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Information

Description

Core Analysis Results Tutorial**Overview**

A **Core Analysis** has multiple ways of helping you find biological insights by automatically using the curated information from the QIAGEN Knowledge Base to put molecules in your 'omics dataset into context. For this tutorial, we will walk through the interpretation of a transcriptomics dataset.

The **Summary** tab displays the top results for all the analyses. See [Analysis Summary Page](#).

Canonical Pathways tab displays your molecules of interest within well-established signaling or metabolic pathways. See [Canonical Pathways Overview](#).

Upstream Analysis tab predicts which upstream regulators (any molecule that can influence the transcription or expression of another molecule) might be activated or inhibited to explain the expression changes in your dataset. See [Upstream Regulator Analysis](#). It also connects upstream regulators into signaling cascades called [Mechanistic Networks](#). If you have Advanced Analytics in IPA this tab can also display [Causal Networks](#), or hierarchical networks or regulators controlled by a master regulator.

Diseases & Functions relates molecules in your dataset to known disease states and biological functions. See [Downstream Effects Analysis Tutorial](#).

Regulator Effects displays hypotheses for how a phenotype, function or disease is regulated in your dataset by activated or inhibited upstream regulators See [Regulator Effects](#).

Networks displays non-directional interaction networks of molecules based on known relationships in the QIAGEN Knowledge Base to your molecules of interest. See [What are Networks?](#)

Use these tools to find insights that are most relevant to your experimental model or question. Note that this tutorial will draw on only a *subset* of these features, due to length considerations.

Scenario

You have RNA microarray data from the white blood cells drawn from children with childhood exacerbated asthma compared to the convalescent state. For this example, we will use expression data from PBMCs from PMID [19620293](#). You would like to know if (and how) the data supports involvement of immune/inflammatory responses in acute asthma attack. You can also find novel gene-to-disease associations that can be followed up and confirmed with future wet bench experiments.

Tasks

Open the **Core Analysis** for your microarray data.

Use the **Summary Page** to quickly identify promising directions for exploration.

Explore the results for areas of importance to your research:

View **Canonical Pathways** that contain significant numbers of genes from your dataset.

Use **MAP** (Molecule Activity Predictor) to predict effects on functional endpoints in a canonical pathway.

Overlay **Biomarkers** that identify genes in the TREM1 Signaling Pathway that are used as efficacy indicators for asthma treatments.

Explore the impact on downstream **diseases and functions**.

Use **Regulator Effects** to see a hypotheses of how activation of certain upstream regulators may lead to outcomes like asthma.

Steps

1) Start IPA

2) In the **Project Manager**, double click My Projects > Example Analyses > Analyses > Childhood exacerbated asthma GSE16032. The analysis opens in a new window displaying the **Summary** tab:

Top Canonical Pathways

Name	p-value	Overlap
TREM1 Signaling	1.46E-09	32.9 % 23/70
Inflammasome pathway	7.57E-09	57.1 % 12/21
Role of Pattern Recognition Receptors in Recognition of Bacteria and Viruses	5.62E-08	23.1 % 30/130
phagosome formation	4.52E-07	23.0 % 26/113
Altered T Cell and B Cell Signaling in Rheumatoid Arthritis	1.56E-06	25.6 % 20/78

Top Upstream Regulators

Upstream Regulators

Name	p-value of overlap	Predicted Activation
Immunoglobulin	4.20E-29	Inhibited
lipopolysaccharide	9.23E-27	Activated
IL13	1.34E-25	
IL4	6.09E-21	Activated
TGFB1	6.26E-20	Activated

Causal Networks

Name	p-value of overlap	Predicted Activation
TH2 Cytokine	7.11E-32	Activated
IFNA8	8.86E-32	Activated
CD300LF	1.94E-31	Inhibited
TH1 Cytokine	1.27E-30	Activated
fluticasone	1.30E-30	

Top Diseases and Bio Functions

Diseases and Disorders

Name	p-value range	# Molecules
Infectious Diseases	1.55E-06 – 9.38E-24	340
Inflammatory Response	1.60E-06 – 7.93E-23	507
Immunological Disease	1.55E-06 – 1.31E-21	623
Respiratory Disease	6.48E-07 – 1.84E-21	180
Inflammatory Disease	1.11E-06 – 1.95E-21	384

Molecular and Cellular Functions

Name	p-value range	# Molecules
Cell-To-Cell Signaling and Interaction	1.60E-06 – 1.91E-26	366
Cellular Function and Maintenance	1.52E-06 – 1.04E-22	382
Cellular Movement	1.27E-06 – 6.12E-21	377
Cellular Development	1.54E-06 – 8.44E-21	520
Cell Morphology	1.46E-06 – 2.03E-20	339

Physiological System Development and Function

Name	p-value range	# Molecules
------	---------------	-------------

Hematological System Development and Function		1.60E-06 – 1.91E-26	425
Tissue Morphology		1.55E-06 – 4.65E-23	314
Immune Cell Trafficking		1.60E-06 – 7.93E-23	254
Hematopoiesis		1.54E-06 – 8.44E-21	225
Lymphoid Tissue Structure and Development		8.84E-07 – 4.10E-16	255

Top Tox Functions

Assays: Clinical Chemistry and Hematology

Name	p-value range	# Molecules
Increased Levels of Red Blood Cells	2.35E-03 – 2.35E-03	16
Increased Levels of ALT	4.33E-01 – 3.31E-03	7
Increased Levels of Hematocrit	7.88E-03 – 7.88E-03	14
Increased Levels of Creatinine	2.99E-01 – 7.93E-02	8
Increased Levels of AST	9.00E-02 – 9.00E-02	3

Cardiotoxicity

Name	p-value range	# Molecules
Cardiac Infarction	4.42E-01 – 2.53E-09	69
Congenital Heart Anomaly	1.00E00 – 6.32E-05	49
Cardiac Pulmonary Embolism	2.09E-01 – 1.65E-03	8
Cardiac Arrhythmia	1.00E00 – 3.01E-03	41
Cardiac Fibrosis	3.33E-01 – 3.65E-03	25

Hepatotoxicity

Name	p-value range	# Molecules
Liver Inflammation/Hepatitis	5.18E-01 – 2.08E-10	64
Liver Damage	4.80E-01 – 1.68E-08	56
Liver Hyperbilirubinemia	1.50E-01 – 8.80E-06	10
Hepatocellular Carcinoma	1.00E00 – 9.40E-05	109
Liver Hyperplasia/Hyperproliferation	1.00E00 – 9.40E-05	506

Nephrotoxicity

Name	p-value range	# Molecules
Renal Damage	4.95E-01 – 3.50E-08	38
Renal Inflammation	5.56E-01 – 4.13E-08	60
Renal Nephritis	5.56E-01 – 4.13E-08	60
Kidney Failure	4.87E-01 – 4.07E-05	47
Renal Necrosis/Cell Death	5.56E-01 – 4.10E-05	63

Top Regulator Effect Networks

ID Regulators	Diseases & Functions	Consistency Score
1 APOE,Ifnar,IRF3,IRF6,IRF7,MET,NOS2,TFEB,TGM2,TLR2	endocytosis by eukaryotic cells (+5 more)	35.833
2 ADAMTS12,BID,IL17R,IL17RA,TNFSF12	accumulation of granulocytes (+9 more)	31.069
3 EGR1,FN1,IL17RA,LDL,LTBP1 (+6 more)	activation of neutrophils,adhesion of granulocytes (+3 more)	28.482
4 CYP2E1,EGR1,FN1,Ifnar (+4 more)	activation of granulocytes (+7 more)	26.167
5 ADAMTS12,Alpha catenin,IL17RA,INSIG1,mir-223,TBX5 (+2 more)	accumulation of granulocytes (+11 more)	23.326

Top Networks

ID Associated Network Functions	Score
1 Cell Cycle, Cellular Assembly and Organization, DNA Replication, Recombination, and Repair	38
2 Hereditary Disorder, Neurological Disease, Organismal Injury and Abnormalities	36
3 Gene Expression, Embryonic Development, Lymphoid Tissue Structure and Development	34
4 Hereditary Disorder, Neurological Disease, Organismal Injury and Abnormalities	34
5 Connective Tissue Disorders, Developmental Disorder, Hematological Disease	32

Top Tox Lists

Name	p-value	Overlap
Increases Renal Nephritis	1.63E-08	36.0 % 18/50
Increases Renal Damage	2.06E-07	27.6 % 21/76
Increases Liver Hepatitis	1.33E-05	27.3 % 15/55
Increases Liver Damage	5.23E-05	19.5 % 22/113
Renal Necrosis/Cell Death	1.66E-04	12.4 % 63/509

Top Analysis-Ready Molecules

Exp Log Ratio ↑

Molecules	Value	Chart
IL1R2*	↑ 5.418	
MS4A4A*	↑ 3.333	
VSIG4	↑ 3.206	
VNN1*	↑ 3.088	
CD163*	↑ 2.736	
GPR34	↑ 2.609	
FKBP5*	↑ 2.550	
C3AR1	↑ 2.515	
CCR2*	↑ 2.468	
HBD	↑ 2.464	

Exp Log Ratio ↓

Molecules	Value	Chart
IFNG	↓ -1.591	
SPTBN1*	↓ -1.531	
WHAMM	↓ -1.371	
DUSP8*	↓ -1.298	
FGF9*	↓ -1.243	
CD160	↓ -1.222	
PLCL1*	↓ -1.184	
ID3	↓ -1.178	
LINC00282	↓ -1.130	
ZNF331*	↓ -1.124	

The tab (shown above) lists the top results for the different types of analyses that are included in Core Analysis. For this dataset:

Several immune-related Canonical Pathways are involved in asthma, such as the TREM1 Signaling and the Inflammasome Pathway. Key cytokines are activated, such as IL4 and IFNA8. Biological processes in the categories of Infectious Disease, Respiratory Disease, and Inflammatory Response are involved. Regulator Effects networks indicate that several upstream regulators appear to drive accumulation of granulocytes and activation of neutrophils. Interaction networks are discovered that are involved in processes such as cell cycle and organismal injury.

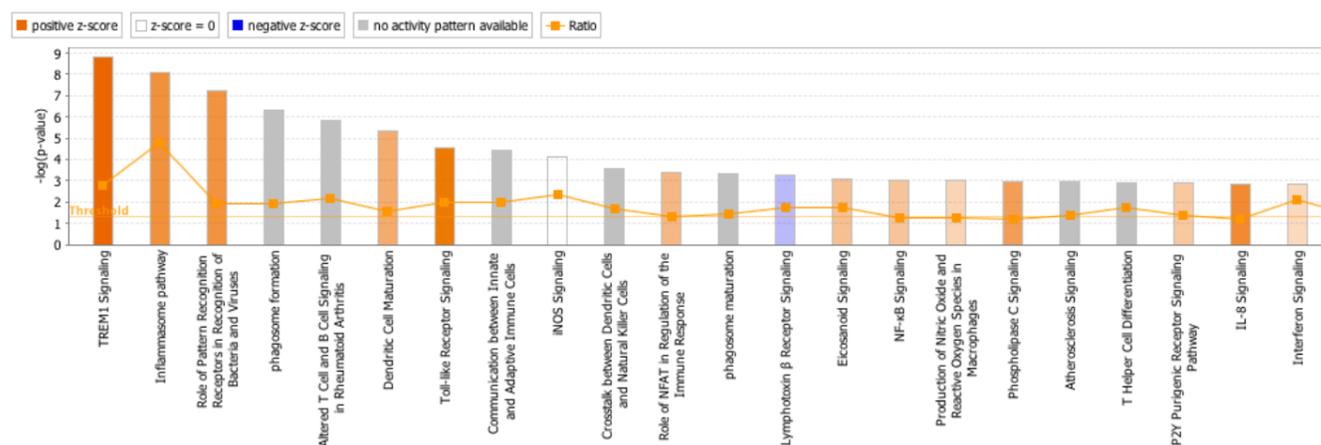
The p-values associated with the results are indicated on a scale in the Summary with red dots. Many of the results represent categories that contain many sub-categories, so a cluster of red dots is shown with each red dot indicating the p-value for each sub-category.

Note: The QIAGEN Knowledge Base is updated on a weekly basis. Details and information in these screen shots may not appear exactly the same as in the current version of this particular analysis.

Canonical Pathways tab

3) Click on the **Canonical Pathways** tab. Canonical Pathways provide information about what is known from the literature to occur on the cellular level in signaling and metabolic cascades. The TREM1 Signaling pathway is predicted to be activated (it has an orange colored bar in the bar chart) based on the expression pattern of the genes in this dataset. TREM1 is an important cell surface signaling molecule involved in the immune response. Click on the orange bar to display results from the TREM1 Signaling pathway.

The **upper pane** shows a bar chart where the height of the bars indicates the significance of the overlap of the molecules in your dataset to the pathways in the QIAGEN Knowledge Base. Significance values calculated based on the Fisher's right tailed exact test and the $-\log(p\text{-value})$ are displayed on the y-axis of the bar chart. The taller the bar, the more significant the overlap of your dataset with the pathway. See [Canonical Pathways for a Dataset](#) to learn more about the calculations that are used. The color of the bars indicates whether the pathway is predicted to be activated (orange bars) or inhibited (blue bars), or if the pathway is ineligible for such an assessment (gray bars). See [Pathway Activity Overlay](#) for more details on the z-score calculation used to color the bars.



The lower pane:

Displays the molecules from your dataset that are members of the TREM1 pathway (which is shown because you clicked on that bar in the bar chart in the previous step). Note that some genes from the dataset may be "inside" groups or complexes that are on the pathway.

23 molecule(s) associated with **TREM1 Signaling** at Childhood exacerbated asthma GSE16032 [Ratio: 23/70 (0.329)] [z-score: 3.962] [p-value: 1.46E-09] [VIEW REPORT](#) [OPEN PATHWAY](#)

Symbol	Entrez Gene Name	Identifier	Expression Value		Expected	Location	Type(s)	Biomarker App	Drug(s)
			Exp Log Ratio	Exp p-value					
<input type="checkbox"/> AKT2*	AKT serine/threonine	1560689_s_at	-0.651	9.25E-03	↑Up	Cytoplasm	kinase		tricitiribine,
<input type="checkbox"/> CASP1*	caspase 1	206011_at*	1.137	1.19E-03	↑Up	Cytoplasm	peptidase		caspase 1
<input type="checkbox"/> CASP5	caspase 5	207500_at	0.754	5.36E-03	↑Up	Cytoplasm	peptidase		
<input type="checkbox"/> CD86*	CD86 molecule	205686_s_at*	0.732	5.32E-03	↑Up	Plasma	transmembrane	efficacy,	abatacept,
<input type="checkbox"/> CXCL8*	C-X-C motif chemokine	202859_x_at*	0.629	2.03E-03	↑Up	Extracellular	cytokine	diagnosis,	
<input type="checkbox"/> FCGR2B	Fc fragment of IgG	210889_s_at	1.052	4.19E-04	↑Up	Plasma	transmembrane		IgG
<input type="checkbox"/> IL10	interleukin 10	207433_at	0.858	1.81E-04	↑Up	Extracellular	cytokine	diagnosis,	
<input type="checkbox"/> IL18	interleukin 18	206295_at	0.767	2.62E-03	↑Up	Extracellular	cytokine	efficacy,	
<input type="checkbox"/> LAT2*	linker for activation of	221581_s_at*	0.769	2.27E-03	↑Up	Plasma	other		
<input type="checkbox"/> MAPK1*	mitogen-activated	1552263_at*	0.931	7.37E-03	↑Up	Cytoplasm	kinase	efficacy	MAP kinase1
<input type="checkbox"/> MYD88	myeloid differentiation	209124_at	0.697	1.25E-03	↑Up	Plasma	other		IMO-8400
<input type="checkbox"/> NFKB2*	nuclear factor kappa B	207535_s_at*	-0.602	8.29E-03	↑Up	Nucleus	transcription		
<input type="checkbox"/> NLR4*	NLR family CARD	1552553_a_at	1.810	4.16E-05	↑Up	Cytoplasm	other		
<input type="checkbox"/> NLRP12*	NLR family pyrin domain	223944_at*	0.870	4.31E-03	↑Up	Cytoplasm	other		
<input type="checkbox"/> PLCG2*	phospholipase C gamma	204613_at*	0.614	1.23E-04	↑Up	Cytoplasm	enzyme		
<input type="checkbox"/> TLR1	toll like receptor 1	210176_at	0.731	2.07E-03	↑Up	Plasma	transmembrane		
<input type="checkbox"/> TLR2	toll like receptor 2	204924_at	1.522	4.45E-05	↑Up	Plasma	transmembrane	diagnosis,	OM 174 lipid
<input type="checkbox"/> TLR4*	toll like receptor 4	232068_s_at*	1.828	3.44E-05	↑Up	Plasma	transmembrane	efficacy	resatorvid, OM
<input type="checkbox"/> TLR5	toll like receptor 5	210166_at	1.208	1.15E-03	↑Up	Plasma	transmembrane		
<input type="checkbox"/> TLR6*	toll like receptor 6	239021_at*	0.614	8.24E-04	↑Up	Plasma	transmembrane		
<input type="checkbox"/> TLR7*	toll like receptor 7	220146_at*	1.245	5.48E-03	↑Up	Plasma	transmembrane		3M-001, UC-
<input type="checkbox"/> TLR8*	toll like receptor 8	229560_at*	1.903	8.66E-06	↑Up	Plasma	transmembrane		VTX-2337,
<input type="checkbox"/> TREM1	triggering receptor	219434_at	1.012	7.74E-05	↑Up	Plasma	transmembrane	efficacy	

Selected/Total molecules: 0/23

The table conveys the following information that can be useful in interpreting the results of your experiment:

The Gene Symbol and Entrez Gene name, which identifies specific molecules in the pathway that are affected.

The observed expression changes in the dataset, in this example, log ratio and p-value.

The identifier uploaded from the dataset.

The expected "direction" (up or down regulated) for the molecule in the pathway, if the pathway were activated. This column is available only for pathways that are eligible for Pathway Activity Analysis.

The cellular location of the molecule and the molecule type.

If the molecule has been identified as a biomarker it will indicate the type of biomarker (diagnosis, efficacy, etc.), which is helpful in determining if the molecule has been studied in a particular disease state that is relevant to the one you are studying.

If there are any drugs that target a given molecule, information on how and in what biological context the molecule has been targeted.

If the pane is too small, you can drag the vertical partition bar up to adjust it.

4) Click the **Open Pathway** button at the top right of the lower pane. A new window displays the canonical pathway diagram.

Canonical pathways are usually directional, following the biological information flow in the cell. Arrows and top to bottom flow indicate upstream and downstream location, respectively.

Dataset molecules that meet the filters and data value cutoff criteria for up- and down-regulation are shaded red and green, respectively.

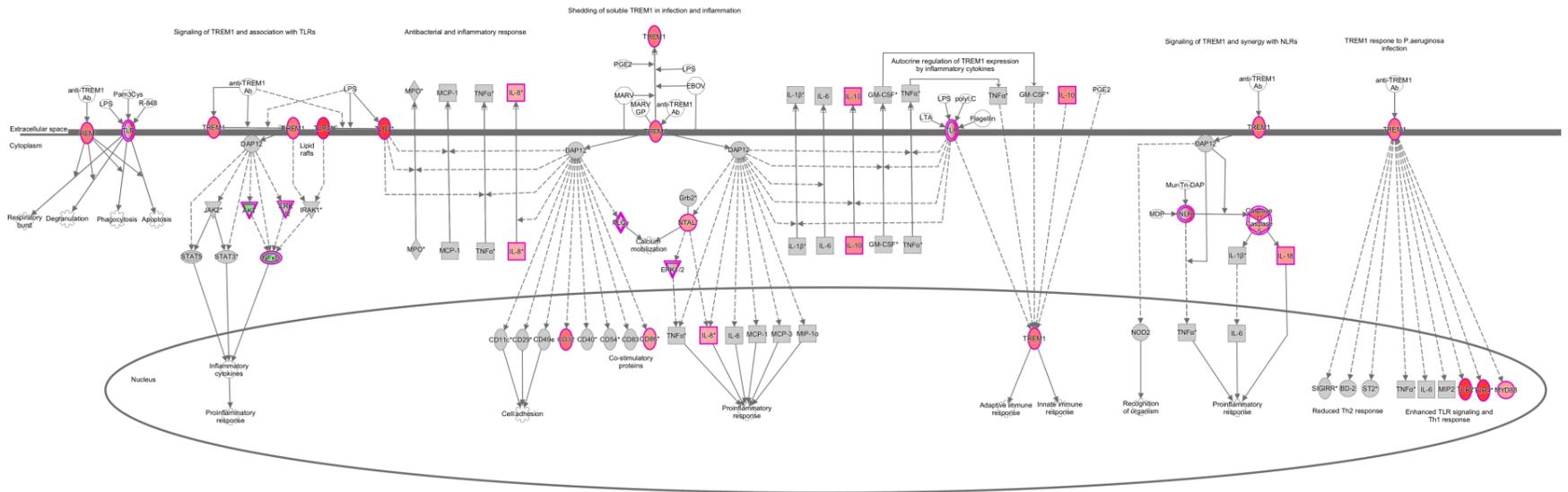
Dataset molecules that are below the cutoff or do not meet filter criteria are shaded grey in the pathway.

Pathway molecules that are not in your dataset are white.

The shapes and positions of the molecules in the pathway define gene type and cellular location, respectively.

Double outlined shapes represent groups of molecules (generally protein families). Groups can be multicolored to indicate they include both up- and down-regulated molecules as their members. Right click and choose "Show Members/Membership" to expand groups to see their individual members.

The pathway shows the biological picture of the activation of the TREM1 pathway, indicating that not only are several of the key receptors up-regulated in asthma, so are a number of downstream effectors.



5) To get a more "causal" picture of the biology represented in the pathway, go to **Overlay > MAP (Molecule Activity Predictor)** and click the Start Prediction button.

Canonical Pathways

TREM1 Signaling

Overlay: **MAP (Molecule Activity Predictor)**

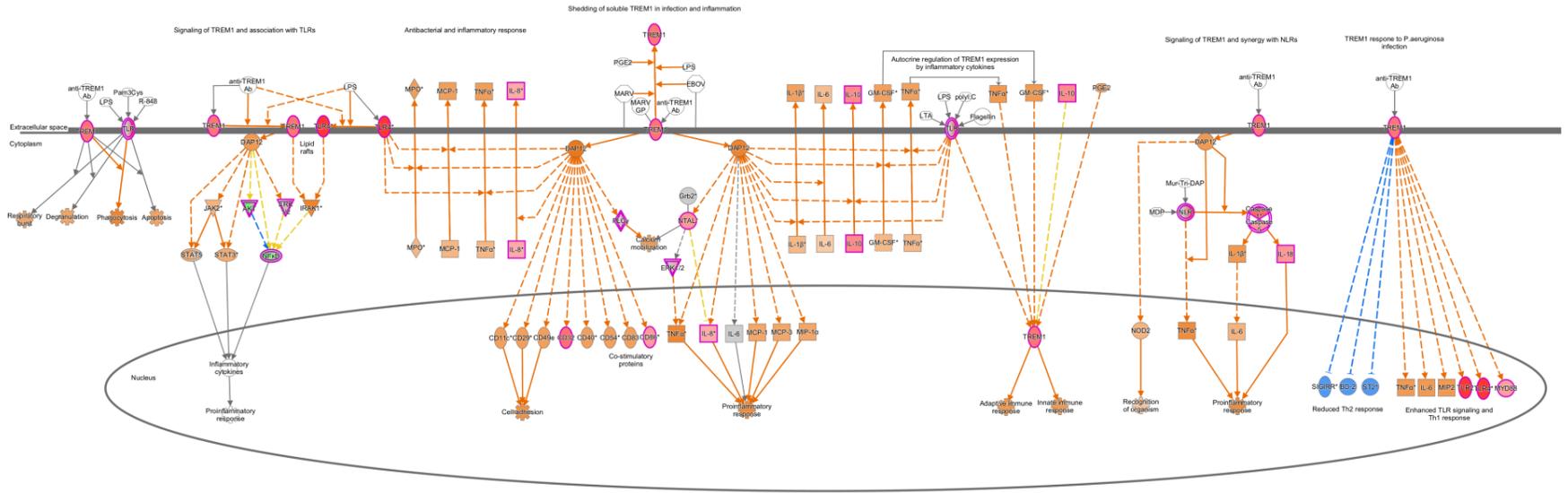
You can predict the upstream and downstream effects of activation or inhibition on other molecules. Begin by overlaying measurement values from a dataset or analysis, or interactively specifying activation in silico below.

Predict effect of dataset or in silico changes

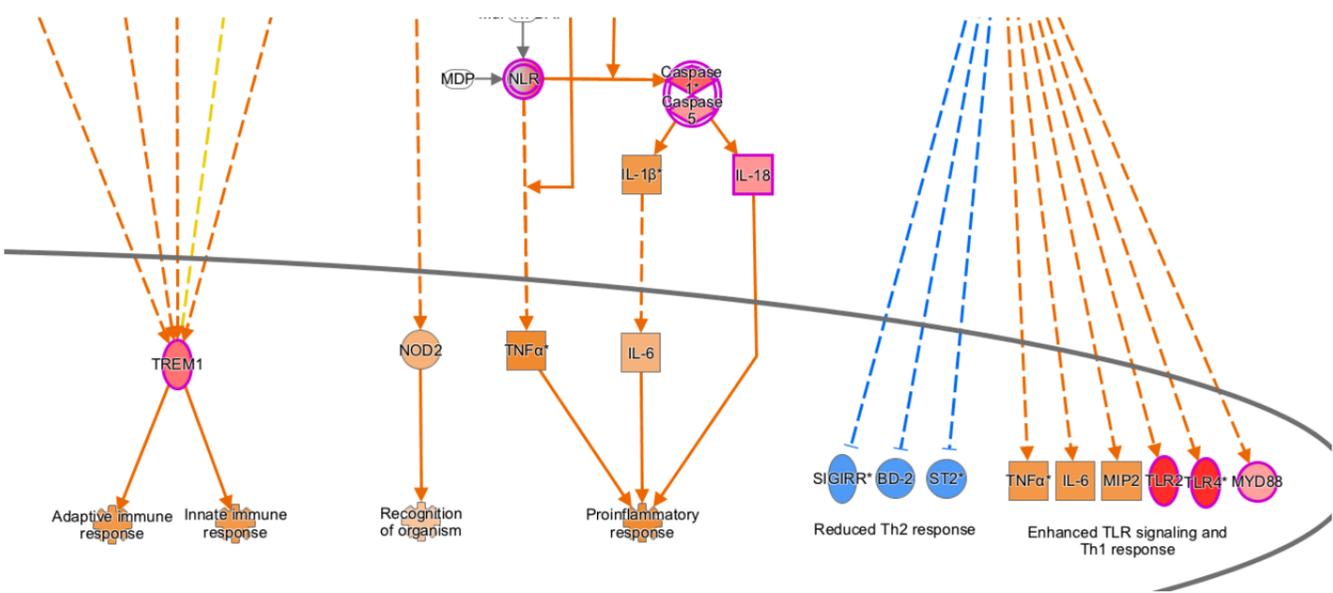
Display prediction legend

Predict effects:
Upstream and Downstream

This predicts the activity of nodes on the pathway that are not part of your dataset-- i.e. it colors as many gray or white nodes in the pathway orange or blue as possible based on the expected influence of the molecules in the dataset that are up or down regulated (i.e. the red and green molecules on the pathway). For example, if a gene is upregulated in your dataset, and the literature states that it activates a downstream gene that it is connected to in the pathway diagram, then that downstream gene is predicted to be activated. Orange nodes indicate predictions of activation, and blue nodes indicate prediction of inhibition.



Zooming in on the bottom of the pathway, you can see that several biological functions related to immunity are predicted to be increased in the pathway. For example, both adaptive and innate immune responses as well as proinflammatory response are predicted to be increased.



6) Double click on any relationship on the pathway to view the curated findings and the literature support for it. Clicking on the line between TNFα and the biological function "Proinflammatory response" brings up this summary:

IPA Relationships: TNF α |Proinflammatory response

Review the information that supports the gene-to-function relationship. Click the plus icon to view the reference information.

PlainText [EXPORT REFERENCES](#)

Ingenuity Relationships

causation [4]

In extracellular space, **TNF- α [TNF]** protein increases **Proinflammatory response**.

15634892 Gerosa F, Gobbi A, Zorzi P, Burg S, Briere F, Carra G, Trinchieri G. The reciprocal interaction of NK cells with plasmacytoid or myeloid dendritic cells profoundly affects innate resistance functions. *J Immunol.* 2005 Jan 15;174(2):727-34.
Source: Ingenuity Expert Findings

In nuclei from myeloid cells, **TNF α [TNF]** protein increases **Proinflammatory response**.

15385460 Lyke KE, Burges R, Cissoko Y, Sangare L, Dao M, Diarra I, Kone A, Harley R, Plowe CV, Doumbo OK, Szein MB. Serum levels of the proinflammatory cytokines interleukin-1 beta (IL-1beta), IL-6, IL-8, IL-10, tumor necrosis factor alpha, and IL-12(p70) in Malian children with severe Plasmodium falciparum malaria and matched uncomplicated malaria or healthy controls. *Infect Immun.* 2004 Oct;72(10):5630-7.
Source: Ingenuity Expert Findings

Binding of HMGB1 protein and TNFA [TNF] protein increases proinflammatory response of cells.

18431461 Klune JR, Dhupar R, Cardinal J, Billiar TR, Tsung A. HMGB1: endogenous danger signaling. *Mol Med.* 2008 Jul-Aug;14(7-8):476-84.
Source: Ingenuity Expert Findings

In nuclei from myeloid cells, **TNF α [TNF]** protein increases **Proinflammatory response**.

16940328 Netea MG, Azam T, Ferwerda G, Girardin SE, Kim SH, Dinarello CA. Triggering receptor expressed on myeloid cells-1 (TREM-1) amplifies the signals induced by the NACHT-LRR (NLR) pattern recognition receptors. *J Leukoc Biol.* 2006 Dec;80(6):1454-61. Epub 2006 Aug 29.
Source: Ingenuity Expert Findings

If desired, click the blue hyperlinks at the left side of each finding to go the NCBI record for the paper.

7) Click the link on the Scroll icon in the pathway tool bar to read the curated report about this pathway.

Zoom:

[View Pathway Report](#)

The pathway report discusses the role of the TREM1 pathway in immune function, and shows that several of the proteins on the pathway are targets in drug trials for asthma:

INGENUITY
PATHWAY ANALYSIS

Canonical Pathway

Report Date: 2016-10-12
Report Version: 401642
Content Version: 28820210 (Release Date: 2016-09-24)

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Canonical Pathway: TREM1 Signaling

Description: The triggering receptor expressed on myeloid cells 1 (TREM1) belongs to the Immunoglobulin (Ig) family of cell surface receptors and is selectively expressed on blood neutrophils, monocytes and macrophages. TREM-1 lacks known signaling motifs in the cytoplasmic domain and thus activation by TREM1 is mediated by a transmembrane adaptor molecule DNAX-activating protein 12 (DAP12), leading to proinflammatory immune responses. The natural ligand for TREM1 is however, unknown.

TREM1 activation triggers the Janus kinase 2 (JAK2), protein kinase B (PKB/AKT) and extracellular signal related kinase (ERK1/2) pathways leading to the phosphorylation of signal transducers of activation of transcription (STAT3, 5) and NF kappa B (NF- κ B). These transcription factors upregulate the expression of genes involved in the inflammatory response. Stimulation of TREM1 by its ligand or toll like receptor (TLR) by lipopolysaccharide (LPS) can lead to an association of TREM1 and TLR. This association leads to the activation of interleukin-1 receptor-associated kinase 1 (IRAK1), which in turn triggers NF- κ B and the proinflammatory response. Engagement and activation of TREM-1 triggers expression and secretion of chemokines and cytokines like monocyte chemoattractant protein 1 (MCP-1) macrophage inflammatory protein-1alpha (MIP-1 α), interleukins (IL-6, -8) and tumor necrosis factor (TNF). Many of these effects are potentiated by LPS. Cytokines like TNF and Granulocyte macrophage colony stimulating factor (GM-CSF) in turn upregulate the expression of TREM1 in an autocrine fashion. The synergy between TLR and TREM1 leads to neutrophil degranulation, phagocytosis and the respiratory burst in addition to the production of proinflammatory cytokines. TREM1 activation also results in the upregulation of cell surface proteins like CD11, CD29 and CD40, CD83 that are involved in cell adhesion and costimulation respectively, as well as phospholipase gamma (PLC γ) mediated Ca²⁺ release. Thus TREM1 activation is involved in diverse aspects of innate and adaptive immune response.

In addition to TLRs, TREM1 also synergizes with a second major class of pattern recognition receptors - the NACHT-LRR receptors (NLR), which recognize intracellular microorganisms. The TREM-1/NLR synergism results in the production of proinflammatory cytokines like TNF, IL-1 β , IL-6 and IL-18 - the latter three via a caspase -1 dependent pathway. Thus TREM-1 acts to amplify signals from both major pathways of pattern recognition- extracellular TLR receptors and the intracellular NLR proteins.

This pathway highlights the important components of TREM1 signaling.

Signaling Pathway Categories: Cellular Immune Response; Cytokine Signaling

Top Functions & Diseases: Cell-To-Cell Signaling and Interaction; Hematological System Development and Function; Immune Cell Trafficking

Molecules: [show all](#) adaptive immune response, Akt, anti-TREM1 Ab, apoptosis, CASP1, Casp1-Casp5, CASP5, CCL2, CCL3, CCL7, CD40, CD83, CD86, cell adhesion, CSF2, CXCL3, CXCL8, DEFB4A/DEFB4B, degranulation, EBOV, ERK1/2, FCGR2B, Flagellin, GRB2, ICAM1, IL10, IL18, IL1B, IL1RL1, IL6, innate immune response, IRAK1, ITGA5, ITGAX, ITGB1, JAK2, L-Ala- γ -D-Glu-meso-diaminopimelic acid, LAT2, lipopolysaccharide, lipoteichoic acid, MARV, MARV GP, mobilization of Ca²⁺, MPO, MYD88, N-acetylmuramyl-L-alanyl-D-isoglutamine, NF κ B (complex), NLR, NOD2, Pam3-Cys, phagocytosis

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Drug Summary - Overview of drugs targeting molecules in Canonical Pathway

Showing 3 of 111 row(s) of Drug data. ([Show All](#))

Drug Name	Targets	Actions	Brand Names	Indications/Status
3M-001	TLR7	agonist		acute lymphocytic leukemia/Phase 2 acute myeloid leukemia/Phase 2 Barrett's syndrome/Unspecified phase
5-fluorouracil/imiquimod [imiquimod]	TLR7	stimulator		
abatacept	CD86	binder	Orencia	allergic asthma/Phase 2 alopecia areata/Phase 2 ankylosing spondylitis/Phase 2

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Target Information - Overview of known drug targets in Canonical Pathway

Showing 3 of 29 row(s) of Target data. ([Show All](#))

Target (Gene Symbol)	Entrez Gene Name	Location	Type	Drug(s)	Species
Akt		Cytoplasm	group	afuresertib, AT13148, ipatasertib, MSC2363318A, ONC-201, SR-13668	Human, Mouse, Rat
AKT1	AKT serine/threonine kinase 1	Cytoplasm	kinase	archexin, ARQ 092, AZD5363, BAY1125976, enzastaurin, GSK2141795, ipatasertib, LY2780301, MK2206, MPT0E028, perifosine, tricinibine, tricinibine phosphate	Human, Mouse, Rat
AKT2	AKT serine/threonine kinase 2	Cytoplasm	kinase	BAY1125976, enzastaurin, tricinibine, tricinibine phosphate	Human, Mouse, Rat

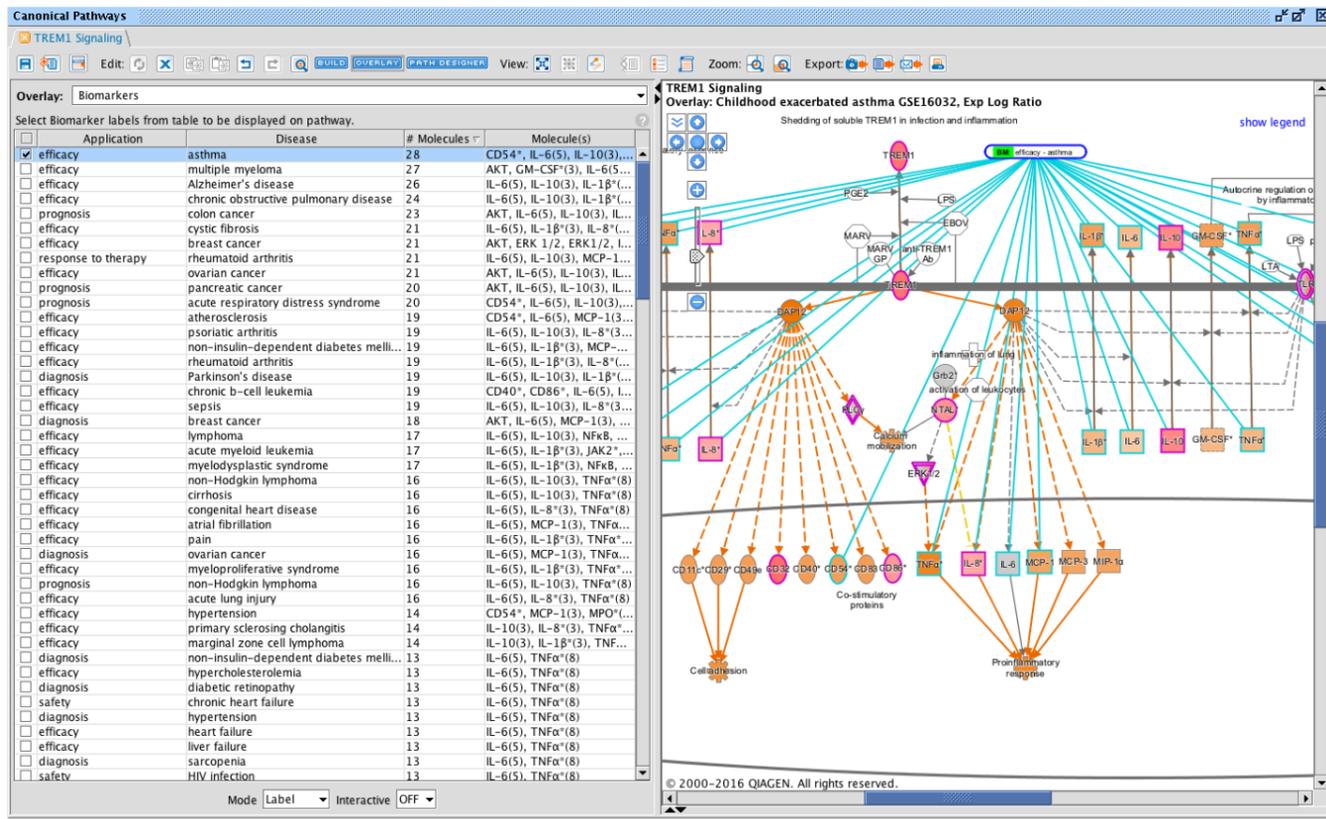
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Supporting References (Show details) - References from which the Canonical Pathway was derived

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8) **Overlay biomarkers** to determine if genes in the TREM1 Signaling pathway are known asthma biomarkers. Go to **Overlay > Biomarkers**. Notice that these known biomarkers are ranked simply based on the number of biomarker molecules found in the given pathway. Clearly the TREM1 pathway is replete with asthma markers, as 28 molecules on the pathway are known biomarkers for asthma efficacy:



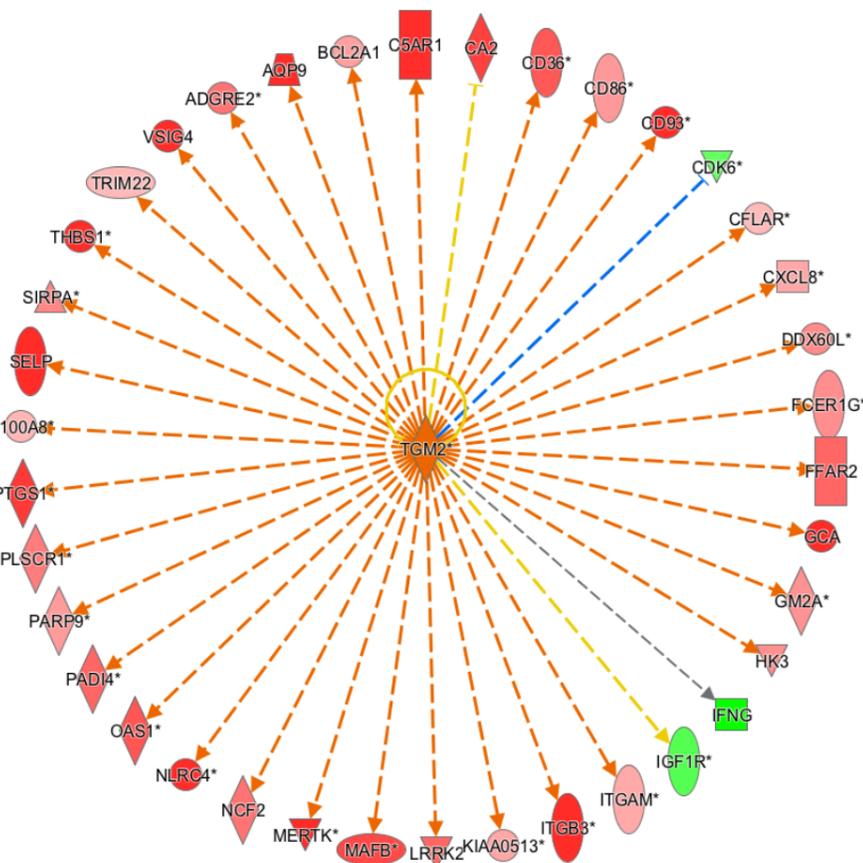
Click the checkbox next to Efficacy - Asthma as shown to add the biomarker tag to the canonical pathway. It will show lines connecting it to the associated genes. Double-clicking a biomarker tag or any of the connecting lines will open the supporting findings.

Upstream Regulators

9) Click on the **Upstream Analysis** tab to see which upstream molecules are predicted to have been activated or inhibited to lead to the expression patterns in your dataset.

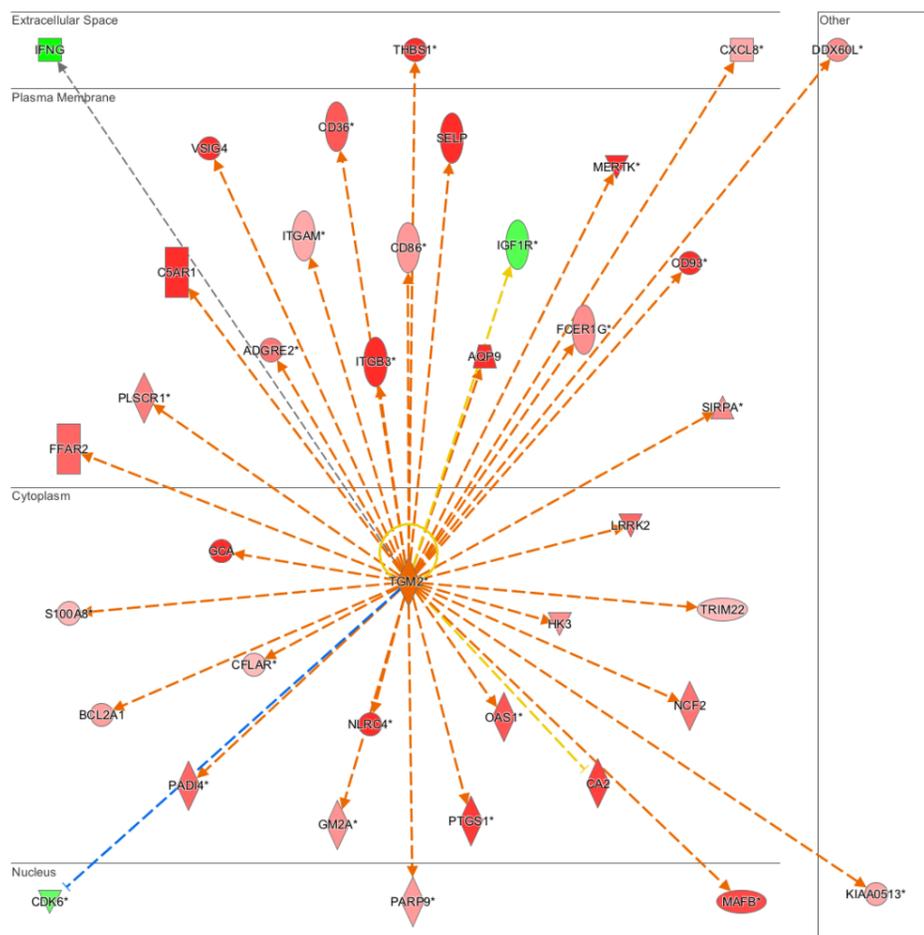
Upstream Regulator	Exp Log Ratio	Molecule Type	Predicted Acti...	Activation z-score	p-value of o...	Target molec...	Mec...
TGM2	+0.251	enzyme	Activated	5.416	8.64E-07	↑ADGRE2, ↑...all 38	369 (17)
IFNG	-1.591	cytokine	Activated	4.714	1.69E-18	↑ABLM3, ↑...all 170	421 (18)
CEBPA	+0.255	transcription regulator	Activated	4.325	1.93E-09	↑ACSL1, ↑...all 64	354 (15)
IRF7	+0.440	transcription regulator	Activated	4.325	3.17E-05	↑CARD16, ↑...all 26	297 (17)
EDN1	-0.225	cytokine	Activated	4.277	1.20E-05	↑ACTA2, ↑...all 32	558 (19)
PRL	-0.036	cytokine	Activated	4.164	2.12E-05	↑ANXA2, ↑...all 39	471 (19)
IL1B	+0.684	cytokine	Activated	4.090	3.19E-09	↑ABCC3, ↑...all 104	392 (14)
Interferon alpha		group	Activated	4.036	5.57E-08	↑AIM2, ↑A...all 51	309 (16)
CSF2	+0.056	cytokine	Activated	4.030	3.62E-12	↑ALOX5, ↑...all 73	403 (16)
MYD88	+0.697	other	Activated	3.995	2.44E-08	↑ACPP, ↑A...all 41	383 (14)
GATA1	+0.419	transcription regulator	Activated	3.991	1.06E-13	↑AHSP, ↑A...all 51	259 (12)
SMARCA4	-0.237	transcription regulator	Activated	3.928	1.56E-05	↑ABHD2, ↑...all 71	525 (20)
TNF	-0.575	cytokine	Activated	3.926	1.02E-17	↑ABCC3, ↑...all 207	469 (16)
IL5	-0.043	cytokine	Activated	3.906	5.69E-06	↑ANXA2, ↑...all 42	454 (19)
TP53	+0.420	transcription regulator	Activated	3.863	2.99E-10	↑ABAT, ↑...all 164	439 (16)
IL4	-0.024	cytokine	Activated	3.829	6.09E-21	↑ABLM1, ↑...all 127	361 (16)
TGFB1	-0.182	growth factor	Activated	3.776	6.26E-20	↑ABLM3, ↑...all 215	510 (18)
CEBPB	+0.417	transcription regulator	Activated	3.775	7.06E-03	↑ACTA2, ↑...all 41	420 (18)
CHUK	-0.027	kinase	Activated	3.769	1.59E-03	↑ACKR3, ↑...all 26	419 (15)
SMARCB1	-0.098	transcription regulator	Activated	3.769	4.10E-03	↑ACSL1, ↑...all 22	302 (7)

Sort by the Activation z-score and filter for upstream regulators that are of the Molecule Type "Genes, RNAs, and Proteins" as shown above. Select the first upstream regulator row (TGM2) and then click the Display as Network button.



This displays the upstream regulator with its targets from the dataset in a circle surrounding it. The expression changes of the molecules in the perimeter are what led to the prediction of the activation of TGM2 as an upstream regulator.

You can change the layout of the network using this button in the toolbar. Choosing Subcellular layout is helpful, because you can see that TGM2 is a cytoplasmic protein, but leads to expression changes in proteins that are secreted out of the cell, are found in the plasma membrane etc.



You can interrogate this network to see what relationship the molecules have to diseases and function. Click the Build button, choose Grow, then click the Diseases & Function tab. This causes IPA to compute the Fisher's exact p-value for the set of molecules on the network against all diseases and functions:

Tool: Grow

Molecules Diseases & Functions

Grow from selected molecules to selected diseases & functions

Indicate diseases or functions related to **Any** of the selected molecules

Consider only categories with names like immun*

Diseases and Functions	p-value	Molecules
systemic autoimmune syndrome	1.36E-18	CXCL8, CA2, IFNG, OAS1, ...all 24
Rheumatic Disease	8.08E-17	CXCL8, CA2, IFNG, OAS1, ...all 23
quantity of leukocytes	2.99E-16	CXCL8, IFNG, CSAR1, CD36, ...all 20
response of neutrophils	4.22E-16	PADI4, IFNG, CXCL8, CSA, ...all 10
response of phagocytes	1.22E-15	IFNG, CXCL8, CSAR1, CD36, ...all 13
inflammation of joint	1.26E-15	CXCL8, CA2, IFNG, CSAR1, ...all 21
response of myeloid cells	1.35E-15	IFNG, CXCL8, CSAR1, CD36, ...all 13
quantity of phagocytes	2.32E-15	IFNG, CXCL8, CSAR1, CD36, ...all 15
chronic inflammatory disorder	2.35E-15	CXCL8, CA2, IFNG, CSAR1, ...all 22
immune response of leukocytes	3.12E-15	IFNG, CXCL8, CSAR1, CD36, ...all 14
differentiation of cells	9.40E-15	CA2, CD36, ITGB3, FFAR2, ...all 26
quantity of mononuclear leukocytes	1.76E-14	CXCL8, IFNG, CSAR1, CD36, ...all 17
function of blood cells	2.43E-14	CXCL8, IFNG, CD36, PTC51, ...all 15
rheumatoid arthritis	1.32E-13	CXCL8, IFNG, CA2, CSAR1, ...all 17
activation of phagocytes	1.45E-13	CXCL8, IFNG, CSAR1, CD36, ...all 14
myocardial infarction	2.10E-13	TGM2, PADI4, CXCL8, IFNG, ...all 13
phagocytosis of leukocytes	3.52E-13	TGM2, PADI4, IFNG, ITGAM, ...all 10
quantity of myeloid cells	3.86E-13	CXCL8, IFNG, CSAR1, CD36, ...all 14
immune response of phagocytes	4.55E-13	TGM2, PADI4, CXCL8, IFNG, ...all 11
quantity of T lymphocytes	6.83E-13	IFNG, CSAR1, CDK6, VSIG4, ...all 14
engulfment of phagocytes	7.27E-13	TGM2, PADI4, IFNG, ITGAM, ...all 10
differentiation of leukocytes	9.95E-13	CXCL8, IFNG, CSAR1, CD36, ...all 15
cellular homeostasis	1.36E-12	IFNG, CA2, CXCL8, OAS1, ...all 22
activation of myeloid cells	1.39E-12	TGM2, CXCL8, IFNG, CSA, ...all 13
immune response of neutrophils	1.46E-12	PADI4, IFNG, CXCL8, ITGAM, ...all 8

Overlay: Childhood exacerbated asthma, GSE16032, Exp Log Ratio

Clearly the set of overlapping diseases and functions has a relationship to immunological and inflammatory disorders. You can add any of the diseases or functions to the network as shown for systemic autoimmune syndrome, which indicates that the disorder is exacerbated by the expression of these genes (the added disease node is orange). You can also use the Add/Remove columns link at the top of the table to add a category column to get a more global picture of the associated functions. You can filter the table well:

Categories	Diseases and Functions	p-value	Molecules
Inflammatory Disease	Categories to include:(use * for wildcard)	2.35E-15	CXCL8, ...all 22
Inflammatory Response	inflam*	1.92E-12	CXCL8, ...all 17
Inflammatory Response	Categories to exclude:	2.64E-12	CXCL8, ...all 15
Inflammatory Response		4.06E-12	CXCL8, ...all 20
Inflammatory Response		3.40E-11	CXCL8, ...all 18
Inflammatory Response,Organismal Injury and		6.53E-11	IFNG, ...all 20
Inflammatory Response,Respiratory Disease	inflammation of respiratory system component	1.21E-08	IFNG, ...all 11
Inflammatory Disease,Inflammatory Response,Or...	Nephritis	4.38E-08	TGM2, ...all 9
Inflammatory Disease,Neurological Disease,Skele...	Multiple Sclerosis	8.07E-08	CXCL8, ...all 9
Inflammatory Response,Organismal Injury and A...	Inflammation of lung	4.36E-07	IFNG, SE, ...all 9
Inflammatory Disease,Neurological Disease,Skele...	relapsed multiple sclerosis	1.56E-06	CXCL8, ...all 6

A number of the other upstream regulators can be found to have direct associations to asthma, such as NR3C1 and IL4. But there are a number of other regulators which have not yet been connected to asthma, and await validation at the bench. See Regulator Effects analysis below as well.

Downstream Effects

10) Click on the **Diseases and Functions** tab to explore the diseases and biological processes which are predicted to be increasing or decreasing based on the pattern of differentially expressed genes in the dataset.

The large labeled boxes in the [treemap](#) shown below are major functional categories in IPA. In this view, it is easy to see the overall directionality of the effects, because orange functions within the categories are predicted to be increasing, and blue decreasing.

From this visualization, many of the individual functions categorized under Immune Cell Trafficking, Inflammatory Response and several others are predicted to be increased, as you would expect if the cells were involved in an immune response involved in acute asthma.

Attachment

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