

Using Mass Spectrometry to Quantitate and Analyze Bioactive Small Molecules

BY

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LIST OF ABBREVIATIONS

^{13}C	Carbon-13
ACN	Acetonitrile
AUC	Area under the curve
CHCA	alpha-cyano-4-hydroxycinnamic acid
DHB	2,5,-dihydrobenzoic acid
DMEM	Dulbecco's Modified Eagle Media
ESI	Electrospray ionization
GC	Gas chromatography
GMF	Grouse's Modified Media Freshwater
GNPS	Global natural products social molecular networking
HPLC	High performance liquid chromatography
IMS	Imaging mass spectrometry
ISP-2	International Streptomyces Project-2
IT	Ion trap
LC	Liquid chromatography
LOD	Limit of detection
m/z	mass to charge ratio

LIST OF ABBREVIATIONS (continued)

MALDI Matrix-assisted laser desorption/ionization

MAN Metabolite association network

MCX Mixed-mode cation-exchange

MeOH Methanol

MFA Metabolite flux analysis

MRM Multiple reaction monitoring

MS Mass spectrometry

MS¹ First level mass spectrometry

MS² Tandem mass spectrometry

MS/MS Tandem mass spectrometry

NAP Network Annotation Propagation

NMR Nuclear magnetic resonance

PBS Phosphate buffered solution

QqQ Triple quadrupole

rRNA Ribosomal RNA

SNF Simple Nutrient Media Freshwater + Fiber

SPE Solid phase extraction

LIST OF ABBREVIATIONS (continued)

SRM	Single reaction monitoring
TFA	Trifluoroacetic acid
TOF	Time of flight mass analyzer
V	Volts

SUMMARY

Metabolomics is the study of small molecules in a biological system. Metabolomics encompasses a wide range of research questions, and therefore requires the use of varied workflows. Chapter 1 is an overall introduction to targeted and untargeted metabolomics, as well as what kinds of experiments fall under each category. The chapter also serves as an introduction to mass spectrometers in the context of their metabolomics applications.

Chapter 2 applies fundamental targeted metabolomics techniques to a multiple reaction monitoring assay of catecholamines in cell media to investigate the role of catecholamines as chemical signals in the primary metastasis of epithelial high grade serous ovarian cancer. This chapter reviews the necessary components of a working assay, with a focus on optimizing sample preparation for translating from an untargeted experiment to a targeted assay.

Chapter 3 details the process of cultivating a unique mammalian-sourced microbial library through untargeted metabolomics coupled with bioinformatics tools. This chapter focuses on the initial steps involved with developing a microbial library for drug discovery, starting with environmental isolation. This resulted in selecting 37 strains for a drug discovery library from 273 bacterial and 50 fungal isolates through pseudotaxonomic diversity and metabolite diversity.

CHAPTER 1. INTRODUCTION TO MASS SPECTROMETRY BASED METABOLOMICS TECHNIQUES

Based on excerpts from Grim, CM; Luu, GT; Sanchez, LM, Staring into the void: demystifying microbial metabolomics, FEMS Microbiology Letters, 2019, 366, (11), fnz135, by permission of Oxford University Press

1.1 Introduction

Metabolites are small molecules that drive biological systems from ecological to single-cell scales, in single cellular and multicellular organisms alike. For example, small molecules can act as chemical communicators among a quorum of similar microbes^{1,2}. Small molecules from nature have garnered attention as sources of novel antibiotics, cancer therapeutics, painkillers, and more³. In mammals, neurotransmitters are particularly small, some less than 200 Daltons, but have relevance in Parkinson's disease⁴, ADHD (Attention-Deficit/Hyperactivity Disorder)⁵, Alzheimer's disease⁶, among other neurological disorders. One way of classifying these molecules is by sorting them by purpose within an organism. Metabolites required for the survival of an organism have been termed "primary metabolites", while metabolites that perform secondary tasks such as chemical defense against predators are termed "secondary metabolites".

Primary metabolism includes well understood biochemical pathways, such as glycolysis. Since the target metabolites are known, the study of primary metabolism is often targeted in nature. A targeted metabolomics experiment requires tracking metabolites with high specificity and sensitivity. The progression of a disease can often be monitored through the abundance of a biomarker in a patient's serum, urine, or blood. These experiments require the accurate measurement of the analyte despite a complex background at low, often times nanomolar concentrations. Mass spectrometry has thus far been a crucial tool in research laboratories for its ability to study complex mixtures⁷.

The secondary metabolites of an organism are often repurposed for pharmaceuticals³. For example, since the discovery of penicillin researchers have isolated microbial chemical

defenses to use them as antibiotics⁸. However, in the search for new antibiotics, challenges arise in finding novel metabolites or the failure to identify “known unknowns” which has led to high rates of rediscovery. Researchers have traveled the globe to search challenging environments in the hope of finding uniquely adapted microbes that produce uncommon chemistry to survive in harsh conditions. However, researchers often still face a plethora of overlapping chemistry in the face of their attempts to diversify their microbial libraries. New mass spectrometry tools exist to help profile the metabolite production of microbes earlier in the process so that less money and time are spent re-discovering common bioactive molecules.

Metabolomics studies fall into different categories that dictate the ideal instrumentation and approach to achieve different research goals. Figure 1.1 shows a basic outline of how a researcher may choose the tools best suited for their research question. This chapter summarizes the cases for common mass spectrometry instrumentation and data analysis workflows. The following chapters will explore specific uses of these workflows to address both targeted and untargeted research questions.

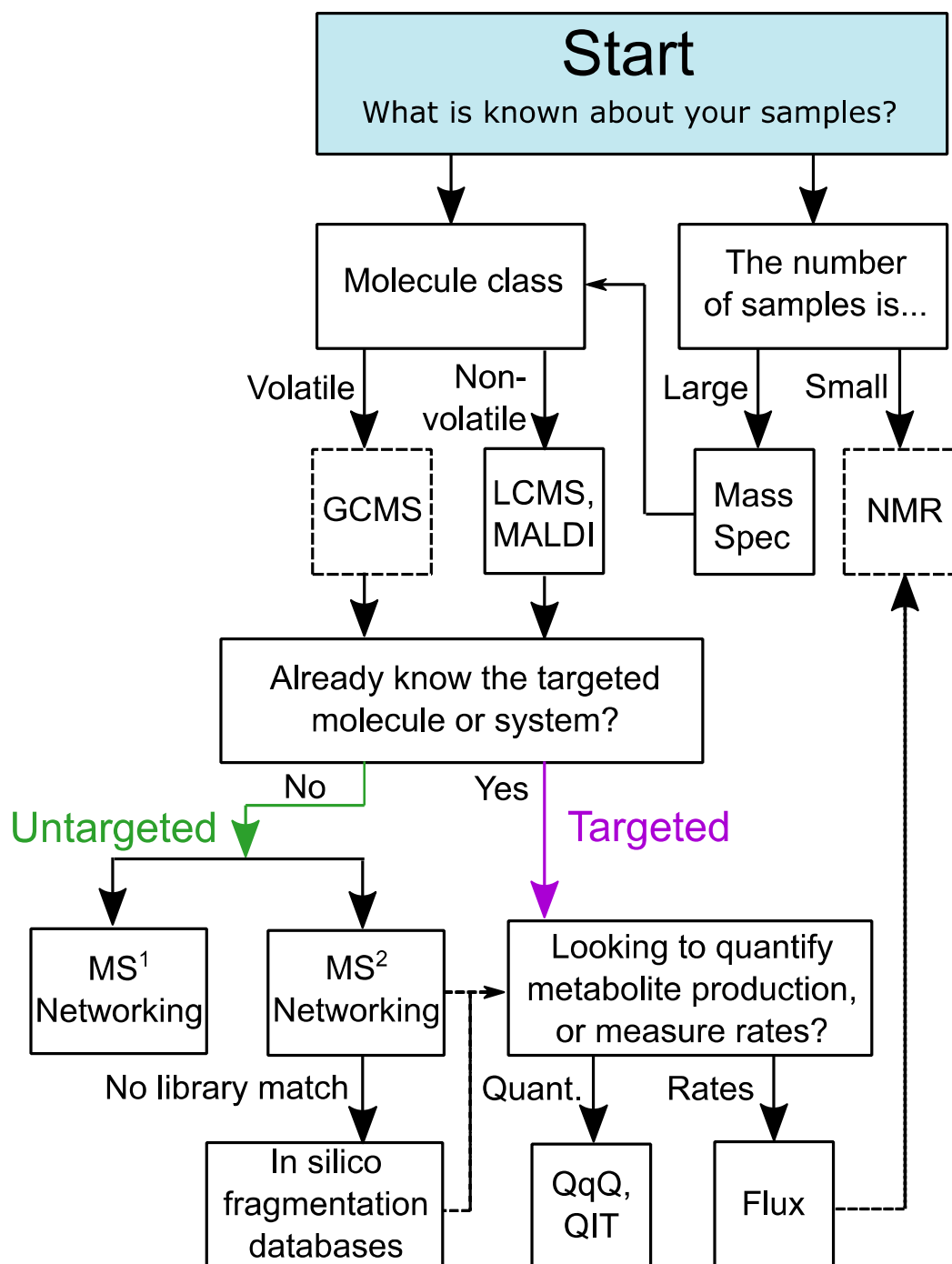


Figure 1.1. Workflow decision tree. Dashed lines indicate methods that are not covered in this chapter. Abbreviation key: GCMS (Gas chromatography mass spectrometry), LCMS (liquid chromatography mass spectrometry), MALDI (matrix assisted laser desorption/ionization), QqQ (triple quadrupole), QIT (quadrupole-ion trap)

1.2 Instrumentation

The use of analytical instrumentation is essential in the field of metabolomics. Nuclear magnetic resonance (NMR) and mass spectrometry (MS) can both be utilized in metabolomics, with MS typically being best suited for large sample sizes and cases requiring high sensitivity⁹. Mass spectrometry-based techniques detect ions, which can be either positively or negatively charged molecules that are denoted as mass to-charge ratios (m/z) each with a corresponding intensity value. Mass spectrometers operate under vacuum and require an ion source to deliver analytes to the mass analyzer and then the detector. The primary methods for introducing samples are by gas chromatography (GC) and liquid chromatography (LC). However, lasers can also be used to desorb ions from a solid phase such as in matrix-assisted laser desorption/ionization (MALDI).

1.2.1 Chromatography

Chromatography helps separate analytes before introducing them to the ion source. This may help simplify what is being ionized simultaneously to increase sensitivity. Additionally, chromatography adds the dimension of time. Analytes spread out across a column based on their affinity to the stationary phase, such that compounds that are detected early on have very low affinity while compounds that elute late have a very high affinity. This often translates to information about an analyte's polarity.

GC-MS is best suited for volatile metabolites or those that can be derivatized as such. Ions in GC-MS are created via electron ionization, a form of hard ionization that produces molecular ions $[M]^+$. During ionization, metabolites fragment and are separated by the mass analyzer. This ability to fragment ions combined with high resolution, sensitivity,

reproducibility, and large compound libraries makes GC-MS the preferred instrument for analysis of volatile metabolites under 1000 Da¹⁰. However, the current limitations include the inability to analyze metabolites, especially unknowns, that don't fit those criteria, such as larger, non-volatile metabolites that are unable to be derivatized into volatile compounds. GC-MS is a robust and well understood form of mass spectrometry, but libraries and tools based on LC-MS are rapidly evolving to better understand the non-volatile chemical space.

Introducing liquid samples to the vacuum environment of a mass analyzer was made possible by the introduction of electrospray ionization (ESI). Electrospray ionization sprays a fine, charged mist of the analyte which causes the solvent to evaporate prior to introduction to the vacuum environment¹¹. The ions are then able to proceed to the mass analyzer under vacuum. Most biomolecules are non-volatile and water soluble, so this was a major breakthrough for the field of metabolomics. Additionally, ESI is considered soft ionization. Analytes, including intact proteins, can be detected as ions of $[M+H]^+$ or $[M-H]^-$ instead of the rigorous fragmentation found in GC-EI-MS.

1.2.2 Mass Analyzers for Targeted and Untargeted Metabolomics

Targeted metabolomics analyzes the differences in a selected few, typically structurally characterized, metabolites. This can mean looking at the biotransformations a drug undergoes after it passes through the body in urine, or how much of a drug was present in the bloodstream of an individual. Untargeted metabolomics instead looks at all or a broad selection of metabolites in a system. A common example of untargeted metabolomics is in the development of an extract library. Extracts, often from environmental sources, are all stored to be tested against a variety of assays for bioactivity. Commonly these assays include the

extract's potential as a chemotherapeutic or an antibiotic. In an untargeted approach, it is unknown what single compound is causing the activity, but analytical chemistry can help determine if it's a known compound, what class it may fall under, or if it's completely unknown. Identifying if an experiment is untargeted or targeted is crucial in determining what mass analyzer is suitable for the experiment (Figure 1.1).

Targeted metabolomics generally looks to quantify the response of a system via a small selection of metabolites. This can be done as a relative measure, where the output would be that condition A elicits three times the response of condition B, or it can be done as an absolute measure. In absolute quantitation, the end result is a specific concentration of the analyte, such as 10 mM or 50 ng/L. Quantitative MS relies on tandem MS to filter and ensure the measurement of only the desired experimental analyte. The most common instrument for quantitative MS is the triple quadrupole, also known as the QqQ. Figure 1.2 shows a schematic depicting how ions travel through the three quadrupoles. First, a complex mixture of ions passes through the first quadrupole. By setting a fixed voltage on the poles, the quadrupole can effectively filter out all ions but ones that have the same m/z value as the analyte with unit specificity¹². The second quadrupole introduces a neutral gas to fragment the precursor ions. The third quadrupole then again filters for a product ion that should be unique to the molecule of interest. This means that an ion was filtered at both the MS¹ and MS² level in order to be detected. This process is termed single reaction monitoring (SRM). Multiple reaction monitoring (MRM) involves the quadrupoles cycling through multiple set voltages to allow ions to pass through in sequence. The number of ions analyzed comes at the cost of sensitivity¹³. Quadrupole based instruments have been the longstanding instrument of choice in

quantitation because of the capability to filter ions and monitor individual m/z values with high sensitivity.

Untargeted experiments, by definition, do not have a molecule of interest to select and quantify in a given experiment. Hence, there is no specific ion or ion fragment that guides data collection. In this case, the mass analyzer used should be capable of collecting data across a wide array of m/z values. Generally, this means using time-of-flight (TOF), ion trap (IT), or orbitrap mass analyzers. Quadrupoles are frequently paired with TOFs and orbitraps, such as in the Q-Exactive, for high resolution and tandem data collection. Depending on the experiment, MS^1 or MS^2 data may be best suited for untargeted analysis.

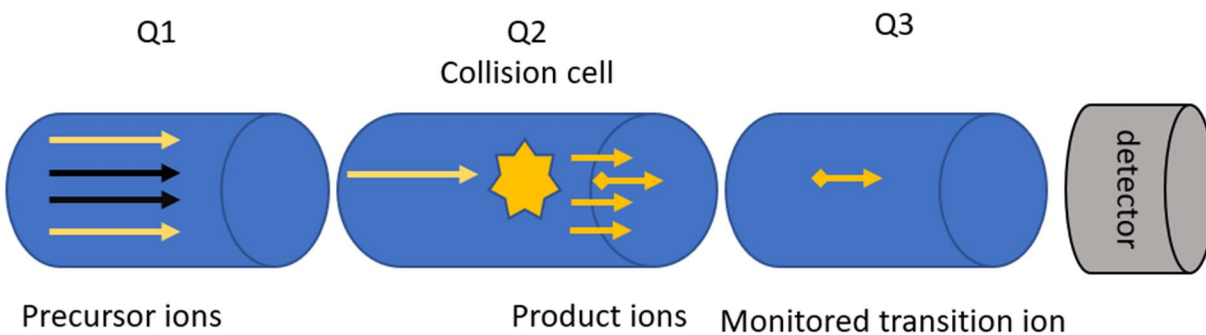


Figure 1.2 Diagram of SRM in a QqQ instruments. Black arrows represent ions that are not selected in the first quadrupole. The long yellow arrows represent ions that are selected in the first quadrupole. The short yellow arrows refer to product ions from the larger selected ion, and the short yellow arrow with the diamond node represents the ion selected in the third quadrupole.

1.3 Stable Isotope Approaches for Determining System Flux

Flux experiments involve using stable isotope labeled media to measure the rates of metabolic processes and exchange of metabolites in a living system¹⁴. While other metabolomics experiments focus on the presence, absence, or differences in metabolites, metabolite flux analysis (MFA) can determine the rate of metabolism and focuses heavily on primary metabolism and falls under the broad category of targeted metabolomics. In flux experiments, media with stable isotope labeled reagents are fed to cell cultures or whole organisms, which in turn ferment the reagents into amino acids and intracellular intermediates with the isotope atoms incorporated into their structure. For example, carbon-13 (¹³C) is the most popular stable isotope for flux analysis labeling^{15,16}. Deuterium is a frequent choice for other techniques requiring heavy labels, but deuterium is known to exchange with hydrogen readily and so is inappropriate for MFA. The location of the isotope on the metabolite's structure provides insight into what metabolic processes led to the formation of the metabolite. What becomes difficult is finding that specifically labeled atom in the structure, as a number of isotopologues can exist¹⁴. It is worth noting that elements have isotopes, whereas molecules have isotopologues for naming convention. Experiments have been designed to measure flux in colonies grown on solid agar¹⁷, the effects of nutrition stress¹⁸, and flux in co-culture¹⁹. Becker and Wittmann have thoroughly described a model MFA experiment using ¹³C, GC-MS, and OpenFlux for *Corynebacterium glutamicum*, which provides step by step details on how a basic experiment was conducted²⁰.

GC-MS and MS² are both valid ways of conducting MFA analysis²⁰. MS² is still being developed and implemented in standard fluxomic workflows, and so while it may be useful in coming years, it is not currently a standard workflow²¹.

1.4 MS¹ Untargeted Prioritization

In GC and LC-MS, analytes are dissolved in a mobile phase and separated by a column before being introduced to the mass spectrometer. MALDI differs in that samples are introduced as co-crystallized solids. Additionally, LC-MS often gives rise to multiply charged species. Since a mass-to-charge ratio is a division of mass by charge, when looking at large proteins the m/z values that correspond to a protein of X Daltons can show up at approximately a third of X, a fourth of X, a fifth of X, and so forth. A MALDI-TOF-MS instrument can analyze molecules up to a very high molecular weight, and typically only produces singly charged species. This means that instead of a protein creating an abundance of multiply charged peaks, it might instead show a small selection of adducts. If the experiment calls for analyzing multiple proteins, MALDI-TOF-MS is generally recommended to avoid needing to deconvolute overlapping signals.

MS¹ networking has been used to identify bacteria based on their protein signatures with minimal sample preparation²². Ribosomal proteins are the most abundant in bacteria, and sequencing the 16S rRNA subunit is the standard identification technique to identify the genus and sometimes species of a bacterium. However, those abundant proteins can also be detected by MALDI-TOF-MS. Although the spectra cannot report a nucleic acid sequence, the spectra can be compared to databases of MALDI-TOF-MS spectra collected of known bacteria to find a match²³.

IDBac is an open source platform using the same principle^{24,25}. After collecting protein data on a selection of microbes, the program uses similarity scoring on the protein spectra to organize strains in a pseudophylogenetic dendrogram. Additionally, MALDI-TOF-MS can collect MS¹ metabolite data from the same sample preparation. The nominal mass is limited in its ability to identify a metabolite, but IDBac creates metabolite association networks (MANs) to show which strains share metabolites with the same nominal mass. Users can use the protein and metabolite data together to choose strains they wish to further investigate. Some examples would be to select strains based on their putative taxonomic distance to known strains or their quantity of unique metabolite signals.

1.5 MS² Networking for Preliminary Compound Identification

MS² networking is a visualization tool that can be used to query the relationships between ions (metabolites) from any number of input files. The relationships are based on both the MS¹ and the product ions from the MS² spectrum. Relative and absolute mass differences and intensities between the products ions within a spectrum are compared across spectra to generate a cosine similarity score. The MS² spectra can be gathered in a number of ways from a cell culture including direct sampling of a developing microbial colony or by creating an extract from agar based colonies or fermentation broth²⁶. This technique can, for example, identify molecules that may share a common core structure but have differing chemical modifications, such as additional amino acids, sugars, or different methylation patterns to name a few. This structural information is captured in an MS² spectrum as differences in mass, which indicate losses of chemical modifications that are shared across related metabolites. This workflow is very powerful because the database of “seed” metabolites is constantly growing allowing for

deeper annotation of metabolites from MS² datasets. A “seed” is the spectrum of a known metabolite that can be found in the GNPS (Global Natural Products Social molecular networking) database, or the spectrum of a metabolite from an in-house database; the terminology seed is meant to represent a known spectrum that can ground the unknown data to a known entity. GNPS also acts as a repository itself, where researchers can upload their own reference spectra to grow the knowledge base²⁷.

Microbiologists have already started using GNPS to identify metabolite group differences across different extraction conditions or different media types^{28,29}. These experiments showed GNPS as a tool to identify major compound classes across many strains and conditions. Bauermeister *et al.* cultured six MAR4 *Streptomyces* strains and were able to compare metabolite production across strains as well as identify metabolite classes produced by the microbes. This led to a correlation of a cluster of novel ions with novel anti-biofilm activity³⁰.

It’s worth noting that connectivity to a “seed” in a molecular network still requires validation rather than considering this a definitive identification. This is an important tool that can help generate hypotheses but should not be considered a standalone resource for identifications. The molecular network is only as good as the data collected and is also heavily reliant on the use of appropriate parameters when setting up the network.

1.5.1 *In silico* Databases Potentially Expand Known MS² Space

An important step in untargeted metabolomics experiments is the dereplication of ‘known unknown’ metabolites that are present among the hundreds/thousands in an extract. *In*

silico database searches have proven to be an invaluable technique towards this end. Through the use of databases such as Massbank³¹, METLIN^{32,33}, GNPS^{27,34,35}, and more, MS² peak data can be used to annotate metabolites and provide putative structural information that was absent in MS¹ data. Despite its usefulness, *in silico* database searches are not absolute, and a compound's identity and structure must be experimentally verified using orthogonal methods such as MS² data, retention time, or NMR.

Although experimental data can provide the highest confidence in matching spectra to identify a metabolite, there is a bottleneck for how quickly an as of yet uncharacterized metabolites can be isolated, fragmented, and uploaded to repositories. *In silico* databases on the other hand have a high number of searchable spectra that may still match to a number of identified structures. Therefore, *in silico* databases are able to assist in metabolite dereplication. Compared to GC-MS, the number of LC-MS/MS spectra that have been curated is much smaller. To fill this gap, *in silico* fragmentation has been used to generate a greater number of mass spectra for comparison. METLIN has implemented *in silico* fragmentation simulations based on the spectra added to the database³⁶. As of 2016, METLIN contained 240,000 molecules where 13,000 had experimental MS/MS data and there were 160,000 *in silico* fragmentation structure predictions³⁷.

While METLIN relies on the strength of their size of accumulated data to develop predictions, GNPS looks to improve confidence in fragmentation predictions through molecular networking. The Network Annotation Propagation tool (NAP) uses the fragmentation of neighboring nodes for the node in question to improve confidence in the structure prediction³⁵. This feature is available through GNPS, but we recommend users familiarize themselves with

the basic molecular network techniques before attempting the experimental features.

Additionally, since both of these tools are bolstered by the community's data, we encourage researchers to upload their own MS/MS data. Additionally, *in silico* fragmentation databases are compatible with any workflow that acquires MS/MS data. *In silico* databases and fragmentation have been more extensively covered in several recent reviews^{38,39}.

1.6 Conclusion

Mass spectrometry is a highly adaptable tool, well suited to both targeted and untargeted metabolomics. Specific instruments are often categorized by which branch they are best suited for, but bioinformaticians and analytical chemists have extended the applications of mass spectrometers through software development and experimental innovation.

The proceeding chapter describes the process of translating the sample preparation of an untargeted experiment incorporating a novel method of *in vitro* MALDI imaging mass spectrometry to a robust LC-MS/MS quantitation assay. The original method was designed to find chemical signals that were upregulated in a tumorigenic environment leading to metastasis in high grade serous ovarian cancer⁴⁰. The MALDI-TOF-MS experiment was successful in generating a list of *m/z* values that were upregulated, however a targeted assay was necessary to further probe the context for the discovery. Despite the seemingly large divide between targeted and untargeted experiments, the two are still mutual partners within metabolomics.

The third chapter employs the use of bioinformatics tool IDBac to optimize a mammalian-sourced microbial drug discovery workflow. IDBac is an example of how a software can expand the uses for a specific MS instrument within the academic community, much like

how GNPS added a layer of information to MS/MS datasets. In this case, reducing the size of a drug discovery library while maintaining metabolite diversity is a cost-saving measure while employing the specific advantages of MALDI-TOF-MS. Chapter three demonstrates the amount by which a library can shrink as it is relevant to the research question.

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CHAPTER 2. ASSAY DESIGN FOR THE QUANTITATION OF CATECHOLAMINES IN AN *IN VITRO* HIGH
GRADE SEROUS OVARIAN CANCER MODEL

2.1 Introduction

High grade serous ovarian cancer is both the most common and most deadly form of ovarian cancer.¹⁻⁴.. More than half of ovarian cancer cases are diagnosed after the tumor has already metastasized, which carries only a 29.2% 5-year survival rate². A previous imaging mass spectrometry (IMS) experiment showed that norepinephrine (**2.1**) was secreted by a healthy murine ovary in the presence of murine tumorigenic oviductal epithelial cell line⁵. However, the study was untargeted and required adaptation to LC-MS/MS to move forward with exploring the catecholamine pathway with both sensitivity and efficiency. The goal of this project was to design an absolute quantitative mass spectrometry-based assay to measure norepinephrine and other catecholamine products from an *in vitro* model of the ovarian cancer microenvironment.

2.1.1 Primary Catecholamines

Norepinephrine belongs to a class of molecules called “primary catecholamines” along with epinephrine (**2.2**) and dopamine (**2.3**). The molecules all belong to the same biosynthetic pathway stemming from metabolism of tyrosine. Dopamine converts into norepinephrine, which converts into epinephrine⁶⁻⁹. An accumulation of norepinephrine within the tumorigenic system may be the result of a malfunction elsewhere within the pathway. Therefore, it was decided to monitor all three of the primary catecholamines in order to develop experiments to contextualize the initial observation.

2.1.2 MRM Quantitative Assays

A standard multiple reaction monitoring (MRM) quantitation assay is a relatively quick and sensitive method for analyzing *in vitro* samples. With a quantitative assay, it would be

possible to compare responses of catecholamines at different time points, different cell lines, or with the presence or absence of key enzymes or receptors within the pathway. Analyzing the cell media of tumorigenic cell lines is an effective way to analyze metabolic pathways without the greater interference of a whole organism¹⁰.

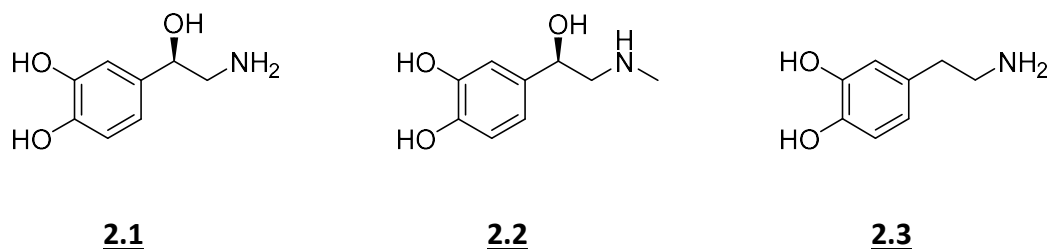


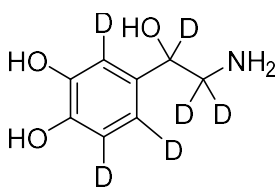
Figure 2.1 provides an overview of the sample preparation in a quantitation assay, as adapted from Duncan, Gale, and Yergey¹¹. The first step is to generate calibration curve standards.

For absolute quantitation via mass spectrometry, internal standards are required¹¹. The most common form of internal standard is a heavy (selectively deuterated) version of the analyte. Compounds **2.4-2.6** are heavy standards for the analysis of compounds **2.1-2.3** respectively. The internal standard behaves nearly identically to the analyte within the instrument, except it will register on the mass spectrometer with a m/z shift, allowing the response to be measured without overlapping signal from the analyte. Since the molecules are nearly identical, whatever perturbations in the system effect the response of the analyte will also similarly affect the response of the internal standard¹¹. Since the concentration of the internal standard is the same across all samples, the ratio of the area under the curve (AUC) of

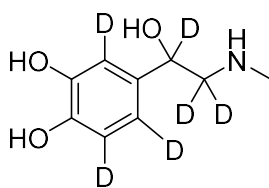
the heavy standard to the AUC of the light analyte acts as a normalized measurement of the analyte response.

2.1.3 Sample Preparation

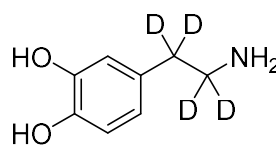
The second step for the analysis of solid samples is to extract the analyte of interest so that it is dissolved in solvent compatible with liquid chromatography (LC) analysis. The internal standard must also be present in the extracts to aid in quantification. In the original *in vitro* model used by Zink *et al.*, cell culture media was mixed with low-melting agarose to fill an 8-well chamber slide⁵, as depicted in **Error! Reference source not found.**. Each resulting well or “plug” represents one sample condition. The cell media used in this experiment is DMEM (Dulbecco’s Modified Eagle Media). Since the media, and therefore the agarose wells, are water based, the analytes are capable of diffusing throughout the sample. Therefore, the internal standards will also diffuse through the sample. We have chosen to add the internal standards, norepinephrine-d₆ (**2.4**), epinephrine-d₆ (**2.5**), and dopamine-d₄ (**2.6**), to the wells before extraction to account for extraction efficiency and ionization efficiency simultaneously during data collection.



2.4



2.5

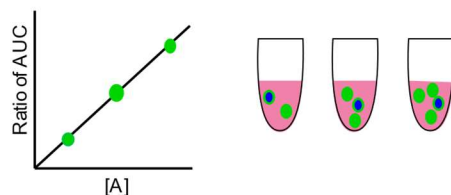


2.6

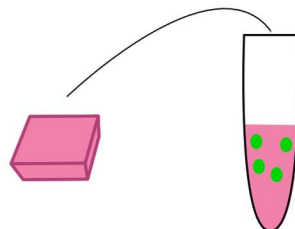
Generally, a targeted experiment involves measuring analytes within a complex mixture, such as blood, plasma, or urine. The biological fluid that carries the analyte is also called the

sample matrix. The matrix may contain proteins, cells, salts, or other compounds that reduce the ionization of the analytes and this reduction in ionization of the analytes is referred to as “matrix effect”. One way of reducing matrix effect is to use solid phase extraction (SPE). The SPE column works by the same principles of liquid chromatography. Analytes are retained by interactions with the solid phase to allow for separation from the sample matrix. When working with water or urine samples, there is often a high volume of matrix containing a very low concentration of analyte. Once again, an SPE column can help by accumulating retained analyte over the introduction of large volumes of sample, and then eluted in a small volume to concentrate the analyte. After sufficient “clean-up”, the samples are ready for introduction to the mass spectrometer either by chromatography or direct injection. Further information about instrumentation can be found in chapter 1, section 1.2.

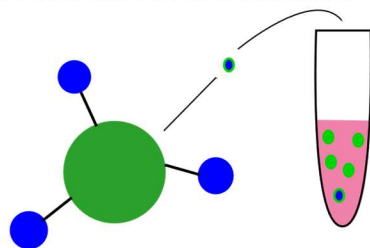
1. Determine calibration points



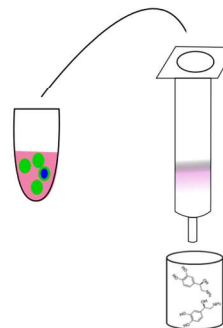
2. Dissolve analyte solids



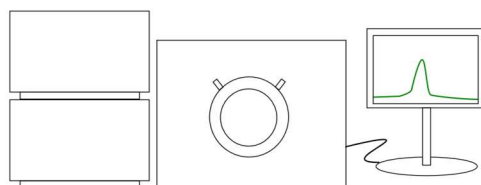
3. Add internal standard



4. Sample work-up



5. Sample analysis



6. Calculation of concentration

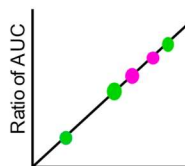


Figure 2.1. Quantitation basic workflow¹¹. In this assay, an internal standard is added (step 3) before the analyte solids are dissolved (step 2). Green represents the analyte, and blue represents deuterium. Pink represents the sample matrix.

2.2 Results and Discussion

2.2.1 Selecting Quantitative Transitions

The first quadrupole of a triple quadrupole (QqQ) or quadrupole-ion trap (QIT) selects intact ions to fragment. The mass of $[M+H]^+$ for norepinephrine is 170.1, but the more stable in-source fragment is m/z 152.1, or $[M+H-H_2O]^+$. This means that there's normally a higher abundance of norepinephrine derived ions at m/z 152.1, leading to a greater abundance of fragment ions. As shown in Figure 2.2, m/z 107 is a unique, abundant product ion. Therefore, $152 > 107$ would seem to be a judicious choice for a quantitative transition. However, while running a sample of pure epinephrine- d_6 , signal was observed for the $152 > 107$ transition. Since this transition state is not unique to norepinephrine, and epinephrine and norepinephrine co-elute by the chosen liquid chromatography method, it is not a viable indicator for the development of this assay.

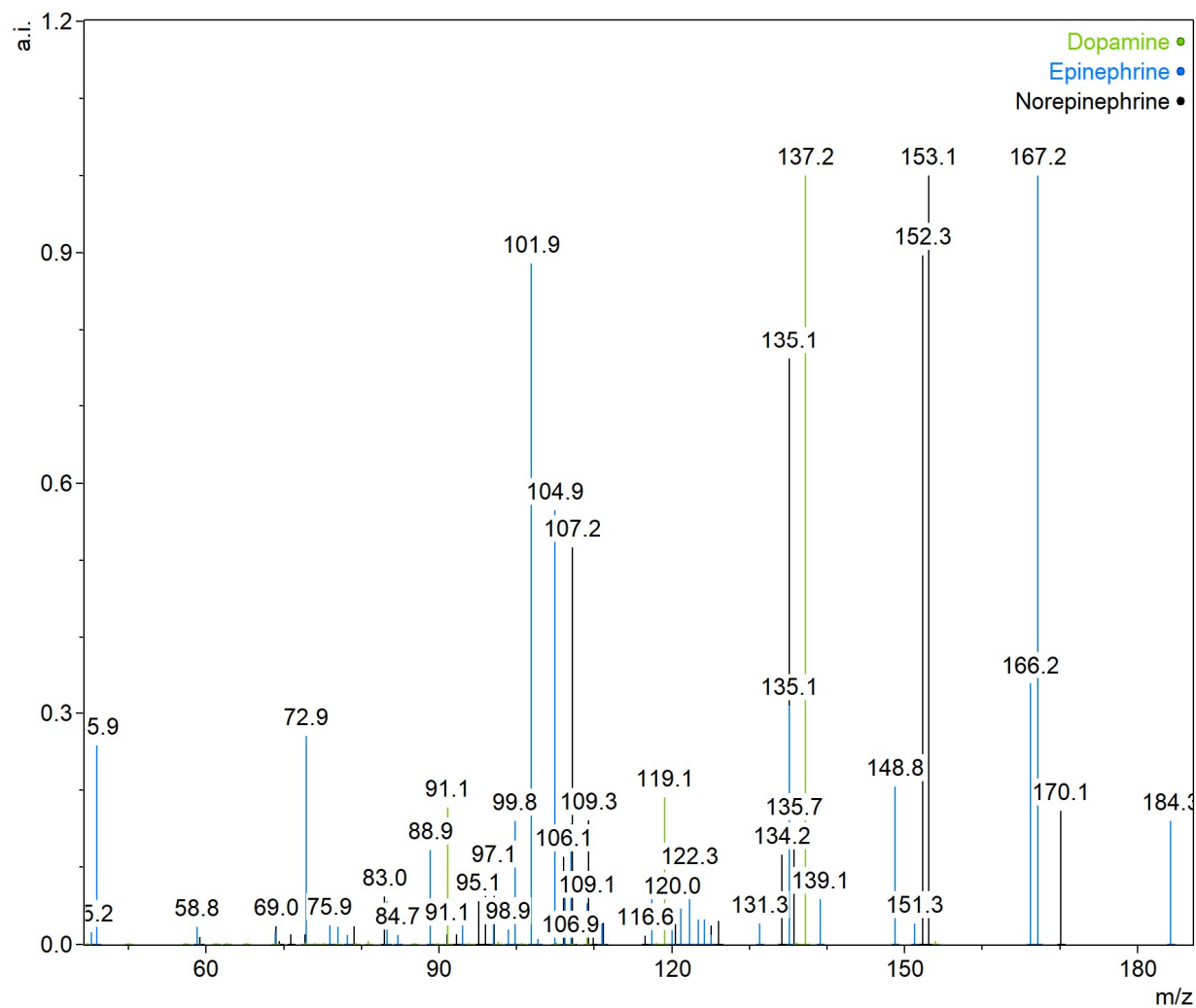


Figure 2.2. Overlaid fragmentation spectra of dopamine (green), epinephrine (blue), and norepinephrine (black). Data was sourced from MassBank¹², and the visualization was created in mMass¹³.

TABLE I

MRM TRANSITION PARAMETERS

	Q1	Q3	Collision	Declustering	Entrance
	(<i>m/z</i>)	(<i>m/z</i>)	Energy (V)	Potential (V)	Potential (V)
Norepinephrine	170.1	107	22	100	9
Dopamine-d4	158.1	94	20	169	11
Epinephrine-d6	190.1	172	11	219	11

Part of selecting transitions for quantitation includes verifying linearity of the response. Transitions were first selected by examining previous literature on catecholamine assays^{14,15} and further investigated by analyzing the analytes' fragmentation patterns (Figure 2.2). After selecting a transition to monitor, the mass spectrometer was tuned to determine the optimal instrument settings specific to each transition by introduction to the spectrometer direct injection. Direct injection bypasses the LC component of the system for the analysis of pure compounds. The parameters used for subsequent LC-MS/MS analysis are displayed in **Table I**.

After setting the method parameters, agarose wells consisting of equal concentration of each included catecholamine were made at 0.1 μM , 10 μM , 33 μM , 66 μM , 100 μM , 150 μM , and 200 μM . Figure 2.3 shows the calibration curve for the agarose extracts, monitored for each of the three catecholamines. Norepinephrine and dopamine-d₄ showed a consistent linear response, but epinephrine-d₆ showed a nonlinear response from 0.1-66 μM for the chosen

transition of $190.1 > 172$. The r^2 value for the dopamine fit was lower, but when looking at the data the fit was mathematically less linear because of the spread of the response for each injection. However, the response associated with epinephrine was more consistent, with less variability in peak area, but follows a non-linear path, as shown in Figure 2.4.

Linearity over a wide range of concentrations was examined. We estimated that tumorigenic conditions would prompt approximately a $10\ \mu\text{M}$ concentration of norepinephrine in a $300\ \mu\text{L}$ well¹⁶. If we assumed that the concentration of epinephrine would be within the same order of magnitude as norepinephrine, then $190.1 > 172$ is not a viable transition due to non-linearity.

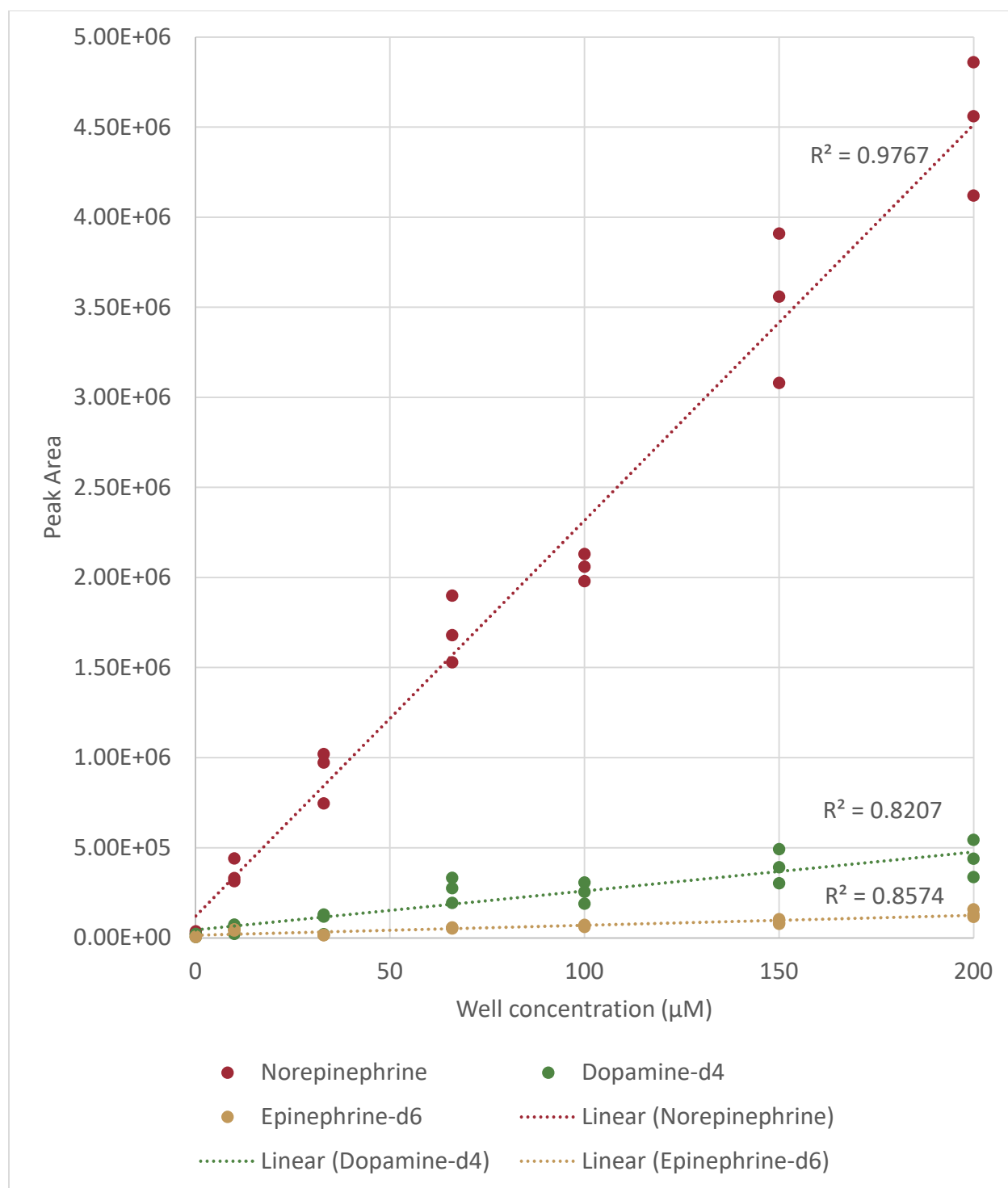


Figure 2.3 Overlay of the calibration curves relating peak area to concentration of an agarose well. Each point refers to a technical replicate for the given concentration.

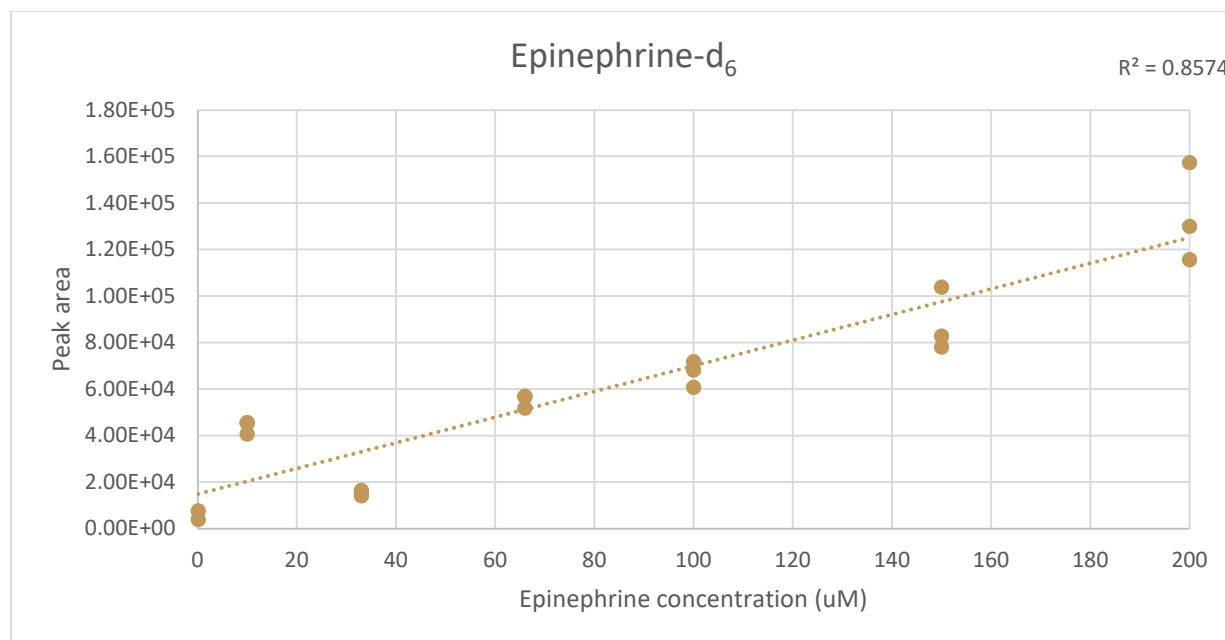


Figure 2.4. Graph of the area under the curve for the transition 190 > 172 for epinephrine stock solution with 3 technical injections each.

2.2.2 Evaluating Extraction Solvent

The catecholamines represented in this assay (**2.1-.3**) are all polar and mildly basic. The pKa for the protonated species of norepinephrine and epinephrine have been calculated to be around 8.5¹⁷. Thus, a selection of eight solvents were chosen to all be polar and include mildly acidic solutions. Agarose plugs were prepared by adding in pure norepinephrine bitartrate salt ($C_8H_{11}NO_3 \cdot C_4H_6O_6 \cdot H_2O$, Sigma-Aldrich) to the cell media and agarose for a final concentration of 10 mM.

TABLE II

EXTRACT INTENSITIES OBTAINED BY MALDI-TOF MS CORRESPONDING TO NOREPINEPHRINE

Solvent condition	Maximum peak height (a.u.)
Acetonitrile:Water:0.1% TFA	1150
Butanol:0.1% TFA	1030
Methanol:Water:0.1% TFA	843.0
Acetonitrile	524.0
Acetonitrile:Water	429.0
Methanol	N/A
Methanol:Water	N/A
Water	N/A

Figure 2.5 shows a comparison of all of the extraction conditions. The two extraction conditions with the highest intensity at m/z 170 are butanol with 0.1% trifluoroacetic acid (TFA)

and a 1:1 ratio of acetonitrile and water with 0.1% TFA. This MALDI-TOF-MS experiment is non-quantitative, so the two highest conditions were also compared by QIT-MS to achieve a higher sensitivity measurement.

The MRM experiment was set-up to monitor norepinephrine across a set of 5 calibration points, including a water blank. Figure 2.6 displays the peak areas for $170 > 107$ for various agarose plug concentrations. The ACN:H₂O:TFA extraction displayed increased extraction efficiency at 100 μ M and 150 μ M, for increased linearity overall. Therefore, ACN:H₂O:TFA was selected as the optimal extraction solvent.

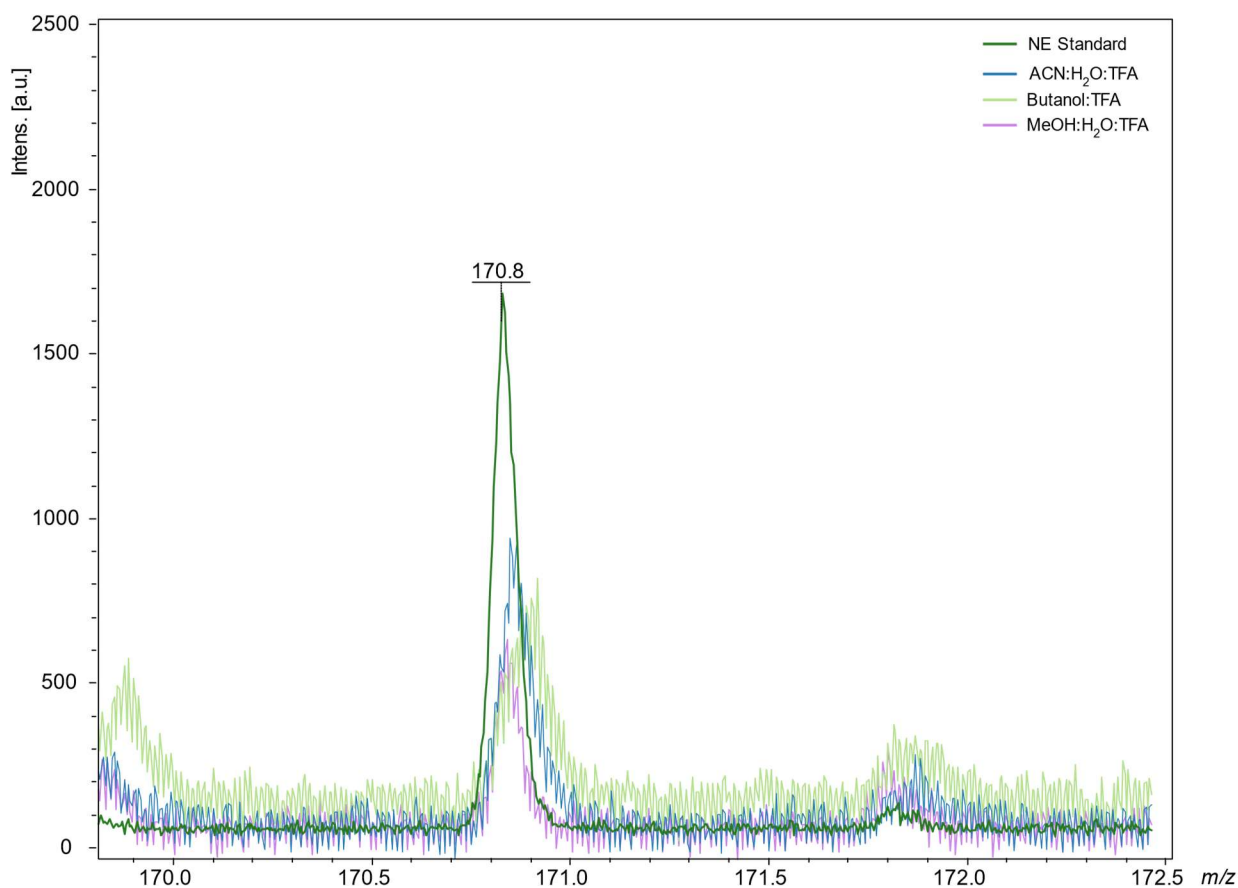


Figure 2.5. MALDI-TOF-MS spectra of agarose extractions at m/z 170, corresponding to norepinephrine.

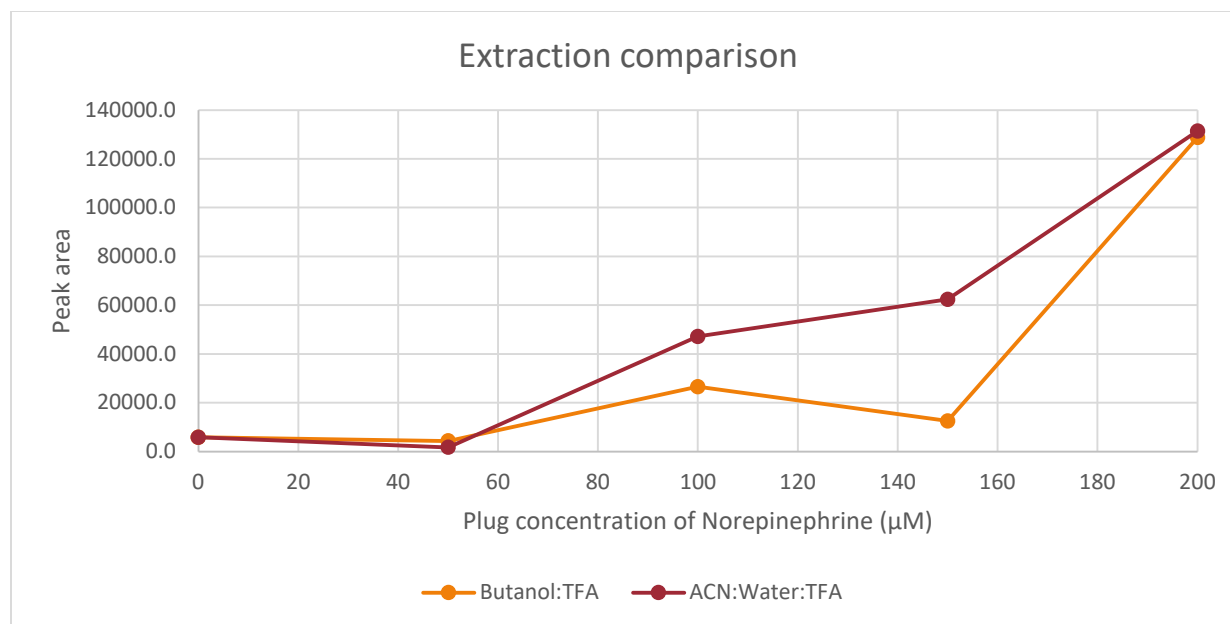


Figure 2.6. Comparison of agarose extraction solvents by MS/MS ($n=1$).

2.2.3 Removal of matrix by Solid Phase Extraction

After preliminary LC-MS/MS runs, the liquid chromatography column required advanced washes to reduce system pressure, indicating the column was being clogged by the agarose extracts. A SPE step, step 4 in Figure 2.1, was therefore added to reduce the need to wash the LC column during experiments.

By weight, DMEM is primarily composed of salts, D-glucose, and an assortment of amino acids¹⁸. The charge state of amino acids can fluctuate from -1 to +2 based on the surrounding pH and the pKa of the side chain. An appropriate sorbent for this assay would need to selectively retain analytes based on pKa. Furthermore, special attention must be made to the pH of samples prior to SPE for proper retention. The Oasis MCX sorbent is a reverse-phase cation-exchange sorbent optimized for basic compounds¹⁹. Proper use includes ensuring an acidic sample solution to protonate bases, performing the appropriate wash steps, followed by elution with a basic solvent¹⁹. A method optimized for a cell media matrix is shown in Figure 2.7.

Samples need to be acidic prior to loading onto the column so that the catecholamines will have a positively charged amine to interact with the negatively charged stationary phase. Limiting the sample preparation to an overall pH of 6 was designed to keep a majority of the amino acids in their neutral, zwitterionic state. Although the ions will have a localized positive charge, their affinity to the stationary phase should be lower than the positively charged catecholamines. Waters Corporation recommends eluting tryptic peptides in 2% ammonium hydroxide in 60:40 water:acetonitrile on MCX columns²⁰, which is unfortunately too similar to the catecholamine elution solvent to be used as a wash. The second wash, 60:40

methanol:acetonitrile, was designed as an alternative in this assay. Additionally, the first wash will remove polar negatively charged ions and polar neutral molecules. The final elution step is a strong base which will remove the proton from the charged catecholamines and free them from the stationary phase.

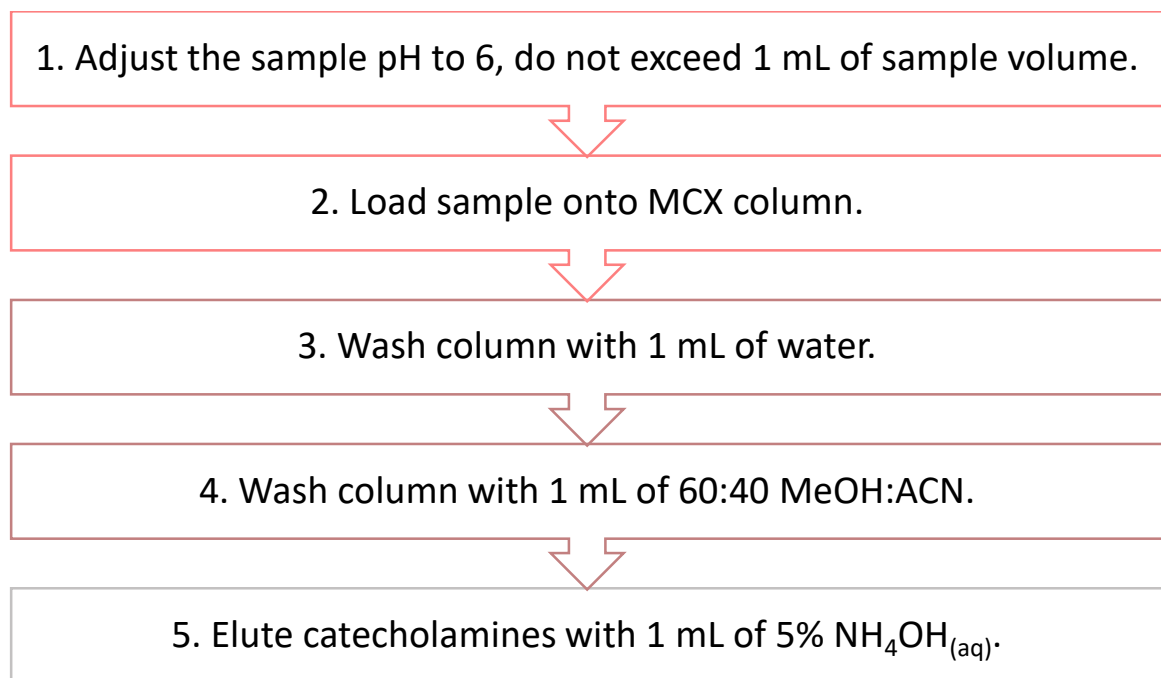


Figure 2.7. SPE method for sample cleanup.

During method development, each stage of the fractionation was collected for analysis by MALDI-TOF MS to ensure no catecholamines were lost during wash steps. Additional washes such as a basified methanolic wash and an acidified aqueous wash were excluded from the protocol based on the observation of minor amounts of catecholamines eluting with the wash. Figure 2.8 shows the MALDI-TOF-MS dried drop spectra of an initial extract overlaid with spectra of the elution fraction and a norepinephrine bitartrate stock solution.

The SPE step was introduced to clean up matrix compounds, namely amino acids and salts. Unfortunately, the mass range for many amino acids was also the range for CHCA:DHB (alpha-cyano-4-hydroxycinnamic acid: 2,5,-dihydrobenzoic acid) matrix crystals (m/z 100-200) so it is difficult to easily spot the presence or absence of amino acids by MALDI-TOF-MS. However, Figure 2.9 shows that there are compounds being removed in the process. The removal of sample matrix will aid in enhancing ionization of the analytes and potentially lower the limit of detection (LOD) for the assay.

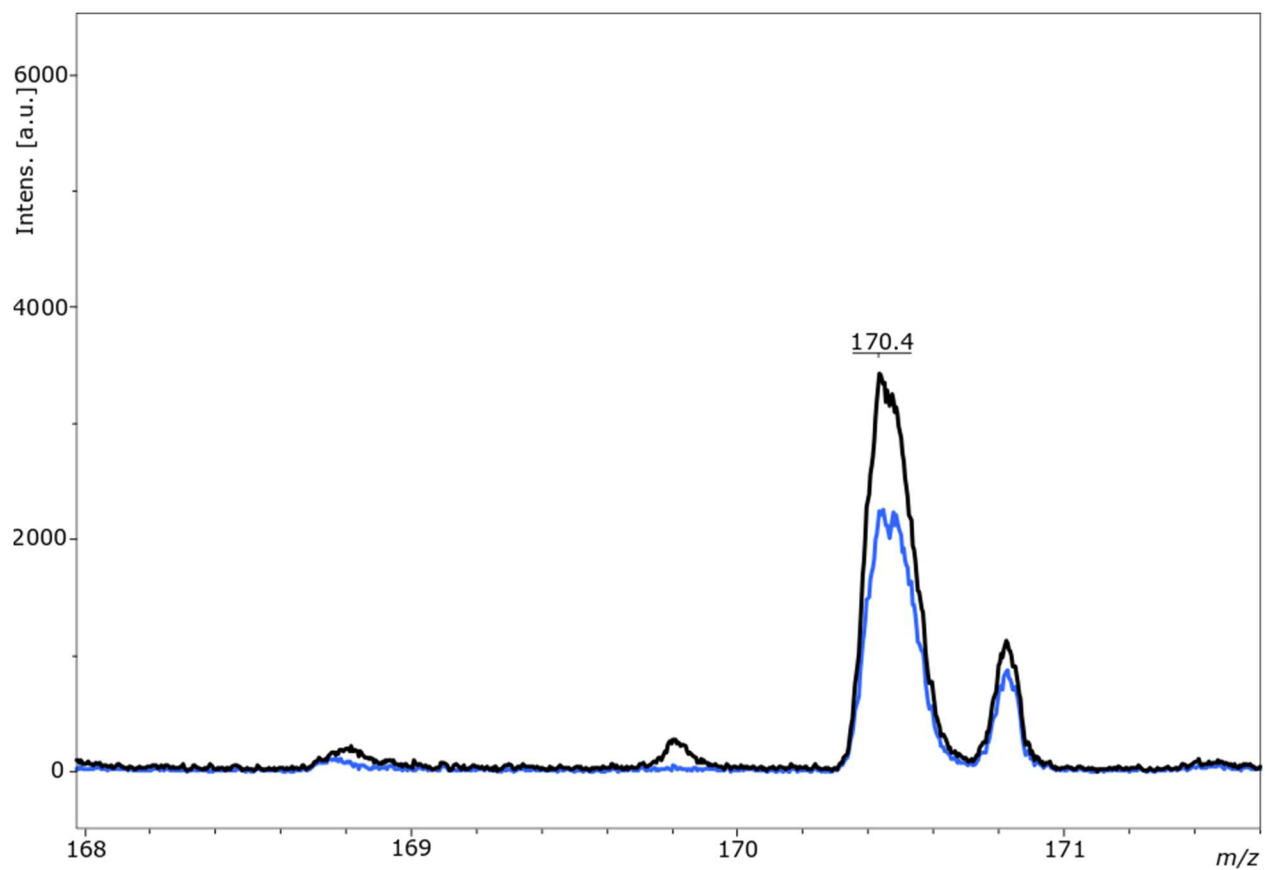


Figure 2.8. An overlay of MALDI-TOF MS spectra for the unfiltered extract (black), and the elution fraction (blue) at m/z 170.

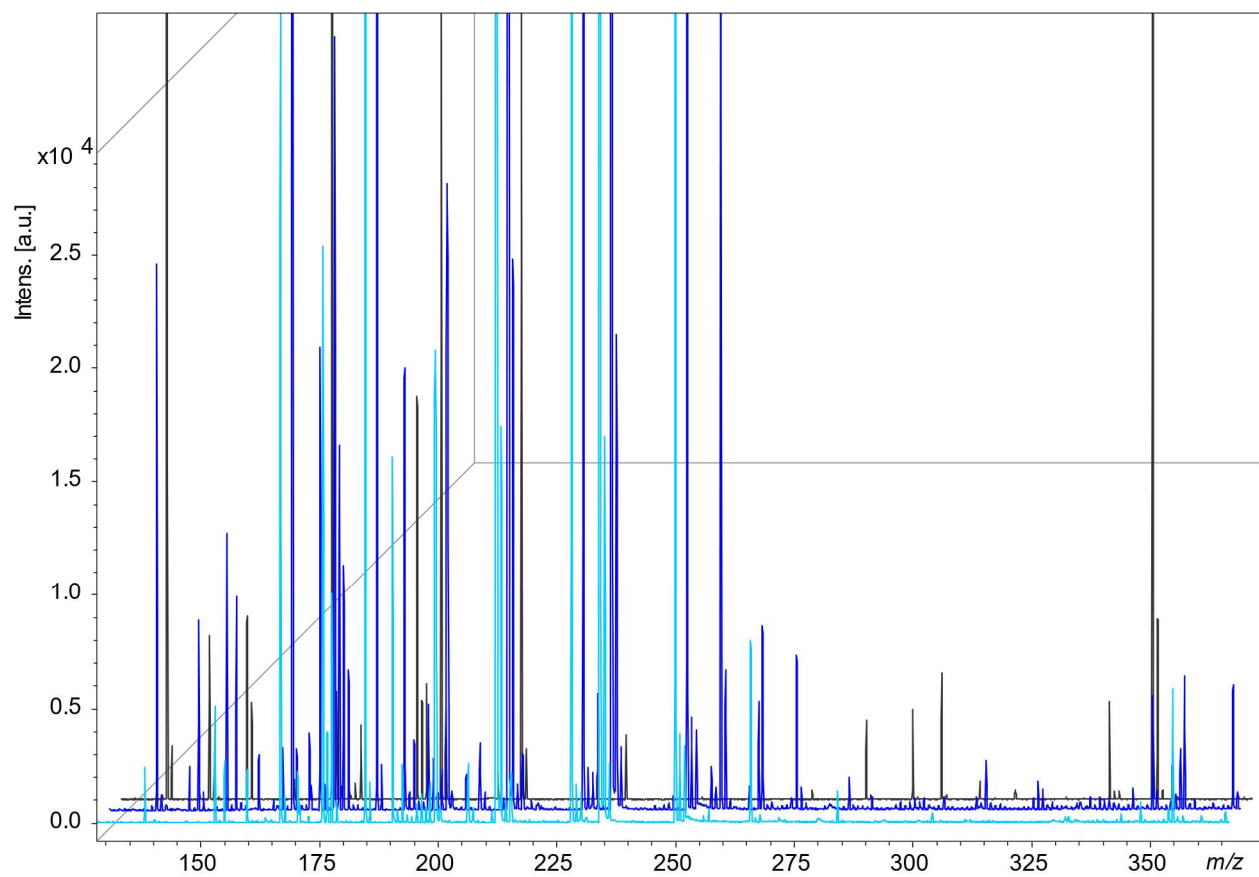


Figure 2.9. Offset overlay of a MALDI-TOF-MS sum spectrum of the original extract (dark blue), elute fraction (light blue), and matrix blank from m/z 150 to m/z 350.

2.3 Conclusion

Extraction and purification of catecholamines from DMEM agarose has been well optimized for the purpose of this assay. However, additional work is necessary to select the ideal MS/MS transitions. Overall, the MRM transition parameters need to be optimized by following researchers during the development of new assays or if using a different instrument, but the sample preparation methods outlined are viable for all catecholamine extractions from DMEM agarose. Additionally, previous literature^{10,21} only describes sample extraction from liquid cell culture, making this the first report of extraction from a solid cell media matrix.

2.4 Experimental

2.4.1 Agarose plugs

Low melting agarose plugs were made following the protocols stated in the original manuscript⁵, but the ITO-coated slide was substituted with a plastic slide and 8-well chamber. To prepare each well, 200 μL of 2X DMEM was mixed with 200 μL of 2% low melting agarose liquified at 70°C for a final concentration of 1X DMEM per well. 300 μL of the mixture was aliquoted per well. If the wells needed to have a concentration of norepinephrine, 6.78 μL of aqueous norepinephrine bitartrate salt solution was added to the aforementioned 400 μL of cell media and agarose before aliquotting 300 μL into the chambered slide. The concentration of the norepinephrine solution was prepared to be 60 times more concentrated than the desired well concentration. The slide was left to incubate overnight at 5% CO_2 and 37 °C to mimic any degradation that would occur during studies with cells.

2.4.2 Extraction

Low melting agarose plugs containing norepinephrine standard were individually placed into 1.8 mL microcentrifuge tubes and frozen at -70 C. Samples were thoroughly lyophilized, and then macerated with a fresh toothpick. 1 mL of 50:50 acetonitrile:water with 0.1% TFA was added to each sample and the extractions were sonicated for 1 h. The extractions were then centrifuged at 10k rpm for 2 min. The supernatant was transferred to a new vial before adding 150 μ L of acetonitrile to dehydrate any agarose that was rehydrated during the extraction process and to ensure all supernatant was transferred. Resulting extracts were dried *in vacuo*.

2.4.3 MALDI-TOF Dry Drop Analysis

MS data was collected in positive reflection mode on an Autoflex Speed LRF mass spectrometer (Bruker Daltonics) over the mass range 100-1000 Da. The smartbeam-II laser (355 nm) was operated at size 3, medium, and 4,000 shots. Laser power was set between 40-60% with a gain of 9.0x. Extracts were spotted on a ground steel 96 target plate in a 1:1 ratio with recrystallized 1:1 CHCA:DHB matrix. The instrument was calibrated manually using phosphorus red. Signals below m/z 80 were suppressed.

2.4.4. Parameter Optimization

Optimization of the collision energy, declustering potential, and entrance potential for standards in **Table I** was accomplished through direct infusion of single analytes at 10 ng/mL on a SCIEX Triple Quad™ 5500. The syringe pump was set to a flow rate of 10 μ L/min. Each parameter was optimized using the ramping feature of the instrument and selecting the voltage that produced the highest intensity over the course of 5 technical replicates.

2.4.4 MRM Analysis

The MRM assay was performed on a SCIEX Triple Quad™ 5500+ LC-MS/MS System interfaced with an Agilent 1200 Infinity series HPLC. Dried extracts were resuspended in 25 μ L of Milli-Q purified water (Millipore Sigma) before injection. The chromatography gradient started at 97% A and increased to 99% A after 3 minutes, where solvent A was water purified by a Milli-Q purification system with 0.1% formic acid buffer, and solvent B was LCMS grade acetonitrile with 0.1% formic acid buffer. A 5 μ L injection volume was used and a flow rate of 0.45 mL/min. A Kinetex Biphenyl 2.1x50 mm column (Phenomenex) was used for liquid chromatography. Data was processed in Skyline version 20.1.0.31^{22,23} and graphed in Microsoft Excel. No weighting methods were used for determining the least square error linear regression.

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CHAPTER 3. DEVELOPMENT OF A UNIQUE MICROBIAL LIBRARY SAMPLED FROM
HETEROCEPHALUS GLABER TISSUE

3.1 Introduction

Naked mole rats (*Heterocephalus glaber*) have been widely researched for their long lifespan^{1,2}, ability to survive hypoxic environments^{3,4}, intolerance to pain⁵, and resistance to cancer^{6,7}. *In vitro* study of mammalian cells, our collaborators and their colleagues have had difficulty with unusual microbial “contaminants” while attempting to culture skin fibroblasts from *H. glaber*. Due to repeated observations of this phenomenon, we hypothesized that the so-called “contaminants” may be commensal microbes.

Screening libraries are collections of chemicals, either pure or as a mixture, used in conjunction with disease specific bioassays to screen for drug leads. Libraries can be comprised of synthetic chemicals as well as natural product extracts or purified compounds. Concerns have arisen regarding the large amount of biomass needed to fully identify an active molecule from natural sources⁸. Researchers are increasingly turning to microbes as powerful, renewable cellular machines for producing bioactive metabolites^{8–10}. Figure 3.1 shows a microbial library generation process as applied to this project.

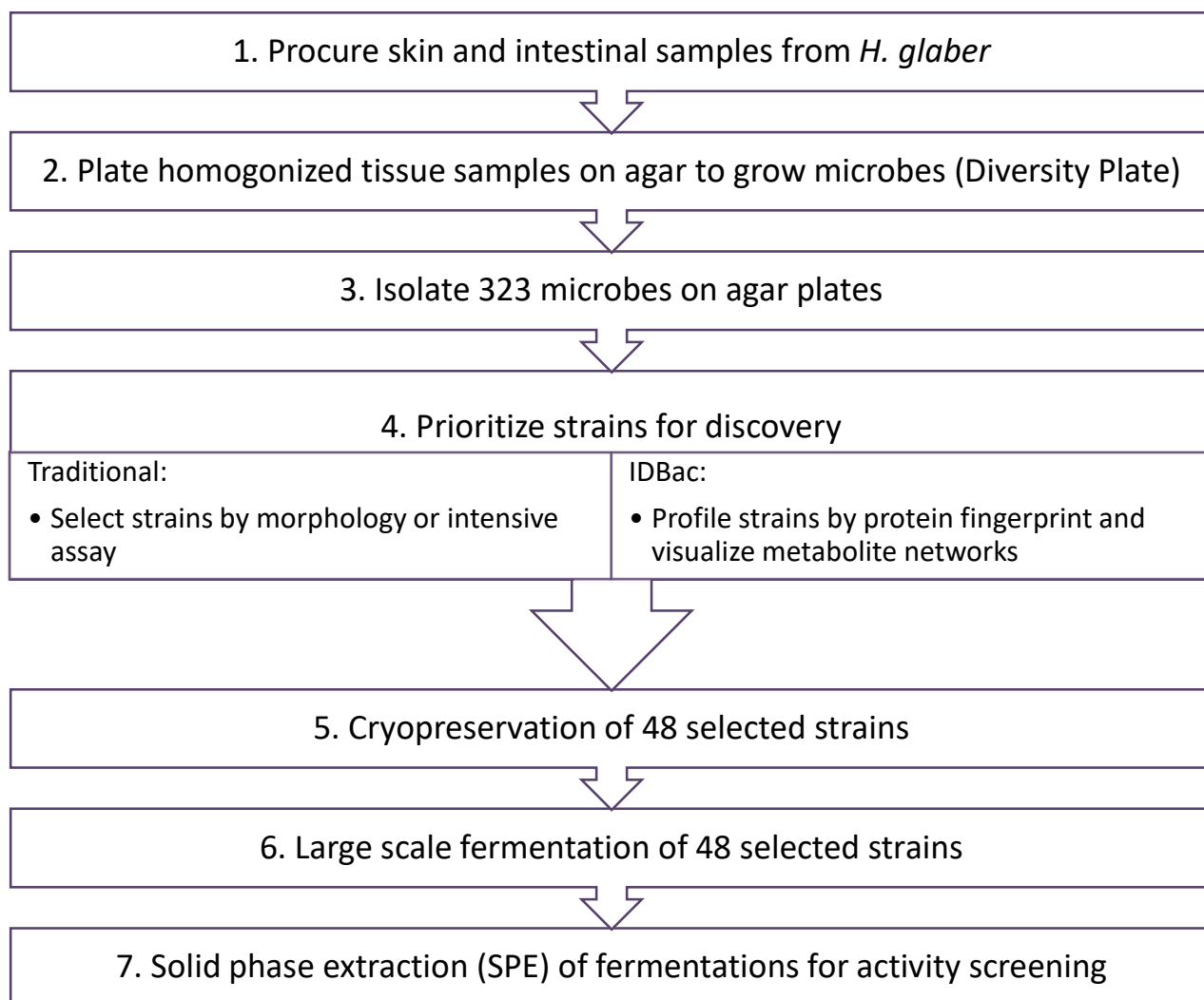


Figure 3.1. Project overview. Steps 6 and 7 are future directions for the project.

A microbial presence in the skin drove the hypothesis for the project but it was decided to also source the intestinal tissue. Mammalian microbiomes have been proven to be viable sources of bioactive molecules^{11–16}, and naked mole rats have been shown to have a unique gut microbiome compared to other mammals¹⁷. The gut has also been linked to overall human health^{18–20}, and the same may be true in *H. glaber*. Even if there is no correlation between the health of *H. glaber* and its microbial partners, the novelty of the *H. glaber* as a niche environment may still lead to unique molecules from a discovery pipeline.

3.1.1 Media Selection can Impact Microbial Diversity in Library Selection

The first step in library generation is the “diversity” plate. A diversity plate provides a broad span of microbes the nutrient combinations they need to propagate in the laboratory setting. The level and type of nutrients (i.e. simple carbon and nitrogen vs complex carbon and nitrogen), or the addition of antibiotics or antifungals determines what microbes may propagate on the plate. If the media is too nutrient rich, the plate may become quickly overgrown with common microbes. Likewise, if the media lacks nutrients, it may be too hostile to foster growth.

Fungal growth media tends to be rich in carbohydrates, which acts as a source of carbon for cellular synthesis²¹. Different species of fungi can utilize simple carbon sources such as glucose or amino acids, while others can use more complex sources like keratin and lignin. The majority of filamentous fungi utilize carbohydrates like monosaccharides, starch, and cellulose²¹. One media condition, simple nutrient fiber plus fiber (SNF + Fiber)²², was designed to mimic a gut environment by utilizing insoluble fiber. The media had previously been shown to be able to cultivate gut-derived bacteria, and we hypothesize that the fiber could be an

ample carbon source for fungi as well²². SNF + Fiber otherwise lacks any other nutrients. Gause Modified Freshwater (GMF)²³ was used as a slightly more nutrient rich medium. In addition, to curb the growth of fast-growing microbes and select for both fungi and bacteria, half of the plates had an anti-fungal agent added while the other half had an antibiotic added.

3.1.2 Protein and Metabolite Profiling (IDBac) with MALDI-TOF MS

After colonies become visible on the diversity plates, they are transferred to individual isolate plates to ensure the purity of the colony. Once the colony has been determined to be pure, the microorganism is profiled by MALDI-TOF MS using the open source platform IDBac²⁴, introduced in section 1.4. Microorganisms form groups based on their protein similarities. Then, users can view the MALDI-TOF-MS metabolite association network (MAN) of any given number of strains to identify unique metabolite production based on features. This process allows for selecting a library that is both taxonomically diverse and varied in metabolite production potential.

3.1.3 Isolate Name Code

The name of each isolate is based on the conditions under which it was isolated. The letters before each isolate ID number stand for the tissue source and media condition. K1 stands for the first skin sample, K2 stands for the second skin sample, M stands for the small intestine and L represents the large intestine. A following G denotes GMF media, while a following F denotes SNF media. Bacteria are numbered by their pseudophylogeny group followed by their identifier within the group. There is no group marker for fungi, only an isolate number. There is no overlap in fungi isolate numbers e.g. only one isolate ends in -001.

3.2 Results and Discussion

Across all of the diversity plates, 323 microbial isolates grew and were profiled. The ultimate goal of the library will be to generate a 384-well plate of fractionated extracts.. This project was designed to be a small experimental library, so one 384-well plate functions as a complete unit for high throughput screening. Since the chosen fractionation scheme results in seven fractions per extract²⁵, there will be room for approximately 48 isolates in the library. Since the project first began as an investigation of a fungi, it was decided to favor isolated fungi within the library for a goal of selecting 24 bacteria and 24 fungi for preservation in cryostorage and subsequent fermentation and fractionation

3.2.1 Fungal Library Prioritization

Of 323 isolates profiled, 50 were identified as fungi based on morphology. Yeast tend to be off-white and mucosal, similar to bacteria, and so it is possible some yeast species may be identified as bacteria based on morphology. However, yeasts are used as biosynthetic hosts for natural product synthesis precisely because they are known for a lack of unique secondary metabolite production²⁶. It is unlikely that a yeast's metabolic diversity would surpass any bacterium it may be grouped with during the profiling steps. Figure 3.2 shows the dendrogram generated from the fungal protein fingerprint along with seed strains with known identities provided by a collaborator lab.

Upon seeding the pseudodendrogram, two major clades arose. The first clade, marked with green lines within the pseudodendrogram in Figure 3.2, seeded with two cheese rind-derived *Penicillium* strains indicating a possible genus identity for other isolates within the clade. It is worth noting that *Penicillium camemberti* did not clade with the other *Penicillium*

species. Mass spectrometry methods for taxonomic identification of microbes is not fully capable of charting higher order phylogeny, and this can sometimes apply even at the genus level²⁷.

Near the bottom of the pseudodendrogram, *Debaryomyces sp.* claded with three isolates. *Debaryomyces* is a yeast, suggesting the possibility that the three isolates may also be yeasts rather than filamentous fungi. The isolates within the green, *Penicillium*-seeded clade were incorporated with the four isolates at the bottom of the pseudodendrogram to form “group A”, and the large clade with no seeded fungi that shares a branch point with group A was designated “group B”.

The MAN for each group was viewed separately with the goal of identifying 24 fungi to prioritize for preservation in a drug discovery library. Group B, shown in Figure 3.3, displays numerous isolates with unique nodes. The highlighted isolates from Figure 3.3 and Figure A.12 were first used to inoculate 90 mm ISP-2 agar plates to generate a large quantity of spores for preservation. However, upon also inoculating ISP-2 agar plates made with 25 mg/L of cycloheximide, resulting bacterial growth indicated that 75% of the chosen isolates listed in **Table III** were a visually homogenous culture of both bacteria and fungi. It was hypothesized that the selected isolates displayed so many unique nodes precisely because a mixture of bacterial and fungal metabolites were being represented at each “single isolate”. It was determined that efforts would be made to isolate strictly the chosen fungal strains, and run another dry drop extraction before finalizing the decision to migrate the strains to cryopreservation.

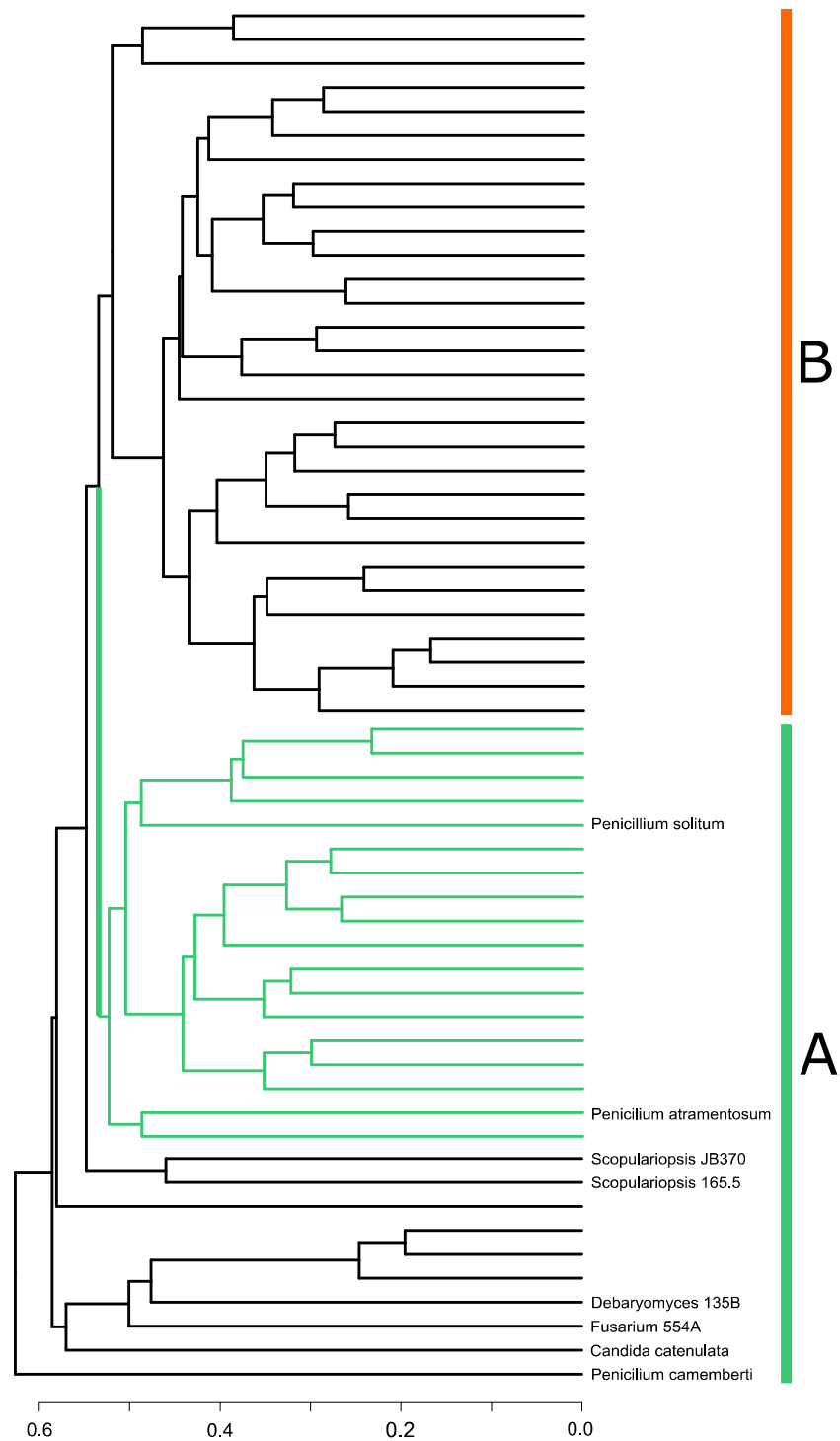


Figure 3.2. Pseudophylogenetic dendrogram created from MALDI-TOF MS protein profiling of isolated fungi seeded with identified fungal strains. Identified strains are denoted by their taxonomic name or identifier.

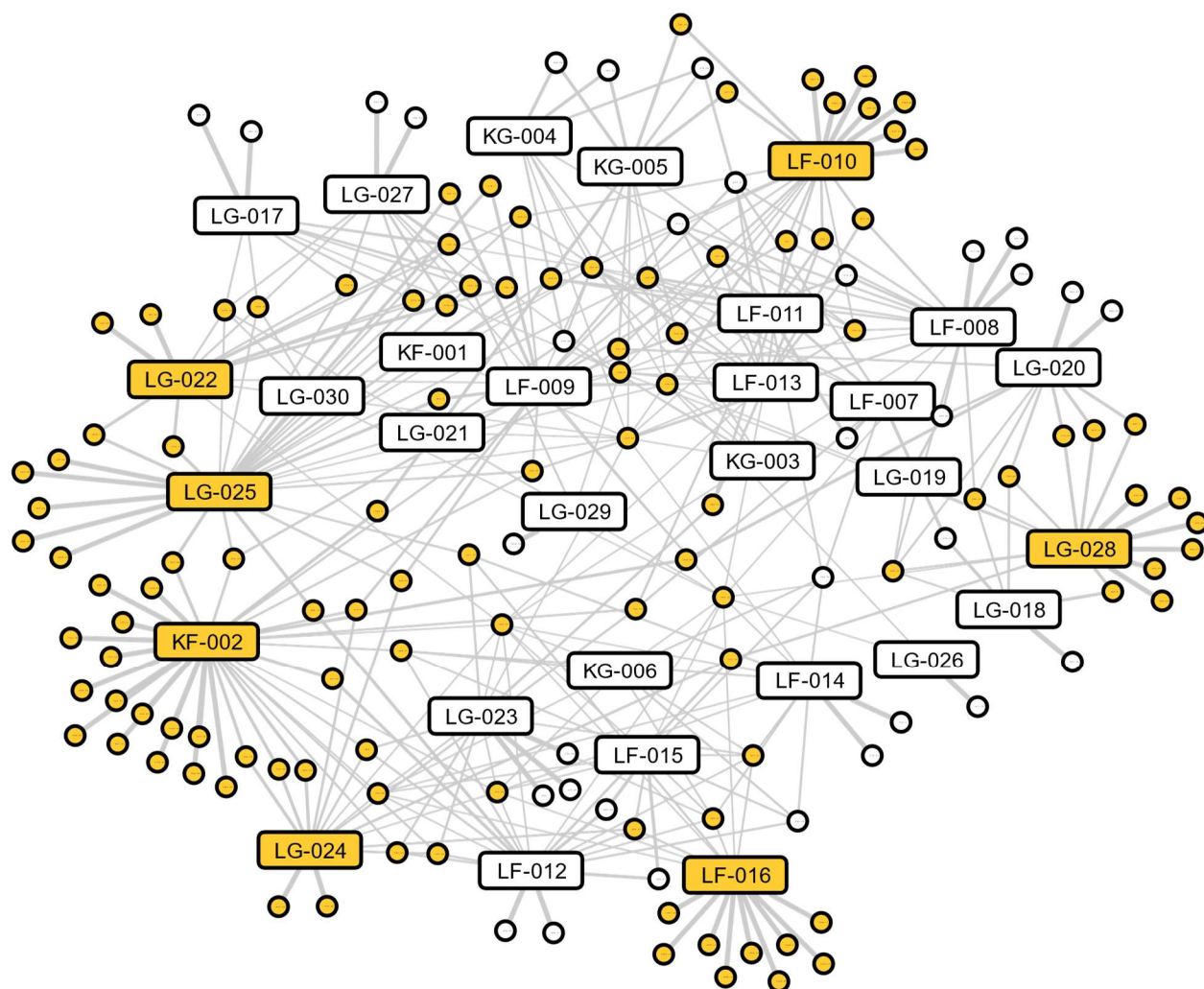


Figure 3.3. Metabolite association network of fungal isolates in clade B. Library isolates and their connected metabolite nodes are in orange.

TABLE III

Library Fungi Metadata					
Group	Isolate name	Tissue source	Media type	Plating method	First inoculated on
A	LF-030	Lg. intestine	SNF	Undiluted spread	7/1/2019
	LG-031	Lg. intestine	GMF	Undiluted spread	7/2/2019
	LG-032	Lg. intestine	GMF	Undiluted spread	7/2/2019
	LF-033	Lg. intestine	SNF	Undiluted spread	7/2/2019
	LF-034	Lg. intestine	SNF	Undiluted spread	6/29/2019
	LG-035	Lg. intestine	GMF	Undiluted spread	7/2/2019
B	LF-010	Lg. intestine	SNF	Undiluted spread	6/29/2019
	LG-022	Lg. intestine	GMF	Undiluted spread	7/2/2019
	LG-025	Lg. intestine	GMF	Undiluted spread	7/2/2019
	LF-016	Lg. intestine	SNF	1:100 spread	09/27/2019
	LG-024	Lg. intestine	GMF	Undiluted spread	7/2/2019
	K1F-002	Skin	SNF	1:100 spread	7/2/2019
	LG-028	Lg. intestine	GMF	Undiluted spread	7/2/2019

3.2.2 Bacteria Library Prioritization

The authors of IDBac have suggested reviewing large sets of microbes by making groupings based on pseudophylogeny²⁸. At the outset of the project it was decided to generate a library with 24 bacterial isolates, therefore based on Costa et al., the aim would be to divide the pseudodendrogram into 12 groups and to select two isolates from each group²⁸. One option would be to cut the pseudodendrogram at a given modularity score, however Figure 3.4 shows how the generated dendrogram has a sloped quality. Cutting the pseudodendrogram at, for example, 0.6 would generate nine groups each consisting of only a few isolates, along with an overwhelming group of over 200 isolates. Instead, it was decided to first evenly divide the pseudodendrogram into 12 even groupings of 22-23 isolates each. After the even cut, boundaries between groups were adjusted to maximize the pseudodendrogram height that determined the division. Essentially, groupings were adjusted through localized height cutting to avoid separating closely related isolates. Figure 3.4 displays the final groups.

The MAN of each group was used to then select which isolates would be preserved based on the total number of connecting metabolite nodes and the number of metabolite nodes unique to only that isolate. Additionally, after selecting two candidates, the amount of overlap between the two isolates was considered. The two isolates should together cover as many nodes of the MAN as possible. An example of one such network and the selected isolates can be found in

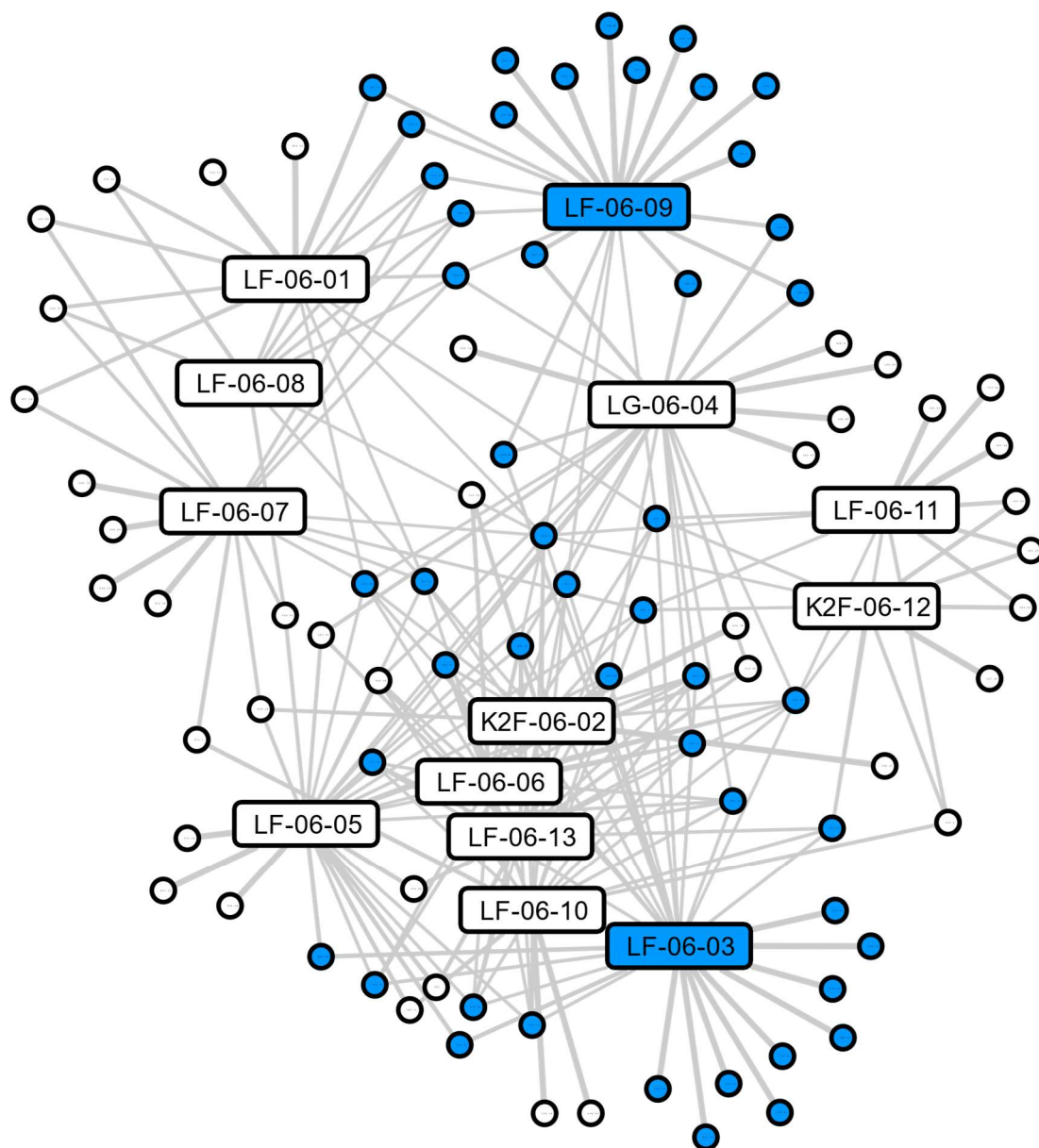


Figure 3.5. LF-06-03 has 30 associated ions, and 9 of them are unique. LF-06-09 also has 9 unique ions, and 22 associated ions total. LF-06-05 in comparison has 28 associated ions, however it was not selected for the library. Despite having more associated ions than LF-06-09, it shares many of its nodes with LF-06-03, and only has 3 unique ions. Therefore, because LF-06-09 would add more overall coverage of the network, including a large number of unique ions, it

was ultimately chosen over LF-06-05. **Table IV** lists all of the chosen strains across bacteria groups 1-12.

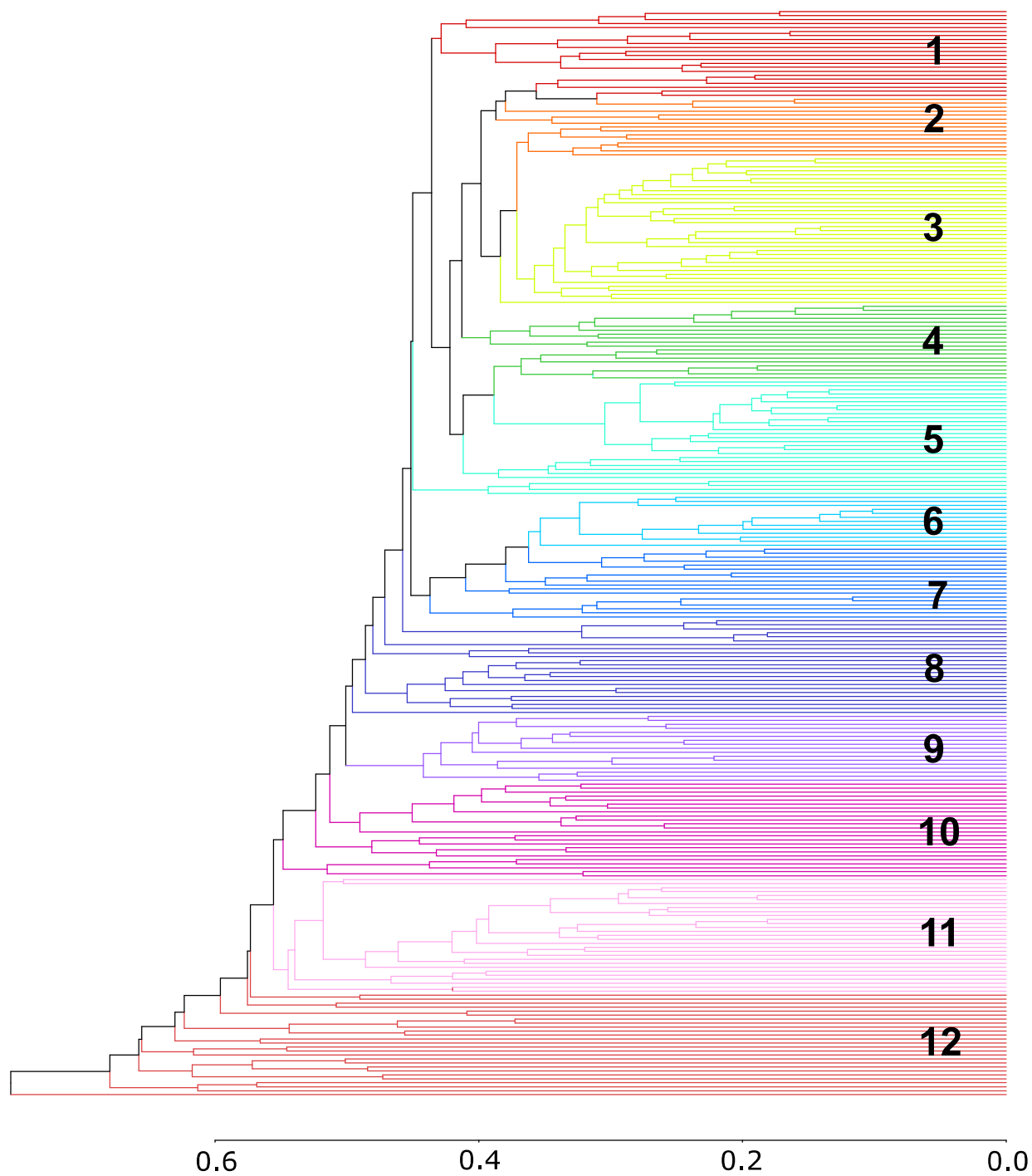


Figure 3.4. Pseudophylogenetic dendrogram created from MALDI-TOF MS protein profiling of bacteria isolates

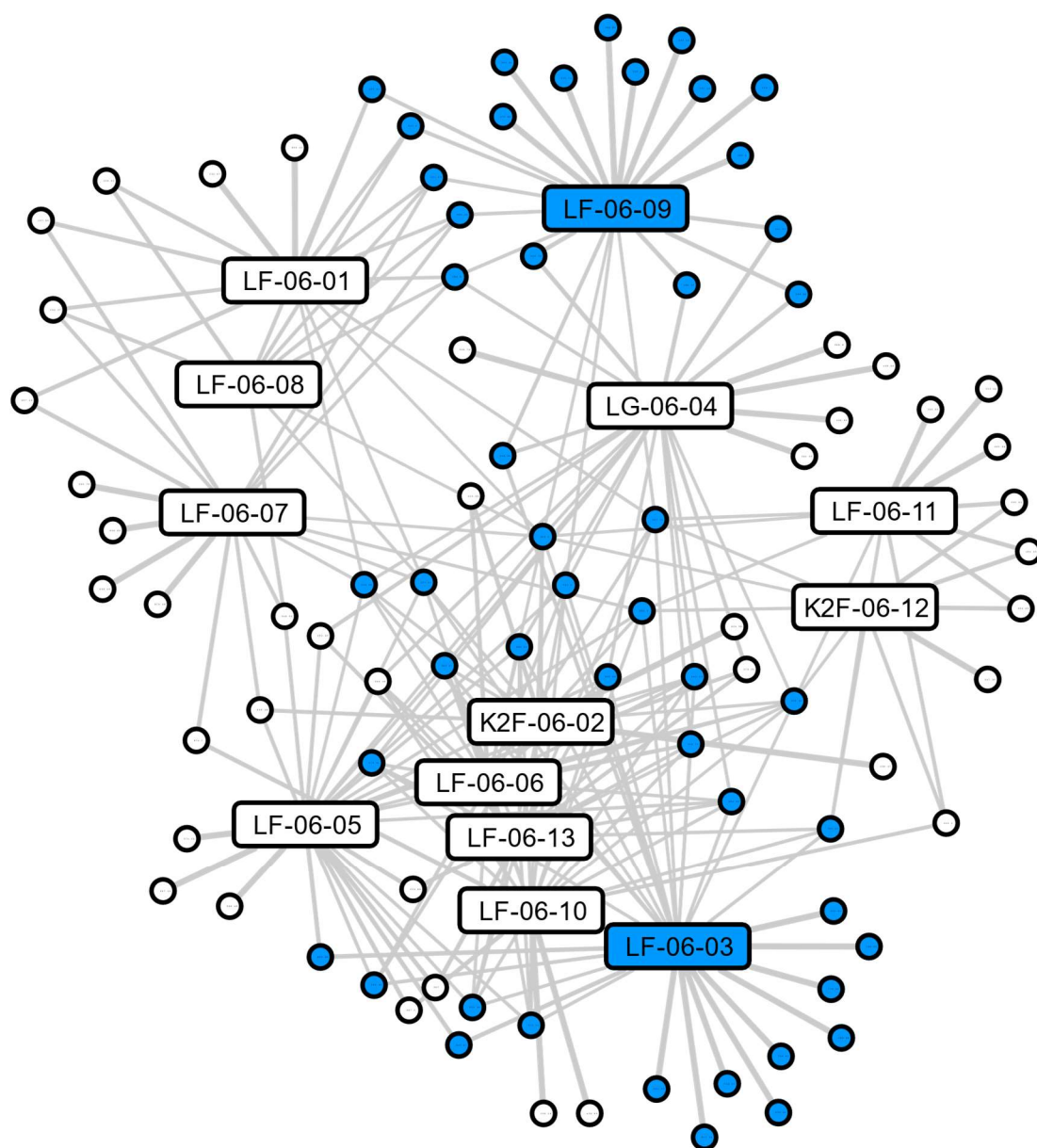


Figure 3.5. MAN of bacteria group 6. Prioritized *s* and their corresponding metabolite nodes are in blue.

TABLE IV

LIBRARY BACTERIA METADATA

Pseudo- phylogenetic Group	Isolate name	Tissue source	Media type	Plating method	First inoculated on
1	K2G-01-01	Skin	GMF	Undiluted spread	7/18/2019
2	K2G-02-09	Skin	GMF	Stamp	8/12/2019
	K2G-02-10	Skin	GMF	1:100 dilution	8/12/2019
3	K2F-03-21	Skin	SNF	1:100 dilution	7/15/2019
	K2F-03-22	Skin	SNF	Stamp	7/25/2019
4	LF-04-10	Lg. intestine	SNF	Undiluted spread	6/29/2019
	K2G-04-17	Skin	GMF	Stamp	7/12/2019
5	MG-05-21	Sm. intestine	GMF	Undiluted spread	9/7/2019
	K1F-05-03	Skin	SNF	1:100 dilution	7/2/2019
6	LF-06-09	Lg. intestine	SNF	Undiluted spread	6/29/2019
	LF-06-03	Lg. intestine	SNF	Stamp	7/15/2019
7	LF-07-13	Lg. intestine	SNF	Stamp	7/11/2019
	LF-07-13	Lg. intestine	GMF	Undiluted spread	6/29/2019
8	LG-08-03	Lg. intestine	GMF	Stamp	7/3/2019
	LG-08-18	Lg. intestine	GMF	Undiluted spread	7/2/2019
9	K2F-09-16	Skin	SNF	Stamp	7/3/2019
	MF-09-07	Sm. intestine	SNF	Stamp	9/27/2019
10	MF-10-07	Sm. Intestine	SNF	Undiluted spread	7/3/2019
	K2F-10-02	Skin	SNF	Stamp	7/3/2019
11	K1F-11-06	Skin	SNF	1:100 dilution	7/2/2019
	K2F-11-20	Skin	SNF	1:100 dilution	7/25/2019
12	LG-12-11	Lg. intestine	GMF	Stamp	8/30/2019
	LF-12-26	Lg. intestine	SNF	Stamp	7/11/2019

3.3 Conclusions

We inoculated 323 microbial isolates from the tissue of a naked mole rat, and selected a diverse set of 48 isolates to be used as an efficient drug discovery library in the future. In order to find microbial producers of unique metabolites, all of the isolates were screened via MALDI-TOF mass spectrometry to make informed correlations between the isolates' taxonomic and metabolomic profiles. It was discovered that many fungal isolates were not yet separated from bacterial stains, and further analysis will need to be done to select an ideal library before cryopreservation. This is the first reported use of sourcing the microbiome of *H. glaber* for a drug discovery library.

3.4 Experimental

3.4.1 Tissue Preparation

Two 1 cm² skin samples were dissected immediately post-mortem from the torso of a 1-year-old male *H. glaber* along with both the small and large intestine by a member of the Thomas Park lab under ACC protocol #18-064. All of the contents of the intestines were left intact. All samples were externally washed with 70% ethanol prior to removal to prevent contamination from the air or interior of the carcass. Each skin sample was added to 5 mL of 1X phosphate-buffered saline (PBS) in a 20 mL conical-bottom centrifuge tube containing 15 sterile 5-mm glass beads. Each half of the intestine was submerged in 30 mL of 1X PBS in a 40 mL conical-bottom centrifuge tube with seven to ten sterile 3-mm glass beads. All four samples were vortexed for approximately 60 seconds.

Figure 3.6 shows photos of each of the tissue samples after bead beating.



Figure 3.6. The four tissue samples, from left to right: skin, skin, small intestine, large intestine.

3.4.2 Diversity Plate Preparation

Each of the PBS solutions described in “Tissue Preparation” were plated on 24 90-mm Petri dishes containing 20 mL of agar media for a total of 96 diversity plates. Four media types were used, with 3 plating methods, all done in duplicate. The four media types were as follows: GMF (20 g starch, 0.5 g NaCl, 0.01 g $\text{FeSO}_4 \cdot 7 \text{H}_2\text{O}$, 0.5 g $\text{MgSO}_4 \cdot 7\text{H}_2\text{O}$, 0.5 g K_2HPO_4 , 1 g KNO_3 , 20 g agar, 1 L Milli-Q water) with 25 mg/L nalidixic acid, GMF with 25 mg/L cycloheximide, SNF+Fiber (20 g Garden of Life Raw Organic Superfood Fiber, 18 g agar, 1 L Milli-Q water) with 25 mg/L nalidixic acid, and SNF+Fiber with 25 mg/L cycloheximide. The PBS solution was plated using serial stamping, an undiluted spread, and a 1:100 dilution spread. The dilution was prepared before any PBS solution was plated, by transferring 100 μL of the tissue homogenate into 10 mL of 1X PBS.

Stamping was done by dipping a sterile swab into the undiluted 1X PBS dilution and lightly dabbing the swab eight times evenly spaced in a circle around the petri dish clockwise

before a final dab in the middle of the plate. For the undiluted and diluted plates, 100 μ L of solution was added to the plate and spread. One plate of each media type (GMF + antibiotic, GMF + antifungal, SNF+Fiber + antibiotic, SNF+Fiber + antifungal) was left untreated to act as a media control and monitored for growth along with the other plates.

3.4.3 Microbial Isolation

After nine days, the diversity plates were monitored for growth daily. As distinct colonies formed, the microbes were transferred to 60 mm plates of ISP-2 media (4 g yeast extract powder, 5 g malt extract powder, 4 g dextrose, 15 g agar, 1 L MilliQ water)²⁹. Bacteria were transferred via a 10 μ L sterile inoculation loop and then quadrant streaked onto a 60 mm agar plate. Fungi were transferred by removing a punch of cells, either by using the back end of a sterile 1 mL pipette tip or by slicing and removing the agar with a sterile 21-gauge sharp needle. The microbes were left to grow on a countertop until growth beyond the inoculation area was observed. If plates appeared to have more than one morphotype, they were re-isolated until pure.

3.4.4 IDBac Sample Preparation

A sterile toothpick was used to sample either a single bacterial colony or approximately a 4 mm² area of a fungal colony and transfer to a 2 mL Eppendorf tube with 5 μ L of HPLC grade trifluoroacetic acid (TFA). The cells were left to lyse for 30 minutes before the addition of 20 μ L of HPLC grade acetonitrile (ACN) and 15 μ L of MilliQ water. The cells were then vortexed for 10 seconds before being centrifuged at 10,000 rpm for 2 minutes. The resulting supernatant (2 μ L) was mixed by pipette with 2 μ L of recrystallized 10 mg/mL CHCA matrix²⁷. Three technical replicates of 1 μ L each were then spotted on a ground steel 396 target plate for analysis in both

positive linear and reflectron modes on an Autoflex Speed LRF mass spectrometer (Bruker Daltonics) equipped with a smartbeam-II laser (355 nm).

3.4.5 MALDI-TOF-MS Sample Acquisition

Samples were collected via AutoXecute methods hosted at DOI: 10.5281/zenodo.1115619. Laser size was set to 3-medium. Laser power for reflectron and linear mode was between 40-70%. Gain for reflectron mode was within 9.0x to 12.0x, and gain for linear mode ranged from 10.0x to 20.0x based on optimization after instrument maintenance. Reflectron mode was calibrated with Peptide Calibration Mix (Bruker Daltonics) and linear mode was calibrated with Protein Calibration Standard I (Bruker Daltonics).

3.4.6 IDBac Parameters

Bacteria Dendrogram: The dendrogram in Figure 3.4 was created by analyzing 273 samples and retaining peaks with a signal to noise ratio above 3.5 and occurring in greater than 70 % of replicate spectra (two of three replicates). Peaks occurring below 2001 m/z or above 15000 m/z were removed from the analyses. For clustering spectra, cosine distance and average (unweighted pair group method with arithmetic mean) algorithms were used.

Bacteria MANs: These MANs were created by subtracting a matrix blank, retaining peaks with a signal to noise ratio above 4, and occurring in greater than 70% of replicate spectra (two of three replicates). Peaks occurring below 200 m/z or above 2000 m/z were removed from the analysis.

Fungi Dendrogram: The dendrogram in Figure 3.2 was created by analyzing 50 samples and retaining peaks with a signal to noise ratio above 4.5 and occurring in greater than 60 % of

replicate spectra (one of three replicates). Peaks occurring below 2001 m/z or above 15000 m/z were removed from the analyses. For clustering spectra, cosine distance and average (UPGMA) algorithms were used.

Fungi MAN: This MAN was created by analyzing 50 samples, subtracting a matrix blank, retaining peaks with a signal to noise ratio above 4.5, and occurring in greater than 70% of replicate spectra (two of three replicates). Peaks occurring below 200 m/z or above 2000 m/z were removed from the analysis.

3.4.7 MassIVE Dataset

The full MALDI-TOF-MS dataset can be accessed at MassIVE accession number MSV000085215.

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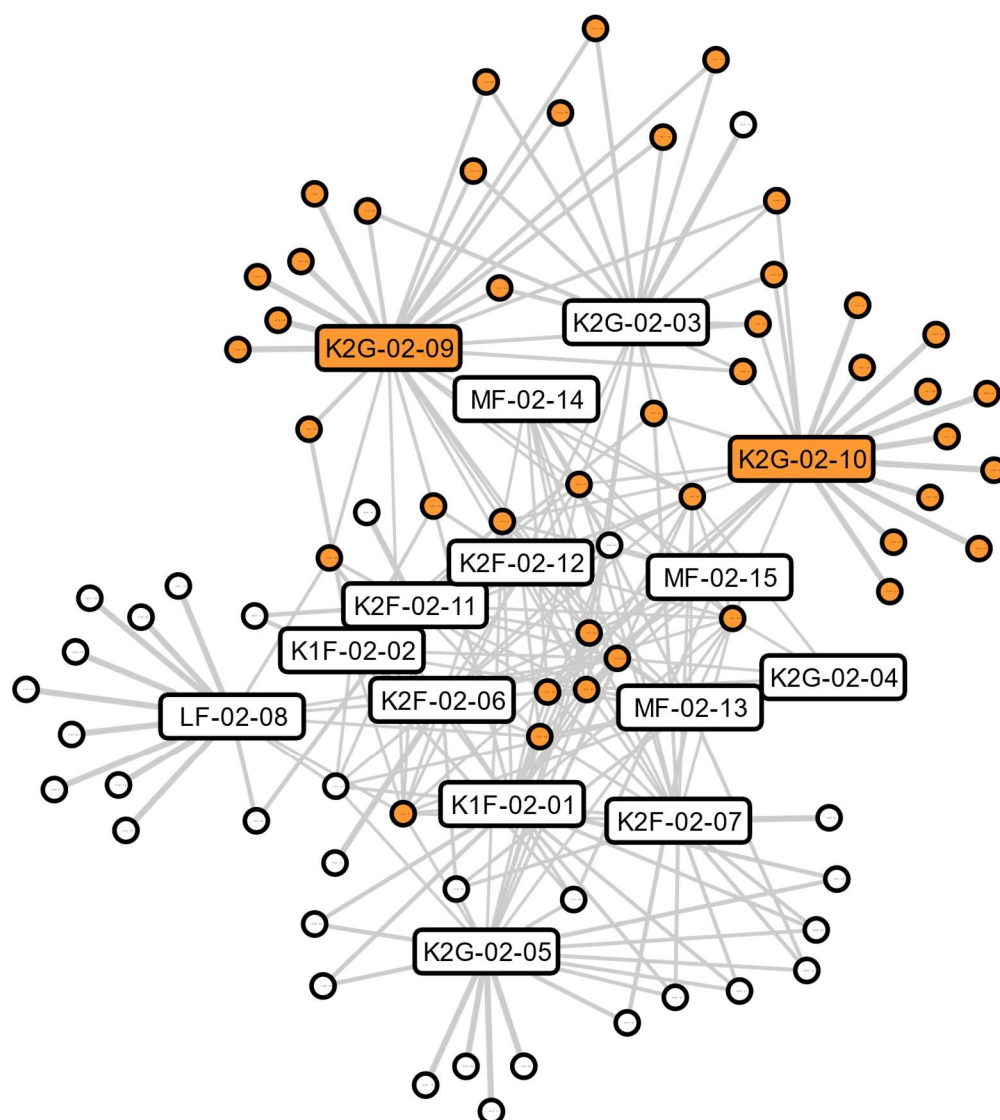


Figure A.2 MAN of bacteria group 2. Chosen strains are highlighted in orange.

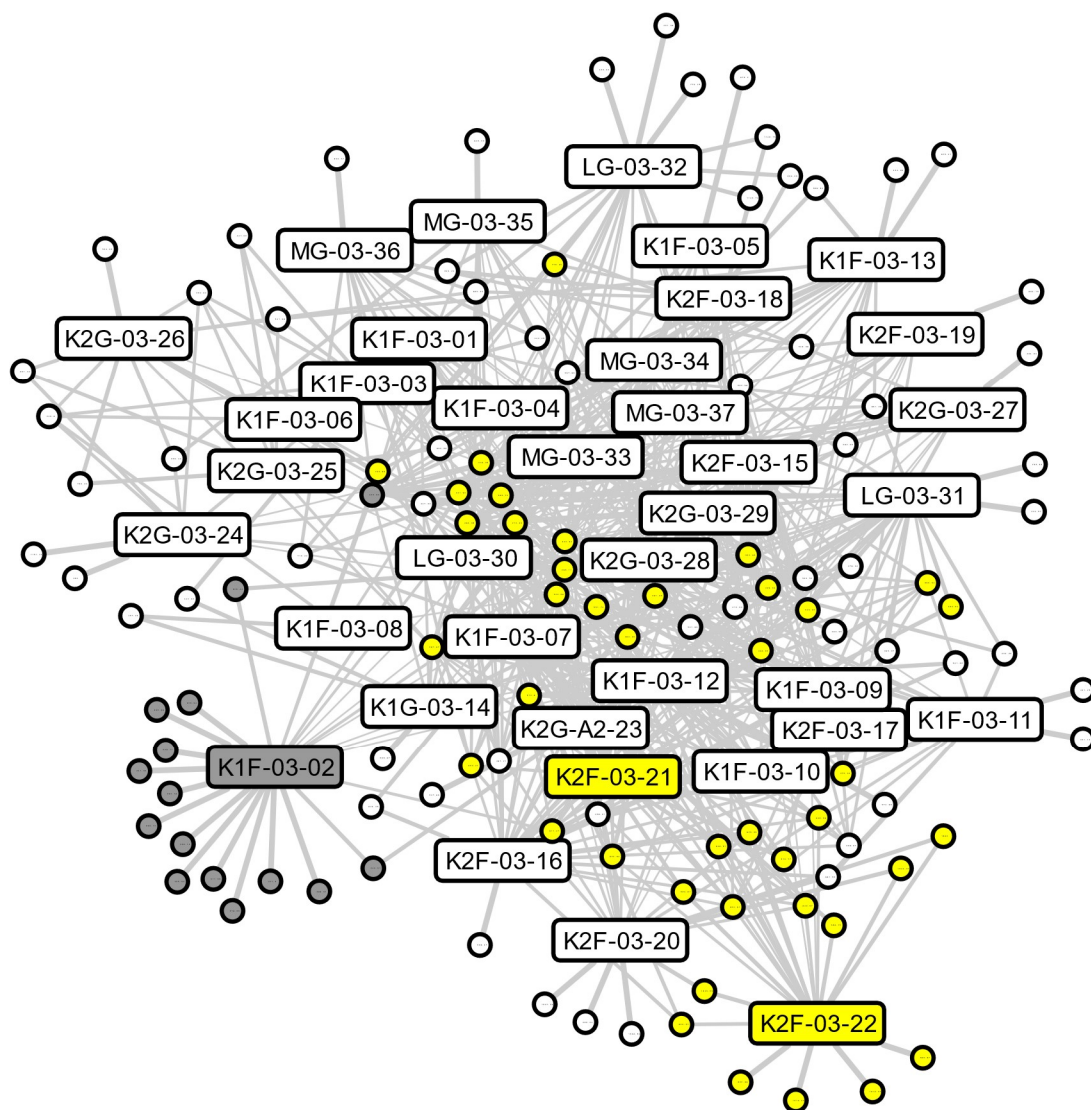


Figure A.3 MAN of bacteria group 3. Chosen strains are highlighted in yellow. Gray marks a chosen strain that failed to grow in liquid media and was discarded.

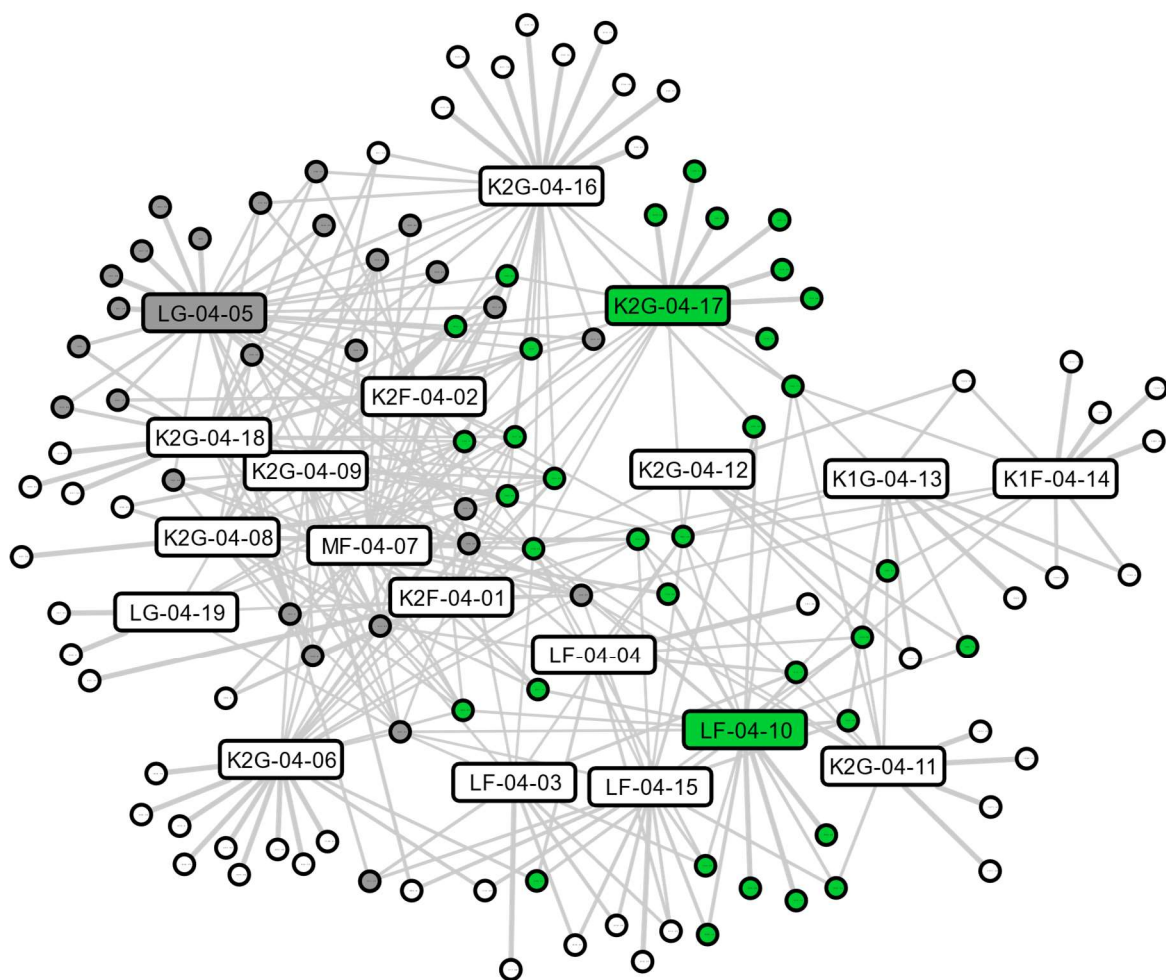


Figure A.4 MAN of bacteria group 4. Chosen strains are highlighted in green. Gray marks a chosen strain that failed to grow in liquid media and was discarded.

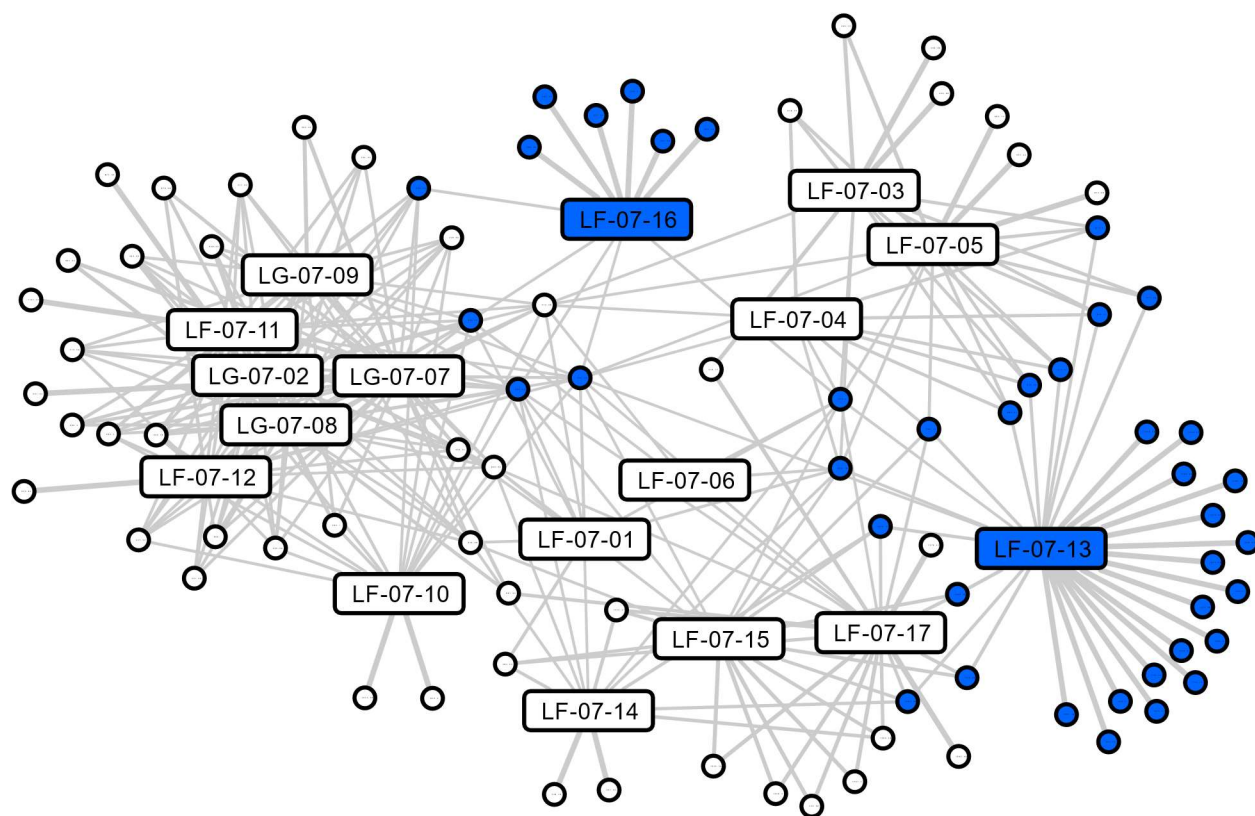


Figure A.6 MAN of bacteria group 7. Chosen strains are highlighted in blue.

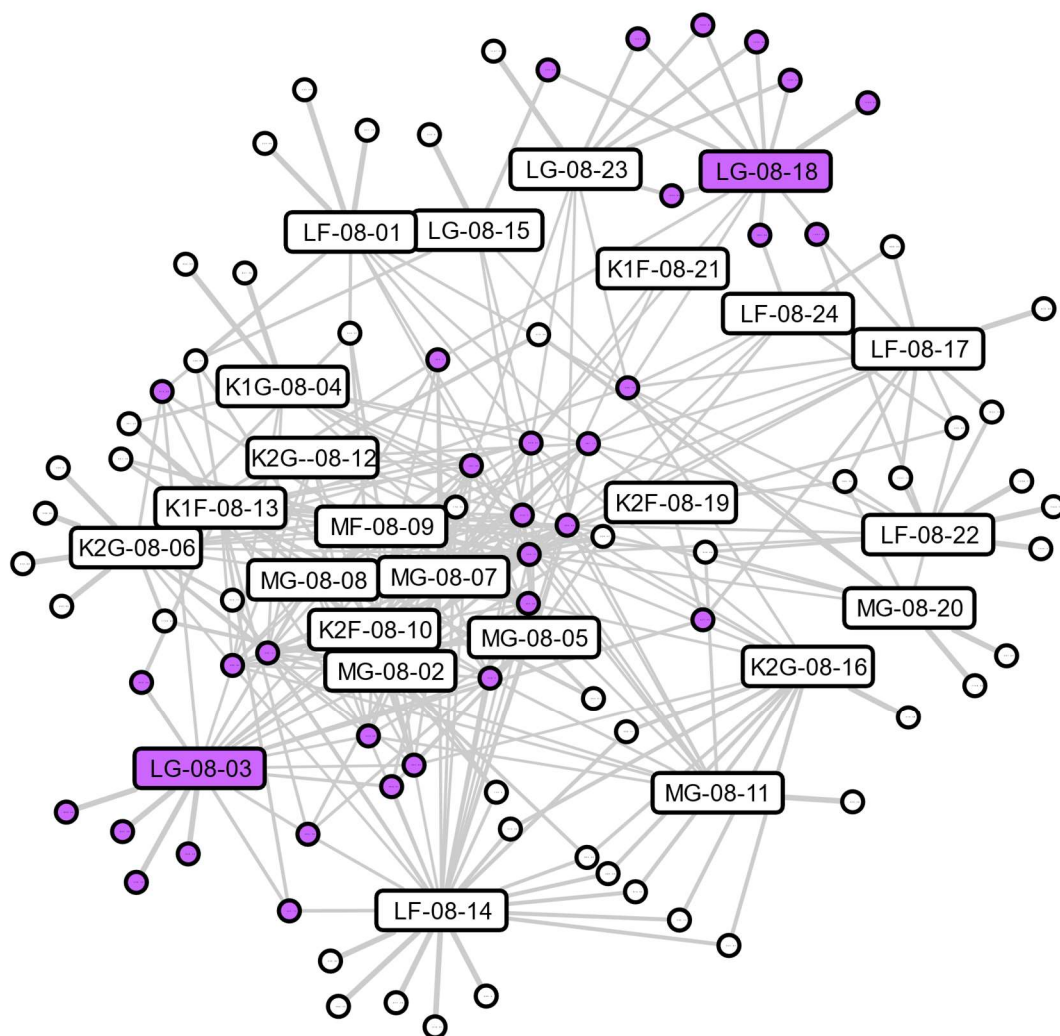


Figure A.7 MAN of bacteria group 8. Chosen strains are highlighted in purple.

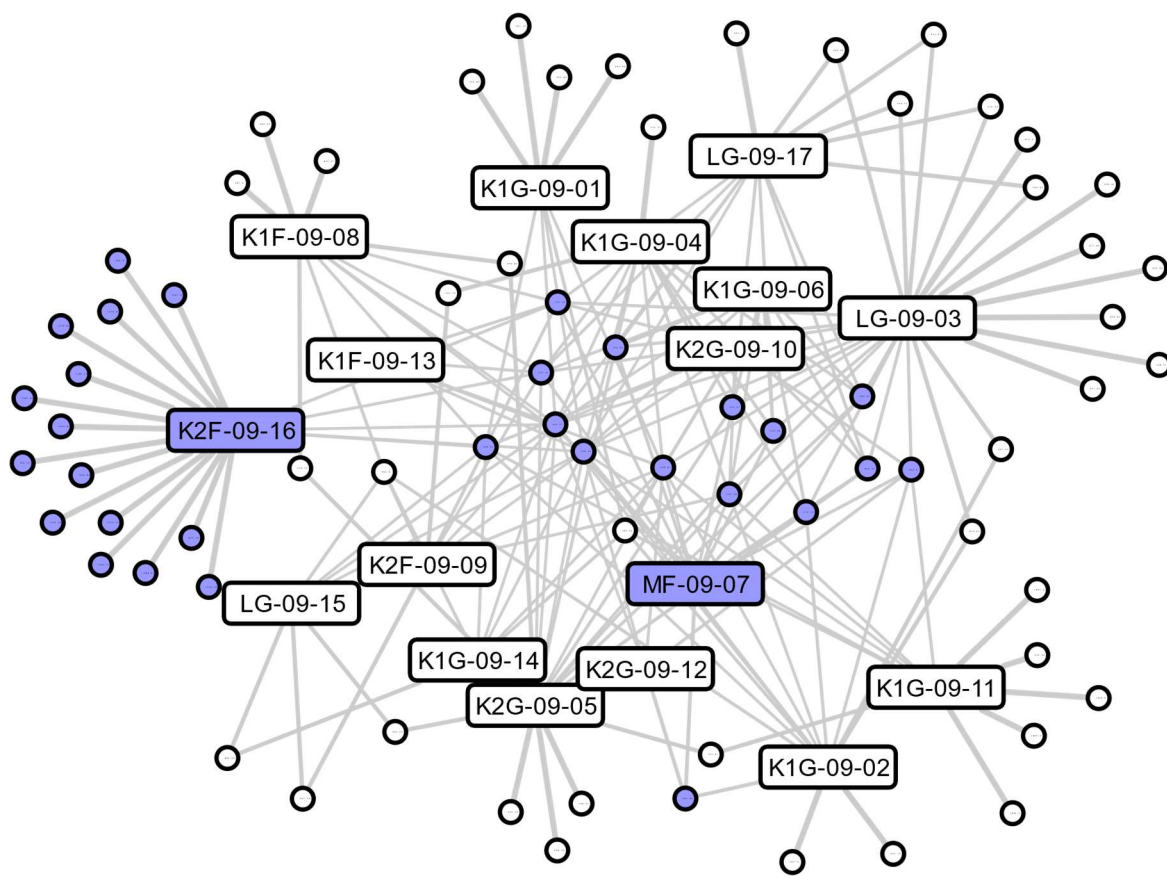


Figure A.8 MAN of bacteria group 9. Chosen strains are highlighted in periwinkle.

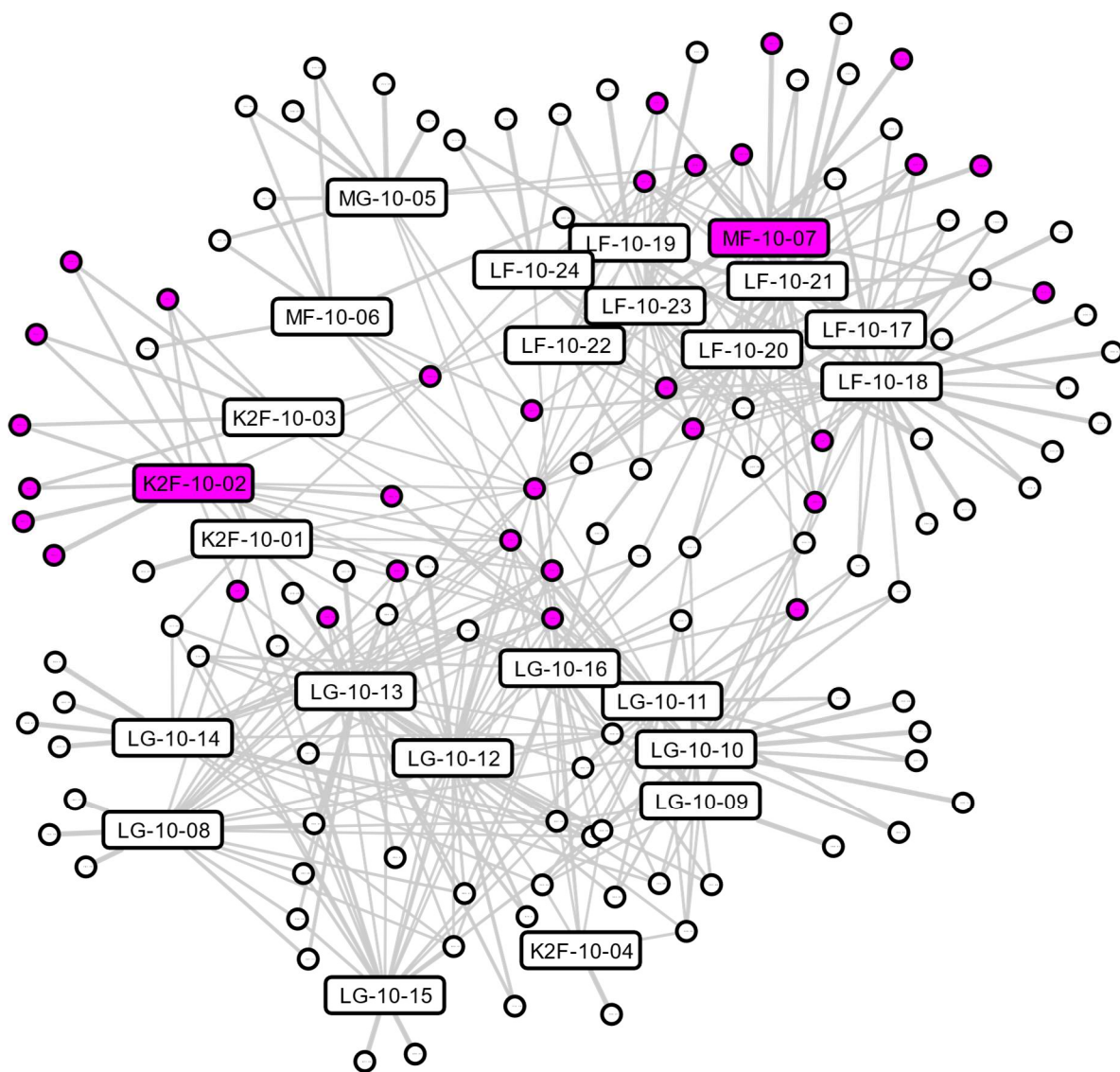


Figure A.9 MAN of bacteria group 10. Chosen strains are highlighted in magenta.

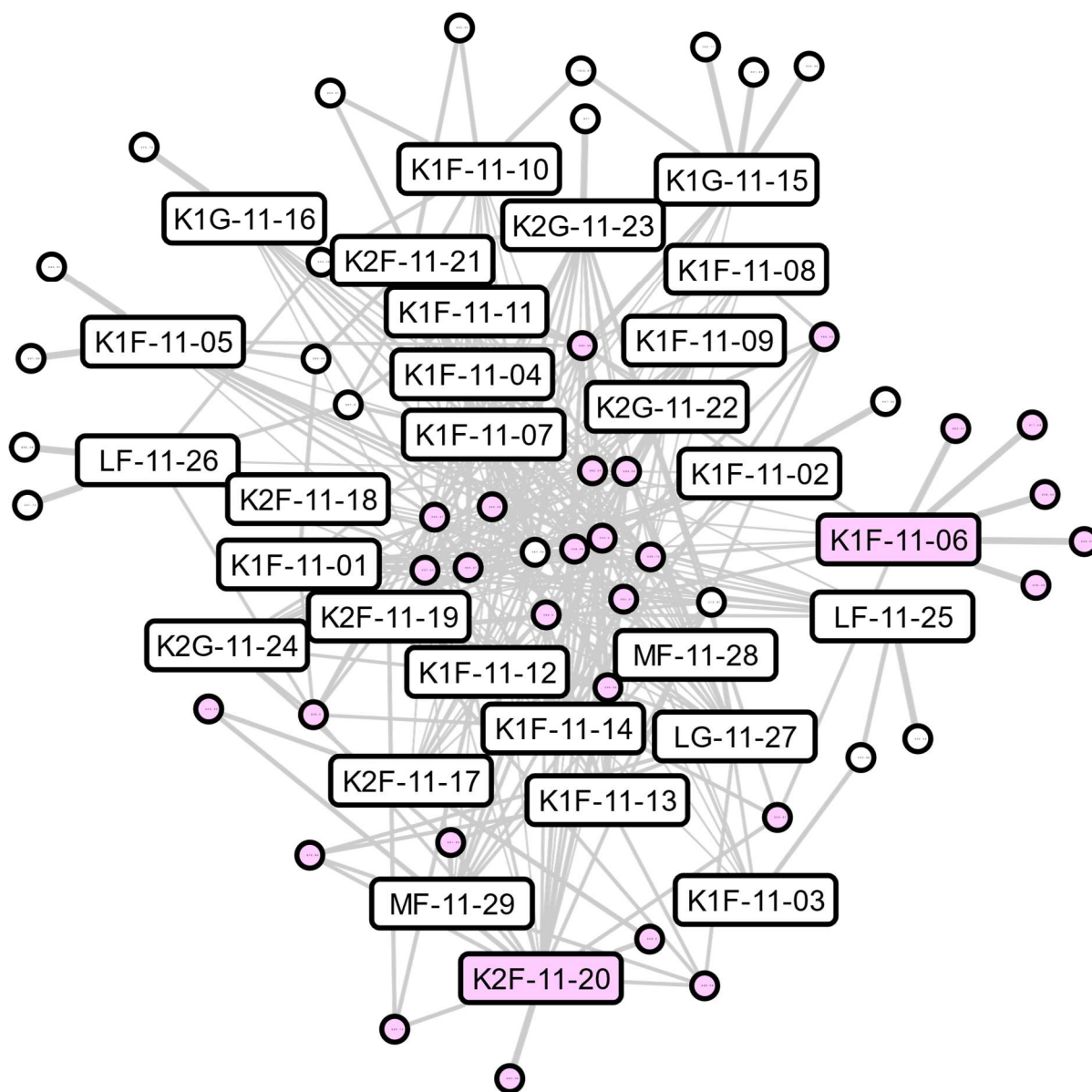


Figure A.10 MAN of bacteria group 11. Chosen strains are highlighted in pink.

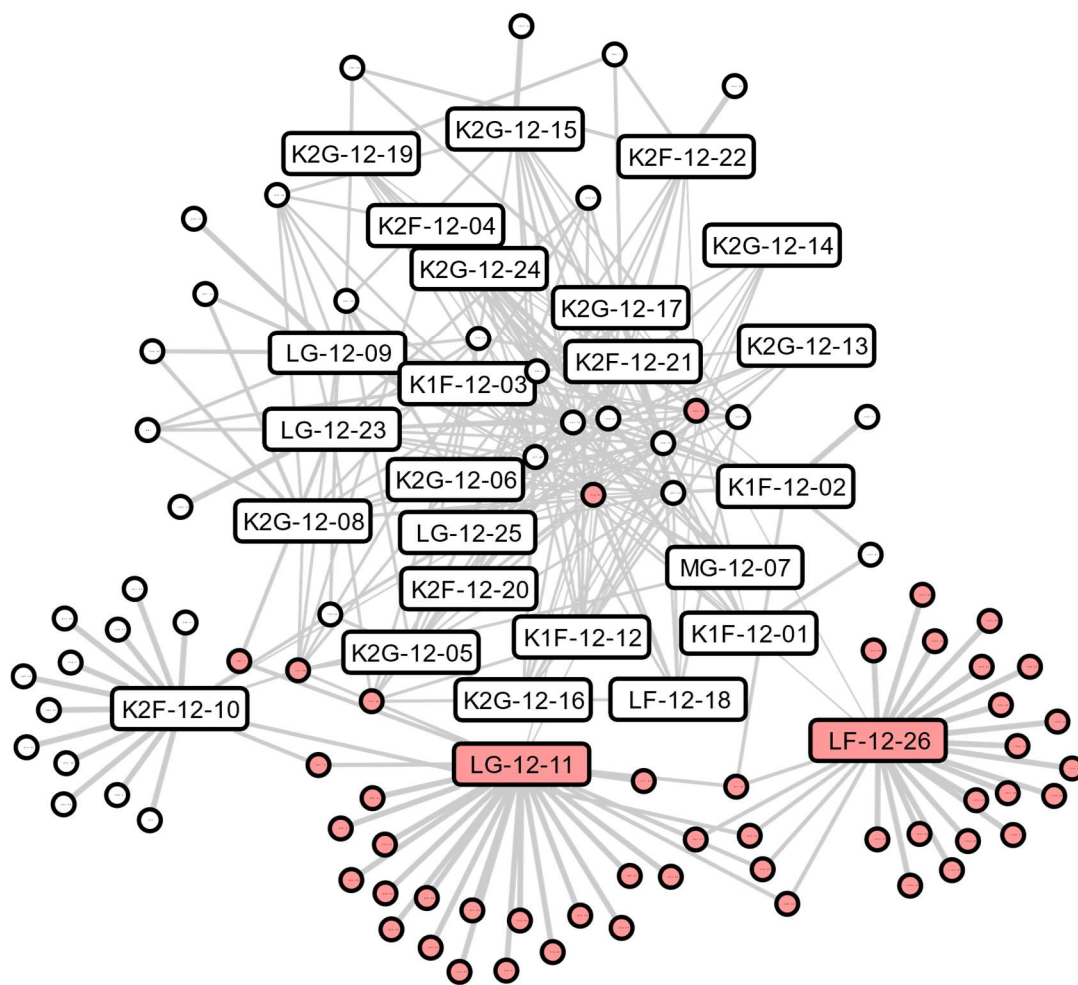


Figure A.11 MAN of bacteria group 12. Chosen strains are highlighted in pink.

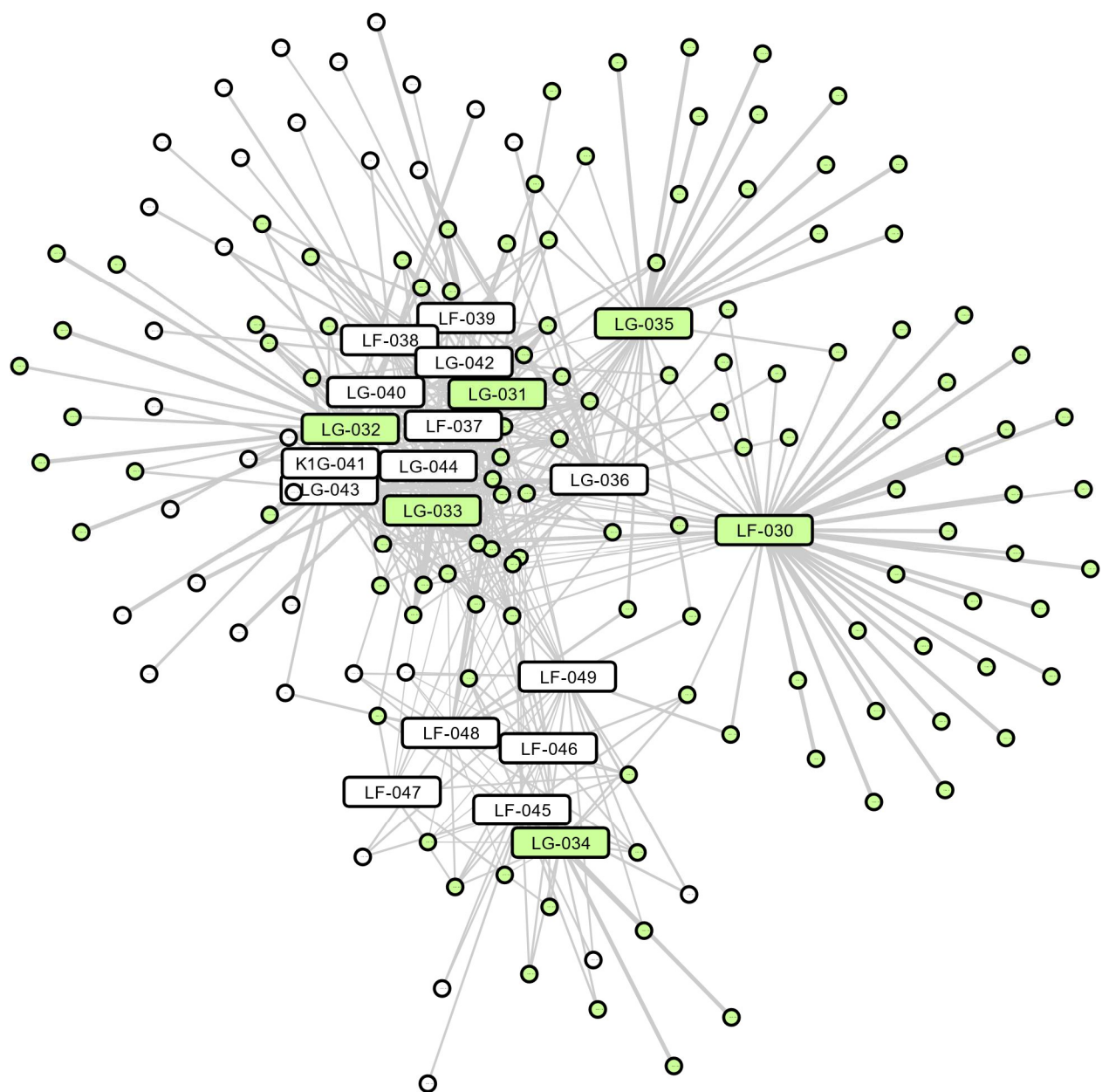


Figure A.12 MAN of Fungi group A. colored notes indicate library strains.

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Publications

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*Denotes equal contribution

Conference Presentations

1. Investigation of the metabolites in ovarian cancer using imaging mass spectrometry, Poster presentation, Chicago Mass Spec Day, July 2019
2. Systematic screening of marine macroorganisms for widespread drug discovery, joint poster presentation, USF Research and Arts Colloquium, April 2016.
3. Finalist, screening of marine macroorganisms for widespread drug discovery, poster presentation, American Institute of Chemical Engineers Southern Regional Conference, University of Alabama, April 2016.
4. Utilization of yeast-based multiplex assay to screen natural products as potential targets for endectocidal drug development, poster presentation, USF Chemistry Departmental Castle Conference, April 2015
5. Utilization of yeast-based multiplex assay to screen natural products as potential targets for endectocidal drug development, poster presentation, USF Oktoberfest Research Presentations, October 2014
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3. Honorable Mention, Raymond N. Castle Conference, USF Chemistry Departmental Conference, 2014
4. USF Honors College Discovery Scholar Scholarship recipient, 2014 – 2017
5. USF Presidential Scholarship, 2013 – 2017
6. Girl Scout Gold Award Recipient, 2013

Outreach

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SACNAS Travel Scholarship Reviewer, May 2019

NSF Outreach trial with the Boys and Girls Club, Jan 2019

Society of Women Engineers, 2014 to 2017. 2016-2017 Outreach Chair.

Give Kids the World Volunteer, 2008 to 2017.