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# Delivering the goods: an environmental survey of secondary metabolites and the microbiome.

# Intro:

The intensity and types of environmental challenges facing plants vary at all spatial scales, from landscape-level scales down to variability among different regions of individual plant bodies. Temporal changes in environmental challenges to plants also range from hourly, to seasonal or even decadal fluctuations in climate, pathogen loads, nutrient availability, and exposure. Plant genomes contain many solutions to these challenges, especially to environmental challenges which have acted as long-term selective pressures on plants. In these times of rapid change, however, the abundance and rythmn of all types of plant stressors has increased, and plant genomes alone may not evolve quickly enough to ensure the success of ecologically or socially important plants in the face of increased pathogen loads, decreased climatic stability, and decreased habitat availability.

Some have proposed that the plant microbiome, with its collective abundance of microbial genetic resources, may represent an important and adaptive addition to the host plant genome. Together, it is proposed, the microbiome and the plant sum to make up the plant “hologenome”, which is more extensive and more dynamic than the plant genome alone1. It remains a mystery how historically important hologenomes were in shaping plant populations, even prior to the current era of rapid anthropogenic change. However, evidence has mounted rapidly that microbiomes are important determinants in plant distributions and ecosystem health2–6. If microbiomes are to harnessed in any way to help plant health in times of rapid change, their spatial and temporal patterns must be better understood.

Here we propose to undertake sampling for the functional microbiome of a host tree species, *Eugenia nesiotica*, across the extent of the BCI ForestGEO plot. Endophytes from both leaf and root tissue of *Eugenia* will be surveyed for the presence of biosynthetic gene clusters, using two survey methods, targeted PCR tests and metabarcoding of two groups of important enzymes for backbone molecules for secondary metabolites, PKS and NRPS7.

## Hypothesis and predictions:

**Hypothesis**: We hypothesize that plants recruit microbial symbionts that supplement or compliment plant genome function to enhance tolerance to stressors.

**Prediction:** Stressors faced by the host plant will predict the relative abundance of “useful” secondary metabolites produced by the microbiome. For instance, we predict that tissues of a drought-stressed tree will contain increased levels osmolytes and antioxidants of microbial origin.

# Methods:

### PKS and NRPS metabarcoding

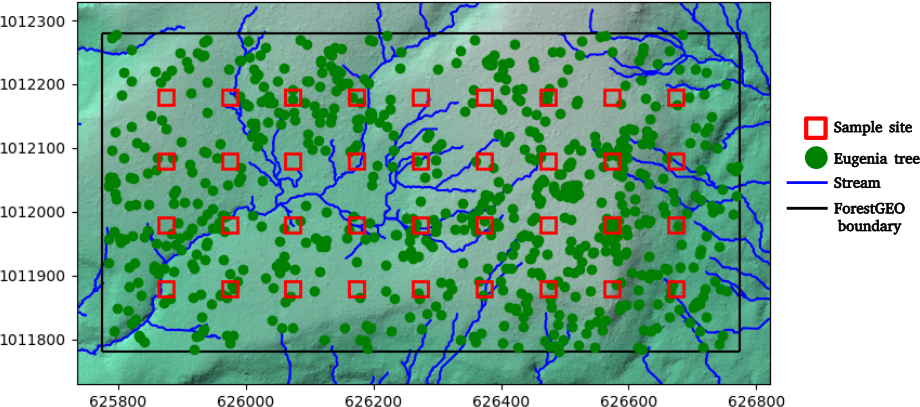
A metabarcode study will broadly target polyketide synthases (PKS) and non-ribosomal peptide synthases (NRPS) with degenerate primers. These data will using be engaged to to get a general picture of some broad classes of secondary metabolites that tree microbiomes are predicted to be producing differentially in response to environmental challenges. These enzymes encode backbone molecules for two of the broadest families of microbial secondary metabolites (polyketides and non-ribosomal peptides)7,8, and exist in diverse homologs that can be used to infer medium-to-fine grain portrait of the secondary metabolome of the plant microbiome. Amplicon library preparation will follow Thomas et al.9, for sequencing on Illumina MiSeq platform. Primer pairs for this survey will be based on Charlop-Powers et al.10. Preliminary surveys of these primers on public databases show very few matches to plant sequences (see [jupyter notebook here](https://nbviewer.jupyter.org/github/danchurch/BCIgeneClusterSurveyIdeas/blob/main/BCIprimerhunts.ipynb)). Though most observations of NRPS and type I PKSes originate from microbial sequences, the possibility exists that that plants are greatly undersampled for type I PKS gene clusters11. Since our hypothesis focuses on the complementary role of the microbiome, we will attempt enhance the PKS- and NRPS-targeting primers with plant-blocking oligonucleotides12.

### Direct metabolite PCR tests

Some key metabolites may not be detected in our metaboarcode survey, either because they are not synthesized using a NR-peptide or polyketide backbone, or because of sampling error. Our metabarcode survey will be augmented with more targeted PCR surveys for the presence of biosynthetic pathways of select microbial products. We will research and develop primer sets for detection of key metabolic steps in the microbial (non-plant) biosynthesis of:

1. Anti-herbivory response: ergot alkaloids13–15 and rugulosins16–18
2. Drought stress response: microbial-origin antioxidants superoxide dismutase19,20 and catalase21,22, microbial-origin osmolyte proline23–25
3. Pathogen response: fungal anti-mycotic compounds echinocandins8,26, polyenols27–29, and strobilurins30,31.

Existing literature and public databases will be use to develop pathway-specific primers each of these enzyme synthesis pathways.



### Host tree:

Subject tree species *Eugenia nesiotica* was selected for its medium population density on the BCI forest dynamics plot (see map), for its history of research at BCI32,33, and for its close relation to economically important species *E. uniflora*, whose transcriptome and genome are subject to ongoing investigation34,35.

### Sampling scheme:

In both survey methods, spatial analysis of the presence of “useful” metabolites produced by plant-associated microbes will be coupled with micro-site data, estimates of disease and herbivory load, and the proximity of conspecific trees. Leaf and root tissue will be collected at 36 locations within the plot according to a spatial grid with a grain of 100m. Epiphyte and endophyte microbial populations will be sampled. Aerial plant tissues will be sampled at upper and lower canopy sites, to test micro-site variability in microbiome expression due to position in the canopy. Three paired samples will be collected from each of three zones of an individual tree, which then will be pooled by zone and by community (endophyte or epiphyte) totaling to 6 genomic DNA samples per tree.

### Sample preparation:

Roots will be gently brushed clean of soil. To sample epiphytes of leaves and roots, leaves and root sections will submerged in 20 ml of 0.1% tween / sterile water mixture, and vortex-ed for 5 minutes. This wash fluid will be centrifuged, and the pellet frozen for later DNA extraction. Root and leaf endophytes sampling will follow Thomas et al.9, using a bleach-based surface sterilization method, followed by tissue homogenization and DNA extraction. Bulk DNA will be purified from host plant tissue and epiphyte wash-pellets using column kits, and selectively PCR-amplified with above PKS and NRPS primers (with added sample barcodes and Illumina platform adapters). Sequences will be sequenced using a Illumina MiSeq platform sequencer. Plant pathogen loads and damage from herbivory will be phenotyped from photos36 of 10 randomly selected from leaves from upper and lower zones of the tree, in addition to photos of leaves harvested for endophyte and epiphyte sampling. 50 grams of soil from each tree will also be harvested for soil moisture estimates.

### Bioinformatic and Statistical Analyses:

Metabarcoding results will be subject to standard bioinformatic pipelines to reduce sequencing artifacts and group highly similar reads37. Following this, amplicon reads will be evaluated and grouped using closest-known biosynthetic gene clusters using the software package eSNaPD10,38 and public databases. The resulting gene-cluster-by-sample-site matrix will be examined standard numerical ecology-based methods for modeling large matrices as dependent objects of environmental data (“regression” of metabarcoding results against environmental data and spatial patterns using redundancy analyses, including MEM and variation-partitioning methods)39–42. Read libraries will also be mined for specific biosynthetic pathways of interest (alkaloids, antibiotics, antioxidants, etc), to the level of specificity allowed by the homology of the polyketide and peptide backbones. and these will be treated as point patterns43, along with the results of the targeted PCR tests. Point patterns of secondary products of interest will analyzed for co-occurrence with environmental challenges, herbivory, pathogen loads, and proximity and size of conspecifics and cluster size of nearest conspecifics.

## Conclusion: pertinence to research on-going at Smithsonian Institute (SI) and Smithsonian Tropical Research Institute (STRI):

One of the greatest contributions to tropical ecology in recent memory by SI and STRI is the empirical testing of the Janzen-Connell hypothesis of conspecific negative density dependence (CNDD), and exploration of the mechanisms behind CNDD44–47. Dr. Allen Herre and colleagues at STRI found that soil-born enemies may be main culprits in the negative density dependence patterns of rare trees in the tropics, but noted that unlike rare trees, highly abundant trees do not show patterns of CNDD32. Insufficient gene flow and genetic variation in the genomes of rare plants trees has been proposed as one mechanism for CNDD33.

The project here proposed suggests another, complementary mechanism - if functional microbiomes are key to pathogen resistance, **then recruitment of beneficial microorganisms into the microbiome may function as a positive-density-dependent feedback.** Dr. Herre and colleagues have observed that many important “horizontally-acquired” microbes can be transmitted to seedlings through litter of older trees48. Abundant trees may therefore be able to transmit their microbiomes to seedlings more successfully than rare trees simply as a function of volume of leaf litter, a positive-density-dependent feedback in seedling success. Clusters of abundant species may represent larger “target” areas not only for pathogens but for environmental recruitment of beneficial microbial partners. Dr. Melissa McCormick at STRI and her colleagues found hypersensitivity by host plants to availability of symbiotic soil fungal partners on the landscape49. This indicates that the availability of mutualistic symbioses may be important to overcoming the inhibitory effects of soil-born enemies and the negative density: both plant and fungal mycorrhizal partners presumably benefit by the presence of the other, a positive density feedback. However, McCormick and colleagues have also examined patterns in taxonomic diversity in fungal endophytes on a scale similar to the study proposed here, and found that taxonomic diversity patterns in fungal microbiome don’t necessarily correlate positively with host post plant diversity50. This and other studies51 indicate that classical taxonomic approaches to microbiome study need to be augmented by *functional* surveys of the microbiome such as here proposed.

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