

Original research

Faecal microbial transfer and complex carbohydrates mediate protection against COPD

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ABSTRACT

Objective Chronic obstructive pulmonary disease (COPD) is a major cause of global illness and death, most commonly caused by cigarette smoke. The mechanisms of pathogenesis remain poorly understood, limiting the development of effective therapies. The gastrointestinal microbiome has been implicated in chronic lung diseases via the gut–lung axis, but its role is unclear.

Design Using an *in vivo* mouse model of cigarette smoke (CS)-induced COPD and faecal microbial transfer (FMT), we characterised the faecal microbiota using metagenomics, proteomics and metabolomics. Findings were correlated with airway and systemic inflammation, lung and gut histopathology and lung function. Complex carbohydrates were assessed in mice using a high resistant starch diet, and in 16 patients with COPD using a randomised, double-blind, placebo-controlled pilot study of inulin supplementation.

Results FMT alleviated hallmark features of COPD (inflammation, alveolar destruction, impaired lung function), gastrointestinal pathology and systemic immune changes. Protective effects were additive to smoking cessation, and transfer of CS-associated microbiota after antibiotic-induced microbiome depletion was sufficient to increase lung inflammation while suppressing colonic immunity in the absence of CS exposure. Disease features correlated with the relative abundance of *Muribaculaceae*, *Desulfovibrionaceae* and *Lachnospiraceae* family members. Proteomics and metabolomics identified downregulation of glucose and starch metabolism in CS-associated microbiota, and supplementation of mice or human patients with complex carbohydrates improved disease outcomes.

WHAT IS ALREADY KNOWN ON THIS TOPIC

⇒ Changes in gut microbiota are associated with chronic obstructive pulmonary disease (COPD) but the underlying host and microbial mechanisms are unclear, limiting the therapeutic applications.

WHAT THIS STUDY ADDS

⇒ Microbiome composition and metabolism is reproducibly correlated with lung and gastrointestinal pathology in experimental COPD.
⇒ Microbiome modifying interventions effectively alleviate disease, including protective effects supplementing smoking cessation.
⇒ Lung and colonic immune changes can be induced by transfer of microbiota from mice with experimental COPD into naïve recipients.
⇒ Nutritional interventions targeting the microbiome in patients with COPD demonstrate efficacy in a small pilot study.

HOW THIS STUDY MIGHT AFFECT RESEARCH, PRACTICE OR POLICY

⇒ Microbiome-targeting therapeutics and nutritional interventions may be developed for COPD, including as supplements to smoking cessation.

Conclusion The gut microbiome contributes to COPD pathogenesis and can be targeted therapeutically.

Chronic obstructive pulmonary disease (COPD) is most often caused by long-term cigarette smoke (CS) inhalation and includes chronic inflammation, airway remodelling and emphysema leading to progressive lung function impairment which persists after smoking cessation.¹ Over 210 million people globally have COPD, causing >3.3 million deaths annually although even these are substantially under-reported.² Pharmacological treatments have little efficacy in reversing disease, suppressing its progression or preventing mortality and have significant adverse effects.³ Mechanisms of COPD pathogenesis remain poorly understood and no singular driver of disease has been discovered, suggesting that chronic effects from a range of factors drive disease pathogenesis over many years.⁴ Development of effective therapies will require targeting this array of factors.

The microbiome is under intense investigation for its immunoregulatory capacity and association with disease.⁵ Most COPD studies have focused on respiratory microbiota, demonstrating increased bioburden, reduced diversity and enrichment of *Firmicutes* and *Proteobacteria*.⁵

The gut hosts the largest and most diverse microbiome of the human body that, depending on its composition, can drive or suppress inflammation, including in the lung.⁶ Alterations in the gut microbiome from antibiotics or diet influence asthma development and respiratory infections.⁶ Chronic CS-exposure induces gastrointestinal histopathology,⁷ and patients with COPD have increased risk of inflammatory bowel diseases⁶ and altered gut microbiome composition.⁸

Transfer of whole microbial communities from healthy individuals through faecal microbial transfer (FMT) is an effective therapy in patients with *Clostridioides difficile* colitis, but its implementation in other diseases is not as well supported.⁹ FMT may therefore have cost-effective benefits in COPD, but further examination is required. A study using a short-term smoke and poly I:C (TLR3 agonist) model showed that FMT prevented emphysema and identified associations with *Bacteroidaceae* and *Lachnospiraceae* families by 16S rRNA gene sequencing,¹⁰ but important controls and analyses linking taxa to pathogenesis were lacking and the model was not representative of human CS-induced COPD. Similarly, *Parabacteroides goldsteinii* lipopolysaccharide (LPS) had protective effects in a murine model of COPD, but broader associations between microbiota and disease were not identified.¹¹ Crucially, neither study identified bacteria associated with COPD in human studies⁸ or demonstrated human translation.

We hypothesised that gastrointestinal microbiota contribute to COPD pathogenesis, and provide a detailed characterisation of the gastrointestinal microbiome in experimental CS-induced COPD using multiomics. FMT protected against experimental COPD through changes in microbiota that correlated with key disease features. Proteomics and metabolomics implicated restoration of bacterial complex carbohydrate metabolism in the protective effects, supported by interventional studies in experimental COPD and human patients with COPD.

METHODS

Mice, CS-exposure, microbiome transfer and diet studies

Female C57BL/6 mice (3–5 weeks old) from the University of Newcastle Animal Service Unit (Newcastle, Australia) underwent 3 weeks of baseline microbiome normalisation

by transferring soiled bedding and co-housing. Microbiome normalisation was not performed for validation experiments (8 weeks CS).

Mice were exposed to normal room air or CS from 12 3R4F cigarettes (University of Kentucky, Lexington, Kentucky, USA) twice per day, 5 days per week, for 8 or 12 weeks as previously described.^{12–20} FMT was administered twice per week by transfer of soiled bedding or oral gavage of faecal supernatants from age-matched air-exposed mice to CS-exposed mice and vice versa (online supplemental figure S1). Experimental controls were used as donors for the FMT to ensure the use of age-matched donors which had undergone microbiome normalisation, minimising variability and confounding factors (online supplemental figure S1).

For antibiotic depletion experiments, mice received antibiotics (ampicillin, metronidazole, neomycin, gentamicin (1 g/L) and vancomycin (0.5 g/L)) in drinking water for 7 days.²¹ Antibiotics were removed and mice received FMT from a separate cohort of air-exposed or CS-exposed donors four times in 10 days. FMT recipient mice were not exposed to CS directly. For diet studies, mice were fed a conventional semi-pure diet (AIN93G) or a resistant starch diet (SF11-025; Specialty Feeds, WA, Australia) ad libitum commencing 2-week prior to CS-exposure and maintained until the end of experiment. Faeces were collected weekly, with at least three fresh faecal pellets collected and stored at –80°C until processing. Airway inflammation, lung and colon histopathology, lung function and gene expression analyses were assessed as previously described.^{7 12–20 22} Detailed methods are provided in the online supplemental file 1.

All experiments were approved by University of Newcastle Animal Ethics Committee.

Analysis of gut microbiome

Faecal microbiome composition, including metagenomics and 16S rRNA gene sequencing, were analysed as previously described⁸ with detailed methods provided in the online supplemental file 1. Briefly, α -diversity was calculated using QIIME V.1.8.0,²³ principal component analysis performed using the R package vegan V.2.5-1²⁴ within metagenomeSeq V.1.22.0²⁵ and differential abundance determined using DESeq2 with Benjamini-Hochberg adjustment. Sparse Partial Least-Squares Discriminant Analysis (sPLS-DA) was conducted using the R package mixOmics V.6.3.2.²⁶ Host phenotypes were tested for association with microbiome composition using the envfit function within the vegan R package. Spearman's rho was calculated using the 'corr.test' function within the R package psych V.1.8.12.²⁷ Non-random relationships between taxa were assessed using the CoNET Cytoscape application.²⁸

Cytometry by time-of-flight analysis

For time-of-flight mass cytometry, single cell suspensions of bone marrow, blood and spleen cells were stained with cell cycle marker iododeoxyuridine (IdU), viability marker cisplatin and surface and intracellular antibodies (online supplemental table S19). Expression levels of 38 markers were measured using a Helios instrument (Fluidigm), transformed with a logicle transformation²⁹ and used for Uniform Manifold Approximation and Projection for Dimension Reduction (UMAP) using the naïve R implementation (umap V.0.2.7). Cells clusters were calculated using Rphenograph (Github JinmiaoChenLab V.0.99.1).³⁰ Detailed methods are provided in the online supplemental file 1.

Cell culture

Raw 264.7 monocytes (1×10^6) were seeded into 12-well plates. Cells were incubated in media alone, or 1:1000 dilutions of sterile-filtered faecal homogenate (100 mg/mL) from CS-exposed or air-exposed mice for 24 hours. During the final 4 hours, LPS (1 µg/mL), monophosphoryl lipid A (4 µg/mL), lipoteichoic acid (1 µg/mL) were added to cell media. Tumour necrosis factor (TNF)-α protein in culture media was assessed using DuoSet ELISA kits (R&D Systems).

Flow cytometry

Colon²¹ and lung tissue^{13 18} were digested to produce single cell suspensions and stained for flow cytometry as previously described, before analysis using an LSRFortessa cytometer (BD Biosciences) and FlowJo software (TreeStar). Details are provided in the online supplemental file 1.

Proteomics of mouse faeces

Faecal proteins were analysed by nanoflow liquid chromatography (Ultimate 3000 RSLCnano, Thermo Scientific) with peptides introduced via an Easy-Spray Nano source coupled to a Q-Exactive Plus Quadrupole Orbitrap mass spectrometer (Thermo Scientific) and subjected to data dependent tandem mass spectrometry (MS/MS). Raw files were processed in Proteome Discoverer V2.1 using the Sequest HT algorithm,³¹ searched against the mouse gut microbiota GigaDB database³² and integrated with metagenomics results using the psych package in R³³ and Diablo mixOmics package. Detailed methods are provided in the online supplemental file 1.

Metabolomics

Metabolomics of caecum contents was performed by Metabolon (Durham, North Carolina, USA) using the Global HD4 MS platform as previously described.⁸ Detailed methods are provided in the online supplemental file 1.

Human interventional study

Sixteen patients with COPD were recruited from June 2019 to March 2020 at the Respiratory Investigation Unit at The Prince Charles Hospital (TPCH) with written and informed consent. Nine participants were randomised into the intervention arm (inulin 10 g daily for 4 weeks) and seven to the placebo group (maltodextrin 10 g daily for 4 weeks). Age, gender, body mass index (BMI), smoking history (pack years) and spirometry (forced expiratory volume in 1 s/forced vital capacity (FEV1) and FEV1% predicted) were similar between the two groups (table 1).

Statistical analysis

Except where specified, data was analysed in GraphPad Prism V9.0 (San Diego, California, USA). Outliers were identified by Grubbs test, and excluded from analysis only if investigators had reported technical errors during sample collection/processing before analysis. Data was analysed by one-way analysis of variance (ANOVA) with Holm-Šidák's post hoc test or Student's t-test, or non-parametric equivalents, for data from mice. Data from the patient with COPD cohort was analysed by Mann-Whitney test (continuous variables) or χ^2 test (categorical variables).

Table 1 Patient characteristics

	Placebo (n=7)		Inulin (n=9)		P value
	Mean	SD	Mean	SD	
Age	69.71	6.50	72.78	6.46	0.18
Gender (male, %)	42.86	N/A	66.67	N/A	0.34
Body mass index	29.68	6.06	28.66	4.85	0.71
Smoking pack years	56.21	20.02	49.73	24.94	0.6
FEV1/FVC	0.48	0.09	0.48	0.13	0.95
FEV1%	55.43	18.04	54.56	16.18	0.98
Freq. exacerbators (n, %)	5 (71)	N/A	3 (33)	N/A	0.13
Medications (n, %)					
Inhaled corticosteroids	5 (71)	N/A	7 (77)	N/A	0.77
Beta agonists	7 (100)	N/A	9 (100)	N/A	>0.99
LAMA	1 (14)	N/A	2 (22)	N/A	0.69
Statin	2 (28)	N/A	6 (67)	N/A	0.13
PPI	4 (57)	N/A	3 (33)	N/A	0.34
ACE	2 (28)	N/A	2 (22)	N/A	0.77
Beta blockers	1 (14)	N/A	1 (11)	N/A	0.85
Angiotensin II receptor antagonist	2 (28)	N/A	1 (11)	N/A	0.37
SSRI/SNRI	2 (28)	N/A	1 (11)	N/A	0.37
Season during intervention (n, %)					
Spring/summer	2 (28)	N/A	3 (33)	N/A	0.84

ACE, angiotensin-converting enzyme inhibitor; FEV1, forced expiratory volume in 1 s; FVC, forced vital capacity; LAMA, long-acting muscarinic antagonists; PPI, proton pump inhibitors; SNRI, serotonin–norepinephrine reuptake inhibitor; SSRI, selective serotonin reuptake inhibitors.

RESULTS

FMT alleviated dysbiosis and disease features in experimental COPD

To define changes in microbiome composition and the effects of FMT, C57BL/6 mice underwent microbiome normalisation to control for variability in starting microbiome composition before exposure to mainstream CS through the nose only for 12 weeks (12 wk CS),^{12–15 18–20} with a subset of mice modelling smoking cessation with 8 weeks of CS-exposure followed by 4 weeks of rest (8 wk CS+4 wk rest). Mice were treated via passive FMT through transfer of soiled bedding from air-exposed mice to CS-exposed mice and vice versa, with controls maintained in their own bedding.

Shotgun metagenomics of faecal samples collected prior to interventions (week 0) and at completion (week 12) recovered 74 metagenome-assembled genomes (MAG)>80% complete (dereplicated at 95% identity) representing 12 families (online supplemental tables 1–4). Microbiome composition was assessed with public genomes within Genome Taxonomy Database release 06-RS202.³⁴ Metagenomic sequencing did not yield adequate sequencing coverage for further investigation of non-bacterial components, with viral signatures detected in <15% of samples and fungal signatures not identified in any samples.

All experimental groups experienced a shift in microbiome composition associated with the maturation of the mice between week 0 and 12 (from 6 to 18 weeks old), with increased α-diversity, enrichment of species belonging to the genus *Prevotella* and family *Muribaculaceae* and depletion of *Akkermansia* and

Duncaniella (a member of *Muribaculaceae*) species (online supplemental figure S2A–C, online supplemental table S5).

While different experimental groups had similar α -diversity at week 12, microbiome composition differed significantly (online supplemental figure S2D,E $p=0.01$, PERMANOVA of Bray–Curtis distances). CS-exposure was associated with increased *Phocaeicola vulgatus*, *Muribaculaceae* species *Duncaniella* sp001689575 and *Amulumurptor* sp001689515, *Desulfovibrionaceae* species *Mailhella* sp003512875 and *Akkermansia muciniphila* and decreased *Lachnospiraceae* species *Muricomes* sp001517425 (figure 1A,B, online supplemental tables S6, S7). CS-exposed FMT mice maintained increased abundance of most CS-associated species compared with air-exposed mice, except for the *Lachnospiraceae* member UBA3282 sp009774575 (MAG FTS36, online supplemental table 6). This species, along with CS-associated *Mailhella* sp003512875 (MAG FTS70), was identified in a multivariate analysis-derived signature distinguishing 12-week CS-exposed mice from both air-exposed and 12-week CS-exposed FMT mice (figure 1C,D, online supplemental table S8). Both species were increased with CS-exposure and decreased with FMT, consistent with a role in disease (figure 1E). Other species contributing to the FMT-associated signal were decreased with CS-exposure and increased with FMT, including obligate anaerobes *Duncaniella dubosii*, *P. goldsteinii*, *Muribaculum intestinale* and *Mucispirillum schaedleri* (figure 1D, online supplemental table S8). Similar evidence of transfer of anaerobic species to air FMT mice (online supplemental table S6) indicates that bedding swaps were an effective method of FMT, including of anaerobes.

CS-exposure induced lung inflammation characterised by increased total leucocytes, neutrophils and macrophages in bronchoalveolar lavage fluid (BALF) and immune cells in parenchyma (figure 1G,H), as well as emphysematous alveolar destruction (figure 1I,J). BALF macrophages remained elevated after smoking cessation at levels comparable to mice exposed to CS for 12 weeks, while other measures of inflammation and alveolar destruction were present but had reduced severity. Impaired lung function, with increased compliance, volume and total lung capacity, was observed after 12 weeks CS-exposure (figure 1K,M).

FMT-treated mice had significantly lower BALF and parenchymal inflammation in mice exposed to CS for 12 weeks, or with smoking cessation (figure 1E–H). Unlike smoking cessation, FMT reduced BALF macrophages and the combination of smoking cessation and FMT had additive effects by further reducing total leucocytes and parenchymal inflammation. Most importantly, FMT alleviated both emphysema and impaired lung function after CS for 12 weeks (figure 1I–M).

Thus, FMT alleviated hallmark features of COPD and improved the resolution of chronic inflammation with smoking cessation.

FMT alters systemic manifestations of COPD

Given our evidence of CS-induced systemic comorbidities,^{7 12} we assessed gut manifestations of COPD and systemic leucocyte populations in mice exposed to CS for 12 weeks, with and without FMT. CS-induced increases in colonic submucosal fragmentation and vascularisation, and colon tissue messenger RNA expression of the microbe-sensing pattern recognition receptors *Tlr3* and *Tlr4* was prevented by FMT, with similar trends for *Tlr2* (figure 2A–E). No significant differences were observed in expression of *Tlr9* (figure 2F).

We employed deep immune profiling using cytometry by time-of-flight with 38 immune cell markers in the bone marrow, blood and spleen. Dimensionality reduction and clustering of bone marrow cells identified 20 clusters with coherent protein expression, but few were altered (figure 3A–C). CS-exposure reduced the abundance of B cells, likely driven by a downregulation of proliferating (IdU⁺) B cells, but these effects were not alleviated by FMT (figure 3D). In blood, 25 clusters were identified (figure 3E–G). B cells were reduced by CS-exposure and partially restored by FMT (figure 3H). CS-exposure also increased blood Ly6C^{lo} monocytes, and FMT restored their abundance to levels in air-exposed mice (figure 3H). In the spleen, 23 clusters were identified (figure 3I–K) with CD8⁺ conventional dendritic cells (cDCs) and B cells reduced while CD107⁺ progenitors and Ly6C^{lo} monocytes were increased by CS-exposure (figure 3L). FMT increased the abundance of CD8⁺ cDCs, CD107⁺ progenitors and Ly6C^{lo} monocytes, partially restoring the CS-induced depletion of CD8a⁺ cDCs but enhancing the impact of CS-induced increases in CD107⁺ progenitors and Ly6C^{lo} monocytes.

Thus, FMT alleviated CS-induced colonic pathology and TLR expression, as well as CS-induced depletion of blood B cells and splenic CD8⁺ cDCs and increases in blood Ly6C^{lo} monocytes.

Correlations of lung and gut pathology with microbiota

We assessed the role of the gut microbiome in pathogenesis by fitting phenotypic data to a β -diversity ordination analysis, identifying significant associations with lymphocytes and emphysema consistent with CS-induced separation in microbiome composition (figure 4A). When smoking cessation groups were excluded to assess associations in more severe disease, total leucocytes and neutrophils were also associated with microbiome composition (figure 4B). Correlations between individual species and phenotypes identified positive correlations between *Mailhella* sp003512875 (FTS70) and inflammation, emphysema, colon vascularisation and *Tlr4* expression (figure 4C, online supplemental table S9). *Muribaculaceae* species *Amulumurptor* sp001689515 and *A. muciniphila* positively correlated with lymphocytes and emphysema while *A. muciniphila* correlated with lymphocytes and colon *Tlr3* expression. *M. intestinale* and *Muribaculaceae* species UBA7173 sp001689685 were negatively correlated with inflammation, emphysema and colon *Tlr4* expression indicating a protective role.

CS-induced changes in microbiota impair responsiveness to TLR4 agonists

Given that CS-associated taxa correlated with colonic TLR expression, we assessed the effects of CS-induced dysbiosis on immune responses *in vitro*. RAW264.7 mouse monocytes were cultured for 24 hours in media alone, or in sterile-filtered homogenates of faeces from mice exposed to CS or air for 12 weeks. During the final 4 hours of incubation, TLR4 agonists LPS and monophosphoryl lipid A (MPLA), or the TLR2 agonist lipoteichoic acid (LTA) were added to directly stimulate TLRs. Faeces from either CS-exposed or air-exposed mice induced TNF- α production compared with media alone, but this was significantly lower with faeces from CS-exposed mice (online supplemental figure S3). LPS and MPLA induced TNF- α production at comparable levels in cells with media or faeces from air-exposed mice, but faeces from CS-exposed mice suppressed LPS-induced and MPLA-induced TNF- α . There were no significant differences in LTA-induced TNF- α production, suggesting that the impairment may be specific to TLR4.

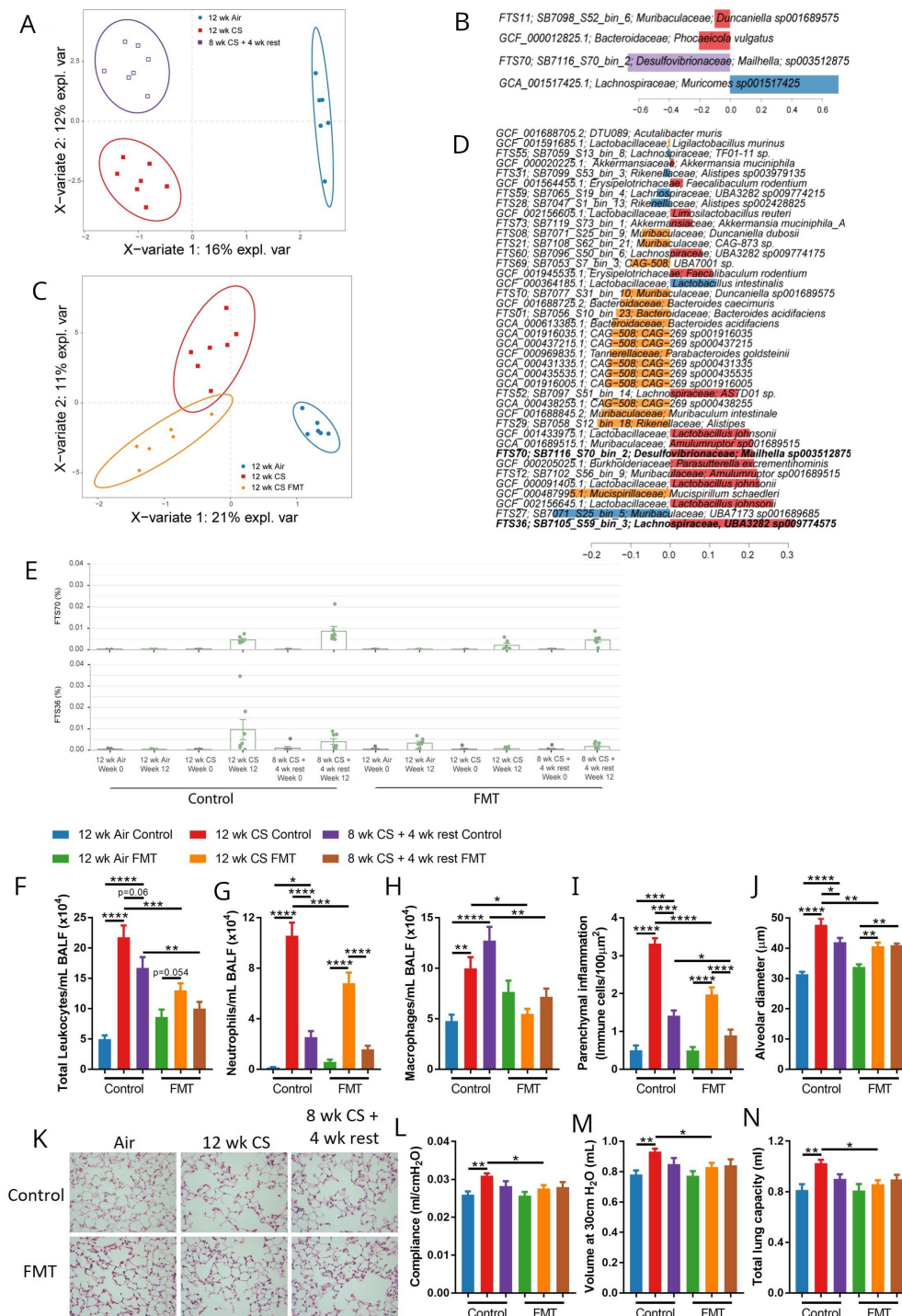


Figure 1 Faecal microbiota transfer (FMT) alleviated dysbiosis and disease features in severe experimental COPD and smoking cessation. Mice were exposed to CS or normal air for 12 weeks, or 8 weeks CS followed by 4 weeks with normal air (8 wk CS+4 wk rest). Mice also received FMT through transfer of soiled bedding or were maintained in their own bedding (Control) for 12 weeks. (A–D) Faecal samples were collected at the end of experiment (week 12) and analysed using shotgun metagenomics. (A) Multivariate analysis (sPLS-DA) of CS-exposed and air-exposed mice at week 12 demonstrating distinction between the groups along component 1 based on relative abundance at the genome level, centred log-ratio transformed and (B) species contributing to that distinction. (C) Multivariate analysis (sPLS-DA) of 12-week CS-exposed control and FMT treated mice, and air-exposed mice, demonstrating distinction between CS-exposed control and FMT mice along component 2 and (D) species contributing to that distinction. (E) Relative abundance of *Lachnospiraceae* FTS36 and *Mallihella* sp003512875 FTS70 were increased with CS-exposure, which were alleviated by FMT. (F–N) Hallmark features of COPD assessed at the completion of the experiment (week 12). FMT alleviated CS-induced increases in (F–H) total leucocytes, neutrophils and macrophages in bronchoalveolar lavage fluid (BALF) and (I) parenchymal inflammation in 12-week CS and 8-week CS+4 week rest mice. FMT also alleviated CS-induced increases in (J–K) emphysema-like alveolar enlargement, and (L–N) lung function parameters of compliance, volume and total lung capacity in 12-week CS mice. N=8 per group (A–N). Data presented as mean±SEM. *= $p<0.05$; **= $p<0.01$; ***= $p<0.001$; ****= $p<0.0001$ using one-way analysis of variance with Holm-Sidak's post hoc analysis (F–N). COPD, chronic obstructive pulmonary disease; CS, cigarette smoke; sPLS-DA, sparse Partial Least-Squares Discriminant Analysis.

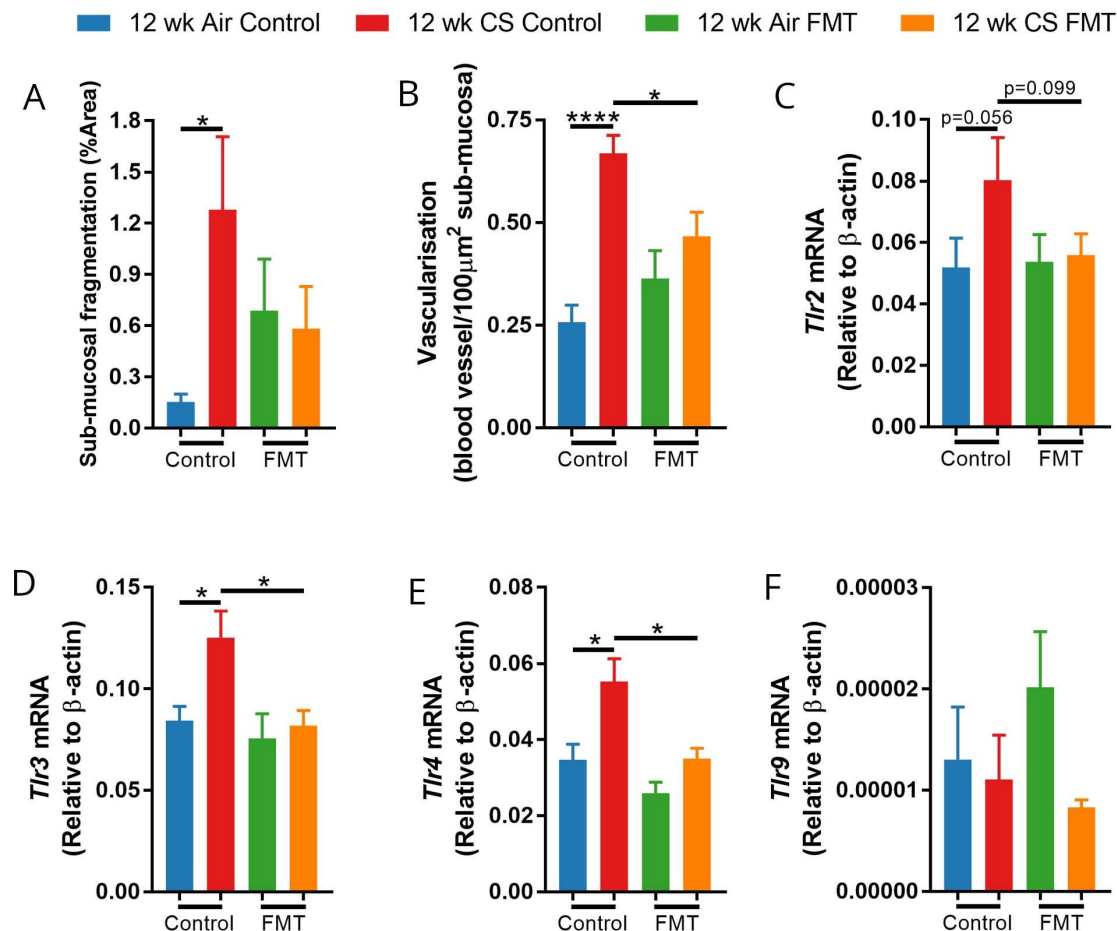


Figure 2 Faecal microbial transfer (FMT) alleviated cigarette smoke (CS)-induced colon histopathology and microbial sensor expression changes. (A–F) Mice were exposed to CS or normal air for 12 weeks, and received FMT through transfer of soiled bedding or were maintained in their own bedding (Control). FMT alleviated CS-induced increases in (A) submucosal fragmentation, (B) vascularisation and expression of (C) *Tlr2* (non-significant), (D) *Tlr3* and (E) *Tlr4* in colon tissue. (F) Expression of *Tlr9* was not altered by either CS or FMT. $N=7-8$ per group. Data presented as mean $\pm$ SEM. * $p < 0.05$ using one-way analysis of variance with Holm-Sidak's post hoc analysis. mRNA, messenger RNA.

CS-associated microbiota reduce colon innate immune cells but promote lung inflammation

Chronic CS exposure induces gastrointestinal inflammation,⁷ and the suppression of TLR4 and other innate responses by CS-associated microbiota may provide a competitive advantage in this environment. To further explore the impact of CS-associated microbiota on innate immunity *in vivo*, mice received antibiotics in drinking water for 7 days to deplete microbial communities. After antibiotics were removed, microbial communities were allowed to recover naturally or were recolonised through oral gavage of faecal homogenates from air or CS-exposed mice. Donors for FMT were mice from separate experiments, and experimental mice in this model received no direct CS exposure. Innate immune cell profiles in the colon and lung were assessed by flow cytometry after 10 days.

There were no changes in innate immune populations when antibiotic-depleted microbiota were recovered naturally or were recolonised with microbiota from air-exposed mice (figure 5A–G; online supplemental figure S4). However, recolonisation with faeces from CS-exposed mice reduced the number of colonic monocytes, immature macrophages (Ly6C+major histocompatibility complex (MHC)-II+) and mature macrophages (Ly6C-MHC-II+), including reduced slow-turnover (T-cell immunoglobulin and mucin domain containing (TIM)4-CD4+) and long-lived, self-renewing tissue resident

macrophages (TIM4+CD4+; figure 5A–E). Similarly, colonic neutrophils and eosinophils were also depleted in these mice (figure 5F,G). To identify the impact of gut microbiota on the gut-lung axis in the absence of direct CS exposure, lung granulocyte populations were also assessed. Recolonisation with faeces from CS-exposed mice promoted lung inflammation, increasing interstitial macrophages and eosinophils (figure 5H,I; online supplemental figure S5).

Thus, microbiota from CS-exposed mice directly exerts differing effects across the gut-lung axis, with suppressed innate immune responses in the colon but induction of macrophage-driven and eosinophil-driven inflammation in the lungs, even in the absence of CS exposure.

Mutual exclusion contributes to the protective effects of FMT

FMT reduced the abundance of *Lachnospiraceae* member *UBA3282 sp009774575* and *Maiella sp003512875* in CS-exposed mice (figure 1C,D, online supplemental table S8). While FMT-mediated modification of local and systemic immune responses may suppress pathogenic taxa, competition from bacteria introduced through FMT may also contribute to protective effects. To assess the role of competitive inhibition, we identified non-random patterns of co-occurrence in our metagenomics data set using the CoNET Cytoscape application.²⁸

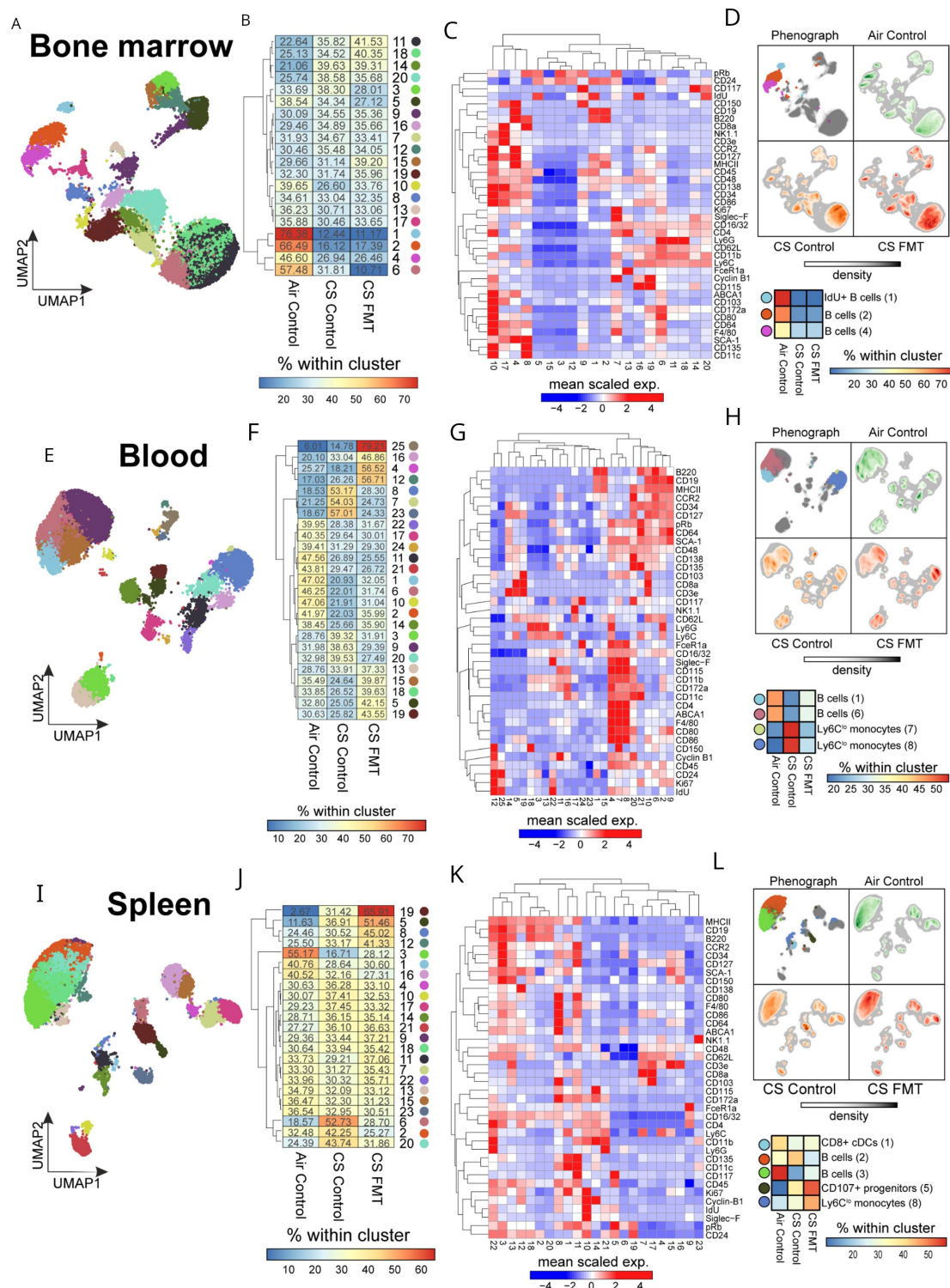


Figure 3 Cigarette smoke (CS)-exposure altered systemic leucocyte populations, and faecal microbial transfer (FMT) alleviated depletion of some blood and splenic populations. (A-I) Mice were exposed to CS or normal air for 12 weeks and received FMT through transfer of soiled bedding or were maintained in their own bedding (Control). Cell abundances were quantified using cytometry by time-of-flight. (A,B) Dimensionality reduction and clustering of bone marrow cells identified 20 clusters of cells (C) based on the scaled mean marker expression. (D) Uniform Manifold Approximation and Projection (UMAP) and a confusion matrix of the normalised frequencies of clusters demonstrated CS-induced depletion of B cells and iododeoxyuridine⁺ B cells which was not alleviated by FMT. (E,F) Dimensionality reduction and clustering of blood cells identified 20 clusters of cells (G) based on the scaled mean marker expression. (H) UMAP and a confusion matrix of the normalised frequencies of clusters demonstrated a CS-induced depletion of B cells and increase of Ly6C^{lo} monocytes, which were alleviated by FMT. (I,J) Dimensionality reduction and clustering of splenic cells identified 23 clusters of cells (K) based on the scaled mean marker expression. (L) UMAP and a confusion matrix of the normalised frequencies of clusters demonstrated CS-induced depletion of CD8⁺ conventional dendritic cells (cDCs) and B cells, and a CS-induced increase of CD107⁺ progenitors and Ly6C^{lo} monocytes. FMT increased the abundance of CD8⁺ cDCs, CD107⁺ progenitors and Ly6C^{lo} monocytes compared with CS control mice, partially reducing CD8⁺ cDCs but enhancing the CS-induced increases in CD107⁺ progenitors and Ly6C^{lo} monocytes. N=6 per group.

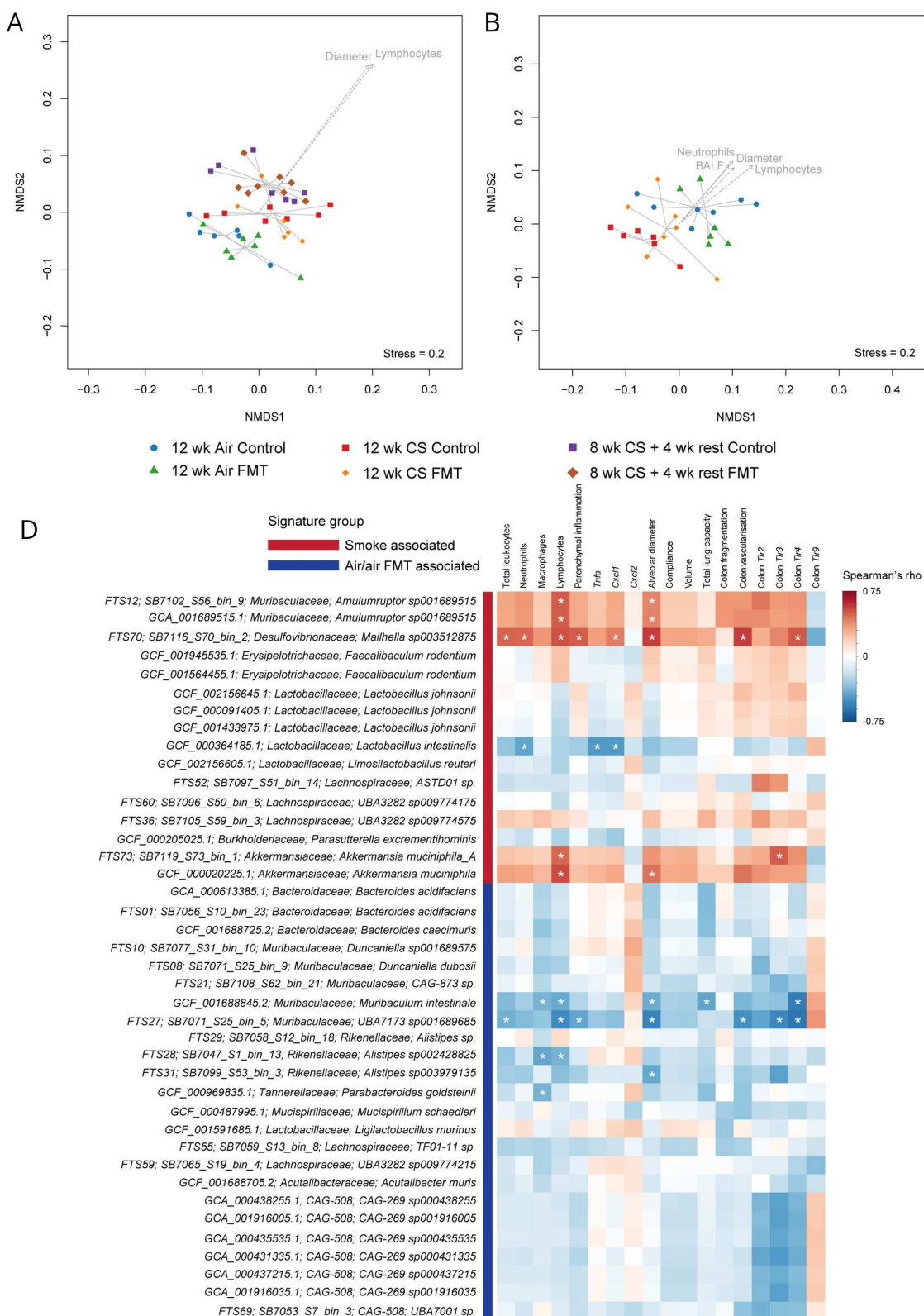


Figure 4 Cigarette smoke (CS)-associated bacterial species positively correlate with phenotypic measurements. (A–C) Mice were exposed to CS (12 wk CS) or normal air (Air) for 12 weeks, or 8 weeks CS followed by 4 weeks with normal air (8 wk CS+4 wk rest). Mice also received faecal microbiota transfer (FMT) of soiled bedding or were maintained in their own bedding (Control). (A, B) Non-metric multidimensional scaling (NMDS) ordination plot of Bray-Curtis distances with phenotypic variables significantly associated with CS and/or FMT fitted to axes. Lung compliance was excluded due to collinearity with volume. (A) Lymphocytes and alveolar diameter were significantly associated with microbiome composition (Benjamini-Hochberg, $p < 0.05$). (B) A similar analysis was performed excluding smoking cessation mice (8 wk+4 wk rest), which also identified total leukocytes in bronchoalveolar lavage (BALF) and neutrophils. (C) Spearman's correlation between phenotypic data and relative abundance of genomes separating CS-exposed mice (red bar) from air-exposed mice or CS-exposed mice receiving FMT (blue bar). Squares marked with an internal asterisk have $p < 0.05$, demonstrating key associations with bacteria. $N = 8$ per group.

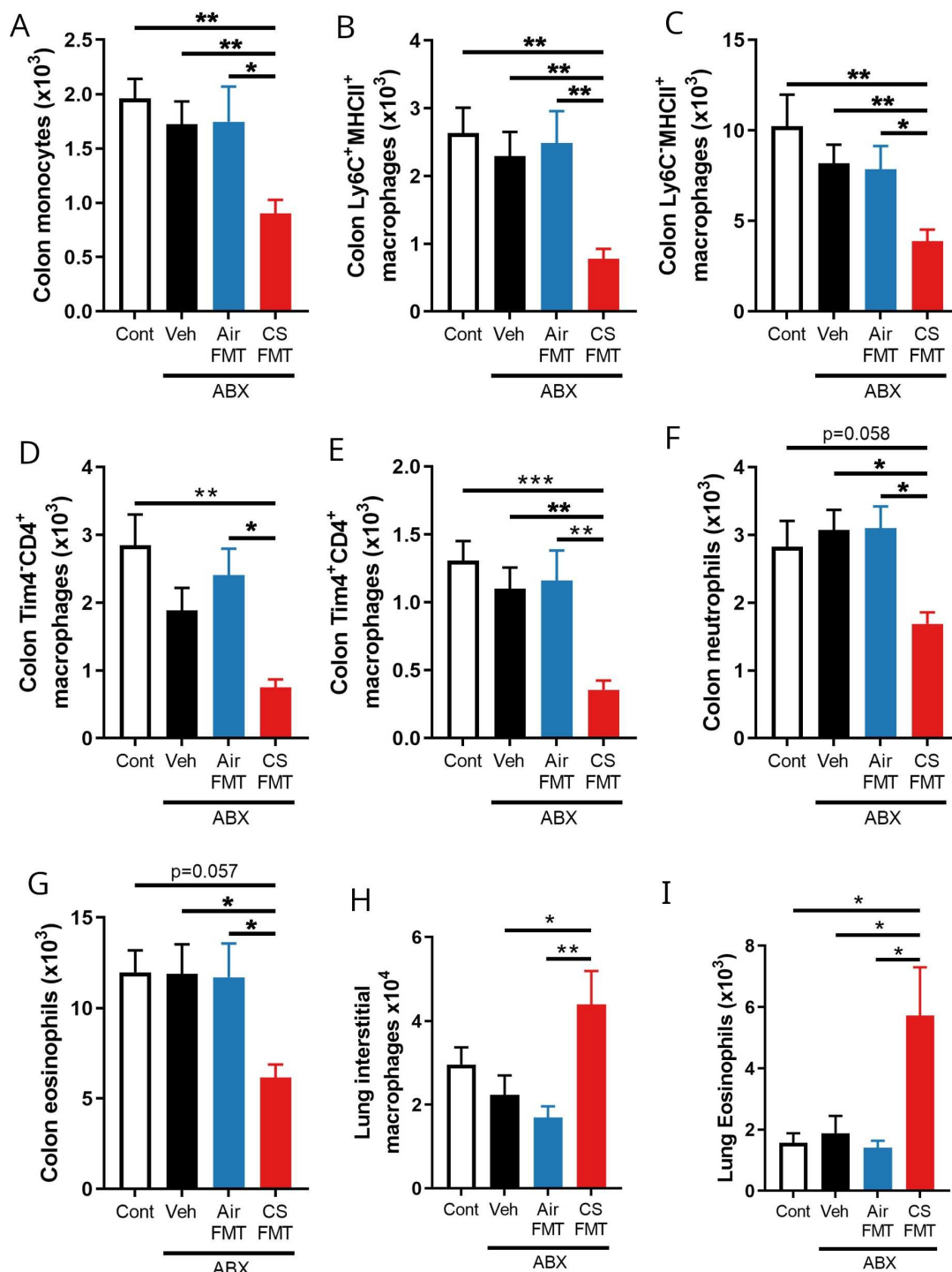


Figure 5 Innate immune cells are reduced in the colon but increased in the lungs after colonisation with microbiota from CS-exposed mice. Microbiota were depleted by antibiotics (ABX) for 7 days before mice were recolonised with faeces from separate air-exposed (Air FMT) or CS-exposed donors (CS FMT) for 10 days. A subset of mice received vehicle (Veh, phosphate-buffered saline+0.05% L-cysteine) to model natural recolonisation. Innate immune cells in the colon and lung were enumerated by flow cytometry. Colon (A) monocytes, (B) immature macrophages and (C) mature macrophages were reduced by CS FMT, including reduction (D) in slow turnover and (E) long-lived, tissue resident macrophages. (F) Colonic neutrophils and (G) eosinophils were also reduced by CS FMT. CS FMT increased (H) lung interstitial macrophages and (I) eosinophils. N=9–11 per group. Data presented as mean±SEM. *= $p<0.05$; **= $p<0.01$; ***= $p<0.001$; using one-way analysis of variance with Holm-Sidak's post hoc analysis. CS, cigarette smoke; FMT, faecal microbial transfer; MHC, major histocompatibility complex.

Of the 424 relationships identified, there were 308 co-presence and 116 mutual exclusion relationships (figure 6; online supplemental table S10). Most CS-associated taxa, including *P. vulgatus*, *Duncaniella* sp001689575, *Amulumpurptor* sp001689515, *Mai-bella* sp003512875 and *A. muciniphila* displayed co-present

relationships with other CS-associated taxa but no evidence of mutual exclusion with air-associated or FMT-associated taxa. Additionally, *Lachnospiraceae* member UBA3282 sp009774575 and air-associated *Muricomes* sp001517425 were not identified in any co-present or mutually exclusive relationships. However,

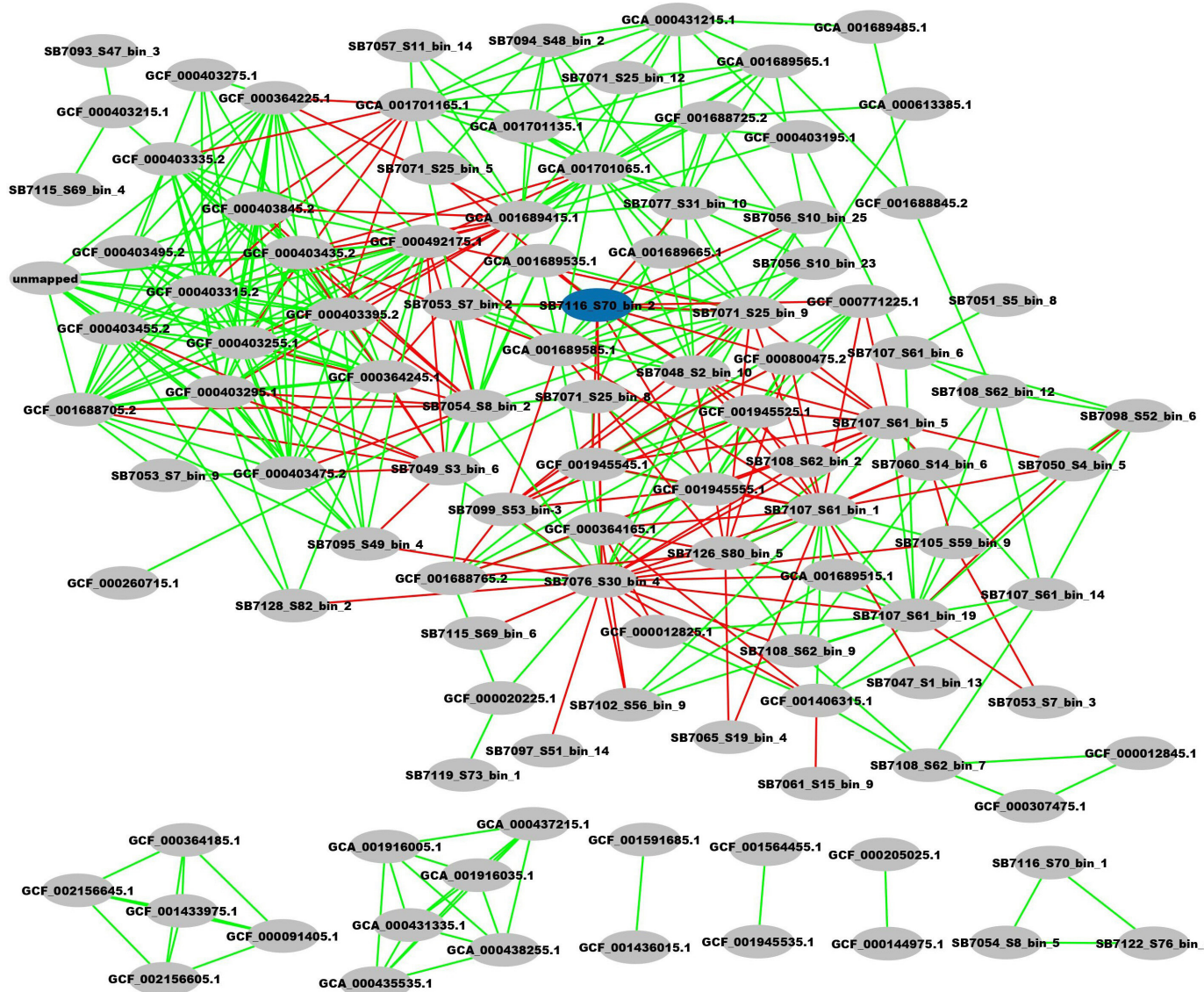


Figure 6 *Mailhella sp003512875* (blue ellipse; SB7116_S70_bin_2) has patterns of mutual exclusion with air/FMT-associated taxa. Mice were exposed to CS or normal air for 12 weeks, or 8 weeks CS followed by 4 weeks with normal air (8 wk CS+4 wk rest). Mice also received FMT through transfer of soiled bedding or were maintained in their own bedding (Control) for 12 weeks. Faecal samples were collected at the end of experiment (week 12) analysed using shotgun metagenomics, and non-random patterns of co-occurrence identified using the CoNET Cytoscape application. Patterns of co-occurrence (308, green lines) were more abundant than patterns of mutual exclusion (116, red lines). CS-associated taxa *Phocaeicola vulgatus*, *Duncaniella sp001689575*, *Amulumpurptor sp001689515*, *Mailhella sp003512875* and *Akkermansia muciniphila* displayed co-present relationships with other CS-associated taxa but no evidence of mutual exclusion with air-associated or FMT-associated taxa. *Lachnospiraceae* member *UBA3282 sp009774575* was not identified in any patterns of co-present or mutually exclusive relationships. However, *Mailhella sp003512875* had 15 non-random patterns indicating mutual exclusion, including with Air/FMT-associated taxa *Duncaniella sp001689575*, *Duncaniella dubosii* and *UBA7173 sp001689685*. N=48 (eight samples per group from six experimental groups).

15 non-random patterns indicating mutual exclusion (13% of all mutual exclusions observed) were identified with *Mailhella sp003512875*, including with Air/FMT-associated taxa *Duncaniella sp001689575*, *D. dubosii* and *UBA7173 sp001689685*. Thus, while not explaining all protective effects, FMT introduced taxa with which mutual exclusion relationships to *Mailhella sp003512875* were observed.

CS-associated taxa are not introduced during smoking

Bacteria have been identified in CS by both culture-dependent and independent methodologies.^{35 36} To determine whether CS-associated taxa were introduced to mice directly during CS-exposure, smoke from a 3R4F research cigarette was bubbled

through sterile phosphate-buffered saline (PBS). This process was repeated independently with four different cigarettes and independent negative controls. Aliquots were subjected to DNA extraction and analysis by 16S rRNA gene sequencing. After quality filtering and adapter trimming, reads from both CS extract (CSE) and PBS samples were exceptionally low (115.8 ± 20.22 reads vs 329.8 ± 503.6 reads). A total of 55 amplicon sequencing variants (ASVs) were identified, none of which were present in >1 CSE sample while absent from PBS. Due to the low biomass, ASVs were collapsed to genera/family level and relative abundances were broadly similar between CSE and PBS samples, although there was significant inter-sample variability (online supplemental figure S6A,B). Notably, there was no

evidence that CS-associated taxa (eg, *Duncaniella* sp001689575 and *Amulumpurator* sp001689515, *Mailhella* sp003512875, *A. muciniphila* and UBA3282 sp009774575) were enriched in CSE samples. CSE were also cultured on Yeast Casitone Fatty Acids agar under anaerobic conditions,³⁷ but no colonies were observed after 48 hours.

Experimental COPD and FMT were consistently associated with bacterial taxa

To validate our findings and explore temporal changes in the microbiome up until disease onset, mice were exposed to CS for 8 weeks with passive FMT through transfer of soiled bedding. Changes in gut microbiota were profiled weekly using 16S rRNA gene sequencing of fresh faecal samples collected directly from each mouse for the duration of the experiment (online supplemental tables 11–12). Microbiota composition in different groups were distinguishable from week 1 and continued to transition over time (online supplemental figure S7A–C). Multivariate analysis using sPLS-DA revealed CS-associated separation at week 8 (online supplemental figure S7D) driven primarily by five ASVs, including two *Muribaculaceae* family members which increased in CS-exposed mice (without FMT) from week 3 but only increased in CS-exposed mice receiving FMT after 6 weeks and decreased in subsequent weeks (online supplemental figure S7E,F).

We also explored microbiome changes in gastrointestinal tissues collected at week 8 and observed similar CS-associated separation in colons, driven by the previously identified *Muribaculaceae* ASVs plus *Desulfovibrionaceae* and *Lachnospiraceae* species (online supplemental figure S8A,D). CS-associated separation was less clear in the caecum (online supplemental figure S8B,E) and was not evident in the ileum (online supplemental figure S8C,F). However, ileum samples did reveal separation between CS-exposed FMT and CS-exposed control mice, with the abundance of distinguishing ASVs in the FMT group resembling air-exposed levels (online supplemental figure S8G,K).

Phenotypic measures were collected at the end of the experiment (week 8), with CS-induced induced inflammation, alveolar destruction and lung function impairment (online supplemental figure S7G–P). FMT blunted the CS-induced increase in BALF cells and completely suppressed changes in total lung capacity (online supplemental figure S7G–I,P). Similar phenotypic changes were observed with active FMT administered by oral gavage, confirming the involvement of gut microbiota (online supplemental figure S9A–F).

Relative abundance of CS-associated *Muribaculaceae* 91ca in faeces at the final time point (week 8) positively correlated with total leucocytes, emphysema and work of breathing (online supplemental figure S10A; online supplemental table S13). ASVs associated with non-smoking had a variable abundance pattern across the experiment and did not correlate with any phenotypic scores (online supplemental figure S10B–D).

Thus CS-induced experimental COPD is reproducibly associated with changes in gut microbiota, particularly a *Muribaculaceae* family member and FMT alleviated hallmark features of disease.

Multimics implicate downregulation of microbial glucose and starch metabolism in experimental COPD

To assess changes in microbial function, we characterised the faecal proteome (online supplemental table S14) and integrated the results with bacterial abundances. Correlations with individual MAGs did not reveal clear associations, but genera

CAG-307, *Staphylococcus*, *Bilophila*, *Aerococcus*, CAG-411, *Corynebacterium*, *Rikenella*, *Lactonifactor*, *Acetivibrio* and *Absiella* were most frequently correlated with the microbial proteome ($R \geq 0.7$; $p < 0.05$; figure 7A, online supplemental table S15).

Of these proteins correlated with bacterial genera, 12 were significantly altered in at least one experimental group (online supplemental table S16). Chaperone protein AELOFPMA_23737 and urease subunit alpha GFDOOBDM_36378 were increased by 12 weeks CS-exposure, but not in smoking cessation and were not alleviated by FMT (figure 7B,C).

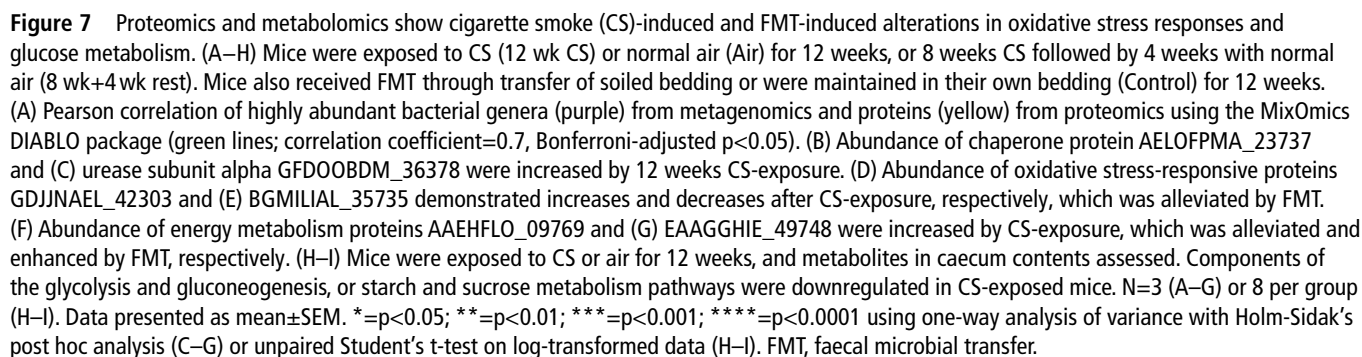
Four proteins were altered by both CS-exposure and FMT (figure 7D–G). Alkyl hydroperoxide reductase subunit C GDJJNAEL_42303, an antioxidant protein, was increased after 12 weeks of CS-exposure and alleviated by smoking cessation and FMT ($p = 0.052$) (figure 7D). Rubredoxin BGMILIAL_35735, also implicated in oxidative stress responses, had a reverse pattern and was decreased after 12 weeks of CS-exposure and after smoking cessation and increased by FMT (figure 7E). Phosphoglycerate kinase AAEHAFLO_09769 and malate dehydrogenase EAAGGHIE_49748, implicated in glucose metabolism and energy generation, were both increased by 12 weeks of CS-exposure and, to a lesser extent, after smoking cessation (figure 7F,G). While FMT alleviated the CS-induced increase in AAEHAFLO_09769, it further increased EAAGGHIE_49748.

Metabolite profiling in the caecum contents of a separate cohort of CS-exposed mice identified reduced abundance of microbiota-derived amino acid metabolites including tryptophan-derivatives indolelactate, indole-3-carboxylic acid, indoleacetate and indolepropionate, phenylalanine derivative phenyllactate and isoleucine-derivative 2-hydroxy-3-methylvalerate (online supplemental figure S11A–F). Isovalerate, produced by microbial fermentation of leucine, was increased (online supplemental figure S11G). Host-derived primary bile acids were not altered by CS (online supplemental figure S12A–E) but secondary bile acids, produced from primary bile acids by microbial biotransformation, were significantly elevated (online supplemental figure S8F–K). Moreover, glycolysis, gluconeogenesis, starch and sucrose metabolism components were downregulated (figure 7H–L) which, while not specific for microbiota, were consistent with changes observed in faecal proteomics. Thus, metabolomics and proteomics implicated a downregulation of glucose and starch metabolism in CS-exposed mice, which was altered by FMT.

Dietary resistant starch alleviated inflammation and emphysema in experimental COPD

Given the changes in microbial glucose and starch metabolism, we assessed whether dietary supplementation with complex carbohydrates improved disease outcomes. Mice were exposed to CS (8 weeks), and fed either control diet or an equivalent diet with all carbohydrates as resistant starch (high amylose maize starch) which is not digested by the host, increasing the availability to microbiota.³⁸ Resistant starch alleviated CS-induced airway inflammation and prevented CS-induced increases in alveolar diameter (figure 8A–E), consistent with the protective effects of FMT.

At the experimental endpoint, faecal microbiota were profiled using 16S rRNA gene sequencing. Resistant starch reduced α -diversity in both air-exposed and CS-exposed



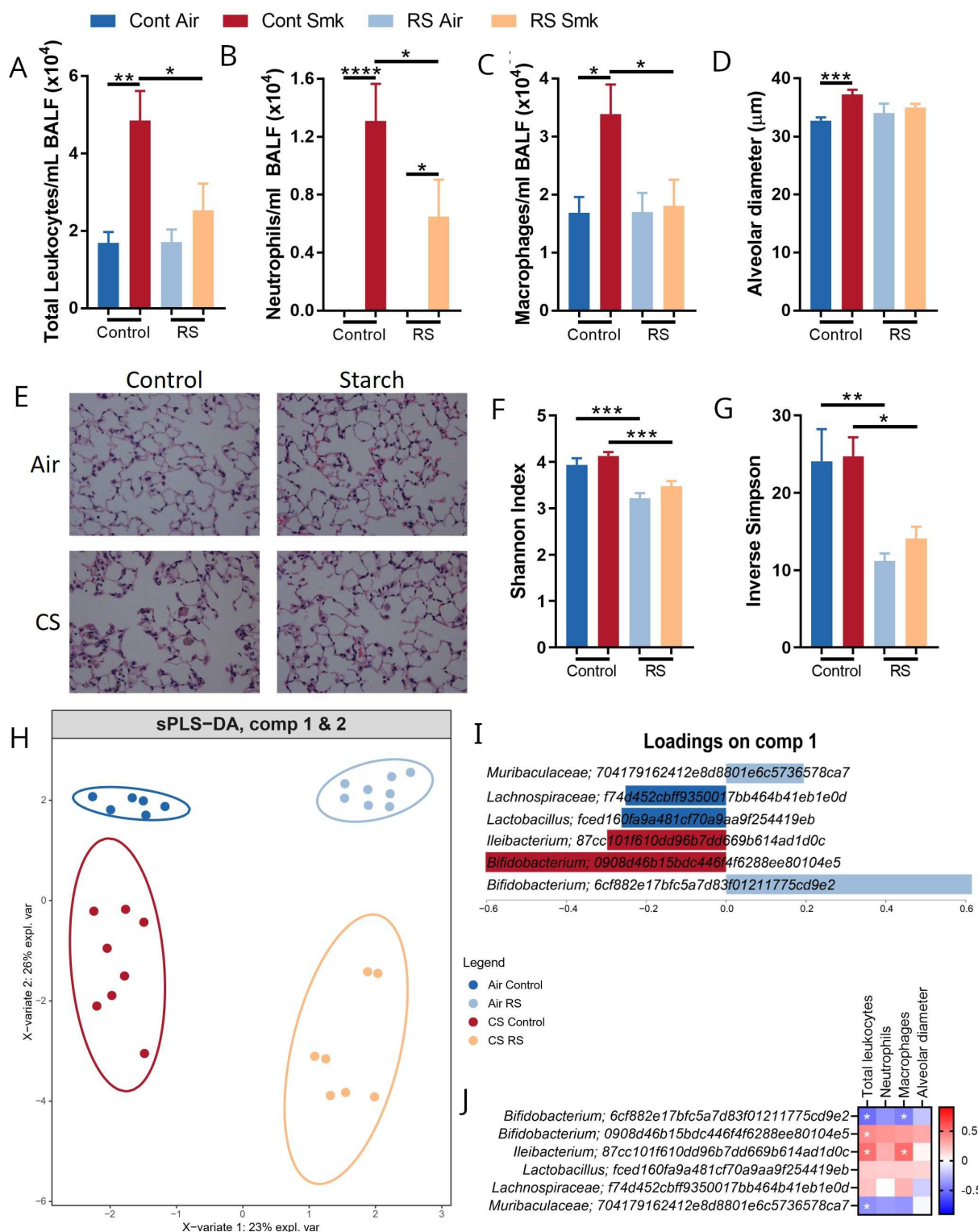


Figure 8 High resistant starch intake alleviated cigarette smoke (CS)-induced inflammation and emphysema. (A–D) Mice were exposed to CS or normal air for 8 weeks. Beginning 2 weeks prior to exposure, mice were fed a control or resistant starch (RS) diet. RS diet alleviated CS-induced increases in (A) total leukocytes, (B) neutrophils and (C) macrophages in bronchoalveolar lavage fluid (BALF) and prevented (D, E) CS-induced alveolar destruction. (E–J) Faecal microbiome was characterised by 16S rRNA gene sequencing. RS reduced α -diversity measure by (F) Shannon and (G) inverse Simpson index. (H) Multivariate analysis (sPLS-DA) indicated RS-induced and CS-induced separation of microbiome samples using PERMANOVA of Bray-Curtis distances, driven largely by (I) altered *Bifidobacterium* amplicon sequencing variants (ASVs). (J) ASVs contributing to separation were correlated with inflammation in BALF. N=6–14 per group. Data presented as mean $\pm$ SEM. * $p < 0.05$; ** $p < 0.01$; *** $p < 0.001$; **** $p < 0.0001$ using (A–G) one-way analysis of variance with Holm-Sidak's post hoc analysis or (J) Spearman's correlation. sPLS-DA, sparse Partial Least-Squares Discriminant Analysis.

Table 2 Exacerbation frequency, COPD Assessment Test (CATest) and St Georges Respiratory Questionnaire symptom scores (SGRQ) in patients with COPD receiving 4 weeks placebo (N=7) or inulin (N=9) supplement

	Placebo		Inulin		P value
	Mean	SD	Mean	SD	
Number of exacerbations	4 (57%)	N/A	1 (11%)	N/A	*0.048
Pretreatment CATest score	14.33	1.16	17.13	7.18	>0.99
Post-treatment CATest score	16.00	1.00	12.88	5.33	>0.99
Δ CATest score	1.67	2.08	-4.25	2.49	*0.01
Pretreatment SGRQ score	40.49	13.08	42.52	16.94	>0.99
Post-treatment SGRQ score	41.91	3.24	36.17	14.51	>0.99
Δ SGRQ total score	1.43	9.94	-6.34	6.07	0.27

*p < 0.05.

COPD, chronic obstructive pulmonary disease.

mice (figure 8F,G), and multivariate analysis using sPLS-DA demonstrated clear separation between all experimental groups (figure 8H; $p=0.01$, PERMANOVA of Bray-Curtis distances). This was driven primarily by a shift in *Bifidobacterium* ASVs, one increased by resistant starch and one decreased (figure 8I). Multiple ASVs correlated with disease features (figure 8J; online supplemental table S17) including the resistant starch-associated *Bifidobacterium* which was negatively correlated with total leucocytes and macrophages in BALF.

Inulin supplementation reduced symptoms and exacerbations in human COPD

To explore whether these findings translated into humans, we performed a small, preliminary pilot study with a randomised, double-blind, placebo-controlled study in patients with COPD receiving supplements of inulin, a common fermentable fibre, for 4 weeks. There were no differences between placebo-treated (N=7) and inulin-treated (N=9) patients in age, gender, BMI, smoking history, lung function, proportion of frequent exacerbators (≥ 2 exacerbations in 12 months), medication use or seasonality of intervention (table 1). Oral inulin resulted in fewer self-reported exacerbations (worsening or respiratory symptoms requiring additional pharmaceutical intervention) during the intervention period (1/9 patients, 11%) than placebo (4/7 patients, 57%) (table 2) and improved health-related quality of life measured by two validated instruments; the COPD Assessment Test (CATest) and the St Georges Respiratory Questionnaire (SGRQ). Both CATest and SGRQ scores were lower (improved symptoms) after inulin treatment than at baseline, exceeding the Minimally Important Clinical Difference.³⁹ There was no reduction in either score after placebo (table 2), but only the change in CATest score was significantly different in inulin-treated patients compared with placebo.

Faecal microbiota composition before and after receiving inulin was assessed by 16S rRNA gene sequencing. Although no differences in α -diversity were observed (figure 9A,B), microbiome composition differed significantly after inulin consumption, primarily driven by increased abundance of *Bacteroides* (figure 9C,D $p=0.01$, PERMANOVA of Bray-Curtis distances).

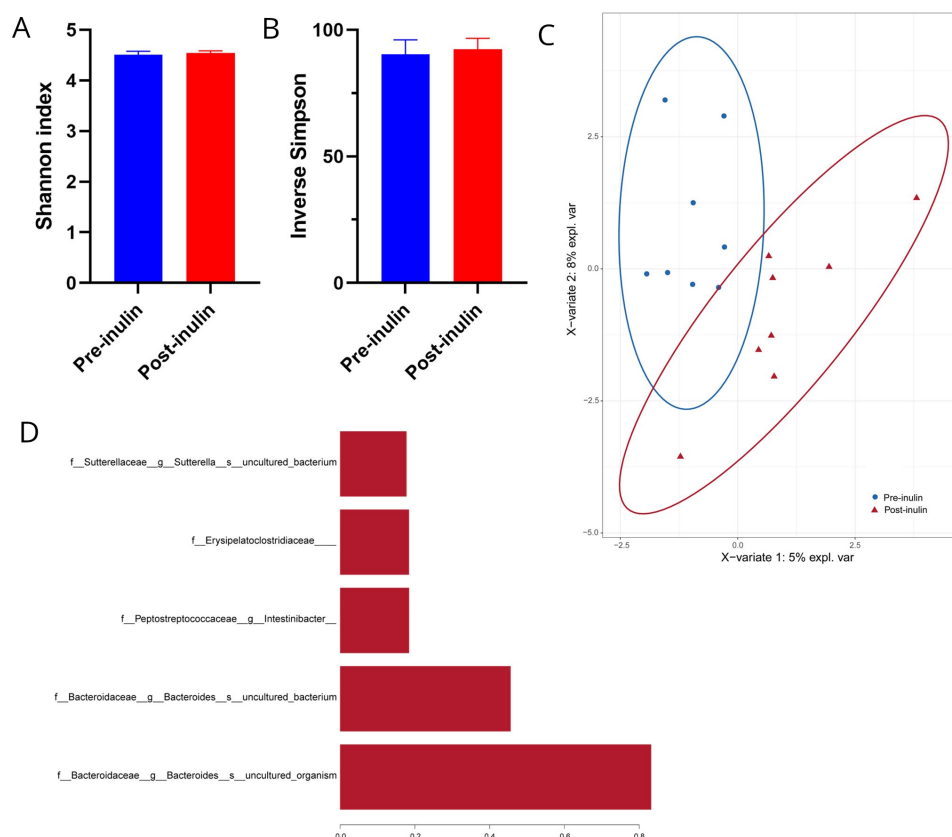


Figure 9 Inulin supplementation altered faecal microbiota in patients with COPD. (A–D) Patients with COPD received an inulin supplement for 28 days. Faecal microbiome composition was assessed before (pre-inulin) and after (post-inulin) inulin supplementation by 16S rRNA amplicon sequencing. (A–B) Shannon and Simpson diversity indices were not altered by inulin supplementation. (C) Multivariate analysis (sparse Partial Least-Squares Discriminant Analysis) demonstrating distinction between pre-inulin and post-inulin samples using PERMANOVA of Bray-Curtis distances and (D) species contributing to that distinction. COPD, chronic obstructive pulmonary disease.

DISCUSSION

CS-induced changes were broadly similar across several experiments, with *Muribaculaceae* *Desulfovibrionaceae* and *Lachnospiraceae* members increased after 8 and 12 weeks and correlating with disease phenotypes. FMT only partially alleviated CS-induced changes in microbiome composition, likely due to continued selective pressure with continuous smoke exposure or incomplete microbiota replacement using the bedding swap FMT. Bedding swaps, while not the most controlled method of FMT,⁴⁰ were used in most experiments as repeated oral gavage affects histopathology and lung function in mice.⁴¹ Nevertheless, FMT reproducibly alleviated inflammation, emphysema, impaired lung function and gastrointestinal pathology and analysis of mice receiving FMT demonstrated effective transfer of bacteria, including obligate anaerobes such as *D. dubosii*, *P. goldsteinii*, *M. intestinale* and *M. schaedleri*. Further, our initial findings were validated by FMT using oral gavage in a smoking model, and transfer of faeces from smoking mice into naïve mice after antibiotic-induced microbiome depletion increased lung inflammation, confirming that gut microbiota have substantial effects on disease. Minor differences in microbiome changes between studies were expected due to natural variability,⁴² and cage/batch effects were controlled by microbiome normalisation. Together, these findings demonstrate effective alleviation of disease by FMT and associations with bacterial taxa across multiple independent experiments.

Reproducibility between experimental models of COPD has proven challenging,^{10 11} as has the correspondence between changes observed in murine models and samples from human patients with COPD. Australian patients with COPD had reduced abundance of *Desulfovibrionaceae* member *Desulfovibrio piger* in faeces, but increased *Lachnospiraceae* species, including *Sellimonas intestinalis*, that negatively correlated with lung function.⁸ Chinese mild COPD cohorts had higher abundance of *Prevotellaceae* while patients with severe disease had lower abundance of *Bacteroidaceae* and *Fusobacteriaceae*,⁴³ and identified that *Bacteroides* sp CAG875, *Christensenella minuta* and *Clostridium* sp Marseille-P2538 were the most abundant species in faeces differentiating healthy donors from patients with COPD.⁴⁴ In a Korean cohort, patients with COPD were differentiated from asymptomatic smokers by 17 bacterial groups including lower *Prevotella copri*, *Dialister succinatiphilus* and *Catenibacterium mitsuokai* and increased *Bacteroides fragilis* and *Bifidobacterium longum*,⁴⁵ highlighting the challenges of identifying a consistent microbial marker of disease especially given the demographic differences between cohorts. Some studies in mice have identified that CS-exposure increased *Lachnospiraceae* abundance,¹¹ while others report decreases.¹⁰ Our new data agrees with that from our previously published Australian cohort and other animal models, identifying a role for *Lachnospiraceae* in disease, but is the first study to implicate *Muribaculaceae* species in patients with COPD or animal models. Although more predominant in the murine microbiome, this family is present in humans and may have roles in disease.⁸ Two *Muribaculaceae* species, *M. intestinale* and UBA7173 sp001689685, negatively correlated with inflammation and emphysema, highlighting the importance of assessing the microbiome at higher taxonomic resolution.¹⁰

The precise causes of dysbiosis in COPD have not been defined. Several bacteria have been identified in CS by both sequencing and culture-based techniques, including *Acinetobacter*, *Bacillus*, *Burkholderia*, *Clostridium*, *Klebsiella*, *Pseudomonas*, *Serratia*, *Campylobacter*, *Enterococcus*, *Proteus*, *Staphylococcus* and *Pantoea* species.^{35 36} However, none of these taxa contribute to

the CS-associated microbiome in our model, and neither 16S rRNA sequencing nor culturing of CSE indicated the presence of any bacteria associated with CS-induced experimental COPD. Rather, dysbiosis is likely induced by products of CS such as nicotine and heavy metals that are present in circulation and accumulate in the gastrointestinal tract, directly influencing microbiota.^{46 47} Additionally, we demonstrated that CS-associated microbiota suppressed local innate immune responses, including blunting TLR4 responses *in vitro* and reducing colonic innate immune populations such as microbiome-dependent tissue resident macrophages *in vivo*.⁴⁸ We previously reported opposing roles for lung TLR4 (pathogenic) and TLR2 (protective) in COPD¹⁸ and provide further evidence here that their signalling in the colon impacts pathogenesis, which is consistent with the protective effects of *P. goldsteinii* mediated by reduced activation of TLR4.¹¹ Moreover, changes in TLR3, which recognises damage associated-molecular patterns, suggests an association with colonic pathology in COPD.⁴⁹ Given that long-term smoking is associated with colonic inflammation,⁷ the local immunosuppressive effects of CS-associated taxa may contribute to their persistence and expansion in these conditions, causing CS-induced dysbiosis while promoting lung inflammation that promotes disease progression.

In mice exposed to CS, FMT alleviated disturbances in systemic immunity including increased blood Ly6C^{lo} (non-classical) monocytes and depleted blood B cells and splenic CD8⁺ cDCs. Non-classical monocytes, associated with tissue repair or autoimmune damage,⁵⁰ are increased in severe COPD⁵¹ and thus the FMT-mediated reduction is likely associated with the suppression of alveolar destruction. CD8⁺ cDCs, particularly in the spleen, are implicated in immune tolerance, phagocytosis and cross-presenting antigens, suggesting an association with increased blood B cells in FMT-treated mice.⁵² Splenic cDC numbers are increased in models of acute lung injury,⁵³ demonstrating a spleen-lung axis regulated by gut microbiota. Restoration of systemic immune responses by FMT is consistent with evidence that FMT from air-exposed mice into naïve mice after antibiotic depletion did not deplete colonic immune cell populations as it did with FMT from CS-exposed mice. Restoration of colonic immune responses likely contributes to FMT-mediated reductions in *Lachnospiraceae* member UBA3282 sp009774575 and *Mailhella* sp003512875 in CS-exposed mice, but other mechanisms also contribute to protective effects including competitive relationships between bacteria. Most non-random relationships between bacteria were co-present relationships between groups of air-associated or CS-associated taxa, demonstrating that environmental conditions and exposures were the dominant influence on overall microbiome composition. However, multiple air/FMT-associated taxa had mutual exclusion relationships with *Mailhella* sp003512875, demonstrating that introduction of these species through FMT contributes to control of CS-associated species. Mutual exclusion with air-associated UBA7173 sp001689685 is particularly noteworthy, given it is negatively correlated with inflammation and emphysema and provides further evidence of protective effects mediated by mutual exclusion. Isolation and culture of these taxa will provide further insight into their relationship.

In the faecal proteome, increased chaperone protein DnaK (AELOFPMA_23737) and urease subunit alpha (GFDOOBDM_36378) after 12 weeks CS may be in response to heavy metals which rapidly reduce after smoking cessation.^{46 54 55} Hydrolysis of urea by urease increases nitrogen availability, which has been linked to the persistence of deleterious bacteria and inflammation.⁵⁵

Alkyl hydroperoxide reductase GDJJNAEL_42303 and rubrerythrin BGMILIAL_35735 deactivate hydrogen peroxide to permit bacterial survival.^{56 57} Increased alkyl hydroperoxide reductase in response to CS-induced oxidative stress³⁸ may explain the expansion of *Desulfovibrionaceae* that thrive in oxidative conditions,⁵⁸ and antioxidants are postulated to maintain the mucous barrier in patients with COPD.³⁸ That rubrerythrin was reduced by CS and increased by FMT indicates that the local environment influences oxidative stress responses and microbial adaptability. For example, rubrerythrin uses iron as a cofactor⁵⁶ whereas alkyl hydroperoxide reductase is repressed in high iron environments.⁵⁷ Many patients with COPD have iron deficiency³⁸ and lower hepcidin levels in CS-exposed mice increase absorption and reduce iron in the intestinal lumen,⁵⁹ and may contribute to the different oxidative stress responses and expansion of pathogenic taxa in COPD.

Phosphoglycerate kinase AAEHAFLO_09769 and malate dehydrogenase EAAGGHIE_49748 are involved in glycolysis and gluconeogenesis^{60 61} and their abundance is increased in nutrient restriction, including by *Lachnospiraceae* spp.^{60 61} Nicotine is an appetite suppressant⁶² and reduced food consumption alters intestinal microbiota and predisposes to *Streptococcus pneumoniae* infection,³⁸ a major cause of COPD exacerbations and lung dysbiosis.⁵

Reduced abundance of microbial amino acid metabolites and increased secondary bile acids may be relevant for disease pathogenesis, particularly given recent evidence of a protective effect of the tryptophan metabolite indole-3-acetic acid in the airways⁶³ and the increasingly recognised role of bile acids as regulators of mucosal immunity.⁶⁴ Metabolomics also confirmed the disruption of carbohydrate metabolism identified by proteomic analysis that was selected for further investigation. Complex carbohydrates like resistant starch and inulin are poorly digested by the host, improving availability to microbiota stimulating anti-inflammatory short chain fatty acids (SCFAs).^{63 68} High fibre intake is associated with lower risk of developing COPD,³⁸ and inulin reduced airway inflammation in asthma.⁶⁵ Resistant starch reduced inflammation and emphysema in experimental COPD, and our pilot study in patients with COPD indicated that even a short period of inulin supplementation may reduce exacerbations and symptoms. However, examination of exacerbation frequency during the short period of intervention and assessment (4 weeks) should be interpreted with caution. Trials in larger cohorts of patients with COPD, and over longer periods of time, are needed to definitively demonstrate alleviation of symptoms and exacerbations and to progress these findings into clinical practice. While inulin is an established prebiotic in humans and improves translatability for future studies,⁶⁵ it is a different fibre to the resistant starch used in mice and therefore affects different bacterial populations. Resistant starch typically enriches *Ruminococcus*, *Bifidobacterium*, *Faecalibacterium prausnitzii* and *Eubacterium rectale*. Our study indicated that resistant starch increased one *Bifidobacterium* ASV which was negatively correlated with inflammation in mice. Inulin increases *Bifidobacterium*, *Anaerostipes*, *Faecalibacterium* and *Lactobacillus*, although there is significant variability between individuals and studies due to the baseline metabolic capacity of microbiota.^{66 67} Inulin primarily increased *Bacteroides* in our human cohort, which may indicate different responses to inulin supplements in patients with COPD compared with other cohorts, but the low sample size and high variability in both microbiome composition and prebiotic responses limits interpretation. Nevertheless, the anti-inflammatory effects of resistant starch associated with *Bifidobacterium* here, evidence of

probiotic *B. longum* in CS-induced inflammation²² and the established bifidogenic activity of inulin suggests greater efficacy of symbiotic *Bifidobacterium*-inulin supplements in COPD. Careful consideration of differences in microbial populations between animal models and patients with COPD, as well as between different cohorts of patients with COPD, will be required for future investigations in larger cohorts.

Jang *et al* showed limited evidence for FMT alleviating whole-body CS-exposure/poly I:C-induced emphysema which were associated, but not correlated, with increased *Lactobacillaceae* and decreased *Bacteroidaceae* identified through 16S rRNA gene sequencing.¹⁰ Whole body CS-exposure coats the fur and grooming will unnaturally affect gut microbiota.⁶⁸ Without appropriate controls (no CS or poly I:C alone) and baseline microbiome normalisation or sampling, it was not possible to discern the different effects of CS and poly I:C, a TLR3 agonist, or determine whether changes in microbiota were due to confounding factors including cage effects,⁴² accelerated ageing in COPD and the maturation of the microbiota,¹ or the effects of PBS/saline vehicle gavage on gut microbiota, immune responses and lung histopathology.⁴¹ That study was not translatable, as it did not assess the most clinically relevant feature of COPD (impaired lung function) or assess efficacy in humans.¹⁰ The lack of human translation has been a limitation of other models of experimental COPD,¹¹ particularly given that experiments rarely consider the impacts in the context of smoking cessation, the current first-line treatment for COPD.

Our nose-only CS-exposures accurately reflects human exposures and disease mechanisms,⁴ and included microbiome normalisation, baseline microbiome assessment and demonstration of therapeutic benefits using FMT by both bedding swap and oral gavage to discriminate between disease-associated changes and those from confounding factors. Higher resolution metagenomic sequencing and correlation of individual taxa with disease features allowed species-level classification to elucidate the different, and sometimes contrasting, effects of family members such as CS-associated *Amuliumraptor sp001689515* and air/FMT-associated *M. intestinale* and *UBA7173 sp001689685* from the *Muribaculaceae* family. The efficacy of FMT was demonstrated in several contexts, including mild and severe disease, gastrointestinal comorbidities and providing additional benefits to smoking cessation by alleviating persistent increases in macrophages that drive progressive disease.¹² Translatability was enhanced by assessing lung function and extending our work in a pilot study to demonstrate clinically relevant alleviation of symptoms in patients. Overall, our study provides strong evidence for the therapeutic efficacy of FMT and complex carbohydrate supplementation which could be acceptable interventions for patients with COPD, that may become more effective when combined with smoking cessation.⁶⁹

A limitation of this and other studies is the lack of assessment of viral and fungal components of the microbiome in COPD. While their role in COPD has not been clarified, gastrointestinal fungi, especially *Candida*, have been implicated in lung cancer⁷⁰ and murine models of allergic airway disease along with changes in lung and gut innate immune cell populations.^{71 72} Fungi and viruses also activate TLRs, including recognition of β -glucans by TLR2, O-linked mannans by TLR4, fungal/viral RNA by TLR3 and 7 and DNA by TLR9,^{73 74} and thus may overlap with several mechanisms identified in this study. Neither viral nor fungal components could be confidently detected in our metagenomics data using read mapping to reference genomes or marker gene detection tools, likely due to the relatively shallow depth (~3 Gb).⁷⁵ Even PCR amplification of the fungal ITS2

region did not yield sufficient material for adequate sequencing coverage, indicating low fungal abundance in our samples. Insufficient material for ITS2 sequencing can commonly occur in ~30% of faecal samples from mice and humans,^{76 77} and faecal fungal abundance is lower in laboratory mice housed in clean conditions.⁷⁸ Future studies should consider the introduction of wild microbiota to laboratory mice to improve capacity for investigation of fungal communities, and use deeper sequencing methods to investigate both fungal and viral components of the microbiota.

CONCLUSIONS

We demonstrate that dysbiosis, including increased *Muribaculaceae*, *Desulfovibrionaceae* and *Lachnospiraceae* species, contributes to disease pathogenesis and FMT alleviates hallmark features of CS-induced COPD. These changes may be affected by microbial energy metabolism, and complex carbohydrate supplementation protects against experimental disease and reduced exacerbations and symptoms in humans. This work opens new avenues towards effective microbiome-targeting therapies in COPD.

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Supplementary Figures:

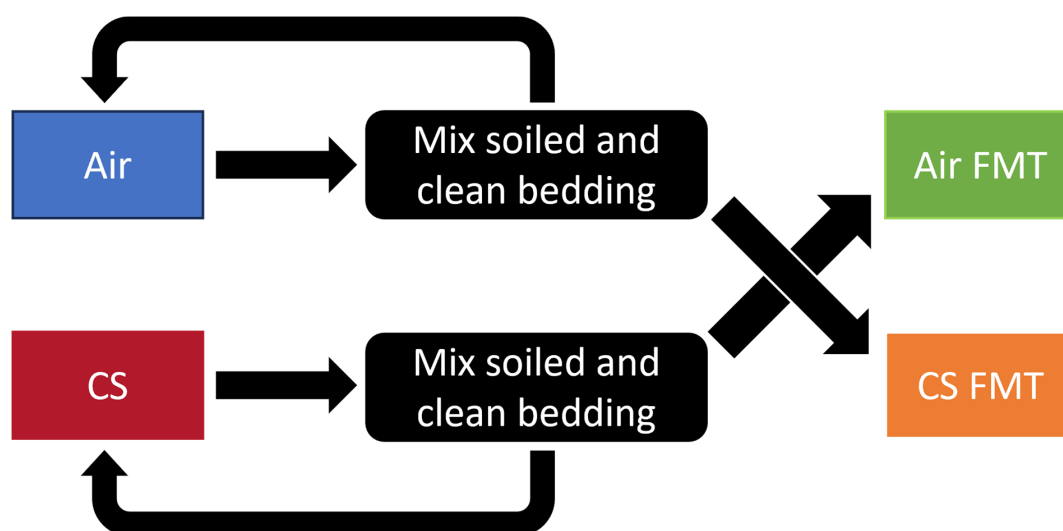
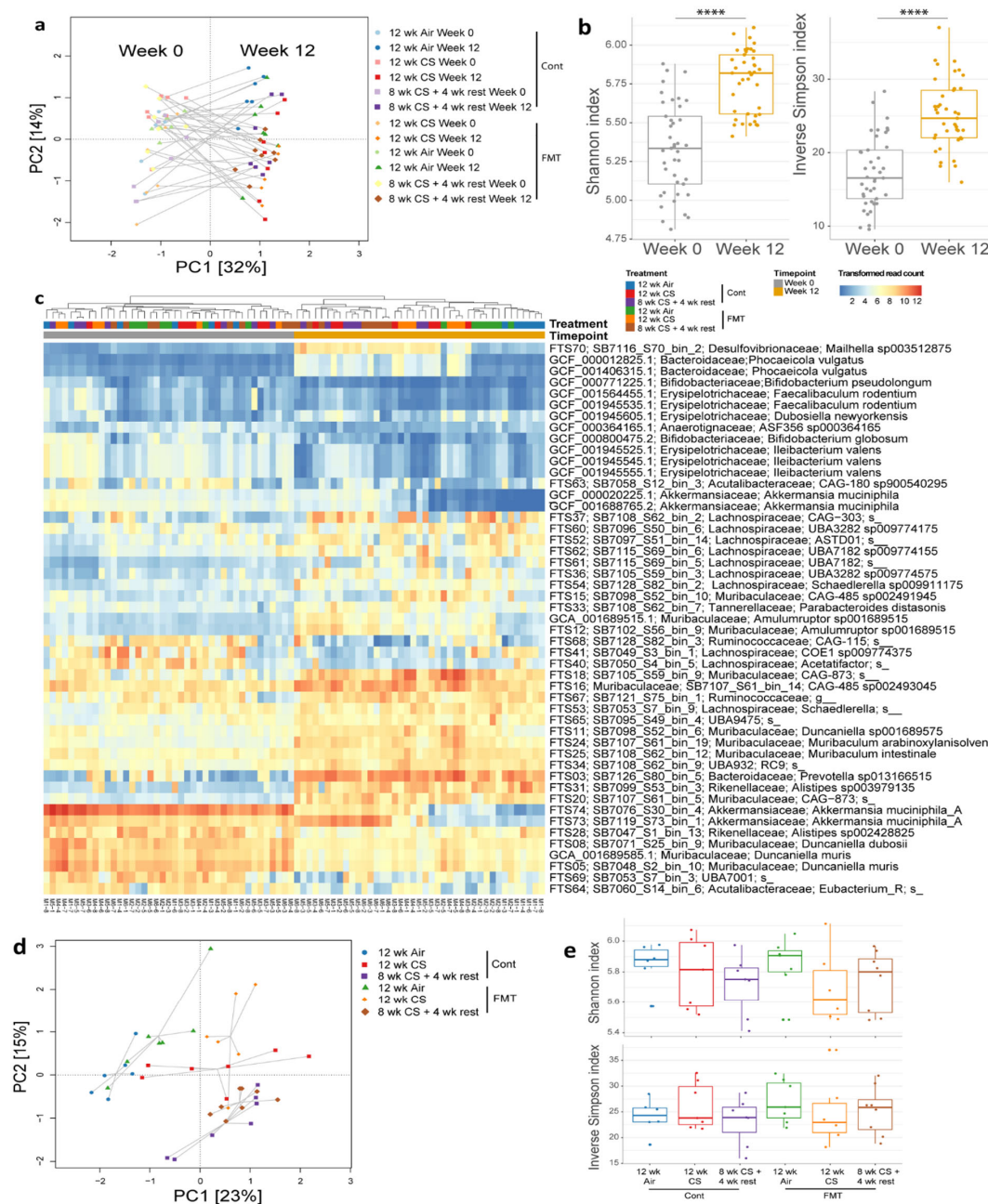


Figure S1: Schematic of fecal microbial transfer (FMT) methodology. To administer FMT, soiled bedding from air-exposed mice was mixed with clean bedding (1:2 ratio) and transferred to the cages of smoke-exposed mice. Conversely, the soiled bedding from 12-week CS-exposed mice was mixed with clean bedding and transferred to the cages of air-exposed mice. Control groups (without FMT) received their own soiled bedding mixed with clean bedding.



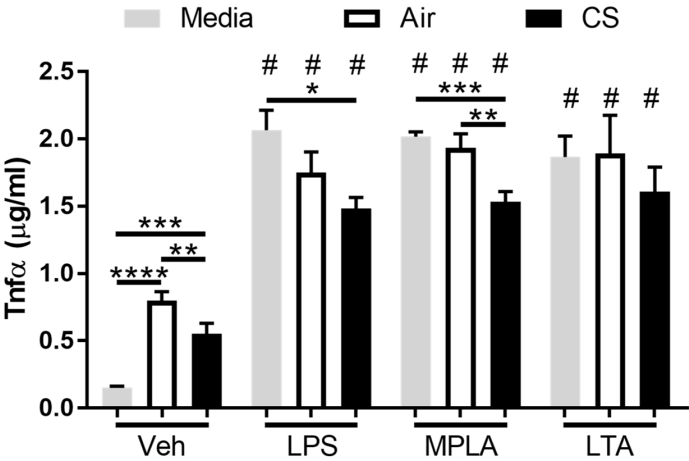
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10 **Figure S2. Fecal microbiome profiles are significantly altered across time with cigarette**
 11 **smoke (CS)-exposure and fecal microbiota treatment (FMT). a-e, Mice were exposed to CS**
 12 **(12 wk CS) or normal air (Air) for 12 weeks, or 8 weeks CS followed by 4 weeks with normal air**

2

13 (8 wk CS+ 4 wk rest). Mice also received FMT through transfer of soiled bedding twice per week
14 or were maintained in their own bedding (Control) for 12 weeks. Fecal samples from baseline
15 (week 0) and at the end of the experiment (week 12) were analyzed using shotgun metagenomics.
16 **(a)** Principal component analysis of all mice demonstrated shifts in microbiome composition with
17 time, smoke-exposure and FMT based on genome level read mapping counts, log-cumulative-
18 sum-scaling (CSS) transformed. **(b)** Simpson and Shannon diversity indices for all mice
19 demonstrated increasing diversity over time based on mapping counts across all genomes
20 transformed using the DESeq2 scaling factor. **(c)** Heatmap displaying significantly different
21 genomes (Benjamini-Hochberg adjusted $P < 0.001$, log2 fold change > 1.5) between week 0 and
22 week 12 as log-CSS transformed read counts. **(d)** Principal component analysis of week 12 mice
23 demonstrated significant separation between groups using PERMANOVA of Bray-Curtis
24 distances. **(e)** Simpson and Shannon diversity indices indicating no differences in diversity
25 between experimental groups at week 12. $N = 8$ per group. **** = $p < 0.0001$ using a Wilcoxon
26 rank sum test.
27

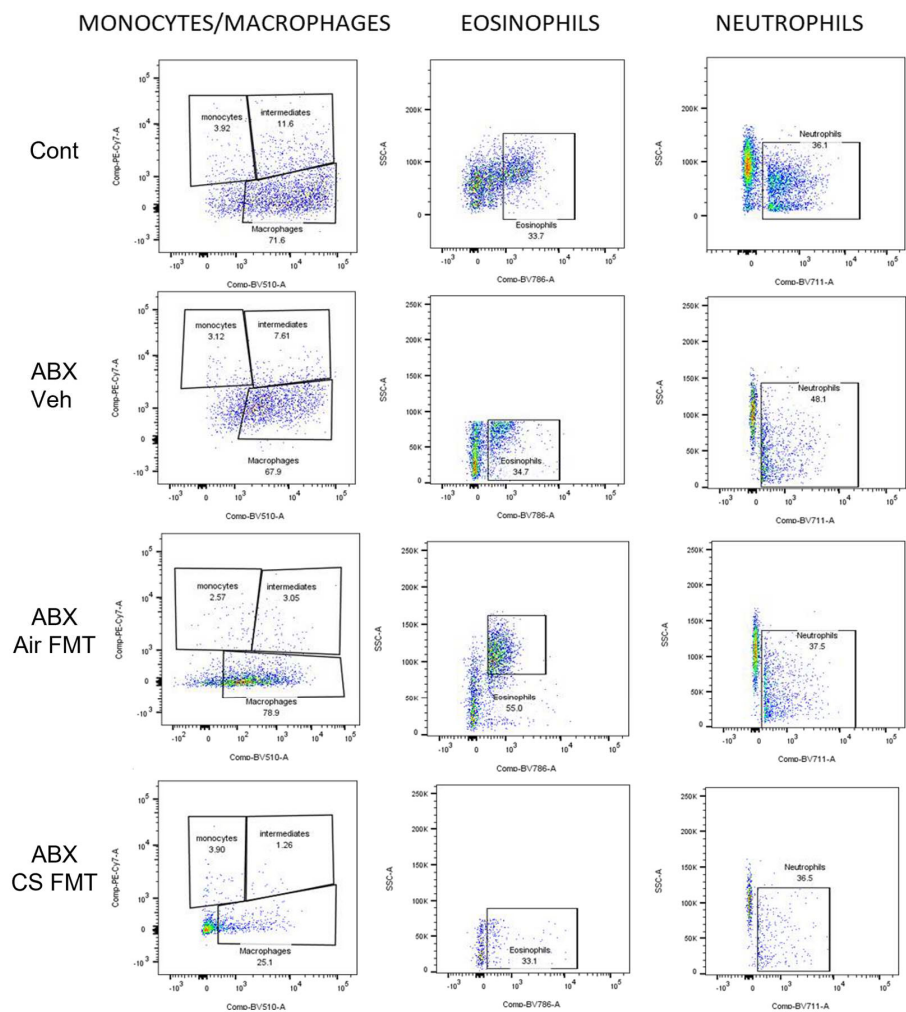
28



29

30 **Figure S3: Cigarette smoke (CS)-induced dysbiosis impaired TLR4 agonist-induced TNFα**
31 **production in monocytes.** Raw 264.7 monocytes were incubated for 24 hours with sterile-filtered
32 faecal homogenates from mice exposed to CS (black) or normal air (white) for 12 weeks, or were
33 maintained in media (grey). In the last 4 hours of incubation, cells were incubated with
34 lipopolysaccharide (LPS), monophosphoryl lipid A (MPLA), lipoteichoic acid (LTA) or vehicle
35 (Veh; 0.5% dimethylsulfoxide). Incubation with faecal homogenates induced TNFα production,
36 which was lower with faeces from CS-exposed mice. LPS- and MPLA-induced TNFα production
37 was also lower in cells incubated with faeces from CS-exposed mice. N = 3-9 per group. Data
38 presented as mean +/- standard error of the mean. ** = p<0.01; *** = p<0.001; **** = p<0.0001;
39 # = p<0.0001 vs. Veh-treated cells using one-way ANOVA with Holm-Sidak's post-hoc analysis.

40



41

42 **Figure S4: Representative FACS plots of colon granulocytes from Figure 5.** Microbiota were

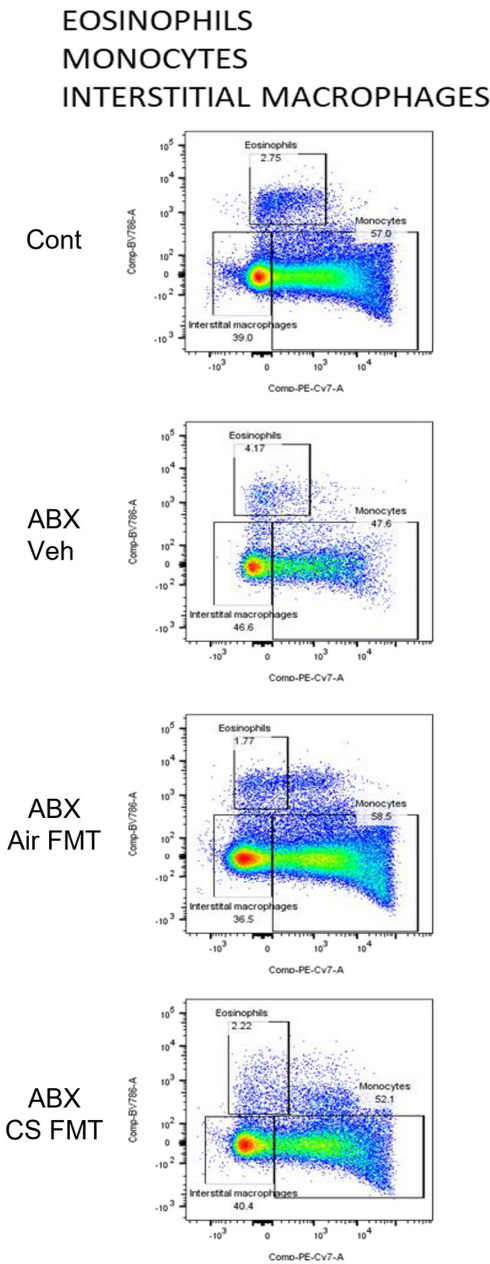
43 depleted by antibiotics (ABX) for 7 days before mice were recolonized with feces from separate

44 air- (Air FMT) or CS-exposed donors (CS FMT) for 10 days. A subset of mice received vehicle

45 (Veh, PBS + 0.05% L-cysteine) to model natural recolonisation. Innate immune cells in the colon

46 were enumerated by flow cytometry, with representative FACS plots displayed here.

47



48

49 **Figure S5: Representative FACS plots of colon granulocytes from Figure 5.** Microbiota were

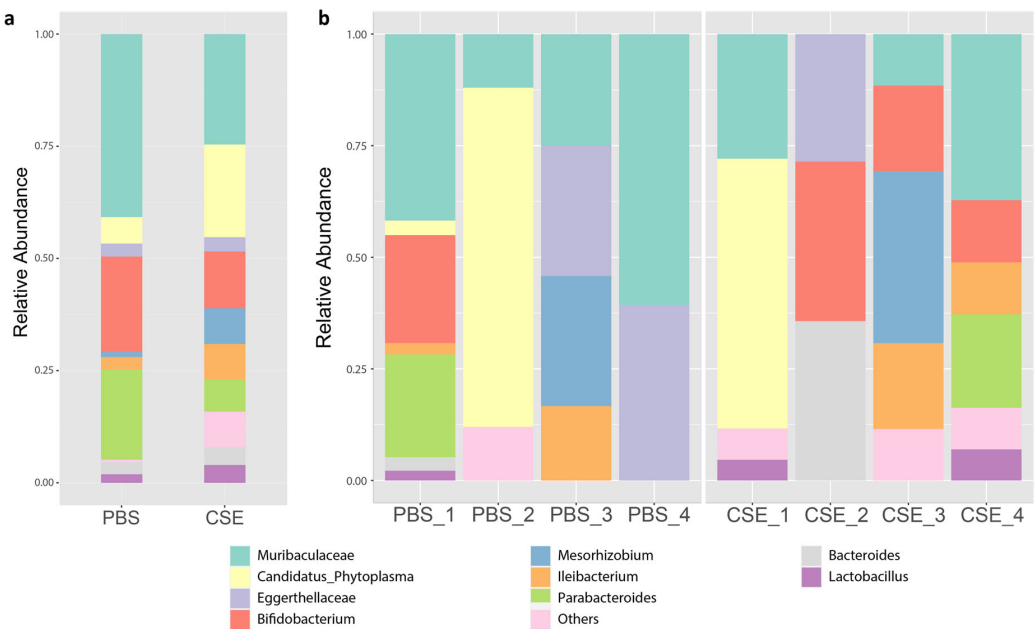
50 depleted by antibiotics (ABX) for 7 days before mice were recolonized with feces from separate

51 air- (Air FMT) or CS-exposed donors (CS FMT) for 10 days. A subset of mice received vehicle

52 (Veh, PBS + 0.05% L-cysteine) to model natural recolonisation. Innate immune cells in the lung

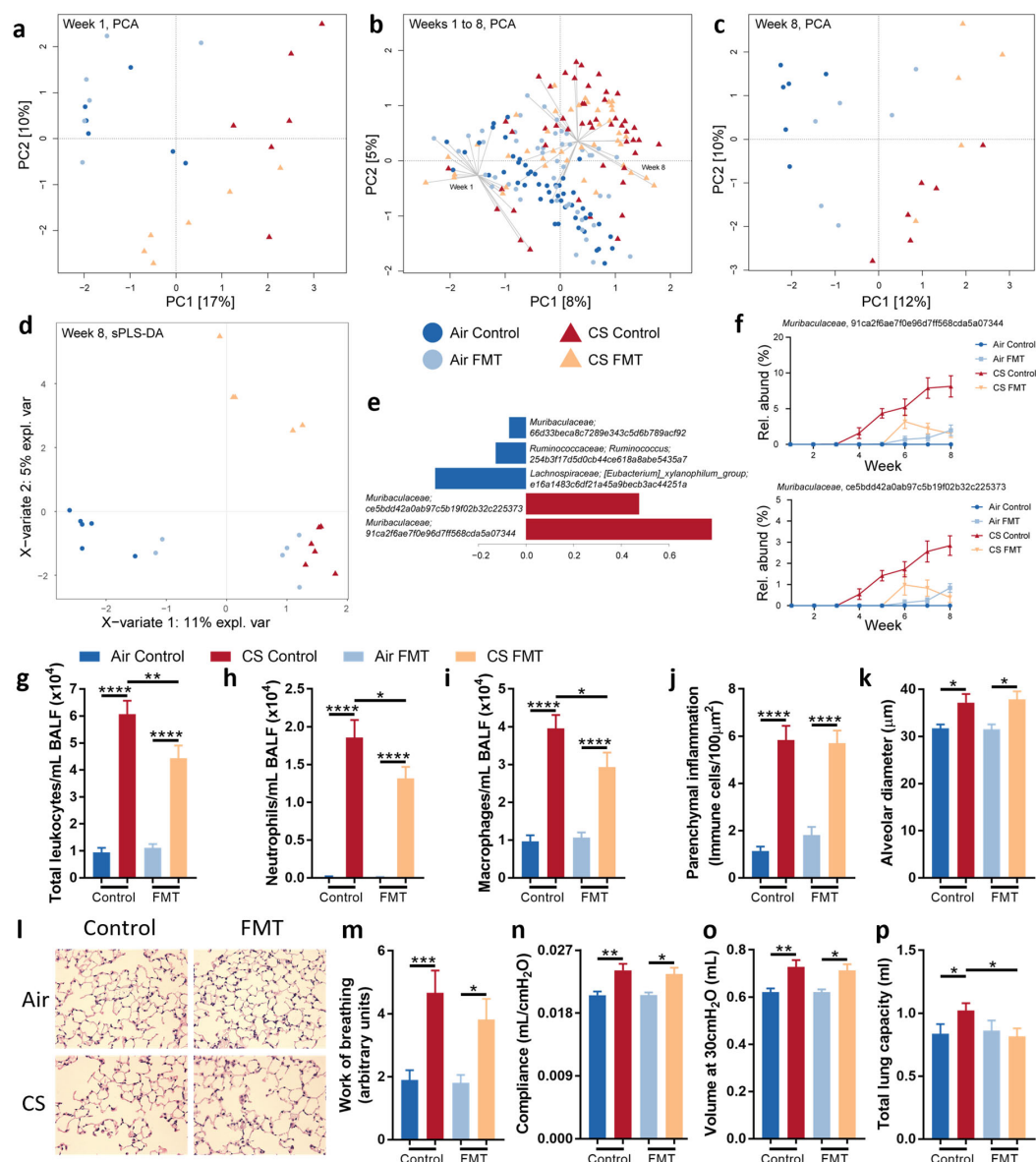
53 were enumerated by flow cytometry, with representative FACS plots displayed here.

54



55

56 **Figure S6: Cigarette smoke extract (CSE) does not differ substantially from phosphate**
57 **buffered saline (PBS) negative control.** CSE generated from 3R4F research cigarettes was
58 assessed by 16S rRNA gene sequencing. **(a)** Overall composition did not differ substantially from
59 negative controls and no cigarette smoke-associated taxa were identified in CSE above levels in
60 negative controls. **(b)** There was significant variability in presence of bacterial genera between
61 samples. N=4.



62

63 **Figure S7. Cigarette smoke (CS)-exposure was associated with a shift in the faecal**
64 **microbiome, and faecal microbiota transfer (FMT) suppressed the development of**
65 **experimental COPD. Mice were exposed to CS or normal air for eight weeks and received FMT**
66 **through transfer of soiled bedding or were maintained in their own bedding (Control). a-f, Faecal**
67 **samples were collected weekly and analyzed by 16S rRNA amplicon sequencing. (a) Principal**
68 **component analysis of mice following one week of CS-exposure and/or FMT, across (b) all eight**

69 weeks of the experiment and at (c) week eight following completion of the experiment
70 demonstrated a clear distinction in microbiota composition of the different experimental groups.
71 (d) Multivariate analysis (sPLS-DA) at week eight demonstrated distinction between air- and CS-
72 exposed groups along component 1, (e) with several species contributing to this separation. (f)
73 Relative abundance of *Muribaculaceae* sequence variants increased in CS-exposed mice but was
74 alleviated by FMT, contributing to separation of sPLS-DA across the eight-week experiment. **g-p**,
75 Hallmark features of COPD assessed at week eight. CS-exposure increased (**g-i**) total leukocytes,
76 neutrophils and macrophages in bronchoalveolar lavage fluid (BALF), (**j**) immune cells in
77 parenchyma, (**k-l**) alveolar diameter, and (**m-p**) lung function parameters of work of breathing,
78 compliance, volume and total lung capacity. FMT alleviated CS-induced increases in BALF total
79 leukocyte, neutrophil and macrophage numbers and total lung capacity. N = 5-6 per group (**a-f**)
80 or 14-18 per group from 3 independent experiments (**g-p**). Data presented as mean +/- standard
81 error of the mean. * = $p < 0.05$; ** = $p < 0.01$; *** = $p < 0.001$; **** = $p < 0.0001$ using one-way
82 ANOVA with Holm-Sidak's post-hoc analysis (**g-p**).

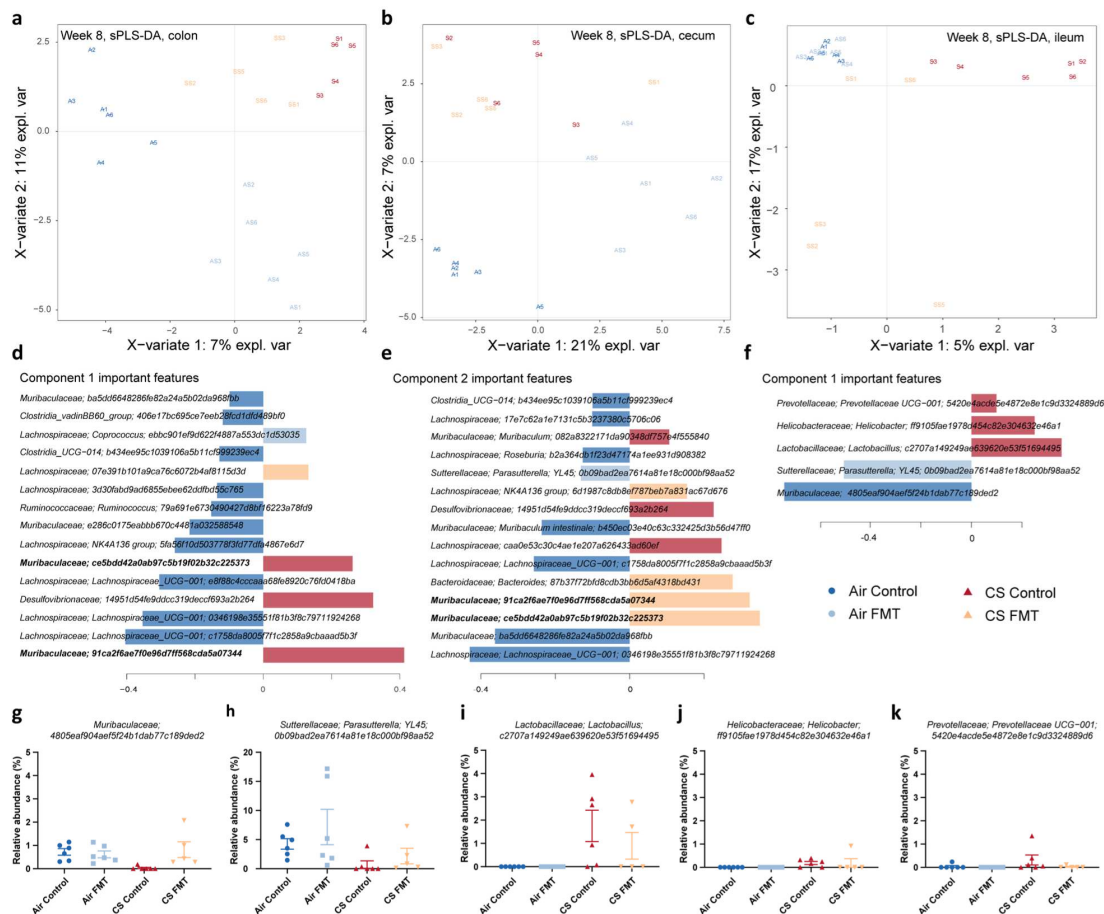


Figure S8. Analysis of ileum samples reveals distinction between cigarette smoke (CS)-exposed groups based on faecal microbiota transfer (FMT) treatment. a-k, Mice were exposed to CS or normal air (Air) for eight weeks, and received FMT through transfer of soiled bedding or were maintained in their own bedding (Control). Multivariate analysis (sPLS-DA) at week eight based on 16S rRNA amplicon sequencing indicated CS-induced separation of microbiome samples, resembling that observed in faecal samples, in the (a) colon and (b) caecum but not the (c) ileum, although the latter did demonstrate a difference between CS control and CS FMT groups. (d) Species contributing to separation along the CS-associated component 1 of a and (e) component 2 of b. (f) Species contributing to separation along the air-associated component 1 of c. (g-k) Relative abundance of air-associated ASVs from the ileum from CS FMT mice trended towards the levels of air-exposed control mice. N = 5-6 per group.

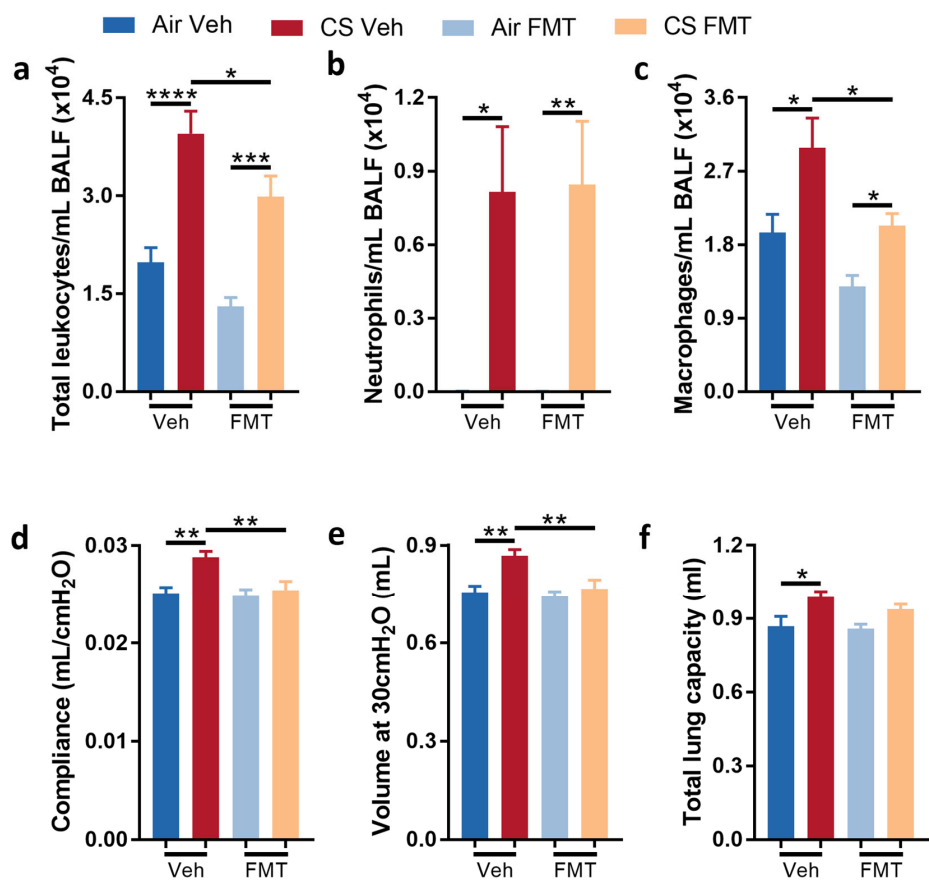


Figure S9. Faecal microbiota transfer (FMT) by oral gavage suppressed the development of experimental COPD. a-f, Mice were exposed to cigarette smoke or normal air for eight weeks and received FMT or vehicle (Veh; phosphate buffered saline + 0.05% L-cysteine) through oral gavage. (a-c) CS-exposure increased total leukocytes, neutrophils and macrophages numbers in bronchoalveolar lavage fluid (BALF), and (d-f) the lung function parameters compliance, volume and total lung capacity. FMT alleviated CS-induced increases in BALF total leukocytes and macrophages, and lung compliance and volume. N = 10-12 per group from 2 independent experiments. Data presented as mean +/- standard error of the mean. * = p<0.05; ** = p<0.01; *** = p<0.001; **** = p<0.0001 using one-way ANOVA with Holm-Sidak's post-hoc analysis.

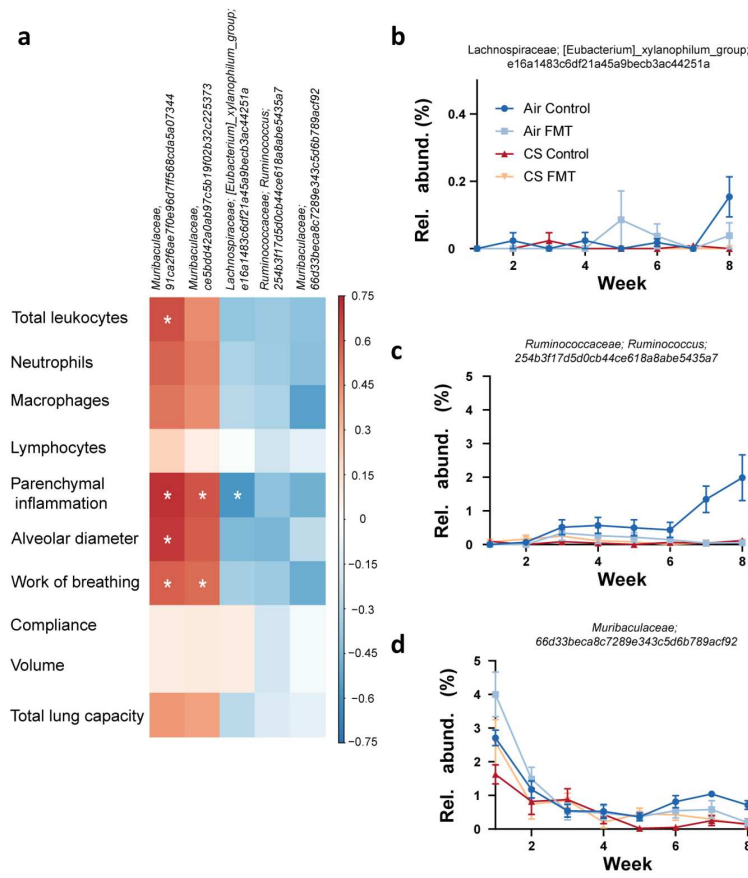


Figure S10: A cigarette smoke (CS)-associated *Muribaculaceae* amplicon sequence variant (ASV) correlates with a subset of phenotypic measures. a-d, Mice were exposed to CS or normal air (Air) for eight weeks and received faecal microbiota transfer (FMT) through transfer of soiled bedding or were maintained in their own bedding (Control). Faecal samples were collected weekly and analyzed by 16S rRNA amplicon sequencing. (a) Spearman's rho calculated between centered log-ratio transformed ASV relative abundance and phenotypic scores indicated correlation between *Muribaculaceae* 91ca. and a subset of phenotypic measures. (b-d) Relative abundances of air-associated ASVs from Fig. S2e across the eight-week experiment were variable. N = 5-6 per group. * = p<0.05 using Spearman's correlation.

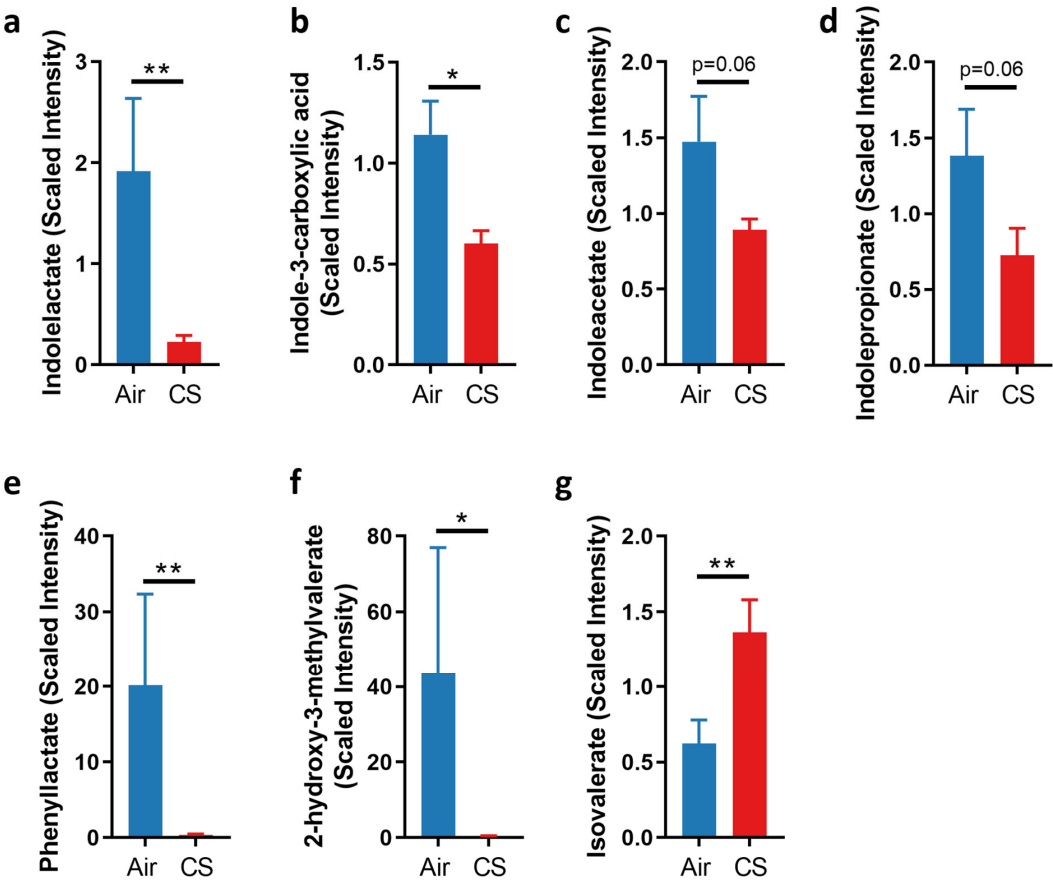


Figure S11: Metabolomics show cigarette smoke (CS)-induced alterations in microbial protein metabolism. a-g, Mice were exposed to CS or air for 12 weeks, and metabolites in cecum contents assessed. Microbial amino acid metabolites were broadly downregulated in response to CS, except for isovalerate which was increased. N = 8 per group. Data presented as mean +/- standard error of the mean. * = p<0.05; ** = p<0.01 using one-way unpaired Student's T test on log-transformed data.

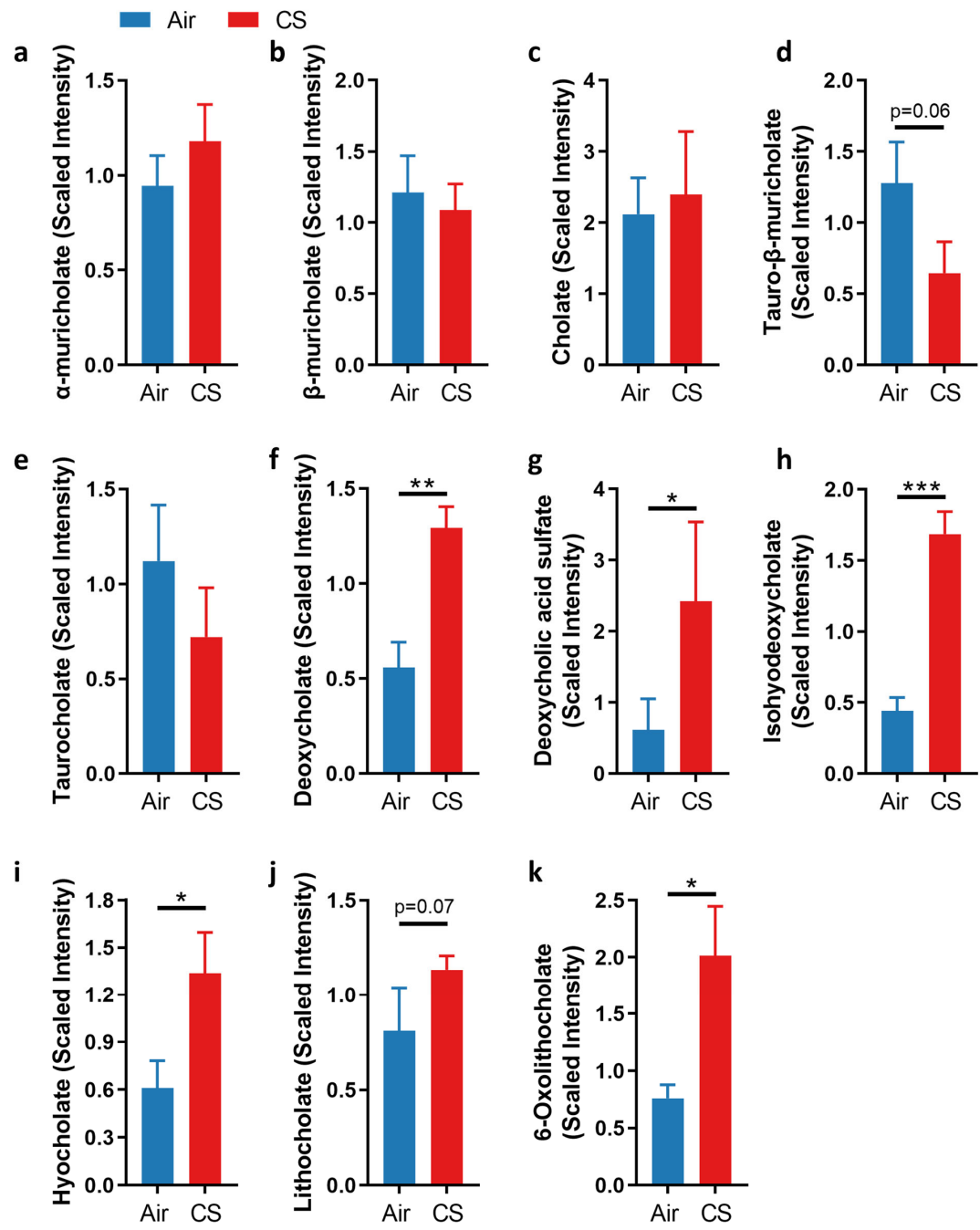


Figure S12: Metabolomics show cigarette smoke (CS)-induced alterations in microbial protein metabolism. a-k, Mice were exposed to CS or air for 12 weeks, and metabolites in cecum

133 contents assessed. (a-e) Host-derived primary bile acids were not altered by CS. (f-k) Secondary
134 bile acids, produced by microbial metabolism of primary bile acids, were increased by CS. N = 8
135 per group. Data presented as mean +/- standard error of the mean. * = $p < 0.05$; ** = $p < 0.01$; *** =
136 $p < 0.001$ using one-way unpaired Student's T test on log-transformed data.

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Supplementary Methods:**Mice, CS-exposure, microbiome transfer and diet studies**

Female C57BL/6 mice (3-5 weeks old) were obtained from the Animal Services Unit, The University of Newcastle (Newcastle, Australia). Upon arriving in the facility, the baseline microbiome was normalized for three weeks by pooling and mixing all soiled bedding and redistributing to all cages (twice weekly) and co-housing mice from different experimental groups weekly. Microbiome normalization was not performed for validation experiments (8 weeks CS with FMT).

Mice were exposed to normal room air or CS from twelve 3R4F reference cigarettes (University of Kentucky, Lexington, KY) in a custom-designed, nose-only inhalation apparatus, twice per day, 5 days per week, for 8 or 12 weeks, as previously described.¹⁻⁹ In some longer experiments, mice exposed to CS for 8 weeks were rested for the final 4 weeks of the experiment to model smoking cessation. Twice per week, soiled bedding from air-exposed mice was mixed with clean bedding (1:2 ratio) and transferred to the cages of smoke-exposed mice. Conversely, the soiled bedding from 12-week CS-exposed mice was mixed with clean bedding and transferred to the cages of air-exposed mice (Fig S1). Control groups (without FMT) received their own soiled bedding mixed with clean bedding. Feces were collected weekly, with at least 3 fresh fecal pellets collected and stored at -80°C until processing.

For FMT by oral gavage, fresh feces were collected from normal air-exposed mice and placed immediately in phosphate buffered saline (PBS) with 0.05% L-cysteine to preserve anaerobic bacteria (100mg/1.5mL). Feces was gently homogenized by hand and debris removed by centrifugation (800xg, 3 min). Resulting supernatants were administered by oral gavage, beginning in the first week of smoke exposure.

For antibiotic depletion experiments, mice were not exposed to CS to assess the direct impacts of CS-associated microbiota. Mice received antibiotics (ampicillin, metronidazole, neomycin, gentamicin [1g/L] and vancomycin [0.5g/L]) in drinking water for 7 days.¹⁰ Antibiotics were removed and mice received FMT from a separate cohort of air- or CS-exposed donors four times in 10 days.

For diet studies, mice were fed a conventional semi-pure diet (AIN93G) or a resistant starch diet (SF11-025; Specialty Feeds, WA, Australia) *ad libitum* commencing 2 week prior to CS-exposure and maintained until the end of experiment. All experiments were approved by University of Newcastle Animal Ethics Committee.

DNA extraction and metagenome assembly and binning

DNA was extracted from 50-100 mg of fecal material using an initial bead beating step and extraction using a Maxwell 16 Research Instrument and Tissue DNA Kit (Promega, USA) according to the manufacturer's protocol. DNA concentrations were measured using a Qubit assay (Life Technologies, USA) and were adjusted to a concentration of 5 ng/ul. Library preparation was performed using Nextera DNA Library Preparation Kits (Illumina, USA) and samples sequenced across two runs on an Illumina NextSeq500 instrument generating an average of 3 Gbp of 150 bp paired-end reads per sample. Each sample was included on both sequencing runs and data combined. Contaminating host reads were removed by mapping against the mouse genome (*Mus musculus* GRCm38.p5) using BWA v0.7.12¹¹ requiring a minimum alignment length of 30 bases and maximum of 15 clipped bases for reads to be considered of mouse origin. Raw host-free reads from each sample were assembled using Spades v3.12.0¹² with the --meta flag. Reads were mapped to each resulting assembly using BamM v1.7.3 (<https://github.com/ecogenomics/BamM>) and bins produced using Metabat v2.12.1,¹³ MaxBin¹⁴ and GroopM2 v2.0.0-1¹⁵ implemented within the ensemble binning tool UniteM v0.0.15 (<https://github.com/dparks1134/UniteM>). Contamination and completeness of bins from all samples were assessed using CheckM v1.0.11.¹⁶ Bins with completeness >80% and contamination <7% were retained and de-replicated using dRep v2.05¹⁷ with default settings (99% identity), skipping quality filtering, to produce the final set of metagenome-assembled genomes (MAG). Taxonomy was determined using GTDB-Tk v0.3.0 and v1.5.0¹⁸ with GTDB releases 04-RS89 and 06-RS202.¹⁹ EukDetect v2022²⁰ and MetaPhlAn v3.01 (database v31_CHOCOPhlan_201901)²¹ were used to detect fungi and viruses (MetaPhlAn only) within metagenomic samples. In addition, raw reads were mapped to fungal and viral reference genomes from NCBI using CoverM v0.6.1 (<https://github.com/wwood/CoverM>) filtering for alignments $\geq 95\%$ identity across $\geq 90\%$ of the read length and genomes with a covered fraction

≥10%. For fungi, all complete or chromosome level genomes were included (247 genomes). For viruses, only genomes with host *Mus musculus* were included (37 genomes).

Metagenomic community profiling

Reads for each sample were mapped to a de-replicated set of 18,342 genomes from NCBI (GTDB release 03-RS83) using BamM with minimum seed length of 25. Genomes with >1x coverage of >1% of the genome and overall coverage of 0.01X, as determined using Mosdepth v0.2.3,²² were retained and combined with a de-replicated set of recovered MAGs for assessment of community composition. Read counts for the final genome set was determined for each sample *via* mapping using BamM with a minimum seed length of 25 bases and filtering for minimum mapping percentage identity of 95%. Per genome read counts were scaled to account for genome size whilst maintaining the raw unmapped read percentage for each sample as a reflection of unrepresented diversity. Relative abundance was calculated using scaled read counts as a fraction of total non-host reads per sample. α -diversity was calculated using QIIME v1.8.0²³ with read counts per-genome normalised to account for library size using DESeq2 v1.20.0.²⁴

Statistical analysis of metagenomic data

Principal component analysis was performed using the R package *vegan* v2.5-1²⁵ on data transformed using log cumulative-sum-scaling (log-CSS) implemented within *metagenomeSeq* v1.22.0.²⁶ Differential abundance between sample groups was determined using DESeq2 using read counts per genome scaled to account for genome size and Benjamini-Hochberg adjustment for multiple comparisons. sPLS-DA analysis was conducted using the R package *mixOmics* v6.3.2²⁷ with centered log-ratio transformed relative abundance (adding pseudo count one order of magnitude below the lowest non-zero value values) with 50x4-fold cross-validation. Host phenotypes were tested for association with microbiome composition using the *envfit* function based on NMDS ordination of Bray-Curtis distances calculated using the *metaMDS* function, both implemented within the *vegan* R package. Significance was determined based on 10,001 permutations and resulting P values were adjusted for multiple comparisons using the Benjamini-Hochberg method. Samples with missing scores were removed. Spearman's rho was calculated using the 'corr.test' function within the R package *psych* v1.8.12²⁸ based on centered log ratio transformed genome relative abundance and raw phenotypic data. Correlation matrix was

produced using the ‘corrplot’ function with the R package corrplot v0.84.²⁹ Non-random relationships between bacteria were identified using the CoNET Cytoscape application.³⁰ Rare taxa (occurrence in <25% of samples) were filtered to avoid spurious relationships, and relationships identified from Pearson and Spearman correlations, mutual similarity, and Kullback-Leibler and Bray-Curtis dissimilarity measures. P-values were combined using Brown’s method and adjusted for multiple comparisons using the Benjamini-Hochberg method.

Culture of CS extract:

CS extract was produced by bubbling smoke from one cigarette through 10mL water or PBS. CS extract (100µL) was cultured on Yeast Casitone Fatty Acids agar plates in a DG250 anaerobic chamber (Don Whitley Scientific) for 48hrs, or analyzed by 16S rRNA amplicon community profiling.

Assessment of experimental COPD: airway inflammation, emphysema-like alveolar enlargement

Airway inflammation was quantified by total and differential enumeration of immune cells in BALF as previously described.^{1, 2, 7} Briefly, two 0.4 ml washes with PBS of the left lung were performed, red blood cells lysed and total inflammatory cells counted, cytopun, air dried and stained with May-Grunwald-Giemsa for differential counts. For histological analysis, lung tissue was perfused with saline *via* cardiac puncture, inflated (500µL) and fixed in formalin prior to mounting, sectioning and staining. Hematoxylin and eosin-stained lung sections were used to assess parenchymal inflammation by counting the number of inflammatory cells in 10 randomised fields of view at 100x magnification,^{1, 7, 8} and emphysema-like alveolar enlargement using the mean linear intercept, with assessor blinded to sample identity.^{7, 31}

Lung function

Mice were anaesthetized with ketamine (100 mg/kg) and xylazine (10 mg/kg). Tracheotomy was performed, and mice cannulated. A BioSystem Forced Maneuvers (Buxco, Wilmington, NC) and a flexivent apparatus (Legacy System, SCIREQ, Montreal, Canada) were used to measure all lung function parameters. Each perturbation was conducted a minimum of three times and the average calculated for each mouse.^{2, 4, 7-9, 31}

264

265 Colon histopathology

266 For histological analysis, colon and ileum tissue was fixed in formalin prior to mounting,
267 sectioning, and staining with H&E. Images were taken at 20x magnification and sub-mucosal
268 fragmentation area measured using ImageJ software. Blood vessels were enumerated along the
269 entire colon and corrected for sub-mucosal length.³¹

270

271 RNA extraction, cDNA synthesis, and qPCR analysis

272 Colon tissues were snap frozen and stored at 80°C. RNA was extracted using a previously
273 described TRIzol (ThermoFisher Scientific) protocol utilizing phase separation in chloroform and
274 RNA precipitation in isopropanol.^{2, 7, 8, 32} Random-primed reverse transcriptions were performed
275 followed by real-time qPCRs with SYBR-green-based detection using a Viia 7 Real Time PCR
276 system.^{2, 7, 8, 32} Specific mRNA transcripts were assessed and expressed as relative abundance
277 compared to β -actin (*Actb*) expression (Table S18).

278

279 CyTOF analysis

280 For time-of-flight mass cytometry, whole blood was collected into ethylenediaminetetraacetic acid
281 (EDTA)-coated tubes and spleen tissue disaggregated using a 70 μ m cell strainer. Bone marrow
282 was collected and immediately incubated with the cell cycle marker iododeoxyuridine (IdU;
283 10 μ M, 1hr, 37°C). For all samples, red blood cells were lysed and cells quantified using trypan
284 blue. Single-cell suspensions were stained with the viability marker cisplatin (5 μ M, 5 minutes,
285 room temperature), quenched (5% foetal calf serum, 5mM EDTA in PBS) and collected by
286 centrifugation (300xg, 3 minutes, 4°C). Fc receptors were blocked with metal conjugated anti-
287 mouse CD16/32 antibody (Table S19, 30 min, 4°C) and quenched before staining with surface
288 antibodies (Table S18, 30 min, 4°C). Cells were washed and fixed (4% paraformaldehyde, 4°C
289 overnight) before permeabilization in methanol (10 minutes, 4°C), washing and staining with
290 intracellular antibodies (Table S19, 45 minutes, room temperature). Samples were washed thrice
291 in permeabilization buffer (Thermo Fisher) and stored in 4% paraformaldehyde at 4°C until
292 analysis using a Helios instrument (Fluidigm).

293

The expression levels of 38 markers (Table S19) measured by CyTOF was used as input for downstream analysis. Before analysis, a maximum of 10,000 cells for each sample were randomly selected. Protein expression was transformed with a logicle transformation³³ (width=0.25, top=16,409, full width=4.5) and used for Uniform Manifold Approximation and Projection for Dimension Reduction (UMAP) calculated using the naïve R implementation (umap 0.2.7) of the algorithm with the following settings: number of neighbors=15, number of components=2, distance metric=Euclidean, minimum distance=0.2. Cells clusters showing similar marker expression were calculated according to the Phenograph algorithm (Rphenograph, Github JinmiaoChenLab v. 0.99.1)³⁴ with standard setting and number of nearest neighbors (k)=60. Data were visualized according to the coordinate on the UMAP dimensions and the cluster assignment of each cell. Confusion matrices were calculated normalizing the number of cells in each experimental group to 1,000 and calculating the contribution of each condition in each cluster to avoid bias from different cell numbers in experimental groups. Finally, a heatmap was generated using the mean marker expression in each cluster scaled for visualization. Data were visualized using the R packages ggplot2 (v. 3.3.2) and pheatmap (v. 1.0.12). All analysis was performed in a dockerized environment based on R 4.0.3 and Bioconductor 3.12 (lorenzobonaguro/flowtools:v2).

Cell culture

Raw 264.7 monocytes (1×10^6) were seeded into 12-well plates and allowed to adhere overnight. Feces from CS- or air-exposed mice were homogenized in DMEM (100mg/mL) and sterile-filtered. Cells were incubated in media alone, or 1:1,000 dilutions of filtered fecal homogenate for 24 hours. During the final 4 hours of incubation lipopolysaccharide (1µg/mL), monophosphoryl lipid A (4µg/mL), lipoteichoic acid (1µg/mL) or dimethyl sulfoxide (0.5%) were added to cell media. Media was collected and stored at -80°C until TNFα protein was assessed using DuoSet ELISA kits (R & D Systems).

Flow cytometry

Excised colons with contents removed were cut into 1cm pieces, placed in pre-digestion buffer (2mM EDTA in Hanks buffered salt solution [HBSS]), and incubated with shaking (150 rpm, 15 min, 37°C). Tissues were washed thrice in HBSS buffer before incubation in digestion buffer

(0.425mg/ml Collagenase V, 0.625mg/ml Collagenase D, 1mg/ml Dispase II [Sigma], 10% FBS in RPMI 1640; 150 rpm, 45 minutes, 37°C). Digested tissue was passed through a 100µm cell strainer, centrifuged (1500 rpm, 7 min, 4°C) and resuspended in FACS buffer (2% FBS, 2mM EDTA in PBS) for staining.

Lung tissue was stored in HEPES buffer (10mM HEPES-NaOH pH 7.4; 150 mM NaCl; 5mM KCl; 1mM MgCl₂; 1.8mM CaCl₂) containing Collagenase D (40U/mL, Sigma-Aldrich) and DNase I (2mg/mL, Sigma-Aldrich) and mechanically disrupted using a gentleMacs machine (m1_lung program, Miltenyl Biotec, NSW Australia). After incubation (150 rpm, 20 min, 37°C), tissues were disrupted again using the gentleMacs machine (m2_lung program), strained using a 100µm cell strainer, and collected by centrifugation (1500 rpm, 5 min, 4°C). Red blood cells were lysed using red blood cell lysis buffer (5 min, 4 °C) before inactivation in FACS buffer, centrifugation and resuspension in FACS buffer.

Single cell isolations (1x10⁶) were collected by centrifuged (1500 rpm, 7 min, 4°C), washed with PBS, and stained with viability dye (Table S19; 30 min, 4°C). Cells were washed with FACS buffer, Fc receptors blocked (BD Biosciences, 15 min, 4°C), and stained with antibodies (Table S20, 30 min, 4°C). Stained cells were washed three times with FACS buffer before analysis using an LSR Fortessa cytometer (BD Biosciences) and FlowJo software (TreeStar).

DNA extraction and 16S rRNA amplicon community profiling

DNA was extracted from fecal samples using the PowerSoil DNA extraction kit according to the manufacturer's instructions (MoBio, Carlsbad, California). For community profiling, the V5-V8 region of the 16S rRNA gene was amplified using a modified two-step PCR protocol.³⁵ The first round 50µl reaction mixture comprised 20ng of template DNA, molecular biology grade water, 5µl 10x PCR buffer (Fisher Scientific), 4µl of 25mM MgCl₂ (Fisher Scientific), 1.5µl BSA (New England Biolabs), 1µl of 10mM dNTPmix (Fisher Scientific), 1µl of 10nM forward primer 803F (803Fa-5': TTAGATACCCTGGTAGTC; 803Fb-5': TTAGATACCCSGGTAGTC; 803Fc-5': TTAGATACCCYHGTAGTC; 803Fd-5': TTAGAGACCCYGGTAGTC; mixed at ratios of (2a:b:c:d), 1 µl of 10nM reverse primer 1392wR (ACGGGCGGTGWGTRC)) and 0.2µl of 5U/ul Taq Polymerase (Fisher Scientific). Cycling conditions were as follows: 95°C for 3 minutes, 30 cycles of 95°C for 30 seconds, 55°C for 30 seconds and 74°C for 30 seconds followed by a final extension of 74°C for 10 minutes. PCR products were re-amplified in a second reaction using

primers containing (5'-3') 454 adaptor sequences, barcode (only for reverse primer), linker sequence and the gene-specific primer sequence (803F and 1392wR: pyroLSSU803F – CCTATCCCCTGTGTGCCTTGGCAGTCTCAGTTAGAKACCCBNGTAGTC pyroLSSU1392wR – CCATCTCATCCCTGCGTGTCTCCGACTCAGACATAGTACGGGCGGTGWGTRC). For the second step reactions, 2 µl of the first step PCR product was added as template to a 50 µl reaction. PCR conditions were the same as those for the first step with the number of cycles reduced to 10. Amplicons were purified using AMPure beads (Beckman Coulter) and pooled in equal ratios. Amplicon pools were sequenced from the reverse primer using 454 GS-FLX Titanium chemistry.

Amplicon sequences were demultiplexed and converted to fastq files using QIIME v1.9.1 scripts `split_libraries.py`, `convert_fastaqual_fastq.py` and `split_sequence_file_on_sample_ids.py`. Reads were then quality filtered and trimmed of adapter sequence using BBDuk v38.41 (<https://sourceforge.net/projects/bbmap/> via BBDuk; `ktrim=l k=15 mink=10 hdist=1 interleaved=f copyundefined=t mm=f qtrim=rl trimq=10 minlen=400`). Sequence variants were defined using QIIME2 v2019.10³⁶ and DADA2 (via `denoise-pyro`).³⁷ Taxonomy was assigned using QIIME2 `feature-classifier`³⁸ `classify-consensus-blast` against the Silva v138 database.³⁹

Proteomics of mouse feces

Feces (100µg) were added to 500µL of sodium bicarbonate, vortexed with 2.5-mm glass beads and supernatants collected (300xg, 5 min, 4°C). Pellets were resuspended in sodium bicarbonate, centrifuged and supernatants collected a further two times. Supernatants were combined and washes repeated to remove debris before final cell collection (14,000xg, 20 min, 4°C). Cells were aliquoted (~30mg wet weight per aliquot) and subjected to enzymatic degradation (20µL lysozyme in buffer containing 20mM TrisHCl, 2 mM EDTA and 1% Triton X-100; 30 min, 37°C). Incubation was halted with sodium bicarbonate and cell debris removed by centrifugation (12,000xg, 20 min, 4°C). Proteins (150µg) were precipitated, resuspended in 8M urea, reduced with dithiothreitol and alkylated with iodoacetamide. Samples were diluted in 50mM ammonium bicarbonate to a final urea concentration of 1M and digested overnight with trypsin (50:1 protein:enzyme, 37°C). Following acidification in 1% trifluoroacetic acid, 20µg of peptides were

dried by vacuum concentration, resuspended in 2% acetonitrile/0.1% trifluoroacetic acid and 500ng injected onto a trapping column for pre-concentration (Acclaim Pepmap100, 20x0.075mm, 3µm C18, Thermo Scientific), followed by nanoflow liquid chromatography (Ultimate 3000RSLCnano, Thermo Scientific). Peptide separation was achieved over a 250x0.075 mm ID, PepMap 2µm EasySpray C18 column using the following mobile phases: water, 0.1% formic acid (solvent A) and 80% acetonitrile incorporating 0.1% formic acid (solvent B). Peptides were resolved using a linear gradient from 2% B to 35% B over 70 min at a constant flow of 300nl/min. Peptides were introduced via an EasySpray Nano source coupled to a Q-Exactive Plus Quadrupole Orbitrap mass spectrometer (Thermo Scientific) and subjected to data dependent tandem mass spectrometry (MS/MS) using the following parameters. Full MS scans were acquired in profile mode, positive polarity at a resolution of 70,000, automatic gain control target 1×10^6 and a max injection time of 50ms. Data dependent MS/MS was then preformed at a resolution of 17,500 on the top 20 ions with a charge ≥ 2 , automatic gain control target of 5×10^5 , max injection time of 120ms, using a stepped NCE of 26 and 30. Dynamic exclusion was set at 15 seconds. .raw files were processed in Proteome Discoverer v2.1 using the Sequest HT algorithm⁴⁰ and searched against the mouse gut microbiota GigaDB database.⁴¹

Integration of proteomics and metagenomics

Metagenomics results were collapsed by adding relative abundances of bacteria from the same genera. Genera and proteins present in $\geq 50\%$ of samples were included for analysis. Both microbiome and proteomic data were integrated through correlation analysis using the psych package in R⁴² and Diablo mixOmics package.

Metabolomics

Caecum contents were collected from mice exposed to CS or normal air for 12 weeks, snap frozen and shipped on dry ice to Metabolon Inc (Durham, NC). Samples were analysed using the Global HD4 mass spectrometry platform using ultra-high performance liquid chromatography. Peaks were identified by comparison to a library of purified standards (or recurrent unknown compounds) containing retention time/index, mass:charge ratio (m/z), and chromatographic data (including MS/MS spectral data), and the relative abundance of each metabolite scaled such that the median value was equal to 1.

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