



Article Sputum Microbiota in Coal Workers Diagnosed with Pneumoconiosis as Revealed by 16S rRNA Gene Sequencing

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Abstract: Coal worker's pneumoconiosis (CWP) is an occupationally induced progressive fibrotic lung disease. This irreversible but preventable disease currently affects millions across the world, mainly in countries with developed coal mining industries. Here, we report a pilot study that explores the sputum microbiome as a potential non-invasive bacterial biomarker of CWP status. Sputum samples were collected from 35 former and active coal miners diagnosed with CWP and 35 healthy controls. Sequencing of bacterial 16S rRNA genes was used to study the taxonomic composition of the respiratory microbiome. There was no difference in alpha diversity between CWP and controls. The structure of bacterial communities in sputum samples (β diversity) differed significantly between cases and controls (pseudo-F = 3.61; *p* = 0.004). A significant increase in the abundance of *Streptococcus* (25.12 ± 11.37 vs. 16.85 ± 11.35%; *p* = 0.0003) was detected in samples from CWP subjects as compared to controls. The increased representation of *Streptococcus* in sputum from CWP patients was associated only with the presence of occupational pulmonary fibrosis, but did not depend on age, and did not differ between former and current miners. The study shows, for the first time, that the sputum microbiota of CWP subjects differs from that of controls. The results of our present exploratory study warrant further investigations on a larger cohort.

Keywords: coal worker's pneumoconiosis; sputum microbiome; lung fibrosis; next-generation sequencing; 16S rRNA genes; *Streptococcus*

1. Introduction

Coal worker's pneumoconiosis (CWP) is an occupationally induced progressive fibrotic lung disease, caused by the deposition of coal mine dust in the lung parenchyma and by the subsequent tissue reaction. This public health problem typically occurs in the coal mining industry, including opencast mining, around the world [1,2]. In many countries, coal is still used as an important source of energy, and coal mining remains a major industry. This irreversible but preventable disease currently affects millions across the world [3]. Numerous studies have shown that inhalation of coal dust containing crystalline silica (silicon dioxide), usually in the form of quartz or silica, is the primary cause of silicosis, leading to progressive pulmonary fibrosis, which is the main clinical and pathological feature of CWP [4]. CWP is associated with an increased risk of malignancy [5], which is consistent with an increase in the baseline level of chromosomal damage in lymphocytes of patients compared with healthy individuals [6,7].



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Copyright: © 2022 by the authors. Licensee MDPI, Basel, Switzerland. This article is an open access article distributed under the terms and conditions of the Creative Commons Attribution (CC BY) license (https:// creativecommons.org/licenses/by/ 4.0/). The main mechanisms of CWP include silica-induced macrophage cytotoxicity, activation of leukocytes to produce active oxygen radicals, and damage to alveolar epithelial cells stimulating fibroblast proliferation. Deregulation of DNA methylation was also pointed out as a possible mechanism of CWP pathogenesis [8]. However, the exact mechanisms of progressive pulmonary fibrosis in CWP remain to be elucidated. In particular, the possible effects of the respiratory tract microbiota on the etiology and pathogenesis of CWP need further investigation.

Numerous recent studies using metagenomic sequencing have shown that the respiratory microbiota plays an important role in maintaining lung health but can differ significantly in various diseases associated with the lungs [9,10]. Changes in the taxonomic composition of respiratory microbiota have been evaluated in patients with various pulmonary disorders: COPD [11], asthma [12], community-acquired pneumonia [13], cystic fibrosis [14], lung cancer [15], and idiopathic pulmonary fibrosis [16].

In the current study, the association between CWP and the taxonomic composition of respiratory microbiota was investigated. These changes may, in turn, be associated with progressive pulmonary fibrosis, which is regarded as the main clinical and pathological feature of CWP. Recently, several reports have highlighted the role of the microbiota in fibrosis affecting several human organs, i.e., intestine [17], cardiac tissue [18], liver [19,20], skin [21], and breast tissue [22,23]. Thus, data on the changes in the microbiome composition in the respiratory tract during CWP pathogenesis may be of importance to clarifying the role of microbiota in lung fibrosis.

To test this hypothesis, an initial analysis of the taxonomic composition of the sputum microbiome of coal miners suffering from CWP and from healthy subjects was performed using 16S ribosomal RNA sequencing.

2. Methods

2.1. Cohort Information

The composition of the bacterial microbiome in sputum samples was studied in 35 patients with CWP diagnosis (men only, average age 58.5 \pm 8.3 years) who were admitted to the Department of Occupational Disease Pathology, Kemerovo Regional Clinical Hospital (Kemerovo, Russian Federation). The diagnosis of CWP (code J60 according to ICD-10) was made on the basis of chest x-ray and spirography. All patients worked as underground coal miners. Of these, 8 (22.9%) were actively working in coal mines at the time of the survey, and 27 participants (77.1%) had ceased working due to the onset of the disease. Mining work experience in CWP patients varied from 18 to 37 years (average value 26.9 \pm 5.3 years). As a control group, we examined 35 healthy male donors at a blood transfusion station, who were residents of Kemerovo (average age 55.7 ± 11.72 years). There were no significant age differences between CWP and controls. Among CWP patients there was one active smoker and among the controls all were non-smokers. The summarized information on CWP patients and controls is shown in Table 1. An individual questionnaire was filled out for each survey participant, containing information about the place and date of birth, profession, exposure to occupational hazards, health status, diet features, medications, X-ray records, and harmful habits (smoking and alcohol use). In the study, inclusion criteria were male \geq 40 years of age and exclusive criteria any acute or chronic condition that would limit the ability of the patient to participate in the study, use of antibiotics within 4 weeks prior to collection, failure to obtain a sputum sample.

Variables	Coal Worker's Pneumoconiosis, <i>n</i> = 35	Healthy Kemerovo Residents (Control), <i>n</i> = 35	
Age (years) (mean \pm SD)	58.5 ± 8.3	55.7 ± 11.7	
Stage of CWP (%):			
Ι	88.6		
II	11.4	_	
Respiratory symptoms (%):			
Cough	22.9	8.6	
Dyspnea	5.7	5.7	
Chronic diseases (%):			
Heart and vessels	34.3	34.3	
Bronchitis	14.3	5.7	
Stomach	14.3	11.4	
Diabetes	5.7	5.7	
Obesity	5.7	2.9	
Living environment (%):			
City	94.3	91.4	
Village	5.7	8.6	
Coal dust occupational exposure (%):			
Current miners	23.0		
Former miners	77.0	-	
Smoking status (%):			
Non-smokers	97.1	94.3	
Smokers	2.9	5.7	
Alcohol status (%):			
Non-drinker	14.3	20.0	
Rare drinker	77.1	48.5	
Medium drinker	8.6	31.5	
Diet (%):			
Vegetarian	0	0	
Non-vegetarian	100	100	

Table 1. Characteristics of the study cohorts.

2.2. Sample Collection, Processing and Storage

To analyze the composition of the respiratory tract microbiome, samples of 2–3 mL sputum from CWP patients and controls were obtained over a period of 15–30 min, prior to all diagnostic or therapeutic procedures. Sputum samples were collected non-invasively through participant-induced coughing, (i.e., without induction) and represented the oropharyngeal secretion. Before sputum collection, all patients and controls were asked to rinse their mouths. Microscopic examination of Giemsa-stained cytological slides was used to and confirm the presence of columnar airway epithelial cells in random sputum samples. The resulting samples were immediately placed in sterile plastic vials and frozen (-20 °C). Frozen samples were transported to the laboratory and stored at -80 °C.

2.3. DNA Extraction, 16S rRNA Amplification and 16S rRNA Sequencing

Procaryotic DNA samples were extracted using FastDNA Spin Kit For Soil (MP Biomedicals) based on the manufacturer's recommendation. From each sample, 500 μ L of sputum was used for DNA extraction. The DNA concentration was monitored using the Qubit[®] dsDNA Assay Kit and the Qubit[®] Fluorometer (Life Technologies, CA, USA). Fifty nanograms of each of the extracted and purified sputum DNAs were used for the subsequent amplification of 16S rRNA genes. Amplification of 16S rRNA was carried out according to the Illumina protocol «Preparing 16S Ribosomal RNA Gene Amplicons for the Illumina MiSeq System». The approximately 500 bp long 16S rRNA amplicons consisted of a fragment within the hypervariable (V3–V4) region of the bacterial 16S rRNA genes. The initial PCR was performed with broad-spectrum 16S rRNA primers. In the next round of PCR, with the index-containing primers, fragments of approximately 630 bp length were produced. The Illumina 600 cycle MiSeq Reagent Kit V3 was used. As negative control, a contamination control sample without biological material was included at the sample preparation stage. DNA extraction was then performed in parallel from all samples. Next, the contamination control underwent 16S amplification and was visualized on a gel. As a positive control, a ZYMO community control probe was used (ZymoBIOMICSTM Microbial Community Standards, Cat No D6300 and D6310, Tustin, CA, USA).

The following 16S rRNA primers were used: forward primer: 5'-TCGTCGGCAGCGTCAGATGTGTATAAGAGACAGCCTACGGGNGGCWGCAG-3'. Reverse primer: 5'-GTCTCGTGGGCTCGGAGATGTGTATAAGAGACAGGACTACHVGGG TATCTAATCC-3'

Amplification was performed using BioMaster Hi-Fi LR 2× ReadyMix DNA polymerase (BiolabMix company, Novosibirsk, Russia). The primer sequence was taken from the recommended library preparation protocol for sequencing on the MiSeq platform https://support.illumina.com/documents/documentation/chemistry_documenta tion/16s/16s-metagenomic-library-prep-guide-15044223-b.pdf (accessed on 27 November 2013).

Cycle conditions were 94 °C (3 min 30 s), followed by 25 cycles of 94 °C (30 s), 55 °C (30 s), 68 °C (40 s), and a final extension at 68 °C (5 min). Libraries (~550 bp) were purified using Agencourt AMPure XP beads (Beckman Coulter, Brea, CA, USA). Dual indices and Illumina sequencing adapters from the Illumina Nextera XT index kits v2 B and C (Illumina, San Diego, CA, USA) were added to the target amplicons in a second PCR step using BioMaster Hi-Fi LR $2 \times$ ReadyMix DNA polymerase (BiolabMix company, Novosibirsk, Russia). Cycle conditions were 94 °C (3 min 30 s), then 8 cycles of 94 °C (30 s), 55 °C (30 s), 68 °C (40 s), and a final extension of 68 °C (5 min). Libraries (~630 bp) were again purified using XP beads. Preparation of 16S rRNA libraries was completed according to the Illumina 16S metagenomic sequencing library protocol. Sample PCR products were pooled in equimolar ratio, purified using XP Beads, and quantified using a fluorometer (Quantus Fluorometer dsDNA, Promega, Madison, WI, USA). Molarity of the libraries was brought to 4 nM, and the libraries were denatured and diluted to a final concentration of 8 pM with a 10% PhiX-DNA spike to align sequencing on the Illumina MiSeq (MiSeq Reagent Kit V3, 600 cycles) [24].

2.4. Taxonomy Quantification Using 16S rRNA Gene Sequences and Statistical Methods

The resulting data were processed using the program QIIME2 [25]. A quality check was carried out with the DADA2 software, the default parameters were used [26] and a sequence library was generated.

Amplicon sequence variants (ASV) sequences were performed with QIIME2, using naive Bayesian classifier models based on a 99% nucleotide composition similarity threshold using the Greengenes (versions 13-8) and SILVA (version 132) reference sequence library, followed by removing singletons (ASVs containing only one sequence).

The total diversity of prokaryotic sputum communities (alpha diversity) is estimated by the number of allocated ASVs (analog of species richness) and Shannon indices $(H = \Sigma p_i \ln p_i, p_i-part of$ *i*-sh species in community) [27]. When calculating sample diversity indices, 498 sequences were normalized (the minimum number of received sequences per sample). The variation in the structure of the bacterial community of different samples (beta diversity) was also analyzed using UniFrac [27]—a method common in microbial ecology that estimates the difference between communities based on the phylogenetic relationships of the presented taxa. We used a version of the weighted UniFrac method. The significance of differences between groups of samples was evaluated by the PERMANOVA method (Adonis). Principal Coordinate Analysis (PCOA) graph construction was carried out by using QIIME2 package.

In addition, to assess the significance of differences in the relative percentage of individual bacterial taxa in the sputum samples, the Mann–Whitney U test was used. Spearman's correlation coefficient was used to calculate correlations [28]. Calculations were performed using the software package STATISTICA.10, Statsoft, Tulsa, OK, USA. The false discovery rate (FDR) correction was used to assess the significance of differences in the relative percentages of individual bacterial taxa taking into account multiple comparisons. Multiple linear regression (MLR) was performed to predict the relationship between the relative abundance of individual bacteria in the sputum of CWP patients and lifestyle/disease factors. Additionally, linear discriminant analysis (LDA), effect size (LEfSe), as well as ANCOMBC analyses, were used to normalize data on observed microbial abundance.

Sequence data has been submitted and is archived in a public NCBI database. Archive number PRJNA820569.

3. Results

In the current study using 16S rRNA sequencing (V3–V4) in CWP patients and control subjects using sputum samples, a total of eight phyla with relative frequencies above 0.1% were identified. The prevailing phyla in our dataset were Firmicutes, Bacteroidetes, Actinobacteria, and Proteobacteria (Figure 1), as expected from previous studies [29,30]. For these eight types of bacteria, there were no differences between patients and controls (Table 2). ASVs in CWP subjects amounted to 142.0 ± 56.04 , while in the control group– $147.43 \pm 43.6 \ (p > 0.05)$. Regarding alpha diversity, neither the number of allocated ASVs nor the Shannon indices showed any significant differences between CWP and controls. Overall, bacterial communities in the two groups in the study were fairly diverse as indicated by the Shannon index at the genus level (6.466 in CWP vs. 6.545 in controls).



Figure 1. Taxonomic structure of the sputum microbiomes from CWP subjects and controls at the phyla level.

Phyla	Controls, Mean \pm SD	CWP, Mean \pm SD	p Value
Firmicutes	47.46 ± 11.25	50.41 ± 14.1	0.39
Bacteroidetes	22.93 ± 10.88	19.83 ± 11.51	0.29
Actinobacteria	10.1 ± 7.91	8.99 ± 5.76	0.85
Proteobacteria	8.24 ± 6.54	11.45 ± 16.56	0.9
Fusobacteria	6.98 ± 4.45	4.56 ± 3.96	0.02
TM7	1.8 ± 1.68	1.53 ± 1.84	0.44
Spirochaetes	0.53 ± 0.96	0.6 ± 0.99	0.53
Tenericutes	0.18 ± 0.74	0.23 ± 0.41	0.1

Table 2. Average percentage abundance of phyla present in «core» microbiome.

Differences in the structure of bacterial communities in sputum samples of CWP and controls are shown in Figure 2. The PERMANOVA (Adonis) test using the distance matrix constructed by the weighted UniFrac method showed a significant difference in the prokaryotic communities in the sputum of healthy subjects as compared to miners with CWP patients (pseudo-F = 3.61; p = 0.004). Sequencing statistics for 22 genera (with no less than 0.1% relative frequency) are summarised in Table 3, alongside the corresponding U-rank Mann–Whitney p values and also taking into account the FDR amendment. Among genera, *Streptococcus, Prevotella (f. Prevotellaceae), Veillonella*, and *Anaerosinus* were the most common in the two pools.

Table 3. Average percentage abundance of genera present in «core» microbiome.

Genus	Controls	CWP	p Value	<i>p</i> Value (FDR)
_	$\mathbf{Mean} \pm \mathbf{SD}$	$\mathbf{Mean} \pm \mathbf{SD}$		
Streptococcus	16.85 ± 11.35	25.12 ± 11.37	0.0003 *	0.0023
Prevotella (f. Prevotellfceae)	15.36 ± 7.83	13.24 ± 8.06	>0.05	0.0114
Veillonella	15.4 ± 11.82	12.03 ± 10.14	>0.05	0.0136
Anaerosinus	15.36 ± 11.86	11.56 ± 10.18	>0.05	0.0159
Selenomonas	4.6 ± 4.39	2.07 ± 2.1	0.02	0.0045
Porphyromonas	3.28 ± 3.5	3.26 ± 5.84	>0.05	0.0182
Actinomyces	5.95 ± 6.75	4.24 ± 3.23	>0.05	0.0205
Megasphaera	2.46 ± 1.43	1.45 ± 1.6	>0.05	0.0227
Alloprevotella	2.2 ± 2.29	2.78 ± 2.92	>0.05	0.0250
Streptobacillus	2.98 ± 2.99	2.3 ± 2.88	>0.05	0.0273
Leptotrichia	3.04 ± 3.01	2.34 ± 2.74	>0.05	0.0295
Granulicatella	1.01 ± 1.39	1.71 ± 1.68	0.03	0.0068
Gemella	2.03 ± 1.93	3.12 ± 2.19	>0.05	0.0318
Rothia	2.27 ± 3.02	2.65 ± 3.05	>0.05	0.0341
Bacillus	1.67 ± 1.7	2.67 ± 2.19	0.04	0.0091
Atopobium	1.41 ± 1.91	1.43 ± 1.39	>0.05	0.0364
Pasteurellaceae	2.34 ± 1.21	3.27 ± 2.35	>0.05	0.0386
Fusobacterium	1.91 ± 1.82	1.65 ± 1.42	>0.05	0.0409
Macellibacteroides	2.01 ± 2.58	1.37 ± 2.58	>0.05	0.0432
Neisseria	3.83 ± 4.91	4.72 ± 12.27	>0.05	0.0454
Bacteroides	1.15 ± 1.6	0.82 ± 1.78	>0.05	0.0477
Prevotella (f. Paraprevotellacacea)	1.17 ± 1.69	0.99 ± 1.41	>0.05	0.0500

Note: * *p* Value lesser than FDR corrected *p*.



Axis 1 (7.77 %)

Figure 2. Two-dimensional diagram constructed by the method of principal components analysis demonstrating the phylogenetic similarity of prokaryotic sputum communities in CWP subjects and controls.

In the sputum of CWP subjects, compared to controls, there was an increase in abundance (by percentage) of the following genera: *Streptococcus* (25.12 ± 11.37 vs. 16.85 ± 11.35 ; p = 0.0003); *Granulicatella* (1.71 ± 1.68 vs. 1.01 ± 1.39 ; p = 0.03) and *Bacillus* (2.67 ± 2.19 vs. 1.67 ± 1.7 ; p = 0.04). At the same time, the genus *Selenomonas* was less represented in the microbiomes of CWP subjects compared to the controls (2.07 ± 2.1 vs. 4.6 ± 4.39 ; p = 0.02). It should be noted that taking into account the FDR correction for multiple comparisons, only differences in *Streptococcus* abundance reached the confidence threshold.

In addition, we applied methods of discriminant analysis to find the most significant characteristics that distinguish groups of data from each other. The LEfSe method revealed an increase in the abundance of a number of taxa both in the CWP group (red) and in the control group (green) (Figure 3). The ANCOMBC method revealed an increase in the absolute abundance, in the CWP group relative to the control, for *Streptococcus* (beta = 1.86, W = 3.94, SE = 0.47, q-value = 0.01).



Figure 3. Barplots representing the effect size for particular taxa in sputum samples in the CWP and control groups. LDA-linear discriminant analysis.

Sequencing statistics for 31 species (which met with a relative frequency of not less than 0.1%) are summarized in Table 4, alongside the corresponding U-rank Mann–Whitney p values and also taking into account the FDR amendment. The only significant difference between the sputum microbiome of CWP patients and controls may be the presence of the most common species, *Streptococcus agalactiae*. The average percentage of *Streptococcus agalactiae* in CWP sputum samples was significantly higher than in controls (25.32 ± 11.55 vs. 16.93 ± 10.92; p = 0.0002).

Given the fact that age can influence the composition of the microbiome, Spearman's coefficient to assess the effect of age on the representation of all bacterial taxa was used. In the total sample collection (CWP patients and control subjects), an age-correlated increase in the abundance of the genus *Streptococcus* was observed; however, the significance of this increase was found only at the confidence limit (r = 0.217; p = 0.076). The correlation between age and the representation of other genera in the sputum samples was also insignificant.

To assess the possible correlations of disease duration to the content of different bacteria in sputum, Spearman's coefficient was used. No significant correlations were found between the duration of the disease and the prevalence of any bacterial genus or species. In addition to the duration of pneumoconiosis, patients were further divided into two subgroups: working miners (n = 8) and former miners (n = 27), who stopped working due to the onset of the disease. We were also unable to find significant differences in the abundance of any genus or species of bacteria between these subgroups.

Species	Controls	CWP	p Value	<i>p</i> Value (FDR)
	$\mathbf{Mean} \pm \mathbf{SD}$	$\mathbf{Mean} \pm \mathbf{SD}$		
Streptococcus agalactiae	16.93 ± 10.92	25.32 ± 11.55	0.0002 *	0.0016
Anaerosinus glycerini	14.27 ± 12.03	11.5 ± 10.22	>0.05	0.0048
Selenomonas bovis	4.28 ± 4.28	1.98 ± 1.99	0.03	0.0032
Megasphaera micronuciformis	2.35 ± 2.87	1.66 ± 1.71	>0.05	0.0065
Prevotella histicola	1.44 ± 1.92	2.64 ± 3.33	>0.05	0.0081
Actinomyces hyovaginalis	4.51 ± 5.78	2.19 ± 1.67	>0.05	0.0097
Granulicatella balaenopterae	1.63 ± 1.54	1.69 ± 1.66	>0.05	0.0113
Atopobium rimae	1.41 ± 1.91	1.44 ± 1.38	>0.05	0.0129
Prevotella pallens	1.54 ± 1.96	0.92 ± 1.74	>0.05	0.0145
Rothia terrae	1.84 ± 2.56	2.44 ± 2.97	>0.05	0.0161
Macellibacteroides fermentans	1.93 ± 2.43	1.38 ± 2.59	>0.05	0.0177
Bacteroides nordii	1.25 ± 1.58	0.75 ± 1.56	>0.05	0.0194
Prevotella tannarae	0.52 ± 1.08	0.7 ± 1.27	>0.05	0.0210
Lachnoanaerobaculum orale	0.56 ± 0.74	0.55 ± 0.63	>0.05	0.0226
Prevotella intermedia	0.44 ± 1.01	0.48 ± 0.7	>0.05	0.0242
Prevotella nigrescens	0.37 ± 0.61	0.22 ± 0.39	>0.05	0.0258
Prevotella nanceiensis	0.61 ± 1.09	0.43 ± 0.58	>0.05	0.0274
Bulleidia moorei	0.23 ± 0.3	0.29 ± 0.43	>0.05	0.0290
Clostridium bolteae	0.22 ± 0.5	0.32 ± 0.56	>0.05	0.0306
Porphyromonas endodontalis	0.67 ± 1.42	1.49 ± 5.07	>0.05	0.0323
Vestibaculum illigatum	0.5 ± 1.27	0.31 ± 0.65	>0.05	0.0339
Clostridium acidurici	0.27 ± 0.8	0.26 ± 0.73	>0.05	0.0355
Mycoplasma zalophi	0.3 ± 0.87	0.29 ± 0.43	>0.05	0.0371
Moryella indoligenes	0.27 ± 0.64	0.2 ± 0.53	>0.05	0.0387
Peptostreptococcus anaerobius	0.2 ± 0.54	0.23 ± 0.49	>0.05	0.0403
Treponema amulovorum	0.11 ± 0.24	0.07 ± 0.22	>0.05	0.0419
Oribacterium sinus	0.39 ± 0.61	0.13 ± 0.27	>0.05	0.0435
Leptotrichia trevisanii	0.1 ± 0.25	0.14 ± 0.31	> 0.05	0.0452
Lactobacillus hamsteri	0.09 ± 0.43	0.11 ± 0.62	>0.05	0.0468
Bergeyella zoohelcum	0.32 ± 0.78	0.13 ± 0.2	>0.05	0.0484
Campylobacter rectus	0.08 ± 0.2	0.11 ± 0.24	>0.05	0.0500

Table 4. Average percentage abundance of species present in «core» microbiome.

Note: * *p* Value lesser than FDR corrected *p*.

In addition, multiple linear regression (MLR) was performed to predict the relationship between the relative abundance of individual bacteria in CWP patients' sputum and lifestyle/disease factors. As potential confounders, we considered age, current respiratory symptoms (cough, shortness of breath), living environment, current coal dust occupational exposure, alcohol use, cardiovascular disease (ischemia, arterial hypertension, etc.), bronchitis, stomach diseases (ulcer, gastritis, etc.), diabetes and obesity. In MLR using these potential confounders for patients, a significant regression was found for g. *Streptococcus* (*p*-value 0.007), *Streptococcus agalactiae* (*p*-value 0.005) with cough as the current respiratory symptom (*p*-value 0.03).

4. Discussion

Differences in bacterial populations in the human airways in health and disease have already been recognized as a possible contributing factor in the pathogenesis of diseases of the respiratory tract; however, the bacterial population in the airways of subjects diagnosed with coal worker's pneumoconiosis (CWP) has not yet been investigated. Only a single recent publication reported increased levels of *Lachnospiraceae* and *Lachnoclostridium* in patients with silicosis compared to controls, but this was only demonstrated in the gut microbiome [31]. In assessing beta diversity, in the current study we observed an increase in the genus *Streptococcus* and, in particular, of the *Streptococcus agalactiae* species in individuals with CWP diagnosis (Tables 3 and 4).

Streptococcus agalactiae (also known as group B streptococcus or GBS) is an important opportunistic bacterium that can cause pneumonia, sepsis, and meningitis in newborns and in patients with weakened immunity [32,33]. Cases of invasive GBS infections are frequently reported in the elderly and immunocompromised adults, including patients with diabetes mellitus, alcoholism, and cancer [34]. In the respiratory tract, GBS sometimes contributes to community-acquired pneumonia and empyema in adults [35]. When GBS causes pulmonary infections, it is usually defined as part of polymicrobial pneumonia [36]. GBS bacteria effectively attach to pulmonary epithelial cells and are capable of invasion. This is initiated by attachment to extracellular matrix molecules such as agglutinin, fibronectin, fibrinogen, and laminin, which facilitate their attachment to host cell surface proteins, such as integrins. Thus, the invasive potential of GBS is influenced by changes in the surface proteome of the host cells, which can be caused by various lung pathologies [37]. The molecular mechanisms of cytopathology caused by GBS bacteria in patients are currently under intensive investigation. It was shown that GBS induces the generation of reactive oxygen species (ROS) and loss of mitochondrial membrane potential [38]. In human endothelial cells, ROS species are generated via the NADPH oxidase pathway, and this is accompanied by cytoskeletal reorganization through the PI3K/Akt pathway and is generally associated with pathogen penetration, which provides evidence for the involvement of oxidative stress in S. agalactiae pathogenesis [39]. Separately, it is important to point out that the 97% sequence similarity threshold that we used based on the results of sequencing gives good reliability values for resolution at the genus level, but is insufficient for resolution at the species level, especially for streptococcal species that have a highly homologous 16s rRNA gene sequence. For this reason, at this stage of the study, we cannot unequivocally state that the representation of *S. agalactiae* is increased in the sputum of patients, since our 16S rRNA sequences may also include other types of streptococci. In the future, the use of quantitative PCR will allow us to answer this question.

In addition to pneumoconiosis, diseases leading to tissue fibrosis in the lungs include idiopathic fibrosis and cystic fibrosis. An increase in the abundance of *Streptococcus* in samples of the airway microbiome has been found for patients with idiopathic pulmonary fibrosis [40,41], but not for patients with cystic fibrosis [42]. Data on *Streptococcus* in the respiratory tract of COPD patients remains controversial. Different investigators report on the one hand, an increase in the representation of *Streptococcus* in the microbiota of patients with COPD compared to healthy controls [43,44] and on the other hand that no differences are observed in the abundance of this taxa with regard to this pathology [45–47].

Since pneumoconiosis is associated with the risk of cancer [7], it is interesting to consider the presence of *Streptococcus* in the respiratory tract of lung cancer patients. We recently reported a significant increase in the representation of *Streptococcus, Bacillus, Gemella*, and *Haemophilus* in the sputum of lung cancer patients compared to control subjects [48]. In addition to a significant increase in *Streptococcus*, we also observe an increase in the representation of the genera Bacillus and Gemella in the sputum of miners with CWP compared to control subjects (Table 3). However, this increase did not reach the threshold of significance in our study. In previous studies of patients with lung cancer, it was found that airway brushings of lung cancer patients show an increase in *Streptococcus* spp. along with a decrease in α -diversity in the affected lung microbiome, as compared with brushings from the contralateral non-cancerous lung [49]. An increase in the prevalence of Streptococcus was reported for sputum samples from LC patients [30,50,51], BALF samples [52], bronchoscopy [53], saliva [54], and lower airway [55]. Thus, it can be suggested that bacteria of the *Streptococcus* genus in the respiratory tract microbiome play an important role in the pathogenesis of many diseases of the bronchopulmonary system by inducing oxidative stress, inflammation, and genome instability. As follows from our data, these diseases also include pneumoconiosis induced by coal and silicate dust.

Current analysis showed no significant age-dependent differences in the representation of the *Streptococcus* taxa but that exposure to occupational hazards, such as coal dust may be the causative factor for the variation in sputum microbiomes. At this stage of the study, we did not establish a relationship between the composition of the sputum microbiome and the duration of CWP disease, and no difference in the content of certain bacterial taxa in the sputum of active and former miners was detected. It may be expected that further research based on larger study populations, will help to significantly increase the confidence levels for these factors and will contribute to a better understanding of the role of the respiratory tract microbiome in the pathogenesis of CWP.

5. Conclusions

This pilot study for the first time presents the results of an analysis of sputum microbiomes in a small group of current and former coal miners suffering from CWP, living in Kuzbass, a coal-mining region in Russia. The method of massive parallel sequencing of 16S rRNA genes has been used for the first time to obtain a taxonomic characteristic of the sputum samples from the coal miners with CWP diagnosis. A significant increase in *Streptococcus* spp. in the sputum of CWP patients compared to controls was found. These changes in microbial communities may contribute to the development of CWP. Based on these findings, future studies will be focused on the specific interactions between pathogen and host involved in CWP pathogenesis. In particular, digital droplet PCR will be used to test the predictability of *Streptococcus* and *Streptococcus agalactiae* (or other Streptococcal species) abundance in sputum samples from CWP patients and controls.

Author Contributions: V.G.D. and L.V.M. conceived the study; V.G.D. and L.V.M. wrote the manuscript; E.D.B., V.P.V., S.A.P. performed laboratory work; P.S.D. carried out bioinformatics and statistical analyses; V.I.M. provided samples and managed clinical data. A.V.L. critically revised the manuscript. All authors have read and agreed to the published version of the manuscript.

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Institutional Review Board Statement: The study was conducted in accordance with the Declaration of Helsinki (1964, amended 2008) of the World Medical Association, and approved by the Ethics Committee of Kemerovo State University (Protocol Code No. 17/2021; 5 April 2021).

Informed Consent Statement: Consent for publication was obtained from each person in the study. All participants (CWP patients and controls) were informed about the aim, methodology, and possible risks of the study; informed consent was signed by each donor. All authors approved the final version of the manuscript.

Data Availability Statement: The datasets obtained and analyzed during the current study are available from the corresponding author on request.

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References

- 1. Han, S.; Chen, H.; Harvey, M.A.; Stemn, E.; Cliff, D. Focusing on Coal Workers' Lung Diseases: A Comparative Analysis of China, Australia, and the United States. *Int. J. Environ. Res. Public Health* **2018**, *15*, 2565. [CrossRef]
- Blackley, D.J.; Halldin, C.N.; Laney, A.S. Continued Increase in Prevalence of Coal Workers' Pneumoconiosis in the United States, 1970–2017. Am. J. Public Health 2018, 108, 1220–1222. [CrossRef]
- Leonard, R.; Zulfikar, R.; Stansbury, R. Coal mining and lung disease in the 21st century. *Curr. Opin. Pulm. Med.* 2020, 26, 135–141. [CrossRef]

- Beer, C.; Kolstad, H.A.; Søndergaard, K.; Bendstrup, E.; Heederik, D.; Olsen, K.E.; Omland, Ø.; Petsonk, E.; Sigsgaard, T.; Sherson, D.L.; et al. A systematic review of occupational exposure to coal dust and the risk of interstitial lung diseases. *Eur. Clin. Respir. J.* 2017, *4*, 1264711. [CrossRef] [PubMed]
- Tomaskova, H.; Jirak, Z.; Splichalova, A.; Urban, P. Cancer incidence in Czech black coal miners in association with coalworkers' pneumoconiosis. Int. J. Occup. Med. Environ. Health 2012, 25, 137–1344. [CrossRef] [PubMed]
- Volobaev, V.P.; Sinitsky, M.Y.; Larionov, A.V.; Druzhinin, V.G.; Gafarov, N.I.; Minina, V.I.; Kulemin, J.E. Modifying influence of occupational inflammatory diseases on the level of chromosome aberrations in coal miners. *Mutagenesis* 2016, *31*, 225–229. [CrossRef] [PubMed]
- Druzhinin, V.G.; Apalko, S.V.; Baranova, E.D.; Volobaev, V.P.; Drobchik, T.Y.; Larionov, A.V.; Hill, E.G.; Chasovskikh, E.V. Micronuclei in blood lumphocytes of existing and former coal miners: Evaluation of the effect of anthracosilicosis. *Ecol. Genet.* 2019, 17, 55–62. (In Russian) [CrossRef]
- Zhang, N.; Liu, K.; Wang, K.; Zhou, C.; Wang, H.; Che, S.; Liu, Z.; Yang, H. Dust induces lung fibrosis through dysregulated DNA methylation. *Environ. Toxicol.* 2019, 34, 728–741. [CrossRef]
- Dickson, R.P.; Erb-Downward, J.R.; Martinez, F.J.; Huffnagle, G.B. The Microbiome and the Respiratory Tract. *Annu. Rev. Physiol.* 2016, 78, 481–504. [CrossRef]
- 10. Caverly, L.J.; Huang, Y.J.; Sze, M.A. Past, Present, and Future Research on the Lung Microbiome in Inflammatory Airway Disease. *Chest* **2019**, *156*, 376–382. [CrossRef]
- 11. Qi, Y.J.; Sun, X.J.; Wang, Z.; Bin, Y.F.; Li, Y.H.; Zhong, X.N.; Bai, J.; Deng, J.M.; He, Z.Y. Richness of sputum microbiome in acute exacerbations of eosinophilic chronic obstructive pulmonary disease. *Chin. Med. J.* 2020, 133, 542–551. [CrossRef] [PubMed]
- 12. Kozik, A.J.; Huang, Y.J. The microbiome in asthma: Role in pathogenesis, phenotype, and response to treatment. *Ann. Allergy Asthma Immunol.* **2019**, 122, 270–275. [CrossRef]
- Wootton, D.G.; Cox, M.J.; Gloor, G.B.; Litt, D.; Hoschler, K.; German, E.; Court, J.; Eneje, O.; Keogan, L.; Macfarlane, L.; et al. A *Haemophilus* sp. dominates the microbiota of sputum from UK adults with non-severe community acquired pneumonia and chronic lung disease. *Sci. Rep.* 2019, *9*, 2388. [CrossRef] [PubMed]
- 14. Acosta, N.; Heirali, A.; Somayaji, R.; Surette, M.G.; Workentine, M.L.; Sibley, C.D.; Rabin, H.R.; Parkins, M.D. Sputum microbiota is predictive of long-term clinical outcomes in young adults with cystic fibrosis. *Thorax* 2018, 73, 1016–1025. [CrossRef] [PubMed]
- 15. Maddi, A.; Sabharwal, A.; Violante, T.; Manuballa, S.; Genco, R.; Patnaik, S.; Yendamuri, S. The microbiome and lung cancer. *J. Thorac. Dis.* **2019**, *11*, 280–291. [CrossRef]
- 16. Hewitt, R.J.; Molyneaux, P.L. The respiratory microbiome in idiopathic pulmonary fibrosis. *Ann. Transl. Med.* **2017**, *5*, 250. [CrossRef]
- Jacob, N.; Jacobs, J.P.; Kumagai, K.; Ha, C.W.Y.; Kanazawa, Y.; Lagishetty, V.; Altmayer, K.; Hamill, A.M.; Von Arx, A.; Sartor, R.B.; et al. Inflammation-independent TL1A-mediated intestinal fibrosis is dependent on the gut microbiome. *Mucosal Immunol.* 2018, 11, 1466–1476. [CrossRef]
- 18. Li, X.; Geng, J.; Zhao, J.; Ni, Q.; Zhao, C.; Zheng, Y.; Chen, X.; Wang, L. Trimethylamine N-Oxide Exacerbates Cardiac Fibrosis via Activating the NLRP3 Inflammasome. *Front. Physiol.* **2019**, *10*, 866. [CrossRef]
- Moghadamrad, S.; Hassan, M.; McCoy, K.D.; Kirundi, J.; Kellmann, P.; De Gottardi, A. Attenuated fibrosis in specific pathogen-free microbiota in experimental cholestasis- and toxin-induced liver injury. *FASEB J.* 2019, 33, 12464–12476. [CrossRef]
- Weiskirchen, R.; Weiskirchen, S.; Tacke, F. Recent advances in understanding liver fibrosis: Bridging basic science and individualized treatment concepts. *F1000Research* 2018, 27, F1000 Faculty Rev-921. [CrossRef]
- 21. Tagliari, E.; Campos, L.F.; Campos, A.C.; Costa-Casagrande, T.A.; Noronha, L. Effect of probiotic oral administration on skin wound healing in raTS. *Arq. Bras. Cir. Dig.* **2019**, *32*, e1457. [CrossRef] [PubMed]
- Buchta Rosean, C.; Bostic, R.R.; Ferey, J.C.M.; Feng, T.Y.; Azar, F.N.; Tung, K.S.; Dozmorov, M.G.; Smirnova, E.; Bos, P.D.; Rutkowski, M.R. Preexisting Commensal Dysbiosis Is a Host-Intrinsic Regulator of Tissue Inflammation and Tumor Cell Dissemination in Hormone Receptor-Positive Breast Cancer. *Cancer Res.* 2019, *79*, 3662–3675. [CrossRef] [PubMed]
- Yang, J.; Ding, C.; Dai, X.; Lv, T.; Xie, T.; Zhang, T.; Gao, W.; Gong, J.; Zhu, W.; Li, N.; et al. Soluble Dietary Fiber Ameliorates Radiation-Induced Intestinal Epithelial-to-Mesenchymal Transition and Fibrosis. *JPEN J. Parenter. Enteral. Nutr.* 2017, 41, 1399–1410. [CrossRef] [PubMed]
- Caporaso, J.G.; Kuczynski, J.; Stombaugh, J.; Bittinger, K.; Bushman, F.D.; Costello, E.K.; Fierer, N.; Peña, A.G.; Goodrich, J.K.; Gordon, J.I.; et al. QIIME allows analysis of high-throughput community sequencing data. *Nat. Methods* 2010, 7, 335–336. [CrossRef]
- Bolyen, E.; Rideout, J.R.; Dillon, M.R.; Bokulich, N.A.; Abnet, C.C.; Al-Ghalith, G.A.; Alexander, H.; Alm, E.J.; Arumugam, M.; Asnicar, F.; et al. Reproducible, interactive, scalable and extensible microbiome data science using QIIME 2. *Nat. Biotechnol.* 2019, 37, 1091. [CrossRef]
- Callahan, B.J.; McMurdie, P.J.; Rosen, M.J.; Han, A.W.; Johnson, A.J.; Holmes, S.P. DADA2: High-resolution sample inference from Illumina amplicon data. *Nat. Methods* 2016, 13, 581–583. [CrossRef]
- Lozupone, C.; Knight, R. UniFrac: A new phylogenetic method for comparing microbial communities. *Appl. Environ. Microbiol.* 2005, 71, 8228–8235. [CrossRef]
- 28. Akoglu, H. User's guide to correlation coefficients. Turk. J. Emerg. Med. 2018, 18, 91–93. [CrossRef]

- 29. Dickson, R.P.; Huffnagle, G.B. The Lung Microbiome: New Principles for Respiratory Bacteriology in Health and Disease. *PLoS Pathog.* **2015**, *11*, e1004923. [CrossRef]
- Cameron, S.J.S.; Lewis, K.E.; Huws, S.A.; Hegarty, M.J.; Lewis, P.D.; Pachebat, J.A.; Mur, L.A.J. A pilot study using metagenomic sequencing of the sputum microbiome suggests potential bacterial biomarkers for lung cancer. *PLoS ONE* 2017, 12, e0177062. [CrossRef]
- 31. Zhou, Y.; Chen, L.; Sun, G.; Li, Y.; Huang, R. Alterations in the gut microbiota of patients with silica-induced pulmonary fibrosis. J. Occup. Med. Toxicol. 2019, 14, 5. [CrossRef] [PubMed]
- Manning, S.D. Molecular epidemiology of Streptococcus agalactiae (group B Streptococcus). Front. Biosci. 2003, 1, 1–18. [CrossRef] [PubMed]
- 33. Furfaro, L.L.; Chang, B.J.; Payne, M.S. Perinatal Streptococcus agalactiae Epidemiology and Surveillance Targets. *Clin. Microbiol. Rev.* **2018**, *31*, e00049-18. [CrossRef]
- Pimentel, B.A.; Martins, C.A.; Mendonça, J.C.; Miranda, P.S.; Sanches, G.F.; Mattos-Guaraldi, A.L.; Nagao, P.E. Streptococcus agalactiae infection in cancer patients: A five-year study. *Eur. J. Clin. Microbiol. Infect. Dis.* 2016, 35, 927–933. [CrossRef] [PubMed]
- 35. Phares, C.R.; Lynfield, R.; Farley, M.M.; Mohle-Boetani, J.; Harrison, L.H.; Petit, S.; Craig, A.S.; Schaffner, W.; Zansky, S.M.; Gershman, K.; et al. Active Bacterial Core surveillance/Emerging Infections Program Network. Epidemiology of invasive group B streptococcal disease in the United States, 1999–2005. *JAMA* 2008, 299, 2056–2065. [CrossRef] [PubMed]
- Farley, M.M.; Harvey, R.C.; Stull, T.; Smith, J.D.; Schuchat, A.; Wenger, J.D.; Stephens, D.S. A population-based assessment of invasive disease due to group B Streptococcus in nonpregnant adults. *N. Engl. J. Med.* **1993**, *328*, 1807–1811. [CrossRef]
- 37. Sharma, P.; Lata, H.; Arya, D.K.; Kashyap, A.K.; Kumar, H.; Dua, M.; Ali, A.; Johri, A.K. Role of pilus proteins in adherence and invasion of Streptococcus agalactiae to the lung and cervical epithelial cells. *J. Biol. Chem.* **2013**, *288*, 4023–4034. [CrossRef]
- da Costa, A.F.E.; Moraes, J.A.; de Oliveira, J.S.S.; Dos Santos, M.H.B.; da Silva Santos, G.; Barja-Fidalgo, C.; Mattos-Guaraldi, A.L.; Nagao, P.E. Reactive oxygen species involved in apoptosis induction of human respiratory epithelial (A549) cells by Streptococcus agalactiae. *Microbiology* 2016, 162, 94–99. [CrossRef]
- Oliveira, J.S.S.; Santos, G.D.S.; Moraes, J.A.; Saliba, A.M.; Barja-Fidalgo, T.C.; Mattos-Guaraldi, A.L.; Nagao, P.E. Reactive oxygen species generation mediated by NADPH oxidase and PI3K/Akt pathways contribute to invasion of Streptococcus agalactiae in human endothelial cells. *Mem. Inst. Oswaldo Cruz.* 2018, 113, e140421. [CrossRef]
- 40. Molyneaux, P.L.; Cox, M.J.; Willis-Owen, S.A.; Mallia, P.; Russell, K.E.; Russell, A.M.; Murphy, E.; Johnston, S.L.; Schwartz, D.A.; Wells, A.U.; et al. The role of bacteria in the pathogenesis and progression of idiopathic pulmonary fibrosis. *Am. J. Respir. Crit. Care Med.* **2014**, *190*, 906–913. [CrossRef]
- Han, M.K.; Zhou, Y.; Murray, S.; Tayob, N.; Noth, I.; Lama, V.N.; Moore, B.B.; White, E.S.; Flaherty, K.R.; Huffnagle, G.B.; et al. COMET Investigators. Lung microbiome and disease progression in idiopathic pulmonary fibrosis: An analysis of the COMET study. *Lancet Respir. Med.* 2014, 2, 548–556. [CrossRef]
- 42. Skolnik, K.; Nguyen, A.; Thornton, C.S.; Waddell, B.; Williamson, T.; Rabin, H.R.; Parkins, M.D. Group B streptococcus (GBS) is an important pathogen in human disease- but what about in cystic fibrosis? *BMC Infect. Dis.* **2017**, *17*, 660. [CrossRef]
- 43. Pragman, A.A.; Kim, H.B.; Reilly, C.S.; Wendt, C.; Isaacson, R.E. The lung microbiome in moderate and severe chronic obstructive pulmonary disease. *PLoS ONE* **2012**, *7*, e47305. [CrossRef]
- Einarsson, G.G.; Comer, D.M.; McIlreavey, L.; Parkhill, J.; Ennis, M.; Tunney, M.M.; Elborn, J.S. Community dynamics and the lower airway microbiota in stable chronic obstructive pulmonary disease, smokers and healthy non-smokers. *Thorax* 2016, 71, 795–803. [CrossRef]
- 45. Cabrera-Rubio, R.; Garcia-Núñez, M.; Setó, L.; Antó, J.M.; Moya, A.; Monsó, E.; Mira, A. Microbiome diversity in the bronchial tracts of patients with chronic obstructive pulmonary disease. *J. Clin. Microbiol.* **2012**, *50*, 3562–3568. [CrossRef]
- Haldar, K.; George, L.; Wang, Z.; Mistry, V.; Ramsheh, M.Y.; Free, R.C.; John, C.; Reeve, N.F.; Miller, B.E.; Tal-Singer, R.; et al. The sputum microbiome is distinct between COPD and health, independent of smoking history. *Respir. Res.* 2020, 21, 183. [CrossRef]
- Wang, Z.; Liu, H.; Wang, F.; Yang, Y.; Wang, X.; Chen, B.; Stampfli, M.R.; Zhou, H.; Shu, W.; Brightling, C.E.; et al. A Refined View of Airway Microbiome in Chronic Obstructive Pulmonary Disease at Species and Strain-Levels. *Front. Microbiol.* 2020, 30, 1758. [CrossRef]
- Druzhinin, V.G.; Matskova, L.V.; Demenkov, P.S.; Baranova, E.D.; Volobaev, V.P.; Minina, V.I.; Larionov, A.V.; Titov, V.A.; Fucic, A. Genetic damage in lymphocytes of lung cancer patients is correlated to the composition of the respiratory tract microbiome. *Mutagenesis* 2021, *36*, 143–153. [CrossRef]
- Liu, H.X.; Tao, L.L.; Zhang, J.; Zhu, Y.G.; Zheng, Y.; Liu, D.; Zhou, M.; Ke, H.; Shi, M.M.; Qu, J.M. Difference of lower airway microbiome in bilateral protected specimen brush between lung cancer patients with unilateral lobar masses and control subjects. *Int. J. Cancer.* 2018, 142, 769–778. [CrossRef]
- Hosgood, H.D., 3rd; Sapkota, A.R.; Rothman, N.; Rohan, T.; Hu, W.; Xu, J.; Vermeulen, R.; He, X.; White, J.R.; Wu, G.; et al. The potential role of lung microbiota in lung cancer attributed to household coal burning exposures. *Environ. Mol. Mutagen.* 2014, 55, 643–651. [CrossRef]
- Ran, Z.; Liu, J.; Wang, F.; Xin, C.; Shen, X.; Zeng, S.; Song, Z.; Xiong, B. Analysis of Pulmonary Microbial Diversity in Patients with Advanced Lung Cancer Based on High-throughput Sequencing Technology. *Chin. J. Lung Cancer* 2020, 23, 1031–1038. (In Chinese) [CrossRef]

- Hasegawa, A.; Sato, T.; Hoshikawa, Y.; Ishida, N.; Tanda, N.; Kawamura, Y.; Kondo, T.; Takahashi, N. Detection and identification of oral anaerobes in intraoperative bronchial fluids of patients with pulmonary carcinoma. *Microbiol. Immunol.* 2014, 58, 375–381. [CrossRef]
- 53. Liu, Y.; O'Brien, J.L.; Ajami, N.J.; Scheurer, M.E.; Amirian, E.S.; Armstrong, G.; Tsavachidis, S.; Thrift, A.P.; Jiao, L.; Wong, M.C.; et al. Lung tissue microbial profile in lung cancer is distinct from emphysema. *Am. J. Cancer Res.* **2018**, *8*, 1775–1787.
- Zhang, W.; Luo, J.; Dong, X.; Zhao, S.; Hao, Y.; Peng, C.; Shi, H.; Zhou, Y.; Shan, L.; Sun, Q.; et al. Salivary Microbial Dysbiosis is Associated with Systemic Inflammatory Markers and Predicted Oral Metabolites in Non-Small Cell Lung Cancer Patients. J. Cancer 2019, 10, 1651–1662. [CrossRef]
- 55. Tsay, J.J.; Wu, B.G.; Badri, M.H.; Clemente, J.C.; Shen, N.; Meyn, P.; Li, Y.; Yie, T.A.; Lhakhang, T.; Olsen, E.; et al. Airway Microbiota Is Associated with Upregulation of the PI3K Pathway in Lung Cancer. Am. J. Respir. Crit. Care Med. 2018, 198, 1188–1198. [CrossRef]