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High throughput screening of fungal phytopathogens caught in Australian forestry insect surveillance traps

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Post-border surveillance for forestry's high priority pests and pathogens is conducted routinely through established programs focused on the main points-of-entry and across the major plantation growing regions. Currently, most diagnostic protocols used to identify fungal phytopathogens sampled during these surveys rely on traditional methods, such as morphological examination and DNA barcoding techniques. This stepwise process from isolation to species identification is often regarded as slow, expensive, and limited due to the need for disease manifestation and/or comprehensive expertise for rapid and accurate detection. In this study, we applied a recently validated high-throughput, dual-marker amplicon sequencing approach on insect surveillance traps from across Australia to assess its performance for the targeted surveillance of the *Ophiostomatales*, an order of fungi comprising notable phytopathogens which are vectored by bark beetles. By using a recently validated assay we were able to confidently characterize a range of Ophiostomatalean taxa known to be present in Australia, while reporting eight first detections from environmental DNA. Our study demonstrates the value of targeted multi-barcode amplicon sequencing for high-throughput screening of fungi caught in post-border surveillance traps, in addition to emphasizing research priorities that require further investigation before such methods can be implemented routinely for biosecurity.

KEYWORDS

forest biosecurity, environmental DNA sequencing, metabarcoding, fungal diagnostics, bark beetles, *Ophiostomatales*

1. Introduction

The combination of climate change and international trade and travel continues to drive organisms into new environments, increasing the likelihood of global range expansion and subsequent establishment of invasive pests and pathogens (Pyšek et al., 2010; Bebber, 2015). Invasive pests and pathogens pose a significant threat to forest ecosystems globally (Food and Agriculture Organization of the United Nations [FAO], 2020). Fungi are particularly worrying as they constitute arguably the largest threat to global biodiversity (Fisher et al., 2012), having enormous impacts on forest ecosystems through large scale population declines and, in severe cases, functional extinctions (Santini et al., 2013; Hughes et al., 2017; Bonello et al., 2020). As a result, forest biosecurity systems have been brought under review to address the needs for better harmonization, international collaboration, and continued investment to ensure better preparedness in the face of these invasive threats (Bonello et al., 2022). In Australia, a National Forest Pest Surveillance Program (NFPSP) has recently been established to fill gaps in post-border surveillance activities specific to the forest industry (Carnegie et al., 2022).

To date, post-border detections of exotic forest pests in Australia (Carnegie and Nahrung, 2019) and elsewhere (Wilson et al., 2004; Tobin et al., 2014; Gould, 2015) have largely been made through general/passive surveillance activities, such as those reported by the public, researchers, or industry. While effective, detections made through passive surveillance methods have a lower probability for successful eradication because the incursions will have likely extended beyond the initial phases of establishment by the time it is reported (Wilson et al., 2004; Carnegie and Nahrung, 2019; Paap et al., 2022). The NFPSP aims to increase the chance of early detection and eradication with a newly proposed framework encouraging better coordination across surveillance activities and risk-based analysis targeting entry pathways (Carnegie et al., 2022). Ideally, this will aim at detecting forestry's high priority pests (Plant Health Australia [PHA], 2022b) at high-risk sites through tree health assessment and/or insect trapping. The success of post-border surveillance activities are, however, fundamentally dependent on the underlying diagnostics framework.

At present, most diagnostic protocols targeting fungal phytopathogens rely on traditional microbial isolation techniques followed by DNA barcoding or, in certain cases, pathogen-specific molecular assays such as quantitative PCR performed on symptomatic tissue (ities). This is due to a dependence on symptom development which needs to be recognized in-field, before skilled technicians or pathologists work through stepwise processes to make a reliable detection (Mumford et al., 2016). Screening for potential vectors through insect trapping does not require symptom development and offers a better opportunity National Plant Biosecurity Diagnostic Network [NPBDN], 2022). Both diagnostic approaches are, however, limited by their targeted nature and not well suited for early detection during general surveillance activ to detect fungal pathogens that may be cryptic during the early stages of invasion. To this end, innovative approaches which rapidly screen for forest phytopathogens using high-throughput molecular techniques are increasingly desirable (Tremblay et al., 2018; Aguayo et al., 2021; Berube et al., 2022).

While species-specific assays, such as quantitative PCR or loop-mediated isothermal amplification (LAMP), currently offer the most rapid and sensitive means of detecting pathogens from asymptomatic tissue (Luchi et al., 2020), they remain targeted to single/few species and are unlikely to be used in general surveillance activities which routinely monitor against lengthy target lists (Piombo et al., 2021). In this regard, high-throughput and comprehensive screening tools are needed to survey for the multitude of invasive pathogens that may be present within asymptomatic tissues or caught within surveillance traps (Bulman et al., 2018; Piombo et al., 2021).

Amplicon sequencing (metabarcoding) of environmental DNA (eDNA), defined here as the total DNA extracted from a given substrate (Taberlet et al., 2018), offers a robust, cost-effective, and high-throughput screening approach for early detection and broad-spectrum monitoring of exotic pests and pathogens (Comtet et al., 2015; Bulman et al., 2018). It is well established that eDNA metabarcoding is suitable for screening bulk insect trap catches, for both pests and pathogens, with several studies demonstrating the opportunities of applying the technique within existing surveillance frameworks (Malacrino et al., 2017; Batovska et al., 2018; Tremblay et al., 2018, 2019; Piper et al., 2022). Despite this, the reliability of detection and accuracy of identification when targeting fungal phytopathogens remains an important challenge (Lücking et al., 2020). Careful decision-making and optimization are required across various phases of an amplicon sequencing pipeline to achieve species-level resolution (Cobo-Díaz et al., 2019; Tedersoo et al., 2019).

We recently established a modular, dual-barcode amplicon sequencing pipeline to improve on high throughput diagnostics of the *Ophiostomatales* (Trollip et al., 2022), an economically important order of fungi that exhibits many of the key challenges to fungal species identification (Hulcr et al., 2020; Lücking et al., 2020). The *Ophiostomatales* are well-known for their associations with bark and ambrosia beetles (Curculionidae: Scolytinae, Platypodinae), benefiting from transport by their insect vectors (Six, 2012). While some of these associations can become catastrophic in novel environments (Fisher et al., 2012; Wingfield et al., 2017), others result in less severe disease and cause blue stain (sap stain) in the wood of economically valuable tree hosts (Seifert et al., 2013). Consequently, Australian forest biosecurity routinely targets bark and ambrosia beetles through insect trapping using semiochemical lures (Bashford, 2008), monitoring the already established pests in plantation forests [e.g., *Ips grandicollis* (Eichhoff) which vectors the blue stain fungus *Ophiostoma ips* (Rumbold)] and screening for potential exotics at high-risk sites (e.g., *Ips* and *Dendroctonus* spp.; Plant Health Australia [PHA], 2022b) (Carnegie et al., 2018, 2022). At present, diagnosis of trap catches during forest biosecurity surveillance is only conducted on insects, leaving an opportunity for Australian biosecurity agencies to improve their capacity by incorporating methods that can simultaneously screen for phytopathogens within the same traps. It is well established that semiochemical-baited insect traps are effective for fungal surveillance (Malacrino et al., 2017; Tremblay et al., 2018, 2019; Berube et al., 2022).

The aim of this study was to evaluate the efficacy of our dual-barcode amplicon sequencing approach for high throughput screening of fungi caught within insect surveillance traps collected across diverse geographic regions with variable environmental

conditions and collection methods. To achieve this, we screened 45 insect panel traps collected from already established forest biosecurity surveillance programs across seven Australian states and territories. Our objectives included (i) an initial assessment of sample preparation for this specific trap type and (ii) to undertake targeted surveillance of the *Ophiostomatales*. This survey functions as a proof-of-concept for how high throughput sequencing approaches might be implemented in post-border surveillance activities for Australian forest pests and pathogens, and to identify where future research activities will continue to improve on their utility.

2. Materials and methods

2.1. Trap collection and sample processing

During post-border surveillance activities performed in the summer of 2020–2021, state agencies across Australia submitted insect panel-trap catches from 45 different locations (Figure 1 and Table 1). Trap catches were predominantly shared from within existing surveillance frameworks, including high risk site surveillance around ports-of-entry and in *Pinus* plantations where forest health surveillance activities are routinely conducted (Carnegie et al., 2018, 2022). Intercept panel traps (Alpha Scents, Inc, OR, USA and ChemTica Internacional, Heredia, Costa Rica) contained semiochemical lures designed to attract beetles of the Cerambycidae and Curculionidae families (Table 1), with insects collected into a preservative drowning fluid (Table 1) within collection cups attached at the bottom of each trap. Trap catches were collected monthly, with the preservative fluid replaced at this time. Trap catches, comprising a mix of target insect groups and bycatch, were either stored at 4°C or ambient temperature before being transported to the laboratory for further analysis.

Bulk trap catches were stored at 4°C for 10 to 17 months until they could be processed. For each trap catch, the sample contents were agitated by swirling and vortexing for 1 min before filtration through a sterile 100 µm Falcon mesh filter (Corning, AZ, USA). This resulted in the division of each trap catch into two subsamples, either comprising of a standardized volume of the preservative fluid (50 ml; excess discarded) or the bulk insect specimens, respectively. Preservative fluid subsamples were vacuum filtered through an EZ-Fit™ Filtration Unit (0.45 µm MCE membrane) using the EZ-Stream™ pump (Merck, Darmstadt, Germany). Vacuum filtration was performed for 5 min per sample or until all 50 ml of the preservative fluid passed through the filter membrane. Membrane filters for each sample were then cut in half using a sterile scalpel blade, with one half being used immediately for DNA extraction and the other stored as a backup at –20°C. Bulk insect specimen subsamples were stored in absolute ethanol at 4°C until DNA extractions and morphological assessment could be performed.

2.2. DNA extractions

DNA extractions were performed on both preservative fluid and bulk insect specimen subsamples in this study. An initial

evaluation of sample processing was performed for these trap types using a subset of 15 traps collected from the Southern Tablelands in New South Wales (NSW; Tumut region, $n = 12$) and the Australian Capital Territory (ACT; $n = 3$) (Table 1 and Figure 1). Extractions from preservative fluid subsamples were performed using the DNeasy® PowerSoil® Pro Kit (Qiagen, Hilden, Germany). Membrane filter halves (prepared as described in section “2.1. Trap collection and sample processing”) were suspended in 800 µl of Solution CD1 in PowerBead Pro tubes and incubated at 65°C for 10 min. Filter homogenization and spore lysis was then performed using a FastPrep 24 system (MP Biomedicals, Santa Ana, CA, USA) with the velocity set at 4 m/s for two 30 s cycles. The rest of the protocol followed the manufacturer’s instruction. Extractions performed on bulk insect specimen subsamples were also conducted with the DNeasy® PowerSoil® Pro Kit, however the sample preparation and lysis phase of the protocol was amended to accommodate a non-destructive workflow so that insect specimens could be retained for downstream morphological assessment (Batovska et al., 2021). First, the storage ethanol was removed from each bulk insect specimen subsample using a 1,000 µl pipette followed by air-drying in a laminar flow. After drying, the contents were resuspended in a 10:1 mix of Solution CD1 (800 µl minimum) and Proteinase K (Qiagen; 80 µl minimum) ensuring all insects were thoroughly submerged in the lysis solution. Samples were then vortexed and incubated at 56°C overnight. After incubation, the lysate from each sample was transferred to a PowerBead Pro tube and incubated at 65°C for 10 min with further mechanical lysis performed on the FastPrep system. Bulk insect subsamples containing the voucher specimens were then resuspended in absolute ethanol and stored at 4°C.

2.3. PCR amplification, library preparation and sequencing

Dual-barcode amplicon sequencing was performed as described in Trollip et al. (2022) using adapter-modified primer pairs of ITS1Fngs (5'-GGTCATTTAGAGGAAGTAA-3') and ITS2ngs (5'-TTYRCKRCGTTCTTCATCG-3') (Tedesoo et al., 2015, 2018) to target the internal transcribed spacer 1 (ITS1) barcode, and EF1087Fngs (5'-GTTCGAGGCTGGTATCTCC-3') and EF646Rngs (5'-GTCTCRATACGGCCRAC-3') (Trollip et al., 2022) to target the translation elongation factor 1-alpha (TEF1α) barcode. PCR amplification was performed independently for each barcode, with three PCR replicates amplified per sample. Successful amplifications were confirmed using gel electrophoresis (2% w/v agarose) before libraries were prepared for Illumina sequencing. Briefly, amplicons were cleaned and normalized using a SequalPrep™ Normalization Plate Kit (Thermo Fisher Scientific, MA, USA) before equal volumes of eluted product from each amplified barcode were pooled according to sample. Dual unique indexes, along with the remainder of the Illumina adapter sequence, were then annealed to each sample’s amplicon pool using qPCR. Indexing reactions consisted of 10 µl of 5 × Phusion HF Buffer (Thermo Fisher Scientific), 1 µl of dNTP mix (10 mM), 1 µl of 1/1,000 SYBR Green I Mix, 0.5 µl Phusion DNA polymerase (Thermo Fisher Scientific), 4 µl of

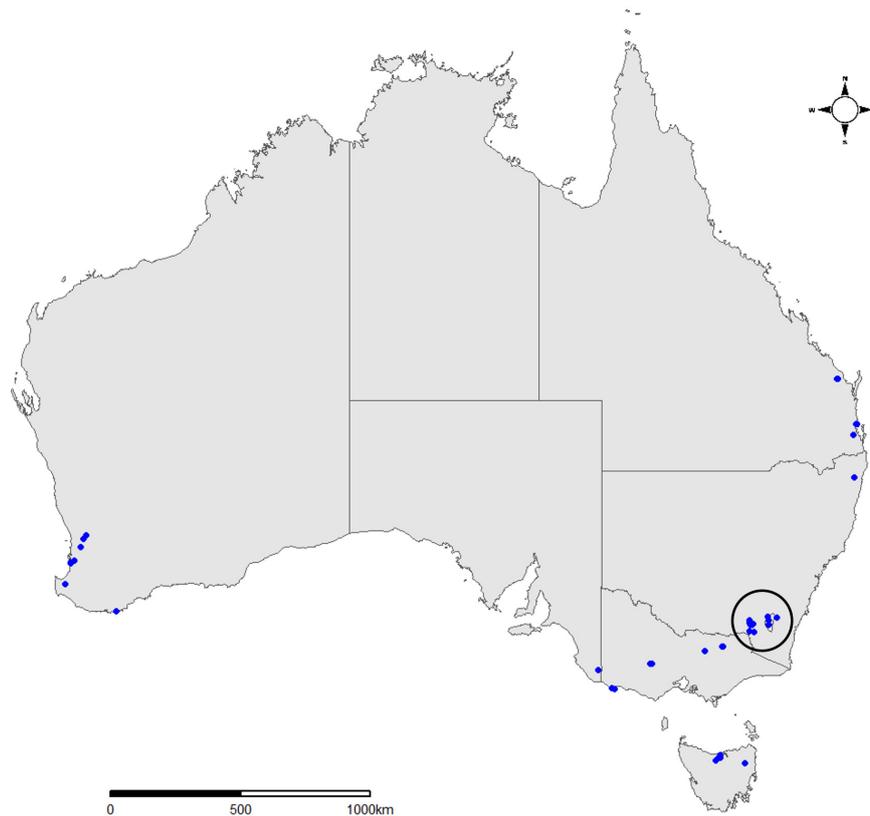


FIGURE 1

Map of Australia showing 45 sampling locations across seven states and territories from which trap samples were collected during the current study. Samples encircled from the Central Tablelands represent a 15-trap subset used to evaluate sample processing and barcode amplification from forestry surveillance traps. This figure was produced in R using the ggplot2, ggmap, ozmaps and ggsn R packages (Kahle and Wickham, 2013; Wickham, 2016; Sumner, 2021).

sample-specific indexes (2.5 μ M), and 10 μ l (5–10 ng template) of sample-pooled amplicon product. Cycling conditions were 98°C for 30 s, followed by seven cycles of 98°C for 10 s, 65°C for 30 s, and 72°C for 30 s. Indexed amplicon libraries were cleaned and normalized for a second time using SequelPrep™, before being pooled equally to generate the final library. Final library QC was performed using a 2200 TapeStation (Agilent Technologies, Santa Clara, CA, USA), with final purifications completed using ProNex® Size-Selective Purification beads (Promega, WI, USA). Sequencing was performed in-house on an Illumina Miseq platform using V3 chemistry (2 × 250 bp) with a 15% PhiX spike-in.

2.4. Bioinformatics and data analysis

Raw sequencing reads were initially demultiplexed according to sample using bcl2fastq v2.2.0, before being further split into each barcode using the PCR primers as indexes with Seal in BBTools v38.9 (Bushnell, 2017). Demultiplexed reads were then trimmed of primers and sequencing adapters using BBDuK (BBTools suite) and filtered according to sequence quality profiles. Any reads and read tails that were determined to be of low-quality [based on expected error models (Edgar and Flyvbjerg, 2015)] or that contained ambiguous bases were removed. Filtered reads were then processed in DADA2 v1.20 (Callahan et al., 2016) with

inferred amplicon sequence variants (ASVs) undergoing *de novo* chimera removal (removeBimeraDenovo). Sequence variants were length filtered (100–450 bp) before an initial taxonomic assignment performed using IDTAXA (Murali et al. (2018); trained on in-house reference sets (Trollip et al., 2022). The TEF1 α reference database was expanded in this study to include available fungal sequences from the NCBI RefSeq database (O’Leary et al., 2016). All downstream analysis and data processing was performed using phyloseq v1.36.0 (McMurdie and Holmes, 2013) and tidyverse v1.3.1 (Wickham et al., 2019) R packages. Final ASVs were filtered according to the following criteria for detection: (i) a relative read abundance (RRA) value above the calculated switch rate threshold [determined by estimating the proportion of sequences that switched at both ends of the amplicon (Wilcox et al., 2018)] and (ii) presence in at least 2/3 PCR replicates (Alberdi et al., 2017). Before performing alpha diversity assessments, the breakaway v4.7.6 R package was used to estimate the number of unobserved species across samples, both before and after merging PCR replicates (Willis and Bunge, 2015; Willis et al., 2020). Alpha diversity metrics were then calculated on the merged datasets (i.e., taxa present if in 2/3 PCR replicates) using phyloseq, with breakaway and the DivNet v0.4.0 R packages used for hypothesis testing and batch-wise comparisons (Willis and Martin, 2022). Linear mixed-effect models [betta and bettarandom (Willis et al., 2015, 2017)] were used to test if any observed differences in the alpha diversity could

TABLE 1 Metadata for the trap batches collected and screened in this study.

Australian states and territories	Collection batch	Growing region	Collection date/s	Traps in batch	Trap number	Preservative fluid ^c	Semiochemical lure ^d
Australian Capital Territory	ACT	–	March 2021	3	1–3	Anti-freeze	Ipsenol + α -Pinene
New South Wales	NSW 1	Tumut	October 2020	12	4–15	Glycol	Ipsenol + α -Pinene
	NSW 2	Whiporie	February 2021	3	23–25		
Queensland	QLD 1	Beerburum	February 2021	4	17–20	Glycol	Superlure
	QLD 2	Elliott		2	26, 27		
South Australia	SA	Nangwarry	March 2021	1	28	Glycol	Ipsenol + α -Pinene
Tasmania	TAS	–	February 2021	5	42–46	Glycol	Ipsenol + α -Pinene
Victoria	VIC 1	Portland	February 2021	2	21, 22	Glycol	Ipsenol + α -Pinene
	VIC 2	Benalla	March 2021	1	29		
	VIC 3	Shelley	May 2021	2	30, 31		
	VIC 4	Mount Lonarch	March 2021	2	40, 41		
Western Australia ^{a,b}	WA 1	Myalup	January 2021, March 2021	4	32–35	Anti-freeze	SIRNOC
	WA 2	The Lakes	2020–2021	3	36–38		
	WA 3	Albany	April 2021	1	39		

Table data summarizes the locations across major growing regions and/or points-of-entry, date of collection, the number of traps collected per batch, individual trap numbers, as well as the type of preservative fluid and lure used.

^aThe single trap sample from Albany, WA was a pooled sample of 3 traps collected from across the region.

^bEach of the three traps of batch WA 2 represents a pooled collection over the entire trapping season.

^cGlycol = a 1:1 mixture of 100% propylene glycol and water. Anti-freeze = a 1:1 mix of water and a car coolant composed of either ethylene and/or propylene glycol bought at the discretion of biosecurity officer/s.

^dSIRNOC lure (Alpha Scentis, Inc.) is composed of alpha pinene (+), alpha pinene (-) and beta pinene (-). The “Superlure” is composed of fuscumol, fuscumol acetate, geranyl acetone, monochamol, 3-hydroxyhexan-2-one, anti-2,3-hexanediol, prionic acid and 2-methylbutanol, (\pm -)-ipsdienol, cis-verbenol, ipsenol, α -pinene and ethanol.

be explained by the time traps spent in storage, with storage time considered a fixed effect and the collection batches considered random effects. All statistical analyses and figures were conducted or generated within the R v4.1.0 programming environment using the software listed above (R Core Team, 2013; Wickham, 2016). Finally, *Ophiostomatales*-specific ASVs were further curated in Geneious Prime[®] where additional reference-based chimera filtration was performed with UCHIME v4.2.40 and USEARCH v8.1 (Edgar, 2010). The remaining *Ophiostomatales* ASVs were aligned to publicly available reference sequences using MAFFT v7.450 (Katoh and Standley, 2013), where any remaining artifactual sequences were removed. Species-level taxonomic assignment was determined phylogenetically using RaxML v8.2.11 (Stamatakis, 2014).

3. Results

3.1. Sample processing and barcode amplification

Results from the initial sample processing comparison, performed on the 15 traps collected in Tumut, NSW and the ACT, showed that PCR amplification was most consistent from the preservative fluid DNA extractions compared to the bulk insect specimen subsamples (Table 2). This was consistent for both primer pairs, with the ITS1 primer pair performing most efficiently with successful amplification for all preservative fluid extracts. The TEF1 α primer pair had an amplification success rate of 93% on the preservative fluid extracts, however, the efficiency of amplification was deemed lower than the ITS1 as five samples were observed with weak amplification (Table 1 footnote). For the bulk insect specimen sub-samples, PCR amplification was far less consistent with success rates of 40% for ITS1 and 20% for TEF1 α . Additional PCR attempts were made using diluted template (1/10 and 1/100) to try and improve the recovery of amplicons; however, amplification remained unsuccessful (data not shown). Due to this, PCR amplification and sequencing of the remaining samples was performed on the preservative fluids only. Amplification results for the expanded dataset were consistent for the ITS1 barcode in 45/45 samples and the TEF1 α barcode in 44/45 samples. The PCR efficiency of the TEF1 α primer pair was again deemed lower than that of the ITS1 pair, with nine of the samples observed to have weaker amplification (data not shown).

3.2. Fungal composition of forest surveillance traps

A total of 21,205,095 reads were demultiplexed from the MiSeq run, with 10,151,570 ITS1 reads and 1,443,255 TEF1 α reads recovered after quality filtering. Read distributions differed for the two barcodes with 73,562 (\pm 35,027 SD) ITS1 sequences and 10,852 (\pm 6,258 SD) TEF1 α sequences recovered per sample. Reads assigned to an unclassified *Malassezia* ASV were detected in one of the blank extractions and the contaminant was subsequently removed from the dataset. For each sample, the number of observed ASVs was within the bounds of the species richness

TABLE 2 PCR amplification success of preservative fluid and bulk insect specimen subsamples performed on the initial 15 trap subset sourced from Tumut, NSW (NSW1) and the ACT.

Barcode target	Primer pair	Sample type	Successful PCR ^a
ITS1	ITS1Fngs-ITS2ngs	Preservative fluid	15/15
		Bulk insect specimens	6/15 (2)
TEF1 α	EF1087Fngs-EF646Rngs	Preservative fluid	14/15 (5)
		Bulk insect specimens	3/15 (1)

^aNumber in parentheses represents the number of samples with weak amplification according to gel electrophoresis (i.e., only faint band observed).

projections from *breakaway*, confirming appropriate sequencing depths were achieved (Supplementary Tables 1, 2). After merging PCR replicates and accepting only detections present in 2/3 replicates, a total of 3,422 and 1,683 ASVs were recovered for the ITS1 and TEF1 α datasets, respectively. The observation counts of ASVs remained at $81 \pm 13\%$ per sample for the ITS1 dataset, and $55 \pm 13\%$ per sample for the TEF1 α dataset (Supplementary Tables 1, 2) after merging.

The ITS1 dataset provided the most complete characterization of the fungal communities caught within the forest surveillance traps, where a total of 29 classes and 465 genera were identified. The top 10 most abundant fungal classes were the Dothideomycetes (Lowest 0.02 – (Mean 23.49 ± 16.44) – Highest 64.93%), Saccharomycetes (0.04 – (21.39 ± 28.78) – 99.20%), Tremellomycetes (0.04 – (12.82 ± 13.41) – 46.40%), Sordariomycetes (0.003 – (8.55 ± 14.13) – 87.75%), Cystobasidiomycetes (0.006 – (9.20 ± 12.99) – 55.52%), Eurotiomycetes (0.02 – (5.96 ± 7.30) – 35.07%), Agaricomycetes (0.001 – (4.50 ± 7.00) – 32.95%), Microbotryomycetes (0.04 – (8.27 ± 15.62) – 79.46%), Leotiomycetes (0.002 – (2.22 ± 4.01) – 19.12%), and the Pezizomycetes (0.01 – (7.04 ± 18.47) – 91.53%) (Figure 2A). Alpha diversity comparisons revealed significant differences in the observed richness and Shannon diversity explained by both storage time and collection batch, however the effect of storage time was only minor when collection batch was accounted for as a random effect (Estimated decrease in richness after storage of 0.67 ± 0.06 , $p < 0.001$; Shannon diversity decrease after storage of 0.003 ± 0.0001 , $p < 0.001$) (Supplementary Tables 3, 4). Batch-wise comparisons revealed that traps collected in NSW1 (Richness: 165.01 ± 32.68 ; Shannon diversity: 2.82 ± 0.02 ; both with $p < 0.001$), TAS (Richness: 163.00 ± 45.34 , $p = 0.001$; Shannon diversity: 1.86 ± 0.32 , $p < 0.001$), VIC1 (Richness: 152.82 ± 71.69 , $p < 0.05$; Shannon diversity: 1.10 ± 0.54 , $p < 0.001$), VIC2 (Richness: 87.27 ± 102.05 , $p < 0.05$; Shannon diversity: 2.47 ± 0.52 , $p = 0.05$), and VIC4 (Richness: 74.44 ± 71.69 , $p = 0.001$; Shannon diversity: 0.44 ± 0.60 , $p < 0.001$), contained significantly fewer species and lower diversity than the ACT (Figure 2B), while only NSW2 (Richness: 435.97 ± 58.53 , $p < 0.05$) exhibited greater richness (Supplementary Tables 3, 4).

Taxonomic classification was more difficult for the TEF1 α dataset where a large proportion of reads remained unclassified at the fungal class rank within each sample

(Figure 3A). Despite this, our current TEF1 α reference database allowed for the recovery of Sordariomycetes ($0.43 - (14.34 \pm 16.50) - 75.56\%$), Dothideomycetes ($0.07 - (11.26 \pm 20.02) - 80.26\%$), Eurotiomycetes ($0.28 - (21.54 \pm 40.16) - 81.76\%$), Exobasidiomycetes ($0.53 - (1.70 \pm 1.66) - 2.87\%$), Taphrinomycetes ($0.21 - (0.57 \pm 0.45) - 1.08\%$), and Leotiomycetes ($0.10 - (0.11 \pm 0.004) - 0.11\%$). Alpha diversity assessments for the TEF1 α merged dataset again exhibited minor but significant effects of storage time when the collection batch was treated as a random effect (Estimated richness decrease after storage of 0.36 ± 0.02 , $p < 0.001$; Shannon diversity increase after storage of 0.02 ± 0.0002 , $p < 0.001$) (Supplementary Tables 5, 6). Batch-wise comparisons in the TEF1 α dataset showed greater richness for the traps collected in NSW2 (91.33 ± 16.16 , $p < 0.05$), SA (120.53 ± 28.10 , $p < 0.05$), TAS (119 ± 13.54), VIC1 (107.84 ± 19.79), VIC3 (136.55 ± 20.65), WA1 (93.72 ± 13.99 , $p = 0.005$), WA2 (95.71 ± 18.78 , $p < 0.05$), and WA3 (144.24 ± 28.07 , $p = 0.001$) (Supplementary Table 5). While Shannon diversity estimates in the TEF1 α dataset appear to be lower for traps collected from NSW1, QLD1, QLD2, VIC2, and VIC4 (Figure 3B), batch-wise comparisons of diversity were inconsistent possibly due to the targeted and filtered nature of this dataset (data not shown; Supplementary Table 6).

3.3. Diagnostics of *Ophiostomatales* species

Ophiostomatales reads were recovered from 23 of the traps surveyed (51%), with 18 traps having *Ophiostomatales*-positive detections for both barcodes. The proportion of ITS1 reads assigned to the target group ranged from 0.005 to 8.89% (mean = 2.23%, SD = 2.32%) in 22 traps, while the proportion recovered for the TEF1 α barcode ranged from 0.11–66.84% (mean = 11.99%, SD = 15.44%) in 19 traps. Following downstream processing and phylogenetic analysis, sixteen unique ITS1 ASVs and 22 unique TEF1 α ASVs were recovered for the *Ophiostomatales*. Thirteen *Ophiostomatales* taxa, detected by 12 and 18 of the ITS1 ASVs and TEF1 α ASVs, respectively, have previously been recorded and are known from culture in Australia (Table 3 and Supplementary Figures 1, 2). These included a *Ceratocystiopsis* sp. (Taxon 1; Trollip et al. (2021)), *Graphilbum ipis-grandicollis*, *Gra.* cf. *rectangulosporium* [Taxon 4; Trollip et al. (2021)], *Leptographium huntii*, *L. radiaticola*, *Masuyamyces pallidulus*, *Ophiostoma angusticollis*, *Ophiostoma fasciatum*, *O. ips*, *Ophiostoma tsotsi*, *Raffaelea deltoideospora*, *Sporothrix stenoceras*, and *S. pseudoabietina* (Nkuekam et al., 2011; Trollip et al., 2021; Farr and Rossmann, 2022). The remaining eight unique ASVs (four for ITS1 and four for TEF1 α) correspond to taxa that have not been recorded in Australia. Here, we treated each novel ASV as an independent detection until more evidence can be gathered to link independent barcodes to a single taxon (i.e., a fungal culture). The eight novel eDNA sequences detected correspond to two previously described taxa [*Dryadomyces sulphureus* (= *Raffaelea sulphurea*) and the invalidly named *Hawksworthiomyces sequentia* ENAS (de Beer et al., 2016b)], five potentially novel or undescribed taxa (*Ceratocystiopsis* sp. AUS2-ITS1, *Ceratocystiopsis* sp. AUS3-TEF1 α , *Hawksworthiomyces* sp. AUS1-TEF1 α , *Sporothrix* sp. AUS1-ITS1, *Sporothrix* sp. AUS2- TEF1 α) and one taxon that cannot be

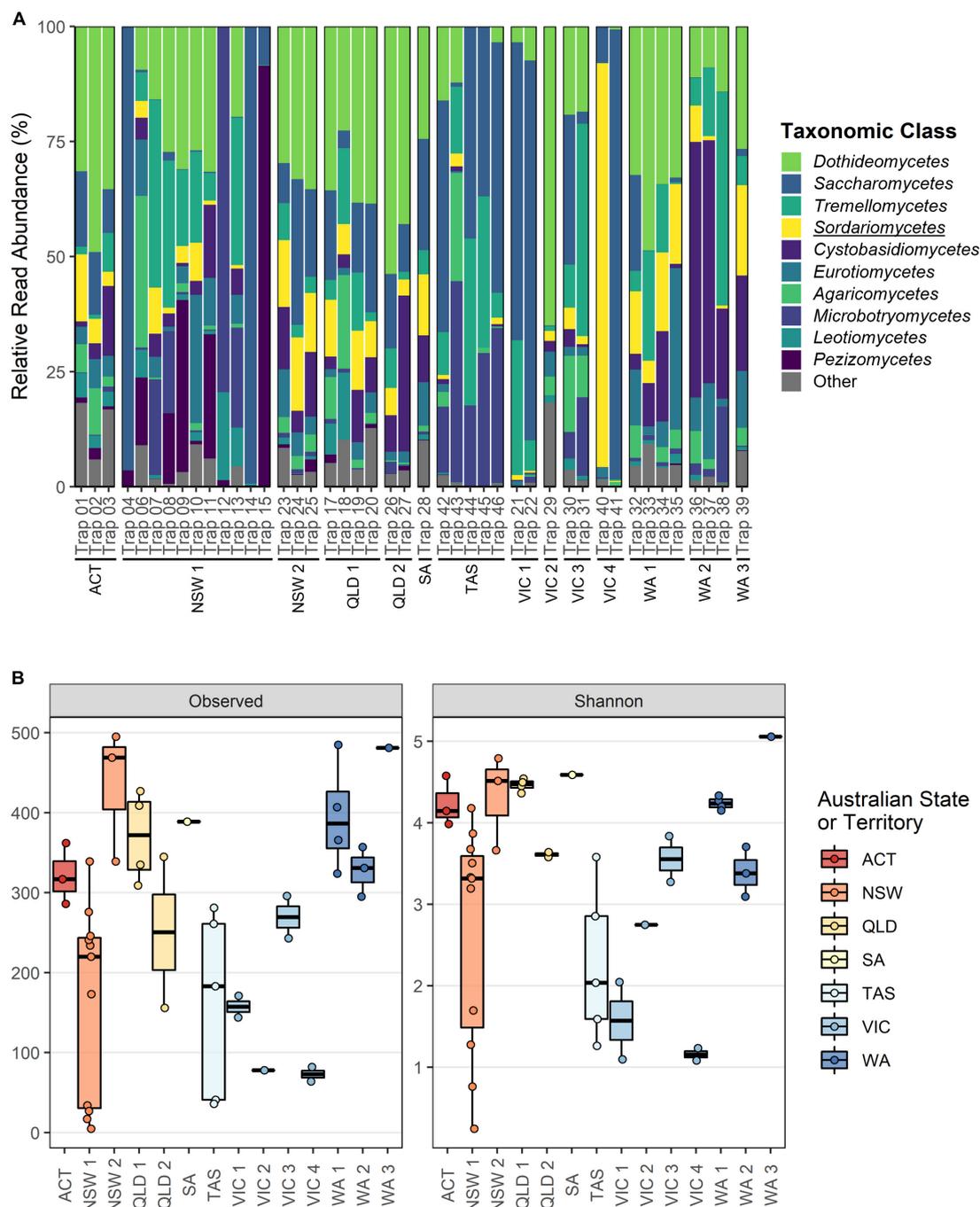


FIGURE 2

Overview of fungal class composition and alpha diversity metrics for the ITS1 dataset separated according to collection batch. (A) Relative read abundance (RRA) of the top 10 taxonomic classes. All taxa outside of the top 10 and any unclassified ASVs have been grouped under "Other." The class wherein the targeted *Ophiostomatales* reside is underlined. (B) Observed richness and Shannon diversity metrics with batches colored according to Australian State or Territory.

distinguished using the ITS sequence alone (*Grossmannia* sp. AUS1-ITS1).

The most common *Ophiostomatales* taxa detected included *Ceratocystopsis* sp. 1 (Taxon 1), *O. ips*, *S. pseudoabietina*, *Gra. ipis-grandicollis*, and *Gra. cf. rectangulosporium*, which were detected across all 6 mainland states and territories sampled (Table 3). For these, the ITS barcode detected the *Ceratocystopsis* sp. (Taxon 1) in 19 sites, *Gra. ipis-grandicollis* in 13 sites, and *S. pseudoabietina*

in 12 sites across all regions, while *O. ips* was not detected in Victoria and *Gra. cf. rectangulosporium* was not detected by this barcode. The TEF1 α barcode detected *O. ips* in 15 sites and *Gra. cf. rectangulosporium* in 10 sites across all states, with the *Ceratocystopsis* sp. (Taxon 1) (12 sites) not recovered in the ACT, and *Gra. ipis-grandicollis* and *S. pseudoabietina* both found in only four sites across three of the states (Table 3). The remaining detections were less common, where only *L. huntii*, *M. pallidulus*,

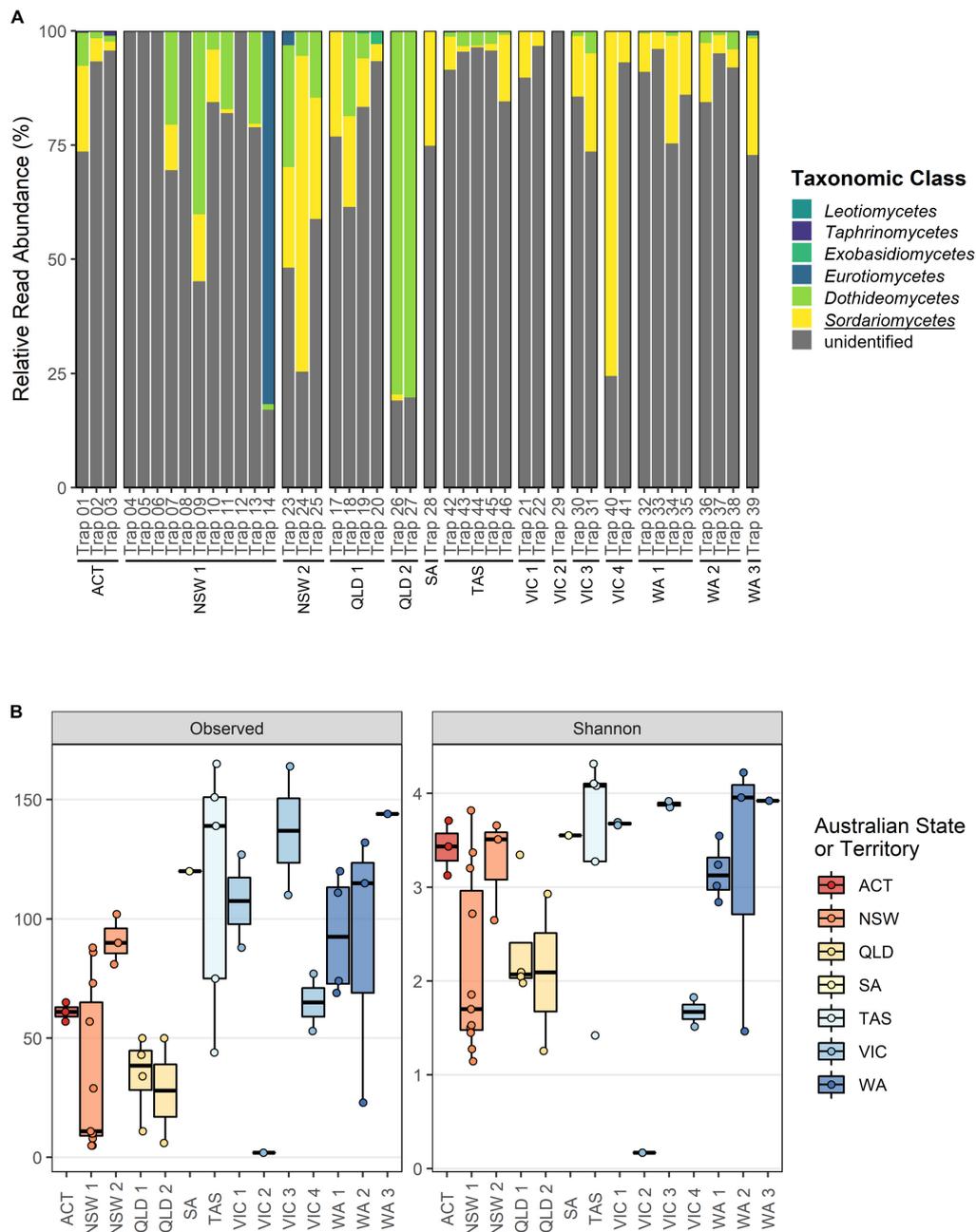


FIGURE 3 Overview of fungal class composition and alpha diversity metrics for the TEF1 α dataset separated according to collection batch. **(A)** Relative read abundance (RRA) at taxonomic class rank. All ASVs have been agglomerated at the class rank, with all unclassified reads grouped as “unidentified.” The class wherein the targeted *Ophiostomatales* reside is underlined. **(B)** Observed richness and Shannon diversity metrics with batches colored according to Australian State or Territory.

O. angusticollis, *O. fasciatum*, *R. deltoideo spora*, *Ceratocystiopsis* sp. AUS2-ITS1 and *Ceratocystiopsis* sp. AUS3-TEF1 α were observed from more than a single site. No *Ophiostomatales* ASVs were detected in the Tasmanian traps collected during this study.

4. Discussion

The modular, multi-barcode amplicon sequencing approach implemented in this study demonstrates the utility of high

throughput sequencing for diagnostics of fungal phytopathogens caught within insect surveillance traps. These results are consistent with previous studies in the rapidly expanding field of metabarcoding for biosurveillance. Notably, by using a validated pipeline for improved diagnostics of the *Ophiostomatales* (Trollip et al., 2022), we were able to accurately characterize the *Ophiostomatales* sequences captured in forestry surveillance traps from across Australia. This dataset represents the largest geographic survey targeting the *Ophiostomatales* in Australia to date, revealing novel eDNA sequences while establishing a

TABLE 3 Detections of *Ophiostomatales* taxa from surveillance traps collected across mainland Australia.

Species/Taxon	ACT (n = 3)		NSW (n = 15)		QLD (n = 6)		SA (n = 1)		VIC (n = 7)		WA (n = 8)		Number of states (total sites)	
	ITS	TEF	ITS	TEF	ITS	TEF	ITS	TEF	ITS	TEF	ITS	TEF	ITS	TEF
Previously recorded taxa														
<i>Ceratocystiopsis</i> sp. (Taxon 1) ^a	+		+	++	+	++	+	+	+	+	+	+	6	5
	(3)		(3)	(3)	(6)	(3)	(1)	(1)	(3)	(3)	(3)	(2)	(19)	(12)
<i>Graphilbum ipis-grandicollis</i> Trollip, Dinh & Edwards	+		+	++	+	+	+		+	+	+		6	3
	(1)		(3)	(1)	(6)	(2)	(1)		(1)	(1)	(1)		(13)	(4)
<i>Gra.</i> cf. <i>rectangulosporium</i> (Taxon 4) ^a		+		+		+		++		++		+	-	6
		(1)		(3)		(2)		(1)		(2)		(1)		(10)
<i>Leptographium huntii</i> (Robinson-Jeffrey)	+						+	+			+	+	3	2
	(2)						(1)	(1)			(1)	(1)	(4)	(2)
<i>L. radiaticola</i> (JJ Kim, Seifert & GH Kim)							+	+					1	1
							(1)	(1)					(1)	(1)
<i>Masuyamyces pallidulus</i> (Linnakoski, de Beer & Wingfield)								+		+		+	-	3
								(1)		(1)		(1)		(3)
<i>Ophiostoma angusticollis</i> (Wright & Griffin)			+								++	+	2	1
			(1)								(2)	(1)	(3)	(1)
<i>O. fasciatum</i> (Olchow & Reid)				++				+		+			-	3
				(2)				(1)		(1)				(4)
<i>O. ips</i> (Rumbold)	+	+	+	++	+	+	+	+	+	++	+	+	5	6
	(3)	(2)	(2)	(3)	(6)	(4)	(1)	(1)	(3)	(2)	(2)	(2)	(14)	(15)
<i>O. tsotsi</i> Grobbelaar, De Beer & Wingfield					+								1	-
					(2)								(2)	
<i>Raffaelea deltoideospora</i> (Olchow & Reid)			+	+	+							+	2	2
			(3)	(3)	(3)							(1)	(6)	(4)
<i>Sporothrix stenoceras</i> (Robak)											+	+	1	1
											(1)	(1)	(1)	(1)
<i>S. pseudoabietina</i> Wang, Lu & Zhang	+		+	++	+		+		+	++	+	+	6	3
	(1)		(3)	(1)	(4)		(1)		(1)	(2)	(2)	(1)	(12)	(4)
Novel eDNA detections^b														
<i>Ceratocystiopsis</i> sp. AUS2-ITS1							+		+		+		3	-
							(1)		(1)		(1)		(3)	
<i>Ceratocystiopsis</i> sp. AUS3-TEF1 α								+		+			-	2
								(1)		(1)				(2)
<i>Dryadomyces sulphureus</i> (Batra) -TEF1 α						+							-	1
						(1)								(1)
<i>Grosmannia</i> sp. AUS1-ITS1					+								1	-
					(1)								(1)	

(Continued)

TABLE 3 (Continued)

Species/Taxon	ACT (n = 3)		NSW (n = 15)		QLD (n = 6)		SA (n = 1)		VIC (n = 7)		WA (n = 8)			
	ITS	TEF	ITS	TEF	ITS	TEF	ITS	TEF	ITS	TEF	ITS	TEF	TEF	
<i>Hawksworthiomyces sequentia</i> ENAS De Beer, Duong & Wingfield – ITS1											+		1 (1)	-
<i>Hawksworthiomyces</i> sp. AUS1-TEF1α												+		1 (1)
<i>Sporothrix</i> sp. AUS1-ITS1											+		1 (1)	-
<i>Sporothrix</i> sp. AUS2- TEF1α													+	1 (1)
Total number of taxa	5	2	6	7	5	8	4	4	8	10	11	11	11	
Total number of ASVs	5	2	6	12	7	6	4	4	10	13	11	11		

ASV detections for each taxon are represented by +, with the number of symbols corresponding to the number of sequence variants detected. The value in parentheses represents the number of sites at which a taxon was detected within each state.

^aAs described in Trollip et al. (2021).

^bNovel eDNA detections were assigned a code if no known taxon could be assigned. AUS# = Australian taxon number. ITS1/TEF1α = barcode the taxon was detected with.

baseline for the target group within Australian *Pinus* plantations. Additionally, our results highlight aspects of the protocol that require further investigation and optimization to ensure sensitivity and reproducibility when implementing high-throughput sequencing approaches within post-border surveillance systems.

A primary objective of this study was to investigate how sample processing might impact on the recovery of fungal DNA from insect traps during surveillance activities. Considering that the traps are currently used to monitor insect pests, our methodology focused on speed, scalability, and compatibility for targeting phytopathogens, and consequently incorporated non-destructive approaches when working with the insect specimen subsamples. Previous studies have established that semiochemical-baited insect traps are effective for fungal surveillance, whether sampling the preservative fluids alone (Tremblay et al., 2018, 2019; Berube et al., 2022) or through destructive sampling of the insect-specimens (Malacrino et al., 2017; Miller et al., 2019; Rassati et al., 2019). To our knowledge, however, the impact on recovering targeted fungi from insect trap subsamples (i.e., extracting from preservative fluids vs. insect specimens) is yet to be assessed. In this study we were able to consistently recover sequenceable amplicons from the preservative fluids, whereas the results of the non-destructive extractions were less reliable despite using a universally recommended DNA extraction kit (Lear et al., 2018). This may be explained by a combination of different factors, including variabilities in substrate composition (Lindahl et al., 2013), differences in insect biomass and tissue types impacting on spore recovery (Aguayo et al., 2018; Martoni et al., 2022), as well as degradation (Weigand et al., 2021; Tedersoo et al., 2022). For instance, the Tumut region (NSW1) recorded 200–300 mm of rainfall for October 2020 (Bureau of Meteorology, 2020) which could have reduced the efficacy of insect specimen preservation through dilution (Martoni et al., 2021; Weigand et al., 2021).

Due to the inability of recovering sequencable amplicons from the insect subsamples, we were unable to address our initial objective of comparing the different subsampling approaches. Our results do, however, substantiate the utility of preservative fluids for forest surveillance activities. This substrate provides a consistent and stable source from which to sample when targeting fungi and other microbes, allowing for standardization of an approach that extends beyond only screening for insect-vectored pathogens (Berube et al., 2022). While both the anti-freeze and glycol mixtures used in the current study showed no apparent difference in their efficacy for long-term storage (>300 days at 4°C), further investigation is needed to test for within sample differences that might exist for these substrates. Optimization and standardization of sampling materials and protocols is important for ensuring no significant bias is introduced by different collectors or across states, especially if HTS is to be implemented routinely within post-border surveillance activities (Nilsson et al., 2019; Martoni et al., 2022). Additionally, future research efforts should continue to optimize the non-destructive approach investigated in this study as this would improve our understanding of fungal community differences when extracting from preservative fluids vs. insect specimens for forestry surveillance. This is particularly important for ensuring that sampling approaches are compatible for both entomologists and pathologists, while remaining appropriate for the biology of any chosen targets. For example, bark and ambrosia beetles are known to have complex interactions with fungi, in some

cases carrying them in specialized structures known as mycangia which can be positioned internally or externally (Six, 2012). It is currently unknown whether the non-destructive or preservative fluid extractions will obtain DNA of all fungal associates of these beetles, including those that reside in internal mycangia. Ensuring that the extraction methods appropriately “capture” all fungal associates within the trap is critical for the analytical sensitivity of these diagnostic tools. In the longer term, using methodologies that retain voucher specimens of captured insects is especially important for regulatory agencies looking to use high-throughput sequencing approaches (Martins et al., 2019; Batovska et al., 2021), while continued harmonization of sampling methods and laboratory protocols will unlock opportunities to monitor and compare communities from a more holistic viewpoint (Jactel et al., 2020).

In this study, targeted surveillance for the *Ophiostomatales* resulted in the detection of 13 taxa already known to be present in Australia, along with eight novel eDNA sequences from taxa not previously recorded. Naturally, the high-throughput approach used in this study uncovered greater diversity than observed in recent surveys across south-eastern Australia, where several first records were found through traditional diagnostics approaches (Trollip et al., 2021). While this is not an unexpected result, because the surveillance conducted in this study covers all major plantation forest regions of Australia and is expected to catch a diverse assemblage of beetles, it does highlight the efficiency and scale with which HTS approaches can be implemented. *Ceratocystiopsis* sp. (Taxon 1), *Gra. ipis-grandicollis*, *Gra. cf. rectangulosporium* (Taxon 4), *O. ips*, and *S. pseudoabietina* continue to represent the most common species recorded in Australia to date and are repeatedly found with their vector *Ips grandicollis* (Trollip et al., 2021, 2022). Our results confirm that the distribution of these fungi extends across all major *Pinus* growing regions on Australia's mainland where *I. grandicollis* is known to occur. The less frequent detections of the *Pinus*-associated *L. huntii*, *L. radiaticola*, *M. pallidulus*, *O. angusticollis*, *O. fasciatum*, and *R. deltoideospora* are likely explained by the fact they have only previously been isolated from beetle galleries, other less common bark beetles or blue-stained wood chips (Trollip et al., 2021). The last of the previously recorded species included *O. tsotsi*, detected from two biosecurity surveillance traps in Brisbane, Queensland, and *S. stenoceras*, recorded from WA's Albany region. *O. tsotsi* was first reported in Australia in 2011 where it was isolated from several *Eucalyptus* spp. (Nkuekam et al., 2011), while *S. stenoceras*, a ubiquitous fungus, has been recorded from bark beetles and soils globally (de Beer et al., 2016; Farr and Rossmann, 2022). Noticeably, no *Ophiostomatales* taxa were observed from Tasmania, where no bark or ambrosia beetles were captured in the traps sampled for this study.

The eight novel *Ophiostomatales* eDNA sequences reported here should not be surprising considering that most were recorded from regions of Australia where the baseline *Ophiostomatales* diversity remains understudied. For example, the *Ceratocystiopsis* sp. AUS2-ITS1 and *Ceratocystiopsis* sp. AUS3-TEF1 α sequences detected in SA, VIC, and WA, indicates the presence of a lineage closely related to *Ceratocystiopsis* sp. (Taxon 1). Both sequences provide further evidence of a link between the initial morphological records by Stone and Simpson in the late 1980s and an undescribed taxon from Canada (Stone and Simpson, 1987, 1990; Plattner et al., 2009; Trollip et al., 2021). Also, the detections

of *Dryadomyces sulphureus*-TEF1 α and *Grosmannia* sp. AUS1-ITS1 in Queensland are likely explained by the unexplored beetle-fungus associations that are present along the north-eastern coast where beetle assemblages show high species diversity (Stone et al., 2010), which would likely be reflected by their fungal associates. All novel *Ophiostomatales* eDNA sequences not previously recorded from Australia were considered in a biosecurity context, following guidelines in the Emergency Plant Pest Response Deed (Plant Health Australia [PHA], 2022a), and reported only as an indication of presence until a fungal culture, or taxonomic equivalent, can confirm status. Furthermore, novel ASV sequences should continue to be treated independently, even in cases where both the ITS1 and TEF1 α barcodes support the presence of the same taxon or lineage until a culture can properly link them. Future studies should look to address the growing need for systematic surveys of fungus-beetle associations across Australia, particularly focusing on the *Ophiostomatales* and their insect vectors, if we are to better interpret occurrence patterns or associations that may be detected through environmental sequence data.

While the dual-barcode approach used in this study was primarily focused on the *Ophiostomatales*, it did allow for some general observations of the overall fungal communities caught within the forest surveillance trap fluids. This was mainly achieved by the ITS1 barcode, which benefits from extensive publicly available sequence data, where the observed fungal communities were largely consistent within each collection batch. The only exceptions were a handful of traps from NSW1 and VIC4 (Figure 2A). The 10 most abundant fungal classes recorded by ITS1 in this study were consistent with previous research efforts that focused on bark beetle associates and are all known to dominate forest ecosystems (Miller et al., 2019). Notably, the TEF1 α barcode reported comparable proportions of the Sordariomycetes to ITS1 and demonstrates scope for characterizing diverse fungal communities with this alternative barcode. This is particularly promising considering that, at present, the TEF1 α barcoding region described by Stielow et al. (2015) lacks a comparable database to the rRNA region (Meyer et al., 2019), and the primer pair used in this study was predominantly developed on *Ophiostomatales* reference sequences (Trollip et al., 2022). The TEF1 α primer pair shows great potential beyond our current target group as future research looks to amplify multiple barcodes and alternative regions, independent of the rRNA (Cobo-Díaz et al., 2019; Hulcr et al., 2020). Nevertheless, in this study we chose to limit our species-level reporting to the pre-defined target, the *Ophiostomatales*, on which our current assay was developed and validated (Trollip et al., 2022). Any non-target or incidental detections outside of the scope of our experiment should be treated cautiously (Darling et al., 2020), especially while HTS-based surveillance transitions from general biomonitoring to high-throughput diagnostics. It is imperative that future research efforts continue to strive toward standardized and validated workflows which ensure sensitivity, reliability and confidence for end-users who are required to make informed decisions when using such data (Darling et al., 2020; Lebas et al., 2022; Massart et al., 2022). Multi-barcode amplicon sequencing, specifically using modular workflows (Swift et al., 2018; Piper et al., 2019; Trollip et al., 2022), offers adaptability for high-throughput forest diagnostics. These assays can be tailored for target-, or industry-specific objectives, and by incorporating validated diagnostic barcodes for species of concern, surveillance

programs (such as Australia's NFSP) could begin to effectively screen against long priority pest lists, rapidly characterize entire disease complexes within a single assay, or survey insect pests and fungal pathogens from within the same surveillance trap.

Data availability statement

The full bioinformatics workflow and required R packages are available at https://alexpiper.github.io/iMapPESTS/local_metabarcoding.html. Representative ASVs of the *Ophiostomatales* taxa reported in this study are available for the ITS1 (OQ221141-OQ221156) and TEF1 α (OQ241187-OQ241208) barcodes. Given that the data generated in this study originated from post-border biosecurity surveillance activities, all raw sequencing data supporting the conclusions of the article will be made available by the authors upon reasonable request.

Author contributions

CT, AC, BR, and JE conceptualized the study. CT, AC, and DS coordinated sample collection. CT performed laboratory work, bioinformatics and statistical analyses with supervision and guidance from AP, JK, FM, QD, and RM. CT wrote the draft manuscript with review and editing from AC, AP, JK, FM, BR, and JE. All authors read and agreed the final version of the manuscript.

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Conflict of interest

DS is currently employed by ArborCarbon.

The remaining authors declare that the research was conducted in the absence of any commercial or financial relationships that could be construed as a potential conflict of interest.

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Supplementary material

The Supplementary Material for this article can be found online at: <https://www.frontiersin.org/articles/10.3389/ffgc.2023.1149755/full#supplementary-material>

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