Staging and classification of biliary atresia: an analysis of disease states

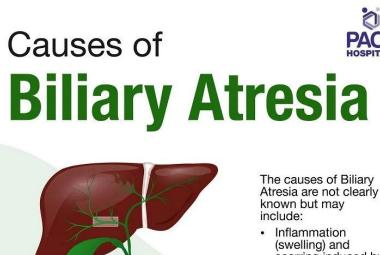
•••

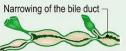
Katie Moyer, Vivek Kaimal, Cristina Pacheco, Reena Mourya, Huan Xu, Pranavkumar Shivakumar, Ranajit Chakraborty, Marepalli Rao, John C Magee, Kevin Bove, Bruce J Aronow, Anil G Jegga & Jorge A Bezerra

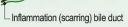
Thomas Sirchi 239007

The disease

- Biliary atresia (BA) affects infants and young children.
- Characterized by destructive inflammatory process in intra- and extrahepatic bile ducts.
- Leads or starts with fibrosis, progressive narrowing, and obliteration of bile ducts.
- About 270 cases are diagnosed each year in Europe







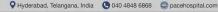


Normal bile duct

Inflammation bile duct

- scarring induced by immune system disorders
- Viral infections
- Hazardous chemical exposure
- · Gene mutations (alterations)



















The Pipeline

The data

Unsupervise d learning

Supervised Learning

Differential expression

Functional enrichment

Data retrieval

Metadata

incorporation

PCA and UMAP

Clustering

Removal of biased

probes

Random Forest

LASSO

Elastic NET

KNN

SCUDO

Feature selection

Heatmap most

important features

pathfindeR

David

Gprofiler

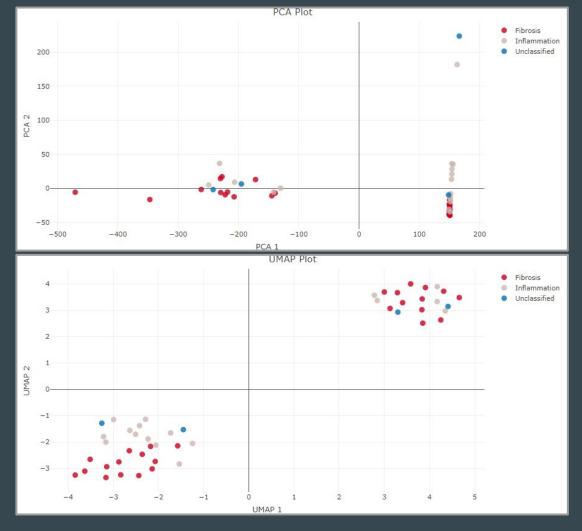
STRING

Literature

annotation

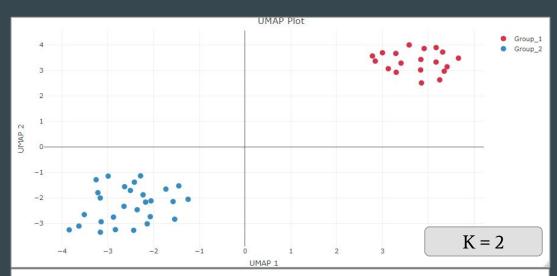
Unsupervised learning: PCA & UMAP

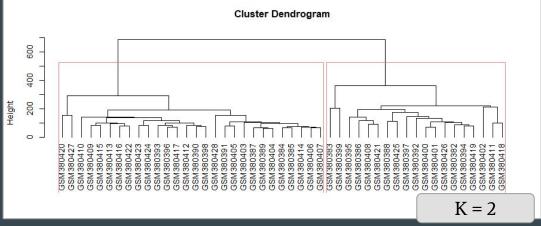
Clear division, not in the metadata



Clustering

Clear division, not in the metadata





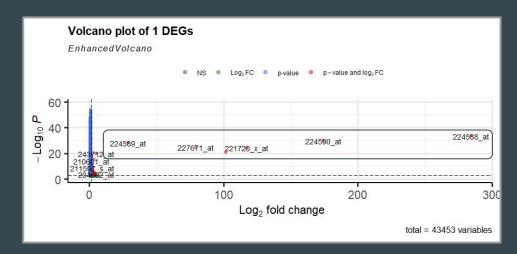
X-inactive specific transcript (XIST)

The probes shown on the volcano plot code for the XIST gene.

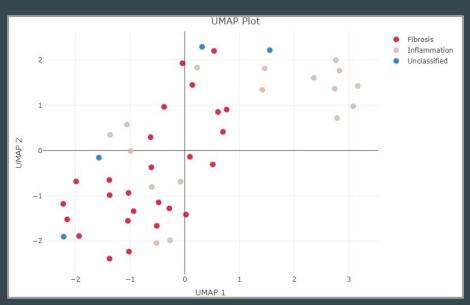
- Plays a crucial role in the process of X-chromosome inactivation in female mammals.
- This gene is located on the X chromosome and produces a long non-coding RNA (lncRNA) molecule

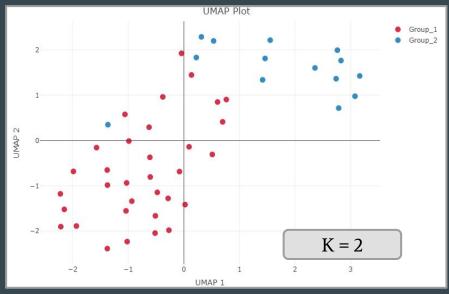
Removal of those probes assures to mitigate

Male - Female differences

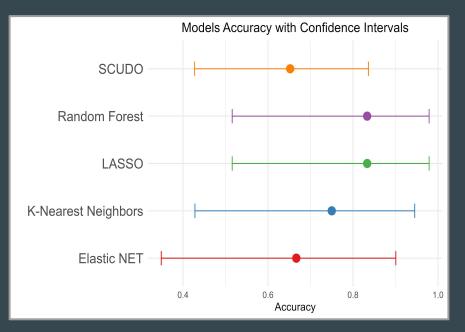


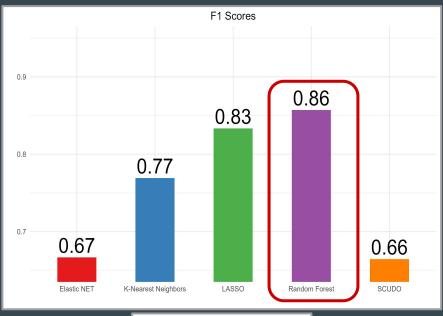
New UMAP and PAM clustering





Supervised Learning: Models performance and predictions

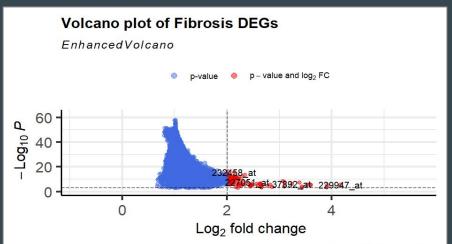


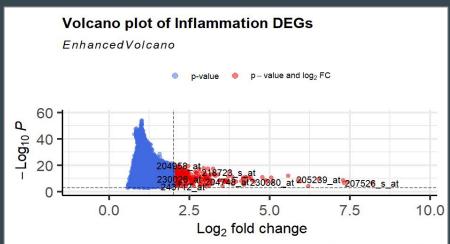


 $F1 = \frac{2 \times Precision \times Recall}{Precision + Recall}$

Differential expression analysis

LIMMA with a p-value threshold of 0.001 and a logFC threshold of 2.0



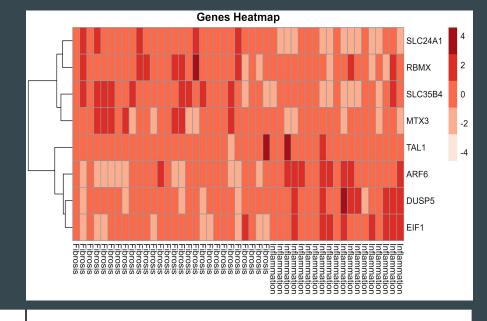


206 genes have been identified as differentially expressed in inflammation

74 genes have been identified as differentially expressed in fibrosis

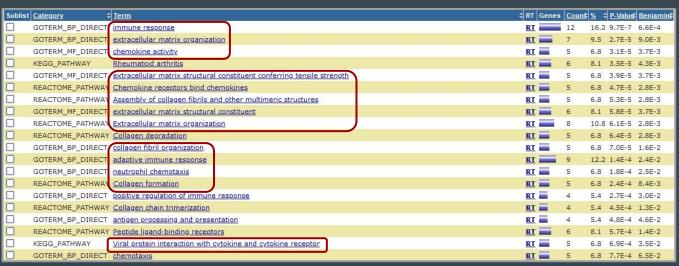
DEGs and Importance

- DEGs and Random forest importance
- Pheatmap library
- Two groups, diving almost as molecular groups



SLC35B4/24A1 (Solute Carrier Family)	Targets for hepatobiliary transformation in animal models
EIF1 (Eukaryotic Translation Initiation Factor 1)	Involved in regulation of translational initiation
ARF6 (ADP Ribosylation Factor 6)	Susceptibility locus at chromosome 14q21.3
DUSP5 (Dual Specificity Phosphatase 5)	Marker to monitor perinatal exposure to environmental toxin

David Fibrosis



	I DOMINATE OF THE PARTY OF THE		ari Vissel sizare				
Sublist	Category \$	<u>Term</u> :	⇒ RT Genes	Count	<u>%</u>	-Value	Benjamint
	GOTERM_BP_DIRECT	inflammatory response	RT ====	31	15.0 1	.8E-17	3.2E-14
	GOTERM_BP_DIRECT	neutrophil chemotaxis	RT ==	15	7.3 7	.7E-14	6.6E-11
	REACTOME_PATHWAY	Interleukin-10 signaling	RT =	12	5.8 3	.9E-11	2.5E-8
	GOTERM_BP_DIRECT	negative regulation of apoptotic process	RT ==	25	12.1 2	.3E-10	1.3E-7
	GOTERM_MF_DIRECT	protein binding	RT	169	82.0 4	.0E-9	1.7E-6
	KEGG_PATHWAY	IL-17 signaling pathway	<u>RT</u> =	13	6.3 1	.0E-8	1.6E-6
	GOTERM_BP_DIRECT	cellular response to tumor necrosis factor	<u>RT</u>	13	6.3 1	.1E-8	4.6E-6
	KEGG_PATHWAY	TNF signaling pathway	<u>RT</u> =	14	6.8 1	.4E-8	1.6E-6
	GOTERM_BP_DIRECT	positive regulation of apoptotic process	RT ==	18	8.7 2	.3E-8	7.9E-6
	GOTERM_BP_DIRECT	positive regulation of cell migration	<u>RT</u> =	17	8.3 2	.8E-8	8.0E-6
	GOTERM_BP_DIRECT	chemotaxis	<u>RT</u>	12	5.8 3	3.3E-8	8.0E-6
	GOTERM_BP_DIRECT	cellular response to interleukin-1	RT =	10	4.9 9	.5E-8	2.0E-5
	REACTOME_PATHWAY	Immune System	RT	58	28.2 1	.2E-7	3.6E-5
	GOTERM_BP_DIRECT	immune response	<u>rt</u> =	21	10.2 1	.5E-7	2.7E-5
	GOTERM_BP_DIRECT	positive regulation of cell population proliferation	RT ===	21	10.2 1	.6E-7	2.7E-5
	REACTOME_PATHWAY	Cytokine Signaling in Immune system	RT ===	32	15.5 1	.7E-7	3.6E-5
	REACTOME_PATHWAY	Neutrophil degranulation	RT ===	24	11.7 2	.9E-7	4.7E-5
	GOTERM_BP_DIRECT	negative regulation of cell population proliferation	<u>RT</u> =	19	9.2 3	.3E-7	5.2E-5
	REACTOME_PATHWAY	Interleukin-4 and Interleukin-13 signaling	<u>RT</u> <u></u>	12	5.8 3	.6E-7	4.7E-5
	GOTERM_BP_DIRECT	monocyte chemotaxis	<u>rt</u> \overline	8	3.9 3	3.7E-7	5.3E-5

David Inflammation

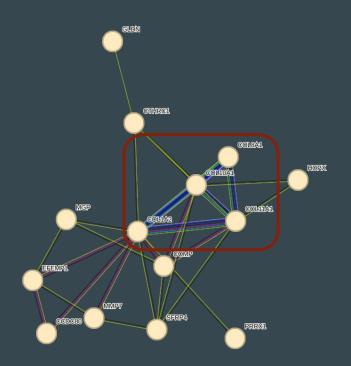
ID	Source	Term ID	Term Name	p _{adj} (query_1)
1	GO:MF	GO:0005201	extracellular matrix structural constituent	4.408×10 ⁻⁵
2	GO:MF	GO:0008009	chemokine activity	3.462×10 ⁻⁴
3	GO:BP	GO:0030198	extracellular matrix organization	1.394×10 ⁻⁴
4	GO:BP	GO:0009888	tissue development	5.385×10 ⁻³
5	GO:BP	GO:0042127	regulation of cell population proliferation	2.950×10 ⁻²
6	GO:BP	GO:0045785	positive regulation of cell adhesion	3.573×10 ⁻²
7	GO:BP	GO:0030593	neutrophil chemotaxis	3.864×10 ⁻²
8	REAC	REAC:R-HSA-2022090	Assembly of collagen fibrils and other multimeric structures	8.207×10 ⁻⁴
9	REAC	REAC:R-HSA-1650814	Collagen biosynthesis and modifying enzymes	3.500×10 ⁻²
10	REAC	REAC:R-HSA-1474290	Collagen formation	5.781×10 ⁻³
11	REAC	REAC:R-HSA-8948216	Collagen chain trimerization	6.578×10 ⁻³
12	GO:MF	GO:0045236	CXCR chemokine receptor binding	1.419×10 ⁻²

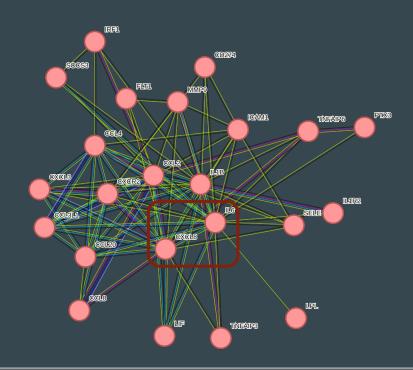
Gprofiler Fibrosis

Gprofiler Inflammation

ID	Source	Term ID	Term Name	p _{adj} (query_1)
1	GO:MF	GO:0140375	immune receptor activity	6.944×10 ⁻³
2	GO:BP	GO:2000351	regulation of endothelial cell apoptotic process	1.511×10 ⁻²
3	GO:CC	GO:0005615	extracellular space	3.411×10 ⁻¹⁰
4	GO:MF	GO:0005125	cytokine activity	9.204×10 ⁻⁷
5	GO:MF	GO:0008009	chemokine activity	2.836×10 ⁻⁴
6	REAC	REAC:R-HSA-6783783	Interleukin-10 signaling	4.532×10 ⁻¹⁰
7	REAC	REAC:R-HSA-6785807	Interleukin-4 and Interleukin-13 signaling	2.469×10 ⁻⁵
8	REAC	REAC:R-HSA-168256	Immune System	8.389×10^{-5}
9	KEGG	KEGG:04657	IL-17 signaling pathway	7.025×10 ⁻⁸
10	KEGG	KEGG:04933	AGE-RAGE signaling pathway in diabetic complica	1.493×10 ⁻³
11	GO:MF	GO:0050786	RAGE receptor binding	9.332×10 ⁻⁴
12	GO:BP	GO:0097529	myeloid leukocyte migration	1.176×10 ⁻¹¹
13	GO:BP	GO:0048661	positive regulation of smooth muscle cell prolifera	1.673×10 ⁻⁴
14	GO:BP	GO:0000165	MAPK cascade	1.087×10 ⁻⁵
15	GO:BP	GO:0002685	regulation of leukocyte migration	9.032×10 ⁻⁸

STRING (confidence = 0.7), networks show significantly more edges

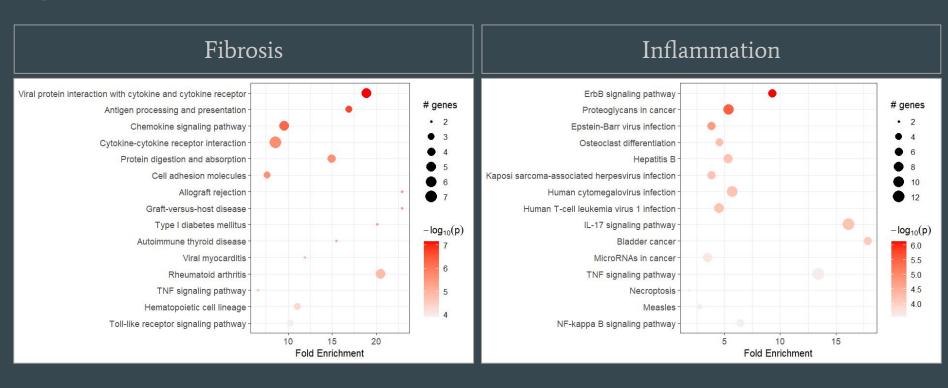




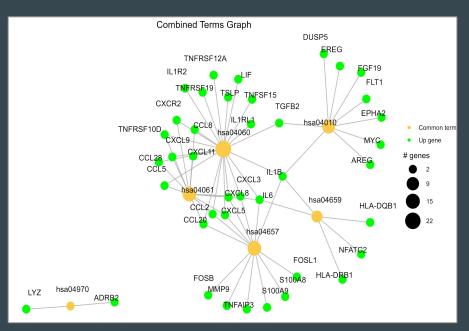
Fibrosis cluster about collagen and fibrillar collagen: COL1A2.

Inflammation cluster about chemokines and interleukins: IL6, CXCL8.

pathfindeR with KEGG



pathfindeR Combined



Pathway ID	Pathway Name
hsa04060	Cytokine-cytokine receptor interaction
hsa04061	Viral protein interaction with cytokine and cytokine receptor
hsa04657	IL-17 signaling pathway
hsa04659	Th17 cell differentiation
hsa04010	MAPK signaling pathway
hsa04970	Salivary secretion

Conclusions

Machine Learning Integration and Classification:

- Enhanced histological classification with machine learning models.
- Achieved good performance with the Random Forest model.
- Successfully classified four samples that were previously unclassified.

Molecular Insights and Feature Importance:

- Identified key genes (SLC24A1, ARF6, DUSP5) of high importance.
- Confirmed relevance of known genes and potential for new targets.
- DEG filtering revealed patterns linked to disease stages.
- Machine learning provided insights into potential therapeutic targets.

Conclusions

Functional Enrichment and Pathway Analysis:

- DAVID and gProfiler identified terms related to collagen organization and inflammation.
- STRING analysis revealed distinct gene clusters per molecular group.
- pathfindR highlighted significant terms related to viral infections and inflammation.
- Identified key pathways such as MAPK (DUSP5), ARF6, and salivary secretion (ADRB2).

Future Directions and Implications:

- Molecular insights are valuable for understanding BA.
- Future research should involve more complex models, additional data, and novel platforms.
- Exploring inflammatory states and specific gene interactions may reveal more about BA mechanisms and inform treatment strategies.

Thank you