# scientific data 

## DATA DESCRIPTOR

# 500 metagenome-assembled microbial genomes from 30 subtropical estuaries in South China 

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As a unique geographical transition zone, the estuary is considered as a model environment to decipher the diversity, functions and ecological processes of microbial communities, which play important roles in the global biogeochemical cycle. Here we used surface water metagenomic sequencing datasets to construct metagenome-assembled genomes (MAGs) from 30 subtropical estuaries at a large scale along South China. In total, 500 dereplicated MAGs with completeness $\geq 50 \%$ and contamination $\leq 10 \%$ were obtained, among which more than one-thirds ( $n=207$ MAGs) have a completeness $\geq 70 \%$. These MAGs are dominated by taxa assigned to the phylum Proteobacteria ( $n=182$ MAGs), Bacteroidota ( $n=110$ ) and Actinobacteriota ( $n=104$ ). These draft genomes can be used to study the diversity, phylogenetic history and metabolic potential of microbiota in the estuary, which should help improve our understanding of the structure and function of these microorganisms and how they evolved and adapted to extreme conditions in the estuarine ecosystem.

## Background \& Summary

The estuary is the intersection of fresh water, land and sea water, where fresh water and sea water with different properties are mixed, and a large number of nutrients and terrestrial microbes are input and accumulated. The complex condition leads to diverse biocoenosis in the estuarine environment. Microorganisms, such as Bacteria and Archaea, are widely distributed, abundant, and play key roles in biogeochemical cycle of carbon, nitrogen, sulfur and phosphorus as well as microbial food web in estuarine ecosystems ${ }^{1-3}$. As one of the most productive ecosystems in the world ${ }^{4}$, the strong natural and anthropogenic gradients in estuaries make them ideal niches to study microbial community structure and its associated functions.

The recent development of high-throughput sequencing technology such as 16 S rRNA gene and metagenome sequencing can identify large amounts of unknown taxa, analyze the characteristics of uncultured microorganisms, and thus has promoted studies of microbial diversity, community assembly, adaptation, evolution and function ${ }^{2,3}$. The research of microbial community structure in various estuaries such as Chesapeake Bay ${ }^{5}$, Delaware estuary ${ }^{6}$, Columbia estuary ${ }^{7}$, and estuaries of Sundarbans (i.e., Mooriganga, Thakuran, Matla, and Harinbhanga ${ }^{8}$ has been carried out. Microbiological studies have also been conducted in several major estuaries in China, such as Yellow River, Yangtze River, Qiantang River and Pearl River ${ }^{1,9-12}$. These studies provided insights into spatial-temporal variations of microbial communities and their responses to environmental changes in estuarine ecosystems. However, despite the increasing knowledge of biodiversity process in the estuarine ecosystem, our understanding of the distributions and ecological preferences and functions of estuarine microbiome across broad spatial scales remains surprisingly limited.

Here we present 500 metagenome-assembled genomes (MAGs) reconstructed from 90 surface water metagenomic samples in 30 subtropical estuaries which span the estuary of 30 major rivers in Guangdong and Guangxi, South China, a range of $\sim 1300 \mathrm{~km}$. All of these MAGs were estimated to be $>50 \%$ complete with $<10 \%$ contamination. Among them, $41.40 \%$ (207) have a completeness $>70 \%$ and $13.20 \%$ (66) have a completeness $>90 \%$, while $75.80 \%$ (379) have low ( $<5 \%$ ) contamination and $4.00 \%$ (20) have no contamination. Together, high-quality MAGs (Completion $>90 \%$ and Contamination $<5 \%$ ) account for $12.2 \%$ (61) and medium-quality MAGs (Completion $\geq 50 \%$ and Contamination $<10 \%$ ) account for $87.8 \%$ (439). The

[^0]| Domain | Phylum | Count | Propotion (\%) |
| :---: | :---: | :---: | :---: |
| d__Archaea | p__Thermoplasmatota | 6 | 1.2 |
|  | p__Thermoproteota | 3 | 0.6 |
| d__Bacteria | p__Proteobacteria | 182 | 36.4 |
|  | p__Bacteroidota | 110 | 22 |
|  | p__Actinobacteriota | 104 | 20.8 |
|  | p__Patescibacteria | 21 | 4.2 |
|  | p__Planctomycetota | 19 | 3.8 |
|  | p__Verrucomicrobiota | 18 | 3.6 |
|  | p__Cyanobacteria | 10 | 2 |
|  | p__Firmicutes | 6 | 1.2 |
|  | p__Chloroflexota | 5 | 1 |
|  | p__Campylobacterota | 4 | 0.8 |
|  | p__SAR324 | 2 | 0.4 |
|  | p__Acidobacteriota | 2 | 0.4 |
|  | p__Margulisbacteria | 1 | 0.2 |
|  | p__Armatimonadota | 1 | 0.2 |
|  | p__Bdellovibrionota_C | 1 | 0.2 |
|  | p__Marinisomatota | 1 | 0.2 |
|  | p__Gemmatimonadota | 1 | 0.2 |
|  | p__Nitrospirota | 1 | 0.2 |
|  | p__Desulfobacterota_B | 1 | 0.2 |
|  | p__Eisenbacteria | 1 | 0.2 |

Table 1. Relative proportion of phyla in MAGs reconstructed from the subtropical estuaries, South China.
draft genomes were classified into 491 bacteria and 9 archaea. A vast majority of them belong to the phyla Proteobacteria (36.4\%), Bacteroidota (22\%), and Actinobacteria (20.8\%) (Table 1; Fig. 1). However, only 62 ( $12.4 \%$ ) could be classified to current known taxa at species level with 438 ( $87.6 \%$ ) representing currently uncultured species. For fully utilizing the genome data, statistics of quality control on metagenomic raw reads is provided in Supplementary Table S1. Assembly information is provided in Supplementary Table S2. Predicted taxon for each MAG, as well as bin statistics (e.g., completeness, contamination, size and N50), are provided in Supplementary Table S3. MAGs abundance in each estuary is provided in Supplementary Table S4 and associated environmental variables is given in Supplementary Table S5.

To the best of our knowledge, this is the largest number of microbial genomes from the largest number of estuaries to be reported in a single study, which should help facilitate future studies in understanding the structure and function of these microorganisms and how they evolved and adapted to the extreme conditions of the estuarine ecosystems.

## Methods

Sample sites and sample collection. A total of 90 surface water samples were collected in December 2018 from 30 sites that spanned the estuary of 30 main rivers in South China, a range of $\sim 1300 \mathrm{~km}$ (Fig. 2). At each estuary, triplicate samples were collected, approximately $30-50 \mathrm{~m}$ apart. 500 mL water was filtered for the metagenome sequencing through $0.22-\mu \mathrm{m}$ pore polycarbonate membranes (Millipore Corporation, Billerica, MA, USA), as most prokaryotes are larger than that size. The filtration was performed within $4 \sim 8 \mathrm{~h}$ and the filter membranes were quick-frozen in liquid nitrogen and then stored at $-80^{\circ} \mathrm{C}$ until DNA extraction.

DNA extraction, metagenomic sequencing and assembly. Total microbial DNA was extracted using a FastDNA Spin Kit for Soil (MP Biomedicals, CA, USA) following the manufacturer's instructions. The quality and concentration of extracted DNA were evaluated by agarose gel electrophoresis (1\%) and Qubit ${ }^{\circledR}$ dsDNA Assay Kit in Qubit ${ }^{\circledR} 2.0$ Flurometer (Life Technologies, CA, USA). All extracted DNA was stored at $-20^{\circ} \mathrm{C}$ for further applications.

A total amount of $1 \mu \mathrm{~g}$ DNA per sample was used as input material for the sequencing preparations. Sequencing libraries were generated using NEBNext ${ }^{\circledR}$ Ultra ${ }^{\text {TM }}$ DNA Library Prep Kit for Illumina (NEB, USA) following manufacturer's recommendations and index codes were added to attribute the sequences to each sample. Briefly, the DNA sample was fragmented by sonication to a size of 350 bp , then DNA fragments were end-polished, A-tailed, and ligated with the full-length adaptor for Illumina sequencing with further PCR amplification (2 circles). At last, PCR products were purified (AMPure XP system) and libraries were analyzed for size distribution by Agilent2100 Bioanalyzer. After cluster generation, the library preparations were sequenced (Paired-end $2 \times 150 \mathrm{bp}$ ) on an Illumina NovaSeq. 6000 platform in Microeco, Shenzhen, China. After sequencing, the raw reads were filtered using kneadData v0.7.4. (https://bitbucket.org/biobakery/kneaddata/wiki/Home) with options (-trimmomatic-options "ILLUMINACLIP: TruSeq. 2-PE.fa:2:40:15 SLIDINGWINDOW:4:20 MINLEN:50"-bowtie2-options "-very-sensitive- dovetail -db Homo_sapiens"). About 6 Gb (giga base pairs)


Fig. 1 Phylogenetic tree of the MAGs constructed by maximum likelihood method using a concatenated alignment of 120 conserved bacterial markers. Concentric rings moving outward from the tree show the completeness, and contamination and inferred phylum. The bar plot shows the size of the MAGs.
of clean metagenomic data was generated for each sample, resulting in a total of $\sim 580 \mathrm{Gbp}$ data. Trimmed metagenomic reads were co-assembled for samples from the same estuary using MEGAHIT v1.2.9 with the default settings ${ }^{13}$. The quality of the metagenomic assemblies assessed with tools like metaQUAST v 5.0.2 ${ }^{14}$.

Genome binning and refinement. Genome binning and refinement were all conducted in metaWRAP $1.3^{15}$. In details, contigs were clustered into metagenomic bins using metaWRAP binning module ( - maxbin2-concoct-metabat2 options). The resulting bins were then refined with metaWRAP's bin_refinement module (-c $50-\mathrm{x} 10$ options). To increase the completion of the bins, and reduce contamination, metaWRAP reassemble_bins module(-c $50-\mathrm{x} 10$ options) was used by extracting the reads belonging to each bin, and reassembling the bins with SPAdes v3.10.1 with the-carefull setting ${ }^{16}$. These decontaminated bins were then dereplicated using dRep v2.6.2 ${ }^{17}$ with parameters: -sa 0.95 -nc 0.30 -comp 50 -con 10 . The bins were then quantified with the Quant_bins module (default parameters) ${ }^{18}$. First, Salmon v0.13.1 ${ }^{19}$ (quasi-mapping-based mode-libType IU-meta options) was used to produce abundance values (TPM) for each contig. Then, the overall abundance of the bin in each sample was calculated by taking the length-weighted average of the contig abundances.

Taxonomic classification and genome tree construction. The taxonomy of the 500 MAGs (bins) were classified using GTDB-Tk v1.3.0 ${ }^{20}$ with the GTDB r202 ${ }^{21}$. Phylogenetic relationships among the 491 bacterial MAGs or nine archaeal MAGs were inferred by constructing a maximum-likelihood tree using 120 bacterial and 122 archaeal marker genes identified in GTDB-Tk. In detail, bacterial and archaeal reference trees are inferred from the filtered 120 and 122 phylogenetically informative markers, respectively. The bacterial reference tree is inferred with FastTree v2.1.10 ${ }^{22}$. under the WAG model. The archaeal reference tree is inferred with IQ-Tree v1.6.9 ${ }^{23}$ under the PMSF model, a rapid approximation of the C 10 mixture model ( $\mathrm{LG}+\mathrm{C} 10+\mathrm{F}+\mathrm{G}$ ), using


Fig. 2 Map of the sampling estuaries. HGH, Huanggang river estuary; HJD, Hanjiangdong river estuary; HJW, Hanjiangwaisha river estuary; RJ, Rongjiang River river estuary; LJ, Lianjiang river estuary; WKH, Wukanhe river estuary; LH, Luohe river estuary; HJH, Huangjianghe river estuary; DAH, Danaohe river estuary; DJN, Dongjiangnan river estuary, HM, Humen mouth; JM, JiaoMen mouth; HQM, Hongqimen mouth; HEM, Hengmen mouth; MDM, Modaomen mouth; JTM, Jitimen mouth; HTM, Hutiaomen mouth; YM, Yamen mouth; MYJ, Moyangjiang river estuary; HJF, Huangjiangfengonghe river estuary; JJ, Jianjiang river estuary; JZJ, Jiuzhoujiang river estuary; BSH, Baishahe river estuary; NKJ, Nankangjiang river estuary; NLJ, Nanliujiang river estuary; DFJ, Dafengjiang river estuary; QJ, Qinjiang river estuary; MLJ, Maolingjiang river estuary; FCJ, Fangchengjiang river estuary; XMJ, Ximenjiang river estuary.

FastTree v2.1.10 to infer an initial guide tree. Both trees contain non-parametric bootstrap support values. The tree was viewed and annotated using Itol ${ }^{24}$ (https://itol.embl.de).

## Data Records

The raw sequence data are available on the NCBI Sequence Read Archive (PRJNA730330) ${ }^{25} .500$ MAGs, the genome trees are available in figshare ${ }^{26}$. They have been appropriately specified in the text where required.

## Technical Validation

To validate the completeness and contamination of the genomes, we accessed the number of marker genes present in all MAGs using CheckM v1.1.3 ${ }^{27}$ (checkm lineage_wf-tab_table -g -x faa -e 1e-10-1 0.7). It provides robust estimates of genome completeness and contamination by using collocated sets of genes that are ubiquitous and single-copy within a phylogenetic lineage. Completeness and contamination scores are estimated by detecting the presence and number of single-copy marker genes in the draft genome. An uncontaminated and complete MAG will have all of these marker genes present just once in the genome. This final catalog comprises of only those genomes that met specific quality thresholds (i.e., completeness $\geq 50 \%$ and contamination $<10 \%$ ) as described in the manuscript. Additionally, to improve the quality (i.e., increasing completion and reducing contamination), the bins were reassembled in metaWRAP.

## Code availability

Custom scripts were not used to generate or process this dataset. Software versions and non-default parameters used have been appropriately specified where required.

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## References

1. Zhu, Y. G. et al. Continental-scale pollution of estuaries with antibiotic resistance genes. Nat. Microbiol. 2, 16270, https://doi. org/10.1038/nmicrobiol.2016.270 (2017).
2. Hutchins, D. A. \& Fu, F. Microorganisms and ocean global change. Nat. Microbiol. 2, 17058, https://doi.org/10.1038/ nmicrobiol. 2017.58 (2017).
3. Wang, J. \& Jia, H. Metagenome-wide association studies: fine-mining the microbiome. Nat. Rev. Microbiol. 14, 508-522, https://doi. org/10.1038/nrmicro. 2016.83 (2016).
4. Kan, J., Suzuki, M. T., Wang, K., Evans, S. E. \& Chen, F. High temporal but low spatial heterogeneity of bacterioplankton in the Chesapeake bay. Appl. Environ. Microbiol. 73, 6776-6789, https://doi.org/10.1128/Aem.00541-07 (2007).
5. Bouvier, T. C. \& del Giorgio, P. A. Compositional changes in free-living bacterial communities along a salinity gradient in two temperate estuaries. Limnol. Oceanogr. 47, 453-470, https://doi.org/10.4319/lo.2002.47.2.0453 (2002).
6. Campbell, B. J. \& Kirchman, D. L. Bacterial diversity, community structure and potential growth rates along an estuarine salinity gradient. ISME J. 7, 210-220, https://doi.org/10.1038/ismej. 2012.93 (2013).
7. Fortunato, C. S., Herfort, L., Zuber, P., Baptista, A. M. \& Crump, B. C. Spatial variability overwhelms seasonal patterns in bacterioplankton communities across a river to ocean gradient. ISME J. 6, 554-563, https://doi.org/10.1038/ismej.2011.135 (2012).
8. Ghosh, A. \& Bhadury, P. Exploring biogeographic patterns of bacterioplankton communities across global estuaries. MicrobiologyOpen 8, https://doi.org/10.1002/mbo3.741 (2019).
9. Zhang, C. J., Chen, Y. L., Pan, J., Wang, Y. M. \& Li, M. Spatial and seasonal variation of methanogenic community in a river-bay system in South China. Appl. Microbiol. Biotechnol. 104, 4593-4603, https://doi.org/10.1007/s00253-020-10613-z (2020).
10. Yu, T. et al. Characteristics of Microbial Communities and Their Correlation With Environmental Substrates and Sediment Type in the Gas-Bearing Formation of Hangzhou Bay, China. Front. Microbiol. 10, https://doi.org/10.3389/fmicb.2019.02421 (2019).
11. Zhou, L. et al. Stochastic determination of the spatial variation of potentially pathogenic bacteria communities in a large subtropical river. Environ. Pollut. 264, 114683, https://doi.org/10.1016/j.envpol.2020.114683 (2020).
12. Zhou, L. et al. Environmental filtering dominates bacterioplankton community assembly in a highly urbanized estuarine ecosystem. Environ. Res. 196, 110934, https://doi.org/10.1016/j.envres.2021.110934 (2021).
13. Li, D., Liu, C. M., Luo, R., Sadakane, K. \& Lam, T. W. MEGAHIT: an ultra-fast single-node solution for large and complex metagenomics assembly via succinct de Bruijn graph. Bioinformatics 31, 1674-1676, https://doi.org/10.1093/bioinformatics/btv033 (2015).
14. Mikheenko, A., Saveliev, V. \& Gurevich, A. MetaQUAST: evaluation of metagenome assemblies. Bioinformatics 32, 1088-1090, https://doi.org/10.1093/bioinformatics/btv697 (2016).
15. Uritskiy, G. V., DiRuggiero, J. \& Taylor, J. MetaWRAP-a flexible pipeline for genome-resolved metagenomic data analysis. Microbiome 6, https://doi.org/10.1186/s40168-018-0541-1 (2018).
16. Prjibelski, A., Antipov, D., Meleshko, D., Lapidus, A. \& Korobeynikov, A. Using SPAdes de novo assembler. Curr. Protoc. Bioinf. 70, e102, https://doi.org/10.1002/cpbi. 102 (2020).
17. Olm, M. R., Brown, C. T., Brooks, B. \& Banfield, J. F. dRep: a tool for fast and accurate genomic comparisons that enables improved genome recovery from metagenomes through de-replication[J]. ISME J. 11, 2864-2868, https://doi.org/10.1038/ismej.2017.126 (2017).
18. Uritskiy, G. et al. Halophilic microbial community compositional shift after a rare rainfall in the Atacama Desert. ISME J. 13, 2737-2749, https://doi.org/10.1038/s41396-019-0468-y (2019).
19. Patro, R., Duggal, G., Love, M. I., Irizarry, R. A. \& Kingsford, C. Salmon provides fast and bias-aware quantification of transcript expression. Nat. Methods 14, 417-419, https://doi.org/10.1038/nmeth. 4197 (2017).
20. Chaumeil, P. A., Mussig, A. J., Hugenholtz, P. \& Parks, D. H. GTDB-Tk: a toolkit to classify genomes with the Genome Taxonomy Database. Bioinformatics 36, 1925-1927, https://doi.org/10.1093/bioinformatics/btz848 (2020).
21. Parks, D. H. et al. A complete domain-to-species taxonomy for Bacteria and Archaea. Nat. Biotechnol. 38, 1079-1086, https://doi. org/10.1038/s41587-020-0501-8 (2020).
22. Price, M. N., Dehal, P. S. \& Arkin, A. P. FastTree 2-approximately maximum-likelihood trees for large alignments. PloS one 5, e9490, https://doi.org/10.1371/journal.pone. 0009490 (2010).
23. Nguyen, L. T., Schmidt, H. A., Von Haeseler, A. \& Minh, B. Q. IQ-TREE: a fast and effective stochastic algorithm for estimating maximum-likelihood phylogenies. Mol. Biol. Evol. 32, 268-274, https://doi.org/10.1093/molbev/msu300 (2015).
24. Letunic, I. \& Bork, P. Interactive Tree Of Life (iTOL) v5: an online tool for phylogenetic tree display and annotation. Nucleic Acids Res. 49, W293-W296, https://doi.org/10.1093/nar/gkab301 (2021).
25. NCBI Sequence Read Archive https://identifiers.org/ncbi/insdc.sra:SRP320016 (2021).
26. Zhou, L., Huang, S., Gong, J., Xu, P. \& Huang, X. 500 metagenome-assembled microbial genomes from 30 subtropical estuaries in South China. Figshare https://doi.org/10.6084/m9.figshare.14717061.v4 (2021).
27. Parks, D. H., Imelfort, M., Skennerton, C. T., Hugenholtz, P. \& Tyson, G. W. CheckM: assessing the quality of microbial genomes recovered from isolates, single cells, and metagenomes. Genome Res. 25, 1043-1055, https://doi.org/10.1101/gr.186072.114 (2015).

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## Author contributions

Methodology: Lei Zhou, Shihui Huang, and Jia-yi Gong; Writing - original draft: Lei Zhou and Xian-de Huang; Funding acquisition: Lei Zhou, Peng Xu, and Xian-de Huang; Writing - review \& editing: Peng Yu.

## Competing interests

The authors declare no competing interests.

## Additional information

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