

# Integrated amplicon sequencing and transcriptomic sequencing technology reveals changes in the bacterial community and gene expression in the rhizosphere soil of *Asparagus cochinchinensis*

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## Abstract

*Asparagus cochinchinensis* has been recognized as an edible and medicinal herb in China, which has resulted in high market demand. Therefore, optimizing its cultivation practice is essential to maintain and enhance production levels. In this study, we simultaneously employed amplicon sequencing and transcriptomic sequencing technology to analyze the rhizosphere bacterial community structure and gene expression level in *A. cochinchinensis* collected from the main production areas. Results showed that a total of 50 phyla, 138 classes, 343 orders, 490 families, and 875 genera were identified, of which Proteobacteria and *RB41* were the most abundant at the phylum and genus level, respectively. Transcriptomics results showed that the gene expression was related to biotic stress resistance, plant growth and development, and polysaccharide metabolism, which was supported by qPCR results. These findings offer valuable insights into the utilization safety and quality improvement of *A. cochinchinensis*.

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## Introduction

*Asparagus cochinchinensis* is a well-known medicinal plant within the genus *Asparagus* in the family Liliaceae. Species in the *Asparagus* genus are extensively utilized in both pharmacology and food industries<sup>[1,2]</sup>. Recently, *A. cochinchinensis* has been recognized as an edible and medicinal herb in China, resulting in an upward trend in market demand. According to the recommendation of the Chinese Pharmacopoeia (2020), *Asparagi Radix* is derived from the dried root and tuber of *Asparagus cochinchinensis* (Lour.) Merr.<sup>[3]</sup>. Modern pharmacology studies have reported that *A. cochinchinensis* shows cough-suppressing and expectorant properties, which aids in managing airway inflammation and other related conditions<sup>[4]</sup>. However, previous studies mainly focused on its botanical, traditional, phytochemical, and pharmacological aspects<sup>[5,6]</sup>. Currently, *A. cochinchinensis* is mainly planted in Sichuan Province (Neijiang City) and Guangxi Zhuang Autonomous Region. There are few studies discussing the topic regarding its sustainable cultivation practices. Cultivation as the initial stage of the whole medicinal plant production chain, remarkably influences its quality and warrants further investigation. In recent years, it has been reported that the rhizosphere microbiome greatly contributes to the growth and development of medicinal plants<sup>[7]</sup>. Qu et al. studied the root secretion metabolites and soil microbiome of five medicinal plants. They found that the bacterial and fungal profiles of these five medicinal plant rhizosphere soil samples were biologically distinct<sup>[8]</sup>. Among these metabolites, ten root secretion metabolites significantly influenced the distribution of bacteria and fungi<sup>[9]</sup>. Liu et al. studied the

rhizosphere microbiome of *Glehnia littoralis* and observed that the rhizosphere microbial community formed a complex symbiotic network that positively impacted the growth and development of *G. littoralis*<sup>[10]</sup>. Wang et al. explored the effect of cultivation process on *Atractylodes lancea*, and they indicated that the autotoxic allelopathic substances in the rhizosphere of *A. lancea* altered rhizosphere soil microbiome, leading to continuous cropping disorders<sup>[11]</sup>. These studies collectively suggest that the soil bacterial community significantly affect the growth and development of medicinal plants. Additionally, the rhizosphere microbiome might also influence the natural product synthesis of medicinal plants. Zhu et al. analyzed the relationship between the rhizosphere bacterial community and the accumulation of polysaccharides of *Dendrobium officinale*, and they indicated that the relative abundance of *Pandoraea*, which was identified from the rhizosphere bacterial community, was associated with polysaccharide level. *Pandoraea* might promote the production of *D. officinale* polysaccharide production<sup>[12]</sup>. The function of rhizosphere species was also reported. Ng et al. studied the promotive effect of *Bacillus subtilis* and *Pseudomonas fluorescens* on the polysaccharide level of *Pseudostellaria heterophylla*. Results showed that the polysaccharide level increased by 38% and 253% after incubation with these strains<sup>[13]</sup>. Based on the above studies, rhizosphere microbiome not only affect the growth and development of medicinal plant, but might also be related to the synthesis of active compounds. Therefore, it is crucial to reveal the rhizosphere microbiome structure and discuss the potential relationship between the rhizosphere microbiome and *A. cochinchinensis*.

In this study, we simultaneously applied amplicon sequencing and transcriptomic sequencing technology to analyze the rhizosphere microbiome structure and gene expression in *A. cochinchinensis*. This work advances our understanding of the rhizosphere microbiome resource and provides references for the development of functional rhizosphere strains for *A. cochinchinensis*.

Materials and methods

Sample collection

The samples of *A. cochinchinensis* and their associated rhizosphere soil were collected from Neijiang City, Sichuan Province, China, with detailed collection information provided in Table 1. It has been reported that Neijiang City has an average annual temperature ranging from 15–28 °C and the annual rainfall is about 1,000 mm. The plant samples were used for transcriptomic analysis. After collection, the tender stems and tuberous roots were excised, and the surface of the samples were washed with RNase-free water. The samples were then immediately dried using absorbent paper and cut into 50–100 mg pieces. These pieces were rapidly frozen in liquid nitrogen and placed into pre-cooled threaded Eppendorf tubes, where they were stored at –80 °C. All sampling tools were sterilized and disinfected before using. The rhizosphere soil samples were collected from three towns including Yangjia Town (YRB), Shuangcai Town (CRB), and Guobei Town (SRB), Neijiang City. The soil was removed from the roots, followed by the collection of soil tightly adhering to the roots using a sterile brush. The rhizosphere soil (within 0.2 cm of the roots) was filtered through a 2 mm sieve, and samples from each area were mixed. Next, plant debris was removed and carefully placed into sterilized bags. All soil samples were rapidly frozen in liquid nitrogen and stored at –80 °C until further analysis.

High throughput sequencing of the rhizosphere bacterial community

Approximately 1.0 g of frozen soil sample was transferred into a sterilized 15 ml centrifuge tube containing 1.5 g of grinding beads. Bacterial DNA extraction was conducted using the EZNA® Soil DNA Kit (D5625, Omega Bio-Tek, Inc.) according to the manufacturer's protocol. The extracted DNA was quantified using a Nanodrop spectrophotometer (ND ONE, Genes Ltd.). Primers were designed based on conserved regions within the 16S rRNA gene and incorporating sample-specific barcode sequences. The bacterial 16S rRNA gene was amplified by PCR using the primers 16S V3V4-F 5'-CCTACGG

GGNGGCWGCAG-3' and 16S V3V4-R 5'-GACTACHVGGGTATCTAA TCC-3'. The PCR products were purified using magnetic bead-based purification methods. Fluorescence-based quantification was performed on the recovered PCR products. Based on the quantification results, samples were mixed in appropriate proportions based on the sequencing requirements. Library preparation was conducted using the VAHTS Universal DNA Library Prep Kit for Illumina V3 and VAHTS DNA Adapters set3-set6 for Illumina from Novogene. The libraries were subjected to quality control and sequenced on the NovaSeq 6000 platform. Barcodes were used to identify the mixed samples and obtain individual sample sequence data. Low quality reads were filtered using Trimmomatic software (Version 0.39), with a window size of 20 bp and an average quality value greater than 20 within the window. Primer sequences were identified and removed using cutadapt software (Version 3.5), resulting in clean sequences. Each sequence generated by DADA2 was termed a feature or amplicon sequence variant (ASV). Low abundance ASVs were filtered using QIIME2 (Version 2022.3), and each ASV was annotated at various taxonomies. This process enabled the determination of bacterial community composition within the samples. R software (Version 3.1.1) with relevant packages was used to evaluate the alpha diversity and beta diversity based on the Bray-Curtis distance. The bacterial composition at the phylum, class, order, family, and genus levels were displayed using the barplot package, which was conducted to reveal differences between samples. Additionally, based on the Spearman correlation matrix, co-occurrence network topology analysis was conducted using the R Igraph package.

RNA extraction, cDNA library construction, sequencing, de novo assembly, and annotation

Total RNA was extracted from the tuberous roots and stems of *A. cochinchinensis* for cDNA library construction, with each treatment performed in triplicate. RNA integrity and quality were evaluated using the NanoDrop One spectrophotometer (NanoDrop Technologies, Wilmington, DE, USA) and a Qubit 3.0 Fluorometer (Life Technologies, Carlsbad, CA, USA). mRNA was enriched using oligo (dT) magnetic beads, and double-stranded cDNA was synthesized. The library preparation was completed using the MGIEasy RNA Library Prep Kit for BGI. The fragment size and concentration of the library were assessed using an Agilent 2100 Bioanalyzer. Sequencing was performed on the BGI high-throughput sequencing platform DNBSEQ-T7. After filtering out raw data connections, poly-N sequences, and low-quality reads, clean reads were obtained. To obtain valid data for subsequent analyses, the raw data was filtered

Table 1. Information for the *A. cochinchinensis* cultivation soil samples and different tissues used in this study.

Sample number	Group	Sampling date	Sampling area	SAMN number
YRBR1	YRBR	2024/4/24	China: Neiliang: Yangjia Town 29.525142N 105.163067E	SAMN43440143
YRBR2	YRBR	2024/4/24	China: Neiliang: Yangjia Town 29.525142N 105.163067E	SAMN43440144
YRBR3	YRBR	2024/4/24	China: Neiliang: Yangjia Town 29.525142N 105.163067E	SAMN43440145
YRBS1	YRBS	2024/4/24	China: Neiliang: Yangjia Town 29.525142N 105.163067E	SAMN43440146
YRBS2	YRBS	2024/4/24	China: Neiliang: Yangjia Town 29.525142N 105.163067E	SAMN43440147
YRBS3	YRBS	2024/4/24	China: Neiliang: Yangjia Town29.525142N 105.163067E	SAMN43440148
CRB1	CRB	2024/4/24	China: Neiliang: Shuangcai Town 29.749611N 105.109195E	SAMN43415523
CRB2	CRB	2024/4/24	China: Neiliang: Shuangcai Town 29.749611N 105.109195E	SAMN43415524
CRB3	CRB	2024/4/24	China: Neiliang: Shuangcai Town 29.749611N 105.109195E	SAMN43415525
YRB1	YRB	2024/4/24	China: Nelliang: Yangjia Town 29.766105N 105.353238E	SAMN43415526
YRB2	YRB	2024/4/24	China: Nelliang: Yangjia Town 29.766105N 105.353238E	SAMN43415527
YRB3	YRB	2024/4/24	China: Nelliang: Yangjia Town 29.766105N 105.353238E	SAMN43415528
SRB1	SRB	2024/4/24	China: Neiliang: Guobei Town 29.525142N 105.163067E	SAMN43415529
SRB2	SRB	2024/4/24	China: Neiliang: Guobei Town 29.525142N 105.163067E	SAMN43415530
SRB3	SRB	2024/4/24	China: Neiliang: Guobei Town 29.525142N 105.163067E	SAMN43415531

with fastp (Version 0.21.0) and fastqc (Version 0.11.9)<sup>[14,15]</sup>. *De novo* assembly was conducted using Trinity (Version 2.11.0), and resulting transcripts were clustered using cd-hit (Version 4.8.1) to obtain universal genes (unigenes)<sup>[16,17]</sup>. The unigenes were annotated using two methods: 1) sequence similarity search using diamond blastp (Version 2.0.6.144; parameters: e-value 1e-5) for comparison; 2) motif similarity search, structural domain prediction using hmmscan (Version 3.3.2; parameters: e-value 0.01) to obtain conserved sequences, motifs, and structural domains of proteins<sup>[18,19]</sup>. Comprehensive functional annotation of the unigenes were performed using seven databases: Nr, Pfam, Uniprot, KEGG, GO, KOG/COG, and PATHWAY<sup>[20–24]</sup>.

### Differential expression unigene analysis

Differential expression analysis was conducted on the counts of expressed unigenes obtained from quantitative expression measurements across various samples. DESeq2 software was used for differential expressed gene analysis (Version 1.26.0)<sup>[25]</sup>. The significance threshold was initially set to  $\text{padj} < 0.05$  and  $|\log_2\text{FoldChange}| > 1$ . The  $\text{padj}$  value was used for analyzing significant differences. The objective was to determine whether the functions of the differentially expressed unigenes were concentrated in specific functional categories. Common functional classifications included GO and KEGG Pathway. The enrichment analysis was performed for hypergeometric tests to identify GO terms and KEGG pathways that were significantly enriched among the differentially expressed unigenes relative to all annotated unigenes. The clusterProfiler software (Version 3.14.3) was utilized for differential expression unigene enrichment analysis<sup>[26]</sup>.

### Quantitative real-time-PCR

The RNA products of the tuberous roots and stems of *A. cochinchinensis* were extracted using the FastPure Plant Total RNA Isolation Kit (Polysaccharides & Polyphenolics-rich, RC401-01, Vazyme, Nanjing, China). Reverse transcription was performed using the RevertAid™ First Strand cDNA Synthesis Kit, with DNase I (RT-01022, Thermo Scientific, Waltham, USA). The TB Green® Premix Ex Taq™ II (Tli RNaseH Plus, RR82LR, TaKaRa, Beijing, China) was used to perform qRT-PCR. The mRNA expression of *Unigene11685*, *Unigene10718*, *Unigene11242*, and *Unigene10145* were evaluated by the  $2^{-\Delta\Delta C_t}$  method and normalized with *EF1 $\alpha$* . The primer sequences were designed in Sangon Biotech (Sangon, Shanghai, China, Table 2).

### Statistical analysis

The data in this study were subjected to statistical testing with SPSS software (Version 22.0) and RStudio (Version 2024.04.2+764). Significant differences were analyzed by one-way ANOVA. A  $p$ -value less than 0.05 ( $p < 0.05$ ) in these analyses indicated a significant difference.

**Table 2.** Designed primer information in this study.

Primers	DNA sequence (5' to 3')
U10703-F	ACCTACCGCCATCACTCAAC
U10703-R	CCCGAACCAATCCCTGAAATACC
U11242-F	CTTGATATAGACGATCCTGACACTTGG
U11242-R	TGGGAAGCGATTATTGAAACCTCTG
U11685-F	ATGGCGGTCTCGGTCATACTAC
U11685-R	GGCATTGGATTGGATTGGATCTG
U10145-F	GCCAAGTGAGTTGCCAGGTTCT
U10145-R	GTTTATGTATCAGGTCGTCTTGCTTTG
U10718 -F	GGTAACTTGGCACTTAAAGCGATATAG
U10718 -R	TACCTTCTCGTGTACCATTAAACAACCTC
EF1 $\alpha$ -F	CTGGCCAGGGTGGTTCATGAT
EF1 $\alpha$ -R	TAAGTCTGTTGAGATGCACC

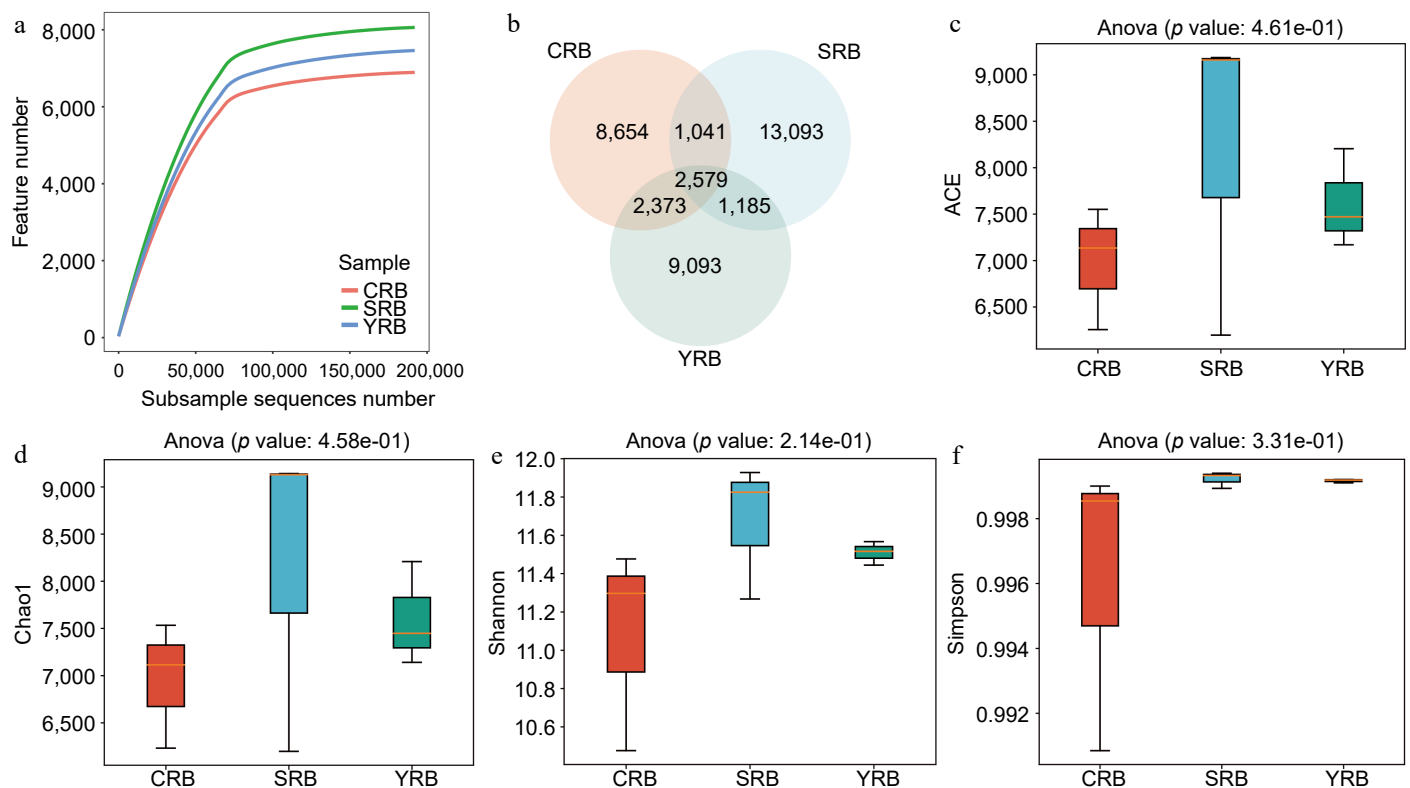
## Results

### Rhizosphere bacterial community diversity in *A. cochinchinensis* rhizosphere soil samples collected from various areas

In this study, high-throughput sequencing technology was employed to analyze the rhizosphere bacterial composition of samples collected from three areas. A total of 3,372,071 raw reads were observed from the nine samples. After quality control, 3,181,644 clean reads were obtained. The raw sequencing data was uploaded to the National Center for Biotechnology Information Sequence Read Archive database with accession numbers CRB1 (SAMN43415523), CRB2 (SAMN43415524), CRB3 (SAMN43415525), YRB1 (SAMN43415526), YRB2 (SAMN43415527), YRB3 (SAMN43415528), SRB1 (SAMN43415529), SRB2 (SAMN43415530), and SRB3 (SAMN43415531). Rarefaction curve analysis results indicated that the sequencing depth in each sample was sufficient to reflect the bacterial community (Fig. 1a). A venn diagram shows the number of shared or unique ASVs across different groups, revealing 2579 shared ASVs in various groups. The numbers of unique ASVs in the CRB, SRB, and YRB groups were 8654, 13093, and 9093, respectively. Notably, the SRB group exhibited the highest number of unique ASVs, while the CRB group had the lowest number (Fig. 1b). The alpha diversity analysis was performed to evaluate the richness and diversity of the bacterial communities. The indices of alpha diversity are listed in Supplementary Table S1. The average ACE indices for CRB, SRB, and YRB groups were 6,982, 8,181, and 7,615, respectively (Fig. 1c). The average Chao1 indices were 6,960, 8,156, and 7,599 (Fig. 1d). The average Shannon indices were 11.08, 11.67, and 11.51, while the average Simpson indices were 0.996, 0.999, and 0.999 (Fig. 1e, f). No significant differences were observed in the Shannon, Simpson, Chao1, and ACE indices among various groups. The SRB group exhibited the highest Shannon, Chao1, and ACE indices followed by the YRB group, while the CRB group consistently showed the lowest values across these indices.

### Bacterial community composition in *A. cochinchinensis* rhizosphere soil samples collected from various areas

Based on the high-throughput sequencing data, a total of 50 phyla, 138 classes, 343 orders, 490 families, and 875 genera were identified in this study. At the phylum level, the dominant phyla included Proteobacteria, Actinobacteria, Planctomycetota, Actinobacteriota, Chloroflexi, Gemmatimonadota, Bacteroidota, Verrucomicrobiota, Myxococcota, and Methylomirabilota with relative abundances greater than 1% (Fig. 2a). Among these phyla, Proteobacteria, Actinobacteria, and Planctomycetota were dominant, with relative abundances ranging from 17.8%–23.1%, 14.8%–24.4%, and 8.2%–12.6%, respectively. The most abundant bacterial classes in this study included Gammaproteobacteria, Vicinamibacteria, Alphaproteobacteria, Phycisphaerae, Blastocatellia, Bacteroidia, Planctomycetes, Gemmatimonadetes, Thermoleophilia, and Verrucomicrobiae. Gammaproteobacteria, Vicinamibacteria, and Alphaproteobacteria were dominant, with relative abundances of 10.2%–12.8%, 6.7%–17.0%, and 6.3%–12.6%, respectively (Fig. 2b). At the order level, Vicinamibacteriales had the highest relative abundance, while the CRB group had lower levels of Vicinamibacteriales compared to the other groups, the SRB group had a lower relative abundance of Pyrinomonadales compared to the other groups (Fig. 2c). At the family level, Gemmatimonadaceae had the highest relative abundance, with the highest relative abundance in the CRB group, whereas Nitrosomonadaceae had the highest average relative abundance in the YRB group. Other abundant families



**Fig. 1** The diversity and richness of rhizosphere bacterial communities of *A. cochinchinensis* rhizosphere samples collected from various areas. (a) Rarefaction curve reveals the sequencing depth in each sample, the horizontal coordinate reflects the number of sequences randomly selected, and the vertical coordinate reflects the number of features obtained based on the number of sequences. Each curve represents a sample/group and is marked with a different color. (b) Venn diagram illustrates the numbers of ASVs in the CRB, SRB, and YRB groups. (c) ACE indices in the CRB, SRB, and YRB groups based on QIIME2 software, the horizontal coordinate reflects the group name, and the vertical coordinate reflects the ACE index value. (d) Chao1 indices in the CRB, SRB, and YRB groups based on QIIME2 software, the horizontal coordinate reflects the group name, and the vertical coordinate reflects the Chao1 index value. (e) Shannon indices in the CRB, SRB, and YRB groups based on QIIME2 software, the horizontal coordinate reflects the group name, and the vertical coordinate reflects the Shannon index value. (f) Simpson indices in the CRB, SRB, and YRB groups based on QIIME2 software, the horizontal coordinate reflects the group name, and the vertical coordinate reflects the Simpson index value.

included Gemmatimonadaceae, Nitrosomonadaceae, Vicinamibacteraceae, Pyrinomonadaceae, WD2101\_soil\_group, and Sphingomonadaceae (Fig. 2d). At the genus level, *RB41* had the highest relative abundance, with the highest average relative abundance in the CRB group. *RB41*, *Vicinamibacteraceae*, and *WD2101\_soil* had relative abundances ranging from 1.7%–6.7%, 2.6%–4.8%, and 1.8%–6.0%, respectively (Fig. 2e).

### Differences of bacterial community of *A. cochinchinensis* rhizosphere soil samples collected from various areas

PCA, PCoA, and NMDS analyses were used to assess the similarity and differences between various groups. Using the binary Jaccard algorithm, the first component in the PCA plot explained 23.25% of the variance, with the CRB, SRB, and YRB groups clustering separately (Fig. 3a). Similarly, based on the PCoA analysis result, the first component explained 21.78% of the variance, and the three groups also formed distinct clusters (Fig. 3b). Based on the NMDS analysis result, CRB, SRB, and YRB groups were clustered separately, and the bacterial community composition was similar between the YRB and CRB groups (Fig. 3c). According to the Bray-Curtis distance algorithm and average clustering for bacterial community analysis at the genus level, the hierarchical clustering tree further revealed that the YRB and CRB groups were clustered, while SRB group formed a separate cluster (Fig. 3d). The histogram indicated that the bacterial communities at the genus level were different, the dominant

bacterial composition was similar with differences in relative abundance. The heatmap also further supported this result (Fig. 3e).

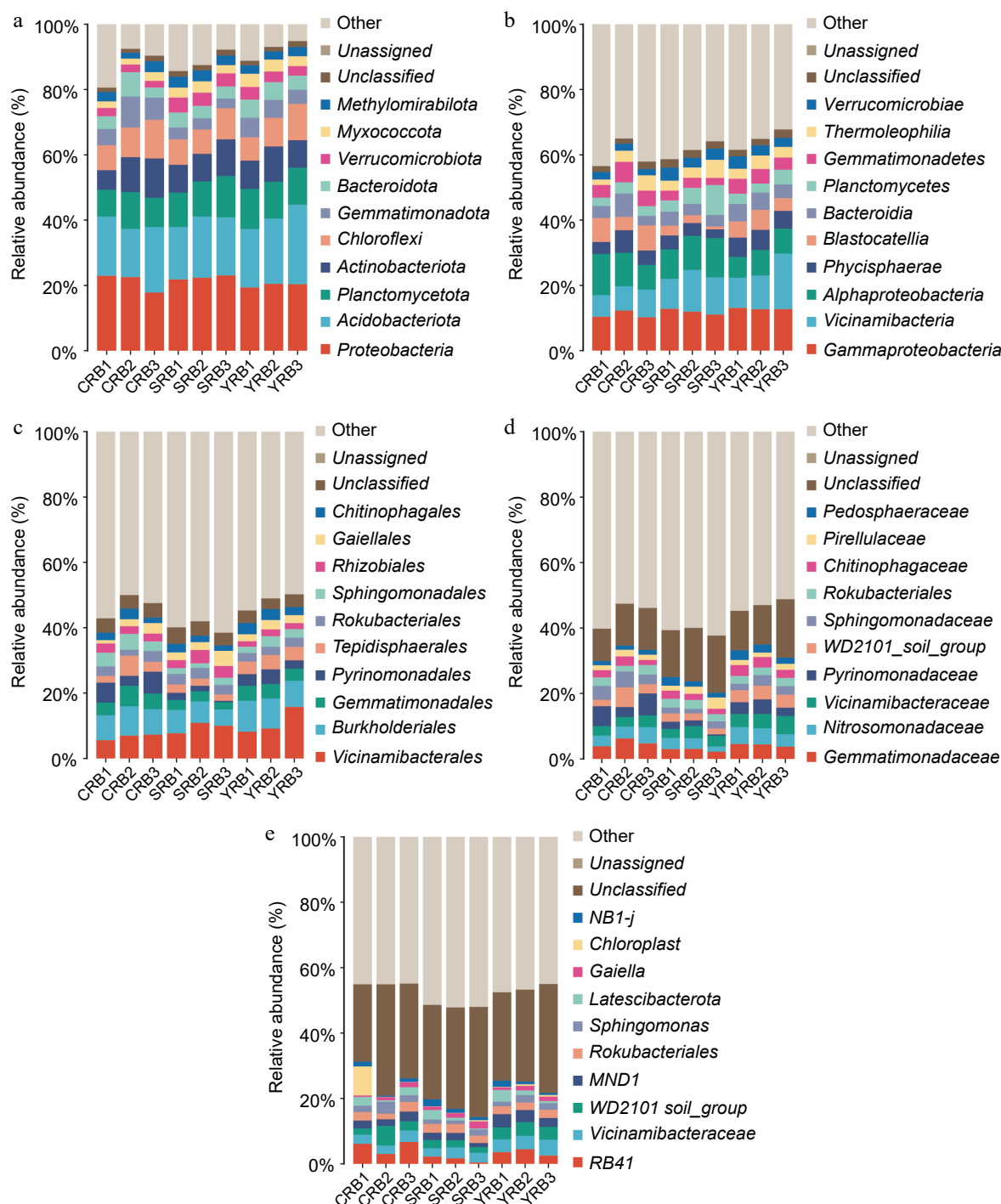
### Correlation of the bacterial community of *A. cochinchinensis* rhizosphere soil samples collected from various areas

To explore the interactions between bacterial taxa in different groups, correlation network graphs were constructed to analyze the interaction between the top 50 abundant taxa at the genus level across all samples. These networks consisted of an average of 50 nodes and 472 connecting lines. From the results of the analysis, *RB41* and *Niastella* in the CRB group were positively correlated to the whole bacterial community, while *Subgroup\_7* and *Rubrobacter* were negatively correlated to the whole bacterial community (Fig. 4a). In the YRB group, *Rubrobacter*, and *Gaiella* were positively correlated with the whole bacterial community, while *TK10* showed negative correlation, and *Candidatus\_Udaeobacter* and *Gaiella* were positively correlated with the whole bacterial community in the SRB group (Fig. 4b, c). *JG30-KF-AS9* and *Candidatus\_Kaiserbacteria* showed negative correlation.

### RNA-seq data revealed the gene expression pattern of *A. cochinchinensis*

RNA-seq analysis of the tuberous roots and stems of *A. cochinchinensis* was performed. Based on the sequencing result, an average of 6.27 GB of clean data was obtained from 68.91 GB of raw reads from all samples. The GC content was 46%, and the Q30 value was 97%.

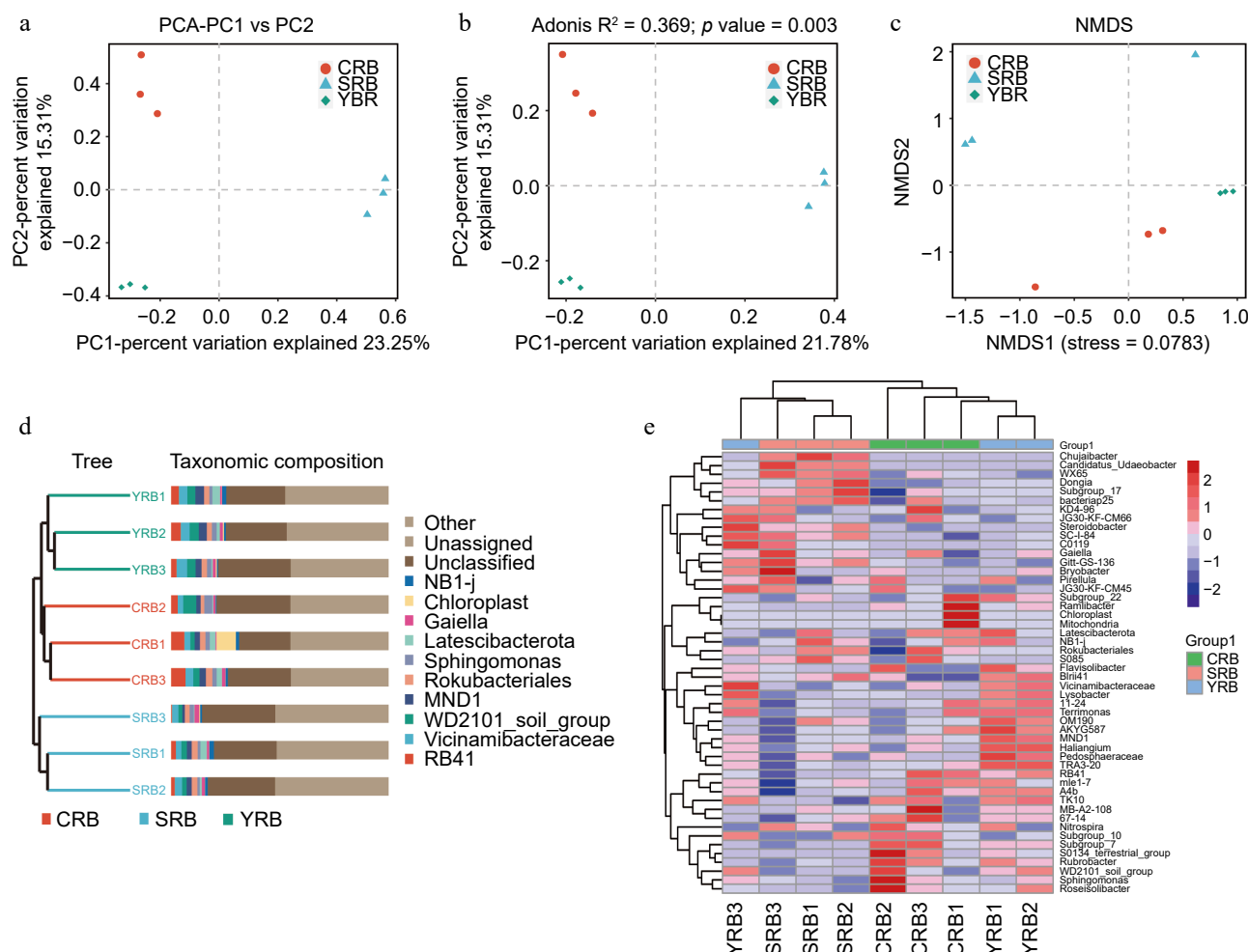




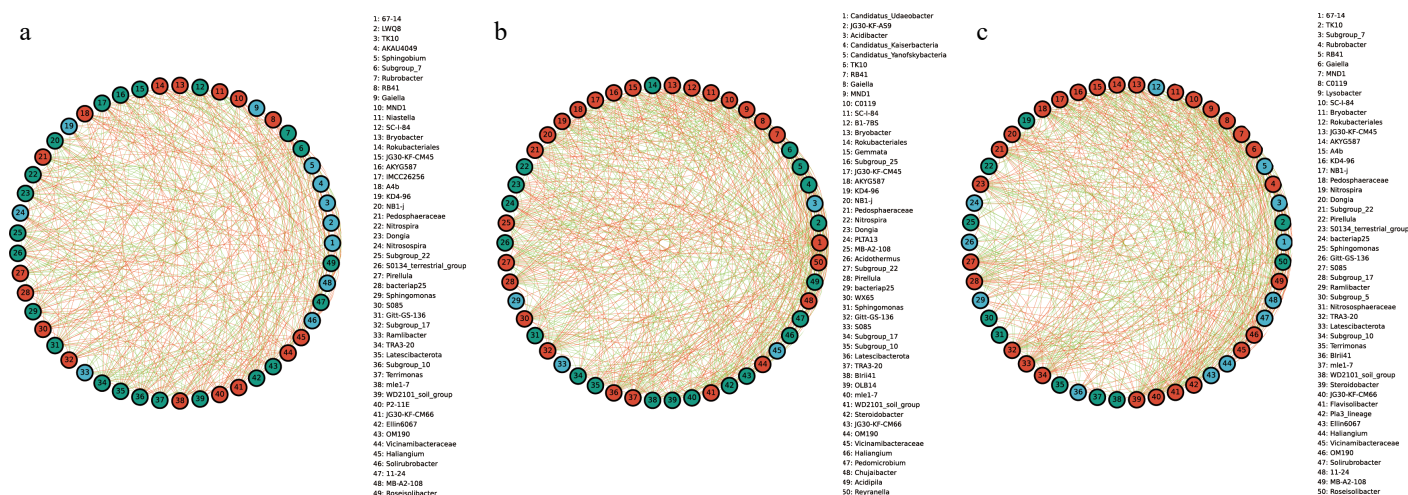
**Fig. 2** Bacterial community composition in the *A. cochinchinensis* rhizosphere samples collected from various areas. (a) The relative abundances of the ten top dominant taxa at the phylum level. (b) The relative abundances of the ten top dominant taxa at the class level. (c) The relative abundances of the ten top dominant taxa at the order level. (d) The relative abundances of the ten top dominant taxa at the family level. (e) The relative abundances of the ten top dominant taxa at the genus level.

A total of 240,208,194 raw reads were generated (Supplementary Table S2). After removing low-quality sequences, 240,208,064 clean reads remained. The clean reads were assembled into 254,407 transcripts. The data was uploaded to the National Center for Biotechnology Information Sequence Read Archive database with accession numbers YRBS3 (SAMN43440148), YRBS2 (SAMN43440147), YRBS1 (SAMN43440146), YRBR3 (SAMN43440145), YRBR2 (SAMN43440144), and YRBR1 (SAMN43440143). These transcripts were further clustered using cd-hit (Version 4.8.1) [13], resulting in 127,262 unigene

sequences. The sequence length ranged from 201 to 16,535 bp, with an average length of 910.2 bp and an N50 length of 1,661 bp. The GO database was used to categorize the functions of predicted unigenes into three categories: cellular component (CC), molecular function (MF), and biological process (BP) [17]. Of the 23,299 unigenes corresponding to known protein families, the top 20 annotations in each GO slim secondary classification were selected for analysis. The function of most unigenes were predicted for cellular components, followed by molecular functions and biological processes. Within



**Fig. 3** Beta diversity of bacterial community in the *A. cochinchinensis* rhizosphere samples collected from various areas. (a) Principal component analysis of gut bacterial community among different groups. (b) Principal co-ordinates analysis of soil bacterial community among different groups. (c) Non-metric multi-dimensional scaling analysis of soil bacterial community among different groups. (d) Unweighted pair-group method with arithmetic mean combined with species abundance histograms analysis among different groups based on the binary Jaccard distances. (e) Heatmap clustering at the genus level and the top 50 species among different groups.



**Fig. 4** The species correlation network of bacterial communities of *A. cochinchinensis* rhizosphere samples collected from various areas. Spearman rank correlation analyses were performed on the top 50 species in the (a) CRB, (b) SRB, and (c) YRB in terms of genus-level abundance, and correlation networks were constructed by filtering data with correlations greater than 0.6 or less than -0.6 and p-values of less than 0.05. The green lines represent negative correlations, and red lines represent positive correlations. The bold lines represent the closer correlations between two genera.

cellular components, the unigenes were predicted for membrane, nucleolus, and cytoplasm. For biological processes, the unigenes were predicted for phosphorylation, proteolysis, and regulation of DNA-templated transcription. For molecular functions, the unigenes were predicted for ATP binding, metal ion binding, and DNA binding (Fig. 5a). KEGG annotation of unigene sequences categorized the pathways involving the largest number of unique transcripts as global and overview maps, involved in carbohydrate metabolism, amino acid metabolism, and translation, transcription, and signal transduction pathways (Fig. 5b).

### Differentially expressed gene analysis between tuberous roots and stem tissues in *A. cochinchinensis*

For the differential expression analysis of unigenes across various tissues, the threshold of  $\text{padj} < 0.05$  and  $|\log_2\text{FoldChange}| > 1$  was applied. The difference in unigene expression between the stem and tuberous root tissues were observed. Specifically, with a total of 10,807 differentially expressed unigenes between the stem and tuberous roots tissues, with 3,652 unigenes upregulated and 7,155 genes downregulated in tuberous roots compared to stems (Fig. 5c). Functional analysis based on the GO and KEGG databases revealed that the most enriched GO pathways in tuberous roots involved DNA-binding transcription factor activity, cellulose synthase (UDP-forming) activity, and cell wall biogenesis (Fig. 5d). KEGG analysis indicated enrichment in pathways including ribosome, pentose phosphate pathway, fatty acid metabolism, and glycolysis /gluconeogenesis (Fig. 5e). Based on the RNA-seq data, four differentially expressed unigenes related to the growth and biotic stress resistance between the tuberous root and stem samples of *A. cochinchinensis*. Significant differences of all these four genes were observed through qPCR results, which supported transcriptomics data (Fig. 6).

## Discussion

The topic of the rhizosphere microbiome has attracted wide attention, which mainly discusses the interaction between plant roots and the rhizosphere microbial community involving bacteria, fungi, archaea, protists, and viruses. Current studies have demonstrated that the composition and characteristics of the rhizosphere microbiome exhibits specificity across different plants. Li et al. studied the rhizosphere microbial community of *Astragalus membranaceus* and found that the dominant bacterial phyla in the roots of *A. membranaceus* were *Proteobacteria* and *Actinobacteria*<sup>[27]</sup>. Li et al. analyzed the microbial community of *Panax notoginseng*, indicating that *Chloroflexi* and *Verrucomicrobia* were dominant<sup>[28]</sup>. Furthermore, Zuo et al. observed differences in the composition of bacterial phyla when comparing the rhizosphere microbial community of *Dendrobium* plants. These findings underscore the specificity of rhizosphere microbial communities in different plant species<sup>[29]</sup>. Additionally, the function of rhizosphere microbiome exhibits differences under various conditions. Zhang et al. reported a close relationship between ecological environments and bacterial population structures. Their research on ginseng cultivated in farmland and forests revealed notable differences in the gene functions of the rhizosphere communities between ginseng samples grown in these two environments, which played significant roles in promoting the growth, development, and health of ginseng<sup>[30]</sup>. Meanwhile, Jiao et al. indicated that heavy metals inhibited the growth of corn, and affected the diversity of corn rhizosphere soil microbial community, as well as their metabolic pathways, and metabolites<sup>[31]</sup>. Liu et al. conducted precipitation control experiments on four plant species and discovered that microbial functions were primarily influenced by water availability rather than being species-dependent. Changes in water levels caused by precipitation may impact the element

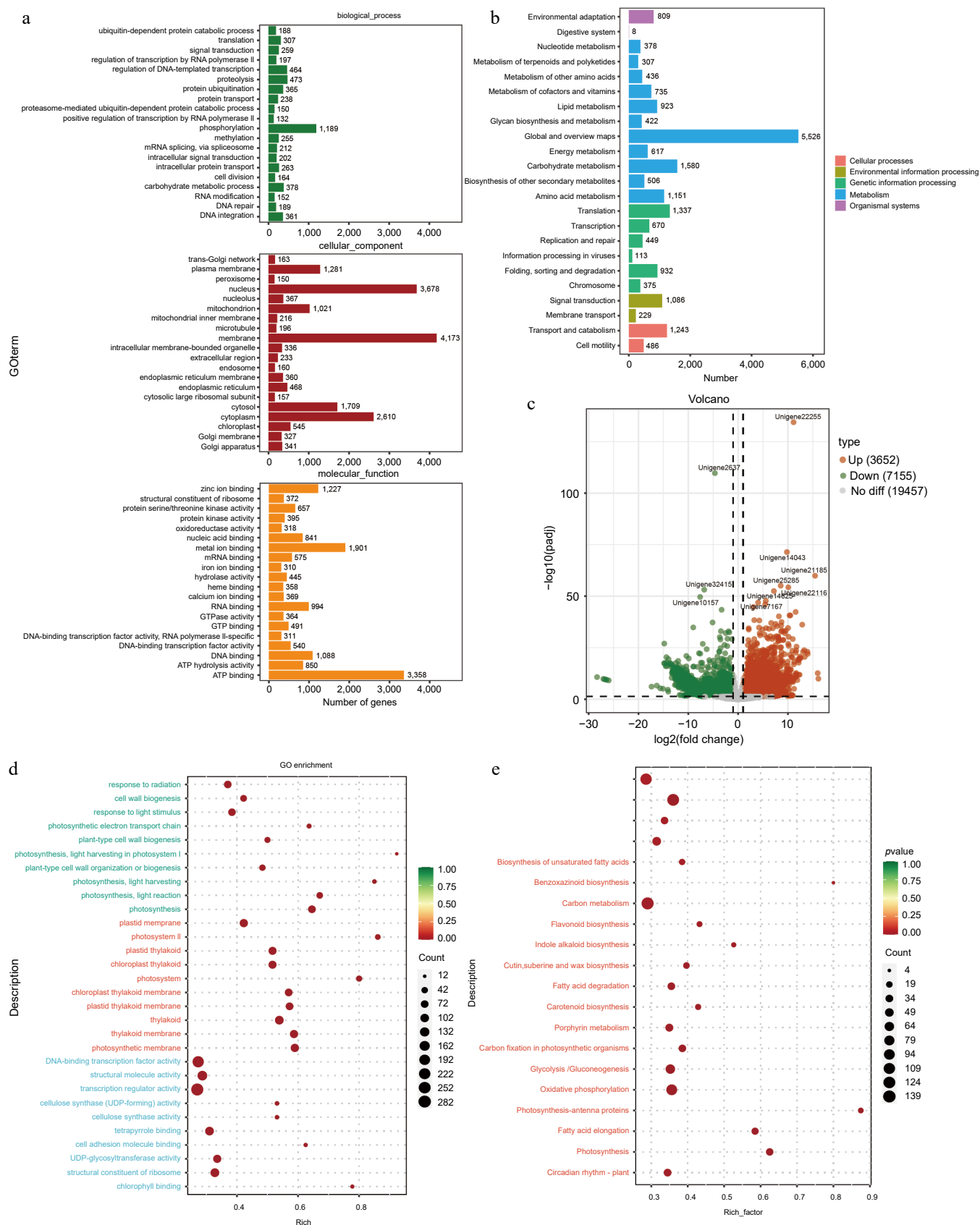
cycle of ecosystems by altering the soil microbial community<sup>[32]</sup>. Furthermore, De Francisco Martínez et al. identified novel cold-tolerance-related genes from the rhizosphere microorganisms of two Antarctic cold-tolerant plants<sup>[33]</sup>. These findings collectively demonstrate the influence of rhizosphere microorganisms on different host plants across various habitats and the functional diversity of microbial communities. Thus, it is essential to reveal the relationship between the rhizosphere microbiome and host plants to provide a basis for optimizing the cultivation process in agricultural production systems. In this study, we employed amplicon sequencing and transcriptomics technology to investigate the rhizosphere bacterial community structure of *A. cochinchinensis* and analyze the gene expression of host plants. Results showed that *RB41*, *Vicinamibacteraceae*, and *WD2101\_soil*, and all these bacterial genera were common in medicinal plants, which was similar with previous studies<sup>[34–38]</sup>. Additionally, the transcriptomics data demonstrated that the rhizosphere bacterial community not only promoted the growth of *A. cochinchinensis* but also assisted the host plants in resisting the biotic and abiotic stress. It has been reported that the rhizosphere microbiome has the potential to influence natural product synthesis in medicinal plants. The rhizosphere is an important place for compound exchange between medicinal plants and soil<sup>[39,40]</sup>. Rhizosphere microorganisms had an important impact on the growth and the formation of bioactive compounds<sup>[41]</sup>. Jamwal et al. observed that the metabolites and rhizosphere microbial community in plants of *Coleus barbatulus* in different developmental stages are changing<sup>[42]</sup>. Similarly, Chen et al. demonstrated that the composition of rhizosphere microorganisms had a profound influence on key bioactive compounds, including dicarboxylates, organic acids, and polysaccharides. It has been reported that polysaccharides are the major bioactive compounds in *A. cochinchinensis*<sup>[43]</sup>. Fan et al. have highlighted the role of the rhizosphere microbiome in affecting polysaccharide accumulation in *Codonopsis pilosula*<sup>[44]</sup>. Ding et al. conducted metagenomic sequencing on the rhizosphere microorganisms of three *Astragalus* species, revealing that KEGG annotations of these microbial communities were enriched in metabolic pathways related to triterpenoids, flavonoids, and polysaccharides<sup>[45]</sup>. There are however limited studies focusing on the relationship between *A. cochinchinensis* polysaccharide synthesis and rhizosphere microbiome. Given these findings, further exploration is warranted to identify beneficial rhizosphere bacteria that may enhance polysaccharide synthesis in *A. cochinchinensis*. This study offers valuable insights for optimizing the cultivation and medicinal quality of *A. cochinchinensis*.

## Author contributions

The authors confirm contribution to the paper as follows: project administration: Zhang X, Song C; resources: Zhang X, Liu X, Tang X, Wang L, Chen J, Liu H, Liang S, Wang X; formal analysis: Zhang X, Yang S, Liu C; methodology: Zhang X, Yu J, Liu X, Tang X, Wang L, Chen J, Luo H, Liang S, Wang X, Song C; data curation: Zhang X; writing – original draft: Zhang X, Yang Y, Yu J, Song C; writing – review & editing: Zhang X, Yu J, Song C; investigation: Yang S; conceptualization, Funding acquisition: Song C. All authors reviewed the results and approved the final version of the manuscript

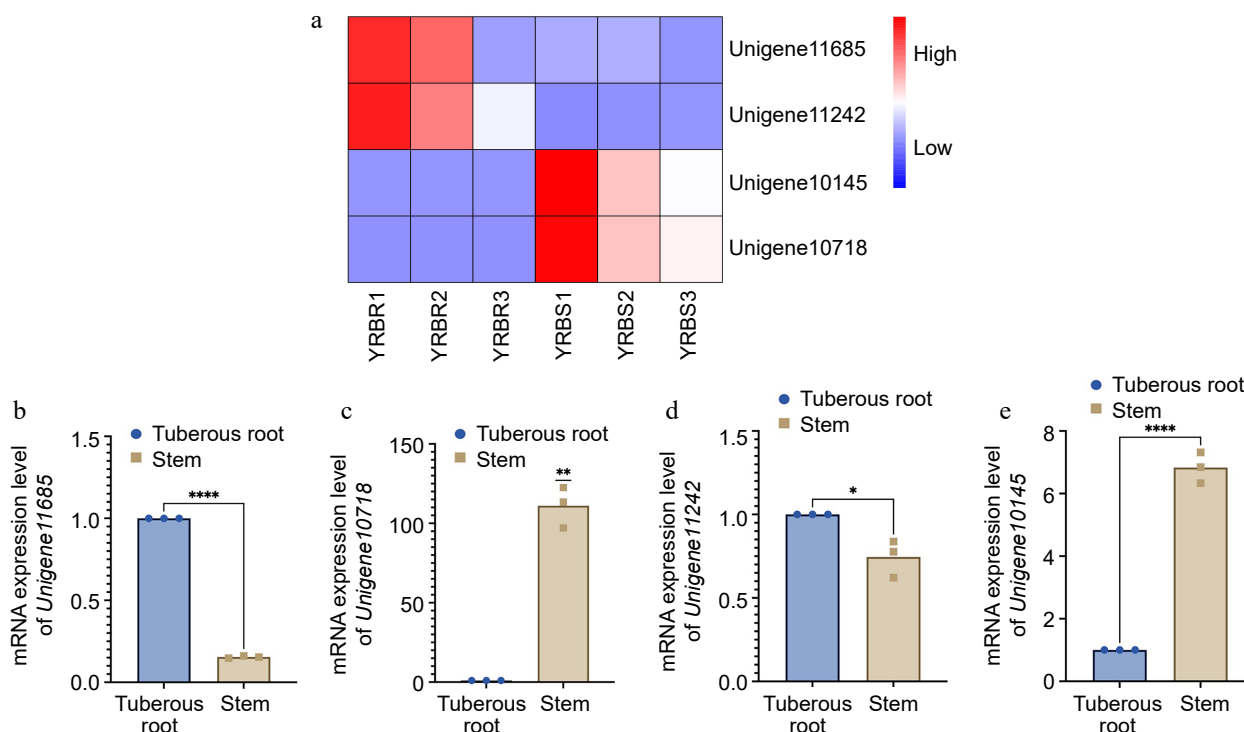
## Data availability

The data that support the findings of this study are available in the National Center for Biotechnology Information Sequence Read Archive database repository with accession numbers SAMN43415523- SAMN43415531, SAMN43440143-SAMN43440148.



**Fig. 5** Functional prediction of different tissues of *A. cochinchinensis*. (a) Function prediction of different tissues of *A. cochinchinensis* based on the Gene Ontology database. (b) Function prediction of different tissues of *A. cochinchinensis* based on the Kyoto Encyclopedia of Genes and Genomes database. (c) Comparing the differential expression of roots and stems unigene. (d) Gene function prediction of differentially expressed unigenes based on the Gene Ontology database. (e) Gene function prediction of differentially expressed unigenes based on the Kyoto Encyclopedia of Genes and Genomes database.





**Fig. 6** mRNA expression level of four differentially expressed genes in different tissues of *A. cochinchinensis*. (a) Comparison of expression level of *Unigene11685*, *Unigene10718*, *Unigene11242*, and *Unigene10145* in tuberous root and stem samples of *A. cochinchinensis* based on heatmap. (b) mRNA expression level of *Unigene11685* normalized with *EF1α* in the tuberous root and stem samples of *A. cochinchinensis*. (c) mRNA expression level of *Unigene10718* normalized with *EF1α* in the tuberous root and stem samples of *A. cochinchinensis*. (d) mRNA expression level of *Unigene11242* normalized with *EF1α* in the tuberous root and stem samples of *A. cochinchinensis*. (e) mRNA expression level of *Unigene10145* normalized with *EF1α* in the tuberous root and stem samples of *A. cochinchinensis*. \*  $p < 0.05$ , \*\*  $p < 0.01$ , \*\*\*\*  $p < 0.0001$ .

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## Conflict of interest

The authors declare that they have no conflict of interest. Dr. Chi Song is the Editorial Board member of *Medicinal Plant Biology* who was blinded from reviewing or making decisions on the manuscript. The article was subject to the journal's standard procedures, with peer-review handled independently of this Editorial Board member and the research group.

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