

The Key Role of Pseudomonas-Associated Bacteria in Obtaining Nutrition from Bamboo for the Giant Panda (Ailuropoda melanoleuca)

Ruihong Ning

Chengdu Medical College

Caiwu Li

China Conservation and Research Center for the Giant Panda

Siyuan Zhang

Chengdu Medical College

Yu Zhang

Chengdu Medical College

Yunong Gan

Chengdu Medical College

Desheng Li

China Conservation and Research Center for the Giant Panda

Zhan Gao

China Conservation and Research Center for the Giant Panda

Wei Li

China Conservation and Research Center for the Giant Panda

Zhufeng Qu

China Conservation and Research Center for the Giant Panda

Fenghui Sun

Chengdu Medical College

Wei Guo (Seguochina2005@126.com)

Chengdu Medical College

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Abstract

Background

The gut microbiota of giant pandas plays a vital role in obtaining nutrition from bamboo. However, despite the importance of cellulose and lignin in bamboo digestion, low cellulase activity has been observed in the panda's gut, and no specific pathway has been implicated in lignin digestion by gut microbiota of pandas. As a result, the mechanism by which pandas obtain nutrients is still controversial, and it is necessary to determine whether the gut microbiome of giant pandas has the potential to degrade lignin. Here, the metabolic pathways for lignin degradation in pandas were explored by comparing 209 metagenomic sequencing data from wild species with different feeding habits. The degradation potential of lignin by the most dominant bacteria genus (*Pseudomonas*) were also explore, including identifying lignin degradation pathway in its genome, detecting lignin enzyme activity, and assessing the efficiency and products of lignin metabolism.

Results

The central pathways of lignin degradation, including beta-ketoadipate and homogentisate pathway, were found to be enriched in the gut microbiota of wild pandas. Additionally, the gut microbiome of wild bamboo-eating specialists was also enriched with genes from pathways implicated in degrading ferulate and p-coumarate (lignin derivative) into Acetyl-CoA and Succinyl-CoA via protocatechuate branch of beta-ketoadipate pathway, potentially providing the raw materials for metabolism in pandas. Specifically, *Pseudomonas*, as the most dominant gut bacteria genus, was found to be the main bacteria to provide genes involved in the central pathways of lignin degradation and metabolic pathway of lignin derivatives (ferulate and p-coumarate). Herein, three *Pseudomonas*-associated species isolated from the feces of wild pandas exhibited the laccase, lignin peroxidase, manganese peroxidase activity, as well as extracellular lignin degradation ability in vitro. Based on these results, we propose a potential mechanism for pandas to obtain nutrition from bamboo.

Conclusions

Notably, our findings indicate that the gut microbiome of pandas, particularly the *Pseudomonas* bacteria, play a crucial role in obtaining nutrition from lignin. This study provides novel insights into the adaptive evolution of pandas with respect to lignin metabolism.

Introduction

Phylogenetic studies have shown that the giant panda (*Ailuropoda melanoleuca*) is an herbivorous mammal belonging to the bear family [1, 2]. Giant pandas have undergone a series of evolution to adapt to the special bamboo diet, including ecological [3-5], morphological [6-8], and genetic [9-11] adaptations [4, 12]. Notably, giant pandas lack genes for lignin, cellulose, and hemicellulose degrading pathways in their genome [9]. As a result, conservation biologists have been curious to know how giant

pandas obtain nutrition from bamboo. A complete metabolic pathway for cellulose and hemicellulose degradation has been detected in the intestine of giant pandas [13]. Therefore, the gut microbiota is considered the main route through which giant pandas obtain nutrition from bamboo diet [9]. However, the gut microbiota of giant pandas has not shown an obvious convergent evolution phenomenon according to their feeding habits, and they still retain a gut microbiota structure like that of carnivores [14–16]. Although the abundance of hemicellulose-degrading bacteria in the gut of giant pandas increases after eating bamboo [17], the abundance of genes related to cellulose and hemicellulose degradation in their gut microbiome is significantly lower than that in other herbivores [13, 14, 17]. Many previous studies have confirmed that the gut microbiota of giant pandas is distinct from that of red pandas and clusters closer to those of the black bears and carnivores [14–16]. However, Huang et al. found a close similarity in gut microbiota structure among bamboo-eating pandas (giant and red pandas), confirming that food factors drove the convergent evolution of gut microbiota of pandas [18]. Therefore, it remains controversial whether the gut microbiota of giant pandas can adapt to their highly specialized diet [14-16]. Indeed, studies on the adaptation of gut microbiota of giant pandas to the bamboo diet have mainly focused on captive populations. Considering the obvious difference in gut microbiota composition between captive and wild giant pandas [13, 19, 20], more studies should be performed to uncover mechanisms behind the adaptation of gut microbiota of wild giant pandas to the bamboo diet.

Bamboo is mainly composed of lignin, cellulose, and hemicellulose [21]. Lignin attaches to cellulose and hemicellulose in the cell wall and protects them from degradation. Therefore, the gut microbiota of giant pandas needs to first oxidize and decompose lignin before digesting the other components of the bamboo diet. To date, the precise mechanism of lignin degradation by giant pandas is unclear. Perenniporia medulla-panis is a fungus found in the gut of giant pandas and exhibits lignin peroxidase activity [22]. Lac51 is a gene encoding laccase, an enzyme that can degrade a variety of lignin monomer phenols via oxidation. The sequence similarity alignment of *Lac51* cloned from the fecal microbiotas of giant pandas is closely related to multi copper oxidase gene of *Pseudomonas* sp. [23]. However, only a few studies have assessed the potential role of gut microbiota of giant pandas in lignin degradation [17]. The community structure of microbiota in wild giant pandas is significantly different from that of captive giant pandas [19]. Specifically, Streptococcus and Enterobateriaceae are the most abundant genera in captive pandas [16, 19, 24], while Pseudomonas is the most dominant genus in the gut of wild pandas [19, 20, 25]. Similarly, Pseudomonas is the most dominant bacteria in the gut of the wild population of red pandas that feed on bamboo [26]. Most Streptococcus and Enterobateriaceae are pathogens or opportunistic pathogens [27-29]. Pseudomonas is one of the few bacteria that can efficiently degrade cellulose and lignin [30, 31]. For instance, Pseudomonas sp. strain ys-1p, Pseudomonas fluorescens, Pseudomonas PKE117, and Pseudomonas putida can degrade several lignin monomers and dimers [32-35]. Davinia et al. revealed that Pseudomonas putida KT2440 secretes vesicles carrying enzymes that can degrade lignin and its derivatives [36]. Whether the Pseudomonas-associated bacteria in wild giant pandas can potentially degrade lignin is still unclear.

A recent study on the adaptation of gut microbiome of giant pandas to different diets and habitats before and after reintroduction identified four genes encoding enzymes, including catalase peroxidase, dioxygenase, quinone reductase, and triacylglycerol lipase, in the metagenome of fecal microbiotas of the giant pandas [37]. Pseudomonas, Enterococcus, and Lactococcus are the bacteria enriching these genes. The abundance of the three bacterial genera significantly increased after the release of giant pandas into the wild [37]. Yao et al. [20] and Tang et al. [25] also reported that *Pseudomonas* abundance in the gut of captive giant pandas significantly increased after their release into the wild, suggesting that the diversity of gut microbiota of giant pandas increased after reintroduction in the wild. Notably, Pseudomonas is the main cyanide-degrading bacteria in the gut of bamboo-eating pandas (wild giant and red pandas), revealing their adaptation to the bamboo diet [38]. Collectively, these studies suggest that Pseudomonas is the main bacteria responsible for the adaptation of giant pandas to the bamboo diet. However, it is still not clear whether the Pseudomonas in the gut of giant pandas can degrade lignin. Also, the specific metabolic pathway of degrading lignin has not been elucidated. In this study, the potential role of the gut microbiome of giant pandas in lignin degradation was explored via metagenomic sequencing to reveal the metabolic function of gut microbiota involved in lignin degradation in bamboo-eating pandas (giant and red pandas), herbivorous, omnivorous, and carnivorous. High-quality individual draft genomes (bins) were produced based on the shotgun metagenomic assemblies. The genomes were then used to reveal the metabolic pathways related to lignin degradation. In addition, Pseudomonas-associated strain were isolated from the feces of giant pandas. Lignin degradation and extracellular secretion ability of ligninolytic enzymes was also verified in vitro.

This study reveals the specific metabolic pathway of lignin degradation in giant pandas and provides new evidence for adaptive evolution of giant pandas to the bamboo diet. Therefore, these findings provide a basis for future studies on giant pandas, which is important for their protection.

Results

Metagenome sequencing data

A total of 319,158,629 paired-end metagenome reads were generated in this study. In addition, 5,347,754,253 published raw paired-end reads from 209 wild individual samples (representing bambooeating pandas, herbivores, omnivores, carnivores) were retrieved from the public database for comprehensive analysis. A total of 5,117,016,424 clean paired-end reads were retained for subsequent analysis after filtering out the low-quality reads (Q <20) and host sequences (Table S4). The final contigs of each species were obtained through De novo assembly (see Table S5 for detailed information). Non-redundant gut microbiome gene set of each species was created from contigs. Principle Coordinate Analysis (PCoA) of predicted metagenomic function (KO genes) based on Bray-curtis distance showed that wild bamboo-eating pandas formed distinct clusters from other animals (Adonis R2= 0.473; P-value= 0.001), implying that they have a distinct function (Fig. S1). Moreover, wild bamboo-eating pandas harbored a distinct gut microbiome from captive bamboo-eating pandas. Furthermore, gut microbiome in carnivores was distinct from that of herbivores and omnivores. Overall, the functions of the gut microbiome of different animals clustered according to their food habits, indicating their various roles.

Metabolic pathways related to lignin degradation in gut microbiotas of giant pandas

Pathway analysis based on the genome of microbiotas derived from feces of wild bamboo-eating pandas (giant and red pandas) showed enrichment of the complete metabolic pathway of central aromatic intermediates, including Catechol branch of beta-ketoadipate pathway (Cbβ-KAP) (Fig. 1A, Fig.S2), Protocatechuate branch of beta-ketoadipate pathway (Pbβ-KAP) (Fig. 1A, Fig.S3) and homogentisate pathway of aromatic compound degradation (Fig. 1B, Fig.S4). Notably, 4hydroxyphenylpyruvate dioxygenase (HPD), ring-1, 2-phenylacetyl-CoA epoxidase subunit (paaE) homogentisate 1, 2-dioxygenase (HGD), maleylacetoacetate isomerase (maiA), fumarylacetoacetase (Fah), fumarylacetoacetate hydrolase (faaH) catalyzed reactions in homogentisate pathway (Fig. 1B). Catechol-1, 2-dioxygenase (catA), muconate lactonizing enzyme (catB) and muconolactone isomerase (catC), involved in catalysis of degradation reactions in Cbβ-KAP, were significantly enriched (Fig. 1A). Pbβ-KAP was metabolized through Protocatechuate 3,4-dioxygenase alpha chain (pcaG), Protocatechuate 3,4-dioxygenase beta chain (pcaH), 3-carboxy-cis, cis-muconate cycloisomerase (pcaB), 4-carboxymuconolactone decarboxylase (pcaC, pcaL), and 3-oxoadipyl-CoA thiolase (pcaF). Protocatechuate and Catechol are degraded to 3-Oxoadipate enol-lactone and share the same metabolic pathway. The degradation process is catalyzed by Beta-ketoadipate enol-lactone hydrolase (pcaD, pcaL), 3-oxoadipate CoA-transferase subunit A (pcal), and 3-oxoadipate CoA-transferase subunit B (pcaJ) (Fig. 1A). The final product of Cb_b-KAP and Pb_b-KAP degradation is Succinyl-CoA and Acetyl-CoA, whereas Phenylpyruvate is eventually degraded to fumarate through the homogentisate pathway. Succinyl-CoA, Acetyl-CoA and Fumarate can join the tricarboxylic acid (TCA) cycle for biosynthesis of valuable products.

Lignin monomers, including ferulate and *p*-Coumarate, are finally degraded to protocatechuate through a series of metabolic reactions. Protocatechuate is further degraded to Succinyl-CoA through Pbβ-KAP (Fig. 1C). Degradation of ferulate is catalyzed by feruloyl-CoA synthase (Fcs), feruloyl-CoA hydratase (FerB), vanillin dehydrogenase (Vdh), vanillate monooxygenase (VanA), and vanillate monooxygenase ferredoxin subunit (VanB). Eruloyl-CoA synthase (Fcs), feruloyl-CoA hydratase (FerB), vanillin dehydrogenase (Vdh), and p-hydroxybenzoate 3-monooxygenase (pobA) catalyze degradation reactions of *p*-Coumarate. These genes were identified in the genome of microbiotas derived from the feces of wild giant pandas.

Taxonomic assignment of enzymes revealed that a large proportion (26.08% on average) of the predicted genes were present in species within the *Pseudomonas* genus bacteria. For example, 8 out of 12 vdh, 16 out of 54 vanA, and 38 out of 163 pcaD genes in this study were homologous to genes from *Pseudomonas*-associated bacteria (Table S6). This finding indicated that predicted genes coding for lignin derivatives or monomer-digesting enzymes were mainly contributed by *Pseudomonas* genus.

Heatmap analysis showed that the average abundance of genes involved in the metabolism of central aromatic intermediates, *p*-Coumarate, and ferulate, was highest in wild bamboo-eating pandas compared

with the abundance in other animals (carnivores, omnivores, herbivores, and captive bamboo-eating pandas) (Fig. 1D). The relative abundance of these genes was significantly higher in wild bamboo-eating pandas than in other animals (Fig. 2).

Individual draft genomes (bins) of gut microbiotas from wild giant pandas

Fifty-two high-quality individual draft genomes (bins) were obtained from gut metagenomic data of wild giant pandas (Fig. 3A, Table S7). The average abundances of Bin 1, Bin 17, Bin 33, and Bin 52 were the highest in the gut of wild giant pandas (Fig. 3A). Bin 1, Bin 17, and Bin 33 were classified as *Pseudomonas*, whereas Bin 52 was identified as *Pseudomonas fluorescens* according to the principle of the closest strain annotation. Analysis at the genus level showed that *Pseudomonas* (Fig. S5A) and captive giant pandas (Fig. S5B), respectively. Further comparison showed that the relative abundance of *Pseudomonas*-associated bacteria in the gut was significantly higher in wild giant panda (mean ± SD: 0.35.80±0.22.70) and wild red panda (0.58.34±0.26.52) than in captive red pandas (0.0014±0.0013), captive giant pandas (0.0016±0.0024), herbivores (0.0013±0.0003), omnivores (0.0004±0.0001) and carnivores (0.0005±0.0004) (Fig. 3B).

The draft genome of *Pseudomonas*-associated (Bin 1, Bin 17, Bin 33, and Bin 52) (Table S8) and *Achromobacter*-associated (Bin 9 and Bin 44) (Table S9) strains encoded all genes implicated in Cbβ-KAP and Pbβ-KAP metabolism and homogentisate pathway (Fig. 3C; Table 1). The genome of *Oxalobacteraceae* (Bin 14 and Bin 51) comprised all genes implicated in Cbβ-KAP and Pbβ-KAP metabolism, whereas the genome of *Janthinobacterium* (Bin 26) encoded all genes involved in Cbβ-KAP metabolism and homogentisate pathway (Table 1). In addition, the genome of *Comamonadaceae* (Bin 29) and *Alcaligenaceae* (Bin 45) encoded all genes implicated in Cbβ-KAP metabolism, whereas the genome of *Flavobacterium* (Bin 27), *Stenotrophomonas sp.* LM091 (Bin 11), *Flavobacterium sp.* 140616W15 (Bin 15) encoded all genes involved in homogentisate pathway (Table 1). Pob A, vanA, vanB, and vdh genes were identified in the genome of *Pseudomonas*-associated Bin (Table S8).

Phylogenetic analysis of Pseudomonas genomes in wild giant pandas

The four closest neighbors of Bin 1, Bin 17, Bin 33, and Bin 52 were *Pseudomonas fluorescens* PfO-1, *Pseudomonas syringae pv. phaseolicola* 1448A, *Pseudomonas putida* KT2440, and *Pseudomonas aeruginosa* PAO1, respectively, as identified through RAST. A total of 40 complete genomes of different strains (10 per species, Table S3), belonging to *Pseudomonas fluorescens, Pseudomonas syringae, Pseudomonas putida*, and *Pseudomonas aeruginosa*, were randomly selected to construct phylogenetic trees for further classification of *Pseudomonas strains* in the wild giant pandas. Phylogenetic analysis showed that Bin 52 (*Pseudomonas fluorescens*) clustered in the *Pseudomonas fluorescens* clade, and Bin 1, Bin 17, and Bin 33 (*Pseudomonas*) clustered into a single clade in the phylogenetic tree (Fig. 4). Similar to *Pseudomonas* genomes in this study (Fig.3C), the genome of *Pseudomonas fluorescens* PfO-1 (Fig. S6A), *Pseudomonas syringae* pv. *phaseolicola* 1448A (Fig. S6B), P*seudomonas putida*

KT2440 (Fig. S6C), and *Pseudomonas aeruginosa* PA01 (Fig. S6D) encoded all enzymes implicated in the metabolism of central aromatic intermediates.

Laccase-like multicopper oxidase genes were present in the gut of wild giant pandas

Approximately 150 bp fragments between regions I and II of laccase-like multicopper oxidase genes, which are crucial in lignin degradation, were obtained from the DNA derived from feces of wild giant pandas through PCR amplification (Fig. S7). A total of 228 colonies from three samples (76 per sample) of wild giant pandas were chosen at random for sequencing. Nucleotide sequences obtained were used as query sequences for BLAST search to obtain the sequences deposited in GenBank nucleotide database. The BLAST results indicated that most (except for colony 2) PCR products highly corresponded to the laccase-like multicopper oxidase genes of bacteria. The results showed that 76% of products of colonies had a high similarity (> 90%) to gene sequences retrieved from GenBank nucleotide database. Notably, only two clones had less than 80% similarity compared with nucleotide sequences deposited in the GenBank (Table S10). The most consistent alignment results obtained from GenBank sequences were considered as the possible species of laccase fragments. A total of 228 colony products were identified as Verrucomicrobiaceae HC12, Verrucomicrobiaceae ONA9, Pseudomonas Antarctica, Janthinobacterium svalbardensis, Pseudomonas gingeri, Agrobacterium, Flavobacterium sp, Acinetobacter sp, Pseudomonas azotoformans, Pseudomonas fluorescens, Pseudomonas putida, Klebsiella pneumonia, Stenotrophomonas sp, Caulobacter sp, Brevundimonas diminuta, Brevundimonas vancanneytii, Brevundimonas naejangsanensis, and Sphingosinicella sp (Table S10).

Eighteen published laccase-like multicopper oxidase gene sequences were retrieved for phylogenetic analysis (Table S11). Phylogenetic results indicated that the laccase-like multicopper oxidase genes in the feces of wild giant pandas clustered into 19 clades. Notably, each representative clone fragment showed a close evolutionary relationship with the reference sequences obtained from GenBank (Fig. 5A). *Pseudomonas*-associated sequences were abundant in feces libraries of the wild giant pandas, where they accounted for 83.33% of the total number of sequences (Fig. 5B). The proportion of *Pseudomonas fluorescens*-associated sequences was the highest (about 46%), followed by the abundance of *Pseudomonas putida-* and *Pseudomonas azotoformans*-associated sequences, accounting for 19% and 15%, respectively (Fig. 5B, Table S10). In addition, predicted genes coding for multicopper oxidase enzymes were identified in the genome of *Pseudomonas-*associated Bin (Table S8). These results imply that laccase-like multicopper oxidase gene in the gut of giant pandas may be mainly contributed by bacteria in the *Pseudomonas* genus.

Isolation and identification of Pseudomonas-associated bacteria

Three most abundant *Pseudomonas*-associated bacteria were obtained based on colonial morphology difference in the *Pseudomonas*-associated bacteria solid screening medium after three repeated cultivations. These *Pseudomonas*-associated isolates were closely matched to *Pseudomonas* putida strain cqsH1 (99%), *Pseudomonas* sp. strain QW16-14 (99%), and *Pseudomonas* oryzihabitans strain h-2 (99%) based on 16S rRNA gene sequence homology.

Lignin degradation capability of single Pseudomonas-associated bacteria

Three *Pseudomonas*-associated isolates grew well on a solid medium with lignin as the sole carbon source (Fig. 6A). The growth and lignin-degrading curves of *Pseudomonas* putida, *Pseudomonas* sp., and *Pseudomonas* oryzihabitans are shown in Fig. 6B. The lag phase, exponential growth phase, stationary phase, and decline phase occurred at 0-12 h, 24-72 h, 84-120 h, and 132-168 h, respectively. The initial average absorbance of the lignin culture medium was 4.47 at 280 nm, which decreased to 3.21 after 7 days of incubation. The degradation rates for *Pseudomonas* putida, *Pseudomonas* sp., and *Pseudomonas* oryzihabitans after 7 days were 19.66%, 19.14%, and 18.17%, respectively (Fig. 6C).

Extracellular ligninolytic enzymes of *Pseudomonas*-associated isolates

The three *Pseudomonas*-associated isolates secreted extracellular ligninolytic enzymes and produced a transparent ring after 24 h of incubation on aniline blue solid medium (Fig. 7A). However, most aniline blue was decolorized by bacteria after 48 h (Fig. 7A). The aniline blue degrading curve of *Pseudomonas* putida, *Pseudomonas* sp., and *Pseudomonas* oryzihabitans are shown in Fig. 7B. The degradation rates of Pseudomonas putida, Pseudomonas sp., and Pseudomonas oryzihabitans after 96 h were 90.74%, 88.97%, and 86.77%, respectively. The enzyme production curve of three *Pseudomonas*-associated isolates is shown in Fig. 8C-E. The results showed that the three *Pseudomonas*-associated isolates secreted considerable reactive Lac, LiP, and MnP. Furthermore, the three Pseudomonas-associated isolates showed a similar characteristic of secreting extracellular ligninolytic enzymes in the culture medium with bamboo powder as the sole carbon source. The activities of Lac, LiP, and MnP reached the maximum values after 72 h. The maximum values of Lac, LiP, and MnP were 688.25 U/L, 3065.37 U/L, and 493.22 U/L, respectively, for Pseudomonas putida, 653.75 U/L, 2700.54 U/L, and 439.53 U/L, respectively, for Pseudomonas sp., and 459.12 U/L, 2668.54 U/L, and 404.26 U/L, respectively, for Pseudomonas oryzihabitans. The average Lac activity of Pseudomonas putida, Pseudomonas sp., and Pseudomonas oryzihabitans was 309.51, 251.58, and 209.52 U/L, respectively (Fig. 7C), while the average MnP activity was 300.08, 269.84, and 258.13 U/mL, respectively (Fig. 7E). The average LiP activity of *Pseudomonas putida*, *Pseudomonas* sp., and *Pseudomonas* oryzihabitans was 2130.5, 2083.12, and 2069.75 U/mL, respectively (Fig. 7D).

Metabolites of degraded lignin by Pseudomonas-associated isolates

Compared with the control (0 day), LC-MS analysis detected new peaks in the culture medium after treatment with *Pseudomonas*-associated strain for three and seven days (Figure S8). The PCA score plot displayed three clusters that corresponded to different cultivation time with *Pseudomonas*-associated strain. Metabolites in the culture medium after treatment with *Pseudomonas*-associated strain were significantly different compared with the control (0 day) (Figure 8A). Distinctive clustering in the metabolite composition of the supernatant was detected between the control group (0 days) and the culture group (Figure S9). A total of 229 and 245 differential metabolites were detected between 0 day vs. 3 day and 0 day vs. 7 day, respectively, of which 201 were shared. Compared with the control (0 day), cluster analysis indicated that the 201 metabolites in the culture medium were significantly different after

treatment with *Pseudomonas*-associated strains for 3 and 7 days (Figure 8B). LC-MS analysis detected 14 of 25 substances for the products in the beta-ketoadipate and homogentisate pathway (Figure 1). Among them, cis,cis-muconate, hydroxybenzaldehyde, vanillate, phenylpyruvate, 2-hydroxyphenylacetate, homogentisate, fumarylacetoacetate, and fumaric acid were only identified in the supernatant of culture medium after treatment with *Pseudomonas* putida, *Pseudomonas* sp., and *Pseudomonas* oryzihabitans (separately) for 3 days and 7 days (Table 2). Lignin derivatives, such as, catechol, ferulate, 4-coumarate, and protocatechuate, were detected in the control group (Table 2). Nevertheless, the relative quantification of the above derivatives was higher on the 3rd and 7th days (Figure 8C).

Discussion

The present study is the first to reveal the complete metabolic pathway involved in the degradation of lignin derivatives by the gut microbiome of giant pandas. The β-ketoadipate pathway, comprising the catechol and protocatechuate branches, is the major microbial degradation pathway for lignin-derived aromatic compounds [39, 40]. Gut microbiota, such as Rhodococcus jostii RHA1 [41], Brevibacillus thermoruber [42], Bacillus ligniniphilus [43], and Pseudomonas putida KT2440 [44], degrade lignin mainly through β-KAP. In addition, the homogentisate pathway plays a key role in the degradation of ligninderived aromatic compounds [45]. Degradation pathways for 4-hydroxyphenylacetate [45], phenylalanine, and tyrosine eventually join the homogentisate pathway [46]. Notably, genes involved in the β -KAP and homogentisate pathways were enriched in the gut microbiotas of wild giant pandas. The relative abundance of these genes was significantly higher in wild giant pandas than in captive giant pandas, herbivores, carnivores, and omnivores. However, the abundance of these genes in wild giant pandas was similar to those in another bamboo-eating panda (red panda). This indicates that the gut microbiome of wild giant pandas plays a key role in catalyzing cleavage of the benzene ring through the metabolic pathway involved in breakdown of central aromatic intermediates. Of note, the gut microbiotas of wild giant pandas showed significantly enriched genes encoding several enzymes that catalyze the degradation of ferulic and p-Coumarate to protocatechuate, which subsequently join the PbB-KAP pathway. Ferulic and p-Coumarate are lignin monomers [47, 48] produced during the initial degradation reaction of lignin catalyzed by laccase [49]. Laccase-like multicopper oxidase gene, which plays a similar role in degrading lignin [50], was also abundant in the feces of wild giant pandas, and was mainly contributed by *Pseudomonas*-associated bacteria. Therefore, most genes implicated in lignin degradation were identified in the genome of *Pseudomonas*-associated bacteria of wild giant pandas. Notably, the final product of lignin degradation (fumarate, acetyl-CoA, and succinyl-CoA) can enter the TCA cycle to produce nutritionally important intermediates. The lignin in bamboo is mainly of the HGS type (phydroxyphenyl (H), vanillin (G), syringaldehyde (S)), containing a considerable amount of p-coumarate and ferulic [51]. The β-ketoadipate pathway is the major lignin-degrading pathway in bacteria isolated from erosive bamboo slips [52, 53]. This explains why the gut microbiotas of wild giant pandas had significantly high expression levels of genes encoding enzymes implicated in *p*-coumarate and ferulic degradation through the β-ketoadipate pathway. Previous studies have shown that the genome of gut microbiotas of captive giant pandas lacks genes encoding enzymes involved in lignin degradation [17].

The reason for these findings is that the gut microbiome composition of captive giant pandas is significantly different from that of wild giant pandas [19]. For instance, the dominant gut bacteria in wild giant pandas is *Pseudomonas*, whereas the dominant gut bacteria of captive giant pandas are *Enterobacteriaceae* and *Streptococcus* [19, 20]. Moreover, red pandas that exclusively eat bamboo exhibit a similar phenomenon as wild and captive population [26]. The results of the present study suggest that the study of the adaptive evolution of animal intestinal flora to diet should be conducted using wild populations rather than captive populations.

The genome of *Pseudomonas*-associated strains (Bin 1, Bin 17, Bin 33, and Bin 53) in the feces of giant pandas can encode all enzymes involved in the metabolism of central aromatic intermediates. Pseudomonas is the most efficient lignin degradation bacterium [54]. The four closest neighbors of Pseudomonas-associated strains in the feces of giant pandas were Pseudomonas fluorescens PfO-1, Pseudomonas syringae pv. phaseolicola 1448A, Pseudomonas putida KT2440, and Pseudomonas aeruginosa PA01. Pseudomonas fluorescens [55], Pseudomonas putida [36] and Pseudomonas aeruginosa [56] have high potential in lignin degradation. Although Pseudomonas syringae is generally considered a plant pathogen [57], its genome contains genes involved in the β-Ketoadipate and homogentisate pathway [58]. Furthermore, in vitro culture experiments showed that three Pseudomonasassociated strains isolated from the feces of wild giant pandas have a lignin degradation ability. Besides, multi-copper oxidases gene was identified in Pseudomonas syringae [59], further confirming that Pseudomonas-associated bacteria in the feces of wild giant panda play a crucial role in lignin degradation. *Pseudomonas* is the most dominant bacterium in the gut of wild giant pandas [19, 20]. In this study, the abundance of *Pseudomonas*-associated Bins (Bin 1, Bin 17, Bin 33, and Bin 53) was higher than other Bins, implying that Pseudomonas is the main lignin-degrading bacteria in the gut of giant pandas. Laccase-like multicopper oxidase genes in the gut of giant pandas are mainly derived from Pseudomonas, consistent with the findings on microbiome abundance. Fang et al. reported that multicopper oxidase in the gut of giant pandas was derived from *Pseudomonas* sp, and the enzyme showed activity for oxidative degradation of lignin [23]. The three Pseudomonas-associated strains could also secret extracellular Lac, Lip, and MnP in the culture medium with bamboo powder as the sole carbon source. This is the first study to show that *Pseudomonas*, the most dominant bacteria genus in the gut of giant pandas, can degrade lignin.

Based on these findings, we proposed a potential model for lignin degradation by gut microbiotas of giant pandas (Fig. 9). In this model, lignin is first depolymerized by extracellular Lac, Lip, and MnP to form lignin derivatives or monomers (p-Coumarate, Ferulate, Phenylpyruvate, etc). Lignin monomers (*p*-Coumarate and Ferulate, etc.) are then degraded into aromatic compounds, such as Vanillate, 4-Hydroxybenzoate, and 2-Hydroxyphenylacetate. The aromatic compounds are further metabolized to form fumarate, acetyl-CoA, and succinyl-CoA through the β-Ketoadipate pathway and homogentisate pathway. Finally, fumarate, acetyl-CoA, and succinyl-CoA enter the TCA cycle of giant pandas for the biosynthesis of metabolically important products. *Pseudomonas* is the most dominant bacteria in the gut of wild giant pandas [19, 20]; thus, it plays a key role in this model.

Pseudomonas putida KT2440, a model strain for lignin degradation, can package enzymes implicated in lignin degradation in outer membrane vesicles (OMVs) and release them into the extracellular space to catalyze lignin degradation [36]. Moreover, several Pseudomonas strains, including Pseudomonas aeruginosa [60] and Pseudomonas syringae [61], can release OMVs to facilitate the metabolism of lignin in vitro. This indicates that Pseudomonas-associated strains in the gut of giant pandas potentially release OMVs containing enzymes involved in lignin degradation into the extracellular space (in the gut of giant pandas) to degrade lignin. In this study, experiments of aniline blue degradation and enzyme activity determination in vitro confirmed that Pseudomonas-associated strains can release extracellular ligninolytic enzymes to degrade lignin. In addition, *Pseudomonas*-associated strains can degrade lignin into some important raw materials, such as fumarate, acetyl-CoA, and succinyl-CoA, for the TCA in vitro. Notably, LC-MS did not detect acetyl-CoA and succinyl-CoA in the culture solution, indicating low content or other detection methods are needed. However, these results indicate that the gut of giant panda may assimilate the final products (fumarate, acetyl-CoA, and succinyl-CoA) of lignin degradation in the extracellular space into the TCA cycle to obtain nutritionally important compounds. This hypothesis effectively explains how giant pandas potentially obtain nutrition from low nutrition bamboo diet. However, additional studies are needed to verify if the intestinal tract of giant pandas can absorb lignin degradation products for biosynthesis metabolism. In summary, this study provides new insights into how giant pandas obtain nutrition from bamboo.

Conclusion

The gut microbiome of wild giant pandas exhibits high expression levels of genes implicated in lignin derivative degradation pathways. These pathways include Catechol branch of beta-ketoadipate pathway, Protocatechuate branch of beta-ketoadipate pathway, homogentisate pathway of aromatic compound degradation, and pathways for degradation of lignin monomers (*p*-Coumarate and Ferulate) into the important raw materials for TCA via beta-ketoadipate pathway, such as Acetyl-CoA and Succinyl-CoA. All these pathways can be found in the genome of the most dominant bacteria genus, *Pseudomonas*-associated strains. Furthermore, results showed that *Pseudomonas*-associated strains isolated from the feces of pandas can degrade extracellular lignin *in vitro*. Overall, these results indicate that the gut microbiome of wild giant pandas plays a crucial role in lignin degradation. The most dominant bacteria, *Pseudomonas*, is the main lignin-degrading bacteria in giant pandas. These findings provide novel insights into the adaptive evolution of giant pandas from the perspective of lignin metabolism.

Materials and Methods

Sample collection and individual information

Fecal samples were collected from both wild and captive animals. Fresh fecal samples were collected from wild giant pandas (*Ailuropoda melanoleuca*) (n = 7) and red pandas (*Ailurus fulgens*) (n = 5) in Fengtongzhai Nature Reserve (Ya'an, Sichuan Province, China). In addition, fecal samples from captive giant pandas (n = 7) and red panda (n = 5) were collected from the China Conservation and Research

Center and Bifengxia Ecological Zoo (Ya'an, Sichuan Province, China), respectively. The samples were immediately frozen in liquid nitrogen, then transferred to – 80 °C refrigerator for further analysis.

Previous studies showed that the seven fecal samples from wild giant pandas were from seven different animals [62, 63]. The five wild panda samples have also been confirmed to be from different individuals via GPS-collared. The detailed sample information, including age, gender, and population, is shown in Table S1.

DNA extraction and sequencing

Total microbial genomic DNA was extracted from the fecal samples using the PowerFecal DNA isolation kit (QIAGEN Inc., Valencia, CA, USA), according to the manufacturer's instructions. The DNA concentration and purity were measured using the Qubit (Thermo Fisher Scientific, Waltham, MA, USA). Agarose gel electrophoresis was performed to assess the DNA quality. The DNA samples that met the criteria of metagenomic sequencing were used for library preparation: (1) DNA concentration of > 15 ng/uL; (2) total amount of DNA > 6 μ g; (3) non-contaminated and intact DNA fragment.

Shotgun metagenomic DNA libraries were constructed based on the Illumina TruSeq DNA Sample Prep V2 Guide (Illumina, Inc; San Diego, CA, USA), with shearing to 300–400 bp fragments. Shotgun metagenomic sequencing was performed on Illumina platform using paired-end 2 × 150 bp chemistry (Novogene, Beijing, China).

Downloading of gut metagenomic data

The composition of gut microbiotas is significantly different between captive and wild animals. In this study, only datasets from wild mammals were included for joint analysis, except for bamboo-eating animals. Meanwhile, only the metagenome data generated by Illumina platform were downloaded to avoid the possible bias caused by different sequencing platforms. Gut metagenomics data of six red pandas and eight giant pandas from the study of Zhu et al. (2018) were retrieved from the National Centre for Biotechnology Information's Sequence Read Archive (SRA) [38]. The previously published mammalian metagenome data set representing herbivores (n = 97), omnivores (n = 10), and carnivores (n = 14) were downloaded from Levin et al.'s study [64]. In addition, the published data of tiger (n = 6) [65], black rhinoceros (n = 17) [66], David's deer (n = 30) [38], Gazella Subgutturosa (n = 4) [67], Chinese pangolin (n = 5), and Malayan pangolin (n = 5) were also downloaded from the Genome Sequence Archive (GSA) [68]. The published raw metagenome sequences used in this study was showed in Table S2.

Shotgun metagenomic sequence analysis

Adapter sequences were removed from all the sequence data using Cutadapt v1.9.1 software [69]. Raw paired-end reads were processed by Trimmomatic software to filter out low-quality sequences using a sliding window (5 bp bases) [70]. The criteria for quality control were as follows: (1) a sequence was

removed once its average quality within the window fell below Q20; (2) sequences containing any Nbases were filtered out; (3) reads that were below 50 bp in length were dropped; (4) only paired-end reads were retained. Animal feces often contain cells shed from their intestines. Therefore, total bacterial DNA extracted from animal feces usually contain the DNA of the host, indicating that the metagenomic sequencing data may contain part of the host genome sequences. After quality control, Bowtie2 [71] was employed to blast the dataset of each species to their genome sequences. Sequences with > 90% similarity to the genome sequences of the host were removed.

High-quality reads were assembled into contigs using metaSPAdes with the default parameters [72]. Assembled contigs with more than 500 bp in length were retained for subsequent analyses. Open reading frames (ORFs) of gut microbiome of each species were predicted from the assembled contigs using MetaGeneMark ORFs ark v 2.8 [73]. The non-redundant gut microbial gene set of each species was generated with a 90% identity cutoff using CD-HIT v4.8.1 [74]. Data were randomly sub-sampled to the minimum number of sequences in all samples using seqtk-master

(https://github.com/lh3/seqtk) before aligning the sequences against the non-redundant panda gut microbiome gene set to reduce biases caused by different sequence depths. Bowtie2 [71] was used to align the paired-end clean reads of each sample against their non-redundant gut microbial gene set. Gene abundance A(g) in each sample was determined according to Qin et al. [75] as follows:

$$A(g) = \frac{N(g)}{L(g)}$$

where 'N denotes the number of paired-end clean reads mapped to a gene, 'L' denotes gene length, and 'g' denotes gene.

The relative abundance of each gene in each sample (RA(g)) was calculated as follows:

$$RA(g) = \frac{N(g)}{L(g)} * \frac{1}{\sum_{i=1}^{n} \frac{Ni}{Li}}$$

Non-redundant gut microbial gene set of each species was aligned against the Kyoto Encyclopedia of Genes and Genomes (KEGG) online database to annotate the gene functions [76]. GHOSTX searches were selected as the assignment method, and other parameters were set as default [77].

Kraken2 was used for taxonomic classification on the clean reads to explore the relative abundance of *Pseudomonas*-associated bacteria in the gut of bamboo-eating pandas and other mammals.

Extraction of individual draft genomes (bins) from metagenomic data of wild giant pandas

High-quality individual draft genomes (Bins) of gut bacteria of wild giant pandas were obtained from shotgun metagenomic assemblies using MetaWRAP (v1.2) to explore the gut strain of giant pandas with

the potential to degrade lignin [78]. Metagenomic assemblies were binned into draft genomes using the Binning module in three metagenomic binning software (MaxBin2, metaBAT2, and Concoct). All bins recovered from the assembly were combined to obtain a single complete bin set using Bin_refinement module. Furthermore, Reassemble_bins module was used to improve N50, completion, and reduce contamination of the bins. Reassembled bins with completeness > 70% and contamination < 5% were considered as high-quality individual draft genomes (Bins) for subsequent analysis. Estimation of bin abundances across samples was performed using the metaWRAP-Quant_bins module. Bins were uploaded to RAST (Rapid Annotations using Subsystems Technology) web server for the identification of species and potential functions of the genome. All parameter options were set at default, and the closest neighbors of bins (with the highest' Identify Score') were considered as the possible species. A phylogenetic tree was constructed based on all bins using PhyloPhIAn 3.0 tool [79]. Furthermore, 40 complete (Table S3) *Pseudomonas* genomes representing the four closest neighbors of *Pseudomonas*-associated strains recovered by binning were downloaded for phylogenetic analysis.

Laccase-like multicopper oxidase gene cloning and sequencing in the feces of giant pandas

Laccase-like multicopper oxidase gene in the feces of wild giant pandas was cloned and sequenced to explore the presence of laccase gene that can degrade lignin into derivatives in the gut microbiotas in giant pandas. Three fresh fecal samples of wild giant pandas were used for the extraction of genomic DNA. Total genomic DNA of microbiotas was extracted from feces following as described in section 2. Laccase-like multicopper oxidase gene short fragments (cbr1- cbr2) were amplified via PCR using the degenerate primers Cu1AF (5' -ACM WCBGTY CAY TGG CAY GG-3') and Cu2R (5' -G RCT GTGGTA CCA GAA NGT NCC-3') [80]. The PCR volume was 25 μ L comprising 2.5 μ L 10×ExTaq Buffer, 2.0 μ L dNTP Mixture (5 mM each), 1 μ L forward primer (10 μ M), 1 μ L reverse primer (10 μ M), 0.5ul Taq DNA polymerase (10 U/ μ), 1.0 μ L DNA template (25 ng) and 17.0 μ L ddH2O. Thermocycling parameters included an initial denaturing step at 95°C for 3 min, 30 cycles of 95°C for 30 s, 55°C for 30 s, and 72°C for 60 s and a final extension at 72°C for 5 min.

PCR products were detected using 2.0% agarose gel electrophoresis, and fragments (~ 150 bp) were obtained for subsequent analysis. Bands with the expected fragment size were purified using Axygen Gel Extraction Kit (Axygen, Silicon Valley, USA) according to the manufacturer's instructions. The purified target fragment was ligated into the pMD19-T vector system (TaKaRa, Dalian, China) and transformed in *Escherichia coli* DH5α cell according to the manufacturer's instructions. A total of 50–85 positive clones per fecal sample were selected. The size of the inserted bacterial laccase-like gene fragment was evaluated via PCR, and sequencing was performed by Beijing Genomics Institute (BGI, Shenzhen, China) using primer RV-M. Nucleotide sequences were manually proofread, and similar sequences were obtained through BLAST search in GenBank. Nucleotide sequences were translated into amino acid sequences and were aligned using ClustalW program [81]. Sequence analysis and phylogenetic reconstruction were performed using Mega 11 software [82]. The neighbor-joining tree was reconstructed based on the

Maximum Composite Likelihood method. The strains encoding Laccase-like multicopper oxidase enzyme were identified as different species.

Isolation and identification of Pseudomonas-associated bacteria

Pseudomonas-associated bacteria were screened in fresh fecal samples of wild giant pandas. Briefly, 1 g fecal sample was dissolved in 100 mL sterile saline solution to form a fecal solution. The fecal solution was diluted 10^{-7} times using sterile saline solution for single colony plotting on an agar plate. The diluted fecal solution (200µL) was inoculated into *Pseudomonas*-associated bacteria solid screening medium containing 16.0 g/L gelatin peptone, 10.0 g/L Acid hydrolyzed casein, 10.0 g/L K SO , 1.4 g/L Mg₂Cl, 0.2 g/L cetyltrimethylammonium bromide (CTMAB), 10 mL/L glycerol, 15 mg/L Nalidixic acid, and 12.0 g/L agar. Single colonies were obtained after incubating the sample at 37 °C for 24 h. The single colonies were re-streaked thrice to obtain a pure colony for strain identification.

Single colonies were incubated with LB liquid medium at 37 °C, 200 rpm for 24 h. The LB liquid medium was then centrifuged at 12,000 × g for 5 min, and the supernatant was discarded. DNA isolation kit (Tiangen Biotech, Beijing, China) was used to extract genomic DNA of the bacterial cells following the manufacturer's instructions. The 16S rDNA sequence was amplified with bacterial universal primers 27F (AGAGTTTGATCCTGGCTCAG) and 1492R (GGTTACCTTGTTACGACTT). The amplified fragment was sequenced by Tsingke Biotechnology Co., Ltd. The obtained sequences were compared with the NCBI sequences using the BLAST website.

Determination of lignin-degrading capability of Pseudomonas-associated bacteria

The isolated *Pseudomonas*-associated strains were inoculated into modified M9 solid medium containing 6.78 g/L Na2HPO4, 3 g/L KH2PO4, 0.5 g/L NaCl, 1 g/L NH4CL, 2 mM MgSO4, 100 μ M CaCl2, 100 μ M MnSO4, 100 μ M CuSO4, 100 μ M FeSO4, 50 μ M ZnSO4, 2 g/L kraft lignin, and 12.0 g/L agar (with lignin as the sole carbon source; PH = 7.0). The cell growth and lignin degradation were determined to test the ligninolytic activity of the *Pseudomonas*-associated strains. *Pseudomonas*-associated strains were inoculated in Luria Broth (LB) medium and cultured at 37 °C and 200 r/min for 24 h while shaking. The culture broth was centrifuged at 12,000 × g for 5 min to obtain the bacterial cells. The bacterial cells were inoculated into 100 mL of modified M9 medium (PH = 7.0) containing 2 g/L kraft lignin (initial OD600 = 0.1), then cultured at 37 °C, 200 rpm for 7 d. Fermentation broth (1 mL) was taken daily to measure the growth curve and lignin-degrading capability of the *Pseudomonas*-associated strains. Cell growth was determined at OD600, then the growth curve of *Pseudomonas*-associated strains was plotted. The lignin degradation rate was evaluated by monitoring the decrease in A280. Degradation ratio was calculated as follows: Degradation% =(A_{control}-A_{sample})/A_{control}*100%. All the experiments were performed in triplicates.

The decolorization capability for aromatic dyes and enzyme assays

The capability of the screened isolates to decolorize the aromatic dyes with almost similar structures to lignin fragments was determined to further evaluate whether *Pseudomonas*-associated strains can secrete extracellular ligninolytic enzymes to degrade lignin. The *Pseudomonas*-associated strains were inoculated into Aniline blue solid medium (PH = 7.0) containing10 g/L glucose, 10 g/L peptone, 0.04 g/L water-soluble aniline blue, and 12 g/L agar. The decolorization circle of the *Pseudomonas*-associated strains on aniline blue solid medium was confirmed, then the bacterial cells with OD600 = 0.1 were inoculated into 50 mL aniline blue liquid medium (PH = 7.0). The samples were cultured at 37 °C and 200 r/min for 96 h while shaking. The aniline blue liquid medium without bacteria was used as the blank control. The culture medium was centrifuged at 12,000 rpm for 5 min every 24 h, and the absorbance of supernatants was measured at 600 nm. The decolorization ratio was calculated as follows: Decolorization % = (A_{control}-A_{sample})/A_{control}*100%. All the experiments were performed in triplicates.

The *Pseudomonas*-associated strains were inoculated into 100 mL modified M9 medium (PH = 7.0) containing 2 g/L bamboo powder at an initial OD600 of 0.1, then cultured at 37 °C and 200 r/min for 7 days while shaking. The culture medium was centrifuged at 12,000 rpm for 2 min every 24 hours, and the supernatants were used to determine extracellular ligninolytic enzymes. Laccase (Lac) activity was determined by measuring the oxidation of ABTS (2,2-azino-bis (3-ethylbenzothiazoline-6-sulphonic acid)) to ABTS radical at 420 nm using a reaction mixture (200 µL) containing 180 µL ABTS (0.5 mmol/L) (dissolved in 0.1 mmol/L HAc-NaAc buffer solution, pH 4.5), and 20 µL cell-free supernatant. Lignin peroxidase (LiP) was evaluated by monitoring the oxidation rate of Azure B at 651 nm using a reaction mixture (200 µL) containing 180 µL Azure B (0.5 mmol/L) (dissolved in 0.1 mmol/L HAc-NaAc buffer solution, pH 4.5), and 20 µL cell-free supernatant. Lignin peroxidase (LiP) was evaluated by monitoring the oxidation rate of Azure B at 651 nm using a reaction mixture (200 µL) containing 180 µL Azure B (0.5 mmol/L) (dissolved in 0.1 mmol/L HAc-NaAc buffer solution, pH 4.5), and 20 µL cell-free supernatant. Mn-peroxidase (MnP) activity was evaluated by monitoring the oxidation of 2,6-DMP (2,6-dimethyl phenol) to coerulignone at 469 nm using a 200 µL reaction mixture containing 180 µL 2,6-DMP (0.5 mmol/L) (dissolved in 0.1 mmol/L HAc-NaAc and MnSO4 buffer solution, pH 4.5) and 20 µL cell-free supernatant. One unit of enzyme activity was defined as the amount of enzyme required to oxidize 1 µmol substrate per minute. All fermentation tests were performed in triplicate.

Determination of degradation products of lignin by Pseudomonas-associated strains

The centrifugal supernatants of lignin fermentation medium of *Pseudomonas*-associated strains were filtered using a 0.2 µm filter membrane, and quality control samples were prepared. Metabolites of the supernatants were extracted before LC-MS detection, as previously described [83]. Each sample was run in triplicate, and the stability of mass spectrometry was evaluated. LC-MS detection was performed as previously described [84, 85]. The raw data were first converted to mzXML format via MSConvert in ProteoWizard software package (v3.0.8789) [86] and processed using XCMS [87] for feature detection, retention time correction and alignment. The metabolites were identified via accuracy mass (< 30 ppm) and LC-MS data, which were matched with Human Metabolome Database (HMDB; http://www.hmdb.ca/) [88], MassBank (http://www.massbank.jp/ [89], LipidMaps (https://www.lipidmaps.org) [90], mzcloud (https://www.mzcloud.org) [91] and KEGG (https://www.kegg.jp) [92]. The robust LOESS signal correction (QC-RLSC) was applied for data normalization to correct any systematic bias [93]. Ion peaks with relative

standard deviations (RSDs) less than 30% in QC were kept ensuring proper metabolite identification. Relative quantification was estimated based on the ratio of the respective peak areas to the total peak areas.

Statistical Analysis

Significant differences in the relative abundance of genes between different diet groups were determined using the Kruskal-Wallis test followed by Dunn's multiple comparisons post-hoc test. All statistical analyses were performed using GraphPad Prism 7 software (GraphPad Software, Inc., USA). Figures were generated using "pheatmap" packages [94] and different functions ("boxplot," "barplot," "pie", and "plot") in R base (version 3.6.1) [95]. Multivariate statistical analysis and modeling were conducted using the resultant datasets from LC-MS procedures via the RopIs software [96]. Models were built based on principal component analysis (PCA) to explore dataset variations. The P-value and variable importance projection (VIP) produced by OPLS-DA were applied to discover the significance of metabolites (P value < 0.05 and VIP values > 1) in different groups.

Declarations

Ethics approval

This study was approved by the Institutional Animal Care and Use Committee of CMC-B20190922.

Availability of data and materials

The raw data of metagenome sequences in this study have been deposited into Sequence Read Archive (SRA) in NCBI with the accession bioproject number PRJNA356809.

Declaration of Competing Interest

The authors declare no conflict of interest.

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Authors' contributions

W.G. designed this study. W.G. and R.N. wrote the manuscript. W.G., R.N., S.Z., and F.S. analyzed the data. W.G., R.N., S.Z., H.W., Y.Z., and Y.G. performed the laboratory experiments. C.L., D.L., Y.H., Z.G., W.L., and Z.Q. collected fecal samples of wild giant and red pandas. All authors read the manuscript, and approved submission of the final draft.

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Tables

Table 1. Species and Potential functions identification of individual bacteria draft genomes in wild giant pandas.

Binning Closest Genome ID Name

Catechol branch of βketoadipate pathway

Protocatechuate branch of βketoadipate pathway Homogentisate pathway of aromatic compound degradation

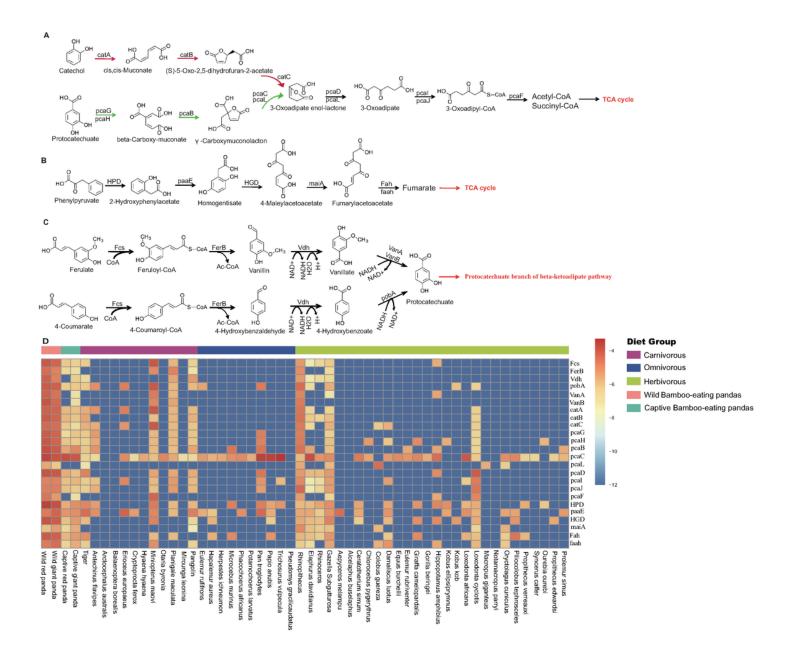
Bin1	Pseudomonas	catA, catB, catC	pcaG, pcaH, pcaB, pcaC, pcaL, pcaD, pcal, pcaJ	HPD, HGD, maiA, Fah, faah, paaE
Bin2	Microbacteriaceae	NA	NA	NA
Bin3	Cellvibrionaceae	NA	NA	NA
Bin4	Lysinibacillus	NA	NA	NA
Bin5	Streptococcaceae	NA	NA	NA
Bin6	Rhizobiales	NA	NA	NA
Bin7	Leuconostoc	NA	NA	NA
Bin8	Microterricola viridarii	NA	NA	NA
Bin9	Achromobacter	catA, catB ,catC	pcaG, pcaH, pcaB ,pcaC, pcaL, pcaD, pcal, pcaJ	HPD, HGD, maiA, Fah, faah
Bin10	Flavobacterium	NA	NA	HPD, HGD, maiA, Fah, faah
Bin11	Stenotrophomonas sp. LM091	NA	NA	HPD, HGD, maiA, Fah, faah
Bin12	Comamonas	catB ,catC	NA	NA
Bin13	Hafniaceae	NA	NA	NA
Bin14	Oxalobacteraceae	catA, catB ,catC,	pcaG, pcaH, pcaB ,pcaC, pcaL, pcaD, pcal, pcaJ	NA
Bin15	Flavobacterium sp. 140616W15	NA	NA	HPD, HGD, maiA, Fah,, faah
Bin16	Enterobacteriaceae	NA	NA	NA
Bin17	Pseudomonas	catA, catB, catC	pcaG, pcaH, pcaB, pcaC, pcaL, pcaD, pcal, pcaJ	HPD, HGD, maiA, Fah, faah,

Bin18Sphingebacterium sp. PN2-97-29NANANABin19LysinibacillusNANANABin20YersiniaNANANABin21Culibacterium achesNANANABin22ClostridiumNANANABin23Sphingobacterium AlcaligenaceaeNANANABin24AlcaligenaceaeNANANABin25YersiniaceaeNANANABin26JanthinobacteriumcatA, catB, catCNANABin27FlavobacteriumNANANABin28LeuconostocNANANABin30RahnellaNANANABin31JanibacterNANANABin32LeuconostocNANANABin33JanibacterNANANABin34JanibacterNANANABin35JanibacterNANANABin36ClostridumNANANABin37StreptococcaseeNANANABin38StreptococcaseeNANANABin39EnterobacteriaceeeVNANABin39LipsinbacillusNANANABin39JanibacterNANANABin39JanibacterNANANABin39JanibacterNANANABin39JanibacterNA <th></th> <th></th> <th></th> <th></th> <th></th>					
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Bin21Cutibacterium acnesNANANABin22ClostridiumNANANABin23SphingobacteriumNANANABin24AlcaligenaceaeNANANABin25YersiniaceaeNANANABin26JanthinobacteriumcatA, catB, catCNAHPD, HCD, maiA, Fah, faahBin27FlavobacteriumNANANABin28LeuconostocNANANABin30RahnellaNANANABin31JanibacterNANANABin32LeuconostocNANANABin33SeudomonascatA, catB, catCNANABin34LeuconostocNANANABin35JanibacterNANANABin36LeuconostocNANANABin37SreudomonascatA, catB, catCpcaG, pcaH, pcaB, pcaC, 	Bin19	Lysinibacillus	NA	NA	NA
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Bin30RahnellaNANABin31JanibacterNANABin32LeuconostocNANABin32LeuconostocNANABin33PseudomonascatA, catB, catCpcaG, pcaH, pcaB, pcaC, pcaL, pcaD, pcaL, pcaDHPD, HGD, maiA, Fah, faah, faah,Bin34ArthrobacterNANABin35JanibacterNANABin36ClostridiumNANABin37StreptococcaceaeNANABin38Streptococcus pasteurianusNANABin39EnterobacteriaceaeNANABin40ComamonascatB, catCNANABin41LysinibacillusNANABin42CutibacteriumNANA	Bin28	Leuconostoc	NA	NA	NA
Bin31JanibacterNANAHPD,HGD,FahBin32LeuconostocNANANABin33PseudomonascatA, catB, catCpcaG, pcaH, pcaC, pcaD, pcaD, pcaC, pcaD, pcaC, pcaD, pcA, pcA, pA, pcA, pA, pA, pc	Bin29	Comamonadaceae	catA, catB, catC	NA	NA
Bin32LeuconostocNANABin32LeuconostocNANABin33PseudomonascatA, catB, catCpcaG, pcaH, pcaB, pcaC, pcaL, pcaD, pcaI, pcaJ,	Bin30	Rahnella	NA	NA	NA
Bin33PseudomonascatA, catB, catCpcaG, pcaH, pcaB, pcaC, pcaL, pcaD, pcal, pcaD,HPD, HGD, maiA, Fah, faah,Bin34ArthrobacterNANANABin35JanibacterNANAHPD,HGD,FahBin36ClostridiumNANANABin37StreptococcaceaeNANANABin38Streptococcus pasteurianusNANANABin39EnterobacteriaceaeNANANABin40ComamonascatB,catCNANANABin41LysinibacillusNANANABin42CutibacteriumNANANA	Bin31	Janibacter	NA	NA	HPD,HGD,Fah
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Bin35JanibacterNANAHPD,HGD,FahBin36ClostridiumNANANABin37StreptococcaceaeNANANABin38Streptococcus pasteurianusNANANABin39EnterobacteriaceaeVVVBin40ComamonascatB,catCNANABin41LysinibacillusNANANABin42CutibacteriumNANANA	Bin33	Pseudomonas	catA, catB, catC	pcaB, pcaC, pcaL, pcaD,	HPD, HGD, maiA, Fah, faah,
Bin36ClostridiumNANABin37StreptococcaceaeNANANABin38Streptococcus pasteurianusNANANABin39EnterobacteriaceaeVVVBin40ComamonascatB,catCNANABin41LysinibacillusNANANABin42CutibacteriumNANANA	Bin34	Arthrobacter	NA	NA	NA
Bin37StreptococcaceaeNANANABin38Streptococcus pasteurianusNANANABin39EnterobacteriaceaeBin40ComamonascatB,catCNANABin41LysinibacillusNANANABin42CutibacteriumNANANA	Bin35	Janibacter	NA	NA	HPD,HGD,Fah
Bin38Streptococcus pasteurianusNANANABin39EnterobacteriaceaeBin40ComamonascatB,catCNANABin41LysinibacillusNANANABin42CutibacteriumNANANA	Bin36	Clostridium	NA	NA	NA
pasteurianusBin39EnterobacteriaceaeBin40ComamonascatB,catCNABin41LysinibacillusNANABin42CutibacteriumNANA	Bin37	Streptococcaceae	NA	NA	NA
Bin40ComamonascatB ,catCNANABin41LysinibacillusNANANABin42CutibacteriumNANANA	Bin38		NA	NA	NA
Bin41LysinibacillusNANANABin42CutibacteriumNANANA	Bin39	Enterobacteriaceae			
Bin42 <i>Cutibacterium</i> NA NA NA	Bin40	Comamonas	catB ,catC	NA	NA
	Bin41	Lysinibacillus	NA	NA	NA
	Bin42		NA	NA	NA

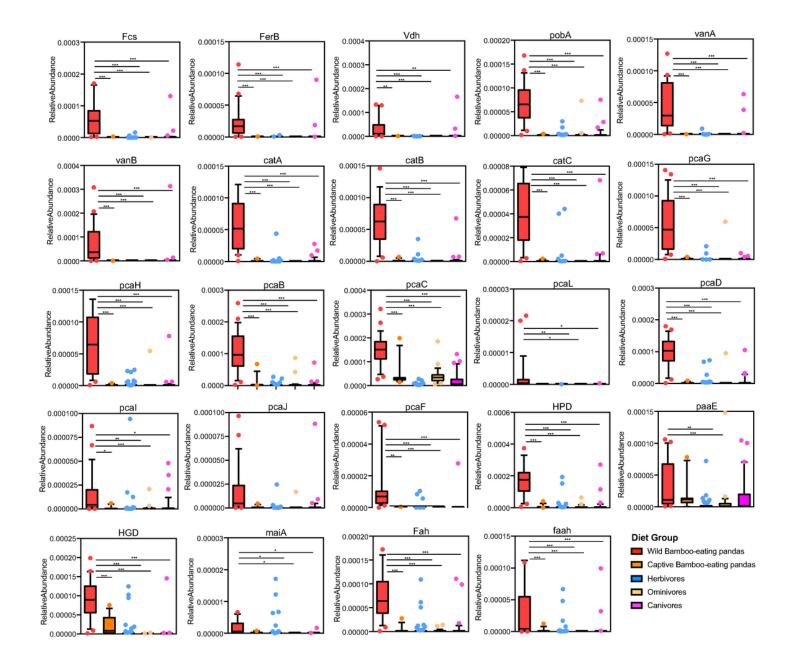
Bin43	Sphingobacterium sp. PM2-P1-29	NA	NA	HPD, HGD
Bin44	Achromobacter	catA, catB ,catC	pcaG, pcaH, pcaB ,pcaC, pcaL, pcaD, pcal, pcaJ	HPD, HGD, maiA, Fah, faah
Bin45	Alcaligenaceae	catA, catB ,catC	NA	NA
Bin46	Clostridiaceae	NA	NA	NA
Bin47	Helicobacteraceae	NA	NA	NA
Bin48	Clostridium	NA	NA	NA
Bin49	Microbacteriaceae	NA	NA	NA
Bin50	Cellvibrionaceae	NA	NA	NA
Bin51	Oxalobacteraceae	catA, catB ,catC	pcaG, pcaH, pcaB, pcaC, pcaL, pcaD, pcal, pcaJ	NA
Bin52	<i>Pseudomonas fluorescens Pf0-1</i>	catA, catB, catC	pcaG, pcaH, pcaB, pcaC, pcaL, pcaD, pcal, pcaJ	HPD, HGD, maiA, Fah, faah,

Table 2. Partial aromatic compounds identified in the control alkali lignin medium (0 day) and the alkali lignin medium degraded by *Pseudomonas*-associated strain for 3 and 7 days. The related intermediate-terminal metabolites in the metabolic pathway are shown in Figure 1.

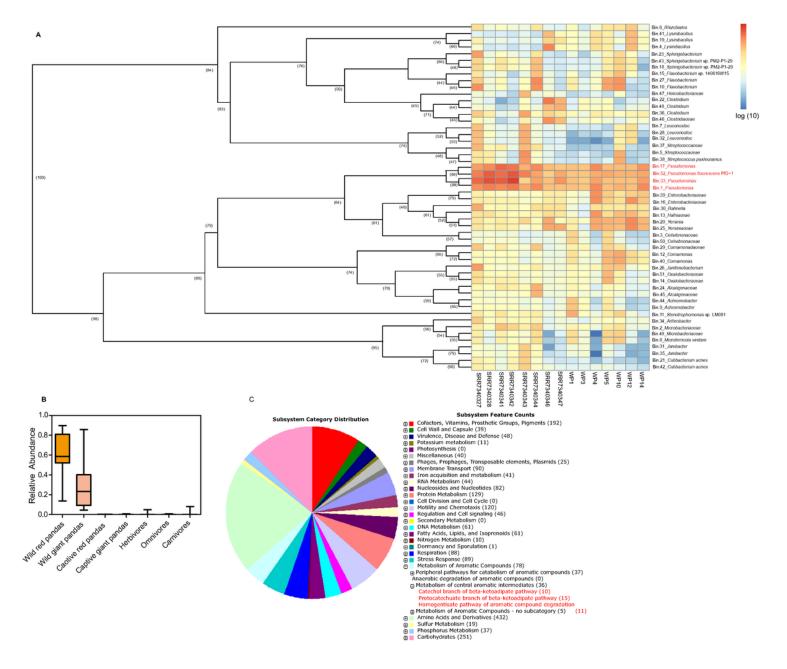
Name	Chemical formula	Control	3d	7d
Catechol	OH 0H 00090	+	+	+
Ferulate	HO L COL494	+	+	+
4-Coumarate		+	+	+
Protocatechuate	он соо230	+	+	+
cis,cis-Muconate	HO OH C02480	-	+	+
Hydroxybenzaldehyde	HO C00633	-	+	+
4-Hydroxybenzoate	HO HO CO0156	+	+	+
Vanillate	OH 0 CH3 0 OH 006672	-	+	+
Vanillin	0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0	+	+	+
Phenylpyruvate		-	+	+
2-Hydroxyphenylacetate	ОН _О 0Но 0Но 0Но 0Но 0Но	-	+	+
Homogentisate	HO COD511	-	+	+
Fumarylacetoacetate		-	+	+
Fumaric acid		-	+	+



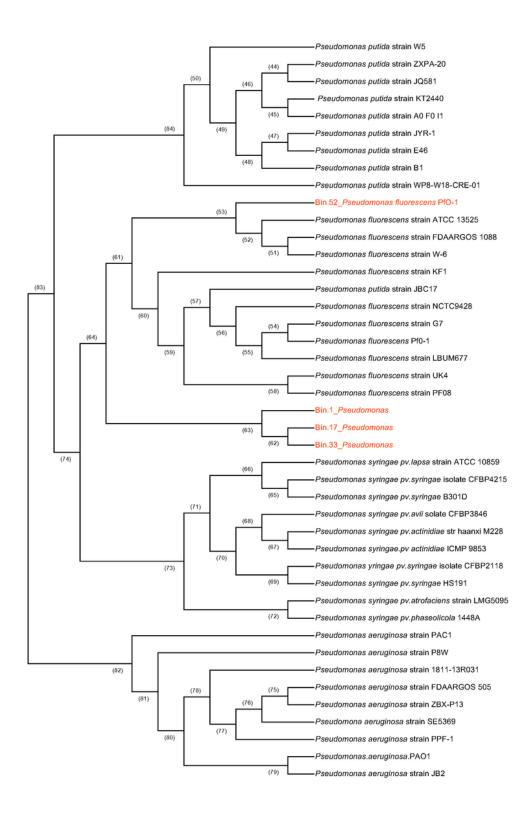
Metabolic pathways involved in lignin derivative degradation enriched in the gut microbiome of wild giant pandas. (A) Beta-ketoadipate pathway, red arrows represent the catechol branch of beta-ketoadipate pathway, green arrows represent the protocatechuate branch of beta-ketoadipate pathway, and black arrows represent parts of the metabolic pathway shared by catechol and protocatechuate branch; (B) Homogentisate pathway for degradation of aromatic compounds; (C) The metabolic pathway for catabolism of lignin monomers p-coumarate and ferulate into protocatechuate, which can be further degraded into Acetyl-CoA and Succinyl-CoA through protocatechuate branch of beta-ketoadipate pathway; (D) Average abundance of genes coding for enzymes implicated in the Beta-ketoadipate pathway, homogentisate pathway, p-coumarate and ferulate degradation pathway in the gut of bambooeating pandas, terrestrial mammals and wood-feeding insects presented as a heat map (log₁₀ abundance).



Relative abundance of genes encoding enzymes for Beta-ketoadipate pathway, homogentisate pathway, *p*-coumarate, and ferulate degradation pathway in the gut of bamboo-eating pandas, terrestrial mammals, and wood-feeding insects presented as box plots. (p values were calculated by the Kruskal-Wallis test followed by Dunn's multiple comparisons post-hoc test: p* < 0.05, p**< 0.01, and p*** < 0.001).



(A) Phylogenetic analysis of the high-quality strain draft genomes (bins) from metagenome sequencing and the relative abundance of these strains in the gut of wild giant pandas. The left panel shows the Neighbor-joining tree using high-quality strain draft genomes (bins) obtained from metagenome sequencing. The numbers in parentheses on the phylogenetic tree represent the bootstrap value of the nodes. The right panel represents the average abundance of each strain draft genomes (bins) in the gut of wild giant pandas. (B) Functional annotation of the genome of Pseudomonas-associated Bin (Bin 1, Bin 17, Bin 33, and Bin 52) in the gut of wild giant panda. Annotation results of the four genomes were similar, thus one Bin (Bin 1) was selected to present the functional annotation results.



Phylogenetic analysis of Pseudomonas-associated genome (Bin) in the gut of wild giant panda and the published genomes of *Pseudomonas fluorescens, Pseudomonas syringae pv. phaseolicola, Pseudomonas putida* and *Pseudomonas aeruginosa.* Genomes of 10 different strains for each *Pseudomonas* species were included in the phylogenetic tree (Neighbor-joining) analysis. The numbers in parentheses on the phylogenetic tree represent the bootstrap value of the nodes.

Phylogenetic analysis and composition of bacterial species with laccase-like multicopper oxidase gene derived from the feces of wild giant pandas. (A) Neighbor-joining phylogenetic tree constructed using laccase-like multicopper oxidase gene sequences from the feces of wild giant pandas and the most similar alignment sequence obtained from Genebank. One laccase-like multicopper oxidase gene sequence was randomly selected from each identified bacterial species. The numbers on the phylogenetic tree represent the bootstrap value of the node. (B) Pie plots showing the composition of bacterial species of laccase-like multicopper oxidase gene in the feces of wild giant pandas.

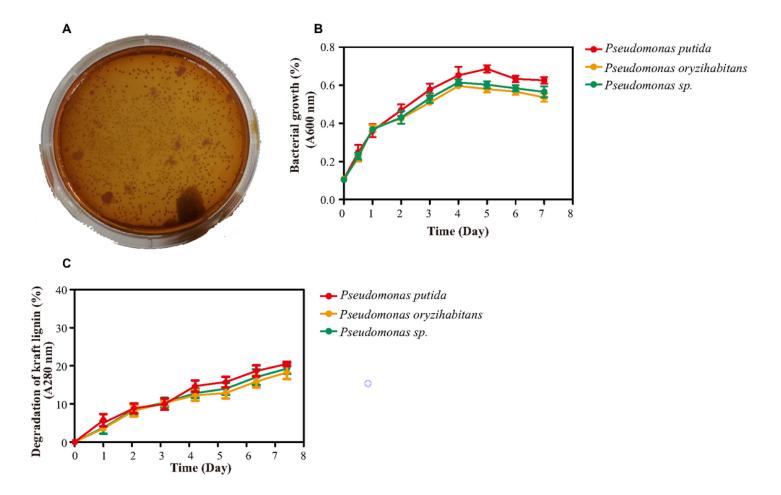
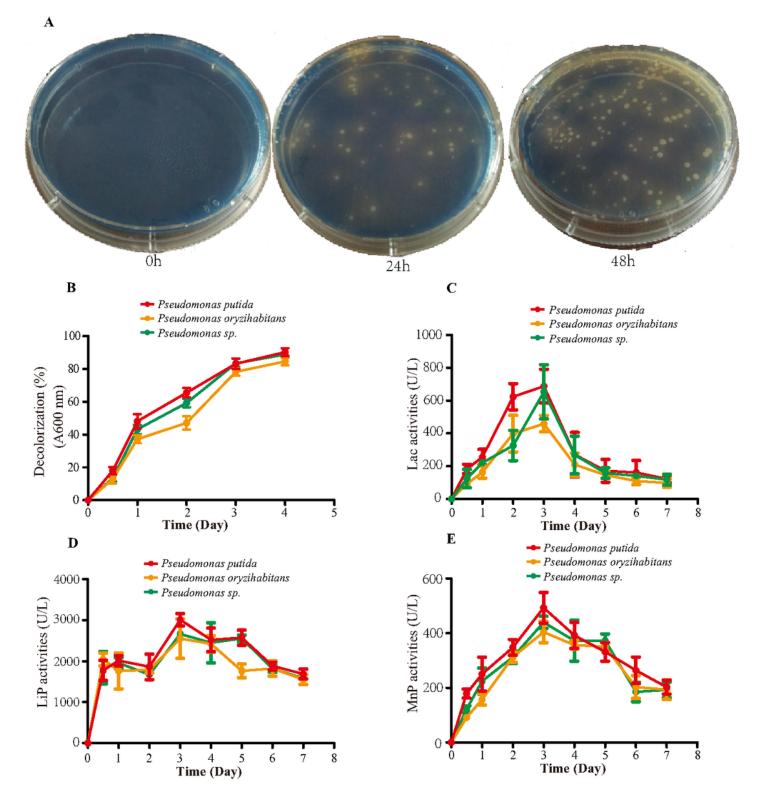


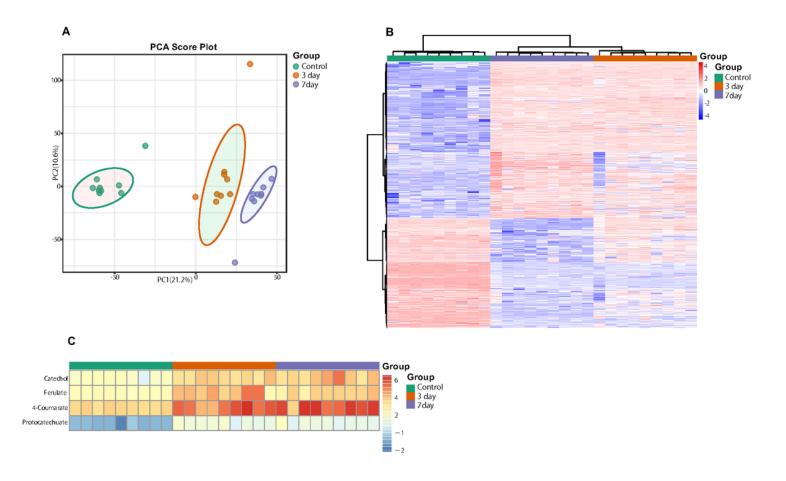
Figure 6

Experimental results of lignin degradation by *Pseudomonas*-associated strain. (A) Growth of *Pseudomonas*-associated strain on solid medium with lignin as the sole carbon source; (B) The growth curve of *Pseudomonas*-associated strain on liquid medium with lignin as the sole carbon source; (C) Lignin-degradation curve.

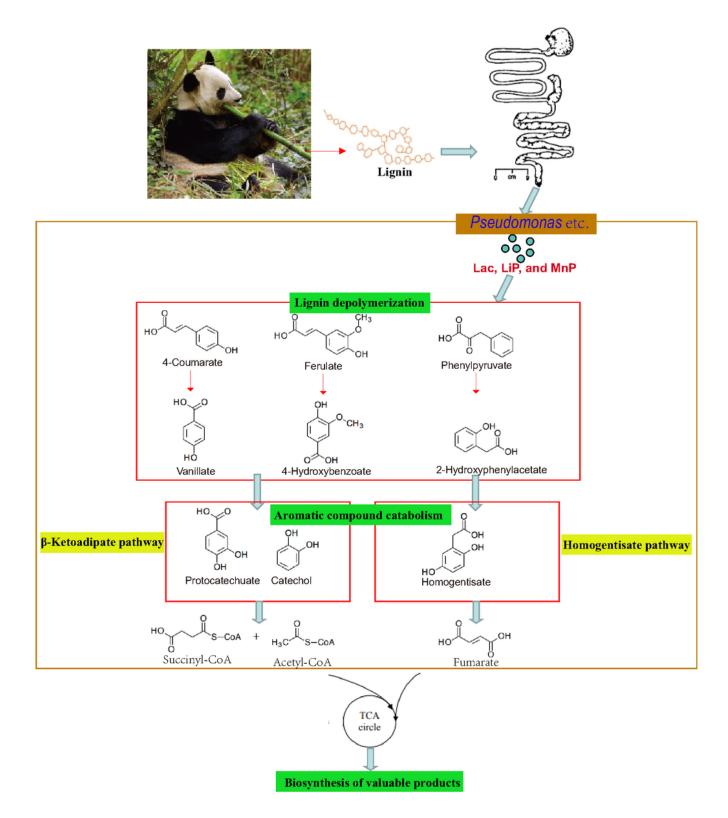




Analysis of extracellular enzyme activity of *Pseudomonas*-associated strain. (A) Decolorization efficiency of *Pseudomonas*-associated strain on aniline blue solid medium; (B) Decolorization rate curve of aniline blue b; Activities of the Lac (C), LiP (D), and MnP (E) during the incubation of *Pseudomonas*-associated strain.



The LC-MS analysis for lignin liquid culture solution. (A) Scores plots of principal component analyses of identified products in lignin culture solution at 0, 3, and 7 days; (B) Heatmap of differential contents in lignin culture solution at 0, 3, and 7 days; (C) Heatmap of catechol, ferulate, 4-coumarate, and protocatechuate in lignin culture solution at 0, 3, and 7 days.



The putative model for lignin degradation by gut microbiotas of giant pandas based on the findings of this study. *Pseudomonas* genus bacteria play a central role in the model.

Supplementary Files

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• SupplementalMaterial.pdf