Published online 2022 June 21.

**Research Article** 

# Gut Microbial Profile in Patients with Pancreatic Cancer

Xue-Yuan Wang<sup>1</sup>, Zao-Xi Sun<sup>2</sup>, Emmanuel Costantine Makale<sup>2</sup>, Zheng-Ke Sun<sup>2</sup>, Tai-Cheng Wang<sup>3</sup>, Zhi-Ying Hou<sup>4</sup>, Xing-Yue Yu<sup>1</sup>, Wenfang Long<sup>1</sup>, Guan-Kui Du <sup>1</sup><sup>5</sup> and Hairong Huang <sup>1</sup><sup>1,\*</sup>

<sup>1</sup>Department of Epidemiology, International School of Public Health and One Health, Hainan Medical University, Haikou, China

<sup>2</sup>Department of General Surgery of the First Affiliated Hospital, Hainan Medical University, Haikou, China

<sup>3</sup>Department of General Surgery of the Second Affiliated Hospital, Hainan Medical University, Haikou, China

<sup>4</sup>Department of Epidemiology, Kangda College of Nanjing Medical University, Nanjing, China

<sup>5</sup>Department of Biochemistry and Molecular Biology, Hainan Medical University, Haikou, China

<sup>\*</sup> Corresponding author: Department of Epidemiology, International School of Public Health and One Health, Hainan Medical University, Haikou, China. Email: jialidog1@163.com.

Received 2022 January 16; Revised 2022 May 01; Accepted 2022 May 26.

# Abstract

**Background:** Pancreatic cancer is a lethal tumor with a poor prognosis. The connection between pancreatic cancer and gut microbiota is less reported.

**Objectives:** This study analyzed microbial characteristics in patients with pancreatic cancer from the tropical area of China and explored the potential impact of the characteristic microflora on pancreatic cancer.

**Methods:** Stool samples and blood test indices of participants were collected in Hainan, China. Metagenomic sequencing was used to analyze the gut microbiota characteristics. The R corrplot package was used to analyze the correlation between gut microbiota and blood test indices.

**Results:** The microbial community in pancreatic cancer were clustered together and significantly separated from controls. The Simpson index was increased significantly in pancreatic cancer compared to controls. The abundances of butyrate-producing bacteria (Anaerostipes hadrus, *Lachnoclostridium phocaeense*, and *Romboutsia ilealis*), *Bifidobacteria*, and [*Eubacterium*] *eligens* were significantly decreased, while *Fusobacterium*, *Enterobacter*, and *Enterococcus* were significantly increased in pancreatic cancer. *Prevotella copri* may have a vital role in the bacterial interaction network. Pathways connected to metabolism, environment (bacterial secretion system), genetic information (protein export and ribosome), and human diseases (infectious diseases and drug resistance) were increased in the pancreatic cancer group. Butyrate-producing bacteria (butyrate-producing bacterium *SS3/4, A. hadrus, R. intestinalis*, and *Faecalibacterium prausnitzii*) and *Bifidobacteria* were significantly negatively correlated with the neutrophil-to-lymphocyte ratio. **Conclusions:** The gut microbiome was distinct in patients with pancreatic cancer from the tropical area of China. Changes in intestinal flora abundance and metabolic pathways may play an essential role in the occurrence and development of pancreatic cancer.

Keywords: Pancreatic Cancer, Gut Microbiota, Metagenomics, Tropical Area

## 1. Background

Pancreatic cancer is a lethal tumor with a poor prognosis, whose five-year survival rate is less than 10% (1). The incidence rate of pancreatic cancer was 14th, and the mortality rate was seventh worldwide among 36 cancers in 185 countries in 2018 (2). In 2015, the incidence and mortality rates of pancreatic cancer in China were 10th and sixth, respectively (3). The morbidity and mortality of pancreatic cancer in China are steadily increasing (4). Therefore, research on preventing and controlling pancreatic cancer is fundamental.

Recent studies have shown that microorganisms colonized in the human body play an important role in regulating metabolism and immune response (5, 6). Some studies showed that intestinal flora is associated with tumors such as liver cancer and colorectal cancer (7-10). However, the connection between pancreatic cancer and intestinal flora is less reported. Ren et al. showed that gut microbial diversity, probiotics, and butyrate-producing bacteria were decreased, but the opportunistic pathogens and lipopolysaccharides-producing bacteria were increased in pancreatic cancer (11). Half et al. stated that *Veillonellaceae* and *Akkermansia* were significantly increased in pancreatic cancer, so the ratio of Bacteroidetes to Firmicutes was higher in pancreatic cancer patients than in the control groups (12). There are discrepancies in these studies. One of the important reasons for the difference is environmen-

Copyright © 2022, Author(s). This is an open-access article distributed under the terms of the Creative Commons Attribution-NonCommercial 4.0 International License (http://creativecommons.org/licenses/by-nc/4.0/) which permits copy and redistribute the material just in noncommercial usages, provided the original work is properly cited.

tal factors, which might result in discrepant results (13). Therefore, the gut microbial characteristics of clinical pancreatic cancer in different environments are worth exploring. Currently, there is no relevant research in tropical areas. Hainan is the only tropical area in China.

## 2. Objectives

In this study, we took registered residents in Hainan as the research subjects, studied the diversity, composition, and functional prediction of gut microbiota in pancreatic cancer patients by metagenome sequencing technology, and analyzed the correlation between clinical blood test results and the abundance of intestinal flora in pancreatic cancer. The results can be used as additional evidence of the relationship between intestinal bacteria and pancreatic cancer, providing theoretical support for the prevention and treatment of pancreatic cancer.

#### 3. Methods

#### 3.1. Objects

This is a cross-sectional comparative study. Pancreatic cancer patients formed the case group, and non-pancreatic cancer patients established the control group. Pancreatic cancer was diagnosed according to comprehensive guidelines for diagnosing and treating pancreatic cancer (version 2018) (14). The diagnostic methods mainly included tumor markers such as CA19-9, imaging examinations such as CT, MRI, and PET-CT, and pathological biopsy. Both groups were Hainanese individuals, and their dietary habits are mainly seafood, vegetables, and fruits, with low salt and less oil. The two groups were matched for gender, age, nationality, smoking, drinking, and background diseases such as hypertension and diabetes. The participants were recruited from June to December 2020 in the Affiliated Hospital of Hainan Medical University. Inclusion criteria were (1) Hainanese individuals, (2) new patients, (3) dietary habits of mainly seafood, vegetables, and fruits, with low salt and less oil, and (4) informed consent. Exclusion criteria were (1) gastrointestinal diseases such as gastritis and colitis, (2) having other cancers, (3) drugs and/or probiotics consumption within eight weeks before stool collection, and (4) lack of essential information and clinical data such as sex, age, and blood examination.

## 3.2. Materials

The study utilized a questionnaire survey while collecting stool samples. The questionnaire included necessary information such as age, sex, nationality, and information on lifestyle, dietary habits, and background diseases. The disposable fecal collection box collected the fresh fecal samples of about 5 g. They were transferred immediately from the hospital to the laboratory in an ice bag using insulating polystyrene foam containers and then stored immediately at -80°C. Data including routine blood examination and blood biochemistry tests were collected. The indices included Red Blood Cell (RBC) counts, neutrophil counts, lymphocyte counts, hemoglobin (Hgb), hematocrit (HCT), lymphocyte ratio, monocyte ratio, serum sodium (Na<sup>+</sup>), serum calcium (Ca<sup>2+</sup>), Aspartate Aminotransferase (AST), Indirect Bilirubin (IBIL), Neutrophil-to-Lymphocyte Ratio (NLR), and plasma crystal osmotic pressure. Patients' information on drug use, smoking, and drinking was collected. Plasma crystal osmotic pressure (mmol/L) [=2\*(K<sup>+</sup>+Na<sup>+</sup>)+BG+BUN], serum potassium (K<sup>+</sup>), and sodium (Na<sup>+</sup>) were measured by the ion-selective electrode. Blood Glucose (BG) was measured by the glucose oxidase method, and Blood Urea Nitrogen (BUN) was measured by Urea Nitrogen (UN) reagents.

## 3.3. DNA Extraction and Sequencing

The DNA was extracted through a fecal genomic DNA rapid extraction kit (Beijing BioTeke Corporation). The process was as follows: sampling 0.3 g, adding reagents according to the instructions, thoroughly mixing, water bath, centrifugation, purification, elution, and dissolution. After measuring the concentration and purity of the DNA, it was stored in a -80°C refrigerator. Metagenomic sequencing was performed on the Illumina NovaSeq platform at the Shanghai Personal Biotechnology Co., Ltd. Sequencing adopted the Whole Genome Shotgun (WGS) strategy. Briefly, the extracted DNA was randomly broken into short fragments and constructed a library of insert fragments with suitable lengths. Then, paired-end (PE) sequencing was carried out. Fast QC was used for quality control of the raw data generated by sequencing. The original data were screened and filtered to obtain a clean data set that could be used for subsequent analyses. Finally, the genome was assembled by MegaHit.

## 3.4. Bioinformatics Analysis

The PLS-DA model was constructed based on the species composition spectrum and grouping data by R3.3 software. QIIME2 software calculated alpha diversity, including Chao1, ACE, Shannon, and Simpson indices. Wilcoxon rank pair test was also performed by QIIME2 software. The SciPy package of Python 3.6.5 was used to test the relative abundance distribution. The Benjamin-Hochberg method calculated the false discovery rate (FDR). Log2 (fold change value) > 2 and P < 0.05 showed significant differences between the groups. Mothur software calculated the

Spearman rank correlation coefficient among the top 50 dominant species, and the correlation network was constructed for the dominant species with | Rho | > 0.8 and P < 0.01, imported into Cytoscape software for visualization. The LEfSe analysis was carried out through the Galaxy online analysis platform. The R corrplot package analyzed the correlation between gut microbiota and blood test indices. KOBAS annotated the KEGG signaling pathway.

## 3.5. Statistical Analysis

Statistical analysis was done using SPSS version 20 (IBM SPSS Statistics for Windows, Version 20.0. Armonk, NY: IBM Corp). Continuous variables were compared between two groups using a *t*-test. Fisher's exact test compared categorical variables. The alpha value = 0.05 and P < 0.05 showed statistically significant differences.

#### 4. Results

## 4.1. Details of Enrolment

Thirty fecal samples from 15 pancreatic cancer patients and 15 controls were collected. After a strict pathologic diagnosis and exclusion process, eight patients with pancreatic cancer and eight controls were utilized for DNA extraction, library construction, metagenomic sequencing, data quality control, and analysis (Figure 1).

#### 4.2. General Information

Eight pancreatic cancer patients and eight controls were included (Table 1). There were no significant differences in age, gender, smoking, drinking, and nationality between the two groups (P > 0.05).

#### 4.3. Gut Microbial Diversity and PLS-DA Analysis

The raw sequencing data of each sample were collated through the Illumina NovaSeq high-throughput sequencing platform. Sequencing fragments of 16 fecal samples ranged from 10035777600 to 12908544600, with an average of 11029185656.25. The number of base pairs of highquality sequences was 9.98 - 12.8 billion, and the average was 11.0 billion. The percentage of high-quality sequences among the total original sequences was between 99.44% and 99.77%, with an average of 99.6%. The number of reads ranged from 66.9 to 86.1 million, with an average of 73.5 million. The refraction curve showed that the number of reads approached 30 million in both groups, and the sequencing results were sufficient to reflect the species composition of the samples (Figure 2A). Alpha diversity was analyzed (Figure 2B1-4). Figure 2B-4 shows that the Simpson index significantly increased in pancreatic cancer compared to controls (P = 0.021). The PLS-DA analysis showed

that the microbial community in pancreatic cancer was clustered together and significantly separated from controls (Figure 2C).

## 4.4. Composition and Difference of Gut Microbial Communities

Taxonomic analysis showed 59 phyla, 597 families, 3273 genera, and 22109 species in pancreatic cancer and control groups. At the phylum level, Firmicutes, Bacteroidetes, Proteobacteria, Actinobacteria, and Fusobacteria were the dominant bacteria phyla (Figure 3A-1). At the family level, *Bacteroidaceae, Enterobacteriaceae, Lachnospiraceae, Streptococcaceae, Ruminococcaceae,* and *Veillonellaceae* were the dominant bacterial families (Figure 3B-1). Phylum *Deinococcus-Thermus* was significantly decreased (P < 0.05) in pancreatic cancer patients compared to controls (Figure 3A-2). Families *Peptostreptococcaceae, Nocardioidaceae, Thermaceae,* and *Catenulisporaceae* were significantly decreased (P < 0.05) in pancreatic cancer (Figure 3B-2).

At the species level, *Prevotella copri*, Faecalibacterium, Escherichia coli, Klebsiella pneumoniae, and Bacteroides vulgatus were the dominant bacteria (Figure 3C-1). *Planctomyces sp. SH-PL14*, *Bacillus pumilus*, *Acidaminococcus*, *Burkholderia stabilis*, and *Paenibacillus sp. FSL H7-0357* were increased in the pancreatic cancer group. However, a total of 18 species, including *Paraoerskovia marina*, *Catenulispora acidiphila*, *Romboutsia ilealis*, and *Bifidobacterium*, were decreased in the pancreatic cancer group, but all of them did not reach statistical significance by the Benjamin-Hochberg method (P > 0.05) (Figure 3C-2). The correlation between floras was further analyzed. *Prevotella copri* was the core species in the seven dominant phylum association networks (Figure 3D).

## 4.5. Characteristic Bacteria and Differences in Microbial Function

The linear discriminant analysis (LDA) effect size (LEfSe) method was utilized to find the characteristic bacteria between pancreatic cancer and control groups. As shown in Figure 4A, the families *Pasteurellaceae*, *Pasteurellales*, and *liumatobacter* and the genera *liumatobacte* coccineus, Gordonibacter pamelaeae, Enterococcushirae, Fusobacterium periodonticum, Dickeya solani, Streptococcus phage ipp28, and Chlorobiun phaeobitrioies were significantly increased in the pancreatic cancer group. In contrast, Firmicutes, Clostridiale, Ruminococcaceae, and Bifidobacteria were remarkably decreased in the pancreatic cancer group based on LDA selection. The species Anaerostipes hadrus, Bifdobacterium pseudocatenulatum, Lachnoclostridium phocaeense, Romboutsia ilealis, and Bifidobacteriumbreve were decreased in the pancreatic cancer



group. However, Gordonibacter pamelaeae, Enterococcushirae, Chlorobium phaeovibrioides, Streptococcus phage IPP28, Dickeya solani, and Fusobacterium periodonticum were increased in the pancreatic cancer group. Further, we compared the differences in the relative abundance of butyrate-producing bacteria (Anaerostipes hadrus, Lachnoclostridium phocaeense, and Romboutsia ilealis) and pathogens (Fusobacterium and Enterobacter) (Figure 4B,C). The relative abundance of the butyrate-producing bacteria (Anaerostipes hadrus, Lachnoclostridium phocaeense, and Romboutsia ilealis) were decreased in the pancreatic cancer group. The relative abundance of the proinflammatory bacteria (Fusobacterium and Enterobacter) was increased in the pancreatic cancer group.

The microbial function alteration was identified using

Picrust 2 to predict the metabolic pathways. Based on LDA analysis, 44 pathways were significantly changed. Compared with controls, pathways connected with metabolism, environmental (bacterial secretion system), genetic information (protein export and ribosome), and human diseases (infectious diseases and drug resistance) were increased in the pancreatic cancer group (Figure 4D). Most pathways are involved in the metabolism of various substances.

The number of amino acid synthesis pathways was significantly increased in the pancreatic cancer group (taurine and hypotaurine metabolism, glutathione metabolism, degradation and biosynthesis of the leucine and isoleucine, and aspartate and glutamate metabolism). In addition, the number of carbohydrate

Characteristic	Statistical Description		Statistical Tests	P-Value
	Pancreatic Cancer (N = 8)	Control (N = 8)		
Age (mean $\pm$ SD)	$63.13 \pm 13.11$	$61.13 \pm 12.14$	<i>t</i> = 0.317	0.756
Gender			Fisher's exact method	1.000
Male	6 (75.0)	5 (62.5)		
Female	2 (25.0)	3 (37.5)		
Nationality			Fisher's exact method	1.000
Han	7 (87.5)	8 (100.0)		
Li	1(12.5)	0 (0.0)		
Smoking			Fisher's exact method	1.000
Yes	2 (25.0)	3 (37.5)		
No	6 (75.0)	5 (62.5)		
Drinking			Fisher's exact method	0.569
Yes	3 (37.5)	1(12.5)		
No	5 (62.5)	7 (87.5)		

<sup>a</sup> Values are expressed as No. (%) unless otherwise indicated.

metabolism pathways (amino sugar and nucleotide sugar metabolism, butanoate metabolism, C5 branched dibasic acid metabolism, fructose and mannose metabolism, galactose metabolism, glyoxylate and dicarboxylate metabolism, pentose phosphate pathway, and propanoate metabolism) were increased in the pancreatic cancer group compared to the controls. Meanwhile, the number of metabolisms of cofactor and vitamin pathways (pantothenate and CoA biosynthesis, ubiquinone, and other terpenoid quinone biosyntheses) was significantly increased. Moreover, the number of pathways, including glycan biosynthesis, terpenoid backbone biosynthesis, purine metabolism, and streptomycin biosynthesis, was increased in the pancreatic cancer group.

# 4.6. Relationship Between Blood Test Indices and Gut Microbiota

As shown in Figure 5A, 18 bacterial species were significantly associated with 12 blood indices. Clostridiales, Clostridum sp. SY8519, Bifidobacterium longum, and Roseburia Blautiahansenii were positively correlated with the lymphocyte count. Bacteroidales were negatively correlated with the number of lymphocytes. Faecallibacterium prausnitzii, butyrate-producing bacterium SS3/4, and Roseburia were positively correlated with the lymphocyte ratio. Anaerostipes hadrus, Eubacteriumeligens, Faecallibacterium prausnitzii, Bifidobacterium longum butyrate-producing bacterium SS3/4, Prevotella intermedia, and Roseburia were negatively correlated with the NLR. The differences in the relative abundance of three anti-inflammatory bacteria

negatively correlated with NLR were further analyzed (Figure 5B). Anaerostipes hadrus and [Eubacterium]eligens were decreased, while Prevotella intermedia was increased in the pancreatic cancer group.

## 5. Discussion

Over the past decade, the role of microorganisms in tumors has attracted much attention from researchers. To our knowledge, the results of the connection between pancreatic cancer and gut microbiota are inconsistent. Even the specific gut microbial profile in pancreatic cancer from the tropical area has never been reported. For the first time, this study illustrates the gut microbial profile in pancreatic cancer from the tropical area by metagenome sequencing.

The study showed that the microbial community in pancreatic cancer is clustered together and significantly separated from controls. Gut microbial diversity is significantly decreased, partly attributed to the increased Simpson index in pancreatic cancer versus controls. These results are consistent with Ren et al.'s (11) and Half et al.'s (12) studies. The gut microbial profile is unique in pancreatic cancer from tropical areas; the abundances of Enteroides, Cancoccus hirae, and Fusobacterium periodonticum were significantly increased, while Firmicutes, Ruminococcaceae, and Bifidobacteria were significantly decreased in the pancreatic cancer group.

We found that Prevotella copri is critical in dominant bacteria's interaction network. However, Ren et al. found



Figure 2. Gut microbial diversity and PLS-DA analysis between pancreatic cancer patients and controls

that *Firmicutes*, *Proteobacteria bacteria*, probiotics, and butyrate-producing bacteria were decreased in pancreatic cancer, whereas pathogenic bacteria and LPS-producing bacteria were increased (11). Half et al. showed that *Veillonellaceae* and *Akkermansia* were significantly increased, and *Ruminococcal* was significantly reduced in the pancreatic cancer group (12). Therefore, this study found the characteristic bacteria associated with pancreatic cancer, which is not entirely consistent with previous studies. This may be explained by regional influences, especially in tropical regions.

In our study, the gut microbiota in the pancreatic cancer group was associated with inflammatory markers (lymphocyte count, lymphocyte ratio, and NLR). The abundances of butyrate-producing bacteria and *Bifidobacteria* were significantly negatively correlated with the NLR. NLR represents the balance between neutrophils and lymphocytes (15). An increased NLR is closely related to the poor prognosis of pancreatic cancer (16, 17). Our previous study also found higher NLR as an independent risk factor of prognosis in pancreatic cancer (18). In this study, probiotics (butyrate-producing bacteria and Bifidobacterium) decreased pancreatic cancer at species, genus, and family levels. Butyrate can reduce the expression of proinflammatory cytokines such as interferon  $\gamma$  (Inf- $\gamma$ ), tumor necrosis factor- $\alpha$  (TNF- $\alpha$ ), or interleukins (ILs) such as IL-6 and IL-8 (19, 20).

Butyrate also regulates the expression of genes en-



coding proinflammatory cytokines, growth factors, Heat Shock Proteins (HSPs), and inflammation-inducing enzymes by inhibiting Nuclear Factor  $\kappa$ B (NF- $\kappa$ B) (21). Furthermore, the butyrate induces the synthesis of active transglutaminases and antimicrobial peptides. It also protects the intestinal mucosa from bacterial infection and inflammation (22-24). Yang found that *Bifidobacterium* is involved in the intestinal barrier and can activate phagocytes and NK cells (25). *Bifidobacterium* also increases the number of intestinal IgA-producing plasma cells and regulates immune function (26). *Bifidobacterial* exopolysaccharides (B-EPSs) can stimulate macrophages to produce TNF- $\alpha$  and inhibit tumor cells from growing (27). Therefore, reducing these probiotics may play a role in the progression of pancreatic cancer.

Moreover, this study showed that opportunistic pathogens increased at the species and genus levels in pancreatic cancer. It has been reported that *Fusobacteria* could combine with the epithelial cadherin (E-cadherin) of colorectal cancer (CRC) cells; then, it will activate the  $\beta$ -catenin signaling pathway and increase the expression of oncogenes and proinflammatory genes (28). This process can also activate NF- $\kappa$ B to increase the expression of proinflammatory genes, thus causing an inflammatory response (29). Quah pointed out that Fusobacterium causes an increase in inflammatory cytokines through the P38 MAPK signaling pathway (30). Kostic showed that Fusobacterium could recruit immune cells to form a proinflammatory microenvironment (31). Enterococcus had been reported to be associated with pancreatic inflammation. A significant increase in Enterococcus may lead to the impaired intestinal barrier function and translocation infection caused by pathogenic bacteria in acute pancreatic inflammation (32). Thus, the increase in the abundance of conditionally pathogenic bacteria may also be an essential factor in promoting the development of pancreatic cancer.

Our study showed that the taurine and hypotau-



Figure 4. Characteristic bacteria and differences in microbial function between pancreatic cancer and control groups



rine metabolism pathways were enriched in pancreatic cancer. Metabolome studies have shown that taurine metabolism significantly differs between pancreatic cancer and healthy groups (33). It is enriched at high levels in pancreatic cancer (34). The ADO/taurine axis can activate the NF- $\kappa$ B pathway and promote tumors (35). In addition, other studies have shown that taurine can inhibit the expression of Wnt5a by enhancing methylation of the promoter and promoting the invasion and proliferation of tumors (36). At the same time, the research showed that the expression level of cathelicidin antimicrobial peptide (CAMP) was high in the advanced pancreatic cancer group (37), consistent with our study. Furthermore, the tumorassociated macrophages secreted CAMP in response to tumor growth factor- $\beta$ , enhancing the growth and proliferation of cancer cells (37). In addition, our study found that pathways of drug resistance, especially the antineoplastic pathway, were increased in the pancreatic cancer group. Other researchers also proved that resistance to gemcitabine could result from the expression of a long isoform of bacterial enzyme cytidine deaminase (38, 39).

Pancreatic cancer is a rare disease. This study defined strict inclusion and exclusion criteria. To the study deadline, eight patients with pancreatic cancer were included. The control group was screened in the same hospital at the same time. After some conditional restrictions and matching, only eight patients were enrolled. The small sample size is an area of concern in this study, which may have affected the accuracy of the results. This study used metagenome sequencing; as far as we know, there is no strict regulation on how large the sample size must be before the results can be applied. As a cross-sectional study, although the sample size is few, we used the restriction and matching design to control the apparent confounding bias, such as no drugs and/or probiotics consumption within eight weeks before stool collection. Furthermore, the subjects were Hainanese individuals registered in Hainan, with dietary habits of mainly seafood, vegetables, and fruits, and low salt and oil intake, which improved the representation of participants and controlled the selection and confounding factors. Similar to most studies that just showed a correlation between gut microbiota and a specific disease, our study could not verify the causal relationship between intestinal bacteria and pancreatic cancer. Therefore, it is necessary to verify the possible causal relationship between intestinal flora and pancreatic cancer using the disease animal model.

#### 5.1. Conclusions

To our knowledge, this is the first report to illustrate gut microbial characteristics in pancreatic cancer from the tropical area of China through metagenomic sequencing. The gut microbial profile is unique in pancreatic cancer. Some butyrate-producing bacteria and Bifidobacteria were decreased, and some conditional pathogenic bacteria (Fusobacterium and Enterobacter) were increased. The bacterial species are not completely consistent with previous studies. It is encouraging to find if Prevotella copri plays a crucial role in the bacterial interaction network. The abundances of butyrate-producing bacteria (butyrateproducing bacterium SS3/4, Anaerostipes hadrus, Roseburia intestinalis, and Faecalibacterium prausnitzii) and Bifidobacteria were significantly negatively correlated with the NLR. Pathways connected to metabolism, environment (bacterial secretion system), genetic information (protein export and ribosome), and human diseases (infectious diseases and drug resistance) increased in the pancreatic cancer group. The above results suggest that the diversity of intestinal flora may be related to the environment. Gut microbial alterations can influence metabolic function and host immunity and affect the occurrence and development of pancreatic cancer.

## Footnotes

**Authors' Contribution:** Study concept and design: H.H., G.D., Z.S., and X.W.; Acquisition of data: X.W., Z.H., and X.Y.; Analysis and interpretation of data: X.W., H.H., G.D., Z.S., and W.L.; Drafting of the manuscript: X.W. and E.C.M.; Critical manuscript revision for important intellectual content: H.H., G.D., Z.S., and W.L.; Statistical analysis: X.W. and H.H.; Administrative, technical, and material support: G.D, Z.S., Z.S., T.W., and E.C.M.; Study supervision: H.H..

**Conflict of Interests:** The authors declare no conflicts of interest.

**Ethical Approval:** The Ethics Committee of Hainan Medical University approved this study based on the protocol described in the Methods section (No. HYLL2020016).

**Funding/Support:** This work was supported by the Key Research Projects of Higher Education Institutions of Hainan Province (No. Hnky2019ZD-23) and supported by the Hainan Provincial Natural Science Foundation of China (821RC745).), innovative research projects for graduate students in Hainan Province (Qhys2021-375), and innovative research projects for undergraduate students in Hainan Province (X201911810051).

**Informed Consent:** We have obtained written informed consent signed by the study participants and uploaded a copy of the informed consent form on the journal website.

## References

- Siegel RL, Miller KD, Jemal A. Cancer statistics, 2020. CA Cancer J Clin. 2020;70(1):7-30. doi: 10.3322/caac.21590. [PubMed: 31912902].
- Bray F, Ferlay J, Soerjomataram I, Siegel RL, Torre LA, Jemal A. Global cancer statistics 2018: GLOBOCAN estimates of incidence and mortality worldwide for 36 cancers in 185 countries. *CA Cancer J Clin.* 2018;**68**(6):394–424. doi: 10.3322/caac.21492. [PubMed: 30207593].
- Zeng H, Chen W, Zheng R, Zhang S, Ji JS, Zou X, et al. Changing cancer survival in China during 2003-15: a pooled analysis of 17 populationbased cancer registries. *Lancet Glob Health*. 2018;6(5):e555–67. doi: 10.1016/S2214-109X(18)30127-X. [PubMed: 29653628].
- Cai J, Chen HD, Lu M, Zhang YH, Lu B, You L, et al. [Trend analysis on morbidity and mortality of pancreatic cancer in China, 2005-2015]. *Zhonghua Liu Xing Bing Xue Za Zhi*. 2021;42(5):794–800. Chinese. doi: 10.3760/cma.j.cn112338-20201115-01328. [PubMed: 34814469].
- Ni Y, Lohinai Z, Heshiki Y, Dome B, Moldvay J, Dulka E, et al. Distinct composition and metabolic functions of human gut microbiota are associated with cachexia in lung cancer patients. *ISME J.* 2021;15(11):3207-20. doi: 10.1038/s41396-021-00998-8. [PubMed: 34002024]. [PubMed Central: PMC8528809].
- Rajagopala SV, Vashee S, Oldfield LM, Suzuki Y, Venter JC, Telenti A, et al. The human microbiome and cancer. *Cancer Prev Res (Phila)*. 2017;**10**(4):226–34. doi: 10.1158/1940-6207.CAPR-16-0249. [PubMed: 28096237].
- Oh NS, Lee JY, Kim YT, Kim SH, Lee JH. Cancer-protective effect of a synbiotic combination between Lactobacillus gasseri 505 and a Cudrania tricuspidata leaf extract on colitis-associated colorectal cancer.

*Gut Microbes*. 2020;**12**(1):1785803. doi: 10.1080/19490976.2020.1785803. [PubMed: <u>32663105</u>]. [PubMed Central: PMC7524312].

- Rattan P, Minacapelli CD, Rustgi V. The microbiome and hepatocellular carcinoma. *Liver Transpl*. 2020;26(10):1316–27. doi: 10.1002/lt.25828. [PubMed: 32564483].
- Song M, Chan AT, Sun J. Influence of the gut microbiome, diet, and environment on risk of colorectal cancer. *Gastroenterology*. 2020;**158**(2):322–40. doi: 10.1053/j.gastro.2019.06.048. [PubMed: 31586566]. [PubMed Central: PMC6957737].
- Tripathi A, Debelius J, Brenner DA, Karin M, Loomba R, Schnabl B, et al. The gut-liver axis and the intersection with the microbiome. *Nat Rev Gastroenterol Hepatol.* 2018;15(7):397–411. doi: 10.1038/s41575-018-0011z. [PubMed: 29748586]. [PubMed Central: PMC6319369].
- Ren Z, Jiang J, Xie H, Li A, Lu H, Xu S, et al. Gut microbial profile analysis by MiSeq sequencing of pancreatic carcinoma patients in China. *Oncotarget*. 2017;8(56):95176–91. doi: 10.18632/oncotarget.18820. [PubMed: 29221120]. [PubMed Central: PMC5707014].
- Half E, Keren N, Reshef L, Dorfman T, Lachter I, Kluger Y, et al. Fecal microbiome signatures of pancreatic cancer patients. *Sci Rep.* 2019;9(1):16801. doi: 10.1038/s41598-019-53041-4. [PubMed: 31727922]. [PubMed Central: PMC6856127].
- Rothschild D, Weissbrod O, Barkan E, Kurilshikov A, Korem T, Zeevi D, et al. Environment dominates over host genetics in shaping human gut microbiota. *Nature*. 2018;555(7695):210–5. doi: 10.1038/nature25973. [PubMed: 29489753].
- Pancreatic Cancer Committee of Chinese Anticancer Association. [Comprehensive guidelines for the diagnosis and treatment of pancreatic cancer (2020 version)]. *Zhonghua Wai Ke Za Zhi*. 2021;**59**(2):81-100. Chinese. doi: 10.3760/cma.j.cn112139-20201113-00794. [PubMed: 33378799].
- Templeton AJ, McNamara MG, Seruga B, Vera-Badillo FE, Aneja P, Ocana A, et al. Prognostic role of neutrophil-to-lymphocyte ratio in solid tumors: a systematic review and meta-analysis. J Natl Cancer Inst. 2014;106(6):dju124. doi: 10.1093/jnci/dju124. [PubMed: 24875653].
- Bausch D, Pausch T, Krauss T, Hopt UT, Fernandez-del-Castillo C, Warshaw AL, et al. Neutrophil granulocyte derived MMP-9 is a VEGF independent functional component of the angiogenic switch in pancreatic ductal adenocarcinoma. *Angiogenesis*. 2011;14(3):235–43. doi: 10.1007/s10456-011-9207-3. [PubMed: 21442180]. [PubMed Central: PMC3688040].
- Szkandera J, Stotz M, Eisner F, Absenger G, Stojakovic T, Samonigg H, et al. External validation of the derived neutrophil to lymphocyte ratio as a prognostic marker on a large cohort of pancreatic cancer patients. *PLoS One*. 2013;8(11). e78225. doi: 10.1371/journal.pone.0078225. [PubMed: 24223776]. [PubMed Central: PMC3817201].
- Zou Y, He ZJ, Ma Y, Shen YM, Sun ZK, Wang TC. Analysis of clinical characteristics and risk of metastasis of 352 patients with pancreatic cancer inthe tropical region of Hainan Province. *Chin J Dis Control Prev.* 2020;**24**(5):610–3. doi: 10.16462/j.cnki.zhjbkz.2020.05.022.
- Fung KY, Cosgrove L, Lockett T, Head R, Topping DL. A review of the potential mechanisms for the lowering of colorectal oncogenesis by butyrate. *Br J Nutr.* 2012;**108**(5):820–31. doi: 10.1017/S0007114512001948. [PubMed: 22676885].
- 20. Vinolo MA, Rodrigues HG, Hatanaka E, Sato FT, Sampaio SC, Curi R. Suppressive effect of short-chain fatty acids on production of proinflammatory mediators by neutrophils. *J Nutr Biochem*. 2011;**22**(9):849– 55. doi: 10.1016/j.jnutbio.2010.07.009. [PubMed: 21167700].
- Donohoe DR, Collins LB, Wali A, Bigler R, Sun W, Bultman SJ. The Warburg effect dictates the mechanism of butyrate-mediated histone acetylation and cell proliferation. *Mol Cell*. 2012;48(4):612–26. doi: 10.1016/j.molcel.2012.08.033. [PubMed: 23063526]. [PubMed Central: PMC3513569].
- 22. Arvans DL, Vavricka SR, Ren H, Musch MW, Kang L, Rocha FG, et al. Luminal bacterial flora determines physiological expression of intestinal epithelial cytoprotective heat shock proteins 25 and 72. *Am J Physiol Gastrointest Liver Physiol.* 2005;288(4):G696-704. doi:

10.1152/ajpgi.00206.2004. [PubMed: 15528251].

- D'Argenio G, Calvani M, Della Valle N, Cosenza V, Di Matteo G, Giorgio P, et al. Differential expression of multiple transglutaminases in human colon: impaired keratinocyte transglutaminase expression in ulcerative colitis. *Gut.* 2005;**54**(4):496–502. doi: 10.1136/gut.2004.049411. [PubMed: 15753534]. [PubMed Central: PMC1774462].
- Kiehne K, Brunke G, Wegner F, Banasiewicz T, Folsch UR, Herzig KH. Defensin expression in chronic pouchitis in patients with ulcerative colitis or familial adenomatous polyposis coli. *World J Gastroenterol.* 2006;12(7):1056–62. doi: 10.3748/wjg.v12.i7.1056. [PubMed: 16534846]. [PubMed Central: PMC4087897].
- Yang YX, He M, Hu G, Wei J, Pages P, Yang XH, et al. Effect of a fermented milk containing Bifidobacterium lactis DN-173010 on Chinese constipated women. *World J Gastroenterol*. 2008;14(40):6237-43. doi: 10.3748/wjg.14.6237. [PubMed: 18985817]. [PubMed Central: PMC2761588].
- Ruiz L, Delgado S, Ruas-Madiedo P, Sanchez B, Margolles A. Bifidobacteria and their molecular communication with the immune system. *Front Microbiol.* 2017;8:2345. doi: 10.3389/fmicb.2017.02345. [PubMed: 29255450]. [PubMed Central: PMC5722804].
- Kim Y, Lee D, Kim D, Cho J, Yang J, Chung M, et al. Inhibition of proliferation in colon cancer cell lines and harmful enzyme activity of colon bacteria by Bifidobacterium adolescentis SPM0212. *Arch Pharm Res.* 2008;**31**(4):468–73. doi: 10.1007/s12272-001-1180-y. [PubMed: 18449504].
- Rubinstein MR, Wang X, Liu W, Hao Y, Cai G, Han YW. Fusobacterium nucleatum promotes colorectal carcinogenesis by modulating E-cadherin/beta-catenin signaling via its FadA adhesin. *Cell Host Microbe*. 2013;14(2):195-206. doi: 10.1016/j.chom.2013.07.012. [PubMed: 23954158]. [PubMed Central: PMC3770529].
- Lee P, Tan KS. Fusobacterium nucleatum activates the immune response through retinoic acid-inducible gene I. J Dent Res. 2014;93(2):162-8. doi: 10.1177/0022034513516346. [PubMed: 24334410].
- Quah SY, Bergenholtz G, Tan KS. Fusobacterium nucleatum induces cytokine production through Toll-like-receptor-independent mechanism. *Int Endod J.* 2014;47(6):550–9. doi: 10.1111/iej.12185. [PubMed: 24102075].
- Kostic AD, Chun E, Robertson I, Glickman JN, Gallini CA, Michaud M, et al. Fusobacterium nucleatum potentiates intestinal tumorigenesis and modulates the tumor-immune microenvironment. *Cell Host Microbe*. 2013;14(2):207–15. doi: 10.1016/j.chom.2013.07.007. [PubMed: 23954159]. [PubMed Central: PMC3772512].
- Tan C, Ling Z, Huang Y, Cao Y, Liu Q, Cai T, et al. Dysbiosis of intestinal microbiota associated with inflammation involved in the progression of acute pancreatitis. *Pancreas*. 2015;44(6):868–75. doi: 10.1097/MPA.00000000000355. [PubMed: 25931253].
- 33. Wang AS, Lodi A, Rivera LB, Izquierdo-Garcia JL, Firpo MA, Mulvihill SJ, et al. HR-MAS MRS of the pancreas reveals reduced lipid and elevated lactate and taurine associated with early pancreatic cancer. NMR Biomed. 2014;27(11):1361–70. doi: 10.1002/nbm.3198. [PubMed: 25199993]. [PubMed Central: PMC5554431].
- Battini S, Faitot F, Imperiale A, Cicek AE, Heimburger C, Averous G, et al. Metabolomics approaches in pancreatic adenocarcinoma: tumor metabolism profiling predicts clinical outcome of patients. *BMC Med.* 2017;15(1):56. doi: 10.1186/s12916-017-0810-z. [PubMed: 28298227]. [PubMed Central: PMC5353864].
- Shen D, Tian L, Yang F, Li J, Li X, Yao Y, et al. ADO/hypotaurine: a novel metabolic pathway contributing to glioblastoma development. *Cell Death Discov*. 2021;7(1):21. doi: 10.1038/s41420-020-00398-5. [PubMed: 33483477]. [PubMed Central: PMC7822925].
- Tian H, Cong P, Qi R, Gao X, Liu X, Liu H, et al. Decreased invasion ability of hypotaurine synthesis deficient glioma cells was partially due to hypomethylation of Wnt5a promoter. *Biocell*. 2017;41(1):27. doi: 10.32604/biocell.2017.41.027.

Jundishapur J Microbiol. 2022; 15(4):e122386.

- Sainz BJ, Alcala S, Garcia E, Sanchez-Ripoll Y, Azevedo MM, Cioffi M, et al. Microenvironmental hCAP-18/LL-37 promotes pancreatic ductal adenocarcinoma by activating its cancer stem cell compartment. *Gut.* 2015;64(12):1921–35. doi: 10.1136/gutjnl-2014-308935. [PubMed: 25841238].
- Geller LT, Barzily-Rokni M, Danino T, Jonas OH, Shental N, Nejman D, et al. Potential role of intratumor bacteria in mediating tumor resistance to the chemotherapeutic drug gemcitabine. Sci-

ence. 2017;**357**(6356):1156–60. doi: 10.1126/science.aah5043. [PubMed: 28912244]. [PubMed Central: PMC5727343].

 Choy ATF, Carnevale I, Coppola S, Meijer LL, Kazemier G, Zaura E, et al. The microbiome of pancreatic cancer: from molecular diagnostics to new therapeutic approaches to overcome chemoresistance caused by metabolic inactivation of gemcitabine. *Expert Rev Mol Diagn*. 2018;18(12):1005–9. doi: 10.1080/14737159.2018.1544495. [PubMed: 30392417].