

Epigenetic and transcriptional plasticity drive inflammation in Cystic Fibrosis

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21 **Abstract**

22 The relative contributions of acute and chronic inflammatory mechanisms in diseases
 23 characterised by persistent bacterial infection are unclear, despite important consequences
 24 for the development of novel anti-inflammatory therapies. Here, we examined longitudinal
 25 transcriptional and epigenetic changes in patients with Cystic Fibrosis (CF), a genetic
 26 condition characterised by persistent bacterial lung infections that drive progressive
 27 inflammatory lung damage. We find that sudden clinical deteriorations in lung health, termed
 28 Acute Pulmonary Exacerbations (APEs), are linked to innate immune signalling driven by
 29 bacterial components, and accompanied by changes in macrophage function. Treatment of
 30 patients with intravenous antibiotics results in rapid modification of myeloid cell gene
 31 expression and epigenetic state, towards that of healthy volunteers, suggesting that
 32 repeated acute inflammatory episodes play an important role in CF inflammatory lung
 33 damage.

34 Main Text

35 Introduction

36 Cystic fibrosis (CF) is a genetic disease associated with dysfunction of the CF transmem-
 37 brane conductance regulator (CFTR). Impaired mucociliary clearance leads to the colonisa-
 38 tion of the lungs by bacterial pathogens, and a pattern of inflammation and infection is es-
 39 tablished. Cytotoxic products released by neutrophils to kill bacteria damage tissues and are
 40 associated with lung function decline (1).

41 Inflammation in CF has generally been regarded as chronic and continuous in nature (1; 2).
 42 However, there is also clinical evidence that cyclical processes are involved. CF patients
 43 frequently undergo episodes of rapid clinical deterioration (APEs) (3) that require prolonged
 44 (usually 14 days) treatment with intravenous antibiotics. It is not clear to what extent inflam-
 45 mation during such periods diverges from the baseline observed when the patient is clinically
 46 stable. CF therefore provides a unique opportunity to understand the temporal dynamics of
 47 inflammation in humans in the context of long-term bacterial infection.

48

49 Results and Discussion

50 We collected whole blood samples from adult CF patients ($n = 13$) infected with *Pseudomo-*
 51 *nas aeruginosa* (PsA) at 3 timepoints: at an APE onset, at the end of intravenous antibiotic
 52 treatment, and when returned to stable clinical baseline (at least 30 days after the APE
 53 onset; **Fig. 1a**; **Dataset 01**). Analyses were performed on peripheral blood myeloid cells,
 54 due to the technical difficulty in isolating lung myeloid cells. Treatment with intravenous an-
 55 tibiotics was associated with resolution of systemic inflammation and an improvement in

lung function (**Fig. 1b**). The study group was broadly representative of the clinical characteristics of the entire cohort attending the adult CF centre from which they were recruited (**Fig. 1c**).

Gene expression of peripheral neutrophils and monocytes isolated from CF patients on each sampling day was directly compared with samples from age- and sex-matched healthy volunteers (HV; $n = 8$; **Fig. 1d**). We observed a marked decrease in differential gene expression between CF and HV across the time series, with more differentially expressed (DE) genes at the onset of exacerbation (day 0), than on subsequent sampling days (days 14 and 30+) (**Fig 1d; Datasets 02 & 03**). The median expression level of DE genes across patients was responsive to antibiotic treatment, indicating that the effect was distributed across individuals and was not driven by subset of the study patients (**Fig 1e**).

Given the clear temporal dynamics of DE genes (**Fig 1f**), we used unsupervised hierarchical clustering to identify 7 gene clusters in neutrophils (**Fig. 1g**). Clusters of upregulated genes in CF were enriched for processes involved in innate immune response (**Fig. 1h**), while downregulated clusters showed no immune-specific enrichment.

We also found differences in the functional properties of monocyte-derived macrophages between day 0, day 14, and day 30+, with decreased intracellular killing and greater inflammatory cytokine production observed in Day 14 cells (**Fig. 1i**).

We next assessed changes at the epigenetic level, in both cell types, through genome-wide profiling of H3K27ac by chromatin immunoprecipitation (ChIP-seq). Differentially acetylated regions (DAR) were identified for each time point by direct comparison with HV samples (**Datasets 04 & 05**). Again, fewest DAR were identified in both neutrophils and monocytes at Day 14, indicating that IV treatment causes a reduction of the effects observed at day 0.

79 Two genes (*TLR5* and *MAP3K20*) had coordinate changes in transcription and H3K27ac at
80 all time points in neutrophils (**Fig. 2a; Dataset 06**), while no genes had coordinate changes
81 in monocytes (**Dataset 07**). In total, four DARs were assigned to *TLR5* (see **SI Appendix,**
82 **Materials and Methods; Fig. 2a**), spanning the promoter region and a distal site located 57
83 kb upstream that has not previously been associated with *TLR5* expression (**Fig. 2b**).

84 We integrated all data layers for neutrophils and monocytes using a multi-omics factor anal-
85 ysis (MOFA) (**4**) and identified eight latent factors accounting for 30-60% of the variance
86 across data types (**Figure 2c**). Several clinical variables were significantly correlated with
87 latent MOFA factors (**Figure 2c**), indicating that phenotypic variation among CF patients is
88 reflected at gene expression level and in the epigenetics of innate immune cells.

89 As all patients in this study were infected with PsA, we sought to assess the extent to which
90 direct detection of PsA by myeloid cells was responsible for the observed changes. Neutro-
91 phils from three healthy volunteers were exposed to PsA or purified PsA flagellin (a known
92 ligand for TLR5 (**5**)) and their transcriptional response was determined using RNA-seq.

93 As expected, the majority (86%) of the genes responsive to flagellin were also differentially
94 expressed in PsA-exposed cells ($P = 8.5 \times 10^{-207}$) (**Fig. 2d**). PsA-upregulated genes over-
95 lapped with genes overexpressed significantly in CF neutrophils at day 0 (86 genes, $P = 5.0$
96 $\times 10^{-13}$) and at day 30+ (7 genes, $P = 0.01$), but not at day 14 (2 genes, $P = 0.31$; **Fig. 2;**
97 **Dataset 08**).

98 Genes upregulated in PsA-exposed neutrophils and CF patient neutrophils formed a highly
99 connected functional network (STRING (**6**) protein-protein interaction enrichment P value $<$
100 1×10^{-16}). The largest connected component of this network included a subset of genes
101 directly responsive to flagellin exposure (**Fig. 2e**), implicating this bacterial product as an
102 important inflammatory driver. Moreover, 23 of the 88 proteins encoded by the genes in the

103 network are responsive to one or more approved drugs (**Dataset 09**) and a further 6 to
104 therapeutic compounds not yet approved (**Dataset 09**).

105 In summary, we have shown that the epigenetic, transcriptional, and functional properties of
106 CF patient myeloid cells fluctuate temporally during the infection cycle. Additional work is
107 required to determine whether or not the inflammatory program changes between periods
108 of disease exacerbation. Peripheral myeloid cells have a short half-life, and the lack of per-
109 sistence of epigenetic changes between sampling dates indicates that the bone marrow
110 progenitors of these cells are not permanently modified during the infection cycle. Although
111 we cannot exclude the possibility that antibiotic treatment directly modulates gene expres-
112 sion in these cells, the reduction in the expression of pro-inflammatory genes persists be-
113 tween Days 14 and 30+, indicating that it is more probably an indirect result of the effect of
114 reducing bacterial burden in the lung.

115 We observed increased TNF production and decreased bacterial killing from macrophages
116 isolated at Day 14 relative to the other sampling days. This may indicate that macrophages
117 at Day 14 are sensitised to the presence of bacteria, due to the reduction in the infection
118 burden caused by the administration of intravenous antibiotics. However, further work is
119 necessary to establish the exact basis of such fluctuations. We also defined a functionally
120 connected network of genes, which is directly responsive to PsA exposure. Targeting this
121 network could therefore offer a rational strategy for reducing inflammation-associated pa-
122 thology in CF.

123

124 **Materials and Methods**

125 Patients with Cystic Fibrosis (CF) were enrolled from Royal Papworth Hospital, UK, after
 126 written informed consent (ethical approval REC 19/EE/0241). Peripheral blood samples
 127 were collected and processed using established protocols from the BLUEPRINT consortium
 128 (7). For the cytokine production assay, experiments were conducted using previously
 129 described methods (8). A detailed description of all Materials and Methods is available in **SI**
 130 **Appendix, Materials and Methods.**

131

132 **Data, Materials, and Software Availability**

133 Sequence data have been deposited at the European Genome-phenome Archive (EGA)
 134 under accession number EGAS00001006421. Data files and code used for clustering and
 135 functional enrichment of differentially expressed genes can be accessed at
 136 <https://doi.org/10.5281/zenodo.7113651>.

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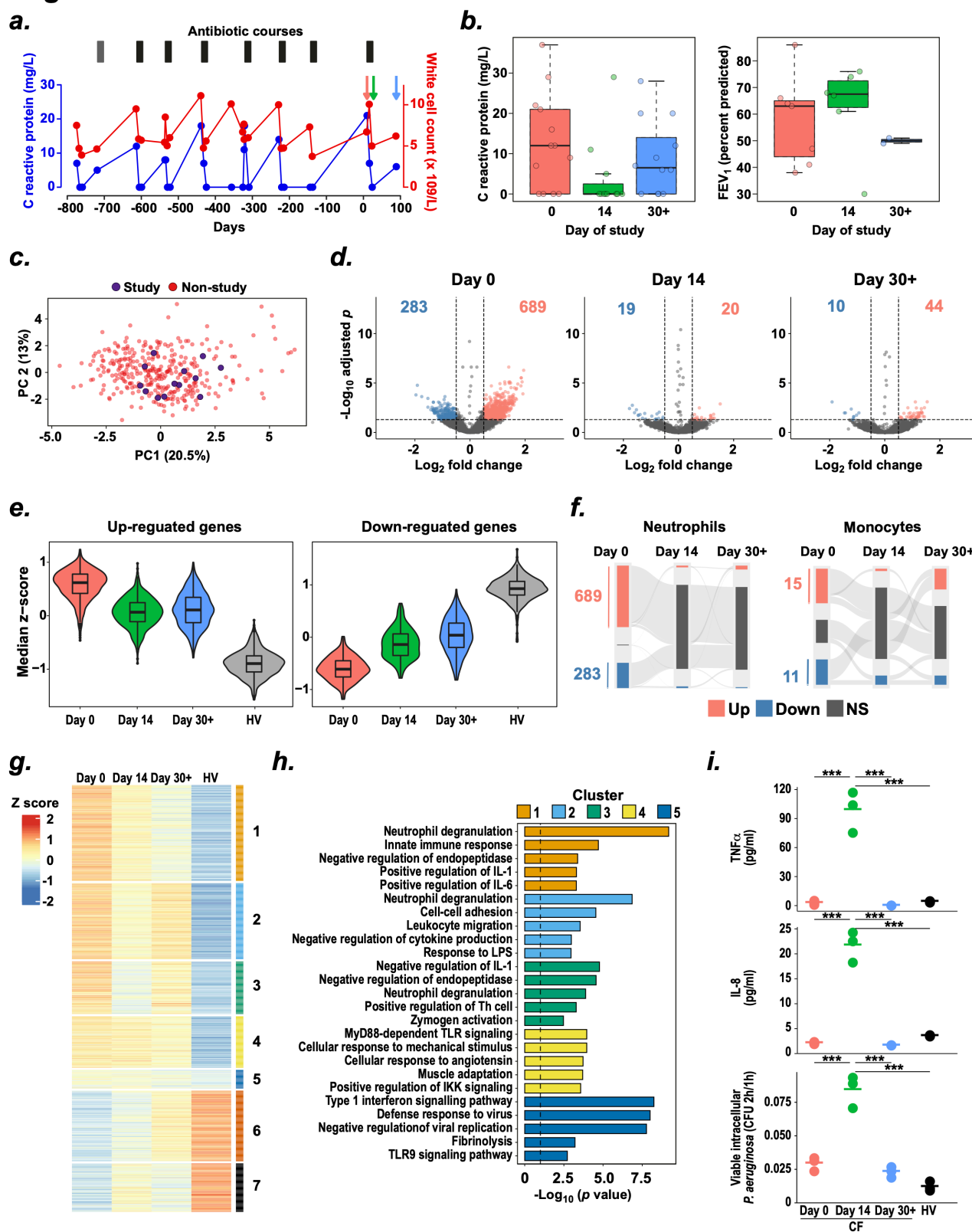
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163

164

Figure 1



166 **Figure 1. (a)** Longitudinal fluctuation in c-reactive protein (CRP) and total white blood cell
167 (WBC) count over a three-year period in a patient recruited for this study. Red arrow
168 indicates Day 0; green arrow indicates day 14; blue arrow indicates Day 30+. **(b)** Levels of
169 CRP and percent predicted forced expiratory volume in one second (FEV₁ % predicted) for
170 study patients at each sampling day. **(c)** Principal component analysis of clinical data set for
171 354 CF patients. Patients selected for inclusion in this study are indicated in purple. **(d)**
172 Volcano plots show differentially expressed (DE) genes in CF neutrophils relative to HV at
173 the indicated time points. **(e)** Median expression levels of DE genes by time point and study
174 group. **(f)** Gene expression dynamics in neutrophils and monocytes. Parallel set diagrams
175 show the changes in numbers of DE genes over sampling days **(g)** Unsupervised
176 hierarchical clustering of neutrophil DE genes. Median z-scores per sample group are
177 plotted for each gene. Gene clusters are indicated with coloured bars above heatmap. **(h)**
178 GO term functional enrichment of neutrophil DE genes by cluster. Dashed line represents
179 adjusted *P* value = 0.05. **(i)** Cytokine release and intracellular bacterial killing by monocyte-
180 derived macrophages. Assays were performed at 4 hours following exposure of cells to PsA.
181 Data from an example CF patient is shown. *** Adjusted *P* < 0.001, one-way ANOVA with
182 Tukey's post-hoc test.

184 **Figure 2. (a)** RNA-Seq and ChIP-Seq log2 fold change values by sampling day. All
185 differentially acetylated H3K27Ac peaks assigned to genes are shown as dots. Red dots,
186 increased in CF relative to HV; blue dots, decreased in CF relative to HV; grey dots, no
187 significant difference. **(b)** RNA-Seq and H3K27Ac ChIP-Seq coverage of the *TLR5* gene (left
188 panel) and distal region of acetylation (right panel). The total coverage is shown as the
189 median fragments per kilobase per million (FPKM) for RNA-Seq, and median reads per
190 kilobase per million (RPKM) for ChIP-Seq, in 40 bp bins. **(c)** Correlation of clinical data with
191 MOFA factors. Abbreviations: ALP, alkaline phosphatase; ALT, alanine transaminase; Asp,
192 Aspergillus; AUC, area under curve; BMI, body mass index; CRP, C-reactive protein;
193 FEV1 % pred., forced expiratory volume in one second (FEV1) percent of predicted; IgE,
194 immunoglobulin E; IgG, immunoglobulin G; RAST, radioallergosorbent; WBC, total white
195 blood cell count. **(d)** UpSet plot shows number of neutrophil upregulated genes in *P.*
196 *aeruginosa*-infected HV neutrophils, flagellin-exposed HV neutrophils, and in CF at Day 0,
197 Day 14, and Day 30+. **(e)** Functional network of genes commonly upregulated in *P.*
198 *aeruginosa*-infected HV neutrophils (PsA) and CF neutrophils, constructed using
199 interactions from the STRING database. A subset of genes in this network are also
200 upregulated upon flagellin (flg) exposure. NS, not significantly differentially expressed.