TITLE: Functional metagenomic discovery of novel tetracycline and acarbose resistance

genes from low biomass samples using METa assembly

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ABSTRACT

A significant challenge in the field of microbiology is the functional annotation of

sequence novel genes from microbiomes. The increasing pace of sequencing technology

development has made solving this challenge in a high-throughput manner even more

important. Functional metagenomics offer a sequence-naïve and cultivation-independent

solution. This forward genetics approach relies on the creation of functional metagenomic

libraries (aka shotgun cloning) in which a microbial host such as *E. coli* is transformed with

vectors containing metagenomic DNA fragments and, optimally, expresses any captured genes

into a corresponding phenotype. These libraries can be screened or selected for a function of

interest, such as antibiotic resistance, allowing the captured metagenomic DNA to be linked to a

phenotype regardless of the sequence's novelty. Unfortunately, most methods for constructing

functional metagenomic libraries require large input masses of metagenomic DNA, putting many

sample types off limits to this toolset. Here, we show that our recently developed functional

metagenomic library preparation method, METa assembly, can be used to prepare useful

libraries from much lower input DNA masses. Standard methods of functional metagenomic

library preparation generally call for 5 µg to 60 µg of input metagenomic DNA. Here, we

demonstrate that the threshold for input DNA mass can be lowered at least to 30.5 ng, a three-

log decrease from prior art. These functional metagenomic libraries, prepared using between

30.5 ng and 100 ng of metagenomic DNA, nonetheless were sufficient to link three MFS efflux

pumps to tetracycline resistance and capture two potential genes for degradation or resistance

to the antidiabetic pharmaceutical acarbose. Our preparation of functional metagenomic

libraries from aquatic samples and a model fecal swab demonstrate that METa assembly can be

used to prepare functional metagenomic libraries from microbiomes that were previously

incompatible with this approach. Functional metagenomic screens and selections are one of the

few high-throughput methods that can link novel genes to functions and here we show that one

of their significant drawbacks, a requirement for large amounts of metagenomic DNA, can now

be overcome.

Figure 1. Functional metagenomics and METa assembly. A) General steps in the preparation and application of a functional metagenomic library: 1) Extraction of metagenomic DNA from a source microbiome. 2) Fragmentation of metagenomic DNA to preferred size range. 3) Packaging of inserts into expression vectors. 4) Transformation of host cells with vector library. 5) Screen or selection of functional metagenomic library for a phenotype of interest. 6) Collection of screened or selected metagenomic fragments. 7) Sequencing of selected inserts. 8) Open reading frame calling and annotation to identify potential genes underlying phenotypes of interest. B) Steps 2) and 3) above are modified in Mosaic ends tagmentation (METa) assembly. Fragmentation is achieved using Tn5 transposase tagmentation with mosaic end sequence oligos. Tagmented DNA is gap filled by polymerase and cloned into an expression vector with matching mosaic end sequences defining the cloning site using assembly cloning (*e.g.,* Gibson assembly). C) and D) The chemical structures of the antibiotic C) tetracycline and the diabetes pharmaceutical D) acarbose.

INTRODUCTION

Microbiomes, assemblages of bacteria, fungi, viruses and other microorganisms, are

host to some of the highest levels of genetic diversity on the planet. Due to the difficulty in

culturing all but a select low percentage of organisms in microbiomes (often estimated to be

~0.1% to 1%) current methods for studying microbiomes rely on DNA extracted from these

communities, metagenomic DNA[1–4]. High-throughput sequencing of metagenomic DNA provides

unrivaled insights into the taxonomic structure of a microbial community and can shed

significant light on its functional capacity as well[3]. However, even with the sequencing

revolution, the genetic diversity of microbiomes has outpaced our ability to accurately annotate

novel microbial genes. Complementing homology and structure-based methods for assigning

function to novel genes is the high-throughput, sequence- and cultivation-naïve method of

functional metagenomics. In this method, functional metagenomic libraries, sometimes referred

to as shotgun cloning libraries, are prepared by fragmenting metagenomic DNA and cloning

those fragments *en masse* into an expression microorganism. Once a library is prepared, the

theoretical goal is that any genes encoded within captured metagenomic DNA fragments are

expressed by the host cell to give a phenotype that can be selected or screened for[4–7] (**Figure**

1A).

While functional metagenomics is a powerful method for linking novel genes to functions

without requiring the growth of non-model organisms in the lab, it has its own shortcomings.

These issues include the potential for gene toxicity, limited expression of foreign genes in model

library hosts such as *E. coli*, limited coverage of target metagenomes, and requirements for

large quantities of metagenomic DNA for preparation of the functional metagenomic library itself.

We recently reported the development of a new method for preparing small insert functional

metagenomic libraries (*i.e.,* 2 kb to 5 kb sized inserts) that we termed METa assembly[8]. This

functional metagenomic library preparation method relies on near-random tagmentation to

fragment input metagenomic DNA and an assembly-based cloning step to replace blunt ligation

(**Figure 1B**). We previously demonstrated that this method circumvents one shortcoming of

functional metagenomics, limited metagenomic coverage, by showing that META assembly

libraries are more efficient at capturing DNA and can produce approximately 80-fold larger

libraries per input DNA mass. We also showed that META assembly could be used to prepare

functional metagenomic libraries encoding tens of gigabase pairs (Gb) of captured DNA using

sub-μg inputs of metagenomic DNA, far lower than the standard 5 μg to 20 μg input[8,9].

Here we build on our earlier results by determining if META assembly can be used to

prepare functional metagenomic libraries with significantly lower quantities of input DNA. We

found that we could consistently prepare functional metagenomic libraries starting with 50 ng or

less input DNA or even DNA extracted from a swab dipped into a standardized human fecal

sample. These libraries were large enough that when selected for resistance to the antibiotic

tetracycline and the anti-diabetic acarbose they yielded novel predicted proteins. These results

suggest that many more rare or low biomass microbiomes can be studied by functional

metagenomics than previously appreciated.

METHODS

Strains and materials

Routine cultivation of *E. coli* was performed at 37°C with shaking at 250 rpm for aeration

in the case of liquid cultures. Cultivation for cloning and starter cultures was performed using

lysogeny broth (Miller) (LB) with the addition of kanamycin at 50 μg/ml to maintain plasmids.

Antimicrobial susceptibility testing was performed using Cation-Adjusted Mueller-Hinton media

(Teknova, 101320-364). All *E. coli* strains used in this study are of the DH10B lineage and carry

plasmids derived from pZE21[8,10,11]. *E. coli* strains were maintained at -80°C in 50% glycerol in

LB. Transposase enzymes and mosaic ends oligos were previously prepared and stored at -

20°C. Standard chemicals used in this study were of high purity for molecular biology and

microbiology. Kanamycin sulfate (VWR Life Science, 75856-68), acarbose (Thermo Scientific

Chemical, 45908-0010), tetracycline (Aldon Corp, TT0070-5GR), ciprofloxacin hydrochloride

(Corning, 61-277-RG), azithromycin dihydrate (TCI, A2076), and florfenicol (Fisher Scientific,

F08115G) were stored at 4°C as powders. Working solutions were stored at -20°C as filtered

aqueous solutions at 50 mg/ml concentration (kanamycin and acarbose), 5 mg/ml in ethanol

(tetracycline), 10 mg/ml in alkaline water (ciprofloxacin), or 50 mg/ml in DMSO (azithromycin

and florfenicol).

Preparation of 50 ng test functional metagenomic libraries

To determine if agarose extraction could be used to purify and size-select DNA following

low mass tagmentation reactions, we used a DNA ladder (ThermoFisher Scientific 1 kb plus,

10787018) to mimic fragmented DNA. Mock tagmentation reactions were made with a final DNA

concentration of 50 ng/µl by combining 1 µl of diluted ladder DNA with 3 µl of autoclaved milliQ

water and 1 µl of 5X Taps tagmentation buffer (50 mM TAPS pH 8.5, 25 mM $MgCl_2$, 50% v/v

DMF in water)[12]. We added 0.25 µl of 1% sodium dodecyl sulfate (SDS) to the 5 µl mock

reactions to simulate reaction quenching followed by 1 µl of 6x DNA dye (15% Ficoll 400 w/v,

0.25% Orange G w/v in water) before loading on to a 0.7% agarose gel in TAE buffer containing

SybrSafe dye (ThermoFisher Scientific, S33102). DNA bands were visualized over blue light

and a clean razor was used to cut out gel portions containing bands greater than 2 kb in length.

DNA was extracted from the agarose gel fragments using a gel extraction kit (New England

Biolabs, T1020L) following manufacturer instructions with modification. Each agarose fragment

was weighed and incubated with four volumes of gel dissolving buffer for at least 10 minutes at

55°C. The dissolved agarose was then applied to the kit's spin column, eluted, then re-applied a

second time. The standard protocol was followed until the elution step where each column was

eluted twice sequentially with 6 μl of 55°C autoclaved milliQ water. DNA concentration was

measured using the Quant-It DNA quantification system (Invitrogen).

To test the preparation of functional metagenomic libraries using 50 ng of input

metagenomic DNA, we used a sample of previously extracted metagenomic DNA from a

Canada goose fecal pellet that was stored at -20°C[8]. For library preparation, we followed our

previously published METa assembly protocol[8]. Briefly, triplicate 5 μl tagmentation reactions

containing 50 ng of metagenomic DNA, 1x TAPS-DMF buffer, and 25 ng of in-house

transposase[12] loaded with mosaic end sequence oligos were incubated at 55°C for 7 minutes

followed by quenching by addition of SDS to reach 0.05% final concentration and incubation for

5 more minutes at 55°C. To each reaction, 2 μl of 6x loading dye was added and fragmented

metagenomic DNA in the range of 2 kb to 9 kb was extracted by gel excision as described

above. The 12 μl of size-selected metagenomic DNA fragments were gap-filled by the addition

of 12 μl of 2x Q5 DNA polymerase (New England Biolabs, M0492S) (pre-incubated at 98°C for

30 seconds) and held at 72°C for 15 minutes before purification by PCR and DNA cleanup kit

(New England Biolabs, T1030S) according to manufacturer instructions with the following

modifications: DNA dissolved in two volumes of binding buffer was applied to the spin column

and the flow-through was applied a second time. At elution, we used two sequential applications

of 6 μl of 55°C autoclaved milliQ water. Metagenomic DNA inserts were cloned into our plasmid

pZE21-ME using 2x NEBuilder HiFi assembly mix (New England Biolabs, E2621L) according to

manufacturer instructions (assuming an insert DNA mass of 15 ng and an average fragment

size of 2 kb). The triplicate libraries were purified by PCR and DNA cleanup kit and the full 12 μl

of purified library DNA was introduced into 25 μl of *E. coli* DH10B electrocompetent cells (New

England Biolabs, C3020K) by electroporation at 1.8 kV in 0.1 mm cuvettes. Actual voltage

reached and the electroporation time constant were recorded for all electroporation reactions to

monitor for failed or low efficiency transformations. All three libraries were rescued post-

transformation by inoculation into 1 ml of SOC outgrowth medium (New England Biolabs,

B9020S) pre-warmed to 37°C and incubated with aeration at 37°C for 1 hour.

Following recovery, 100 µl of 100-fold, 10,000-fold, and 1,000,000-fold diluted cultures

were plated on to LB agar with kanamycin and incubated overnight at 37°C to estimate the

number of unique clones resulting from each transformation. The remaining cells (~1 ml) were

inoculated into 50 ml of LB with kanamycin and incubated at room temperature overnight with

aeration to amplify the functional metagenomic library. Once cultures reached a 1 cm pathlength

optical density at 600 nm (OD_{600}) of between 0.6 and 1.0, cells were collected by centrifugation

at 4,000 rcf for 7 minutes at 4°C. Pellets were resuspended in 10 ml of LB containing kanamycin

and 15% glycerol and stored at -80°C as 1 ml aliquots. The triplicate functional metagenomic

library sizes were determined using the Dantas lab calculator (http://

dantaslab.wustl.edu/LibSizeCalc/) considering unique cell count, determined above, average

metagenomic insert size, and the proportion of clones containing an empty plasmid. This

proportion and the average insert size were determined by performing colony PCR as

previously described on 9 clones from each replicate library[8].

Preparation of an aquatic functional metagenomic library from a Shedd Aquarium

microbiome

Preparation of a functional metagenomic library from aquatic metagenomic DNA began

with a previously extracted metagenomic DNA sample. The initial sample was collected from the

Shedd Aquarium in Chicago in the Underwater Beauty exhibit in May, 2019 and metagenomic

DNA was extracted[13]. A functional metagenomic library was prepared from 50 ng of DNA as

described above and evaluated by colony counting and colony PCR (n=14 colonies) to

determine library size. After preparation of the functional metagenomic library, quantification of

the input DNA by Quant-It revealed that only 30.5 ng of metagenomic DNA had been used as

input for this library and this corrected DNA mass was used in determining library preparation

efficiency.

Functional metagenomic selection for antibiotic resistance

The aquatic functional metagenomic library was selected for resistance to the antibiotics

tetracycline, ciprofloxacin, and florfenicol. Cation-Adjusted Mueller-Hinton agar plates containing

50 µg/ml kanamycin were prepared to contain each antibiotic at the following concentrations

previously established to be inhibitory to *E. coli*[14]: 8 µg/ml tetracycline, 0.5 µg/ml ciprofloxacin,

and 8 µg/ml florfenicol. An aliquot of the functional metagenomic library was thawed from -80°C

storage and a volume of suspension calculated to contain approximately 10-fold more total cells

than the unique cell count was plated on each antibiotic and incubated overnight at 37°C. A

parallel series of agar plates were plated with a similar cell count of *E. coli* cells containing only

empty vectors as a control.

Following incubation, 15 colonies each from the tetracycline and florfenicol selections

were collected and re-streaked onto LB agar plates containing kanamycin and either

tetracycline or florfenicol to obtain pure cultures of resistant strains. Plasmids were extracted

from each strain by miniprep kit (New England Biolabs, T1010L) and whole plasmid sequencing

was performed by Plasmidsaurus using Oxford Nanopore Technology with custom analysis and

annotation. The metagenomic inserts were identified by the presence of flanking mosaic end

sequences (5'- agatgtgtataagagacag-3') and clustered on the Clustal Omega webserver[15] to de-

replicate metagenomic fragments that were captured in the selection more than once. The

metagenomic inserts targeted for further study (TET1, TET3, TET13) were analyzed by the

MetaGeneMark program[16] to identify potential open reading frames which were subjected to

analysis by Basic Local Alignment Search Tool (BLAST)[17,18] against the NCBI non-redundant

database and The Comprehensive Antibiotic Resistance Database (CARD)[19,20]. The annotated

metagenomic DNA fragments were made into schematics using ApE plasmid editor[21] and

Microsoft PowerPoint.

Preparation of a functional metagenomic library from a simulated fecal swab

An aliquot of the ZymoBIOMICS fecal reference standard (Zymo Research, D6323) was

used as a model source for a fecal microbiome swab functional metagenomic library. An aliquot

of fecal standard was thawed from -80°C storage and a sterile swab was dipped into the tube

and gently wrung out against the tube's side as it was removed. Metagenomic DNA was

extracted from the swab using a PowerSoil Pro kit (Qiagen, 47014). The swab was agitated in

800 µl of C1 solution to release material and DNA extraction was performed following

manufacturer instructions with the eluted DNA reduced in volume using a vacuum concentrator.

A functional metagenomic library was prepared by METa assembly as above, with the

tagmentation reaction scaled to match the 100 ng of metagenomic DNA available from the

extraction. Library size was evaluated by colony count and colony PCR as above (n=14

colonies).

Functional metagenomic selection for acarbose modifying genes

To determine conditions appropriate for a functional metagenomic selection for acarbose

modifying genes, we performed an agar dilution assay on minimal media plates containing

maltose as the sole carbon and energy source with acarbose concentrations ranging from 0.5

µg/ml to 256 µg/ml in two-fold jumps[22]. The agar contained 0.4% w/v maltose in a modified

MDG minimal medium[23] consisting of 25 mM Na_2HPO_4, 25 mM KH_2PO_4, 50 mM NH_4Cl, 5 mM

Na_2SO_4, 2 mM $MgSO_4$, 0.2x trace metals solution (Teknova, T1001) and was supplemented with

50 µg/ml leucine *E. coli* DH10B strains are auxotrophic for leucine. Each agar plate was

inoculated with 100 µl of a 0.5 McFarland unit suspension of *E. coli* DH10B harboring an empty

pZE21-ME plasmid. Plates were incubated at 37°C for roughly 48 hours then photographed to

record growth.

A functional metagenomic selection for acarbose modifying genes was carried out as

above with the Zymo fecal library plated to carry an estimated ten-fold excess of cells over

unique clones. The functional metagenomic library was washed once in modified MDG media

with no carbon added by pelleting at 8,000 rcf for 5 minutes, discarding the supernatant, and

resuspending in no carbon media. The washed library was then plated on a minimal media agar

plate containing 0.4% maltose as the sole carbon source supplemented with 50 µg/ml

kanamycin and 256 µg/ml acarbose. In parallel, a suspension of a control *E. coli* strain with

empty vector was washed as above and plated on maltose minimal media agar with acarbose at

a similar predicted titer. The plates were incubated at 37°C overnight. A weather-related power

outage and campus closure resulted in the plates incubating at room temperature for multiple

days. Eight colonies from the Zymo fecal library plate that were judged to be substantially larger

than background colonies on the empty plasmid control plate were picked, re-streaked to obtain

pure cultures, and processed by miniprep kit to obtain their plasmids for sequencing as above.

The plasmid sequences were analyzed as above and the most prevalent sequence (ACA1, the

S-layer homology domain-containing protein) and a sequence with homology to α-amylase

family glycosyl hydrolase (ACA2) were chosen for further study. The predicted amino acid

sequence of ACA1 was further submitted to the Contrastive Learning-Enabled Enzyme

Annotation (CLEAN) webserver for potential annotation[24].

Phylogenetic analyses

For phylogenetic analysis of the putative tetracycline efflux pumps (TET1, TET3, and

TET13), we downloaded CARD amino acid sequences[19,20] for the ontology term "major

facilitator superfamily (MFS) antibiotic efflux pump" and added the functionally-selected efflux

pump amino acid sequences to the FASTA file. For analysis of the predicted acarbose

resistance proteins (ACA1 and ACA7), we compared them to sequences for the *Klebsiella*

grimontii Apg enzyme and a *Raoultella ornithinolytica* α-glycosidase from the same

manuscript[25], additional reported acarbose degradation enzymes (maltogenic amylases from

Thermus sp. IM6501, *Geobacillus strearothermophilus*, and a glycoside hydrolase from

Lactiplantibacillus plantarum)[26–29], as well as acarbose phosphotransferases AcbK

(*Actinoplanes* sp. SE50/110), GacK (*Streptomyces glaucescens*), CatK (*Streptomyces*

coelicoflavus ZG0656), and Mak1 (uncultured bacterium)[30]. Phylogenetic trees were prepared

by aligning the protein sequences using the Clustal Omega web server with default settings[15]

and preparing a maximum likelihood phylogenetic tree from the alignment using IQ-TREE[31,32]

with the "Find best and apply" substitution model, standard 100 replicates bootstrap and

otherwise default settings. The resulting trees were visualized using the Interactive Tree of Life

display and annotation tool[33].

Microbroth dilution antimicrobial susceptibility testing

The antimicrobial resistance of *E. coli* strains expressing metagenomic fragments

encoding TET1, TET3, and TET13 were quantified by microbroth dilution assay[22,34]. Briefly, the

corresponding strains and control *E. coli* carrying an empty plasmid were streaked for single

colonies on LB agar with kanamycin. Approximately 10 colonies for each strain were suspended

in 2 ml of MH media with kanamycin and brought to an OD_{600} of between 0.1 and 0.15 to

correspond to a 0.5 McFarland standard density, then diluted 100-fold in MH broth with

kanamycin. We prepared a 96-well plate (Costar, 3370) to contain 50 µl of MH broth with 50

µg/ml kanamycin and 2-fold increasing gradients of tetracycline, chloramphenicol, or

azithromycin at twice the final target concentration and added 50 µl of the diluted bacterial

suspensions with each strain assayed in quadruplicate. The 96-well plates were sealed with a

Breathe-Easy membrane (MilliporeSigma, Z380059) and incubated at 37°C with shaking for 20

to 24 hours. Culture density was recorded as OD_{600} with a 1 cm pathlength correction using an

Epoch 2 plate reader (BioTek) and concentration-response curves were fitted to four parameter

Hill equations in GraphPad Prism 10.2.3 (GraphPad Software, La Jolla, CA, USA) to determine

50% inhibitory concentration (IC_{50}) values. Differences between IC_{50} values for control *E. coli*

and the tested strains (TET1, TET3, TET13) were evaluated for significance in GraphPad Prism

using Brown-Forsythe ANOVA tests with multiple comparison correction using the Dunnett T3

method.

Acarbose growth inhibition

The growth of a control *E. coli* strain and strains carrying the ACA1- or ACA7-containing

metagenomic DNA fragments were recorded as OD_{600} culture densities following 24 hours of

aerobic growth at 37°C using a BioTek Epoch 2 plate reader. The three strains to be tested were

prepared as above to give a 100-fold dilution of 0.5 McFarland suspension in modified MDG

minimal media without a carbon source added. We prepared a 96-well plate where each row

contained 50 µl of a different media condition (all wells containing kanamycin, acarbose

concentrations are twice the final concentration): MH ± 512 µg/ml acarbose, modified MDG with

0.4% glucose ± 512 µg/ml acarbose, modified MDG with 0.4% maltose ± 512 µg/ml acarbose,

and modified MDG with no carbon added. The cultures were inoculated in quadruplicate for

each condition by adding 50 µl of bacterial suspension in modified MDG no carbon to the above

media. The plate was sealed with a Breathe-Easy membrane and incubated with shaking at

37°C for 24 hours before recording growth. For each media condition, final OD_{600} values were

compared pairwise for each strain with or without acarbose in each media, and significance was

determined in GraphPad Prism as above.

Sequence availability

Sequence files for TET1, TET3, TET13, ACA1, and ACA7 can be found as supplemental files 1

through 5.

RESULTS

METa assembly can reproducibly prepare Gb sized libraries with 50 ng of input

metagenomic DNA

In our manuscript describing METa assembly, we successfully prepared functional

metagenomic libraries using metagenomic DNA input quantities of 200 ng to 300 ng,

approximately 25- to 67-fold less than is commonly used to prepare small insert libraries[8,9]. An

advantage of our tagmentation-based fragmentation approach over physical fragmentation (*e.g.,*

via sonication) is its ready scalability to lower volumes and therefore lower input DNA masses.

However, we suspected that DNA losses during purification and fragment size-selection by gel

excision would make sub-200 ng libraries impractical. Our published METa assembly protocol

tagments metagenomic DNA at a concentration of 10 ng/µl, meaning a 50 ng DNA tagmentation

reaction would have a 5 µl volume where accurate pipetting starts to become difficult. To test the

practicality of extracting useful DNA fragments from this size reaction, we diluted 50 ng of a

Table 1. Summary statistics of functional metagenomic libraries

Library	Input DNA	DNA mass (ng)	Ave insert size (kb)	Proportion	Library size (Gb)	Efficiency (Gb/µg)	Selections[1]
50 ng goose 1	Goose gut microbiome	50	2.48	1.00	0.89	17.8	n/a
50 ng goose 2[2]	Goose gut microbiome	50	2.62	1.00	0.10	2.09	n/a
50 ng goose 3	Goose gut microbiome	50	3.13	1.00	2.46	49.3	n/a
Aquatic	Shedd aquarium	30.5	2.77	0.88	0.72	23.6	TET, CIP, FFC
Fecal	Human fecal reference	100.8	2.70	0.90	1.48	1.48	ACA

[1] no selection (n/a), tetracycline (TET), ciprofloxacin (CIP), florfenicol (FFC), acarbose (ACA)

[2] Replicate sparked during electroporation

DNA ladder in tagmentation buffer in triplicate to a concentration of 10 ng/µl to make mock 5 µl

tagmentation reactions. The triplicate mock fragmented samples were loaded on to an agarose

gel for excision of bands ranging from approximately 2 kb to 10 kb (**Supplemental figure 1A**).

Surprisingly to us, this resulted in 'fragmented DNA' quantities averaging 12.7 ng (± 2.7

standard deviation). Given that the total mass of the DNA bands 2 kb and higher was 16 ng, this

represented a 79% yield and suggested that size-selection of fragmented DNA from a 50 ng

tagmentation reaction could be feasible (**Supplemental figure 1B**).

Next, we tested this explicitly by preparing a 50 ng metagenomic DNA functional

metagenomic library in triplicate. Using previously extracted goose fecal metagenomic DNA as

input[8], we performed three tagmentation reactions and used gel excision to size-select for DNA

fragments between 2 kb and 6.5 kb (**Supplemental figure 2A**). These fragments were used to

prepare three functional metagenomic libraries by META assembly. During the electroporation

step, replicates 1 and 3 reached full voltage (1.8 kV) and had time constants of 4.6 and 4.8

seconds respectively. Replicate 2 reported a voltage of 1.65 kV and a 1.0 second time constant,

consistent with the sample 'sparking'. Across the three libraries we found an average insert size

of 2.75 kb with no empty vectors (**Supplemental figure 2B**), an average library size of 1.15 Gb,

and a library preparation efficiency of 23 Gb/µg of input DNA. Notably, these averages include

replicate 2 with a dramatically smaller library size (100 Mb compared to 890 Mb and 2.5 Gb)

(**Table 1**). If replicate 2 is excluded, the average library size for the 50 ng functional

metagenomic libraries increases to 1.68 Gb with an efficiency of 33.6 Gb/µg.

METa assembly preparation of a functional metagenomic library from a limited, low biomass microbiome

Two areas where input mass is limiting for classic functional metagenomic library preparation are 1) when the microbiome of interest is lower in biomass (*e.g.,* aquatic sources as opposed to soil or fecal samples) and 2) when the microbiome of interest is not readily available for re-sampling. We decided to test low-input mass METa assembly on a microbiome that fulfills both limitations. In a previous project, metagenomic DNA was extracted from the Underwater Beauty exhibit at the Shedd Aquarium in Chicago, Illinois following filtration capture of planktonic cells[13]. We used a sample of previously extracted metagenomic DNA to prepare a functional metagenomic library. Tagmentation was performed on 30.5 ng of metagenomic DNA and fragments from 2 kb to 10 kb were size selected by gel excision. Following transformation, random colonies from the functional metagenomic library were used as template for PCR to amplify captured metagenomic inserts to determine library statistics. We found that this aquatic functional metagenomic library captured 720 Mb of metagenomic DNA with an average insert size of 2.77 kb and with 12% of plasmids not containing an insert. The library preparation efficiency was calculated to be 23.6 Gb/μg (**Table 1**).

To evaluate the usefulness of this functional metagenomic library, we used it to investigate antimicrobial resistance against three antibiotics that are commonly used in aquaculture: tetracycline (**Figure 1C**), florfenicol, and ciprofloxacin[35,36]. The selection on ciprofloxacin did not yield any resistant colonies but colonies were recoverable from both the tetracycline and florfenicol plates. As an initial investigation, fifteen colonies from the tetracycline selection were collected and submitted for whole plasmid sequencing. We dereplicated repeated inserts, which resulted from plating a ten-fold excess of total cells over unique cells, resulting in three *E. coli* strains, TET1, TET3, and TET13. The metagenomic DNA captured in these strains had top nucleotide BLAST hits to *Lentibacter algarum* strain SH36 (95% coverage,

Figure 2. Tetracycline resistance in the Shedd aquarium aquatic microbiome. A) Gene schematics of the TET1, TET3, and TET13 metagenomic DNA fragments with predicted efflux pumps highlighted (TET1 red, TET3 blue, TET13 gray). Other predicted open reading frame are shown as empty arrows and plasmid backbone markers are highlighted as follows: Promoter (green), mosaic end sequence (orange), terminator (red). B) Maximum likelihood phylogenetic tree of CARD MFS efflux pumps with predicted TET1, TET3, and TET13 amino acid sequences highlighted. Microbroth dilution assays performed un quadruplicate for C) tetracycline or D) chloramphenicol resistance with calculated 50% inhibition concentration (IC$_{50}$) values graphed as insets. For statistical comparisons: * $p<0.05$, ** $p<0.005$, and *** $p<0.0005$.

82.09% identity), *Legionella pneumophila* strain NY23 (30% coverage, 66.87% identity), and

Saccharophagus degradans strain FZY0027 (32% coverage, 70.26% identity), respectively.

We annotated the metagenomic DNA fragments captured in these three strains

(**Supplemental figures 3A, 3B, and 3C**) and found them to each contain predicted MFS family

efflux pumps with low full-length amino acid identity (<30%) to their closest match in The

Comprehensive Antibiotic Resistance Database (CARD)[19,20] (**Figure 2A**). When placed on a

phylogenetic tree alongside known antibiotic resistance efflux pumps, TET1 and TET13 cluster

with each other and a variety of efflux pumps with associated chloramphenicol resistance. TET3

instead clusters with known macrolide and tetracycline efflux pumps (**Figure 2B**). The

resistance of all three strains to tetracycline, chloramphenicol, and azithromycin were measured

by microbroth dilution. All three strains showed significantly increased resistance to tetracycline,

with TET1 and TET3 showing growth at 8 µg/ml tetracycline or higher (**Figure 2C**). The *E. coli*

strain expressing the TET3 metagenomic fragment also showed a statistically significant

reduction in chloramphenicol susceptibility, while the strain corresponding to TET1 showed a

significantly increased susceptibility to the antibiotic (**Figure 2D**). None of the three strains

showed significant differences in azithromycin resistance compared to the control strain

(**Supplemental figure 3D**).

METa assembly preparation of a functional metagenomic library from a fecal swab

Another area where the ability to prepare a functional metagenomic library using minimal

DNA resources could be beneficial is within clinical settings. For example, fecal swabs provide

access to the medically important gut microbiome, but extraction of metagenomic DNA from this

sample type falls far below what classic functional metagenomic library preparation methods call

for[14,37]. To test the possibility of medical swabs being used as input for METa assembly, we

turned to the ZymoBIOMICS fecal reference with TruMatrix technology fecal standard. Each

aliquot of this standard is expected to contain 10 mg (wet weight) of human fecal material in 1

ml of DNA/RNA shield preservation buffer. We sampled this microbiome by dipping a sterile

swab into a thawed tube of the material followed by DNA extraction from the swab, resulting in

collection of 100.8 ng of DNA. We prepared and evaluated a functional metagenomic library by

METa assembly as above and found it to contain an average insert size of 2.7 kb, 10% empty

vectors, for a total expected library size of 1.48 Gb and a library preparation efficiency of 14.7

Gb/µg (**Table 1**).

The ZymoBIOMICS fecal standard has already been sequenced and annotated for

known antibiotic resistance genes. As a result, we chose to investigate a different function in the

human gut microbiome, pharmaceutical metabolism. Specifically, we investigated modification

of acarbose (**Figure 1D**), a glucosidase inhibitor used in the treatment of diabetes[38,39]. Previous

work has established that acarbose can be metabolized by bacteria in the human gut

microbiome through modification by phosphorylation[30] and degradation[25–29,29,40] and that

acarbose can prevent growth of *E. coli* when it is cultured with maltose as a sole carbon

source[41]. This suggested to us that a functional metagenomic approach could be used to

investigate acarbose metabolism. We validated this by plating *E. coli* in a lawn on minimal

media agar plates containing maltose as a sole carbon and energy source with acarbose

concentrations ranging from 0.5 µg/ml to 256 µg/ml. We observed confluent growth until an

acarbose concentration of 256 µg/ml at which point almost no colonies were observed

(**Supplemental figure 4A**). In contrast, we confirmed growth in the presence of 256 µg/ml

acarbose when *E. coli* was plated on minimal media containing glucose as a carbon source

(**Supplemental figure 4B**).

Based on these observations, we performed a functional metagenomic selection for

growth on maltose in the presence of acarbose. An aliquot of the Zymo fecal functional

metagenomic library and a similar titer of *E. coli* containing an empty vector were plated on to

minimal media plates containing maltose as the sole carbon source and acarbose at a

concentration of 256 µg/ml. Inclement weather resulted in the selections incubating at room

temperature for multiple days after which both plates showed a lawn of very small colonies.

However, the agar plate that was seeded with the functional metagenomic library contained

several larger colonies, the eight largest of which were collected for further study

(**Supplemental figure 4C**).

Following whole plasmid sequencing of these eight colonies, we found that four

contained identical inserts to each other and refer to the representative strain as ACA1. Protein

BLAST analysis of the single predicted open reading frame of ACA1 annotated it as an S-layer

homology domain-containing protein (**Figure 3A**). We used the CLEAN enzyme function

annotation tool[24] to suggest a potential function. Interestingly, E.C. families 3.2.1.8 and 3.2.1.4

were suggested as potential hits and both belong to the glycosidase family of enzymes. Analysis

of the ACA1 metagenomic DNA fragment by nucleotide BLAST failed to find a high-quality hit,

with the top predicted match (12% coverage, 77.18% identity) being to a bacteriophage and the

top predicted bacterial match (3% coverage, 90.74% identity) being to an *Intestinibacillus*

species. Plasmid sequencing of one of the other colonies (termed ACA7) found a metagenomic

insert predicted to encode three open reading frames (**Supplemental figure 5A**), one of which

was predicted to encode an α-amylase family glycosyl hydrolase gene (**Figure 3A**). Unlike the

enigmatic ACA1, the nucleotide sequence associated with ACA7 had a high-quality hit on

BLAST, *Alistipes onderdonkii* susbspecies *vulgaris* (100% coverage, 97.18% identity).

To orient the predicted effectors of ACA1 and ACA7 we placed them on a maximum

likelihood phylogenetic tree built on the sequences of known acarbose degrading enzymes and

known acarbose phosphotransferase (or kinase) enzymes (**Figure 3B**). As expected, these two

functions, glycan degradation and phosphorylation, segregate on the phylogenetic tree. The

predicted α-amylase ACA7 clusters with the known acarbose degrading enzymes while ACA1

does not appear to cluster with either enzyme group; its exact function is currently under

investigation.

To verify that ACA1 and ACA7 counteract the inhibitory activity of acarbose on *E. coli*

growth on maltose, we compared the growth of the two corresponding strains to an empty

plasmid control *E. coli* strain under a variety of conditions. When cultured in Mueller-Hinton

broth, the addition of acarbose resulted in a decreased final optical density of the control *E. coli*,

potentially reflecting the mixed carbon sources available in that media. In contrast, the ACA1

and ACA7 strains showed no effect on addition of acarbose **(Supplemental figure 5B)**. In

Figure 3. Predicted acarbose degradation enzymes from the human gut microbiome. A) Gene schematics of the ACA1 and ACA7 metagenomic DNA fragments with the predicted effector genes highlighted (ACA1 red and ACA7 blue). Predicted genes captured on the same DNA fragment are shown as empty arrows. Plasmid backbone elements are: Promoter (green), mosaic end sequence (orange), terminator (red). B) Maximum likelihood phylogenetic tree of the predicted proteins for ACA1 (red) and ACA7 (blue) in the context of known acarbose degrading or modifying enzymes (*i.e.,* kinases). C) and D) Culture densities with or without addition of 256 µg/ml acarbose after 24 hours of growth in minimal media with C) 0.4% glucose or D) 0.4% maltose as carbon sources. Pairwise statistical comparisons for the effect of acarbose addition were made for each strain and culture condition: * $p<0.05$, ** $p<0.005$, and *** $p<0.0005$.

minimal media with glucose as a carbon source, all three strains showed essentially no change

in growth in the presence of acarbose, although we observed a minor but statistically significant

decrease in the OD_{600} reached by the control strain (**Figure 3C**). Finally, when grown in minimal

media with maltose as a carbon source, the growth of the empty plasmid *E. coli* strain is nearly

entirely inhibited in the presence of acarbose while ACA1 and ACA7 show no difference in

growth on the addition of acarbose to the media (**Figure 3D**).

DISCUSSION

Here we report that our METa assembly method for preparing functional metagenomic

libraries can be extended to preparing libraries from low biomass and otherwise precious

microbiome samples where metagenomic DNA would otherwise be limiting. As an illustration of

this potential, we prepared modestly-sized functional metagenomic libraries from metagenomic

DNA extracted from an aquarium filtrate and from a simulated fecal swab. We further

demonstrated the utility of the resulting functional metagenomic libraries by applying them to

identify likely tetracycline efflux pumps and novel acarbose degradation genes.

A key result of our experiments is that using METa assembly to prepare functional

metagenomic libraries allows us move the minimal metagenomic DNA input to approximately

two orders of magnitude lower than previous standard protocols[9,14]. We show that, across our

experiments, 50 ng or lower input metagenomic DNA masses can be used to prepare functional

metagenomic libraries encoding near or low Gb libraries (**Table 1**). We propose, and

demonstrate, that this opens new sample types to analysis by functional metagenomic

selections or screens. These include samples that are intrinsically of low biomass, either due to

difficulty in collection or their rarity (*e.g.,* aquatic metagenomes, but also historically banked

microbiomes or samples from remote regions). These also include microbiomes sampled using

medical swabs. While our example used Zymo fecal reference material, the mass of the

metagenomic DNA extracted from the swab and used to prepare a functional metagenomic

library was noticeably less than what has been reported in extractions from other fecal swabs

(*i.e.,* ~100 ng vs ~370 ng). Functional metagenomic selections have been proposed to be useful

for medical applications[42], and our results support that.

Our previous smallest METa assembly input, 200 ng, resulted in a functional

metagenomic library encoding approximately 13.5 Gb with a preparation efficiency of ~68

Gb/µg[8]. Here, our 50 ng, 30.5 ng, and 100.8 ng libraries averaged an efficiency of between 20.5

Gb/µg and 24.1 Gb/µg, depending on the inclusion or exclusion of the library that sparked

during transformation, suggesting that lower input DNA masses may come with decreased

library preparation efficiency (**Table 1**). It is possible that this is due to DNA losses during the

multiple purification steps involved in library preparation. Nevertheless, these levels of efficiency

still perform better than our previously estimated average of 1 Gb/µg using blunt ligation-based

techniques[8].

While the functional metagenomic libraries prepared here are smaller than our previous

high-mass input libraries, their usefulness is demonstrated by their capture of genes for

tetracycline efflux pumps and acarbose degradation (**Figures 2 and 3**). The three studied

predicted tetracycline efflux pumps show homology to previously identified MFS efflux pumps,

with the predicted amino acid sequences having near complete coverage and between 65% and

almost 100% identity to hits in the NCBI database. In contrast, they have only low identity to

CARD hits, with full length alignments showing only 28% to 29% identity. It is common for

proteins identified through functional metagenomic selections for antibiotic resistance to have

high identity to known hits, but for those proteins to not previously have been identified as

having antibiotic resistance potential[43]. Tetracycline efflux pumps are found within many

environments[44–46] but the substrate specificity of novel transporters makes the identification of

potential resistance genes from environmental microbiomes difficult. This is illustrated by the

contrast between the apparent clustering of the TET1 and TET13 MFS pumps with amphenicol

efflux and TET3 with macrolide efflux (**Figure 2B**) and their respective lack of resistance to

chloramphenicol (**Figure 2D**) or azithromycin (**Supplemental figure 3D**). Our work highlights

the important role that functional metagenomic selections can play in filling this knowledge gap.

Similarly, one of the predicted proteins from the acarbose/maltose selection, ACA7, has

very high coverage and identity to known proteins in the NCBI database that lack an acarbose-

modification annotation. On the other hand, the predicted protein from ACA1 only showed

homology proteins with limited annotation as an S-layer homology domain-containing protein.

Interestingly, while ACA7 appears to clearly cluster with known acarbose degrading proteins (a

class which expanded again recently following a functional metagenomic screen for glycan

breakdown[40]), ACA1 does not (**Figure 3B**). Instead, ACA1 may represent a new acarbose

modifying enzyme. When ACA1 is compared against the CARD database by protein BLAST, the

top hit is a chloramphenicol acetyltransferase with an E-value of 0.09. We are currently

investigating the possibility that ACA1 may have transferase activity.

While we see the opening of low biomass functional metagenomic libraries as a

significant advancement, limitations to the method remain. First, this method relies on *E. coli*

cells as the host organism for the functional metagenomic library and therefore is subject to

potential limits on what metagenomic genes can be successfully expressed by this host[47]. In

theory this can be surmounted using alternative hosts[48–50], and the added efficiency of METa

assembly may be more important when using strains with lower transformation efficiency than

commercial *E. coli* strains. Second, there are potentially diminishing returns to preparing smaller

functional metagenomic libraries even if their preparation efficiency increases. However, the

libraries prepared here were shown to still be useful by their capture of potentially novel

antibiotic efflux pumps and acarbose degradation enzymes even without covering more than a

fraction of the genetic diversity found in many microbiomes. Tagmentation reactions are

regularly performed on metagenomic DNA samples with 5 ng or less mass[12,51,52] suggesting that

the limit for functional metagenomic libraries could be pushed even further.

In conclusion, while high-throughput DNA sequencing has helped circumvent the great

plate count anomaly and has revolutionized microbiome research, methods for accurately

annotating novel genes have not managed to keep pace. Machine-learning tools will likely be

able to help with this issue but will also require large amounts of high-quality sequence-function

correlation data. Functional metagenomics is well placed to supply this data and, in the

meantime, can make significant strides in narrowing the sequence annotation gap. We propose

that the fulfillment of this potential by functional metagenomics is significantly aided by the ability

to make the necessary libraries more efficiently and using less input DNA, a precondition that

we demonstrate is reached here.

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