Network analysis of single nucleotide polymorphisms in asthma

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Background: Asthma is a chronic inflammatory disease of the airways with a complex genetic background. In this study, we carried out a meta-analysis of single nucleotide polymorphisms (SNPs) thought to be associated with asthma.

Methods: The literature (PubMed) was searched for SNPs within genes relevant in asthma. The SNP-modified genes were converted to corresponding proteins, and their protein–protein interactions were searched from six different databases. This interaction network was analyzed using annotated vocabularies (ontologies), such as the Gene Ontology and Nature pathway interaction databases.

Results: In total, 127 genes with SNPs related to asthma were found in the literature. The corresponding proteins were then entered into a large protein–protein interaction network with the help of various databases. Ninety-six SNP-related proteins had more than one interacting protein each, and a network containing 309 proteins and 644 connections was generated. This network was significantly enriched with a gene ontology entitled “protein binding” and several of its daughter categories, including receptor binding and cytokine binding, when compared with the background human proteome. In the detailed analysis, the chemokine network, including eight proteins and 13 toll-like receptors, were shown to interact with each other. Of great interest are the nonsynonymous SNPs which code for an alternative amino acid sequence of proteins and, of the toll-like receptor network, TLR1, TLR4, TLR5, TLR6, TLR10, IL4R, and IL13 are among these.

Conclusions: Protein binding, toll-like receptors, and chemokines dominated in the asthma-related protein interaction network. Systems level analysis of allergy-related mutations can provide new insights into the pathogenetic mechanisms of disease.

Keywords: asthma, network, pathway pathogenesis, single nucleotide polymorphisms

Introduction

Asthma is a chronic inflammatory disease of the airways characterized by infiltration and activation of inflammatory cells and by structural changes, including subepithelial fibrosis, smooth muscle cell hypertrophy/hyperplasia, epithelial cell metaplasia, and angiogenesis. These structural changes are believed to correlate with the severity of asthma and to some extent with the development of progressive lung function deterioration. The mechanism underlying airway angiogenesis in asthma and its precise clinical relevance has not yet been completely elucidated.¹

Asthma may best be described as a loosely defined syndrome characterized by respiratory symptoms, airways narrowing, and inflammation. Asthma is a common pulmonary condition that involves heightened bronchial hyperresponsiveness and

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reversible bronchoconstriction, together with acute-on-chronic inflammation that leads to airways remodeling. The most common causes predisposing for asthma include viral upper respiratory tract infections, cigarette smoke, cold temperatures, allergies, pets, and exercise. Symptoms of asthma include wheezing, intercostal and supraclavicular retraction, cough (worse at night), shortness of breath, chest pain, exercise intolerance, and limitation of daily activities, which should alert physicians to a diagnosis of possible asthma or an asthma exacerbation.^{2,3}

Allergic asthma is characterized by a specific pattern of inflammatory attributes driven by IgE-dependent triggering of resident tissue mast cells and characterized by the influx of basophils and eosinophils in inflamed airways. The interaction between inflammatory cells and structural cells in asthmatic airways is complex. Several cytokines and growth factors released during allergic airway inflammation and remodeling are responsible for increasing basal levels of vascular endothelial growth factor in fibroblasts and smooth muscle cells.^{1,4,5}

In spite of its great burden on public health care, our knowledge of the etiologic mechanisms underlying asthma, both genetic and environmental, is still very limited. One of the most promising approaches to expand further our understanding of the disease mechanisms involved is identification of the genetic variation that contributes to the risk of developing asthma.⁶

In recent years, research has mainly focused on detecting the genetic variations that predispose the individual to asthma. Three basic types of genetic study have been undertaken, ie, genetic linkage analysis, searches for focused candidate genes, and the modern genome-wide association studies performed with single nucleotide polymorphism (SNP) chips. Extensive epidemiologic studies have made little progress in determining the individual's susceptibility to asthma. The molecular genetic studies of asthma offer the prospect of defining this susceptibility at a genetic level, and allow more precise studies on the etiology of asthma to be undertaken.⁷⁻⁹

Family studies using linkage methodologies conducted to date have not been very successful in identifying the genetic determinants of this complex disease.¹⁰ The revolution in genotyping technology with high-throughput methods now allows genotyping of greater numbers of SNPs in large cohort genome-wide association studies. Most of the genes uncovered during recent years with the genome-wide approach are novel, and were not even considered in the old candidate gene studies. Asthma is an example of a complex disease where

several common susceptibility alleles affect the disease risk in varying combinations, but in a manner such that each gene contributes only a minor impact.¹¹ The downstream biologic effects of the majority of these genes and their proteins are still unknown. Expression studies of these genes and proteins could allow us to uncover some of their effects.¹²

If some of the unexplained heritability in genome-wide association studies was due to interactions then, rather than discovering interactions per se, one goal might be to use these interactions in order to discover novel genes that act synergistically with other factors without having demonstrable marginal effects.¹³

Materials and methods

Literature search

A literature search in PubMed (<http://www.ncbi.nlm.nih.gov/pubmed>) was carried out in September 2009 with the search phrase “asthma and SNP”. The list obtained was then manually annotated, and a list of SNP modifications in asthma was collected. Even though this is probably not a complete list, it gives a very good idea of the found and proposed SNP modifications linked to asthma. The list of SNPs claimed to be associated with asthma (Table 1) was used for further analyses.

Protein–protein interaction networks

We used a web-based protein interaction network analysis platform (PINA), which integrates protein–protein interaction data from six databases. The Cytoscape757 program provides a network construction tool, which uses our protein–protein interaction data as the baseline. The SwissProt names without the tag “__HUMAN” are used throughout this study, if not otherwise stated.

The gene ontology categories provide a controlled vocabulary to describe the gene and the gene product attributes of any organism. The enrichment analysis of proteins in the various gene ontology (GO Slim) categories was carried out essentially as described elsewhere.¹⁴⁻¹⁶

When the analysis within these annotated categories was carried out, we searched for enriched categories. The basic question was: Is the category “toll-like receptors” or the key phrase “cytokine–cytokine receptor interaction” within the observed SNP-related network enriched when compared with the background set of the whole human proteome? If such a phenomenon was observed, it could suggest that these enriched categories and keywords play a role within the SNP-related protein network found in asthma patients.

Table 1 A list of 127 genes and their corresponding proteins, where SNP(s) have been found to relate to allergic diseases

GENE	PROTEIN	PMID	ALOX5AP	AL5AP	18547289	TGFB1	TGFB1	17333284
ARG1	ARG1	19281908	LTA4H	LKHA4	18547289	IL27	IL27A	17318299
ARG2	ARG2	19281908	CTTN	SRC8	18521921	MYLK	MYLK	17266121
IL10	IL10	19222424	PTGS2	PGH2	18489027	CD40	TNR5	17255560
PPARG	PPARG	19217272	PTPRD	PTPRD	18414509	CCL11	CCL11	17220216
IL1RL1	IL1RL1	19198610	CRHR2	CRFR2	18408560	CYSLTR1	CLTR1	17154652
WDR36	WDR36	19198610	CHI3L1	CH3L1	18403759	CYSLTR2	CLTR2	17154652
IL33	IL33	19198610	IL4	IL4	18396027	CCR5	CCR5	17154652
MYB	MYB	19198610	IFNG	IFNG	18385742	SOCS1	SOCS1	17099141
CTNNA3	CTNA3	19187332	IL18R1	IL18R	18382474	IL9R	IL9R	17083349
TLR2	TLR2	19148143	GSTP1	GSTP1	18335111	RIPK2	RIPK2	17075290
ADAM33	ADA33	19146844	TLR9	TLR9	18312481	CHRM1	ACM1	16931638
CD14	CD14	19096003	NPSR1	B7ZMA2	18305139	IL7R	IL7RA	16890764
HLX1	HLX	19038437	TNC	TENA	18305139	LIFR	LIFR	16890764
TNF	TNFA	19004142	NPPA	ANF	18294255	AOAH	AOAH	16815140
GSTM1	GSTM1	18988661	IL18	IL18	18200581	PDGFRA	PGFRA	16804324
CMA1	CMA1	18973102	INPP4A	INP4A	18187694	NPS	NPS	16790440
CXCR3	CXCR3	18962861	UGRP1	SG3A2	18089940	SFRS8	SFRS8	16738036
CRTL2	GPR44	18946232	CCR3	CCR3	17983872	MMP-9	MMP-9	16631427
IL6	IL6	18810365	IL5RA	IL5RA	17983872	ICAM1	ICAM1	16625213
IL1RN	IL1RA	18810365	FCER2	FCER2	17980418	MBL2	MBL2	16487239
IL21	IL21	18802358	EDN1	EDN1	17960156	DEFB1	DEFB1	16435024
IL1B	IL1B	18773331	TBX21	TBX21	17949803	IL16	IL16	16387589
MYLK	MYLK	18766098	PTGER3	PE2R3	17877755	ALOX5	LOX5	16361798
ORMDL3	ORML3	18754760	HLA-G	HLAG	17847008	C3	CO3	16355111
ACE	ACE	18727619	PHF11	PHF11	17702965	TLR4	TLR4	16215326
CYBA	CY24A	18716406	TGF-beta1	TGFB1	17673695	DPP10	DPP10	15986064
SERPINE1	PAI1	18714537	KAT5	KAT5	17672871	LTA	TNFB	15969671
CCL24	CCL24	18712274	GCLC	GSH1	17643973	GATA3	GATA3	15637551
CCL26	CCL26	18712274	HAVCR1	TIMDI	17570927	CLCA1	CLCA1	15318163
CCL17	CCL17	18691306	ITGB3	ITB3	17556058	TLR10	TLR10	15201134
MS4A2	FCERB	18691306	ADH5	ADHX	17543375	DAP3	RT29	15179560
IL13	IL13	18691306	PTGDR	PD2R	17538632	IL15	IL15	15131572
IL4R	IL4RA	18691306	STAT4	STAT4	17532201	NOS1	NOS1	14767694
TLR7	TLR7	18682521	STAT6	STAT6	17519224	TAPI	TAPI	12640628
TLR8	TLR8	18682521	PTGER2	PE2R2	17496729	CTLA4	CTLA4	12417883
GRK5	GRK5	18622265	PTGER4	PE2R4	17496729	AICDA	AICDA	11544457
CHIA	CHIA	18602573	PTGIR	PI2R	17496729	CCL2	CCL2	11544456
FCER1G	FCERG	18595682	TBXA2R	TA2R	17496729	IKBKAP	ELP1	11281413
FCER1A	FCERA	18595682	TNF	TNFA	17450233	C5	CO5	10973279
ADRB2	ADRB2	18558635	SFTPA2	SFPA2	17407567	CCL5	CCL5	11197694
TLR1	TLR1	18547625	PLAU	UROK	17363771	LTC4S	LTC4S	10970818
TLR6	TLR6	18547625	RNASE3	ECP	17362255			

Notes: The proteins displayed on the common pathway as shown in Figure 2 are marked in bold. The SwissProt names for proteins are used without the tag _HUMAN throughout the study.

Results

We first searched the literature published for 2000–2009 on the association of SNPs with asthma. The search produced 251 articles, from which a list of 127 genes and their corresponding proteins linked to asthma was compiled (Table 1). Next, we generated a protein–protein interaction network for these 127 proteins. The interacting protein partners for each of these proteins were searched with PINA, which integrates six different protein–protein interaction databanks.¹⁷

We first pulled down all the protein–protein interactions of these putatively asthma-related proteins, which resulted in a very large network with 1073 proteins (nodes) and 1421 connections (edges) between them. This large data set was imported to Cytoscape and the network is displayed in Figure 1A.¹⁸ In order to facilitate data mining within the protein–protein interaction network, we created a more stringent query yielding a smaller subset of the original large network. This new limited network contained

Figure 1 **A)** The protein–protein binding network is built on the basis on two sources of data, ie, the asthma pathogenesis literature published on the putative association of single nucleotide polymorphisms within genes coding corresponding proteins and the protein–protein interaction network data for all these proteins. The original dataset from the literature with 127 genes with single nucleotide polymorphism modifications were converted to corresponding proteins (synonymous marked as green and nonsynonymous marked as red nodes). The interacting protein partners for each of these proteins (yellow nodes) were searched using protein interaction network analysis. **B)** The Figure 1A network was modified so that each interacting protein (yellow nodes) binds to at least two (synonymous marked as green and nonsynonymous marked as red nodes) single nucleotide polymorphism proteins. A Cytoscape file can be loaded from the online supporting information. All the proteins are marked with the SwissProt name, but without the tag “_HUMAN” for clarity (see Table 1).

interacting proteins which were bound to at least two other proteins identified to carry asthma-related SNP modifications in their corresponding genes. This data set contained only 309 proteins (96 of which were SNP-related proteins, 213 interacting proteins, and 644 connections between them, Figure 1B).

Because such a network is far too large to be analyzed visually in a meaningful manner, we decided to perform a gene ontology enrichment analysis. When the analysis within the gene ontology-molecular function categories was performed, several strongly enriched classes were observed, as shown in Table 2. “Protein binding” and several of its daughter categories, including receptor binding, cytokine binding, growth factor binding, interleukin binding, transcription factor binding, chemokine binding, and pattern binding, were among the most significantly enriched categories. Furthermore, protein kinase activity, including tyrosine and serine/threonine kinase activity, as well as endopeptidase activity, was significantly enriched within the asthma-related SNP-modified genes and their corresponding proteins.

To utilize this asthma-related network further, we searched for connections between the asthma-related SNP proteins only (Figure 2). The results showed that while 31 of the 96 proteins do not interact with any other SNP protein, 14 interact with each other, 20 form a pair, two triplets, one a quartet, and two larger networks contain several connected asthma-related SNP proteins. The chemokine network includes eight proteins and the toll-like receptor network 13

proteins, all shown to carry asthma-related SNP modifications among their corresponding genes. We expanded to the most interesting networks with PINA and were able to create two new networks in Cytoscape. The chemokine network (green proteins, Figure 3) shows eight asthma-related SNP proteins binding to each other. Furthermore, a great number of other chemokines and their receptors also show an interaction within this network (yellow proteins, Figure 3).

Likewise, the other new subnetwork displays 13 asthma-related SNP proteins interacting with each other (Figure 4). This toll-like receptor/cytokine network also contains a large group of novel proteins, including toll-like receptors and cytokines, as well as signal transduction molecules (yellow proteins, Figure 4). Such an enlarged network of interacting proteins could putatively be used to search for novel proteins having a crucial role in the development of asthma-related inflammatory reactions. Of great interest are the nonsynonymous SNPs, which code for an alternative amino acid sequence of proteins. The red proteins within Figure 4, ie, TLR1, TLR4, TLR5, TLR6, TLR10, and IL4R and IL13, are among these. This small subnetwork of toll-like receptor-related proteins and their 40 interacting proteins was further analyzed. This analysis showed that these 40 interacting proteins have already been reported to carry almost 1000 nonsynonymous SNPs coding for alternative protein sequences (Table 3).

Finally, we manually annotated the whole set of 309 proteins in Figure 1B. A thorough analysis showed that the

Table 2 Gene ontology (GO) enrichment on the SNP-related proteins and their first binding partners

GO-ID	P-value	x	nX	Description
5515	6.5707E-69	256	7023	Protein binding
4871	4.9020E-44	130	2217	Signal transducer activity
6950	8.1435E-63	120	1258	Response to stress
19538	2.1909E-15	98	2768	Protein metabolic process
4872	2.9275E-20	84	1792	Receptor activity
5102	3.3451E-40	77	765	Receptor binding
166	3.8108E-3	50	2146	Nucleotide binding
43234	8.1977E-5	48	1712	Protein complex
7267	3.5518E-16	40	535	Cell-cell signaling
5509	1.1569E-8	40	932	Calcium ion binding
16740	7.7375E-3	40	1693	Transferase activity
30528	5.7626E-3	35	1405	Transcription regulator activity
16265	2.2804E-12	33	485	Death
8219	2.2804E-12	33	485	Cell death
16301	4.3301E-7	33	788	Kinase activity
30154	6.4206E-5	33	1001	Cell differentiation
50222	4.2327E-9	31	580	Protein kinase activity
8233	6.9718E-7	28	616	Peptidase activity
16023	3.4546E-7	21	359	Cytoplasmic membrane-bounded vesicle
6629	1.1188E-2	20	716	Lipid metabolic process
5768	2.8622E-8	18	228	Endosome
8283	1.8276E-5	16	290	Cell proliferation
8092	1.0819E-2	14	437	Cytoskeletal protein binding
30246	8.7209E-4	13	292	Carbohydrate binding
8289	8.3026E-3	13	380	Mol funct lipid binding
9719	5.2362E-6	11	125	Response to endogenous stimulus
3779	2.5076E-2	10	305	Actin binding
16032	3.2732E-5	7	57	Viral reproduction

Notes: Go-ID is the GO class, x represents of the number of observed proteins in our data set, nX is the number of proteins in the background comparison dataset.

most common class of annotations for these proteins was “cytokine–cytokine receptors”. We have generated a network of these proteins in Figure 5 (green proteins) and enlarged the pathway by also including their interacting proteins (yellow proteins, Figure 5). A strong input of other chemokines and signal transduction proteins is also seen here.

Discussion

Asthma is a major burden for health care worldwide. Although the pathophysiology of asthma has been studied intensively during recent years, much more work needs to be done before it will be possible to prevent the onset of symptoms of asthma and to cure patients. A great number of genetic analyses have been conducted with asthma patients describing SNPs in the coding sequences of several proteins and intergenic, nontranslated regions close to protein coding sequences. During recent years, more than 100 candidate

genes harboring these SNP modifications have been associated with bronchial asthma.

Our aim in the present work was to start a systems level analysis of the putative pathogenetic mechanisms involved in asthma. Databases, and especially their integrated and merged data warehouses, allow rapid and convenient access to the data publicly available. Furthermore, it is possible to integrate one's own data on top of the publicly available data and thus enhance the level of information. In this study, starting with 127 SNP-modified genes converted to the corresponding proteins obtained from published data, we could identify a protein–protein interaction network of over 1000 proteins.

Instead of focusing our analysis only on one or a few altered genes or their corresponding proteins, we attempted to generate larger protein–protein interaction networks from our data. No single database alone can provide such a connected network; in order to understand the new systems levels of diseases, we decided to build up an integrated data warehouse combining information from several databases. We are now able to show how the 127 proteins of asthma patients were connected together in a putative network. One of the most prominent observations in this network was that it was strongly enriched with protein binding, signal transduction, and peptidase functions. The network contained several subnetworks enriched with toll-like receptors or chemokines.

The innate immune system responds to invading pathogens by activating a proinflammatory cascade aiming at eradicating the invading agents. Pattern recognition receptors are a crucial part of this innate immune reaction.¹⁹ A variety of intra- and extracellular pattern recognition receptors are known today, of which toll-like receptors are involved in the recognition of molecular structures specific for microbial pathogens.²⁰ Two main categories of toll-like receptors exist, ie, cell surface receptors and receptors localized in the endosome. It is important to make this distinction because surface toll-like receptors bind molecules on the bacterial cell wall, such as bacterial lipopeptides (TLR2) or lipopolysaccharide (TLR4), whereas endosomal toll-like receptors that are activated by microbial nucleic acids are less readily accessible.²¹

Research has shown that toll-like receptors can now be divided into two groups on the basis of their subcellular localization. The first group (TLR1, TLR2, TLR4, TLR5, and TLR6) is present on the surface of the cell, and recognizes lipid structures and, in the case of TLR5, the protein flagellin. The second group (TLR3, TLR7, TLR8, and TLR9) resides intracellularly and recognizes nucleic acids. The reason for

Figure 4 The toll-like receptor network displays 13 asthma-related single nucleotide polymorphism proteins (synonymous marked as green and nonsynonymous marked as red nodes) interacting with each other and their receptors (yellow nodes). The red nodes represent proteins TLR1, TLR4, TLR5, TLR6, TLR10, and IL4R and IL13, which are the nonsynonymous single nucleotide polymorphisms coding for an alternative amino acid sequence of proteins.

the different localization of toll-like receptors may be that TLR1, TLR2, TLR4, TLR5, and TLR6 recognize markers on the surface of pathogens, while TLR3, TLR7, TLR8, and TLR9 recognize nucleic acids derived from the genome of viruses and bacteria. It has become increasingly apparent that the localization and traffic of toll-like receptors within the cell is an important mechanism whereby toll-like receptors sense their ligands. Importantly, the traffic of certain toll-like receptors during signaling can also prevent overactivation of the toll-like receptor signaling pathways.²²

A small proportion of SNP mutations can cause the altered amino acid sequence in the corresponding protein directly. Genes that have previously been shown to have a SNP mutation leading to a change in the actual protein structure and which have also been linked with asthma, are presented in red in Figure 4 (TLR1, TLR4, TLR5, TLR6, TLR10, and IL4R and IL13).^{23–26} Genes colored green have been shown to be asthma-related SNPs, but these SNP mutations do not have any effect on protein structures (IL4, STAT6, CD14, RIPK2,

TLR2, TLR9). There are altogether 54 protein coding genes in the same pathway, with the aforementioned asthma-related genes having a SNP mutation (Figure 4).

The aberrant activation of toll-like receptor pathways, on the other hand, has been implicated in various chronic and autoimmune diseases affecting the gastrointestinal tract, central nervous system, kidneys, skin, lungs, and joints, whereby both exogenous and endogenous ligands have been suggested to act as toll-like receptor activators. The finding that intracellular proteins or the products of protein cleavage can act as endogenous ligands for toll-like receptors supports the hypothesis that toll-like receptors are important in mediating the response not only to infections but also to stress, damage, and death of cells in general.²⁷

New developments in the fields of allergy and immunology have yielded a variety of novel therapeutic approaches in recent years, resulting in more agents at the clinical trial stage as well. Among the therapeutic approaches are the toll-like receptor agonists, immunostimulatory

Table 3 Forty interacting proteins from the toll-like receptor-pathway (Figure 4), which have been reported to carry altogether almost 1000 nonsynonymous single nucleotide polymorphisms coding for alternative protein sequences

Gene	SNP	Type	PMID
TLR1	rs5743594	intronic	18547625
	rs5743595	Intronic	18547625
	rs4833095	nonsynonymous-coding	18547625
TLR2	rs4696480	intronic	18547625
	rs1898830	intronic	18547625
	rs3804099	synonymous-coding	18547625
	rs2289318	upstream	19096003
	rs3775291	nonsynonymous-coding	18547625
TLR4	rs2737190	upstream	18547625
	rs10759932	upstream	18547625
	rs4986791	nonsynonymous-coding	18547625
	rs11536889	downstream/3prime utr	19096003
	rs7045953	upstream	19096003
TLR5	rs5744168	stop gained	18547625
	rs2072493	nonsynonymous-coding	18547625
	rs5744174	nonsynonymous-coding	18547625
TLR6	rs5743789	upstream/intronic	18547625
	rs5743819	nonsynonymous-coding	18547625
TLR7	rs179008	nonsynonymous-coding	18547625, 18682521
TLR8	rs3761624	upstream	18547625
	rs2407992	synonymous-coding	18682521
TLR9	rs187084	upstream	18547625, 19247692
	rs5743836	upstream	18547625
	rs352143	upstream	19096003
	rs352163	upstream	19096003
	rs353547	downstream	19096003
TLR10	rs11096956	synonymous-coding	18547625
	rs4129009	nonsynonymous-coding	18547625
	rs1109657	nonsynonymous-coding	19247692
IL13	rs1881457	within_non_coding_gene/upstream	19796199
	rs1800925	within_non_coding_gene/upstream	19796199
	rs2243204	downstream	19247692
	rs1295686	within noncoding gene/intronic	19247692
	rs20541	nonsynonymous-coding/within_non_coding_gene	19254294
IL4	rs2243250	upstream	18263811
IL4RA	rs1805010	non_synonymous_coding/downstream	19796199
STAT6	rs324015	3prime-utr	19247692
	rs324011	intronic	19247692
RIPK2	rs2293807	downstream	17075290
CD14	rs2569190	3prime-utr/intronic	18925877

oligodeoxynucleotides, oral and parenteral cytokine blockers, and specific cytokine receptor antagonists. However, a much better understanding of the “big picture of this systems inflammatory disease” must still be obtained before more target therapeutic approaches can be designed.²⁸

Compared with the latest reports in which only one gene at a time has been in focus when analyzing the pathogenetic mechanisms of multifactorial diseases like asthma, we have now focused on the entire set of 127 genes and their corresponding proteins. Of these 127 genes, 96 could be connected to a same gene-mRNA-protein and protein-protein interaction network, and were found to be enriched significantly

with protein binding, signal transduction, and endopeptidase activities. Taken together, we showed in this study using our in silico analysis framework and the outside databases that we can increase the level of knowledge by performing systems level analyses of previously characterized genes carrying SNPs related to asthma.

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Disclosure

The authors report no conflicts of interest in this work.

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- ## Disclosure
- The authors report no conflicts of interest in this work.
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