

Microeukaryotic Variation in Local Sediments with the Influence of Sea-Crossing Bridge Construction: A Case Study in East China

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ABSTRACT

Assessing the ecological status of benthic habitats is important in marine ecosystem management. Sea-crossing bridges have been constructed worldwide to facilitate human travels and transactions. However, limited information is available on the ecological impacts of bridge constructions on local microeukaryotic compositions and functional shifts. In the present study, next-generation sequencing and bioinformatics analyses were performed to compare changes of microeukaryotic communities in local sediments influenced by sea-crossing bridge construction. Relatively low levels of alpha diversity and high levels of beta diversity were observed in samples influenced by bridge construction (group EG). The decreased abundance of Chloroplastida and increased abundance of Animalia in group EG suggested that engineering activities induced environmental disturbance, which impaired the ecosystem balance. LEfSe and SIMPER approaches revealed a significant abundance of ectomycorrhizal fungi in group EG, while these taxa were rare in sediments of a control group. Increased abundance and metabolic functions of these rare ectomycorrhizal fungi suggested that rare microeukaryotes should play fundamentally ecological roles in local ecosystems, especially when environmental perturbations occurred. This report is the first to address the ecological impacts of sea-crossing bridge construction on local microeukaryotic communities, which can improve our understanding of local microbial responses to marine infrastructure construction.

Article Information

Received 02 November 2022

Revised 15 December 2022

Accepted 04 December 2022

Available online
(early access)

Published

Authors' Contribution

HJ and TL conceived the study. HJ analyzed the data and drafted the manuscript. TL reviewed the manuscript. JX, HY, and MH collected the samples. All authors read and approved the final manuscript.

Key words

Microbial eukaryotes, Human activity, Marine engineering, 18S rRNA, Ectomycorrhizal fungi

INTRODUCTION

Assessing the ecological status of benthic habitats is important in marine ecosystem management and functional maintenance (Harrison *et al.*, 2021), especially when environmental perturbations or changes occur (Logares *et al.*, 2014). Invisible microorganisms are considered as the most important players in an ecosystem (Graham *et al.*, 2016). Among them, microbial eukaryotes play fundamental and essential roles in marine ecosystem functioning and biogeochemical processes at local and global scales (Falkowski *et al.*, 2008; Caron *et al.*, 2012). Previous studies have shown that microeukaryotes involve

in food webs as primary producers, consumers, and decomposers (Massana *et al.*, 2015; Shulze *et al.*, 2017; Field *et al.*, 1998). In addition, they have significant effects on biogeochemical cycles (Caron *et al.*, 2012). Thus, microbial eukaryotes are crucial to ecological stability and integrity (Shi *et al.*, 2020), which have attracted plenty of attention in studies related to ecosystem disturbance (Jones *et al.*, 2018; Huang *et al.*, 2020; Liu *et al.*, 2021; Philippot *et al.*, 2021).

With rapid urban radiation and economic development, the increasing intensity of global human activities have dramatically shaped and impaired ecosystem diversity, function, and services (Huang *et al.*, 2020; Ellis *et al.*, 2021). Coastal and marine environments are among the most productive ecosystems on Earth. With nearly two thirds of the human populations living in coastal regions, coastal and marine engineering facilities supply plenty of societal needs such as transportation, protection and energy production (Ido and Shimrit, 2015). Human activities have also impaired coastal and marine environments and human health (Chen *et al.*, 2019). As crucial ecological indicators, local microbial communities play important roles in monitoring environmental perturbations in coastal

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0030-9923/2022/0001-0001 \$ 9.00/0



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and marine ecosystems (Zhang *et al.*, 2020). Previous studies have reported the resistance and resilience of microorganisms in response to environmental disturbances (i.e., Logue *et al.*, 2015), showing microbial contributions to ecosystem stability (Huang *et al.*, 2020). Considering the importance of microbes in maintaining ecosystem and ecological balance, expounding and predicting the variation of microbes in ecological changes is a high-priority issue in microbial ecology (Zhou *et al.*, 2010). However, few studies have investigated microbial changes in coastal ecosystems (Huang *et al.*, 2020) characterized by higher population density and human activity intensity exist.

Coastal and marine engineering constructions can result in drastic environmental disturbances, which can further lead to significant changes in composition and abundance of local microbial communities (Grimm *et al.*, 2008). In contrast to research on bacteria, limited information is available on investigations of microbial eukaryote communities. In the present study, local sediment samples originated from Zhoushan, Zhejiang Province, China were collected, and comparative analyses were conducted for the microeukaryotic communities inhabiting these sediments, through 18S rRNA gene amplicon sequencing, to indicate potential ecological impacts related to sea-crossing bridge construction. The results revealed significant changes in the diversity, abundance, and function of microbial eukaryotes in response to sea-crossing bridge construction, which could provide new insights into ecological management of coastal and marine engineering projects.

MATERIALS AND METHODS

Sediment collection

Zhou-Dai Bridge is one of the newly built sea-crossing bridges in Zhoushan, Zhejiang Province, China (Fig. 1). When the bridge was nearly constructed, a total of 42 sediment samples were collected on October 24, 2020 to tentatively evaluate the ecological impacts of bridge building on local microeukaryotic communities, among which 20 samples were from the surrounding sediments as the control group (Group CG, ID numbers A12-A31) and 22 were from the surfaces of sea-crossing cable-stayed bridge pier bodies as the experimental group (Group EG, ID numbers A32-A53). Samples in group CG were collected approximately 10 m away from bridge piers to minimize the effects of the bridge building. The coordinates of sampling sites were from 30.178°N, 121.983°E to 30.193°N, 121.995°E. When sampling, core sediments of each sample were collected to exclude potential seawater contamination, and the weights of the

sediment samples were assured to be sufficient for DNA extraction. The sediment samples were placed in sealed plastic bags and stored provisionally in a portable ice box, and then transferred to the laboratory within 24 h and stored at -80 °C. Given our objectives were mainly to reveal the potential ecological impacts of bridge building process on local microeukaryotic communities, thus general physicochemical factors including total nitrogen, total phosphorus, and heavy metals were excluded in our study. Instead, four physical parameters, including seawater flow velocity (VF), depth from sea surface to sampling site (DEP), drilled pile shaft friction (FR1), and sinking pile shaft friction (FR2) were calculated for further association analyses (Supplementary Table SI).



Fig. 1. Location of Zhou-Dai Bridge and overall sampling strategy (photo by T.L.).

DNA extraction, PCR amplification and sequencing

Total genomic DNA was extracted from a sediment sample using a CTAB method (Zhou *et al.*, 1996). DNA purity and concentration was assessed by using 1% agarose gels. DNA concentration was further diluted to 1 ng/μL using sterile water and stored at -80 °C. The V4 region fragments of 18S rRNA genes were amplified using a universal primer pair V4F (CCA GCA SCY GCG GTA ATW CC)-V4R (ACT TTC GTT CTT GAT YRA) (Stoeck *et al.*, 2010; Hirakata *et al.*, 2019). The PCR reaction system contained 15 μL of Phusion High-Fidelity PCR Master Mix (New England Biolabs), 2 μM of forward and reverse primers, and approximately 10 ng template DNA. Thermal cycling was under the following conditions: Initial denaturation at 98 °C for 1 min, followed by 30 cycles of denaturation at 98 °C for 10 s, annealing at 50 °C for 30 s, and extension at 72 °C for 30 s, with a final extension at 72 °C for 5 min. PCR products were detected using 2% agarose gel electrophoresis with loading buffer and SYB green dye. Mixed in equidensity ratios, PCR products were then purified by using a Qiagen Gel Extraction Kit (Qiagen, Germany).

Sequencing libraries were constructed with a TruSeq DNA PCR-free sample preparation kit (Illumina, USA) and a specific index was added. The library quality was assessed and sequenced on an Illumina NovaSeq6000 platform. The library preparation and Illumina sequencing processes were performed at Novogene Co. in Beijing.

Data analysis

For quality check and filtering of raw data, QIIME v1.9.1 (Caporaso *et al.*, 2010) software was used to obtain clean tags (Bokulich *et al.*, 2013). Chimera sequences were detected and removed by using UCHIME software (Edgar *et al.*, 2011). Sequence analysis was performed by using UPARSE v7.0.1001 software (Edgar, 2013). Sequences were assigned as one operational taxonomic unit (OTU) with the similarity threshold of $\geq 97\%$. For representative sequences of each OTU, the Silva 138 database was used based on Mothur algorithm for taxonomic information annotation, with confidence score ≥ 0.8 (Quast *et al.*, 2013). Alpha diversity indices including Chao1, ACE, and Shannon index were calculated using QIIME v1.9.1 and statistically plotted with R software. To analyze the phylogenetic relationship of identified OTUs and estimate the difference of the dominant species in samples or groups, multiple sequence alignment was conducted using MUSCLE v3.8.31 software (Edgar, 2004). The phylogenetic tree of the top 100 genera was constructed using MEGA software based on neighbor-joining algorithm (Tamura *et al.*, 2013).

Weighted and unweighted unifracs Beta diversity were calculated using QIIME v1.9.1 software. R packages including WGCNA, stat, and ggplot2 were used to perform principal coordinate analysis (PCoA). R package vegan was used to conduct non-metric multidimensional scaling (NMDS) analyses. The statistical analyses of difference (i.e. independent t-test and Wilcoxon tests) between groups were calculated using R software with P value threshold of <0.05 . The ANOSIM and MRPP functions in vegan package were employed to conduct ANOSIM and MRPP analyses, respectively. Linear discriminant analysis (LDA) effect size (LEfSe) (Segata *et al.*, 2011) and SIMPER analyses (Warton *et al.*, 2012) were used to identify differential species between groups. To estimate the potential relatedness between physical parameters and microeukaryotic distribution, the CCA and RDA functions in vegan package were used to perform canonical correspondence analysis (CCA) and distance-based redundancy analysis (dbRDA), respectively. The presumptive functional attributes of related fungi in this study were annotated using FUNGuild (<https://github.com/UMNFun/FUNGuild>) database (Nguyen *et al.*, 2016). The FUNGuild annotations include guild, trophic

mode, and growth morphology; only confidence scores of probable and highly probable were used.

RESULTS

Overall microeukaryotic communities determined by 18S rRNA gene sequencing

A total of 2,642,382 effective tags were retained after quality and chimera filtering, ranging from 42,242 to 69,399 and with an average of 62,914 (Supplementary Table SII). Good's coverage estimates of 99.0%–99.8% were obtained for the sequencing data (Supplementary Fig. S1) and the observed OTUs ranged from 257 to 1805, with an average of 1209 (Supplementary Table SII). Phylogenetic relationship of top 100 genera species was classified as Ascomycota, Basidiomycota, Chlorophyta, Streptophyta, Diatomea, Ciliophora, Arthropoda, Cnidaria, and Chordata at the phylum level (Fig. 2). Eukaryota, Fungi, and Chloroplastida were the top three kingdoms in group CG (77.43% in total), while Eukaryota, Fungi, and Animalia were dominant (71.96% in total) in group EG (Supplementary Fig. S2). The relative abundance of Chloroplastida decreased from 23.68% in CG to 12.85% in EG, with Phyla Streptophyta and Chlorophyta as the main contributors. The alpha diversity estimates, including Chao1, ACE, and Shannon index, were relatively higher in the CG group than in the EG group (Supplementary Table SIII, Supplementary Fig. S3).

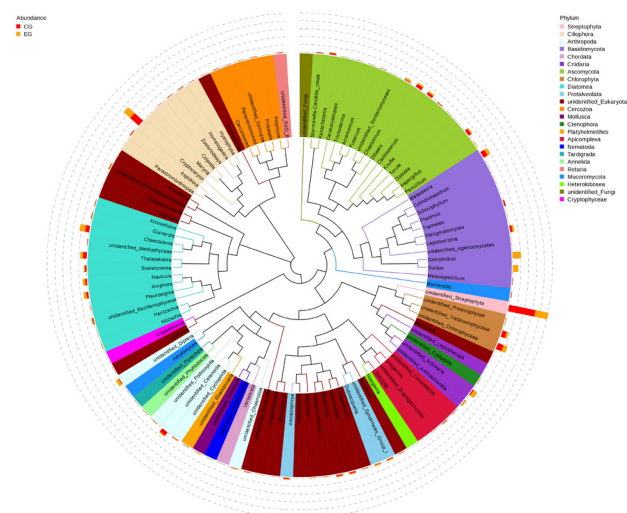


Fig. 2. Phylogenetic relationship of top 100 genera species identified in this study.

Microeukaryotic community composition between groups

Contrary to alpha diversity, the beta diversity index showed that samples in group EG had significantly higher

levels of species dissimilarity (Fig. 3a, b). Consistently, the PCoA and NMDS plotting exhibited similar distribution patterns, also showing higher species variations in group EG (Fig. 3c, d). In addition, samples A43, A44, A45, A50, A51, and A52 in group EG diverged from other samples based on PCoA and NMDS analyses. The results of statistical analyses also confirmed significant differentiation between groups ($P=0.005$ and 0.002 in ANOSIM and MRPP, respectively). To further identify significantly different taxa between groups, t-test was performed at different classification levels. At the phylum level (Fig. 4), four phyla including Streptophyta, Cnidaria, Rotifera, and Picozoa were identified as significantly different taxa. At the species level, a total of 40 species showed significantly different abundance between groups, half of which (22/40) belonged to Eukaryota (Supplementary Table SIII). LEfSe analysis was also implemented to identify unique microorganisms (biomarkers) that differed significantly in abundance between groups (Segata *et al.*, 2011). Our findings showed that kingdom Animalia and affiliated phylum Cnidaria, family Suillaceae and affiliated genus *Suillus*, family Gomphidiaceae, affiliated genus *Gomphidius* and species *Gomphidius roseus* were particularly abundant in group EG, while kingdom Chloroplastida and affiliated phylum Streptophyta, species *Zea_mays* and *Citrullus_lanatus*, and order Calanoida were enriched in group CG (Fig. 5). In addition, our LEfSe results (Supplementary Fig. S4) further revealed that fungi in family Suillaceae and Gomphidiaceae were rare in group CG but enriched in group EG, suggesting that these rare microbial eukaryotes should play ecological roles in increasing abundance or maintaining ecosystem functioning after environmental disturbance (Logares *et al.*, 2014). Additionally, the contributions of eukaryotes at the phylum and species level to variation between groups were calculated using SIMPER analysis (Fig. 6). The phylum-level result showed that Streptophyta, Ciliophora, Basidiomycota, and Arthropoda were differential taxa with a high contribution ratio. Consistent with results of PCoA and NMDS analyses, phylum Basidiomycota was enriched in samples A43, A44, A45, A50, A51, and A52 in group EG, but was rare in the remaining samples. The species-level result also revealed that *Gomphidius roseus*, which belonged to phylum Basidiomycota, was differential species with a high contribution ratio. The fungus *Gomphidius roseus*, which was also identified as one of the biomarkers in LEfSe analysis, was also enriched in samples A43, A44, A45, A50, A51, and A52.

Correlation analyses, including CCA and dbRDA approaches, were performed to detect possible impacts of four physical factors (i.e., VF, DEP, FR1, and FR2) on eukaryotic communities. Our findings showed

significant correlations between the physical factors and microeukaryotic composition (Supplementary Fig. S5), suggesting that local microeukaryotic communities were highly varied during the bridge construction process.

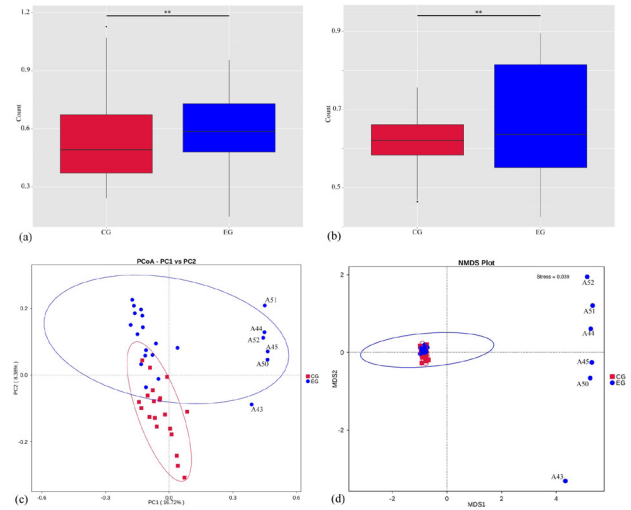


Fig. 3. Beta diversity based on weighted (a) and unweighted (b) unifracs, PCoA plotting (c) and NMDS plotting (d) in this study. Scatters with numbers in (c) and (d) denote divergent samples in group EG. Significance $P < 0.01$.

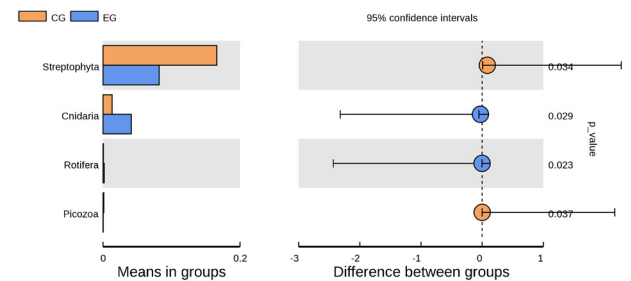


Fig. 4. Significantly different phyla between groups based on t-test approach.

Functional prediction of related fungi

Fungi were identified as contributors and biomarkers in our study, suggesting the ecological importance of these taxa. FUNGuild approach was performed to predict the nutritional and functional groups of the related fungal communities between groups. Our findings showed that nine trophic mode groups were classified, with saprotroph, symbiotroph, pathotroph-symbiotroph, and pathotroph-saprotroph being the major components (Supplementary Fig. S6). In group EG, saprotroph, symbiotroph, and pathotroph-symbiotroph were the top three primary trophic modes, while the top three in group CG were saprotroph, symbiotroph, and pathotroph-saprotroph.

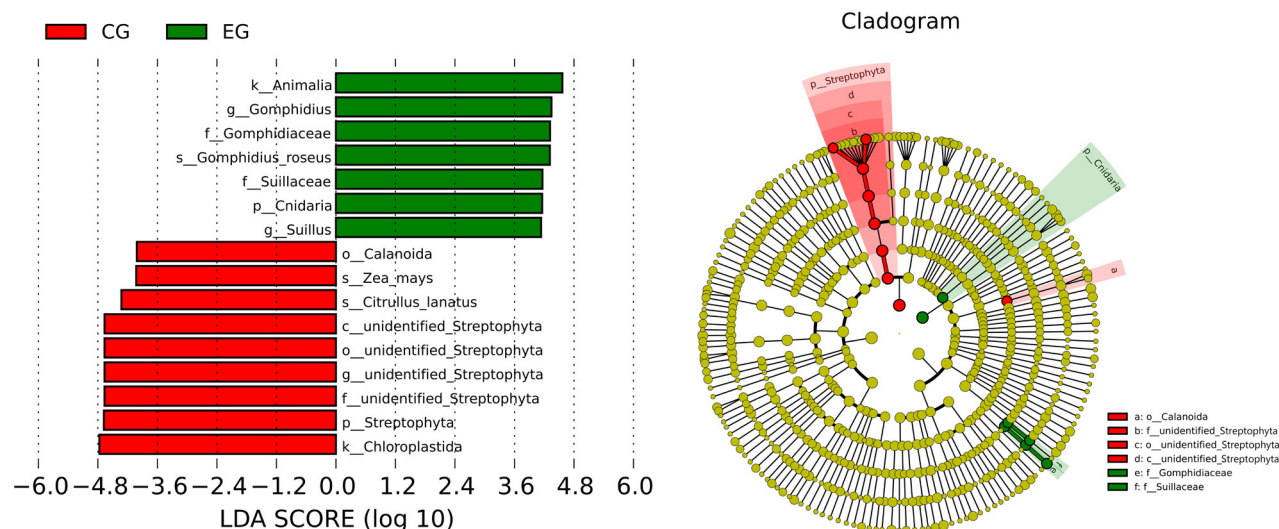


Fig. 5. Microbial biomarkers identified based on LEfSe approach. Left panel histogram showing LDA scores of identified biomarkers; right panel cladogram showing phylogenetic distribution of identified biomarkers.

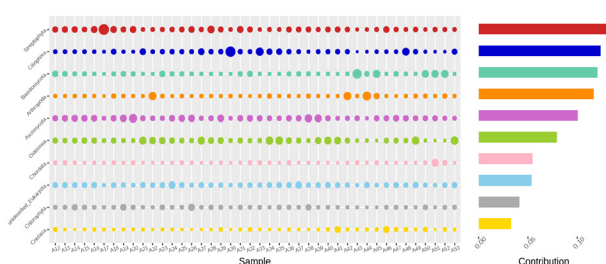


Fig. 6. Top 10 phyla with high contribution ratio identified using Simper.

Significance test (t-test) results showed that pathotroph-symbiotroph was the significantly different trophic mode between groups ($P = 0.035$, [Supplementary Fig. S7a](#)). Furthermore, our results showed that the compositions of ectomycorrhizal and ectomycorrhizal-fungal parasite fungi were significantly varied between the CG and EG groups, both of which were more abundant in group EG ([Supplementary Fig. S7b](#)).

DISCUSSION

The ecosystem biodiversity, function, and service aspects, which are largely supported by the microbial communities, have been dramatically changed, shaped, and impaired by increasing human activities ([Huang et al., 2020](#)). Some studies have reported that the ecosystem function and service depend upon both microbial composition and specific functional groups, suggesting that microbial diversity and function should be connected

to ecosystem function and service aspects ([Pérez-Valera et al., 2015](#); [Galand et al., 2016](#)). In the present study, we investigated the changes of microeukaryotic composition, abundance, and function in local sediments following bridge construction to reveal the effects of human activities such as bridge construction on local microeukaryotic communities.

We found that despite the higher alpha diversity of group CG, group EG had high levels of community dissimilarity and structure heterogeneity. Moreover, significant differentiation of microeukaryotic composition, abundance, and function were observed between groups, revealing the impacts of bridge construction on local microeukaryotic communities. Human activities, such as engineering construction, have tremendously shaped marine and coastal environments, and these changes would affect ecosystem function and services ([Chapin et al., 2000](#)), which in turn could lead to adverse impacts on human societies. Invisible microorganisms play crucial roles in ecosystems and are the most important players in ecosystem function and services ([Graham et al., 2016](#)). As a result, the diversity and functional attributes of local microbial communities were generally affected by human activities. For instance, a previous study showed that microbial communities had been dramatically changed during urbanization in coastal regions ([Huang et al., 2020](#)). In our study, microeukaryotic traits including diversity, structure, abundance, and function were significantly changed during sea-crossing bridge construction, especially the abundance and function of ectomycorrhizal fungi such as *Gomphidius roseus*. The decreased

abundance of chloroplastida and increased abundance of Animalia in group EG suggested that engineering-induced environmental disturbance impaired the ecosystem balance, leading to lower primary productivity. The enriched ectomycorrhizal fungi in group EG could counteract such feedback. Previous studies suggested that ectomycorrhizal fungi could not only transport dissolved nutrients but also mobilize essential plant nutrients directly from minerals through excretion of organic acids (Landeweert *et al.*, 2001). Besides, ectomycorrhizal fungi also had an active part in organic matter decomposition, which could facilitate co-metabolic degradation of recalcitrant organic complexes (Lindahl and Tunlid, 2015). Thus, the increased abundance of ectomycorrhizal fungi could facilitate primary productivity and maintain ecosystem balance.

In this study, ectomycorrhizal fungi in family Suillaceae and Gomphidiaceae were rare subcommunities in group CG samples, but enriched in group EG with significant abundance and functional attributes. Such changes in abundance should be an example to elucidate the ecological importance of rare microbes in the ecosystem. Apart from limited information on marine microeukaryote diversity and composition (Arrigo, 2005; Caron *et al.*, 2009), the ecological roles and functions of rare marine microbes remain unknown. Rare marine microbes are hypothesized to be ecologically redundant taxa that could increase in abundance when environmental disturbance occurs and maintain ecosystem biodiversity, service, and function (Caron and Countway, 2009). Some studies also revealed that rare bacteria could grow exponentially under the right conditions, be metabolically more active than other taxa, and perform crucial ecosystem functions (Jones and Lennon, 2010; Pester *et al.*, 2010). In this study, we also observed the increased abundance and metabolic functions of the rare ectomycorrhizal fungi, suggesting that these ectomycorrhizal fungi may play fundamental roles in local ecosystems. The functional prediction of fungi showed that saprotroph, symbiotroph, and pathotroph-symbiotroph were the main trophic modes in group EG. Hall *et al.* (2003) reported that ectomycorrhizal fungi could be classified as symbionts, saprobes, and pathogens, and the trophic mode might shift depending on the phase in the life cycle of a given fungus. In addition, some bacteria can be associated with ectomycorrhizas and appear to aid the infection process (Hall *et al.*, 2003). As a result, given the extent of ectomycorrhizal fungi increase in group EG, we noted that pathogenic fungi and associated bacteria might infect local flora and fauna, especially seafood and people in local communities, leading to negative health effects on human society. Thus, ecological investigations are needed to monitor the impacts of engineering construction on local microbiota and the flora and fauna biosphere.

Local microbial communities play crucial roles in monitoring environmental perturbations in coastal and marine ecosystems, especially when human activities such as infrastructure engineering were increasingly implemented in recent years. In the current study, by using 18S rRNA gene amplicon sequencing, comparative analyses of microeukaryotic communities were conducted to indicate potential ecological impacts of sea-crossing bridge construction on local microbial community. Significant changes of microeukaryotic communities were detected, providing novel ecological insights into infrastructure construction on local environment. In addition, we detected some rare ectomycorrhizal fungi were enriched with environmental disturbance, suggesting rare microbes should play crucial roles in ecosystem stability. These findings could provide reference information for further ecological engineering construction.

ACKNOWLEDGMENTS

This work was supported by the Research on Self-Repairing Technology of Nano-modified Microorganisms on Concrete Structural Cracks (grant number CJHX22061) and Applied Technology of Microbial Healing Cracks of Seepage Water in SKS201 Wuling Tunnel (grant number 21186006520).

Funding

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IRB approval

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Ethical statement

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Data availability statement

The sequencing data in this study have been deposited in Sequence Read Archive (SRA) database under accession number PRJNA806524.

Supplementary material

There is supplementary material associated with this article. Access the material online at: <http://dx.doi.org/10.17582/journal.pjz/.....>

Statement of conflict of interest

The authors have declared no conflict of interest.

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Supplementary Material

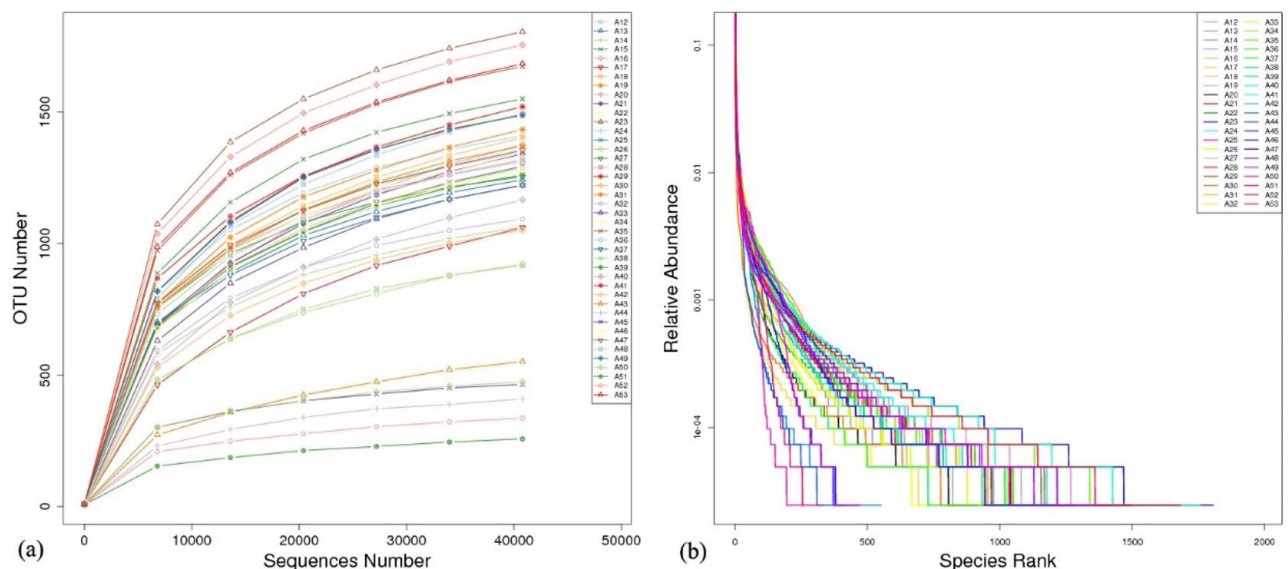
Microeukaryotic Variation in Local Sediments with the Influence of Sea-Crossing Bridge Construction: A Case Study in East China

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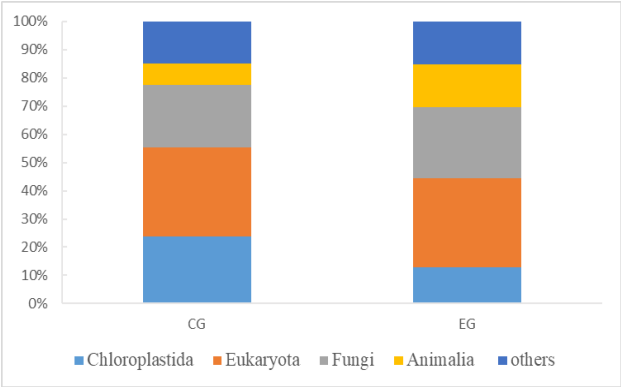
Supplementary Fig. S1. Rarefaction curves (a) and rank abundance curves (b) of all sediment samples.

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0030-9923/2022/0001-0001 \$ 9.00/0

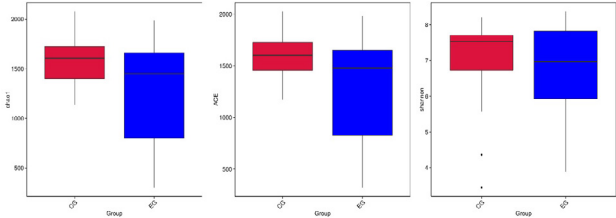


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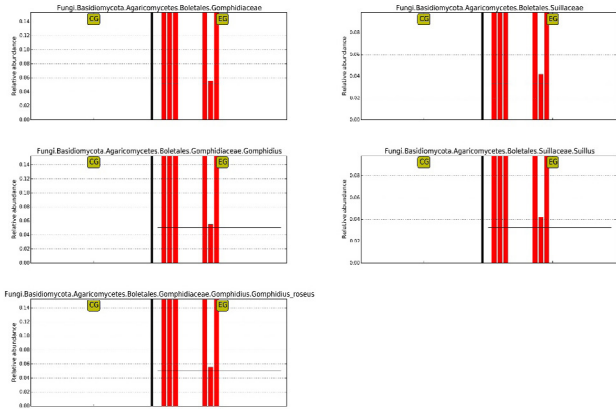
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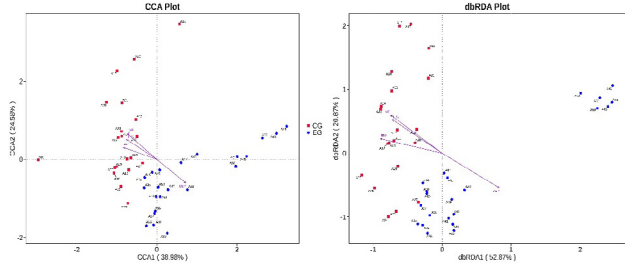
Supplementary Fig. S2. Relative abundance at the kingdom level between two groups.



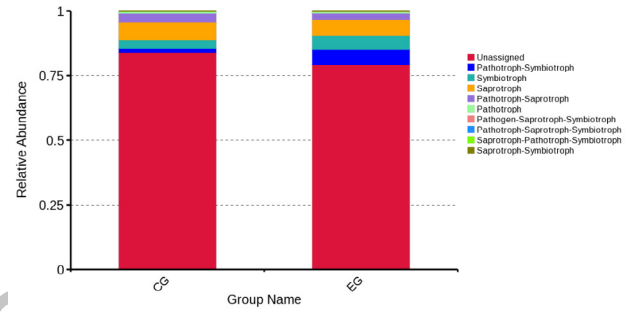
Supplementary Fig. S3. Alpha diversity estimates between two groups.



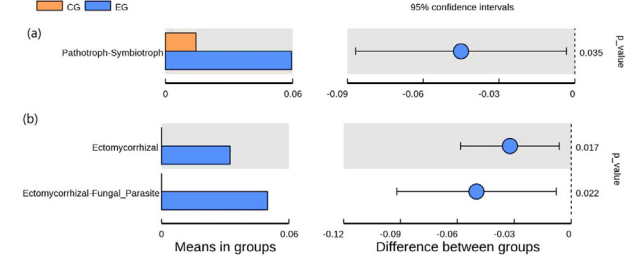
Supplementary Fig. S4. Relative abundance of identified *Fungi* biomarkers in LEfSe analysis.



Supplementary Fig. S5. CCA (left panel) and dbRDA (right panel) plotting showing the relationship between microbial communities and physical factors.



Supplementary Fig. S6. Relative abundance of classified trophic mode groups in this study.



Supplementary Fig. S7. Trophic modes (a) and compositions (b) with different abundance between groups.

Supplementary Table SI. Statistical information of four environmental factors (physical parameters).

Group	Sample ID	Flow velocity (VF, cm/s)	Depth (DEP, m)	Drilled pile shaft friction (FR1, kPa)	Sinking pile shaft friction (FR2, kPa)
CG	A12	16	26	46	50
	A13	17	28	50	47
	A14	17	28	50	47
	A15	16	26	46	43
	A16	14	29	48	44
	A17	14	29	48	44
	A18	9	30	50	42
	A19	7	31	50	40
	A20	16	28	48	47
	A21	17	28	48	47
	A22	16	29	50	44
	A23	14	29	50	44
	A24	6	31	50	40
	A25	16	26	46	50
	A26	14	26	46	50
	A27	17	27	48	48
	A28	17	27	48	48
	A29	16	29	50	47
	A30	16	29	50	47
	A31	15	29	50	47
EG	A32	4	35	37	40
	A33	4	35	37	40
	A34	4	35	37	40
	A35	6	34	35	42
	A36	6	34	35	42
	A37	6	34	35	42
	A38	6	34	35	42
	A39	4	35	37	40
	A40	4	35	37	40
	A41	6	34	35	42
	A42	6	34	36	39
	A43	6	34	36	39
	A44	6	34	36	39
	A45	4	35	40	37
	A46	4	35	40	37
	A47	6	34	36	39
	A48	4	35	40	37
	A49	4	35	40	37
	A50	4	35	40	37
	A51	4	35	40	37
	A52	6	34	36	39
	A53	4	35	40	37

Supplementary Table SII. Information of sequencing data and OUT statistics.

Group	Sample ID	Effective read	Base (bp)	Q20 (%)	Q30 (%)	OUT number
CG	A12	60,119	18,358,689	98.44	95.73	1580
	A13	61,242	18,763,692	98.40	95.58	1462
	A14	47,889	14,684,940	98.42	95.65	1237
	A15	47,780	14,822,905	98.32	95.32	1699
	A16	67,719	20,741,616	98.47	95.81	1626
	A17	65,180	19,913,073	98.46	95.81	1297
	A18	67,748	20,889,990	98.43	95.62	1594
	A19	65,302	20,009,692	98.42	95.68	1601
	A20	65,534	20,046,608	98.46	95.78	1366
	A21	63,947	19,532,364	98.19	95.12	1569
	A22	61,953	19,081,032	97.95	94.30	1694
	A23	63,136	19,277,055	98.30	95.31	1957
	A24	64,267	19,658,311	98.45	95.72	1510
	A25	67,598	20,668,749	98.50	95.76	1459
	A26	66,675	20,318,869	98.44	95.67	1076
	A27	45,636	13,870,218	98.41	95.54	1375
	A28	65,551	20,014,589	98.43	95.74	1455
	A29	69,399	21,165,583	98.44	95.72	1688
	A30	63,672	19,044,583	98.35	95.66	1215
	A31	68,501	20,897,564	98.46	95.72	1553
EG	A32	61,569	18,833,779	98.33	95.45	1256
	A33	60,583	18,303,254	98.13	94.98	1405
	A34	64,072	19,569,050	98.27	95.17	1519
	A35	67,198	20,437,811	98.30	95.34	1833
	A36	62,935	19,292,739	98.38	95.51	1481
	A37	64,270	19,725,414	98.37	95.47	1381
	A38	60,346	18,541,628	98.40	95.55	1066
	A39	54,515	16,681,070	98.37	95.43	1392
	A40	42,242	12,910,355	98.39	95.47	1755
	A41	63,156	19,273,792	98.37	95.47	1667
	A42	65,695	21,836,332	98.00	94.10	1460
	A43	66,407	20,509,388	99.00	96.56	637
	A44	67,389	21,365,516	99.37	97.63	455
	A45	66,763	20,558,745	99.02	96.63	523
	A46	67,020	20,546,092	98.33	95.38	1447
	A47	66,929	20,497,720	98.29	95.25	1540
	A48	66,385	20,169,475	98.19	95.00	1668
	A49	61,848	18,947,392	98.25	95.05	1661
	A50	68,322	21,019,394	98.94	96.43	539
	A51	60,649	18,821,705	99.25	97.42	284
	A52	68,281	21,057,146	99.34	97.53	379
	A53	66,960	20,435,181	98.25	95.12	1860

Supplementary Table SIII. Information of Alpha diversity estimates including Chao1, ACE and Shannon index.

Group	Sample name	Shannon	Chao1	ACE
CG	A12	7.644	1524.754	1588.01
	A13	7.963	1386.354	1464.486
	A14	7.436	1229.634	1244.494
	A15	8.207	1722.74	1729.273
	A16	7.63	1758.727	1785.276
	A17	4.357	1403.909	1432.203
	A18	7.859	1760.619	1753.995
	A19	7.034	1739.049	1725.873
	A20	5.568	1490.9	1499.38
	A21	7.074	1690.394	1666.922
	A22	6.076	1623.687	1682.677
	A23	8.042	2077.502	2026.798
	A24	7.654	1600.532	1586.785
	A25	7.672	1613.757	1615.094
	A26	6.307	1135.333	1171.693
	A27	7.678	1364.053	1388.377
	A28	6.872	1568.325	1510.361
	A29	7.252	1807.33	1770.787
	A30	3.44	1309.048	1260.615
	A31	7.792	1643.902	1669.26
EG	A32	7.087	1195.789	1251.762
	A33	6.195	1353.658	1434.943
	A34	7.694	1607.562	1582.827
	A35	7.976	1937.814	1907.484
	A36	7.963	1438.079	1499.376
	A37	7.669	1462.833	1459.835
	A38	6.041	992.426	1057.848
	A39	7.026	1394.541	1420.754
	A40	8.374	1990	1983.051
	A41	7.475	1737.858	1725.273
	A42	6.272	1548.388	1536.358
	A43	4.366	736.067	747.963
	A44	3.877	491.787	501.082
	A45	5.676	528.909	533.35
	A46	6.899	1610.259	1575.955
	A47	7.87	1674	1619.644
	A48	6.914	1818.248	1780.771
	A49	7.906	1616.319	1665.615
	A50	5.893	569.192	567.752
	A51	4.477	301.386	319.135
	A52	5.535	405.188	423.046
	A53	7.991	1923.028	1903.151