Research Article



Comparative Transcriptomic Analysis Between SARS-COV-2, RSV And Influenza H3N2 Patients Highlights the Use Of IL-6 Inhibitors

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Abstract

Comparative transcriptomic analysis between different respiratory viral infections could help identifying the fundamental targets to improve the administration of these diseases. Especially in the case of COVID-19, unprecedented epidemy has challenged the health system with limited treatment options. Recently, based on studies on thousands of patients, WHO has recommended the use of IL-6 inhibitors in COVID-19. Here, we subjected the patient transcriptome of SARS COV2, RSV, and Influenza A to a comparative analysis. Two gene expression profiles were obtained from the NCBI GEO database. The differential analysis was performed by the R's LIMMA package and Galaxy limma voom tool. To compare gene expression, Ggplot2 and pheatmap packages of R were applied to generate volcano plots and heatmap. Afterward, the KEGG pathway and GO enrichment were analyzed by DAVID online database. We used the STRING database for PPI network construction and hub gene identification. Using the DGIdb, potential drugs were predicted. We found a group of 11 genes, including IL6, IRF7, MX1, IFIH1, OAS1-3, XAF1, IRF9, IFIT1, and IFIT3, ... and... with potentially the highest score in terms of gene expression level in all three viruses. IL-6 is particularly of higher importance with three FDA approved drugs, Siltuximab, Olokizumab, and Clazakizumab. Five other genes including, , IRF7, IFIH1, OAS1, IFIT3, and OAS3, also had potential therapeutic targets for SARS-CoV-2 and IAV H3N2 and RSV.

Keywords: SARS-CoV-2, IAV, RSV, IL6

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Abbreviations:

Abbreviation	Full name
HT RNA-Seq	high throughput RNA sequencing
DEGs	differentially expressed genes
KEGG	Kyoto Encyclopedia of Genes and Genomes
GO	Gene Ontology
РРІ	protein-protein interaction
DGIdb	Drug-Gene Interaction Database
IAVs	influenza A viruses
CoVs	coronaviruses
SARS-CoV	severe acute respiratory syndrome coronavirus
MERS-CoV	Middle East respiratory syndrome coronavirus
RSV	respiratory syncytial virus
ISGs	IFN-stimulated genes
TNF-a	tumor necrosis factor
NK	natural killer
GSE	genomic spatial event database
adj P-value	Adjusted P-values
FDA	Food and Drug Administration
GEO	gene expression omnibus
GDI	Gene-Drug Interaction
STRING	search tool for the retrieval of interacting genes/proteins
IRF7	interferon regulatory factor 7
PRRs	pattern-recognition receptors
IFNs	interferons
JAK	janus kinase
ТҮК	tyrosine kinase
STAT	signal transducer and activator of transcription
PAMPs	pathogen-associated molecular patterns
TLR	Toll-like receptor
RIG-I	retinoic acid-inducible gene I
OAS	2'-5'-oligoadenylate synthetases
IFIT	immunity is the induction of Interferon-induced protein
	with tetratricopeptide repeats
IFIH1	Interferon-induced with helicase C domain 1
MAVS	with mitochondrial antiviral signaling protein
IFN	interferons
IL-6R	IL-6 receptor
gp130	glycoprotein 130

Introduction

Viral respiratory infections have always been affected human's life and being a challenge for public health. Only in around 100 years, we have encountered many epidemics of respiratory viruses such as different types ofInfluenza A (IAV), including; H1N1 in 1918, 2009 and 1977, H2N2 in 1957 and H3N2 in 1968. Another such epidemies are originated from coronal viruses, including SARS-CoV in 2002, MERS-CoV in 2012, and finally SARS-CoV-2 in 2019 [1-3]. Moreover, Respiratory Syncytial Virus (RSV) is another important virus, especially in young children aged under two years, elderly, and immunodeficient. RSV causes infections in the lower respiratory tract, which leads to bronchitis and pneumonia, although it can infect healthy adults asymptomatically [4].

After cell entry of respiratory viruses, secretion of IFN β by the infected cells results in paracrine type I IFN signaling through the IFN α/β receptor, which induces hundreds of interferon-stimulated genes (ISGs) and inflammatory responses that have a crucial role for infection clearance. Still, a disturbance in immunity responses can be turned into a threat to the body. Cytokine storm induction is a common feature among these three viruses. These viruses can trigger the overexpression of inflammatory cytokines interleukin IL-1 β , IL-6, IL-8, IL-12, Tumor Necrosis Factor (TNF- α), etc., from overactivated inflammatory cells such as neutrophils and macrophages, leading to the release of active mediators. Furthermore, effector cells such as T lymphocytes and NK cells are increased by the secreted cytokines to participate in tissue damage [5, 6].

Recent meta-analysis based on thousands of COVID-19 patients show the significance of IL-6 inhibition strategies in the treatment of this disease (ref: doi: 10.1001 / jama.2021.11330).

WHO has accordingly recommended IL-6 inhibitors in the management of OCVID-19 (https://www.who.int/news/ item/06-07-2021-who-recommends-life-saving-interleukin-6receptor-blockers-for-covid-19-and-urges-producers-to-join-efforts-to-rapidly-increase-access) In the current study we found interesting link between the transcriptome of SARS-CoV-2, RSV, and IAV H3N2. Especially, in the case of Covid-19 pandemic, it is of particular importance to study this disease in the context of other respiratory infections and learn from the experience of the decades of combat with these diseases and find proper possibilities in our currently approved antiviral arsenal. The comparative differentially expressed genes (DEGs) showed overlapping hub genes among three viruses for protein-protein interaction, and gene-drug interaction for available chemotherapeutic drugs to suggest potential drugs affecting these three viruses.

Methods

Data collection

Data analysis procedures are showed in Figure 1. Two gene expression profiles were obtained from the NCBI GEO database (https://www.ncbi.nlm.nih.gov/geo/): GSE32138 (GSM796520 to GSM796535) and GSE147507 (GSM4432378 to GSM4432383). The data for GSE147507 was generated by high throughput sequencing, Illumina NextSeq 500 method, while GSE32138 dataset was from microarray method using Agilent G2505C Microarray Scanner. GSE32138 contains triplicate primary human lung epithelium for SARS-CoV-2, and GSE147507 includes human airway epithelial cells for RSV and IAV H3N2 with a total of 3/3, 4/4, and 4/4 mock controls/infected cells culture, respectively, used for the generation of data in the interest datasets for DEGs analysis.

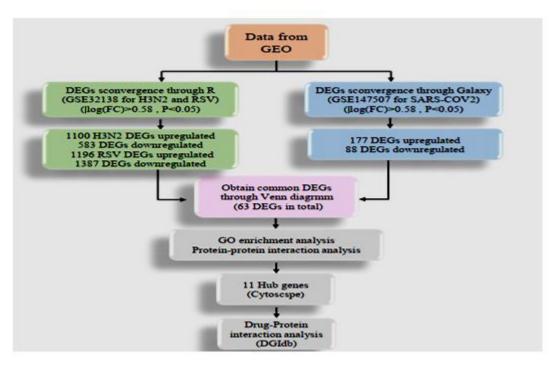


Figure 1: Flowchart of data preparation, processing, and analysis in this study

Sample Sources of the SARS-CoV-2, RSV, and IAV H3N2 Datasets

Generally, the primary target cells of the three viruses were respiratory cells. Triplicate primary human lung epithelium for SARS-CoV-2 and quadruplicate human airway epithelial cells for RSV or IAV H3N2 with 3/3, 4/4, and 4/4 mock controls/ infected cells culture, respectively, were used for the generation of data in the datasets of interest. We selected all sixteen samples, <u>GSM796520</u> to <u>GSM796535</u>, from the GSE32138 dataset and only six samples, <u>GSM4432378</u> to <u>GSM4432383</u>, from the GSE147507 dataset for DEGs analysis.

DEGs identification in the Interest Datasets

In selected samples, we applied two tools for DEGs identification in both datasets; the online Galaxy tool [7] for GSE147507 and R version 1.2.5019 software for GSE32138. The \log_2 fold change (logFC) ≥ 0.58 or ≤ -0.58 were considered for significantly changed DEGs.

Identification of DEGs

Firstly, for microarray data (GSE32138), downloaded the series Matrix file of GSE32138 from the GEO database. Data were then normalized using the normalized quantiles function from R package 'LIMMA' from the Bioconductor project. Afterward, the LIMMA package managed the differential analysis by entering related codes into R (version 3.4.4). Galaxy limma voom tool (https://usegalaxy.org/) used for RNA-seq data (GSE147507) to Identification of DEGs. Adj P-value was supposed to avoid the occurrence of false-positive results. Genes with log FC \geq 0.58 or \leq -0.58 and adj P-value < 0.05 were taken as differentially expressed genes between infected cells and mocks. Ggplot2 and pheatmap packages of R were applied to generate volcano plot and heatmap, respectively, for the visualization of the identified DEGs.

Gene Ontology and GO enrichment analysis

The Database for Annotation, Visualization, and Integrated Discovery (DAVID 6.8; available online: http://david.ncifcrf.gov) is an online database that facilitates users' performance biological analysis from data collection. KEGG pathway and GO enrichment used for analyses were conducted with DAVID. P < .01 to considered statistically significant.

PPI network construction and hub gene (Protein-protein interaction)

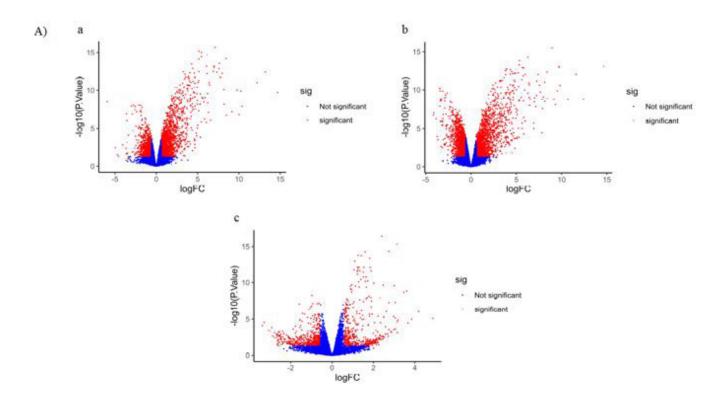
PPI networks of common-DEGs were analyzed by the search tool for retrieving interacting genes (STRING database, V10.5; http://string-db.org/) that predicted protein functional associations and protein-protein interactions. Subsequently, Cytoscape software (V3.5.1; http://cytoscape.org/) was applied to visualize and analyze biological networks and node degrees after downloading analytic results of the STRING database with a confidence score > 0.4.

Drug-hub gene interaction

Based on the hub genes, drugs were selected that served as promising targets using the Drug-Gene Interaction Database (DGIdb; http://www.dgidb.org/search_interactions; version 3.0.2 – sha1 ec916b2). In the present study, the final drug list included just drugs that the FDA permitted. The online tool named Cytoscape was used to construct the interaction network between the potential drugs and the hub genes.

Results

We identified 1682 DEGs for IAV H3N2, 2581 for RSV, and 265 for SARS-CoV-2 with p-value < 0.05. Among 1682 and 2581 DEGs for IAV H3N2 and RSV, 1100 and 1196 genes were upregulated, while 583 and 1387 genes were downregulated. Similar pattern was observed for SARS-CoV-2, in which, out of 265 DEGs, 177 genes were upregulated and 88 genes were downregulated. For identification and visualization of DEGs with statistical significance, drew a volcano plot for each virus (Figure 2 A). Also, Figure 2 B shows the heatmap of 50 top DEGs in each understudied virus. According to our analysis represented in a Venn diagram (Figure 2 C), 63 DEGs shared between these three viruses that 62 genes were overexpressed and only one gene decreased. We listed these common DEGs in more detail in Table 1. The sixty-three/63 DEGs enriched in biological pathways are related to the immune system, response to stimulus, multi-organism process, signaling pathways, biological regulation, locomotion, single-organism process, and reproduction (Figure 3). The common property of IAV H3N2, RSV, and SARS-CoV-2 is the cytokine storm phenomena that trigger inflammation and pathological damages, especially in the respiratory tract through overexpression of inflammatory, antiviral genes, and immune cell proliferation [6]. We found involved interferon signaling genes as prominent among 63 DEGs.



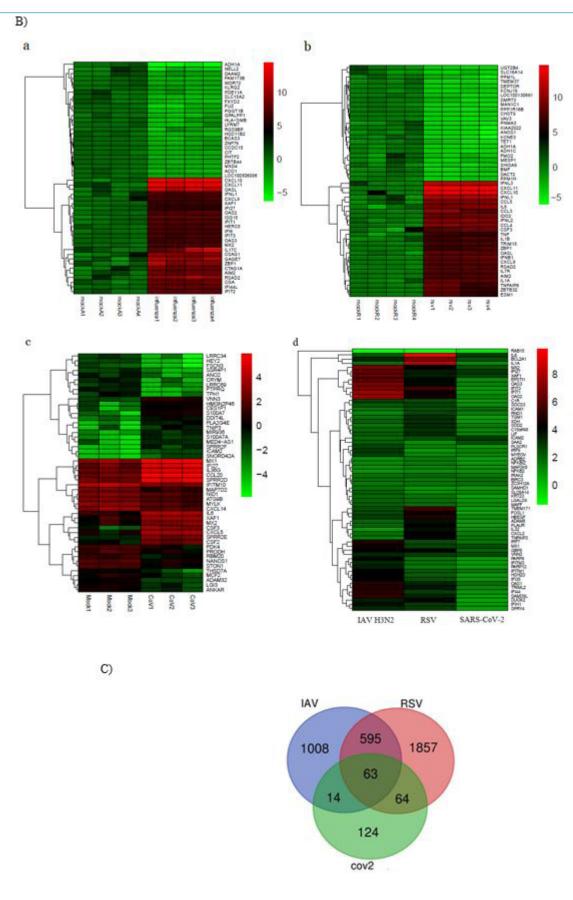


Figure 2: A) The volcano plots presenting DEGs for IAV H3N2 (a), RSV (b), and SARS-CoV-2 (c). The y-axis demonstrates the \log_{10} values of the p-values for the significance of each examined DEG, while the x-axis shows the \log_{10} values of fold changes for each DEG. Red plots depict DEGs with logFCs ≥ 0.58 or ≤ -0.58 , and blue plots displays remained DEGs with no statistical significance. B) Venn diagram of overlapping DEGs for three viruses. C) The heatmap of 50 top DEGs (25 increased and 25 decreased) in IVA H3N2 (a), RSV (b), SARS-CoV-2 (c), and 63 common DEGs (d). Red and green colors reflected upregulation and downregulation, respectively

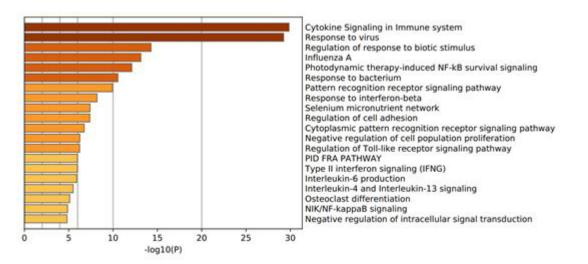


Figure 3: Gene Ontology (GO) enrichment analysis of the common differentially expressed genes

PPI Network and GDI

Protein-protein interactions (PPIs) was investigated as one of the possible ways for the altered transcriptomes to cause the disease. To predict the protein interactions which were derived from DEGs, we submitted all 63 DEGs to the STRING database [8], and the interaction between proteins was constructed with 63 nodes corresponding to the 63 common DEGs between three viruses, 316 edges, average node degree 10, and PPI enrichment p-value <1.0e⁻16 (Figure 3). We also analyzed the shared DEGs by using the Cytoscape tool to visualize further gene interactions [9], in which the genes ranked based on score. The IL-6 and IRF7 genes were identified to contain the highest score among increased DEGs. The last column of Table 1 shows the score of 63 common DEGs between three viruses using Cytoscape analysis, in which 11 DEGs were identified as hub genes with high scores.

We also investigated the interaction of hub genes with available drugs and potential drug ability to construct the genedrug interaction network using DGIdb. <u>IL6</u>, <u>IRF7</u>, <u>IFIH1</u>, <u>OAS1</u>, <u>IFIT3</u>, <u>OAS3</u> were identified as potential targets for RSV, H3N2, and SARS COV2 infection treatment. Among potential drugs, Siltuximab, Olokizumab, Clazakizumab drugs had a therapeutic effect on soluble IL6 through direct interaction with IL-6 protein and abrogating several genes involved IL-6 signaling DGIdb found (Figure 3).

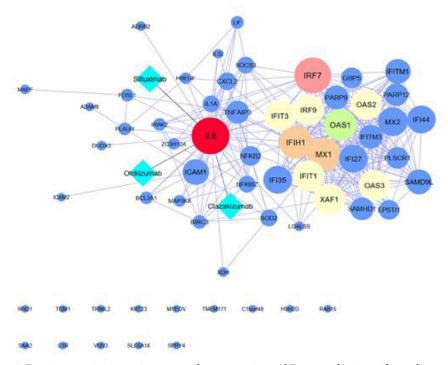


Figure 3: Protein-protein interactions network construction and Drug prediction with p-value <1.0e⁻16

Gene ID	Gene name	logFC	P. value	Adj. p. value	Degree score *
Upregulate	d DEGs				score
IL6	Interleukin 6	3.008258075	2.73E-10	1.99E-07	29
IRF7	Interferon Regulatory Factor 7	1.528321986	1.01E-05	0.001419983	28
MX1	Interferon-induced GTP-binding	2.622073653	8.97E-07	0.000222124	25
171211	protein Mx1	2.022073033	0.9711 07	0.000222121	23
IFIH1	Interferon Induced with Helicase C Domain 1	0.663239405	4.70E-06	0.000793441	25
OAS1	2'-5'-Oligoadenylate Synthetase 1	1.724701628	6.30E-11	5.72E-08	24
OAS2	2'-5'-Oligoadenylate Synthetase 2	1.283179817	8.10E-11	6.79E-08	23
OAS3	2'-5'-Oligoadenylate Synthetase 3	1.332988392	2.19E-08	1.01E-05	23
XAF1	XIAP associated factor 1	2.743618629	5.41E-08	2.19E-05	23
IRF9	Interferon Regulatory Factor 9	1.244703625	1.71E-12	2.48E-09	23
IFIT1	Interferon Induced Protein with	0.913347803	0.000601272	0.032608456	23
	Tetratricopeptide Repeats 1				
IFIT3	Interferon Induced Protein with Tetratricopeptide Repeats 3	0.783392048	0.000315828	0.021512845	23
MX2	MX Dynamin Like GTPase 2	2.49141001	5.36E-06	0.000867141	21
IFI35	Interferon Induced Protein 35	0.986827626	0.0005195	0.030035931	20
IFI44	Interferon Induced Protein 44	0.881349811	9.84E-08	3.40E-05	20
IFI27	Interferon Alpha Inducible Protein 27	3.046637799	7.07E-06	0.001077669	19
ICAM1	Intercellular Adhesion Molecule 1	1.879758892	1.12E-08	5.66E-06	18
IFITM1	Interferon Induced Transmembrane	1.655143451	1.64E-08	7.95E-06	17
11 1 1 1 1 1 1 1	Protein 1				
SAMD9L	Sterile Alpha Motif Domain Contain- ing 9 Like	0.634035551	0.00092164	0.04301708	17
IFITM3	Interferon Induced Transmembrane Protein 3	0.775704598	0.000172631	0.013532281	16
PARP9	Poly (ADP-Ribose) Polymerase Family	0.667454176	2.34E-07	7.27E-05	16
	Member 9				
TNFAIP3	TNF Alpha Induced Protein 3	1.602353044	5.29E-15	2.88E-11	15
SAMHD1	SAM And HD Domain Containing	0.802599368	1.09E-05	0.001504722	15
	Deoxynucleoside Triphosphate Tri- phosphohydrolase 1				
PARP12	Poly (ADP-Ribose) Polymerase Family Member 12	0.777012877	2.75E-07	8.10E-05	15
PLSCR1	Phospholipid Scramblase 1	0.862444838	1.01E-07	3.44E-05	14
GBP5	Guanylate Binding Protein 5	2.051097583	3.12E-05	0.003542176	13
NFKB2	Nuclear Factor Kappa B Subunit 2	0.955892636	2.00E-10	1.56E-07	12
EPSTI1	Epithelial Stromal Interaction 1	1.098081891	1.83E-05	0.002280005	12
CXCL2	C-X-C Motif Chemokine Ligand 2	1.421837035	7.19E-07	0.000184406	10
SOD2	Superoxide Dismutase 2	1.513765318	1.80E-11	1.84E-08	9
IL1A	Interleukin 1 Alpha	1.077593072	9.69E-06	0.001371078	9
SOCS3	Suppressor of Cytokine Signaling 3	1.027635624	2.37E-07	7.27E-05	9

Table 1: 63 shared DEGs between IAV H3N2, RSV, and SARS-CoV-2

NFKBIZ	NFKB Inhibitor Zeta	0.900995776	7.24E-10	4.51E-07	9
BIRC3	Baculoviral IAP Repeat Containing 3	1.638567812	1.85E-11	1.84E-08	6
IRAK2	Interleukin 1 Receptor Associated	1.607708143	7.31E-11	6.37E-08	5
	Kinase 2				
LIF	leukemia inhibitory factor	1.296612746	7.98E-13	1.34E-09	5
PLAUR	Plasminogen Activator, Urokinase	0.691228374	1.13E-06	0.000271087	5
	Receptor				
FOSL1	Fos Proto-Oncogene Like 1	0.59129513	7.53E-07	0.000190836	5
BCL2A1	Bcl-2-related protein A1	2.318598715	1.31E-08	6.49E-06	5
ZC3H12A	Zinc Finger CCCH-Type Containing	1.659683733	4.38E-10	2.81E-07	4
	12A				
HBEGF	Heparin Binding EGF Like Growth	1.290525563	7.48E-13	1.34E-09	4
	Factor				
MAP3K8	Mitogen-Activated Protein Kinase	1.027745148	1.61E-05	0.002072595	4
	Kinase Kinase 8				
IL32	Interleukin 32	1.227557086	6.49E-12	7.07E-09	2
LGALS9	Galectin 9	0.850730372	0.000686534	0.035629464	2
MAFF	MAF BZIP Transcription Factor F	1.129511986	3.07E-10	2.09E-07	2
XDH	Xanthine Dehydrogenase	0.730795237	9.11E-08	3.26E-05	2
ADRB2	Adrenoceptor Beta 2	0.682671108	3.30E-07	9.47E-05	2
DUOX2	Dual Oxidase 2	0.58104455	0.000532824	0.030643714	2
ICAM2	Intercellular Adhesion Molecule 2	2.990121945	0.000270681	0.019344408	1
ADAM8	ADAM Metallopeptidase Domain 8	1.008094337	4.97E-06	0.000820449	1
VNN3	Vanin 3	2.499860771	5.49E-06	0.00087919	0
SAA2	Serum Amyloid A2	2.405120128	3.90E-17	8.49E-13	0
RND1	Rho Family GTPase 1	1.474356937	3.09E-06	0.000590577	0
TRIML2	Tripartite Motif Family Like 2	1.448435304	1.18E-05	0.001602523	0
MYEOV	Myeloma Overexpressed	1.326629482	1.85E-06	0.000415158	0
C15orf48	Chromosome 15 Open Reading Frame	1.231469907	7.89E-13	1.34E-09	0
	48				
SLC6A14	Solute Carrier Family 6 Member 14	1.212804879	2.66E-12	3.22E-09	0
C1R	Complement C1r	0.838868792	0.000613307	0.032926728	0
TGM1	Transglutaminase 1	0.787494457	3.33E-08	1.45E-05	0
HSH2D	Hematopoietic SH2 Domain Contain-	0.77207627	0.001150116	0.049128951	0
-	ing				
TMEM171	Transmembrane Protein 171	0.754251093	0.000381358	0.024593072	0
SPRY4	Sprouty RTK Signaling Antagonist 4	0.672842446	0.000586277	0.032108259	0
KRT23	Keratin 23	0.598854314	2.33E-06	0.000475135	0
Downregula		1	I		
RAB15	RAB15, Member RAS Oncogene	-0.75807553	7.16E-06	0.00108439	0
141015	Family	0.7007000	,	0.00100100	

*11 top genes presented in bold accounts for the highest degree score through Cytoscape analysis

Discussion

In all three viruses the number of upregulated genes are higher than downregulated genes which is associated with... (is this a known phenomenon in virus infection due to release of cytokines for example? – ref?)

The innate immune system provides the first line of protection against viral infections. Invasion and damage of the intrinsic barrier on the airway epithelium, followed by respiratory viral infections such as SARS-CoV-2, IAV, and RSV, lead to the induction of several classes of PRRs, as the viral sensing pathways. Subsequently PRRs trigger, intracellular signaling cascades, and finally, activation of downstream antiviral responses, including the production of IFNs, ISG, and proinflammatory cytokines [10-12].

Rapid advances in microarray technologies and studies of host transcriptome would give scientists new and more comprehensive insights into the procedure to virus-host interactions that enable proposing potential/new therapeutic targets [13, 14]. In the current study, based on the updates in the GSEs (147507, 32138) dataset, our results show that 11 genes of the 63 common genes, including IL6, IRF7, MX1, IFIH1, OAS1-3, XAF1, IRF9, IFIT1, and IFIT3 had the highest score in term of gene expression level (Table 1). Moreover, according to the analysis listed in Table 1, the study revealed that these genes were upregulated in all three viruses.

IFN-I induces immune responses through JAK1/ TYK 2 activation followed by STAT1, STAT2, and IRF9. Also, IFN-1 is capable of activating (STAT3, 4, 5, and 6). It has been seen that when STAT1/2 and IRF9 transcription factors are activated, it leads to the induction of hundreds of IFN-stimulatory genes (ISGs) [15]. Our analysis showed that IRF9 was elevated during infection with the three viruses. We also detected an enhancement in the (IRF7) gene expression SARS-CoV-2 infection and two other respiratory viruses, RSV and Influenza A.

IRF7 plays a significant role in the expression of IFN-I [16]. Viral PAMPs are detected by PRRs such as (TLR3, TLR7/8) as well as the RIG-I-like receptor families (RIG-I and MDA5). This is followed by a cascade of the signaling pathway that leads to activation of downstream transcription factors, including IRF3, IRF7, and NF- κ B, their translocation into the nucleus, and eventually transcription of IFN-I genes [17]. IFI16 and TLR2-mediated signaling pathways are also involved in activating IRF7 [18].

Upregulation of the IRF7 gene in response to RSV infection has been supported by the study conducted by McDonald et al. [19]. Similarly, Lowen and colleagues have shown that IRF7 was upregulated in response to IAV infection in guinea pigs as a host model of this virus [20].

The IFN type I produced by virus-infected cells in response to viral sensors such as intracellular TLRs, RIG-I, and MDA5 confers antiviral immunity in uninfected cells. For instance, 2'-5'-oligoadenylate synthetases (OAS1, OAS2, OAS3), which are the members of the OAS family, are induced by type I interferons in uninfected cells. They bind to viral dsRNA afterward activates RNase L, which degrades viral and cellular RNA and thereby limits viral replication [21, 22]. Consistent with other studies, our analysis showed increased levels of OAS1-3. For instance, it has been demonstrated that OAS1 was remarkably upregulated in human primary epithelial cells infected by IAV (H3N2) [23]. The overexpression of OAS1 in SARS-CoV2 and RSV infections has also been confirmed by Lieberman [24] and Ampuero investigations [25]. Moreover, several studies showed OAS2 was significantly upregulated in the acute phase of IAV, RSV, and SARS-CoV-2 infection [26-28]. In addition, up-regulation of the OAS3 gene was supported by McDonald's study [19], and also Fjaerli et al. observation that represented the OAS3 gene up-regulation in the infected infants [29].

Another mechanism by which IFN-I confer antiviral IFIT genes in the un-infected cells. IFIT genes are distinguished and well-known ISGs with four family members, recognized in humans: IFIT1, IFIT2, IFIT3, and IFIT5 [30]. IFIT proteins are involved in antiviral activities mainly by different mechanisms, including restricting viral RNA translation, recognizing and sequestering the viral RNA, binding to the viral proteins, and reducing virus replication [31]. In the absence of stimuli, most cell types have silent or have a very low level of IFIT genes expression. In contrast, they are prominently induced by type I and type III interferons as strong and weak inducers, respectively. IFIT1 is also involved in inhibiting mRNA translation initiation through direct interactions with eIF3 and hence suppresses cells and virus's protein synthesis. IFIT2 and IFIT3 are essential for modulate apoptosis and cell proliferation [32, 33]. Some observations close to our finding represented an increased expression of IFIT1 and 3 in virus-infected cells and patients [19, 25, 34-37]. IFIH1, also known as MDA-5, is the cytoplasmic sensor that binds to the ds-RNA and, as mentioned above, as a part of innate immune response to viral infection triggers IFN-I induction. During the detection of double-stranded RNAs by the CTD and helicase domains of IFIH1, then conformational changes formed a filament of the molecules and association MAVS, leading to further signaling events that finally boost transcription of IFN-I and ISGs [38]. Our results show up-regulation at the IFIH1 level in agreement with Loganathan et al. study on SARS-CoV-2 infection [22], Cao et al., IAV (H3N2) infected cells, [39], and Okabayashi et al. in RSV infection [40].

The interferon-inducible MX1 gene codes a GTPase protein as part of the antiviral response induced by type I in the infected cells. Mx GTPases have antiviral activity against a large variety of RNA viruses, especially influenza viruses at the entry stage and before viral replication. Generally, Mx protein appears to recognize viruses by sensing nucleocapsid and subsequently captures viral ingredients in the host cells [41, 42]. Other studies confirm the elevated level of MX1 in our findings. For example, Barral-Arca revealed a significant increase in the MX1 gene expression levels when RSV infected patients were compared with healthy controls [43]. Consistently, the upregulation of the MX1 gene has been shown in the cells infected with various influenza subtypes A virus [44] and SARS-CoV-2 infection [22].

X-linked inhibitor of apoptosis XAF1 is an IFN-induced gene that enhances IFN-mediated apoptosis and robustly affects IFN-induced sensitization of the cell to apoptosis. Studies have shown that XAF1 plays a role in pro-apoptotic responses. XAF1 is widely expressed in normal tissues at low basal levels can be upregulated in response to IFN and other cytokines like TNF-α. Early studies have revealed that XAF1 binding to XIAP inhibited XIAP's anti-apoptotic activity [45, 46]. Our analysis close to Loganathan et al. study evaluating revealed an increase in the expression of XAF1 in SARS-CoV-2 infection [22]. IL-6 is a proinflammatory and pleiotropic cytokine produced by several cell types in response to cell damage, and infections control the expansion of chronic inflammatory diseases [47, 48]. For IL-6 signaling onset, first IL6 binds to the non-signal transducing IL-6R, followed by complex formation through gp130.

After IL-6 receptor complex formation, IL-6 initiates a cascade of signaling events primarily related to the JAK/STAT3 activation pathway [49]. Several studies have indicated a strong relationship between IL-6 levels in the serum and future respiratory failure [50]. Recent studies have shown the detrimental effects of IL-6 in viral clearance, supporting viral persistence and hence chronic infections via several mechanisms [51]. The results of our study showed an upregulation in IL6. The elevated level of this proinflammatory cytokine was in agreement with

various studies on RSV [52], SARS-CoV-2 [53], and influenza [54] viral infections. During viral infections, IFN-I and inflammatory cytokines are produced by different cells to limit viral replication and its spreading. If this process is done well, the virus may be eliminated by the immune system. However, these antiviral processes may not turn off in a sensible time frame, do not occur at all, or function efficiently, in which case we will see dysfunction or tissue damage .[55] One of the most important events during infection with respiratory viruses such as Influenza, SARS-COV-2, and RSV is cytokine storm, which causes damage to the lungs. During infection with these three viruses and the occurrence of cytokine storm phenomenon, various cytokines and chemokines are increased and lead to tissue damage. However, among the genes examined in the present study, IL6 is commonly elevated during infection with these viruses [56-58]. Since in the early stages of viral infection, the role of IFN and ISGs against viral infections and the future outcome seems very important, so we assume that they can be used potentially as diagnostic or prognostic markers or as targets for therapy for viruses such as Influenza A, SARS-COV-2, and RSV. Some of the 11 genes studied in this study, such as IL6, which plays a significant role in inflammation, severe tissue damage, and viral persistence, could be potentially used as targeted by appropriate therapeutic drugs. In the current study, Siltuximab, Olokizumab, and Clazakizumab drugs have been detected as an inhibitor of soluble IL6. Siltuximab is an FDA-approved chimeric human-mouse monoclonal antibody against IL-6 that forms high affinity and steady complexes with human IL-6. This drug prevents the binding of human IL-6 to both soluble and membrane-bound IL-6 receptors that can inhibit the formation of the hexameric signaling complex with gp130 on the cell surface and interdiction activating of the JAK-STAT signaling pathway [59]. Given that IL6 initiates the cytokine storm (58) and is a common gene with high expression level among the RSV, Influenza H3N2 and SARS COV2 preventing its production and signaling pathway could be considered for treatment of the patients infected with these three viruses.

Conclusion:

Our analysis identified 61 common DEGs between influenza A H3N2, RSV, and SARS Cov2.. Among them, eleven common hub genes might be the core genes, including IL6, IRF7, MX1, IFIH1, OAS1-3, XAF1, IRF9, IFIT1, and IFIT3. All of them were upregulated, and the overexpression of these genes was associated with inflammatory responses and cytokine storm that can lead to tissues damage. More importantly, IL6, IRF7, IFIH1, OAS1, IFIT3, OAS3 were identified as potential therapeutic targets. IL-6 with 3 FDA-approved drugs, Siltuximab, Olokizumab, and Clazakizumab could be considered as a high-scored potential target for upcoming COVID-19 clinical trials.

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Authors' contributions

MM, PB and KHB designed the manuscript. MM, PB, PH, MMK, AB, AZ, TL, KHB, and AH wrote the first draft. PB and KHB edited the following versions. All authors read and approved the final manuscript.

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Competing interests

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