

Google Summer of Code 2021
CODING PROJECT PROPOSAL
From DNA sequences to metabolic interactions: building a
pipeline to extract key metabolic processes

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1 Synopsis

Twenty years after Covert W.Covert, Bernhard O. Palsson and their colleagues published their review on metabolic modelling of microbial strains [4], the value of this method has been well established.

From the beginning, metabolic modeling has been interwoven with constraining-based methods [12]. The value of randomized sampling in the framework of metabolic modeling has been proved itself over the years [17, 9].

High Throughput Sequencing technologies have allowed process in the genetic information (DNA) in a cost-efficient and easy way, especially for microbial species as their genome is only but a few hundreds of genes long.

Once the complete genome of a species is obtained, the complete reconstruction of the metabolic network of the species is enabled, called genome-scale metabolic models (GEMs) [8].

Such models for all the species present in a microbial community, allows the study of metabolic interactions, thus an insight for the actual microbial interactions [15].

Aim of this project is to integrate the produced data and knowledge of these twenty (and more) years and make use of the randomized flux sampling method to evaluate the metabolic interactions retrieved. To this end, thousands of publicly available reference microbial genomes will be selected, and their automatic metabolic network reconstructions will be implemented. Based on these models, cross-feeding interactions algorithms will be performed for groups of species to extract key metabolic processes. New functions, implementing the recently developed Multiphase Monte Carlo Sampling (MMCS) approach [3] in the framework of the `dingo` library, will make use of the randomized flux sampling concept to evaluate the processes retrieved.

2 The Project

2.1 Background

Microbial communities populate most environments on earth; from the seafloor to the human gut, they literally live everywhere [16]. They play a critical role in shaping the environment as we know it. By driving biogeochemical cycles, bacteria, along with geochemical (abiotic) transformations (atmospheric, tectonic and geothermal), shape Earth's climate [6]. At the same time, *the human body is inhabited by millions of tiny living organisms* having a fundamental part in keeping us healthy [5]. However, up to now scientists are able to cultivate approximately 1% of known Bacteria [19]. Since the growth of various bacteria depends on their interactions with others [22], inferring microbial interactions would strongly support cultivating taxa for the first time, allowing the production of secondary metabolites and their biotechnological applications. At the same time, it would be an essential tool to further expand our understanding regarding the underlying mechanisms governing a range of phenomena, from ecosystem functioning to human health disorders. This is why researchers from all these scientific fields have been studying the community structure of the microbial communities of their interest, focusing both on the taxa present and the metabolic processes (referred as *functions*).

Metabolism is a network of the metabolic pathways that occur in an organism; thus, a metabolic network is the representation of all these pathways [12]. Using the stoichiometry of each reaction, which is always the same in the various species, we convert the metabolic network of an organism into a mathematical model [12]. High Throughput Sequencing has made access to the complete genomic information of an organism rather easy. However, building the complete metabolic network of a species is not that trivial yet [20]. However, over the last few years, automatic reconstruction approaches for building genome-scale metabolic models [10] of relatively high quality have been developed.

The study of metabolic dependencies to infer interspecies microbial interactions has been established the last years (Fig. 2.1), for more see [24].

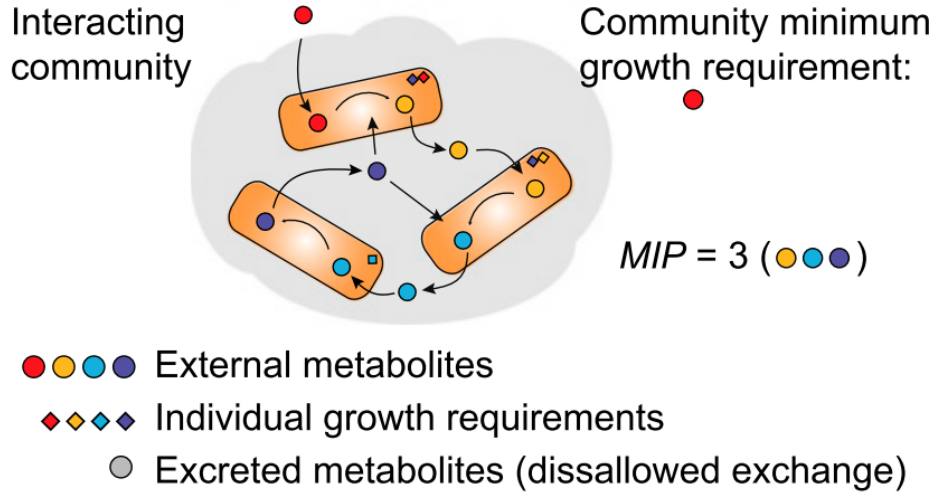


Figure 1: Representation of a microbial community of 3 interacting species. The Metabolic Interaction Potential metric (MIP) quantifies the propensity of a community to exchange metabolites. The figure is part of Fig. 3. at [24]

Constrained-based modelling approaches, such as as Flux Balance Aanalysis and flux randomized sampling, have enabled the investigation of the properties of the possible steady states of a metabolic network, end up with essential insights on metabolism at every level [11, 21]. Microbial communities harbor metabolically interdependent cooperative groups. As a matter of fact, such groups play a key role for species co-occurrence [24].

The aim of this project is to bring together data (reference microbial genomes), automatic genome scale metabolic networks reconstruction tools, cross-feeding interactions algorithms and flux randomized sampling. This way, we will enable the evaluation of the effect of each of the predicted metabolic interactions, to the various species of the community.

2.2 Methodology

This project will make use of third party software tools and will also develop some new functions in the framework of the **dingo** project.

Global repositories including thousands of reference microbial genomes such as the Genome Taxonomy Database (GTDB) will be exploited [14, 13].

Automatic genome scale metabolic networks reconstruction tools, such as AuReMe [1] and CarveMe [10], have been recently developed. Such tools will be implemented to get the corresponding GEMs for the genomes retrieved and/or for the genomed provided by the user.

The SMETANA package, is a set of algorithms looking for possible cross-feeding interactions between the species of the microbial community under study [24]. Such algorithms will be implemented in the GEMs built from the original genomes gathered or/and provided by the user, to return key metabolites.

Flux randomized sampling using the MMCS algorithm of the **dingo** Python library will be implemented to evaluate the effect of the metabolites returned from the previous steps. To this end, new functions

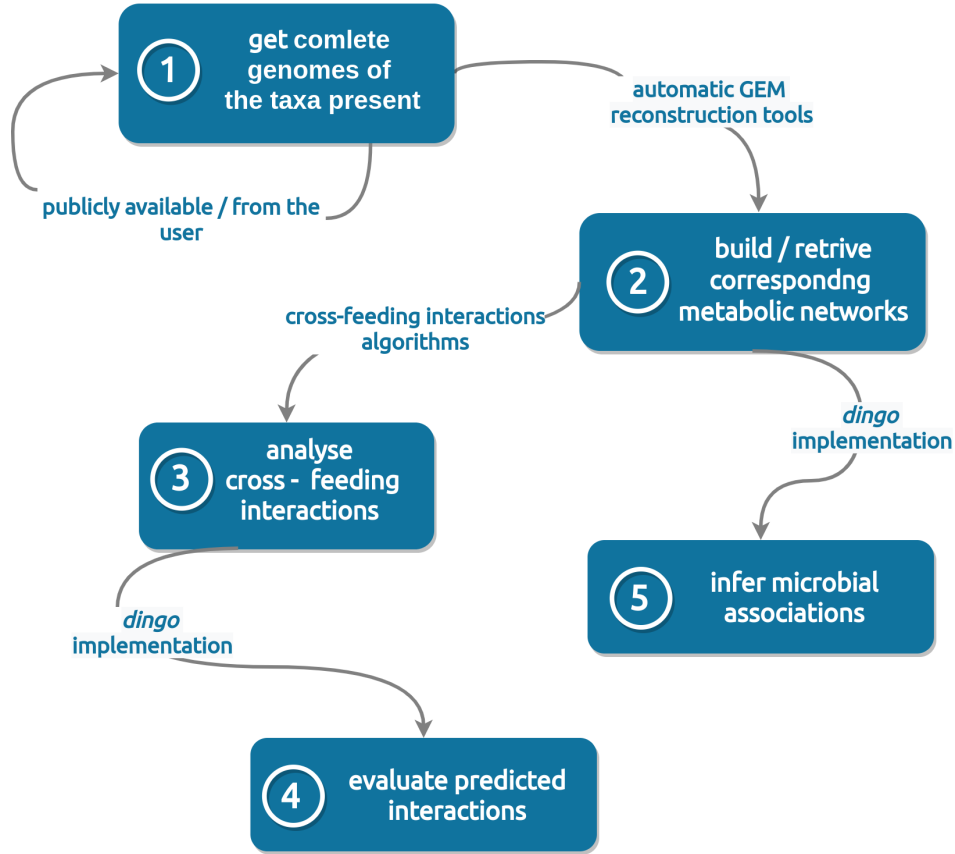


Figure 2: Project’s workflow. This proposal consists of 4 major steps combining third party tools implementation and development of new code in the framework of the **dingo** Python library.

will be developed in the framework of the **dingo** library to check the response of the biomass function [7] or the function of maximization of ATP per unit flux [18] in alterations regarding the corresponding metabolites. Furthermore, functions will be developed to perform sampling in the flux spaces of all the paired associations between the various species, to strengthen these associations, depending on whether they respond robustly on exchange/competition for metabolites.

2.3 Benefits to the Community

The proposed project will benefit the most the **dingo** Python library as an open source code project, as it will provide a thorough use case in a big-data scale and will also develop further functions for the analysis of microbial communities.

Furthermore, it will be essential for biologists studying microbial interactions in all the different biological fields, from ecosystem functioning to precision medicine.

3 Deliverables

The tasks of this project will be split in the two main periods of the project; i.e. the Community Bonding period between April 27 and May 18 (3 weeks) and the Coding period between May 18 and

August (10 weeks). At the end of the coding period, all the implementations described in the Tasks chapter will have been completed.

VolEsti provides a great basis for the implementation of this project. Therefore, a limited amount of time is needed to set up all the necessary coding environments. A number of extra rounding algorithms will be added on the **VolEsti** package in the next few weeks; thus, working with members of the **VolEsti** group would allow us for a simultaneous development of the source code and the interface, without unexpected delays. That said, the coding tasks could start before the bonding period is over. Furthermore, a blog will be created to report the project's progress periodically and a thorough documentation for the **VolEsti** Python interface will be delivered. The **VolEsti** package is under the LGPL-3 license, while the **cobrapy** package is under the GPL and **cobra-toolbox** under the GNU General Public License version 3.0; all open-source. Thus, I intend to give the **VolEsti** Python interface the LGPL-3 license too.

3.1 Bonding period (May 17 - June 7)

During the bonding period everything that I need to know regarding the implementation of T1 will be fully covered. Thus, the source code of **VolEsti** will be thoroughly studied and a step-by-step guide for building the Python interface will be delivered by the end of the second week of the bonding period. In its last week, everything that might be needed to be able to start coding by the end of the bonding period, will be taken care of. Thus, I will build a branch of the current **VolEsti** project at my GitHub repository [18] and a blog to report my progress periodically will be also created in my web page*. The first two weeks will require a strong effort from me in comprehending the architecture of the **VolEsti** source code. To address this, the mentors' guidance will be essential during this time. In the third week of the bonding period, it is my intention to be ready to start coding. I do not count it as a coding week, however I believe I will be ready.

3.2 Coding period (June 7 - August 16)

This 13 weeks period will be split as shown in the following Gantt chart (Figure 3). More specifically:

- **1st - 4th week**

In the first four weeks probably the most time-consuming task will be implemented. I will create the wrapper for the **volesti** package as described in T.1. Meanwhile, I will catalogue all the useful open-cobra visual components and I will include them in this first Python interface of **volesti**. Tests that the wrapper is operating well will be implemented. A first documentation of the Python interface will also be made.

- **5th - 7th week**

In those four weeks, I am going to implement the **cplex** buildings; first for the case of **volesti** source code (C++ interface of the **cplex** package). Tests will be implemented as well.

- **8th - 9th week**

And then, the 2 following weeks for the case of the Python wrapper built on T.1. A thorough report about the coding blocks regarding the “**cplex**” building will be delivered. 10th week By the time being, the **GeomScale** group will have added the extra sampling and rounding methods on the **volesti** source code. Thus, this week I will build those methods based on the **cplex** package as well.

- **11th - 12th week**

In those two weeks, the final wrapper of the **volesti** Python interface will be implemented. This version will include the **cplex**-based methods of **volesti** both those that are currently included and those that

are about to be added by the GeomScale group and the open-cobra visual components mentioned.

- **13th week**

In the last week, the final volesti Python interface will be applied to a series of data as described in T.4. Likewise, the cobrapy library. Their performance will be reported. Finally, a complete documentation for the volesti Python interface will be delivered. This time schedule is a worst-case scenario. In case that something is not going according to the plan, then the extra time that we attempt to earn from the bonding period, will come up as a backup plan.

4 Related work

I have been a member of the GeomScale project over the last year. Based on an idea from the GSoC of 2020 we have been working on sampling the flux space of metabolic networks. I contributed with wrapping the C++ code of `VolEsti` to build the Python interface. Furthermore, I implemented the MMCS method developed by the GeomScale group on metabolic networks of high dimensions [3]. This work was accepted in the proceedings of the 37th Symposium on Computational Geometry.

5 Tests

This proposal is not among the ideas listed in the table of proposed coding projects of the GeomScale group. As the scope of my proposal is close to the Inferring microbial interactions project, and after contacting the mentors, I implemented the tests of the latter project in the framework of my proposal.

You can find my answers on this link and the correspondig code on its correspondig GitHub repository.

6 Biographical information

6.1 Education

- PhD candidate at University of Crete (2018 - ongoing). Dissertation on: “Merging NGS data, knowledge aggregation and data integration techniques, along with ecological network analysis (ENA): an attempt to decipher microbial community ecology and ecosystem functioning by taking advantage of the hypothesis-generating method”.
- MSc in Bioinformatics at the University of Crete (2016 - 2018). Thesis: “eDNA metabarcoding for biodiversity assessment: Algorithm design and bioinformatics analysis pipeline implementation”
- BSc in Biology at the National and Kapodistrian University of Athens (2010 - 2016). Thesis: “Morphology, morphometry and anatomy of species of the genus *Pseudamnicola* in Greece”

6.2 Publications

- Zafeiropoulos, Haris, Anastasia Gioti et al. 0s and 1s in marine molecular research: a regional HPC perspective (**under review in GigaScience journal**)
- Zafeiropoulos, Haris, Anastasia Gioti et al. (2021, April 5). The IMBBC HPC facility: history, configuration, usage statistics and related activities (Version 1.0.0). Zenodo. DOI: 10.5281/zenodo.4665308
- Polymenakou, Paraskevi N., et al. "The Santorini Volcanic Complex as a Valuable Source of Enzymes for Bioenergy." *Energies* 14.5 (2021): 1414. DOI: 10.3390/en14051414

- Chalkis, Apostolos, et al. "Geometric algorithms for sampling the flux space of metabolic networks." arXiv preprint arXiv:2012.05503 (2020) <https://hal.inria.fr/hal-03047049v2>
- Zafeiropoulos, Haris, et al. "PEMA: a flexible Pipeline for Environmental DNA Metabarcoding Analysis of the 16S/18S ribosomal RNA, ITS, and COI marker genes." GigaScience 9.3 (2020): gaaa022.. DOI:10.1093/gigascience/gaaa022

6.3 Programming

- Author of the PEMA workflow
- Very experienced in programming with scripting programming languages (Python, BigDataScript, R).
- Basic knowledge of C++
- Very experienced in container-based technologies (Docker, Singularity)
- Experienced in web

6.4 Personal motivation

Bioinformatics play a great part in almost every aspect of modern biology. That is why after graduating Biology school (BSc), I focused on computer science and Bioinformatics (MSc). I am now quite experienced with a series of scripting programming languages (Python, BigDataScript, R) and with container-based technologies (Docker, Singularity). PEMA [23], a pipeline for the analysis of metabarcoding data, was my first coding project; PEMA has now been published and selected from LifeWatch - ERIC a European Research Infrastructure, to support . PREGO and DARN are ongoing bioinformatics projects in the framework of my PhD.

My research interests focus on ecology and ecosystem functioning at the microbial dimension. As Systems Biology approaches can benefit the most this field, I have spent the last 2 years working on knowledge aggregation and data integration techniques as well as networks analysis are employed.

The last year, I have been working with the GeomScale group on the VolEstipy project and sampling on the flux space of metabolic networks

This GSoC project would allow me to continue working with in the framework of the GeomScale project whilst building the proposed application would benefit me the most, especially as it will allow me to extend the scopus of my PhD dissertation.

Metabolic interactions [2] is now the next big thing in systems biology. In such a study, the changing phenotype of an ecosystem, as derived by the combination of phenotypes of all the ecosystem's individual entities, under different scenarios of constraint values could be predicted. The impact of such studies would be more than significant, especially on a time when nature suffers the most, mostly due to anthropogenic activities. Indicatively possible datasets such a study could be conducted on: "Impact of exogenous nitrogen on the cyanobacterial abundance and community in oil-contaminated sediment: A microcosm study" from Wang et al. [19] (peer-reviewed, no public data, email request) "Discovery of functional gene markers of bacteria for monitoring hydrocarbon pollution in the marine environment - a metatranscriptomics approach" from Knapik et al. [20] (not peer-reviewed, data not published yet) In their study [21], Gossart et al. describe the research framework of such studies we suggest to apply the aforementioned methods .

Coding has been a great part of my everyday routine through the recent years. The different needs of the different types of analysis in biology force you to deal with a great range of computing tasks, such as choosing the proper programming language for each issue, working with High Performance Computing

environments, finding ways to make your code easy-to-use, easy-to-distribute and flexible at the same time. It would be both a profit and an enjoyment for me to be part of the Google Summer of Code and upgrade my so-far programming skills, particularly as `VolEsti` functions have been developed in C++, meaning that I will have to deal with new personal coding challenges.

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