, IL10RA* and *STAT5A*). We also noted that two risk SNPs (*rs2476601* and *rs679574*) showed association in *trans* with genes in the MHC (*HLA-F, G, H*, and *DRB4*), which gave positive enrichment for terms such 'antigen processing and presentation'. These results provide insights into the functions of genes whose expression is affected by the T1D loci.

Validation of trans-regulatory gene interactions

To confirm our results, we searched using the blood eQTL browser (43) for the *trans* regulatory associations we identified at significance threshold FDR *P* < 0.05. Since not all T1D SNPs may be present in this browser, we allowed a 100Kb window for the search of the expression SNP. Two *trans* genes were validated: *UBE2L6* (EBV-B +/− PMA) associated with *rs3184504* and *STAT1* (EBV-B basal) with *rs17696736*. Secondly, we searched in the *trans* regulatory interactions reported by Fairfax *et al*. (44) and validated a further three gene interactions reported in their study: *LOC728823, IP6K2* and *LOC389386* all associated with the T1D SNP *rs705704*. Although many *cis* gene effects were clearly defined from our datasets, validating *trans* genes poses a challenge warranting further investigation.

DISCUSSION

Our results provide a potential molecular basis for disease association at 46 of the 59 identified T1D loci (Table I). Sixteen of these loci contained non-synonymous SNPs in strong LD with the T1D SNP. Thirty-six of the loci showed *cis* effects on 75 nearby genes. The remainder showed statistically significant *trans* regulatory interactions that were substantiated by significant enrichment results (Tables V–VII). These candidate genes can be the focus for further studies. For example, a systems genetics study (45) into candidate gene *CTSH* whose expression was affected by T1D SNP *rs3825932*, supported its product as a novel therapeutic target.

Onengut-Gumuscu *et al*. (13) recently confirmed several previously reported T1D associated SNPs (2, 5) in addition to the identification of four additional new T1D risk SNPs of which one SNP ($rs2611215$) had high significance ($P = 1.817 \times 10^{-11}$) while *P* values of the rest only just exceeded the significance threshold ($P < 5 \times 10^{-8}$). This study found that the associated SNPs localized to enhancer sequences active in thymus, T and B cells, and CD34+ stem cells. Of the four new T1D associated SNPs (13) we were able to establish likely candidacy for *rs2611215* as *TMEM192*.

An important conclusion from our study is that the cell type was important in characterizing T1D SNP function, i.e. eQTLs are cell type-specific. For example, the candidate gene *ERBB3* was highly significantly *cis* regulated in CD8+ T cells but its variation effect was largely undetectable in other cell types. The risk allele associated with *rs4788084* reduced expression of candidate gene *TUFM* exclusively in the CD4+ cells. Similarly, *CLECL1* did not show any effect in EBV-B cell lines but showed highly significant effects in both T cell types tested. Among the weakly detected effects, there was evidence that suggested the risk allele associated with *rs231727* reduced expression in *cis* of a well known candidate (*CTLA4)* exclusively in the CD8+ cells (unadjusted *P*=0.0003, FDR *P* = 0.04) (Suppl. Table II. (i)). Our CD4+/ CD8+ cell type data also assisted in mapping candidate genes at otherwise anonymous T1D SNPs; the most significant of these candidates included *SLC11A1* (*rs3731865*), *C6Orf173* (*rs9388489*) and *C10orf59* (*rs10509540*).

Sixteen transcripts (twelve in *cis*, four in *trans*) were significantly associated with T1D SNPs in both EBV-B and the T cell types tested. Of these, a novel uncharacterized *cis* transcript *LOC728734* (nuclear pore complex interacting protein family, member B8) was identified to be associated with T1D SNP *rs4788084* (chr 16p11.2) where the risk allele decreased expression in all four cell types. The effect direction of *cis* and *trans* regulation by T1D SNPs on genes detected across multiple cell types were found consistent for all SNPs except *C16ORF75*. We also noted that probes associated with candidate genes *DEXI* and *RPS26* also showed strong *cis* regulatory effect in association with T1D risk SNPs *rs12708716* and *rs705704*, respectively, in one or more cell types. However, due to quality control procedures relevant to cross hybridisation problems described in the previous section, these probes were excluded from further analysis. Non-synonymous SNPs may also affect gene expression in *trans*. We found two examples of these: *rs1990760* (chr 2q24.2) in *IFIH1* also affected the expression of *LOC643997* in *trans*; similarly *rs2304256* (chr 19p13.2) in *TYK2* also affected the expression of *ZNF280D* in *trans*.

Pathway analysis identified the "Cytokine-cytokine receptor interaction" pathway with highest confidence. The "Sulfur metabolism" pathway also scored high significance because two genes *SUOX* (*cis*) and *SULT1A2* (non-synonymous) involved in this pathway were identified as candidates in this study. It is also well known that sulfur plays an important role in insulin production (see 46, for review). Furthermore, DAVID enrichment analysis of locus specific *cis* and *trans* transcript perturbations revealed significant enrichment of 48 category terms in 15 of the T1D regions at FDR *P* < 0.05. Among the best enriched terms were 'response to virus', 'acetylation', 'lectin' and 'IL10-anti-inflamatory pathway'. From the enrichment analysis for genes associated with each T1D SNP, upon examination T1D risk SNP *rs17696736* (chr 12q24.12) was notably associated with 'response to virus' and 'antiviral defence' due to *trans* genes that are involved in pro-inflamatory response (such as *MX1*, *MX2*) in the salmonella infection pathway (KEGG Pathway: 05132). In contrast, chemokine gene *CCL5* was highly significantly associated with diabetes loci associated with T1D SNP *rs425105* (chr 19q13.32). These results support evidence found in a recent work (47) suggesting salmonella and chemokine vaccines can prove clinically useful in diabetes management and prevention.

In conclusion, our results confirm systems genetics (12) as a powerful tool for investigating the genetic architecture of complex diseases such as T1D. Many genes were identified whose expression levels were influenced by SNPs associated with T1D susceptibility. These nsSNPs, *cis* and *trans* regulated genes we identified are important candidates for further investigation. So that other researchers can extend the work reported here, we have implemented a web interface (42) allowing users to browse boxplots for the eQTL interactions reported below.

Supplementary Material

Refer to Web version on PubMed Central for supplementary material.

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ABBREVIATIONS

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Figure. 1.

Comparison of gene expression in EBV-B cells between basal and 6 hour PMA stimulated samples. Differences (log_2 fold change) in gene expression are shown on the X axis; Y axis shows $-log_{10}$ (adjusted P values).

Figure. 2.

Screen capture of the Web interface for browsing box-plots and gene networks. URL: [http://](http://www.sysgen.org/T1DGCSysGen/) www.sysgen.org/T1DGCSysGen/

Table I

List of reported T1D SNPs located in 59 non HLA T1D loci. List of reported T1D SNPs located in 59 non HLA T1D loci.

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Ref.: Publication reference. ID: unique T1D loci identifiers. BP: NCBI36 chromosome positions. Ref.: Publication reference. ID: unique T1D loci identifiers. BP: NCBI36 chromosome positions.

 α α β δ 363609 is in LD with reported SNP r s6546909 (r^2 = 0.84) whose p-value is stated. *rs363609* is in LD with reported SNP *rs6546909* (*r* $2 = 0.84$) whose p-value is stated.

*** P-values are derived from Barrett et al. (from Table 2 in (2)). 'NR' denotes no gene of interest was reported at the locus. P-values are derived from Barrett *et al*. (from Table 2 in (2)). 'NR' denotes no gene of interest was reported at the locus.

II, III, IV, S corresponds to Tables where SNPs are featured; II – Non-synonymous LD SNPs, III – cis- interacting genes, or IV – trans- interacting genes, S – Supplementary cis- interacting genes (see
Suppl. Table II(i)). II, III, IV, S corresponds to Tables where SNPs are featured; II – Non-synonymous LD SNPs, III – *cis*- interacting genes, or IV – *trans*- interacting genes, S – Supplementary *cis*- interacting genes (see Suppl. Table II(i)).

Table II

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ID: T1D loci identifiers as in Table I. Genes SULTIA2 and GSDMB, the SNPs cause deleterious changes. The best NS SNP is underlined where possible. ID: T1D loci identifiers as in Table I. Genes *SULT1A2* and *GSDMB*, the SNPs cause deleterious changes. The best NS SNP is underlined where possible.

Association p-values are derived from Barrett et al. (2). Association p-values are derived from Barrett *et al.* (2).

n/a - SNP genotypes not available. n/a – SNP genotypes not available.

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Table III

Cis-genes associated with 24 T1D SNPs with minimum False Discovery Rate, FDR $P < 0.001$. *Cis*-genes associated with 24 T1D SNPs with minimum False Discovery Rate, FDR P < 0.001.

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 $\frac{1}{2}$

ID: T1D loci identifiers as in Table I. ID: T1D loci identifiers as in Table I.

Following notations are used: FDR: **** P < 0.0001, *** P < 0.001, ** P < 0.01, * P < 0.05, ns – not significant, \downarrow risk (effect) allele reduces expression, \uparrow risk (effect) allele increases expression (determined Following notations are used: FDR: **** P < 0.0001, *** P < 0.001, ** P < 0.01, * P < 0.05, ns – not significant, ↓ risk (effect) allele reduces expression, ↑ risk (effect) allele increases expression (determined using beta coefficient) Author Manuscript

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ID: T1D loci identifiers as in Table I. ID: T1D loci identifiers as in Table I. Following notations are used: FDR: **** P < 0.0001, *** P < 0.001, ** P < 0.01, * P < 0.05, ns – not significant, ↓ risk allele reduces expression, ↑ risk allele increases expression. TID SNPs highlighted
also showed stro Following notations are used: FDR: **** P < 0.0001, *** P < 0.001, ** P < 0.01, * P < 0.05, ns – not significant, ↓ risk allele reduces expression, ↑ risk allele increases expression. T1D SNPs highlighted also showed strong *cis*-regulatory effects at FDR P < 0.001.

Table V

Enrichment terms shared between *cis* genes and *trans* genes in association with three T1D SNPs.

ID: T1D loci identifiers as in Table I.

Table VI

Network and Pathway analysis of the list of candidate genes identified in Tables II–IV using GATHER and GENEMANIA (Unadjusted *P* < 0.001).

Table VII

Significant enrichment terms found using DAVID bioinformatics resource.

ID: T1D loci identifiers as in Table I.

PIR: Protein Information Resource, GO: Gene Ontology, BP: Biological process, MF: Molecular Function