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ABSTRACT

Iron rich hot springs could represent analogue environments to early Earth's ferruginous oceans. To investigate the geochemical drivers of microbial community structure and metabolic potential I performed read-based, and metagenome assembled genome (MAG) analyses on sediment samples from five iron-rich hot springs of Japan. The study revealed the reductive pentose phosphate cycle as the most abundant carbon fixation pathway, widespread iron oxidation-reduction, and microbial groups showing high metabolic plasticity involved in the nitrogen and sulfur cycles. Temperature and pH were important drivers of variation. Reconstructed MAGs belonged to 63 previously unreported genera, revealing underexplored taxonomic novelty. A machine learning approach to test the feasibility of using microbial and pathway abundances to predict geochemical parameters in hot springs revealed that only some parameters, can be predicted with accuracy, and that thermophilic microorganisms are important prediction features.

Summary

The Precambrian Supereon, which includes the Hadean, Archean, and Proterozoic eons, represents nearly 85% of Earth's history (Stüeken et al., 2020) but is the period we know the least about due to a sparse rock record. While microorganisms were the dominant life forms on early Earth, much about their physiology and ecology remains unknown. The geochemical context in which organisms survived included an atmosphere with less than 0.1% oxygen compared to current levels and high carbon dioxide, along with oceans rich in dissolved ferrous iron (Catling, 2011; Holland, 1986; Lyons et al., 2014; Planavsky et al., 2011; Stüeken et al., 2020). Other factors such as temperature, pH, salinity, and nutrient availability remain uncertain (Crowe et al., 2014; Fakhraee Katsev, 2019; Krissansen-Totton et al., 2018; Watanabe et al., 2024). With the geochemical constraints we have on Precambrian oceans we can make some inferences on the availability of electron donors and acceptors microbes could have used for energy conservation, but availability does not necessarily mean utilization, and function does not necessarily mean phylogeny. There are still many unknowns about the biosphere of the Precambrian, its connection with the environment and the possible ways we can interpret the geological record. One possible strategy to use the current knowledge on Precambrian environmental conditions and gain insights into the biosphere present at the time, is the use of analogue environment studies. Studied on analogue sites can help us elucidate the material cycling and microbial community structure of Precambrian-like environments and test our geochemical and microbial physiology models and hypotheses. Availability of environments in which soluble iron is present is limited, iron-rich hot springs represent unique opportunities to study microbial communities in such conditions. The focus of this study is CO₂-buffered, naturally occurring, aerial, ferrous iron-rich hot springs. In this work, I use metagenomic techniques to analyze the metabolic potential, material cycling and species-level contribution of Bacteria and Archaea to the biogeochemical cycles in iron rich hot springs of Japan that represent analogues to Precambrian oceans. In Chapter 1, I summarize the current knowledge on Archean environmental conditions, the transition to the Proterozoic marked by the Great Oxidation Event (GOE) and putative roles of microbial communities to the material cycling on early Earth based on modern analogue environments. In Chapter 2, I discuss the metabolic potential of microbial communities in five iron-rich hot springs of Japan (Jinata, Okuokuhachikuro (OHK), Furutobe, Tsubakiyama, Kowakubi) studied with read-based metagenomic sequencing techniques and discuss the findings in the context of amplicon sequencing-based microbial abundance and geochemical conditions. The study reveals that the reductive pentose phosphate pathway is the main carbon fixation pathway in all springs, but performed by different organisms, namely Cyanobacteria, Gammaproteobacteria, and Zetaproteobacteria, no name a few; with the exception of

Kowakubi hot spring, in which the reductive Acetyl-CoA pathway dominated. Salinity and temperature were strong drivers of microbial community structure, whereas salinity, pH and temperature were the main drivers of metabolic potential, with highest temperatures showing more abundance of C-1 compound utilization pathways. Oxygenic photosynthesis and anoxygenic photosynthesis were negatively affected by higher dissolved organic and inorganic carbon as well as temperature, suggesting most phototrophic microorganisms are hindered by these conditions. In Chapter 3, I refine the metagenomic data to reconstruct metagenome assembled genomes (MAGs) to discern the species-specific contributions to the geochemical cycles. The reconstructed 242 MAGs belonged to diverse phyla and include 185 new species. The study revealed that most organisms are able to utilize several electron donors for energy conservation such as iron, sulfide, thiosulfate and organic carbon. In regard to iron, some organisms were able to perform both iron oxidation and reduction, and most of them encoded iron regulation genes, highlighting the importance of iron in the nutrient cycling. The Cyanobacteria MAGs revealed their potential for anoxygenic photosynthesis, further expanding their role in the springs. Several diazotrophs were found from several clades including the Cyanobacteria, Desulfobacterota, Firmicutes and Myxococcota, to name some. Metabolic plasticity is widespread. In Chapter 4, I describe the results of a feasibility study on the use of machine learning (ML) random forests to predict geochemical conditions in a data set of 925 hot springs from New Zealand, obtained from (Power et al., 2018). Microbial community structure at the species level and family level was compared to predicted metabolic pathways as input features on model accuracy. Species level data showed the highest accuracy amongst all datasets for predicting pH and Temperature with an accuracy of around 80% and 60%, respectively. Micronutrient concentrations were poorly predicted by either dataset. The present dissertation highlights hot springs as natural laboratories to test hypotheses on extreme environments on Early Earth, it allowed us to comprehensively understand the carbon, nitrogen, iron and sulfur cycles in Precambrian analogues and the contributions of species to them. Future work should focus on experimentally constraining the rates of assimilation of nutrients with the metabolic pathways uncovered in this study to obtain quantitative values of material cycling.

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