DNA metabarcoding of insects and allies: an evaluation of primers and pipelines

G.-J. Brandon-Mong1,2, H.-M. Gan3,4, K.-W. Sing1,2, P.-S. Lee1,2, P.-E. Lim5 and J.-J. Wilson1,2*

1Museum of Zoology, Institute of Biological Sciences, Faculty of Science, University of Malaya, 50603 Kuala Lumpur, Malaysia: 2Ecology and Biodiversity Program, Institute of Biological Sciences, Faculty of Science, University of Malaya, 50603 Kuala Lumpur, Malaysia: 3School of Science, Monash University Malaysia, Jalan Lagoon Selatan, Bandar Sunway, 47500 Petaling Jaya, Selangor, Malaysia: 4Monash University Malaysia Genomics Facility, Monash University Malaysia, Jalan Lagoon Selatan, Bandar Sunway 47500 Petaling Jaya, Selangor, Malaysia: 5Institute of Ocean and Earth Sciences (IOES), University of Malaya, 50603 Kuala Lumpur, Malaysia

Abstract

Metabarcoding, the coupling of DNA-based species identification and high-throughput sequencing, offers enormous promise for arthropod biodiversity studies but factors such as cost, speed and ease-of-use of bioinformatic pipelines, crucial for making the leap from demonstration studies to a real-world application, have not yet been adequately addressed. Here, four published and one newly designed primer sets were tested across a diverse set of 80 arthropod species, representing 11 orders, to establish optimal protocols for Illumina-based metabarcoding of tropical Malaise trap samples. Two primer sets which showed the highest amplification success with individual specimen polymerase chain reaction (PCR, 98%) were used for bulk PCR and Illumina MiSeq sequencing. The sequencing outputs were subjected to both manual and simple metagenomics quality control and filtering pipelines. We obtained acceptable detection rates after bulk PCR and high-throughput sequencing (80–90% of input species) but analyses were complicated by putative heteroplasmic sequences and contamination. The manual pipeline produced similar or better outputs to the simple metagenomics pipeline (1.4 compared with 0.5 expected:unexpected Operational Taxonomic Units). Our study suggests that metabarcoding is slowly becoming as cheap, fast and easy as conventional DNA barcoding, and that Malaise trap metabarcoding may soon fulfill its potential, providing a thermometer for biodiversity.

Keywords: Arthropoda, biodiversity, COI, high-throughput sequencing, Illumina MiSeq, Malaise trap

(Accepted 19 July 2015; First published online 7 September 2015)

Introduction

Much of our knowledge of biodiversity patterns and changes comes from the data based on mammals, birds and vascular plants (e.g., Gillison et al., 2013). Yet these taxa represent only a fraction of biodiversity; the major component of terrestrial biodiversity comprises insects (Mora et al., 2011). A recent meta-analysis of biodiversity studies revealed the dearth of information about most of the world’s tropical biota (Gillison et al., 2013), highlighting the fact that in order to decipher biodiversity patterns and change the major component can no longer be ignored. The absence of data on insects in biodiversity surveys, with the exception of small
groups of charismatic taxa such as butterflies, dragonflies and dung beetles (e.g., Korasaki et al., 2013; Hart et al., 2014; Zografou et al., 2014), reflects the taxonomic challenges associated with the huge diversity of this group of relatively small-sized organisms (Floyd et al., 2009). Obtaining insect samples is not an obstacle to collecting this data as many efficient sampling techniques have been developed (e.g., Russo et al., 2011, and in particular Malaise traps) but the investment required to sort and classify these samples is prohibitive. Fortunately, modern technology is addressing this impediment. First, conventional (single specimen) DNA barcoding, the use of short cytochrome c oxidase 1 mtDNA (COI) sequences as species identification tags (Hebert et al., 2003), has been applied to rapidly accelerate biodiversity surveys in hyperdiverse insect groups (e.g., ants of Madagascar; Smith et al., 2005). Now, with next-generation-sequencing technologies allowing simultaneously sequencing of DNA fragments from multiple specimens in a bulk mixture of diverse taxa, termed metabarcoding (Yu et al., 2012), the impediment is being alleviated further.

Metabarcoding is simply the pairing of DNA-based species recognition with high-throughput (next-generation) DNA sequencing (HTS) (Ji et al., 2013). Consequently, metabarcoding, like conventional DNA barcoding, relies on ‘universal’ polymerase chain reaction (PCR) primers that can amplify a fragment of a standard DNA region from diverse taxa (Ji et al., 2013). Due to the limitations in the size of DNA fragments sequenced by HTS platforms (see Shokralla et al., 2014), metabarcoding has typically been restricted to targeting short fragments of the COI DNA barcode (e.g., Hajibabaei et al., 2011; Zeale et al., 2011). Prior to the coining of the word ‘metabarcoding’, the idea of ‘mini-barcodes’ had been investigated in the context of degraded DNA samples (Hajibabaei et al., 2006). Hajibabaei et al. (2006) concluded that 135 bp fragments of COI can distinguish most species, but the location of the fragment within the full-length DNA barcode (~658 bp) is important. After further exploration of primer binding sites and species resolution offered by different fragments within the COI DNA barcode region, Meusnier et al. (2008) designed and advocated the use of Uni-MinibarF1 and Uni-MinibarR1 primers (amplifying a 130 bp fragment; see fig. 1) as a universal (eukaryote) primer set for the amplification of mini-barcodes (Meusnier et al., 2008). Zeale et al. (2011) designed and tested primers (ZBJ-ArtF1c and ZBJ-ArtR2c; see fig. 1) which amplify a 160 bp fragment from the 5’ end of COI for application in the study of arthropod prey in bat guano. These primers have since been used extensively in metabarcoding-type studies of diets (e.g., Bohmann et al., 2011; Razgour et al., 2011; Vesterinen et al., 2013; Burgar et al., 2014; Hope et al., 2014; Piñol et al., 2014a). Other primers were designed by Leray et al. (2013) for the analysis of the (metazoan) diets of fish collected at coral reefs, targeting ~330 bp fragments of DNA barcode suitable for ampiclon 454 pyrosequencing.

Besides diet studies, metabarcoding has been applied to environmental monitoring (Hajibabaei et al., 2011). Hajibabaei et al. (2011) collected aquatic insect samples in southern Ontario, Canada, for a test of metabarcoding, targeting a 130 bp fragment of COI (LepF1 primer paired with a newly designed reverse primer – EPT-long-univR; see fig. 1). Metabarcoding of bulk Malaise trap samples took off with Yu et al. (2012) with a ‘biodiversity soup’ study. This study employed primers typically used for DNA barcoding of insects (Folmer et al., 1994; also see Wilson, 2012) for ampiclon 454 pyrosequencing producing sequenced fragments (~400 bp) which when assembled together cover the full-length DNA barcode (~658 bp). Liu et al. (2013) used the same samples (from Yu et al., 2012) to develop a new bioinformatics pipeline ‘SOAPBarcode’ utilizing Illumina (HiSeq 2000) shotgun sequencing of the amplicons; in brief, 150 bp sections of the amplicon are sequenced and then assembled together to form the full-length DNA barcode. The use of metabarcoding as a source of data for conservation policy-making was validated by Ji et al. (2013) who compared metabarcoding datasets against standard biodiversity datasets in Malaysia (metabarcoded Malaise dataset versus birds, dung beetles, ants), China (light trap collected moths, both metabarcoded and morphologically identified) and England (metabarcoded whole pitfall-trap dataset versus ants, spiders, carabid beetles). Like Yu et al. (2012), Ji et al. (2013) used degenerate Folmer primers for ampiclon 454 pyrosequencing, Yang et al. (2014) also followed the protocols of Yu et al. (2012) to test the metabarcoding approach on soil and leaf-litter samples for rapid environmental monitoring in terrestrial ecosystems.

The reliance on ‘universal’ primers and associated biases has been of concern to the early practitioners of metabarcoding. Hajibabaei et al. (2011) reported a taxonomic bias for their LepF1 and EPT-long-univR primer set (fig. 1) as well as biases due to varying abundances of species in bulk samples (Hajibabaei et al., 2011). Yu et al. (2012) reported limitations of the classic ‘Folmer’ barcoding primers, particularly in regard to amplification of hymenopterans. In vitro PCR analyses by Clarke et al. (2014) with five primer sets targeting COI and 16S rDNA suggested a primer bias of COI markers towards lepidopteran and dipteran species with certain orders failing to amplify. In response to these simulated and empirical observations of primer biases in metabarcoding, Zhou et al. (2013) developed a new PCR-free Illumina pipeline for DNA-based

![Fig. 1. Relative positions of primers on the COI barcode region.](https://www.cambridge.org/core/)

Downloaded from https://www.cambridge.org/core. IP address: 54.70.40.11, on 28 Apr 2019 at 11:40:01, subject to the Cambridge Core terms of use, available at https://www.cambridge.org/core/terms. https://doi.org/10.1017/S0007485315000681
biodiversity assessment in bulk samples. Although the PCR-free Illumina pipeline (Zhou et al., 2013) enabled the successful identification of 97% of (73) species in a pooled sample, the produced large amounts (99.47%) of redundant data i.e., sequences not (presently) useful for taxonomic identification purposes, despite a mitochondrial enrichment step. Tang et al. (2014) followed this work omitting mitochondrial enrichment with similar results. Another approach to limit primer biases in metabarcoding has been to use multiplex PCR (multiple primers) prior to amplicon sequencing (Gibson et al., 2014). In order to maximize taxon detection, Gibson et al. (2014) used 11 unique PCR primer sets which all targeted the same 310 bp fragment of the standard COI DNA barcode (see fig. 1).

Despite the advancements in PCR-free and multiplex PCR pipelines, metabarcoding using universal primers for bulk PCR amplification still remains the most cost-effective and time-efficient protocol. PCR-free approaches generate a huge volume of redundant (un-utilized i.e., non-barcode) sequences. Only ~0.53% of raw sequences were mitochondrial sequences (Tang et al., 2014), even after mitochondrial enrichment (Zhou et al., 2013). In addition, shotgun, PCR-free approaches could miss the COI barcode target due to insufficient sequencing (Tang et al., 2014), especially as bulk samples are pooled for cost-efficiency. The payoffs from multiplex PCR deserve a more systematic evaluation – 11 primer sets recovered 91% of the known species in a pooled sample (Gibson et al., 2014), whereas a single primer set has been reported as successfully amplifying 91% of tested taxa (Leray et al., 2013). Such comparisons are likely to be idiosyncratic, but undoubtedly depend on the ‘universality’ of the single primer sets being compared; and it is worth to note that Gibson et al. (2014) did not make a comparison with the standard DNA barcoding primers. In addition, for cost-effective metabarcoding, the Illumina sequencing platform may be preferred due to its low sequencing cost compared with Roche 454 platforms (Yu et al., 2012; Liu et al., 2013; Yang et al., 2014). The Roche 454 platform has an estimated processing cost of US $240–415 per metabarcode sample, whereas the processing cost using an Illumina platform is estimated as half this value (Yu et al., 2012; Liu et al., 2013; Yang et al., 2014). An Illumina Miseq v2 can produce up to 15 million reads per run while Roche 454 FLX Titanium can generate 1 million reads per run (Glenn, 2014; also see Liu et al., 2013; Shokralla et al., 2014). Furthermore, Roche is considered a ‘Zombie platform’ (Glenn, 2014). The objectives of this study were: (1) to test and compare universal primer sets on a diverse set of arthropod orders to establish the optimal primer set for metabarcoding a tropical Malaise trap sample, while also considering the short read-length requirements of the Illumina platform; and (2) to test and compare the ease-of-use and reliability of the outputs from two bioinformatic pipelines (metagenomic and manual) to establish the optimal quality control and filtering pipeline for ‘real-world’ application of arthropod metabarcoding.

**Materials and methods**

*Sample collection, selection and DNA extraction*

A Malaise trap was set at Rimba Ilmu Botanic Garden, University of Malaya, Kuala Lumpur, Malaysia between 7 and 13 June 2014. From the bulk sample collected, 80 morphologically distinct specimens were selected as a test dataset, with the aim of maximizing taxonomic diversity. The specimens were pinned and oven-dried for 24 h. Based on examination of morphological characters (Triplehorn & Johnson, 2005) the test dataset included species from the orders: Lepidoptera, Hymenoptera, Araneae, Blattodea, Coleoptera, Orthoptera, Odonata, Diptera, Hemiptera, Collembola and Mantodea (table 1). Genomic DNA was extracted from the whole bodies of smaller specimens and two legs of larger specimens using a NucleoSpin Tissue kit (Macherey-Nagel, Germany), following the manufacturer’s instructions. A NanoDrop spectrophotometer (NanoDrop 2000c UV-Vis Spectrophotometer, Thermo Scientific) was used for DNA purity and concentration assessment.

**Primer selection and testing: individual specimen**

Four primer sets were retrieved from the metabarcoding literature: (i) ZBJ-ArtF1c/ZBJ-ArtR2c (Zeale et al., 2011), (ii) Uni-MinibarF1/Uni-MinibarR1 (Meusnier et al., 2008), (iii) mlCOlintF/HCO2198 (Leray et al., 2013), (iv) LCO1490/mlColintR (Leray et al., 2013) (table 1). In addition, a new reverse primer (MLepF1_Rev) was designed for use with the standard barcoding primer LepF1 (Hebert et al., 2004). In our previous studies we have found high amplification success with the standard barcoding primer MLepF1 (Smith et al., 2008; also see Wilson, 2012) and noted its binding site around
200 bp from LepF1 (fig. 1). Consequently, we used the program Primer3 Plus (Rozen & Skalsky, 2000) and a set of diverse high-quality insect COI sequences from another study (Wong et al., 2015) to select a 22 bp region slightly downstream of MLepF1 with appropriate structural and physical properties for primer binding. We included two degenerate bases (Ws) to create a reverse version of MLepF1 (named MLepF1-Rev). In preliminary testing with individual specimens, we found very low amplification success with ZBJ-ArtF1c/ZBJ-ArtR2c, therefore we proceeded with LCO1490 (the standard COI barcoding primer) and ZBJ-ArtR2c as an alternative combination.

PCR amplification was performed in a total volume of 25 µl with 0.25 µl of each forward and reverse primer (10 µM), 12.5 µl of Taq98® Hot Start 2X Master Mix (Lucigen, USA), 10 µl ddH2O and 2 µl of genomic DNA. For each primer set, we followed the thermocycling programs recommended by previous studies: for LCO1490/ZBJ-ArtR2c a touch-down program (40 cycles) with annealing temperatures 61–53°C was followed (Zeale et al., 2011); for Uni-MinibarF1/Uni-MinibarR1 a ‘touch-up’ program (40 cycles) with annealing temperatures of 46 and 53°C was followed (Meusnier et al., 2008); for mlColintF/HCO2198 and LCO1490/mlColintR a touch-down program (41 cycles) with annealing temperatures between 62 and 46°C was followed (Leray et al., 2013); for LepF1/MLepF1-Rev we followed ‘COI Fast’ (40 cycles) (Wilson, 2012). Success of PCR amplifications was checked on 2% agarose gels. A clear band of expected length (refer to fig. 1) indicated amplification success, whereas the absence of a band was recorded as PCR amplification failure.

**Primer selection and testing: bulk PCR and Illumina sequencing**

Based on the results from the individual specimen tests (see above), the two primer sets with the highest amplification success were selected and modified to include Illumina sequencing adapters with multiplex identifiers (following Bartram et al., 2011; table 2). The 80 test DNA extracts were pooled (1–15 µl of DNA extract from each specimen depending on the measured DNA concentration) and used for bulk PCR using the two modified primer sets. Initially PCR amplification was performed in a total volume of 25 µl with 0.25 µl of each forward and reverse primer (10 µM), 12.5 µl of Taq98® Hot Start 2X Master Mix (Lucigen, USA), 10 µl ddH2O and 2 µl of pooled genomic DNA. For each primer set, we followed the thermocycling programs from above. Success of PCR amplifications was checked on 2% agarose gels. Amplicons were gel extracted and purified using a NucleoSpin Gel and PCR Clean-up kit (Macherey-Nagel, Germany), following the manufacturer’s instructions. The libraries were quantified using KAPA library quantification kit (KAPA Biosystems, South Africa), normalized, pooled and sequenced on a MiSeq Desktop Sequencer (Illumina, USA) constituting approximately 0.5% of a MiSeq V2 500 cycle kit. Paired-end sequencing was performed at the Monash University Malaysia Genomics Facility.

Subsequently a second Illumina MiSeq run was conducted with amplicons produced by PCR in a total volume of 25 µl with 2.5 µl of each forward and reverse primer (10 µM), 2.0 µl of dNTPs, 0.25 µl of Accura™ High-Fidelity Polymerase (Lucigen, USA), 12.5 µl of Accura™ 2X HF buffer (Lucigen, USA), 3.25 µl of ddH2O and 2 µl of mixed genomic DNA. The thermocycling profile was modified to minimize chimera formation (fewer cycles with longer extension times) during PCR: for [V3]mlColintF/[MID96]HCO2198, 95°C for 2 min;
25 cycles of 95°C for 15 s; 51°C for 30 s; 72°C for 3 min and a final extension of 72°C for 10 min; for [V3]LepF1/[MID95] MLepF1-Rev, 95°C for 2 min; 25 cycles of 95°C for 15 s; 45°C for 30 s; 72°C for 3 min and a final extension of 72°C for 10 min. Five independent PCR products for each primer set were pooled prior to gel extraction. The next steps followed as above with each sample comprising approximately 2.75% of the sequencing run.

**Quality control and filtering pipelines**

Sequencing reads were demultiplexed and adapter-trimmed onboard the MiSeq using the MiSeq Reporter software. This resulted in a ‘raw’ output of two paired-end FASTQ files for each primer set (fig. 2). We followed two pipelines (fig. 2) for quality control and filtering of the paired-end reads: (a) a simplified metagenomics pipeline (by HMG)
incorporating FASTX (Hannon Lab, 2014), PEAR (Zhang et al., 2014), UPARSE (Edgar, 2013) and CD-HIT-EST (Fu et al., 2012); (b) we screened and filtered the reads 'manually' using CodonCode Aligner (CodonCode Corp.) and BioEdit (Hall, 1999) (by GJBM & JJW).

The 80 individual DNA extractions were also used for PCR amplification with Folmer primers (following standard methods; Wilson, 2012) and the products, or alternatively PCR products generated during the individual specimen primer tests (see above), were sequenced by a local company (MyTACG Bioscience). These Sanger sequences are available on BOLD (Ratnasingham & Hebert, 2007) in the public project MBPT. The assembled and dereplicated Illumina reads were ‘BLASTed’ (Altschul et al., 1990) against the Sanger sequences to give an estimate of the species surviving the bulk PCR and Illumina sequencing (i.e., the detection rate based on hits with an e-value of <1e\(^{-100}\)). Additionally, we built neighbor-joining (NJ) trees (in MEGA 6; Tamura et al., 2013) combining the filtered metabarc ode Operational Taxonomic Units (OTU) and Sanger sequences. The single representative of Odonata did not generate a Sanger sequence, but was traced among the OTU by BLAST searches against GenBank.

**Results**

**Primer testing: individual specimen**

The primer sets miCOlntF/HCO2198 and LepF1/ MLePf1-Rev showed the highest amplification success (both 98%), followed by Uni-MinibarF1/Uni-MinibarR1 (63%), LCO1490/ZBJ-ArtR2c (20%) and LCO1490/mlCOlntR (5%) (table 2, fig. 3). Consequently miCOlntF/HCO2198 and LepF1/MLePf1-Rev were used for further evaluation.

**Primer testing: bulk PCR and Illumina sequencing**

FASTQ files related to this study are available in the NCBI short read archive under accession SRR1848965. The first sequencing run produced 106,070 paired-end reads for the LepF1/MLePf1-Rev primer set (28 Mb) and 133,806
paired-end reads for the mlCOIintF/HCO2198 primer set (35M b). The detection rate of input species was 60% for the LepF1/MLepF1-Rev primer set and 64% for the mlCOIintF/HCO2198 primer set (fig. 3). A second sequencing run at greater sequencing depth produced 685,208 paired-end reads for the LepF1/MLepF1-Rev primer set (167 Mb) and 725,930 paired-end reads for the mlCOIintF/HCO2198 primer set (136 Mb) (table 3). The detection rate of input species was 80% for the LepF1/MLepF1-Rev primer set and 90% for the mlCOIintF/HCO2198 primer set (fig. 3). There were notable differences in the abundance of reads produced for each species. LepF1/MLepF1-Rev produced double the number of expected lepidopteran (17 input species) reads (16% of reads) compared with mlCOIintF/HCO2198 (8% of reads) and significantly fewer expected hymenopteran (16 input species) reads (0.1% of reads compared with 0.8%). LepF1/MLepF1-Rev produced a large number of expected hemipteran (4 input species) reads (6% of reads) compared with mlCOIintF/HCO2198 (0.02% of reads) whereas the primers produced comparable number of expected dipteran (19 input species) reads (9% of reads) and expected coleopteran (14 input species) reads (1% of reads).

Quality control and filtering pipelines

After the simple metagenomics quality control and filtering pipeline, 171 OTU were retained for the LepF1/MLepF1-Rev primer set, 44 of which corresponded to the 80 input species (i.e., matched Sanger sequences so were ‘expected’ whereas the remaining 127 OTU did not match Sanger sequences so were ‘unexpected’) and 201 OTU were retained for the mlCOIintF/HCO2198 primer set, 64 of which corresponded to the 80 input species (table 3). After applying the manual quality control and filtering pipeline, 122 OTU were retained for the LepF1/MLepF1-Rev primer set, 67 of which corresponded to the 80 input species and 105 OTU were retained for the mlCOIintF/HCO2198 primer set, 62 of which corresponded to the 80 input species (table 3). The unexpected OTU included contaminants such as Wolbachia and bats for the LepF1/MLepF1-Rev primer set and fungi and mammals for the mlCOIintF/HCO2198 primer set, based on BLAST hits in GenBank (online Supplemental Figure).

Discussion

To move from demonstration technology to a practical, widely employed, biodiversity monitoring tool, Malaise trap metabarcoding, must be (i) easy to understand, (ii) easy to use, (iii) fast and cheap. Our metabarcoding approach using a single primer set targeting a short mini/metabarcode is very similar in essence to conventional DNA barcoding, which has already gained considerable acceptance among conservation practitioners and the general public (e.g., Bucklin et al., 2011; Fišer Pečník & Buzan, 2014; Kress et al., 2015). Our DNA extraction from fresh caught (Malaise trap) specimens and PCR with a single primer set (such as conducted in this study) can be completed in a basic molecular lab in a few hours, while an Illumina MiSeq v2 run takes ~39 h. HTS can be outsourced to commercial companies at reasonable (and dropping) cost (US$2.5 per Mb in Malaysia). Therefore, to obtain 100 Mb for a 1-week Malaise trap sample (~100–300 specimens) can cost around US$250, theoretically US$1–2.5 per specimen or less (DNA extraction and PCR would add approximately US$1 per specimen).

The commercial companies will also provide bioinformatics analysis of the submitted samples up to BLAST hit, however the company may not be familiar with the specific protocols or purpose of the study, so we would always recommend the end-user retains control of the quality control and filtering pipeline. In our view, it is unrealistic to expect the users of applied metabarcoding (e.g., conservation officers in government agencies or NGOs) to master a series of command line programs to analyze their metabarcodes. A specific step-by-step web interface (such as those available for phylogenetic analyses; Dereepr et al., 2008) would be a significant step in the development of metabarcoding as a practical tool. Alternatively, easy-to-use GUI DNA sequence editing software, such as the widely used CodonCode Aligner, can be used to filter moderately sized metabarcodes samples (e.g., weekly Malaise trap collections) and produces similar or better outputs to the ‘conventional’ pipelines adopted from bacterial metagenomics – 1.4 compared with 0.5 expected(Sanger matching)/unexpected (without Sanger matches). Consequently, quality control and filtering of metabarcodes has the potential to be straightforward with considerable room for user input as opposed to the ‘blackbox’ of more complex pipelines (especially those requiring advanced sequence assembly e.g., Liu et al., 2013). Several examples from traditional DNA barcoding studies illustrate the need for careful understanding and review of sequence data by the user (e.g., Wilson & Sing, 2013).

Despite the significant progress made in metabarcoding in recent years, several issues remain. Particularly important issues concern what is considered an acceptable detection rate (influenced both by sequencing depth and difficult to amplify taxonomic groups i.e., PCR bias) and species identification (incorporating species resolution, heteroplasmy and contamination). Further issues relating to species delimitation methods and the completeness of DNA barcode reference libraries for the identification of OTU are also critical (e.g., Wilson et al., 2011).

The detection rate for 80 input species, which is slightly less than that found in a weekly Malaise trap sample in Malaysia (~100–300 species; Ji et al., 2013), was 80 and 60% at 167 and 35 Mb of sequencing output, respectively, for the LepF1/MLepF1-Rev and 90 and 64% at 136 and 53 Mb of sequencing output, respectively, for the mlCOIintF/HCO2198 primer set.

Table 3. Comparison of quality control and filtering pipelines applied to Illumina MiSeq metabarcodes.

<table>
<thead>
<tr>
<th>Primer set</th>
<th>Paired-end reads</th>
<th>METAGENOMIC (PEAR/CD-HIT) PIPELINE OTU</th>
<th>MANUAL (CODONCODE/BIOEDIT) PIPELINE OTU</th>
</tr>
</thead>
<tbody>
<tr>
<td></td>
<td></td>
<td>Expected (Sanger sequence match)</td>
<td>Unexpected</td>
</tr>
<tr>
<td>LepF1/MLepF1-Rev</td>
<td>685,208</td>
<td>44</td>
<td>127</td>
</tr>
<tr>
<td>mlCOIintF/HCO2198</td>
<td>725,930</td>
<td>64</td>
<td>137</td>
</tr>
<tr>
<td></td>
<td></td>
<td>Expected (Sanger sequence match)</td>
<td>Unexpected</td>
</tr>
<tr>
<td></td>
<td>67</td>
<td>55</td>
<td></td>
</tr>
</tbody>
</table>


This is less than detection at >97% in the PCR-free Illumina pipeline of Zhou et al. (2013) and Tang et al. (2014). However, the detection rate is comparable with that reported for bulk amplification with Folmer primers – 81% in the ‘biodiversity soup’ pipeline (Folmer primers and 454 sequencing) of Yu et al. (2012) and 84.9% using Illumina shotgun sequencing of the ‘biodiversity soup’ amplicons (Liu et al., 2013). Considering the size of the sequencing output: >1.1 Gb for Liu et al. (2013), and 13.2–31.7 Gb for PCR-free pipelines (Zhou et al., 2013; Tang et al., 2014), this is an unfair comparison and represents the trade-off between cost and detection.

Previous studies have reported a low detection rate for species of Hymenoptera (Yu et al., 2012; Zhou et al., 2013) and this was also seen in bulk PCR and Illumina sequencing in the current study (25% of hymenopteran species were detected during the first sequencing run). Interestingly, the amplification success rate for hymenopterans using single-specimen PCR was 100% for the best primer sets, LepF1/MLePFl-Rev and mCOI intF/HCO2198, showing that these primers can amplify hymenopteran COI but that there may be a bias during bulk PCR. It has been suggested that species with lower affinities with primer binding sites will yield lower level amplicons and fewer, if any, reads (Hajibabaei et al., 2011); but primer affinity is hard to predict (see Lee et al., 2015). To alleviate or at least minimize taxonomic bias in primer sets, lower PCR annealing temperatures (Ishii & Fukui, 2001; Sipos et al., 2007) and deeper sequencing (Hajibabaei et al., 2011) can be performed (88% of hymenopteran species were detected during the second deeper sequencing run), but may involve a trade-off in terms of non-specific binding and increased cost. Although a high number of hymenopteran species were detected by BLAST hit of the raw dereplicated reads, a large proportion of these reads were filtered out by the quality control pipelines, suggesting that although they were generated during sequencing, the hymenopteran reads were low abundance, low quality or characteristic of ‘error’ sequences. Hymenopteran reference sequences containing the poly-T region that is difficult to sequence cleanly and accurately (Zhou et al., 2013) may be beneficial in guiding quality control and filtering pipelines and help avoid discard of genuine hymenopteran OTU. The bias towards amplification of lepidopteran and dipteran sequences (as reported by Clarke et al., 2014) was seen in this study although appeared less severe for the primer set mCOI intF/HCO2198.

Deagle et al. (2014) argued that because COI has poorly conserved regions for primer design, the list of potential markers for metabarcoding has to be broadened. This is a controversial position as the opportunities for species identification based on the large, curated DNA barcode (COI) libraries (Ratnasingham & Hebert, 2007) would be forfeit (Yu et al., 2012). Quality control and filtering techniques (especially for chimeric sequences) often rely on properties of protein-coding sequences (Yu et al., 2012; Leray & Knowlton, 2015) and the advantages of mitochondrial protein-coding genes for species identification are well-established. Our study supports the previous work that has shown that reasonable detection rates and taxonomic coverage can be achieved with COI metabarcodes.

Although the majority of Sanger and OTU matches were within 2% p-distance, some Sanger sequences fell into clusters with closely grouping OTU on the NJ trees. There are a number of potential explanations for these sequence clusters including chimera formation, PCR or Illumina sequencing errors (Haas et al., 2011; Quail et al., 2012) and mitochondrial heteroplasm (Shokralla et al., 2014). Haas et al. (2011) reported that the number of PCR amplification cycles has a dominant effect on chimera formation. By increasing the PCR extension time, reducing the concentration of template DNA and the number of amplification cycles to the fewest number (approximately 20 cycles) still able to yield sufficient amplicons for sequencing, chimera formation can be alleviated or at least be minimized (Lahr & Katz, 2009; Haas et al., 2011; Stevens et al., 2013). Rapid changes in temperature might produce incomplete products which subsequently anneal to other DNA templates, creating chimeras, thus slowing the PCR ramp speed to 1°C s⁻¹ has been recommended as another modification to inhibit chimera formation (Stevens et al., 2013). Several potential chimeric sequences were observed and removed during our manual filtering steps, for example, when a group of reads showed close matches across a significant portion of the read and major divergences across another portion, or the presence of large gaps and frameshifts. Common methods for chimera filtering rely on analyzing the distribution and abundance of closely matching reads (Boyer et al., 2014) but we have found this approach will significantly reduce the detection rate when low abundance but ‘real’ (100% match to Sanger sequences) sequences are inadvertently filtered (also reported by Yu et al., 2012). Other common approaches for chimera removal rely on reference alignments (Edgar et al., 2011), but this is problematic for datasets consisting of many novel sequences, such as tropical Malaise samples. Other chimera detection methods based on signatures of recombination within a dataset (Martin et al., 2010) may be suitable additions to metabarcoding pipelines.

The observed divergences between Sanger sequences and OTU may partially be explained by heteroplasm (coexistence of multiple mitochondrial haplotypes in an individual) (Magna, Brown, 2010; Shokralla et al., 2014). COI heteroplasm has been documented in many insects species across several orders; Lepidoptera (12% of species examined; Shokralla et al., 2014), Orthoptera (2–24% of individuals observed; Moulton et al., 2010), Hymenoptera (13% of Hawaiian Hylaeus; Magnacca & Brown, 2010) and Diptera (17% of individuals of Drosophila melanogaster; Townsend & Rand, 2004). After our manual filtering pipeline we encountered putative heteroplasmic sequences in 58% of species, representing all the insect orders included, but especially among Diptera. Decreasing the OTU clustering threshold (e.g., to 95% as used by Yu et al., 2010) may mask the presence of heteroplasmic sequences, but could also merge ‘valid’ species showing low COI divergences. Another potential complication is nuclear-mitochondrial pseudogenes (numts). Numts have been highlighted as a potential source of ambiguity for DNA barcoding (Song et al., 2008), however, numts are generally easily spotted among amplified COI sequences by patterns in the amino acid translation (numts are noncoding), and probable numts would be removed when sequences were aligned and translated into amino acids.

Previous metabarcoding studies have all reported high levels of unexpected OTU (sequences) (13–35%; Hajibabaei et al., 2011; Liu et al., 2013). Our study was no exception with 22% (mCOI intF/HCO2198) to 39% (LepF1/MLePFl-Rev) of OTU surviving the metagenomics pipeline being probable contaminants or error sequences. However, this was reduced to <10% using the manual pipeline. Contamination may be caused by environmental DNA in the field or laboratory, if mis-priming (hybridization of sequencing primers during sequencing of libraries) occurs, if residual tissues (gut contents including mammal tissue such as detected in our study, eggs, minute specimens, etc.) or endosymbiotic bacteria (e,g,
Wolbachia are present in pooled samples, or be carried over from previous sequencing runs using the same MID (Hajibabaei et al., 2011; Kircher et al., 2011; Liu et al., 2013; Zhou et al., 2013; Nelson et al., 2014; Shokralla et al., 2014). In a real sample (when no corresponding Sanger sequences are available), it will be difficult to detect contamination and for particularly sensitive work, including that with legal ramifications and the detection of invasive species (Boykin et al., 2012), Yu et al. (2012) suggested that specialized protocols such as those followed in ancient DNA laboratories will be necessary. Another suggestion has been to split samples and conduct independent amplification and sequencing on both halves followed by statistical comparison with detected error reads (Lange, 2015). This of course would massively increase cost. Options for the elimination and detection of contamination without the need (and expense) of using ancient DNA protocols or additional sequencing runs include not reusing MID tags, sequencing extraction blank controls, and including technical replicates on the same HTS run.

Conclusion

Metabarcoding is a young field. Despite the successes of early empirical studies, a number of commentaries, simulations and thought-experiments have been published highlighting perceived shortcomings (e.g., Coissac et al., 2012; Cristescu, 2014; Deagle et al., 2014; Piñol et al., 2014b), in particular, many focus around marker choice. This is reminiscent of the wave of publications following Hebert et al. (2003) (e.g., Rubiniolf et al., 2006) which decreased abruptly after the build-up of empirical data clearly demonstrated the technical plausibility and utility of COI as a standard marker to recognize species boundaries (e.g., Smith et al., 2008b). The single primer sets used here, both targeting the COI barcode region, showed an acceptable detection rate for arthropod biodiversity analysis but there were complications due to putative heteroplasmic sequences and contamination. Considering the higher detection rate, and lower components of unexpected sequences after filtering, the miCOIfnF/HCO2198 primer set seems to best target around which to develop future protocols. To date, most metabarcoding-type studies have focused only on the efficiency of pipelines (detection rate), but factors such as cost, time and ease-of-use of the bioinformatics pipeline, that are crucial for making the leap from demonstration studies to a real-world application have not been realistically addressed. Our study suggests that DNA metabarcoding is slowly becoming as easy, fast and cheap as conventional DNA barcoding, and that Malaise trap metabarcoding may soon fulfill its potential, providing a thermometer for biodiversity.

Supplementary material

The supplementary material for this article can be found at http://www.journals.cambridge.org/BER

Acknowledgements

This study was supported by a High Impact Research Grant (Chancellery) from the University of Malaya to J. J. W. and P. E. L.; K. W. S. and P. S. L. are supported by postgraduate scholarships from the University of Malaya. Yee Yoke Pei (Ecology and Biodiversity Program, UM) and Narong Jaturas (Department of Parasitology, UM) assisted in sample collection and laboratory work. M. Sugumaran (coordinator of Rimba Ilmu Botanic Garden) provided permission to deploy the Malaise trap. H. M. G. is grateful to the Monash University Malaysia Tropical Medicine and Biology Platform for financial and infrastructure support.

References


DNA metabarcoding of insects and allies


Downloaded from https://www.cambridge.org/core, IP address: 54.70.40.11, on 28 Apr 2019 at 11:40:01, subject to the Cambridge Core terms of use, available at https://www.cambridge.org/core/terms. https://doi.org/10.1017/S1000748315000681