

# A discriminative primer design workflow enables selective metabarcoding, demonstrated using long-read sequencing of endophytic fungi

Ting He<sup>1</sup>, Koert Jansonius<sup>1</sup>, Xiao Li<sup>1</sup>, Alison M. Reilly<sup>1</sup>, Bahar Sevgin<sup>2</sup>, Rita Setroikromo<sup>1</sup>, Thomas Hackl<sup>1,2</sup>, Kristina Haslinger<sup>1,\*</sup>

<sup>1</sup>Department of Chemical and Pharmaceutical Biology, Groningen Research Institute of Pharmacy, University of Groningen, Antonius Deusinglaan 1, 9713 AV, Groningen, The Netherlands

<sup>2</sup>Groningen Institute for Evolutionary Life Sciences, University of Groningen, Nijenborgh 7, 9747 AG, Groningen, The Netherlands

\*Corresponding author. Department of Chemical and Pharmaceutical Biology, Groningen Research Institute of Pharmacy, University of Groningen, Antonius Deusinglaan 1, 9713 AV, Groningen, The Netherlands. E-mail: [k.haslinger@rug.nl](mailto:k.haslinger@rug.nl)

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## Abstract

Metabarcoding is a powerful tool to simultaneously identify multiple taxa within a habitat. However, its application to host-associated microbiomes is challenged by substantial co-amplification of host DNA. Here we developed a host-exclusive primer design workflow, to selectively generate amplicons from target taxa while excluding the host. This workflow is centered around a new computational tool, *mbc-prime*, that can generate a list of discriminative candidate primers and score them. We showcase the use of this tool in the design of primers for long-read metabarcoding of endophytic fungi in *Vinca minor*. *Mbc-prime* streamlines the design of fungus-specific primers, enabling efficient and plant-free amplification of fungal rDNA from mixed DNA samples. Our workflow can be used to study the composition of complex host-associated microbiomes. It should be universally applicable for the design of discriminative primers in a user-friendly and practical manner and thus be of use for various researchers in microbiome research.

**Keywords** Amplicon sequencing; host-associated microbiome; Oxford Nanopore Technology; discriminative primer design; metabarcoding, marker genes

## Introduction

High-throughput sequencing of taxonomic marker genes (metabarcoding) has become a widely adopted approach for studying biodiversity and microbial ecology across diverse ecosystems (Abdelfattah et al. 2018). Among the critical steps in any metabarcoding workflow, the choice and design of suitable primers is paramount. An optimal primer set must balance broad taxonomic coverage of target organisms while minimizing amplification of non-target taxa (Tedersoo et al. 2022).

This issue of non-target amplification extends beyond microbial profiling and is encountered across a wide range of ecological and evolutionary studies. For instance, ancient DNA analyses are often compromised by high levels of microbial or environmental DNA that obscure the target sequences (Slon et al. 2017). Marine eDNA studies employing broad-range primers are susceptible to amplification biases, whereby highly abundant eukaryotic taxa, such as dominant zooplankton orders, are preferentially amplified, potentially limiting the detection sensitivity for less abundant taxa within the community (Zhan et al. 2014, Needham and Fuhrman 2016).

This complication is particularly pronounced when studying host-associated target taxa, for instance in dietary metabarcoding, where the DNA of the predator or herbivore can dominate over that of the consumed prey or plant material, leading to biased interpretations (De Barba et al. 2014); or in parasitological diagnostics where scientists frequently struggle with the low abundance of target DNA of helminths or protozoan parasites amidst the overwhelming host background (Avramenko et al. 2015). In bacterial profiling of the human microbiome, a common challenge arises from the co-amplification of host organellar DNA that overwhelms the microbial signals, especially in low-biomass niches such as skin, blood, or mucosal surfaces (Marotz et al. 2018, Karstens et al. 2019). In plant-associated microbiomes, chloroplast and mitochondrial sequences often overshadow bacterial 16S rRNA gene amplification, obscuring true microbial diversity (Lundberg et al. 2013, Beckers et al. 2016). Fungal profiling, using eukaryotic marker genes such as ITS, 18S, or 28S rRNA, introduces additional complexity. These markers often carry regions that are conserved with host ribosomal DNA, making it difficult to avoid co-amplification, particularly in plant tissues such as roots or leaves (Hadziavdic et al. 2014, Gdanetz et al. 2017).

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This can mask the presence of less abundant fungal taxa and complicate downstream ecological interpretation. These examples collectively highlight the widespread impact of non-target DNA interference in amplicon sequencing, emphasizing the need for robust primer design strategies that improve target specificity and minimize the amplification of host DNA to ensure accurate and efficient microbial community profiling.

To overcome these issues, a wide array of bioinformatics tools has been developed to assist in primer design. General-purpose tools like Primer3 (Rozen and Skaletsky 2000), Primer-BLAST (Ye et al. 2012), and PrimerQuest™ offer design capabilities based on melting temperature, GC content, and structural properties. However, they are limited in their ability to assess primer suitability for complex, taxonomically diverse samples. Taxon-aware tools such as PrimerProspector (Walters et al. 2011) and DegePrime (Hugerth et al. 2014) allow for broader coverage assessments and degeneracy control, making them more suitable for environmental applications. Still, these tools generally lack features required for discriminative primer design, such as the ability to target specific taxonomic groups while excluding the host. Other tools, such as BarCrawl (Frank 2009), support barcode design but not primer development; Primer Validator provides taxonomic validation without de novo primer design or advanced scoring systems; and tools like RDP's Probe Match (Cole et al. 2005) or Greengenes' probe search (DeSantis et al. 2006) are limited to database-matching probes without flexible input or customization. Moreover, small-scale tools like Primrose (Ashelford et al. 2002) are not scalable to modern large-scale datasets and lack features like 3'-end scoring, which is important for predicting primer success.

Despite this extensive toolkit, most existing programs are not designed to avoid amplification of host DNA, a key requirement for studying host-associated microbiomes. Therefore, there remains a strong need for specialized computational solutions that can design host-exclusive primers, ensuring that sequencing reads derive primarily from microbial taxa rather than host contaminants (Yap et al. 2020). Such primers are essential for increasing classification sensitivity, reducing bias, and improving the cost-effectiveness of microbial community profiling in host-rich samples.

Here, we present an effective metabarcoding workflow tailored for host-associated microbiomes, beginning with the design of highly discriminative primers using our new computational tool, *mbc-prime*. This tool automatically identifies and ranks candidate primers based on microbial specificity and host exclusivity. We demonstrate the utility of this approach by profiling the community of endophytic fungi associated with the evergreen plant *Vinca minor* (lesser periwinkle), using long-read nanopore sequencing. Our method provides a robust solution for the accurate detection of microbial and other organismal diversity in challenging, host-associated environments.

## Materials and methods

### Sample collection and DNA extraction from the leaves of *V. minor*

For the sampling, 10–15 individual plants of *V. minor* were collected from Italy (45°26'3" N, 8°49'53" W), immediately placed in an ice box, and transported to the lab. After surface sterilization

(He et al. 2023), the plant material was quickly frozen with liquid nitrogen, ground into fine powder, and stored at  $-70^{\circ}\text{C}$  until DNA extraction.

Holobiont genomic DNA was extracted using NucleoSpin® Microbial DNA kit (BIOKÉ, Leiden, The Netherlands) in accordance with the protocols provided by the manufacturer with adjustment. Briefly, 0.2 g ground plant material was suspended in 100  $\mu\text{L}$  elution buffer BE, 40  $\mu\text{L}$  buffer MG and 10  $\mu\text{L}$  liquid proteinase K. The plant cells were disrupted by vortexing at 1000 r/m for 3 mins. After adding 600  $\mu\text{L}$  buffer MG, samples were vortexed for another 1 min at 1000 r/m, followed by 2 min centrifugation at  $13\,000 \times g$ . The supernatant was transferred onto a NucleoSpin® Microbial column, washed twice with buffer BW and B5 buffer, and eluted with 30  $\mu\text{L}$  elution buffer BE. The DNA was quantified with NanoDrop N-100 (ThermoFisher, Waltham, MA, USA) and stored in  $-20^{\circ}\text{C}$  until further use.

### Mock community construction

To generate a synthetic mock community for initial testing of the new primers, we used six species isolated from healthy-looking, surface-sterilized leaves of *V. minor*, collected in Groningen (The Netherlands) in November 2021. Fungal isolation, cultivation, and genomic DNA isolation were executed as described in our previous work (He et al. 2023). The SR1R-Fw/ LR12-R primer pair (D'Andrea et al. 2021) was used to amplify the entire fungal rDNA locus. Reactions were run in volumes of 25  $\mu\text{L}$ , containing 1  $\mu\text{L}$  each of the forward (10  $\mu\text{M}$ ) and reverse primer (10  $\mu\text{M}$ ), 12.5  $\mu\text{L}$  LongAmp Hot Start Taq 2x Master Mix (New England Biolab, MA, USA), 1  $\mu\text{L}$  of DNA template (40 fmol), and 9.5  $\mu\text{L}$  of nuclease-free water. PCR amplification was carried out with the following protocol: initial denaturation of  $94^{\circ}\text{C}$  for 30 s, followed by 30 cycles of  $94^{\circ}\text{C}$  for 10 s,  $65^{\circ}\text{C}$  for 30 s, and  $65^{\circ}\text{C}$  for 5.5 min, followed by a final  $65^{\circ}\text{C}$  extension for 10 min. The PCR products were purified using AMPure XP magnetic beads (Beckman Coulter, California, USA) and the concentrations were determined using Qubit 3.0 (Invitrogen, Waltham, MA, USA), and the Qubit dsDNA HS Assay Kit. These six fungal amplicons were mixed in equimolar ratios to construct the mock community DNA sample with a concentration of 55 ng/  $\mu\text{L}$ .

### Primer design and implementation on endophytic fungi in *V. minor*

To analyze endophytic fungi in *V. minor*, all related sequences were downloaded from the SILVA database (Supplementary File 2, Datasets S1 and S2). For the SSU region, 46 sequences from the Gentianales order were extracted from the downloaded database file (SILVA\_138.1\_SSURef\_NR99\_tax\_silva.fasta.gz.). In the same way, a dataset of 118 fungal sequences was collected, which represents all major fungal divisions (ascomycota, basidiomycota, chytridiomycota, cryptomycota, zoopagomycota, microsporidia, mucoromycota, neocallimastigomycota, and blastocladiomycota). These 164 sequences were aligned using the MAFFT multiple sequence alignment software (output format: fasta format, input order; strategy: L-INS-i; arguments: -ep) (Katoh and Standley 2013). For the LSU region, 62 fungal sequences (Supplementary File 2, Dataset S3) and 6 available Gentianales sequences (Supplementary File 2, Dataset S4) were extracted from the downloaded

database file (SILVA\_138.1\_LSURef\_NR99\_tax\_silva.fasta.gz). Alignment was performed among these 68 sequences in the same way as for the SSU.

Both alignments were analyzed with *mbc-prime* version 0.6.0 ( $-s = 0.5$ ) (see Results section for implementation). We obtained 147 candidate discriminative primers across the SSU (Supplementary File 2, Dataset S5) and 371 across the LSU (Supplementary File 2, Dataset S6). Four candidates for forward primers on the SSU and two reverse primers on the LSU were selected from the initial set based on the following criteria: 1) at least 80% coverage of fungi at zero or one mismatch (*in silico*); 2) at least one mismatch at the 3' end between primer compared to the plant sequences; 3) a GC content between 40 and 60%; 4) approx. 20 base pairs in length. In addition, we manually optimized the length of candidate primer sequences to lower the difference in melting temperatures and the likelihood of heterodimer formation in primer combinations as well as incorporated degenerate positions to maximize the coverage of the target group. Finally, all selected primers were evaluated *in silico* using SILVA TestProbe 3.0 (<https://www.arb-silva.de/search/testprobe/>) with default settings and “fungi” and “Gentianales” or “Chloroplastida” as test sets for the individual primers, as well as OligoAnalyzer™ (Version 3.1, Integrated DNA Technologies, <https://eu.idtdna.com/pages/tools/oligoanalyzer>) for primer pairs before being tested on the DNA samples of a fungal mock community and plant material.

## Amplicon generation for Oxford Nanopore sequencing

The strategy employed in this study involves two steps of PCR. The first PCR was performed with the gene-specific rDNA primers including overhangs that serve as handles for PCR barcoding in the second step (5' TTTCTGTTGGTGCTGATATTGC-[forward primer sequence] 3'; 5' ACTTGCTGCTGCTCTATCTTC-[reverse primer sequence] 3'). All primer pairs used in this study are shown in Supplementary File 1, Supplementary Table S1 and sequences of all primers used in this study are listed in Supplementary File 2, Dataset S7. Each PCR reaction mix consisted of 0.4  $\mu$ M each of the forward and reverse primer, 1  $\times$  LongAmp Hot Start Taq 2 $\times$  Master Mix (New England Biolab, MA, USA), and 40 ng of template DNA, combined in a 25  $\mu$ L reaction. The PCR cycling conditions were as follows: initial denaturation of 94°C for 30 s, followed by 30 cycles of a 10 s hold at 94°C, a 30 s hold at the respective annealing temperature shown in Table 1, and an elongation period specific for the length of the amplicon as shown in Table 1 at 65°C, followed by a final 65°C extension for 10 min. The PCR amplicons were purified using AMPure XP magnetic beads (Beckman Coulter, California, USA) and quantified by Qubit. The second PCR was used to incorporate the Oxford Nanopore barcode sequences with the PCR Barcoding Expansion 1–12 (EXP-PBC001, ONT, Oxford, UK). Each amplicon PCR product was used as a template (100–200 fmol) in the second PCR reaction with 1  $\mu$ L PCR barcoding primer mix (one of BC1–BC12, at 10  $\mu$ M), and 25  $\mu$ L LongAmp Hot Start Taq 2X Master Mix (New England Biolab, MA, USA) in a total of 50  $\mu$ L. The PCR cycling conditions were as follows: initial denaturation of 95°C for 3 min, followed by 15 cycles of 95°C for 15 s, 62°C for 15 s, and an elongation period specific for the length of the amplicon as shown in Table 1 at 65°C, followed by a final 65°C extension for 10 min. The barcoded PCR products were further purified with 0.4x volume of AMPure XP Beads (Beckman Coulter, California, USA). All

barcoded PCR products were pooled in equal amounts to prepare 1  $\mu$ g of DNA in 50  $\mu$ L nuclease-free water for library preparation. The metadata of samples as input for the library preparation is listed in the Supplementary File 2, Dataset S8.

## Library preparation

The library preparation was performed with the ligation sequencing kit V14 (ONT, SQK-LSK114) following the manufacturer's instructions ([https://community.nanoporetech.com/docs/prep-are/library\\_prep\\_protocols/ligation-sequencing-gdna-pcr-barcoding-sqk-lsk114-with-exp/v/pbc\\_9182\\_v114\\_revh\\_07mar2023](https://community.nanoporetech.com/docs/prep-are/library_prep_protocols/ligation-sequencing-gdna-pcr-barcoding-sqk-lsk114-with-exp/v/pbc_9182_v114_revh_07mar2023)). First, the pooled barcoded amplicons underwent end repair with the NEBNext® Ultra II End Repair / dA-tailing Module (NEB, E7546). This reaction consisted of 50  $\mu$ L DNA (1  $\mu$ g), 7  $\mu$ L Ultra II End-prep Reaction Buffer and 3  $\mu$ L Ultra II End-prep Enzyme Mix, and was incubated at 20°C for 5 min and 65°C for 5 min. The end-repaired amplicons were further purified using AMPure XP magnetic beads and quantified by Qubit. For the adapter ligation step, 60  $\mu$ L of the end-repaired and purified amplicons, 25  $\mu$ L Ligation Buffer (LNB), 10  $\mu$ L NEBNext Quick T4 DNA Ligase, and 5  $\mu$ L Ligation Adapter (LA) were combined in a reaction tube and incubated for 10 min at room temperature. After ligation, the samples were mixed with 0.4  $\times$  volume of AMPure XP magnetic beads and then washed twice with 250  $\mu$ L Long Fragment Buffer provided in the ligation sequencing kit. The library was finally eluted with 15  $\mu$ L Elution Buffer (EB), incubated at 37°C for 10 min, and quantified by Qubit. 35–50 fmol of this final prepared library were loaded for sequencing.

## MinION sequencing data analysis

Sequencing was performed on a MinION device with a R10.4.1 flow cell running with MinKNOW software v24.02.10. Basecalling and demultiplexing were performed with Dorado v7.3.9 in super accurate mode. We trimmed reads using NanoFilt v2.8.0 and assessed the read quality using NanoPlot v1.42.0. Subsequently, the FASTQ files were uploaded to EPI2ME Labs v4.1.4, and the “wf-metagenomics” workflow was performed with the following settings: classifier: Kraken2; Kraken2 memory mapping: true; dataset: ncbi\_16s\_18s\_28s\_ITS or SILVA\_138\_1; N taxa barplot: 9; bracken length: 2000; min % identity: 90; min len: 2000; max len: 6000. The mock community samples were analyzed with a custom database built with Kraken2. Relative abundances were further visualized with GraphPad Prism 10.

## Results

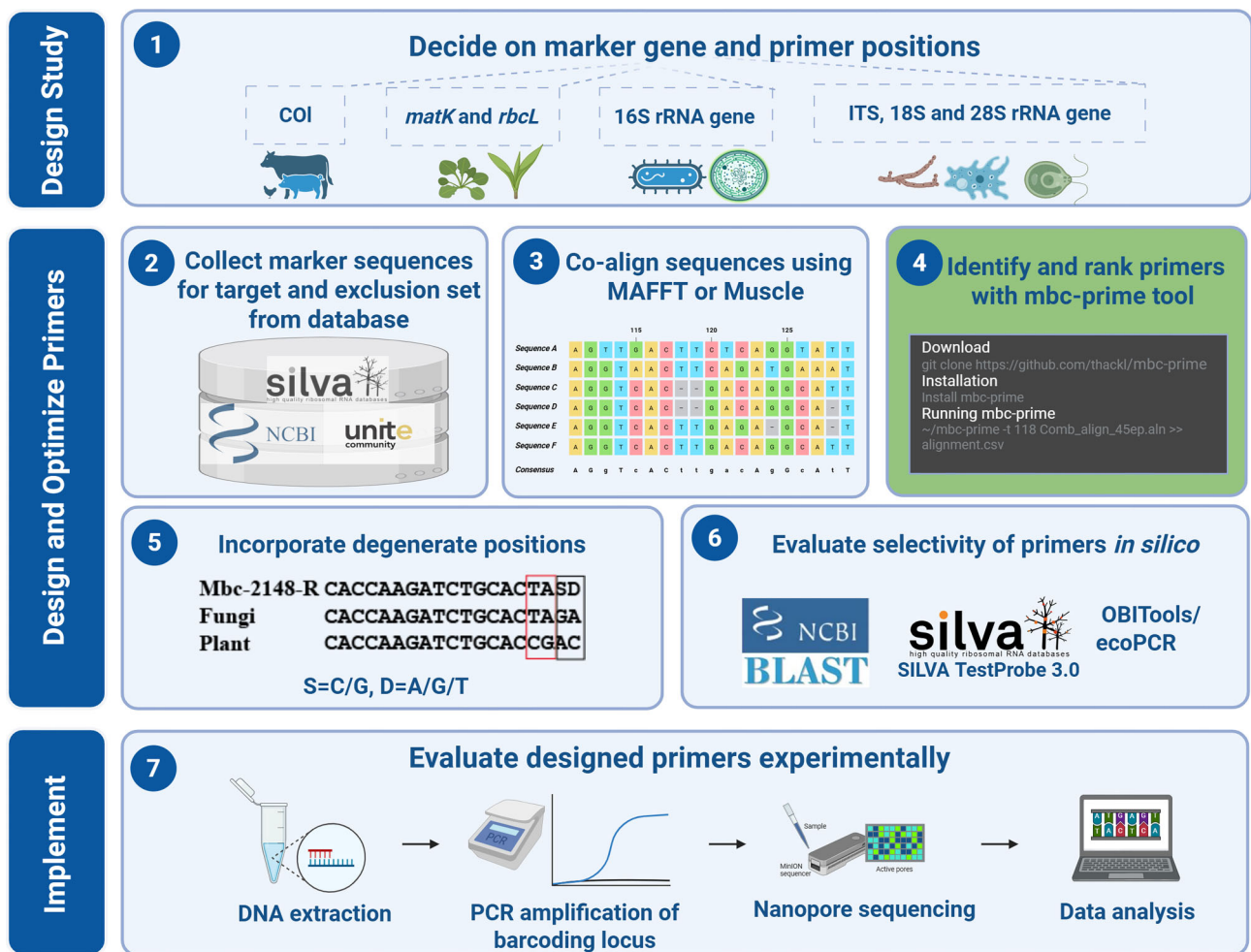
### Rationale and workflow for effective host-exclusive primer design

Metabarcoding relies on the targeted PCR amplification of suitable genomic markers that comprise conserved regions serving as the primer binding sites and variable regions with enough differences to distinguish individual species (Hebert et al. 2003, Hollingsworth et al. 2009, Klindworth et al. 2013). Herein, it is crucial that the primers amplify DNA equally well across a broad target taxonomic range. Therefore, we devised an effective workflow for designing such discriminative primers (Fig. 1) with the help of multiple-sequence alignments (MSAs) of representative sequences of the

Table 1 *mbc-prime* and SILVA TestProbe 3.0 evaluations of all primers used in this study.

| Primer             | Weighted score <sup>a</sup> | Coverage of Fungi <sup>b</sup> |       |       | Coverage of Gentianales <sup>b</sup> |        |        | Chloroplastida |       |       | Source                 |
|--------------------|-----------------------------|--------------------------------|-------|-------|--------------------------------------|--------|--------|----------------|-------|-------|------------------------|
|                    |                             | 0 mm                           | 1 mm  | 2 mm  | 0 mm                                 | 1 mm   | 2 mm   | 0 mm           | 1 mm  | 2 mm  |                        |
| Located on the SSU |                             |                                |       |       |                                      |        |        |                |       |       |                        |
| SSSU-Fngs-F        | –                           | 0.2%                           | 94.8% | 96.4% | 0%                                   | 80.0%  | 88.9%  | 0.5%           | 79.2% | 86.2% | Tedersoo et al. 2015   |
| mbmc-2479-F        | 0.62                        | 77.1%                          | 87.8% | 89.2% | 0%                                   | 0%     | 0%     | 0.3%           | 0.5%  | 2.2%  | This study             |
| mbmc-1840-F        | 0.62                        | 66.1%                          | 83.5% | 90.2% | 0%                                   | 0%     | 2.2%   | 0.2%           | 1.0%  | 8.6%  | This study             |
| mbmc-136-F         | 0.78                        | 80.1%                          | 89.7% | 92.3% | 0%                                   | 0%     | 87.5%  | 25.2%          | 39.0% | 81.3% | This study             |
| mbmc-1220-F        | 0.66                        | 52.1%                          | 82.3% | 88.7% | 0%                                   | 0%     | 11.1%  | 0.7%           | 24.1% | 35.8% | This study             |
| NS1short           | –                           | 60.05                          | 69.5% | 85.2% | 45.0%                                | 50.0%  | 65.0%  | 77.1%          | 87.0% | 90.8% | Wurzbacher et al. 2018 |
| Located on the LSU |                             |                                |       |       |                                      |        |        |                |       |       |                        |
| LR5-Fung           | 0.75                        | 88.0%                          | 91.8% | 93.3% | 0%                                   | 0%     | 100.0% | 0.0%           | 0.0%  | 0.1%  | Tedersoo et al. 2008   |
| mbmc-3159-R        | 0.80                        | 87.0%                          | 92.5% | 95.1% | 0%                                   | 0%     | 100.0% | 11.8%          | 18.1% | 84.7% | This study             |
| mbmc-2148-R        | 0.72                        | 86.4%                          | 92.0% | 94.4% | 0%                                   | 0%     | 0%     | 4.1%           | 8.8%  | 17.6% | This study             |
| LR5-R              | –                           | 90.9%                          | 90.3% | 94.2% | 100.0%                               | 100.0% | 100.0% | 67.9%          | 80.2% | 85.1% | Tedersoo et al. 2015   |
| RCA95m             | –                           | 88.7%                          | 91.3% | 94.2% | 0%                                   | 0%     | 0%     | 0.1%           | 0.1%  | 4.3%  | Wurzbacher et al. 2018 |
| Fun-rOP-R          | –                           | 80.9%                          | 88.4% | 90.8% | 0%                                   | 100.0% | 100.0% | 3.5%           | 73.7% | 84.4% | Wurzbacher et al. 2018 |

a: *mbc-prime* weighted score of the candidate primer before manual modification (optional); b: summative coverage achieved with the final primers determined by SILVA TestProbe 3.0; mm: number of maximum mismatches



**Figure 1** Workflow for effective host-exclusive primer design as described in this study. At the core of the workflow stands the computational tool mbc-prime, which effectively identifies candidates for discriminative primers based on multiple sequence alignments of the target and the exclusion sets (step 4). Figure created in BioRender [He, T. (2024) <https://BioRender.com/d86p488>].

target and exclusion groups. Since finding suitable primer bindings sites in such MSAs by hand is slow and tedious, we also devised the program *mbc-prime* (described in detail in the next section) to help automate this crucial step. The list of discriminative, candidate primers returned by *mbc-prime* is then further inspected for desired properties of the primer, for instance the GC content and melting temperature of the primer and the position of the non-target mismatches within the primer sequence. With the most discriminative primers in hand, additional optimization of the primers can be performed by introducing degenerate bases into the sequence to optimize the coverage among the target sequences. Lastly, the candidate primers should be evaluated for their specificity for target group and efficiency of excluding non-target group.

## Detection of discriminative primer binding sites with *mbc-prime*

*Mbc-prime* is a stand-alone Python program available at <https://github.com/thackl/mbc-prime>. It identifies discriminative primer binding sites based on a user-supplied MSA in FASTA format, comprising DNA sequences of the chosen genomic locus in the tar-

get and exclusion sets. In a first step, the MSA is imported and preprocessed. Terminal gaps, introduced by partial sequences frequently present in amplicon marker databases, are masked to distinguish them from genuine gaps. Columns with >20% (default value) gaps or ambiguous bases (non-ATGCU) are trimmed from the alignment. The trimmed MSA is then analyzed via a sliding window with a size corresponding to the intended primer length (default: 20 bp). For each window and sequence set (target and exclusion), the abundance of each sequence variant is counted, and the most abundant sequence variant of the target set is chosen as primer candidate. Next, for each remaining sequence variant from both sets, the number of mismatches compared to the primer candidate is computed, and the overall percentage of sequence variants with 0, 1, 2, 3, or 4-or-more mismatches are determined. The underlying edit distances are computed via Python bindings of the Edlib library (Šošić and Šikić 2017). This results in two mismatch profiles per primer candidate, one for the target set and one for the exclusion set. An example of promising profiles for a discriminative primer would be the following: 90% and 10% of the target sequences align with 0 and 1 mismatch, respectively, while 10%, 20% and 70% of the exclusion sequences align with 2, 3, and 4-or-more mismatches, respectively.

| Positional Information |     |             |       | Score                        |                               | Mismatch Profiles          |                      | Primer Sequences |                |                 |
|------------------------|-----|-------------|-------|------------------------------|-------------------------------|----------------------------|----------------------|------------------|----------------|-----------------|
| A                      | B   | C           | D     | E                            |                               | F                          |                      | G                | H              | I               |
| group                  | pos | trimmed_pos | score | exc: 0,1,2,3,4+ mismatches   |                               | inc: 0,1,2,3,4+ mismatches |                      | primer forward   | primer reverse | inc-exc-matched |
| # locus 1              |     |             |       |                              |                               |                            |                      |                  |                |                 |
| 1                      | 136 | 46          | 0.78  | [0.0, 0.0, 0.84, 0.88, 1.0]  | [0.71, 0.86, 0.87, 0.87, 1.0] | gccaugcaugcuaaguaua        | tatacttagacatgcatg   |                  |                |                 |
| 1                      | 137 | 47          | 0.76  | [0.0, 0.04, 0.84, 0.88, 1.0] | [0.73, 0.82, 0.87, 0.87, 1.0] | ccaugcaugcuaaguaua         | tatacttagacatgcatg   |                  |                |                 |
| 1                      | 138 | 48          | 0.61  | [0.0, 0.04, 0.84, 0.88, 1.0] | [0.5, 0.76, 0.85, 0.87, 1.0]  | caugcaugcuaaguaua          | tttatacttagacatgcatg |                  |                |                 |
|                        |     |             | 0.56  | [0.0, 0.04, 0.84, 0.88, 1.0] | [0.41, 0.76, 0.84, 0.85, 1.0] | augcaugcuaaguauaac         | gtttatacttagacatgcat |                  |                |                 |
| # locus 2              |     |             |       |                              |                               |                            |                      |                  |                |                 |
|                        |     |             | 0.5   | [0.0, 0.0, 0.84, 0.91, 1.0]  | [0.38, 0.61, 0.75, 0.86, 1.0] | uacuggaauaacgugguua        | ttaccacggttatccaagta |                  |                |                 |
|                        |     |             | 0.51  | [0.0, 0.0, 0.84, 0.91, 1.0]  | [0.41, 0.61, 0.83, 0.88, 1.0] | acuuggaauaacgugguua        | attaccacggttatccaagt |                  |                |                 |
|                        |     |             | 0.5   | [0.0, 0.02, 0.84, 0.93, 1.0] | [0.41, 0.61, 0.83, 0.88, 1.0] | cuuggaauaacgugguua         | aattaccacggttatccaag |                  |                |                 |
| 2                      | 262 | 138         | 0.55  | [0.0, 0.02, 0.84, 0.93, 1.0] | [0.42, 0.7, 0.84, 0.88, 1.0]  | uuggaauaacgugguuauc        | gaattaccacggttatccaa |                  |                |                 |
| 2                      | 263 | 139         | 0.7   | [0.0, 0.02, 0.87, 0.93, 1.0] | [0.59, 0.84, 0.86, 0.88, 1.0] | uggaauaacgugguuauc         | agaattaccacggttatcca |                  |                |                 |
| # locus 3              |     |             |       |                              |                               |                            |                      |                  |                |                 |
| 3                      | 280 | 155         | 0.6   | [0.0, 0.0, 0.91, 0.96, 1.0]  | [0.43, 0.77, 0.82, 0.86, 1.0] | uucuaagagcuaauaaucaugcu    | agcatgtattagctctagaa |                  |                |                 |
| # locus 4              |     |             |       |                              |                               |                            |                      |                  |                |                 |
| 4                      | 337 | 199         | 0.55  | [0.0, 0.0, 0.0, 0.38, 1.0]   | [0.47, 0.62, 0.73, 0.8, 1.0]  | uguauuuuuuagauaaaaaa       | ttttttatctaataataca  |                  |                |                 |
| 4                      | 338 | 200         | 0.54  | [0.0, 0.0, 0.0, 0.02, 1.0]   | [0.46, 0.62, 0.74, 0.83, 1.0] | guauuuuuuagauaaaaaac       | gtttttatctaataataac  |                  |                |                 |
| 4                      | 339 | 201         | 0.54  | [0.0, 0.0, 0.0, 0.0, 1.0]    | [0.46, 0.62, 0.74, 0.83, 1.0] | uuuuuuuuuagauaaaaaac       | ggttttatctaataataa   |                  |                |                 |
| 4                      | 340 | 202         | 0.54  | [0.0, 0.0, 0.0, 0.02, 1.0]   | [0.46, 0.62, 0.79, 0.86, 1.0] | uuuuuuuuuagauaaaaacca      | tggtttttatctaataata  |                  |                |                 |
| 4                      | 341 | 203         | 0.53  | [0.0, 0.0, 0.0, 0.02, 1.0]   | [0.45, 0.61, 0.76, 0.83, 1.0] | uuuuuuuuuagauaaaaacca      | ttggttttatctaataaa   |                  |                |                 |
| # locus 5              |     |             |       |                              |                               |                            |                      |                  |                |                 |
| 5                      | 550 | 362         | 0.5   | [0.0, 0.0, 0.04, 0.8, 1.0]   | [0.34, 0.66, 0.88, 0.95, 1.0] | aacggggaacggggaauuag       | ctaattccccgttaccggt  |                  |                |                 |
| 5                      | 551 | 363         | 0.56  | [0.0, 0.02, 0.78, 0.89, 1.0] | [0.41, 0.74, 0.91, 0.97, 1.0] | acggggaacggggaauuagg       | ccctaattccccgttaccgt |                  |                |                 |
| 5                      | 552 | 364         | 0.56  | [0.0, 0.02, 0.78, 0.89, 1.0] | [0.41, 0.74, 0.91, 0.97, 1.0] | cggggaacggggaauuagg        | ccctaattccccgttaccg  |                  |                |                 |
| 5                      | 553 | 365         | 0.56  | [0.0, 0.02, 0.78, 0.89, 1.0] | [0.41, 0.74, 0.88, 0.95, 1.0] | ggggaacggggaauuagg         | accctaattccccgttacc  |                  |                |                 |
| 5                      | 554 | 366         | 0.57  | [0.0, 0.02, 0.83, 0.93, 1.0] | [0.41, 0.74, 0.88, 0.91, 1.0] | gguuacggggaauuagggu        | aaccctaattccccgttacc |                  |                |                 |
| 5                      | 555 | 367         | 0.56  | [0.0, 0.02, 0.83, 0.91, 1.0] | [0.41, 0.73, 0.84, 0.89, 1.0] | guuacggggaauuaggguuc       | gaaccctaattccccgttac |                  |                |                 |
| 5                      | 556 | 368         | 0.54  | [0.0, 0.02, 0.83, 0.91, 1.0] | [0.38, 0.72, 0.79, 0.86, 1.0] | uuacggggaauuaggguucg       | cgaaccctaattccccgtta |                  |                |                 |
| 5                      | 557 | 369         | 0.54  | [0.0, 0.02, 0.83, 0.91, 1.0] | [0.38, 0.72, 0.79, 0.86, 1.0] | aacggggaauuaggguucga       | tcgaaccctaattccccgtt |                  |                |                 |

Figure 2 Screen shot of an example output file of mbc-prime.

To facilitate easy comparison and ranking of possible primer candidates based on their mismatch profiles, we integrate the two profiles obtained from target and exclusions sets into a single discriminative score, as defined below. The score ranges between 0 and 1, with 1 capturing 100% of target sequences perfectly while excluding 100% of non-target sequences at the given mismatch threshold (default mismatch threshold: mm = 2).

$$\text{score} = \frac{1}{mm} \sum_{k=1}^{mm} \left( \sum_{i=1}^k \text{target}[i] - \sum_{i=1}^k \text{exclude}[i] \right).$$

Finally, candidate primers are filtered for a maximum number of gaps, ambiguous bases and a minimum score (default 0.5), and overlapping primers are grouped together. The filtered list of above-threshold primer candidates is reported together with the positional information, score, mismatch profiles, and sequence in a tab-separated table (Fig. 2). A step-by-step tutorial for using *mbc-prime* is listed in [Supplementary Note S1](#).

## Case study

### Design of metabarcoding primers for endophytic fungi in *V. minor*

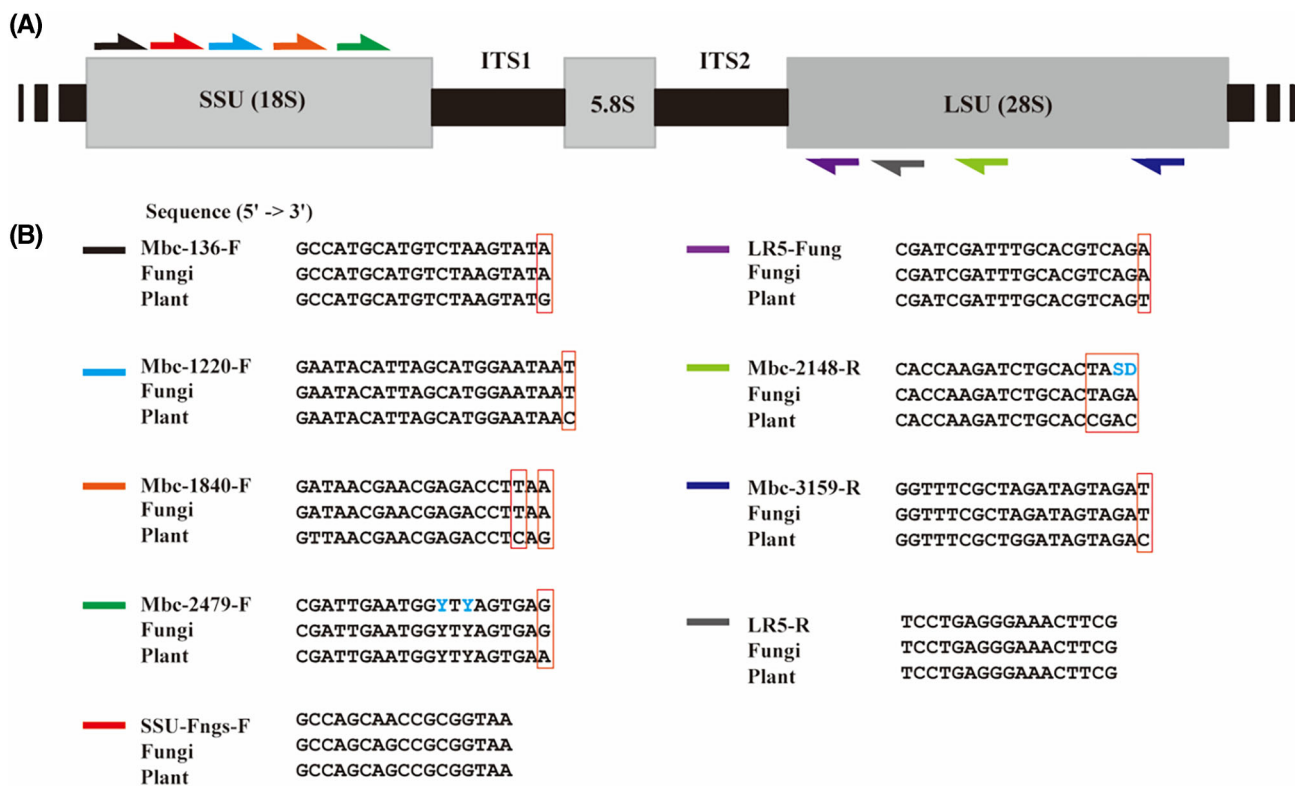
To demonstrate the applicability of our metabarcoding primer design workflow, we targeted the endophytic mycobiome of *V. minor* (lesser periwinkle) with long-read sequencing of the rDNA locus. *V. minor* belongs to the order of Gentianales in the Apocynaceae family and its microbiome remains largely unexplored. In our initial long-read amplicon sequencing experiments with published primers, we observed that 99% of reads were derived from the *V. minor* rDNA locus (data not shown). Therefore, we considered this the ideal test case for the performance of our holobiont metabarcoding workflow.

Since we wanted to target the entire rDNA locus, we created two separate MSAs for SSU and LSU of fungal (target) and Gen-

tianales (plant, exclusion) sequences. Using *mbc-prime* with default settings, we obtained 147 candidate SSU primers and 371 LSU primers. From the list, we selected primers that had scores of 0.6 and higher and carried at least one mismatch at the 3' end when compared to plant (exclusion) sequences. With these criteria we obtained four forward primers in different regions of the SSU and two reverse primers in different regions of the LSU (Fig. 3a). Following careful examination of the GC content, the predicted melting temperatures, and the respective primer binding sites within the MSA, we manually optimized the primer sequences to establish the final sequences (Fig. 3b, [Supplementary File 2, Dataset S7](#)). Primer mbc-1220-F was extended by two nucleobases at the '5 end, mbc-2148-R was shortened by one base at the '5 end, and in mbc-2479-F and mbc-2148-R we incorporated degenerate positions to maximize the coverage of the target group (Fig. S1).

To assess the performance of the final primers *in silico*, we used the SILVA TestProbe 3.0 tool and included several primers from literature for completeness (Tedersoo et al. 2008, 2015, Nilsson et al. 2019, Wurzbacher et al. 2019). All of the new primers show a wide coverage across fungi, similar to the literature primers (Table 1), while showing little coverage of Chloroplastida. Only the literature primers RCA95m and LR5-Fung were similarly discriminative as the new primers. Since the sequence of LR5-Fung was also among the top scoring candidates identified by our *mbc-prime* tool, we included it in our further analyses.

Next, we performed an *in silico* analysis of all possible primer combinations to identify those with favorable parameters, such as a low difference in melting temperatures and a low likelihood of heterodimer formation. Subsequently, we carried the 10 primer pairs that met our quality standards (pairs 1–10; [Table S1, Table S2](#)) and one primer pair from literature ("3.5 kb") (D'Andrea et al. 2021) forward for experimental validation.



**Figure 3** A. Illustration of primer locations in the rDNA locus consisting of SSU, ITS1, 5.8S, ITS2, and LSU; B. Sequence alignments of primers with the consensus sequences of fungi (target set) and plant (exclusion set). Degenerate bases (marked in blue): Y = C/T, S = C/G, D = A/G/T.

### Experimental evaluation of the designed fungus-specific primers using mock community

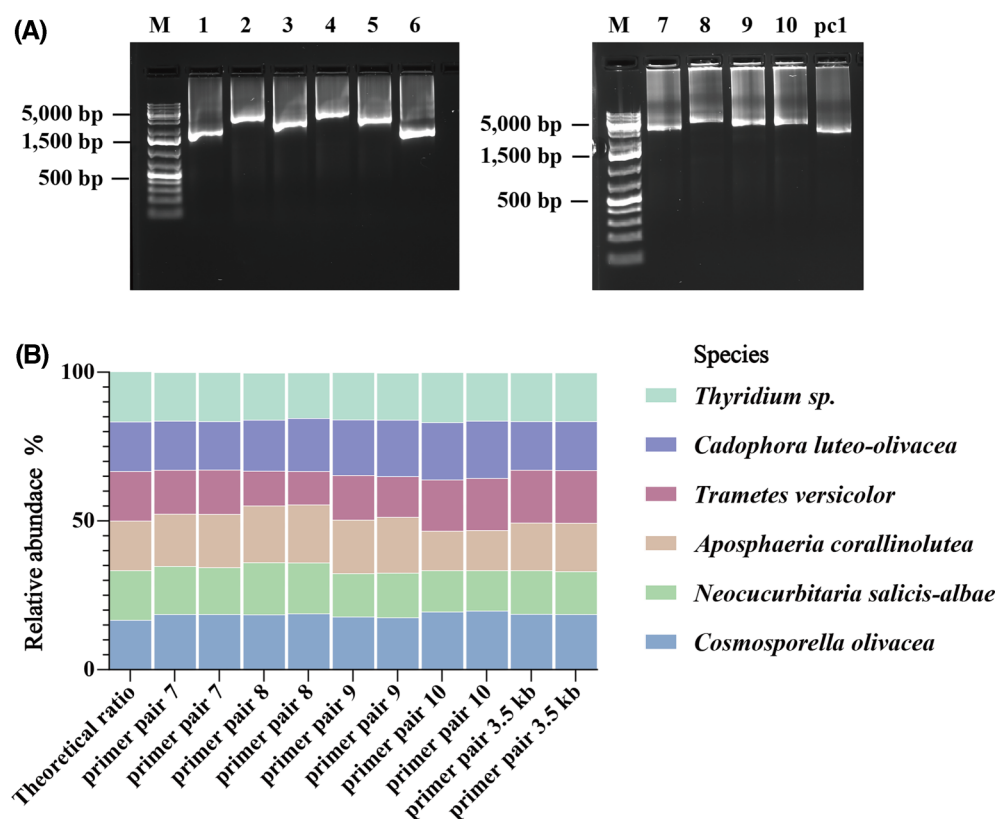
We first generated a standardized DNA sample for a fungal mock community covering a wide taxonomic range and different levels of GC content (Table S3). With this sample, we optimized the annealing temperatures until all primer pairs yielded clear bands of the appropriate molecular weight in agarose gel electrophoresis (Fig. 3a). Since it was our primary goal to generate long amplicons covering both SSU and LSU, we abandoned primer pairs 1–6 after this step and proceeded with multiplex sequencing of the amplicons generated with primer pairs 7–10. All steps from PCR to sequencing were performed with technical duplicates and in parallel to positive (3.5 kbp primer pair) and negative (nuclease-free water) controls. With one sequencing run, we generated 303 100 raw reads with a yield of 1.1 Gb and a mean Qscore of 18 (accuracy: 98%; Table S4).

To classify the trimmed and filtered reads, we applied the *wf-metagenomics* workflow with Kraken2 and a custom database comprising the whole genome sequences of the six isolates (Supplementary Note S2). Using this custom database, 99.80% of reads across all samples were classified, indicating a high sample quality with minimal contamination or sample preparation artefacts, and a high sequencing accuracy. All samples showed similar relative abundances of species, although the basidiomycete *Trametes versicolor* seemed to be underrepresented in amplicons generated with primer pair 8. This indicates that primer pairs 7, 9, and 10 successfully amplify rDNA regions across diverse fungal isolates closely matching the expected composition of the mock community (Fig. 4B).

### Experimental evaluation of the designed fungus-specific primers using genomic holobiont DNA of *V. minor*

Having confirmed the amplification efficacy of the new primer pairs on the mock community, we next assessed their performance in a real-world holobiont scenario. When performing the first amplification step with the *V. minor* holobiont DNA, primer pair 8 failed to yield a high-quality amplicon, likely due to low target DNA abundance (Figure S3). Since we could not overcome this problem despite various optimization attempts (annealing temperature, additives, polymerase), we abandoned primer pair 8.

Proceeding with the remaining primer pairs, we generated a total of 599 335.274 reads with a Qscore greater than 18 in a single multiplex sequencing run (Table S5). When using the same classification approach as before, this time with a custom database comprising the NCBI RefSeq Targeted Loci database (ncbi\_16s\_18s\_28s\_ITS) and the rDNA locus of a closely-related *Vinca* species (NCBI taxon ID 141628), over 92% of all reads were successfully classified (Fig. 5). As expected, a large proportion of the reads generated with the 3.5 kbp primer set is classified as host (*Vinca* sp.), while the new primer pairs did not generate such amplicons. In terms of species composition, the samples generated with the new primer pairs show similar abundances of the most abundant taxa with only slight variations. This demonstrates that our newly-designed primer pairs are indeed able to efficiently discriminate between the rDNA locus of fungi and the host plant and that they provide a similar reflection of the community composition.



**Figure 4** Metabarcoding of a fungal mock community. (a) Gel electrophoresis of rDNA amplicons generated from gene specific primers including overhangs for PCR barcoding; Sample order: M: 1 kb plus DNA ladder; 1–10: amplicons generated with primer pairs 1–10 respectively; pc1: positive control generated with primer pair 3.5 kb. All primer pairs yielded products in the expected size range. (b) Relative abundance of taxonomic groups classified based on the sequencing reads of amplicons generated from the mock community with the different primer pairs (each primer pair has two replicates). Species belonging to the same family are displayed in the same color with different fill patterns. All primer pairs reproduce the overall structure of the mock community, albeit with different biases for different groups.

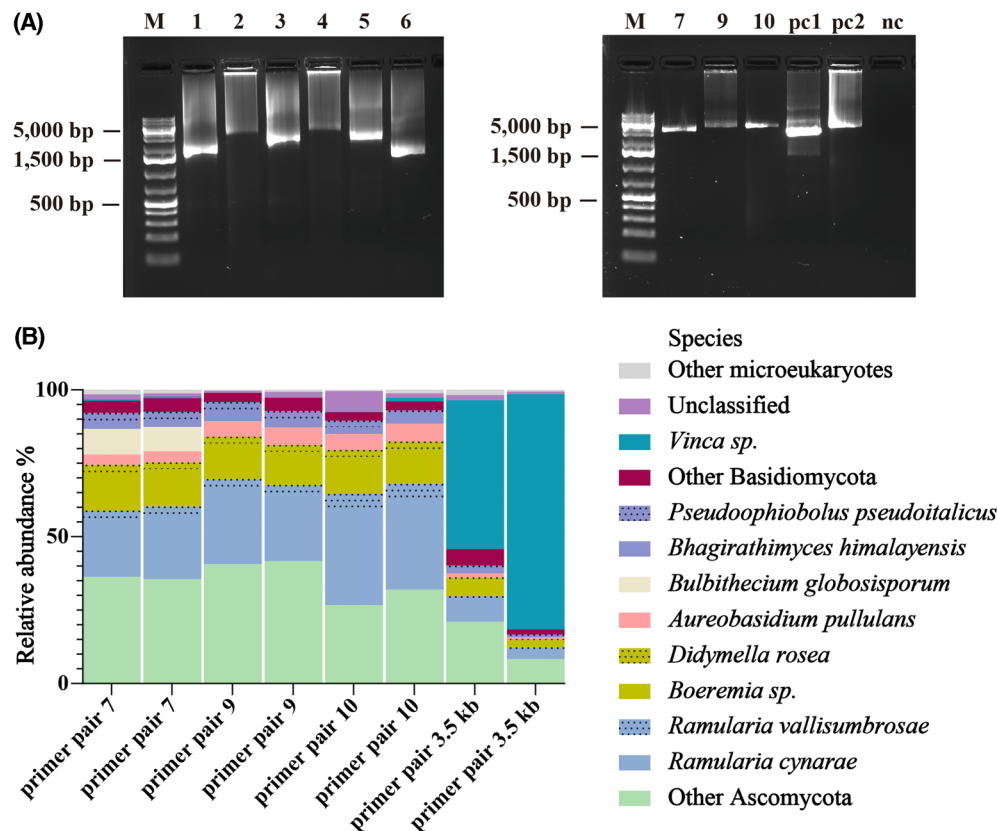
For completeness, we also analyzed the data using the SILVA database, which includes a broader range of eukaryotes. For amplicons generated with primer pair 7 97% of the reads were classified as fungi, with primer pair 9 95%, and with primer pair 10 97%. Overall, none of the amplicons generated with the new primer pairs were classified as plant. Interestingly, a subset of amplicons was assigned to protist taxa, with the highest proportion observed with primer pair 9 (~4.5%). This indicates that while all primer pairs can successfully exclude the plant rDNA locus, primer pair 9 may additionally be useful for targeting other microeukaryotes, which requires further investigation (Figure S3).

## Discussion

In molecular biology, the presence of non-target DNA (e.g. host DNA) can interfere with the detection and analysis of target DNA (e.g. microbial, pathogen, or environmental DNA). This is particularly problematic in fields like microbiome research, clinical diagnostics, and environmental DNA studies, where target DNA is often present in low abundance relative to other DNA. To address this issue, various blocking methods have been developed to selectively suppress the amplification or sequencing of non-target DNA. These techniques involve the use of chemically modified oligonucleotides, such as C3 spacers, locked nucleic acids (LNAs), or peptide nucleic acids (PNAs), which are designed to bind conserved

regions of non-target DNA. The goal is to either direct the binding of blocking oligos to these modified sequences or to physically block polymerase elongation during PCR amplification. However, their effectiveness is often constrained by suboptimal binding affinity of the blocking oligos to non-target templates (Vestheim and Jarman 2008), the requirement for complex experimental optimization (Piñol et al. 2015), and limited cost-effectiveness (Azadnia et al. 2023). Given these limitations, the design of highly specific, discriminative primers could be a more practical and scalable alternative for minimizing non-target amplification.

In this study, we developed fungus-specific primers using a host-exclusive primer design workflow for the generation of long amplicons that almost entirely exclude plant DNA as a cost-effective alternative to the blocking methods. High-throughput sequencing of taxonomic marker genes (metabarcoding) is a widely used and powerful technique for biodiversity and microbial ecology. However, many taxa recovered from natural environments remain unidentifiable at the species level or even higher taxonomic levels when working with short reads (Tedersoo et al. 2020). In this case, long-read sequencing has the potential to overcome this problem. For instance, several of the six fungal isolates used for our mock community cannot be unambiguously identified solely based on ITS sequences (amplified with ITS1F and ITS4; data not shown) because they exhibit 100% sequence similarity with multiple taxa. For instance for the *Thyridium oculorum* isolate, ITS-



**Figure 5** Metabarcoding of holobiont DNA of *V. minor*. (a) Gel electrophoresis of rDNA amplicons generated from gene specific primers including overhangs for PCR barcoding; Sample order in both panels: M: 1 kb plus DNA ladder; 1–10: amplicons generated with primer pairs 1–10 respectively; pc1: positive control generated with primer pair 3.5 kb; pc2: positive control generated with primer pair 10 with fungal mock community; nc: negative control (nuclease-free water). (b) Relative abundance of taxonomic groups classified based on the sequencing reads of amplicons generated from the genomic holobiont DNA of *V. minor* with the different primer pairs on species level (each primer pair has two replicates).

based identification returned matches to both *Phialemonium sp.* and *Thyridium sp.*, thus preventing a definitive taxonomic assignment. However, long-read sequencing results from this study showed that 92.15%–99.17% of all reads derived from this isolate were classified as *T. oculorum*, which is consistent with the identification based on multi-locus classification. This highlights that long-read sequencing can provide superior classification results in metabarcoding studies.

There are, however, still common pitfalls that need to be tackled in the future. First, reducing bias in the relative abundance of sequence variants should be a primary objective in enhancing amplicon sequencing methods. It is well established that even PCR primers designed for broad coverage tend to amplify some phylogenetic subgroups more effectively than others. For instance in this study, the inferred abundance of *T. versicolor* in the mock community was consistently lower than the expected ratio, which might be attributed to amplification efficiencies of the used primer pairs. In the experiments with holobiont DNA, we found that the overall community profiles were fairly consistent across primer pairs, though some minor differences were observed in the dominant taxa. This highlights the importance of evaluating primer performance in both controlled and real-world contexts when selecting appropriate primers for a given study.

Second, the accuracy of taxonomic classification is highly dependent on the reference database employed. Currently, these

databases mainly rely on parts of the rDNA locus. The SILVA database includes near full-length sequences of the small and large subunits of the rRNA operon for both prokaryotes and eukaryotes. However, these sequences are insufficient for amplicons spanning the ITS region and generally allow identification only up to the genus level. Despite this limitation, the SILVA database enables the identification of non-fungal sequences, such as those from Chloroplastida and Metazoa (Quast et al. 2012). Conversely, for classification of fungi, the NCBI RefSeq Target Loci database seems to outperform the SILVA database. This superior performance is due to NCBI's focus on the RefSeq database, which includes ribosomal DNA loci (16S, 18S, 28S rDNA genes, and ITS region) and allows for species-level identification, offering better accuracy and resolution in taxonomic classification of microbes. The limitation of this database, however, is that it is focused on archaea, bacteria and fungi, and does not include plants or other micro- and macro-eukaryotes. Therefore, reads originating from such eukaryotes will either not be classified at all or incorrectly classified as fungi. Furthermore, the fungal tree of life is still rather sparsely sampled for molecular barcodes deposited in the RefSeq database. Thus, significant efforts are still needed to improve the database to enhance its completeness and accuracy. This could involve developing databases tailored for eukaryotic full-length rDNA operons (Krabberød et al. 2025) or customizing databases to meet specific research needs.

## Conclusions

In the current study, we developed a host-exclusive primer design workflow, which is centered around a new computation tool, *mbc-prime*, to design discriminative metabarcoding primers with the option to exclude the DNA amplification of nontarget taxa, for instance the host in host-associated microbiomes. We demonstrated the use of *mbc-prime* when targeting endophytic fungi associated with the medicinal plant *V. minor*. Initially, we evaluated our custom-designed primers *in silico*, demonstrating high coverage of fungal lineages with low coverage to Chloroplastida. Subsequent evaluation using a mock community revealed that the relative abundance of amplicons accurately reflected the community composition of the mock community. Moreover, when applied to plant holobiont samples, our primers successfully targeted a broad range of endophytic fungi while almost completely excluding the plant genome. This specificity has the potential to shed light on the roles of endophytic fungi in plant health and ecosystem dynamics. In addition, this demonstrates that our workflow, incorporating the *mbc-prime* tool, can effectively be used to study the composition of complex host-associated microbiomes. Furthermore, the primer design workflow developed in this study is universally applicable for designing discriminative primers targeting particular phyla, offering a user-friendly and practical solution for researchers in microbiome or other biological systems studies.

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## Author contributions

T.He (Conceptualization, Data curation, Formal Analysis, Funding acquisition, Investigation, Methodology, Visualization, Writing – original draft, Writing – review & editing), K.J. (Investigation, Software, Writing – review & editing), X.L. (Investigation, Writing – review & editing), A.M.R. (Formal Analysis, Investigation, Writing – review & editing), B.S. (Investigation, Writing – review & editing), R.S. (Investigation, Writing – review & editing), T.H. (Conceptualization, Formal Analysis, Methodology, Resources, Software, Writing – original draft, Writing – review & editing), K.H. (Conceptualization, Data curation, Funding acquisition, Project administration, Resources, Supervision, Writing – original draft, Writing – review & editing).

## Supplementary material

Supplementary material is available at *FEMSEC Journal* online.

## Conflicts of interest

None declared.

## Data Availability

The *mbc-prime* tool is available on GitHub (<https://github.com/thackl/mbc-prime>) and on Zenodo (<https://doi.org/10.5281/zenodo.14218237>). Sequencing reads are available on the European Nucleotide Archive (ENA) under the accession PRJEB82852. Other research data are available on Zenodo (<https://doi.org/10.5281/zenodo.16411378>) and in the supplementary files.

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