Rhinovirus Infection Drives Complex Host Airway Molecular Responses in Children With Cystic Fibrosis

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Early-life viral infections are responsible for pulmonary exacerbations that can contribute to disease progression in young children with cystic fibrosis (CF). The most common respiratory viruses detected in the CF airway are human rhinoviruses (RV), and augmented airway inflammation in CF has been attributed to dysregulated airway epithelial responses although evidence has been conflicting. Here, we exposed airway epithelial cells from children with and without CF to RV in vitro. Using RNA-Seq, we profiled the transcriptomic differences of CF and non-CF airway epithelial cells at baseline and in response to RV. There were only modest differences between CF and non-CF cells at baseline. In response to RV, there were 1,442 and 896 differentially expressed genes in CF and non-CF airway epithelial cells, respectively. The core antiviral responses in CF and non-CF airway epithelial cells were mediated through interferon signaling although type 1 and 3 interferon signaling, when measured, were reduced in CF airway epithelial cells following viral challenge consistent with previous reports. The transcriptional responses in CF airway epithelial cells were more complex than in non-CF airway epithelial cells with diverse over-represented biological pathways, such as cytokine signaling and metabolic and biosynthetic pathways. Network analysis highlighted that the differentially expressed genes of CF airway epithelial cells’ transcriptional responses were highly interconnected and formed a more complex network than observed in non-CF airway epithelial cells. We corroborate observations in fully differentiated air-liquid interface (ALI) cultures, identifying genes involved in IL-1 signaling and mucin glycosylation that are only dysregulated in the
CF airway epithelial response to RV infection. These data provide novel insights into the CF airway epithelial cells’ responses to RV infection and highlight potential pathways that could be targeted to improve antiviral and anti-inflammatory responses in CF.

Keywords: cystic fibrosis, RV, airway epithelial cells, transcriptomic, innate immune response

INTRODUCTION

Lung disease is the major cause of morbidity and mortality in cystic fibrosis (CF) (1). Progressive lung damage is associated with mucus obstruction, neutrophilic inflammation, and chronic airway infection and is already evident in the first years of life (2–6). Intermittent pulmonary exacerbations occur in individuals with CF who experience increased respiratory symptoms and reduction in pulmonary function that are responsive to therapy with antibiotics (7). Moreover, the frequency of exacerbations is a predictor of long-term morbidity and irreversible loss of lung function (8, 9). The triggers for these pulmonary exacerbations are not fully understood although it is recognized that lower respiratory infections caused by viruses are likely to play a significant role (10–14).

The most common virus detected in the airway of adults and children with CF is human rhinovirus (RV) (15–19). The clinical impact of RV includes reduction of lung function/FEV1 (15, 20, 21), hospitalization (22), and increased requirement for intravenous antibiotic treatment (11, 14). Recent longitudinal data suggest that RV infection persists for a longer period in individuals with CF compared to non-CF controls (14), a finding consistent with in vitro observations that suggest a defective innate response of epithelial cells to RV (23, 24). The nature of any intrinsic deficiency still remains unclear although some explanations are now emerging (25).

In this study, we hypothesized that the antiviral responses of primary airway epithelial cells (AEC) from children with CF are dysregulated following RV infection. We utilized transcriptome sequencing (RNA-Seq) to assess the gene expression of CF (ΔPhe508del homozygous) and non-CF primary AEC pre- and post-RV infection. Differential expression analysis was carried out to compare the antiviral responses between CF and non-CF AEC. Functional analyses identified diverse biological pathways and complex networks in response to RV infection in CF AEC that were less apparent in non-CF AEC. We performed additional work to validate some of these unique biological pathways using primary differentiated AEC culture models, and data corroborates observations made from the RNA-Seq analysis. Overall, this study provides insights into the global transcriptomic response by non-CF and CF AEC to RV infection and data corroborates observations made from the RNA-Seq analysis. Here, primary AECs were differentiated into ciliated pseudostratified AECs as described previously (29). Briefly, AECs were initially seeded on 0.4-µm polyester membrane culture inserts grown to confluence (Corning, NY, USA) and ALI cultures established. These were maintained for 28 days, and both beating cilia and mucus production were well-established. Prior to ALI validation experiments, inserts were confirmed to have a transepithelial electrical resistance (TEER) measurement >800 Ω/cm².

Human RV Infection and RNA Extraction

To emulate an acute RV infection episode in vitro, we exposed AEC with RV1b (courtesy of P. Wark, University of Newcastle) at MOI 12.5 (23, 30, 31). After 24 h, culture supernatant was collected for cytokine measurement and cell pellets for RNA extraction. RNA was extracted using a PureLink® RNA (Life Technologies) mini kit as per manufacturer instructions. Total RNA was eluted with 30 µL RNase free water with the addition

| TABLE 1 | Patient demographic for subjects used for RNA sequencing analysis including five non-CF children and seven children with CF. |
|------------------|------------------|------------------|
| **Non-CF control** | **Cystic fibrosis** |
| Number of participants | 5 | 7 |
| Mean Age ± sd (yr) | 3.5 ± 1.4 | 2.8 ± 2.3 |
| Age range (yrs) | (1.7–5.4) | (0.2–5.6) |
| Male (%) | 40 | 57 |
| Genotype | Healthy non-CF | p. Phe508del/ p. Phe508del |
| NE Activity (%) | NA | 43 |
| IL-8 Detected in BALs (%) | NA | 100 |
| Microorganisms detected in BAL (%) | NA | 14 (Pseudomonas aeruginosa) |
| PRAGMA Disease (%) | NA | 3.44(2.24–4.16) |

*aNon-CF control were children who underwent elective surgery for non-respiratory-related conditions.
of 1 μL of RNase Inhibitor (Life Technologies). RNA purity and yield were determined using a NanoDrop, and integrity was assessed using an Agilent RNA 6000 Nanochip on an Agilent Bioanalyzer.

RNA Sequencing (RNA-Seq) and Analysis
Samples identified with high purity (1.8–2.0 range A260/280) and quality (RIN > 8.0) were then processed for library preparation. Here, the KAPA Stranded mRNA-Seq kit (KAPABiosystems) was used for mRNA capture and fragmentation (~200–300 bp fragments). RNA fragments were then subsequently reverse transcribed into cdNA strands, followed by adapter ligation and library amplification. Sequencing of these libraries (100 bp, single-end) was performed on the Illumina HiSeq 2500 platform at an average depth of 5.08 ± 1.17 million reads (Figure S1A) per. The quality and quantity of the FASTQ sequence reads were assessed using FastQC (v0.11.3) (32), followed by mapping to the reference genome (Homo sapiens hg19/GRCh37 – Ensembl) using "hisat" (v0.1.6-beta) (33). Gene-level quantification (counts) of hisat alignments was performed using SummarizeOverlaps and, finally, post-alignment QC using Samstat (v1.5.2.) (34). Mapping rates to the human genome were within the expected rate for all samples at 88.2–91.2% (35), and post-alignment quality control using SAMStat 1.5.2 reported an average high quality (mapping quality score of thirty) mapping rate of 89.98% ± 0.67 (Figure S1B).

Bioinformatics and Statistical Analysis
Bioinformatics and statistical analyses were performed on five non-CF and seven CF samples. Statistical analysis was conducted in PRISM 8 (v8.1.2; GraphPad Software Inc., California, USA) and included the Mann–Whitney test to compare the statistical variance between genotype, and the Wilcoxon test was used to compare the statistical difference between paired samples. All subsequent bioinformatic analyses post-alignment were performed in R (v3.4.1) (36). To remove low-abundance genes, only those that had a minimum of 10 counts per sample in at least five or more samples were included, resulting in a total of 12,757 genes analyzed. The R package RUVseq (1.10.0) (37) was applied to normalize RNA-Seq read counts between samples to remove the unwanted variance. Differential gene expression was determined using DESeq2 (v1.16.1) (38) after calculating variance-stabilizing transformation (VST) from the fitted dispersion mean relation to yield count data with constant variance along the range of mean values. We determined those genes with an adjusted p-value ≤ 0.05 and ± 1.5-fold change as statistically and biologically significant, respectively. To visualize the variance between samples, a principal component analysis plot was generated using the plotPCA function in DESeq2 and visualized using ggplot2 (v3.1.0) (39). Next, we identified non-infected baseline non-CF and CF enriched gene ontology (GO) terms from the biological process (BP) using Metascape (http://metascape.org) (40). Visualization of GO term analysis was performed using the GOGPlot (v1.0.2) (41). The GoCircle function was used to highlight gene expression changes within each of the selected terms. The value of the z-score from GOGPlot is calculated as zscore = (up - down) ÷ √count, where up and down were the number of up- and down-regulated genes respectively.

Pathway Analysis and Protein–Protein Interaction Network-Based Enrichment Analysis
Pathway analysis based upon Reactome repositories was performed using Signature Over-Representation Analysis (SIGORA) version 2.0.1. The pathway enrichment by SIGORA was identified according to statistically over-represented Pathway Gene-Pair Signatures (Pathway-GPS) (42). To expose the interactive associations among the DEGs at the protein level, genes obtained from both non-CF and CF responses were mapped using protein–protein interactions (PPI) via NetworkAnalyst (http://www.networkanalyst.ca/). Network Analyst (43, 44) and was based upon IMEX Interactome, a comprehensive, high-quality protein–protein interaction database curated from InnateDB (45) to characterize the relationships and interactions of input genes. The network was built by limiting the original seed proteins only and picking zero order interactions.

ELISA
Cytokine production of interleukin 8 (IL-8) (Becton Dickinson, Biosciences, San Diego, CA), interferon lambda 1, 2, 3 (IFNλ), RANTES (CCL5), interleukin (IL)-1B, Interferon gamma-induced protein 10 (IP-10) (R&D, MN, Minneapolis) in culture supernatant was measured by ELISA. Production of interleukin 6 IL-6 was measured using a time-resolved fluorometry detection system (PerkinElmer, Waltham, MA). Expression of interferon beta (IFNβ) was measured using an AlphaLISA® bead-based assay (PerkinElmer).

Corroboration of RNA-Seq Observations in Fully Differentiated Cultures
Experiments were then performed to assess whether unique pathways identified from the initial RNA-Seq analysis were evident in fully differentiated 3-D cultures. Primary ALI cultures were established and, upon TEER confirmation, were rinsed three times with sterile room temperature 1× phosphate-buffered saline for 10 min at 37°C. Cultures were then subsequently infected with RV1b at MOI 0.1 in 200 μL for 24 h. An MOI

| Table 2 | Patient demographics for subjects used for validation work including six non-CF children and six children with CF. |
|----------|--------------------------------------------------------------------------------------------------------------------------------|
| **Non-CF control** | **Cystic fibrosis** |
| Number of participants | 6 | 6 |
| Mean Age ± sd (yr) | 3.3 ± 0.65 | 2.3 ± 2.3 |
| Age range (yrs) | (2.4–4.0) | (0.2–5.9) |
| Male (%) | 50 | 83 |
| Genotype | Healthy non-CF | p. Phe508del / p. Phe508del |
| NE Activity (%) | NA | 50 |
| IL-8 Detected in BALs (%) | NA | 100 |
| Microorganisms detected in BAL (%) | NA | 50 |
| PRAGMA Disease (%) | NA | 3.53 |

*Non-CF control were children who underwent elective surgery for non-respiratory related conditions.
of 0.1 was chosen based on the lowest infection dose from the literature with no reported cytopathic effects or major disruption to the epithelium (46–48) as this prevents assessment of mucin-related enzymes and other downstream analyses. After 24 h, inserts were harvested in RNA lysis buffer for RNA extraction using the PureLink® RNA (Life Technologies) mini kit as per manufacturer instructions. Total RNA was eluted with 30 µL RNase free water. Genes were chosen from at least two independent pathways identified to be uniquely expressed by AEC in children with CF in response to RV infection and included IL1R2, STS8SIA4, ST6GALNAC2, MAN1A1, and B3GNT8. Gene expression was determined via real-time qRT-PCR (refer to Supplementary Materials and Methods 1.1 and 1.2) using TaqMan® pre-designed primer/probes (ThermoFisher Scientific). Details on all primer probes are listed in Table S8.

RESULTS

Patient Demographics
The demographic information of children participating in this study is shown in Table 1 (RNA sequencing) and Table S1 (entire sample sets, including those additional samples used for ELISA). Non-CF controls were children who underwent elective surgery for non-respiratory-related conditions and did not possess existing lung disease. RNA samples of primary AECs obtained from these children (n = 32, 16 non-CF and 16 CF) were originally collected, both pre- and post-infection with RV in vitro.

RNA sequencing was performed on all samples as summarized in the workflow diagram (Figure 1A). A sample elimination process was carried out to exclude unqualified samples (detailed in Figure S3). Samples that were run on a different sequencer did not pass rigorous quality control for RNA sequencing (mapping quality score >30, n = 3), and those with sequencing depth of less than one million reads (n = 9) were excluded from analysis. Finally, RNA sequencing samples from a total of seven CF and five non-CF children were included for the differential expression analysis by applying a fold change cutoff of ≥1.5-fold. Only one child with CF had detectable microorganisms in bronchoalveolar lavage fluid (BALf) during the time of AEC sampling. The PRAGMA CT score presented as percentage of disease was also conducted to demonstrate the quantitative measurement of disease progression during the time of sampling in children with CF.

Distinct Transcriptional Changes of AEC in Response to RV Infection
The normalized read counts matrix was used to build a non-supervised principal component analysis to visualize the major contributors to transcriptional variation within this data set (Figure 1B). The first principal component (PC1, 69% of the variance) completely separated RV-infected and non-infected AEC, and separation of uninfected or infected CF and non-CF samples was observed on PC2 (8% of the variance), indicating that
patient genotype is the second largest source of variation within the data set.

Modest Transcriptional Differences Between Uninfected CF and Non-CF AEC

To determine whether the AECs transcriptional profiles from children with CF are intrinsically different from non-CF controls, non-infected baseline CF and non-CF AECs were analyzed for differential gene expression. We observed a total of 162 DEGs with absolute fold change $\geq 1.5$ between non-infected baseline CF and non-CF AECs. Among those, 92 genes were significantly downregulated, and 70 genes were significantly upregulated in CF AEC compared to non-CF AEC. To identify in which biological processes the 162 DEGs were involved, we performed gene ontology (GO) term enrichment analysis. The predominant enriched GO term in CF AEC is depicted by a circle plot (Figure 2A). The circle plot highlights the overall gene expression change by showing increased expression in red and decreased expression in blue. The $p$-value of the GO terms is represented by the height of the inner rectangle, which is also colored by $z$-score based on GOPlot formula ($z$score = (number of up-regulated genes − number of down-regulated genes) $\div \sqrt{\text{[gene count]}}$). Analysis identified the cytokine-mediated signaling pathway and type 1 interferon signaling pathway as the GO term with an increased $z$-score. The full list of the top upregulated and downregulated genes is summarized in Table S2. The top DEG from differential expression analysis comparing non-infected baseline CF and non-CF AEC was HLA-DQB1 (HLA Class II GWAS genes). We also identified the top 20 genes with the highest fold change between non-infected CF and non-CF AEC (Figure 2B). These genes were found to be involved in biological processes including type 1 interferon signaling pathway (AIM2, BST2, IFI27), keratin (KRT14), DNA methylation (H19), cell cycle (BEX1), extracellular matrix (COL1A2, COL5A1, COL6A1, COL6A2), cell–cell interaction (LGALS7), signal transduction (FST, LRCH2, LRRN1), calcium ion binding (PCDH20), potassium channel (KCNJ5), transferase activity (NEURL3), and phosphatase activity (PTPRZ1).

CF AEC Have More Transcriptional Changes in Response to RV Infection Than Non-CF AEC

We next analyzed the RNA-Seq data to assess the transcriptomic response of CF and non-CF AECs collected after infection with RV. Comparative analysis of response profiles indicates that AECs from both CF and non-CF differentially modulated the expression of several genes related to the innate antiviral immune response in response to RV infection. The Venn diagram (Figure 3A) was used to compare genes that were uniquely and commonly modulated between CF response (RV-infected CF AEC vs. uninfected CF AEC) and non-CF response (RV-infected non-CF AEC vs. uninfected non-CF AEC) to RV infection. A total of 896 (652 upregulated, 244 downregulated) DEGs were observed in the non-CF response to RV and 1442 DEGs (884 upregulated, 558 downregulated) in the CF response (Figures 3A,B). Candidate genes were ranked according to their extent of differential expression when compared to uninfected samples. Although there was considerable overlap between the groups (778 common DEGs, Figure 3C), there were significantly more unique DEGs specific to the CF response (664) compared with the non-CF response (118). A majority of overlapping DEGs were involved in the core immune response to RV infection, including interferon signaling, interferon regulation, cytokine signaling, cell death, and metabolism. The unique DEGs for both CF and non-CF AEC in response to RV infection are summarized (Tables S3, S4, respectively). The top unique DEG for the non-CF response was CX3CL1, which is an important chemoattractant to attract other immune cells, such as dendritic cells. Other top unique genes for the non-CF response were found to be associated with the cellular component (FAXDC2, ARMCX4, RAB17, TMEM17), DNA repair (BRCA2, RMI2), and cellular metabolism (CBR3, B4GALT3, HS3ST3B1, GIPR). Nevertheless, 46% (664 out of 1442) of DEGs for the CF AEC response to RV infection were found to be unique with the IL-1R2 gene, the IL-1 signaling decoy receptor, being the top unique DEG (4.8-fold change). Other unique genes for the CF AEC response were found to be associated with growth factor (PTN), immune response (NOD2, CCRL2, HMOX1, SLC7A2, SERPINB4), cellular metabolism (MDG5A1, ANGPT1), cytoskeletal regulation (LRCH2), signal transduction (MAPK8IP2, STK32A), and transcription regulation (SPDEF, ZNF488).

RV Infection Drives Common Epithelium-Induced Innate Antiviral Response in CF and Non-CF AEC

Genes that were commonly modulated in CF and non-CF AECs (Table S5) were found to be key drivers of core epithelium-induced innate antiviral response to RV infection. Specifically, RV infection triggered a significant upregulation of type I and III interferons (IFNB1, IFNL1, IFNL2, IFNL3) in both CF and non-CF AECs (Figure 3C). However, it was evident that the fold changes (Log2FC) of IFNB1 (5.8-fold), IFNL1 (5.8-fold), IFNL2 (5.1-fold), and IFNL3 (6.1-fold) in gene expression in response to RV infection were lower in the CF AEC response compared to the non-CF AEC response (IFNB1: 6.9-fold, IFNL1: 7.2-fold, IFNL2: 7-fold, IFNL3: 7.5-fold). Interferon signaling also triggered the induction of a variety of interferon-stimulated genes (ISGs), including Mx1; viperin (RSDA2); and the IFITM, IFIT, and OAS family in both CF and non-CF AECs (Figure 3C).

We extended our analysis to identify the biological pathways corresponding to all DEGs in CF and non-CF AECs in response to RV infection. The full list of enriched biological pathways for CF and non-CF AECs’ antiviral responses are provided in Tables S6, S7, respectively. SIGORA pathway analysis was then performed using gene-pair signature pathway analysis, which only accounts for statistically significant gene pairs unique to the over-represented pathways. This analysis identified 52 and 31 biological pathways responsible for CF and non-CF AEC host responses to RV infection, respectively. Comparing the two, we identified 26 common significantly enriched biological
pathways (Figure 4), which are mainly categorized into five main functions, including (1) cytokine signaling in the immune system, (2) presentation to the adaptive immune system, (3) innate immune system, (4) metabolism or biosynthetic, and (5) signal transduction. Consistently, the core antiviral response was demonstrated by type I and III interferon and other antiviral factors as reported earlier with interferon-α/β signaling, interferon-γ signaling, and interferon signaling being the top

FIGURE 2 | Comparison of non-CF and CF non-infected baseline control. (A) Circular visualization of gene-annotation enrichment analysis of non-infected baseline samples. Statistically significant differentially expressed genes (DEGs) between non-infected CF and non-CF samples were annotated using gene ontology (GO). The circular plot combines gene expression and gene-annotation enrichment data. The outer circle shows a scatterplot for each enriched GO term of the Log2FC of the assigned genes. Red dots indicate upregulation, and blue dots indicate downregulation in CF non-infected control compared to non-CF. The inner ring is a bar plot where the height of the bar indicates the significance of GO terms (log10-adjusted p-value), and color corresponds to the z-score: blue, decreased; red, increased; and white, unchanged. (B) Normalized gene counts of the top 20 DEGs between CF and non-CF non-infected baseline samples with the highest fold change, data points represent individual samples for non-CF controls (turquoise; circle) and CF control (coral; triangle).
FIGURE 3 | Response to rhinovirus infection. (A) Venn diagram comparing the differentially expressed (DEGs) genes between non-CF (teal; non-CF RV-infected vs. non-CF non-infected control) and CF (pink; CF RV-infected vs. CF non-infected control) response to RV1B infection. A total of 778 DEGs were common between non-CF and CF response to RV1B infection. (B) The number (y-axis) and direction of change (upregulated = positive y-axis, downregulated = negative y-axis) of DEGs (|Log2FC| > 1.5, adjusted p-value < 0.05) of non-CF and CF response to rhinovirus infection (x-axis). (C) The relative expression genes that are commonly differentially expressed (|Log2FC| > 1.5, adjusted p-value < 0.05) in airway epithelial cells (AECs) from non-CF and CF individuals when infected with rhinovirus. These

(Continued)
three most enriched pathways associated with cytokine signaling. Other common cytokine responses, such as interleukin 20 family signaling (Figure 4A, Table S6), was over-represented with upregulation of IL22RA1, STAT family (STAT1, 2, 3, and 5A), JAK family (JAK1, JAK2), and the negative regulator of IFN signaling SOCS3. Infection with RV has also significantly increased gene expression of chemokines such as CXCL10, CXCL11, CXCL3, CXCL16, CCL2, CCL5, and CCL20 in both CF and non-CF AECs. Additionally, we also detected transcriptional changes in pathogen recognition receptors, such as TLR3, DDX58 (RIG-I) and IFIH (MDA5), and other genes that regulate innate immune signaling, including IKBKE, IRF7, ISG15, NFKBIA, UBE2L6, UBA7, and DDX58. Genes involved in the over-represented pathway Class I MHC-mediated antigen processing and presentation, such as the gene set of F-box protein, TRIM, and the HERC family were the common DEGs in response to RV infection. The PARP protein family, including PARP4, PARP8, PARP9, PARP10, and PARP14, responsible for the regulation of nicotinamide metabolism and salvaging of cellular redox reactions, were also upregulated in response to RV infection. The changes of genes involved in nucleotide biosynthesis pathway pyrimidine catabolism were observed, including NT5C3A, DPYD, TYMP, and NT5E. Changes in gene expression of caspases (CASP1, 4, 8, 10), which provide pivotal links in cell regulatory networks controlling inflammation and cell death, were also observed in both CF and non-CF AEC post-RV infection.

**CF AEC Transcriptome Reveals More Biological Pathways and a More Complex Network in Response to RV Infection Than for Non-CF AEC**

In addition to the common over-represented pathways induced by RV infection, we observed an additional 26 enriched pathways specific to the CF response (Figure 4A). In addition to the five functions mentioned above, the unique over-represented pathways also fall under another two functions, including extracellular matrix organization and vesicle-mediated transport or transport of small molecules. Additional pathways categorized in cytokine signaling in the immune system, such as interleukin 1, 2, 7, 10, and 15 signaling pathways, were the unique enriched pathways specific to CF response. Genes associated with interleukin 1 family signaling–driven proinflammatory activity are IL36G; receptor antagonist IL36RN; IL1R2; IL1RN; receptor IL1R1; protein phosphatase PTPN12; pellino proteins PELI1, PELI3, and IRAK kinase IRAK2, IRAK3; and key immune and inflammatory response regulator S100A12. Other cytokines with essential immunomodulatory functions, including IL-7, IL-10, IL-15, and IL-2 family signaling, were the significantly over-represented pathways unique for CF response to RV infection. Furthermore, we observed a significant upregulation of the chemotactic factors for neutrophils CXCL1 and CXCL2 in the CF AEC response to RV infection. Downregulation of genes encoding E3 ubiquitin ligases, such as TRIM45 (regulator of TNFα-induced NF-kB-mediated transcriptional activity) and RNF128 (inhibitor of cytokine gene transcription), were also only observed in the CF response. The transcriptional change of the HSPA5 gene was also observed in the CF response as part of major histocompatibility complex (MHC) class I molecules mediated adaptive immune regulation.

Several metabolism/biosynthetic pathways of notable interest to CF airway disease include nucleobase catabolism, inositol phosphate metabolism, synthesis of IP3 and IP4 in the cytosol and tryptophan catabolism, which were all altered in CF response to RV infection (Figure 4A). We observed transcriptional changes of ectonucleotidases in the nucleobase catabolism pathways, particularly ecto-nucleoside triphosphate diphosphohydrolases (ENTPDases) ENTPD3 (downregulated) and ENTPD6 (upregulated). The inositol phosphate metabolism pathway was also found to be altered in CF AECs, namely the downregulation of genes encoding phosphohydrolases NUDT11, phospholipase PLCH2 and PLCD4, kinase ITPKB, and phosphatase INPP4B. We also observed a group of upregulated genes, including KNYU, KMO, IDO1, AADAT, and CCBL1, which are associated with the key biosynthetic process of tryptophan catabolism. Biological pathways regulating metabolism of proteins, notably mucin metabolism (O-linked Glycosylation of mucins and sialic acid metabolism), were also over-represented pathways for the CF response. Additionally, RV infection in CF AEC triggered transcriptional changes of transport of small molecules (including cellular hexose transport, metal ion SLC transporters, transport of amino acids, and SLC-mediated transmembrane transport). We noted transcriptional changes for genes involved in extracellular matrix organization, such as integrin α5 and β6 (ITGA5, ITG6A) and cell adhesion molecule ICAM1.

To better understand the potential functional interaction of DEGs, we also visualized expression and investigated the underlying molecular interactions between genes by generating zero-order PPI subnetworks (Figures S2A,B). The main CF and non-CF PPI subnetwork consisted of functionally enriched pathways that play imperative roles in the host antiviral response to RV infection. The non-CF AEC response subnetwork identified associations of 254 nodes and 565 edges (Figure S2A). We observed 172 genes with a degree more than one interactor, where 27 nodes were observed with ≥10 connections with other nodes. Key hub genes regulating the antiviral response
found included STAT1, STAT2, IRF2, IRF1, ISG15, DDX58, IRF7, RIPK1, IKBKE, and CASP8. Conversely, a more complex CF AEC response subnetwork projected the associations of 493 nodes and 1156 edges (Figure S2B). We observed 320 genes with a degree more than one, where 66 nodes were observed with ≥10 connections with other nodes. The key hub genes regulating the CF AEC response subnetwork included IRF1, ISG15, STAT1, STAT3, HSAP1B, CASP8, TBK1, IKBKE, TRAF2, and CASP8. The key hub genes of both CF and non-CF subnetworks are represented by key regulators related to the innate immune system and cytokine signaling.

**Aberrant Cytokine Production of CF AECs to RV Infection**

In order to validate the transcriptional changes of the enriched cytokine signaling pathways, we measured the levels of key innate and inflammatory cytokine production at 24 h post-RV infection (Figure 5). Although IFNB1 was significantly induced upon RV infection in both cohorts, this is not reflected at the protein level with significantly lower levels (average 10.8-fold) of IFNβ1 (type 1 interferon) released by CF AEC (668.3 ± 576.2 pg/ml; p < 0.05) compared to non-CF AEC (7,265 ± 6,558 pg/ml). As shown in Figure 3C, all type 3 IFN genes (IFNL1, IFNL2, IFNL3) were upregulated post-RV infection. Cytokine levels of the type III interferons IFNλ1, IFNλ2, and IFNλ3 were also significantly elevated in both CF and non-CF AEC infected with rhinovirus. However, levels of IFNλ1 (296.4 ± 293.3 pg/ml) and IFNλ2 (334.6 ± 642.8 pg/ml) produced by CF AEC in response to RV infection were significantly (3.5- to 5-fold) lower when compared to non-CF AEC (1,059 ± 1,170 pg/ml and 1,665 ± 1,932 pg/ml, respectively; p < 0.05). IFNλ3 produced by CF AEC (285.3 ± 287.3 pg/ml) following RV infection was somewhat but nonsignificantly lower compared to that produced by non-CF subject AEC (928.6 ± 997.9 pg/ml). Similar cytokine levels of antiviral chemokines CCL5 (RANTES) and IP10 and pro-inflammatory cytokines, including IL6 were detected in non-infected CF and non-CF AECs, and similar increases in these proteins occurred in response to RV infection. However, IL-8 and IL-1β cytokine production is significantly elevated in non-CF AECs in response to RV infection compared to CF AECs.
Corroboration of Unique Gene Expression Patterns in Response to RV Infection in CF ALI Cultures

To validate results generated from the RNA-Seq in a model that better represents the airway, we assessed the expression of some unique DEGs identified in submerged CF cultures post-RV infection by challenging ALI cultures with the same RV and again assessing gene expression at 24h (Figure 6). Expression of the top unique DEG for the CF response, IL1R2, was validated with a consistent increase in CF ALI post RV infection (9.4-fold over uninfected, p < 0.05; Figure 6). Upregulation of IL1R2 appeared bimodal in non-CF ALI and was not significant (p = 0.30). Furthermore, expression of genes involved in glycosylation of mucins and sialic acid metabolisms, namely sialyltransferase ST8SIA4, ST6GALNAC2, mannosidase MANIA1, and acetylglucosaminyltransferase B3GNT8, was also validated as unique to CF (Figure 6). A significantly higher level of sialyltransferase ST8SIA4 expression (2.2-fold, p < 0.05) was observed in RV-infected CF ALI cultures while ST6GALNAC2 was significantly downregulated (−1.4-fold, p < 0.05) in CF ALI in response to RV infection. Expression changes of these genes were all consistent with the RNA-Seq data from submerged cultures. We observed that these genes did not change expression in non-CF ALI cultures upon infection with RV; ST8SIA4 (3.6-fold, p = 0.14); ST6GALNAC2 (−1.3-fold, p = 0.09); MANIA1 (5.4-fold; p = 0.08); B3GNT8 (1-fold, p = 0.28).
Corroboration of uniquely expressed genes by CF AEC in fully differentiated cultures. Gene expression of the top unique gene IL1R2, as well as mannosidase (MAN1A1), sialyltransferases (ST8SIA4 and ST6GALNAC2), and acetylglucosaminyltransferase (B3GNT8), were uniquely altered in CF air-liquid interface cultures when infected with rhinovirus (RV) for 24 h. Data points represent individual samples for non-CF controls (turquoise), non-CF infected (purple), CF control (coral), and CF infected (green). Expression of all genes were normalized to the housekeeping gene PPIA (49). *Indicates $p < 0.05$ by one-way ANOVA following normal distribution test.
DISCUSSION

To improve knowledge of the underlying epithelial transcriptional responses during infection with rhinovirus, a major respiratory pathogen, we performed RNA sequencing on primary AEC from children with CF and non-CF controls in vitro at baseline and post-RV infection. There are five important findings from this study: (i) There were only modest baseline transcriptional differences between non-infected CF and non-CF AECs prior to exposure to RV, (ii) there was conservation in certain core antiviral responses (e.g., IFN signaling) of CF and non-CF AECs at the transcriptomic level but not the protein level, (iii) CF AECs elicited a larger and more complex transcriptional response compared to non-CF AECs with multiple unique biological pathways represented, (iv) key among these biological pathways are cytokine signaling and biosynthetic pathways (e.g., O-linked glycosylation of mucins) as they are highly relevant to CF lung pathology, (v) we corroborated observations made from the RNA-Seq analysis in fully differentiated cultures and identified genes involved in IL-1 signaling and mucin glycosylation that are only dysregulated in the CF airway epithelial response to RV infection. Collectively, these results identify potential biological pathways and processes that could be contributing to the adverse outcomes typically seen in people with CF during virus infection.

There were only modest baseline transcriptional differences between non-CF and CF AECs. This is most likely reflective of the very early and mild lung disease in the CF cohort. Minimal baseline differences also provide confidence that any difference in the antiviral transcriptional changes that we observed were due to infection. Nevertheless, the top differentially expressed baseline was HLA-DQB1, previously identified in a GWAS study with a high association signal to CF lung disease severity [reviewed in (50)]. Interestingly, the major enriched GO terms for the differentially expressed genes in non-infected baseline samples were denoted by the cytokine-mediated signaling pathway and type 1 interferon signaling pathway. Among these, AIM2 inflammasome (associated with induction of pyroptosis, activation of pro-inflammatory cytokines, and viral suppression) (51, 52) IFI27 (also known as ISG15α) contributes to IFN-dependent perturbation of normal mitochondrial function and enhanced cellular apoptosis (53), and the IFN-dependent antiviral factor BST2 were all significantly downregulated in CF AECs. In response to RV infection, several common responses were found, including interferon signaling. However, the induction of type 1 and 3 interferon genes was lower in CF AEC. This was mirrored by reduced type 1 (IFNβ1) and type 3 interferon (IFNλ1 and IFNλ2) protein in supernatant. The reduction of type 1 and 3 interferon production of CF AECs in response to RV infection could be associated with the negative regulation of interferon signaling by the unique key gene, such as STAT3 (54, 55); however, this requires further characterization. Conversely, the IL-1 family signaling pathway was unique to the CF AECs response to RV infection, but in this case, IL-1β protein was significantly lower in CF supernatant compared to non-CF. This unusual observation could be, in part, mediated by negative regulators of IL-1 signaling expressed in CF AEC, including IL1R2 and IL1RN, pellino protein genes PELI1 and PELI3, together with interleukin 1 receptor-associated kinase IRAK2 and IRAK3. We then assessed the expression of IL1R2 pre- and post-RV infection in a differentiated culture model and made similar observations to those obtained using submerged cultures.

The IL-1 signaling pathway has been suggested as a link between hypoxic cell death and sterile neutrophilic inflammation in CF (56). Both IL-1α and IL-1β were detectable in bronchioalveolar lavage fluids (BALs) of young children with CF in the absence of bacterial infection, highlighting potential for inflammation of the CF airway under sterile inflammation (57). Since S100A12 (key regulator of inflammatory process) is also part of the IL-1 family signaling pathway in CF response to RV infection, we postulate that the CF AECs could be directing from pro-inflammatory IL-1 signaling under sterile inflammation to a hyperinflammatory condition characterized by NF-κB signaling cascades during RV infection. Other evidence suggests the alteration of the inflammatory response with abundant cytokine signaling pathways (interleukin 1, 2, 7, 10, and 15 signaling) in CF AEC post-RV infection could be explained by downregulation of RNF128 genes, which functions as an inhibitor of cytokine gene transcription and could interact with TBK1 (key hub of CF AEC response in our study here) kinase activity to enhance antiviral immunity. We also observed an elevated IL-8 production in both CF and non-CF AECs post-RV infection with higher amounts produced by non-CF AECs compared to CF AECs. Our IL-8 results agree with a previous study that also utilized primary AEC cultures in a similar infection setting (58) but contrasts with another that observed elevated inflammatory mediator release by the CF AECs (23). Overall, the over-represented cytokine signaling pathways suggest a unique and prominent role in regulating inflammation in CF AECs when infected with RV. However, with conflicting observations in this area, elucidating the complexity of the inflammatory response with associated cell death in CF AECs warrants further investigation.

We also identified over-represented metabolic pathways in CF AEC in response to RV infection specifically involved in the regulation of immunity, including inositol phosphate metabolism and synthesis of IP3 and IP4 in the cytosol, suggesting an altered CF airway microenvironment after RV infection. The induction of inositol phosphate has previously been related to endoplasmic reticulum expansion and Ca2+ storage, resulting in Ca2+-dependent transcriptional activity of inflammatory mediators (59), which could contribute to hyperinflammatory responses seen in the CF AEC to viral infection (23). Upregulation of extracellular ectonucleotidase in the inositol phosphate metabolism pathway was found to cause depletion of ATP concentrations, reduction of air–surface liquid volume, ASL collapse, and failure in mucociliary clearance may trigger CF lung disease exacerbations as shown previously in a model of respiratory syncytial virus infection (60). Another metabolic pathway, tryptophan catabolism, was also one of the over-represented pathways in the CF AECs following RV infection. Tryptophan metabolism has been previously found to be dysregulated in CF AEC (61) and has been implicated in Pseudomonas aeruginosa infection, oxidative stress, and Th17 hyperinflammation (62, 63). Alteration of
tryptophan metabolism results in accumulation of kynurenine and anthranilate, which could subsequently disrupt the homeostatic balance of the host’s innate immune system and reduce the antimicrobial activity of airway epithelium.

Other identified biosynthetic pathways associated with RV infection in CF include sialic acid metabolism and O-linked glycosylation of mucins. Sialic acids are a family of negatively charged monosaccharides that are commonly expressed as the terminal residues in glycans of the glycoconjugates on the epithelial cell surface lining the airways and are also major components of secreted mucins in the airway. Previous studies have identified increased fucosylation and decreased sialylation in cultured AEC while a contrary observation was reported in CF sputum (64–67). As a key player that contributes to the rheological properties of mucus, aberrant sialic acid metabolism may worsen the pathological conditions of CF. O-linked glycosylation is a post-translational modification process and occurs within the endoplasmic reticulum (ER) and Golgi complex. The enzymes in ER and Golgi complex regulate glycosylation of N-glycans and O-glycans by successively adding to and then remodeling mucin oligosaccharides prior to transport to cell membranes for tethering or secretion. Here, alteration of genes encoding glycosyltransferases, such as N-acetylgalactosaminyltransferases, N-acetylgalactosaminytransferases, and galactosyltransferases, were reported from our RNA-Seq analysis. We corroborated a number of these as unique to the AEC response to RV in children with CF. Changes in these glycosyltransferases could potentially alter the O-glycans on cell surfaces and, thus, affect interactions with airway pathogens and irritant exposures. Emerging evidence suggests alteration of mucin glycosylation is a response to infection and inflammation and might induce extended conformational changes to prevent damage from proteolytic enzymes (68). Although the impact that CFTR mutations has on mucin biomolecules is unknown, our results suggest that RV infection could be a potential mechanism that contributes to changes in mucin glycosylation that are exclusive to CF and that might influence mucosal barrier function. A previous investigation has demonstrated that a surplus of unfolded proteins that results from blocked glycosylation leads to prolonged ER stress and activation of the unfolded protein response (UPR) causing cell death (69). Previous in vitro work using an immortalized cell line discovered a pronounced reprogramming of host cell metabolism toward an anabolic state, including upregulation of glucose uptake, glycolysis, nucleotide synthesis, and lipogenesis (70). Considering most of the metabolic changes found in this study occur post-RV infection, future studies integrating the transcriptomic signature patterns with analyses of the metabolites produced by CF AEC in response to RV infection will provide significant insight into the exact metabolic changes that occur during infection.

Interestingly, RV infection in CF AECs results in the upregulation of a group of SLC transporter genes, including upregulation of CP (ferrooxidase), SLC41A2 (magnesium transporter), SLC30A1 (zinc transporter), and SLC39A8 (zinc transporter) and downregulation of SLC39A10 (zinc transporter) and SLC40A1 (iron-regulated transporter). Increasing total iron and zinc has previously been associated with airway inflammation in CF (71). These results suggest that RV infection in the CF airway is associated with the presence of redox active biometals. A previous study (72) has suggested that the dysregulation of iron homeostasis is accompanied by a respiratory virus infection, which, in turn, facilitates pseudomonas biofilm growth. Understanding the mechanistic link of virus infection to the alteration of the cellular microenvironment and instigation of secondary infection might aid in development of new treatment.

We acknowledge some limitations in the experimental design. First, we only analyzed transcriptomics at the 24-h time point, primarily due to the limited number and expansion of primary cells established from each patient. However, early optimization of our infection model did assess the transcriptional changes earlier (data not shown), and the greatest transcriptional change identified occurred at the 24-h time point. Although methodologies now exist to assist with primary AEC expansion in vitro (29), its effects at the transcriptomic level remain unknown and, thus, the use of unaltered primary airway cells remains a significant strength of this study. Future investigations could possibly include additional time points to better appreciate the transcriptional signature changes over the full course of RV infection as well as the long-term consequence of viral infection on CF AECs. Second, this study utilized a laboratory strain of rhinovirus (RV1b), which might exert differential effects on CF AEC compared to clinically derived isolates known to cause exacerbations in this cohort. With different RV serotypes causing infection in CF airways (10), future studies may identify whether innate immune responses may be serotype-specific. Similarly, comparison studies to other viruses (respiratory syncytial virus, influenza) would also assist in our understanding of the contribution of early-life viruses to CF disease progression. Finally, the simplified monolayer cell culture model of basal CF AECs may be regarded as a limitation, but basal cells are the primary target of RV (73). While monolayer cultures may oversimplify the multicellular interactions of epithelial (ciliated, goblet, basal, secretory cells) and immune cells (dendritic cells, neutrophils), it is an important, repeatable model with low methodological variation, and we were able to validate genes in differentiated AEC. Overall, we are highly confident that limitations are minor and that our results provide new insight into new therapeutic targets for treating acute viral infections in CF that can be validated in future transcriptomic studies assessing differentiated AEC models.

In conclusion, this study shows that, at the transcriptomic level, CF AECs induce a complex and unique set of responses when infected with RV in vitro that have implications for lung disease progression in CF. Despite type I, II, and III interferon signaling being involved in the core CF antiviral response, IFNs protein levels were lower in CF AEC when compared to non-CF AEC. Metabolic and biosynthetic pathways were unexpectedly integrated with the core CF antiviral response, and multiple key regulatory molecules of antiviral response were dysregulated in CF AEC, revealing new potential to modulate CF AEC innate immunity to RV infection. Future work will explore whether these regulatory molecules are potential targets for therapy.
unique to RV and may be leveraged to reduce the impact viral infections have on lung disease progression CF.

**DATA AVAILABILITY STATEMENT**

Raw datasets have been uploaded to GEO, with accession number GSE138167.

**ETHICS STATEMENT**

The study was approved by the St. John of Gods Human Ethics Committee (SJOG#901) and Perth Children's Hospital Ethics Committee (#1762) and written informed consent was obtained from parents or guardians.

**AUTHOR CONTRIBUTIONS**

Conceptualization: SS and AK. Funding acquisition: SS, AK, RH, and ST. Methodology: K-ML, EG, AL, TL, PA-R, TI, and TR. Supervision: SS, AK, TL, and LG. Validation: K-ML, ES, and LG. Manuscript writing: K-ML, LG, and AK. Manuscript review: SS, RH, TL, PA-R, ES, ST, TI, and EG. All authors contributed to the article and approved the submitted version.

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**SUPPLEMENTARY MATERIAL**

The Supplementary Material for this article can be found online at: https://www.frontiersin.org/articles/10.3389/fimmu.2020.01327/full#supplementary-material

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Conflict of Interest: The authors declare that the research was conducted in the absence of any commercial or financial relationships that could be construed as a potential conflict of interest.

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