TESTING THE UTILITY OF THE CONSORTIUM FOR THE BARCODING OF LIFE'S TWO ‗AGREED UPON' PLANT DNA BARCODES, *matK* AND *rbcL*

By

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ABSTRACT

DNA barcoding is the use of short standardized regions of DNA to identify unknown specimens to species. Currently, the zoological community has agreed that *cytochrome oxidase I subunit 1* (*COI*), a mitochondrial gene region, will serve as the barcode region for all animal taxa. Due to oftentimes complicated evolutionary histories of plants, the plant barcoding community has had a much harder time agreeing on a gene region or regions that should be used to barcode the various land plant lineages. This is in large part due to poor reproductive barriers, which allows for chloroplast sharing between closely related species.

In the summer of 2009, the Consortium for the Barcoding of Life's Plant Working Group (CBOL, PWG) announced that portions of two coding chloroplast gene regions (cpDNA), *matK* and *rbcL*, would serve as the DNA barcode for all land plants. This recommendation was accompanied by CBOL's call for continued testing of these two regions, along with other potential gene regions that may prove to be more effective barcode regions, such as noncoding cpDNA regions like *trnH*-*psbA*.

Originally, this project was focused on the utility of three noncoding cpDNA regions (*trnH*-*psbA*, *trnL-trnL-trnF*, and *trnS*-*trnG-trnG*) at identifying species from the genus *Prunus* L. Upon the announcement by CBOL, additional sequence data was generated for *matK* and *rbcL* using the same *Prunus* taxa to determine how well these two regions would delimit species compared to the three noncoding cpDNA regions. In addition to this, sequence data for *matK* and *rbcL* were generated for 27 angiosperm taxa and compared to 34 previously tested noncoding chloroplast gene regions to determine their relative genetic variability. This broader study enabled me to directly compare the genetic variability of these two coding regions to that of noncoding regions.

My results for the broader study demonstrate that *matK* and *rbcL* contain less genetic variability than noncoding regions. Based on the number of potentially informative characters (PIC), *matK* was the 25th most variable region and *rbcL* was the 34th most variable region out of 36 regions tested. This low level of genetic variability in these two regions may make it difficult to identify closely related species. I recommend further study of the 34 previously tested noncoding cpDNA regions to determine their respective utility as plant DNA barcodes. For the direct species identification tests using *Prunus*, I found that no region alone or in combination

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was able to discriminate > 50% of species, and noncoding cpDNA regions typically outperformed the Consortium for the Barcode of Life's combination of *matK*+*rbcL*.

DEDICATION

This work is dedicated to my daughter, Ella, who has shown me what unconditional love and true courage are all about.

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CHAPTER I

TESTING THE UTILITY OF THE CONSORTIUM FOR THE BARCODING OF LIFE'S TWO 'AGREED UPON' PLANT DNA BARCODE REGIONS

Introduction

More than 250 years ago, Linnaeus began his quest to name Earth's living organisms using a hierarchical system that clusters organisms into smaller and smaller groups based on the presence or absence of certain morphological features. This system is still in use today and research to identify and name life on earth continues in earnest. Identification of many plant species can be quite challenging due to the imperfect nature of morphology, i.e. flower color or leaf shape can grade through a species range making it difficult to determine how many species are contained in a particular genus (see Shaw and Small, 2005 for example). While many vascular plant species (e.g., *Acer rubrum* L.), as well as, certain plant groups (e.g. ferns and gymnosperms) are relatively easily identifiable, others require that a suite of morphological characters are present to properly identify (e.g., Asteraceae Bercht. & J.Presl, Cyperaceae Jussieu, and Poaceae (R. Brown) Barnhart). For the majority of the year, these hard to identify plant species lack discriminating morphological characters, such as flowers or fruits, making them unidentifiable and increasing the chance that they will go unnoticed (or undocumented) during a typical field season. However, recent advances in technology have enabled biologists to more easily obtain DNA sequences, which could be used to identify species regardless of time of year.

First proposed by Hebert et al. (2003), DNA barcoding is the use of short DNA sequences (\leq 750 bp) for rapid species identification. The potential impacts of DNA barcoding are far reaching and could benefit a number of related disciplines including: ecology, conservation biology, and population genetics (Tautz et al., 2003; Blaxter et al., 2005; Valentini et al., 2009). Despite widespread criticism (Will and Rubinoff, 2004; Mitchell, 2008; Seberg and Petersen, 2009; Spooner et al., 2009), DNA barcoding efforts have garnered the attention, support, and funding from various organizations including the Alfred P. Sloan Foundation, which has committed \$150 million to the Barcode of Life endeavor (Alfred P. Sloan Foundation Website, http://www.sloan.org/program/7). This funding has aided in the creation of two consortia, the Barcoding of Life Database (BOLD) and the Consortium for Barcoding of Life (CBOL). The purpose of these two consortia is to 1) serve as a repository for DNA barcode sequence data, 2) assist in the decision process to choose gene regions that will serve as DNA barcodes a gene region or regions, (3) improve and standardize the analytical methods by which the gene regions can be compared to one another to ensure that the best region(s) are selected, and (4) development of a probabilistic search algorithm, so that researchers can compare sequences from unknown samples to known samples.

According to Kress and Erickson (2008), in order for a gene region to be considered as a potential DNA barcode it must meet the following three criteria: (1) contain significant specieslevel genetic variability and divergence, (2) possess conserved flanking sites for universal primer development for wide taxonomic application, and (3) have a relatively short sequence (\leq 750 bp) in order to facilitate current capabilities of DNA sequencing (this would also facilitate amplification and sequencing from older preserved material since DNA degrades over time).

Currently, zoologists have identified a portion of the mitochondrial *cytochrome oxidase subunit 1* gene (*COI*) to serve as their "universal" barcode (Hebert et al., 2004; Ward et al., 2005), with several other regions suggested for taxa that are more difficult to discriminate (e.g., tardigrades and nematodes; see Meyer and Paulay, 2005 for details). In plants, it is well understood that the evolutionary rate of the mitochondrial genome (mtDNA) tends to be highly conserved (Clare, 2008; Fazekas et al., 2008), thus this genome does not offer enough genetic variability to discriminate between closely related species. Plant mitochondrial genomes also tend to experience high rates of structural rearrangements and gene duplication events, making it difficult to design universal primers that will work across the various land plant lineages. Thus far, identifying a plant DNA barcode region that will work as well as *COI* in animals has eluded the plant barcoding community. Researchers have turned their attention to the nuclear and chloroplast genomes, both of which accumulate mutations faster than the mitochondrial genome does in plants.

The nuclear genome might seem to be the obvious choice since it accumulates mutations faster than either the chloroplast or the mitochondrial genome. Early on, the nuclear ribosomal internal transcribed spacer region (nrITS) was suggested for plant barcoding since it is the most widely sequenced region in plants (Kress et al., 2005) and there is already a framework in place to build a DNA barcode database, however, the nuclear genome has several functional limitations. Because it is biparentally inherited, individuals can be heterozygous at a particular locus. This could make it more difficult to capture genetic variation within a species since an individual can possess two different copies of a gene. Nuclear genes are also often present in multiple copies, so using a region from the nuclear genome could drive the costs of plant barcoding up because many more alleles would have to be sequenced in order to correctly

measure (and associate) variation within a single species. For these reasons, the plant barcoding community has focused on the chloroplast genome (cpDNA).

Like the mitochondrial genome, cpDNA is effectively haploid and non-recombinant. This makes cpDNA a better choice than the nuclear genome because there are no issues related to heterozygosity. cpDNA also accumulates mutations at least three times faster than the plant mitochondrial genome (Wolfe et al., 1987; Palmer et al., 2000), so there are more potentially informative characters to identify closely related species. However, since cpDNA accumulates mutations slower than *COI* does in animals researchers have had to test large portions of the chloroplast genome in order to find gene regions that contain high levels of interspecific variation to differentiate between closely related species. It has become widely accepted that no single chloroplast region contains enough genetic divergence to discriminate the \sim 300,000 species of land plants (Kress and Erickson, 2007; Taberlet et al., 2006; Fazekas et al., 2009), and that plant DNA barcoding is going have to be a multilocus approach. Several multilocus barcode (MBC) approaches have been proposed (Chase et al., 2005; Newmaster et al., 2006; Chase et al., 2007; Kress and Erickson, 2007; Fazekas et al., 2008), but none of these have been shown to have the level of success of *COI*. Fazekas et al. (2009) recently demonstrated that species discrimination only improved slightly when more than three loci were concatenated, which suggests that potential barcode regions need to contain high amounts of genetic variability between species since simply adding more sequence data does not mean improved species delimitation. While the plant barcoding community has narrowed its focus to the chloroplast genome, the debate on whether to use coding or noncoding gene regions continues (Kress et al., 2005; Devey et al., 2009; Ford et al., 2009; Fazekas et al., 2010).

There are advantages and disadvantages associated with using either coding or noncoding cpDNA regions. Noncoding regions are less evolutionary constrained (see Kelchner, 2000), thus mutations can accumulate faster with virtually no effect on phenotype. This could mean more informative nucleotide characters to distinguish between closely related species. Noncoding regions also contain insertion-deletion characters (indels), which have been shown to help resolve phylogenetic relationships at lower taxonomic levels (Gielly and Taberlet, 1994). On the other hand, those championing the use of coding regions argue that current sequencing technology, sequence alignment algorithms, and database search algorithms makes the use of noncoding regions too difficult for barcoding purposes (Ford et al., 2009). This is largely due to the presence of polynucleotide runs in noncoding regions, which can make obtaining quality sequences and aligning sequences more difficult than coding regions (Devey et al., 2009; Ford et al., 2009). However, Kress et al. (2005) argue that sequence alignment issues should not be considered in a barcoding scheme since technology will continue to improve and resolve these issues.

For those in the plant barcoding community advocating the use of noncoding regions, previous work by Shaw et al. (2005; 2007) provides a foundation. Shaw et al. (2005; 2007) tested the relative genetic variability of 34 noncoding cpDNA regions across ten phanerogam lineages. Based on Shaw et al. (2005), and their own empirical testing of nine noncoding regions, Kress et al. (2005) proposed using nrITS and the *trnH*-*psbA* intergenic spacer from the chloroplast genome. As noted above, nrITS currently presents many challenges for barcoding and has therefore been largely avoided by the plant barcoding community. However, *trnH-psbA* was found to be one of the most variable noncoding cpDNA regions in terms of percent variability (Shaw et al., 2005). This is an attractive barcode region because it has a relatively

short length (avg. aligned length 465 bp; to Shaw et al., 2005), which makes amplification and sequencing generally straightforward. *TrnH-psbA* does have functional limitations that complicate its use as a barcode region, such as poly- or mononucleotide repeats. These repeat regions result in $trnH-ps\bar{b}A$ being too short ($<$ 400 bp) in many plant lineages, thus reducing the number of variable nucleotide characters present, or too long in other plant lineages (> 1000 bp) making it more costly to sequence (Chase et al., 2007). Nucleotide repeats also reduce the clarity and reliability of DNA sequences, which can make it exceedingly difficult to successfully sequence the entire region regardless of size (Devey et al., 2009). For these reasons, many in the barcoding community have advocated the use of coding cpDNA regions.

Ford et al. (2009) note that coding regions are easily alignable and the amino acid code can be utilized to ensure that a nuclear pseudogene was not unintentionally sequenced, which they argue outweigh the use of noncoding regions. Using the tobacco chloroplast genome (*Nicotiana tabacum* L.), Ford et al. (2009) looked at 41 of the 81 coding regions to determine which regions would be most suitable for DNA barcoding. After further testing 12 loci on 98 samples, representing the major land plant lineages (mosses, liverworts, gymnosperms, and angiosperms), they determined that six of the twelve regions merited further evaluation based on amplification success and sequence variability, including *matK*, which has been touted early on in the search for a plant DNA barcode (Ford et al., 2009). However, none of the regions tested by Ford et al. (2009) were shown to be as variable as the noncoding regions tested by Shaw et al. (2005; 2007).

Since the publication of Kress et al. (2005), numerous articles have discussed the advantages and disadvantages of a number of other cpDNA regions. But none of these studies have unequivocally shown any chloroplast gene regions to be as effective for plant barcoding as

COI has for animals. Table 1 summarizes the plant DNA barcode studies since 2005 and shows the taxonomic groups used and gene regions tested. These previous studies can be broken down into two types, (1) those that have looked at a suite of coding and noncoding regions using a few species from a variety of genera and families (Fazekas et al., 2008; Ford et al., 2009; Hollingsworth et al., 2009) and (2) those more narrowly focused on testing how well different gene regions identify species within a single family or genus (Ziegenhagen et al., 2005; Edwards et al., 2008; Newmaster et al., 2008; Nitta, 2008; Seberg and Petersen, 2009; Spooner et al., 2009; Newmaster and Ragupathy, 2009). Sampling in many of the latter studies was also geographically restricted. Both types of studies are important for identifying and measuring the utility of potential plant DNA barcodes. Studies with a broad taxonomic scope are useful for determining which regions will amplify and sequence easily across the land plant lineages, while the studies with a narrow taxonomic scope are useful at determining the fine scale utility of each potential plant DNA barcode region at discriminating closely related species. In either type of study, it is possible that intra- and interspecific variation was not accurately captured for a particular species due to a lack of multiple samples per species (see Fazekas et al., 2008) or because sampling did not mirror the geographic distribution for a particular family, genus or species (see Newmaster et al., 2008). It is paramount that within, and between, species genetic variability is measured for a species prior to setting up an efficient barcoding system since genetic variability is backbone of DNA barcoding. Zhang et al. (2010) argue that most DNA barcoding projects have wildly under-sampled at the species level, thus have not measured the genetic variation thoroughly enough to correctly define and identify a species. They also point out that sampling is idiosyncratic, i.e., the number of samples needed per species is unique to that particular species (Zhang et al., 2010); couple this with that fact that most species in BOLD are

represented by fewer than 10 sequences (Hajibabaei et al., 2007) and the chances are high that many species searches might return a misleading species assignment for an unknown sample.

The PWG (2009) looked at seven candidate loci (three coding and four noncoding cpDNA regions) across 907 samples representing all the major land plant lineages to determine which loci amplified the best, produced the cleanest bidirectional reads, and discriminated the most species. They reduced their dataset to only samples that were successfully sequenced for all seven loci tested, so species identification tests were performed on 397 samples (all angiosperms) rather than the original 907 samples (PWG, 2009). It was formally announced in the *Proceedings of the National Academy of Sciences* (CBOL PWG, 2009) that the barcoding community had "agreed upon" the use of portions of two coding cpDNA regions, *matK* and *rbcL*, as the genetic markers to serve as the barcodes for all land plants. As the PWG noted (2009), *matK* is one of the fastest evolving coding regions (Hilu et al., 2003). However, developing universal primers for *matK* has been difficult and the PWG (2009) noted their own difficulties with generating *matK* sequence data for non-angiosperm taxa. On the other hand, *rbcL* is the best characterized cpDNA gene, but not the most variable region, and its inclusion had more to do with its ease of amplification and sequence recovery rather than its ability to discriminate species. The PWG (2009) did find that species resolution was better when *rbcL* was included in multilocus tests. Both regions have been tested in combination with other regions prior to the PWG's announcement and the results appear mixed (Kress and Erickson, 2007; Fazekas et al., 2008; Hollingsworth et al., 2009), with neither *matK* nor *rbcL*, alone or in combination, having been shown to be as successful as *COI* is in animal taxa.

It is well established that most chloroplast gene regions will identify an unknown sample to genus (see Newmaster et al., 2006; Kress and Erickson, 2007), but the goal of DNA barcoding

is species level identification. Large-scale studies with wide phylogenetic/taxonomic scope, such as the PWG (2009), are important in determining amplification and sequencing success of the various regions tested. These types of studies included only a few congeneric or conspecific samples, which may alleviate many of the issues facing plant DNA barcoding, including the potential overlap in genetic distances between closely related species due to hybridization or chloroplast sharing. According to Fazekas et al. (2009), plant barcoding is going to continue to be a challenge due to incomplete lineage sorting and high rates of chloroplast sharing between closely related species. Unlike animals, many plant groups display poor reproductive barriers, with many species are indeed capable of forming interspecific hybrids and in some cases intergeneric hybrids (Rieseberg, 1997). Since species boundaries tend to be porous in plants, defining what is a species can be challenging (Shaw and Small, 2004; 2005) and can add to the difficulty of using DNA barcodes for species identification or association. This lends to the idea that plants are just harder to barcode than animals (Fazekas et al., 2009).

Study Objective

The over arching objective of my research was to compare the utility of *matK* and *rbcL* as plant DNA barcodes to other noncoding cpDNA regions. Previous research looked at the utility of a number of gene regions in one of two manners 1) broad phylogenetic scope (many families or genera with few species) or 2) narrow phylogenetic scope (one family or genus with many species). There are currently no protocols for study design within the barcoding community, so I tested *matK* and *rbcL* in the above two manners to follow the study designs from previous DNA barcoding studies (Table 1).

Using the same samples and methods of Shaw et al. (2005; 2007), the first part of this research was broad in phylogenetic scope in order to measure amplification, sequencing, and

alignability success for *matK* and *rbcL* across angiosperm lineages. Within this framework, comparisons between closely related species were used to measure the relative genetic variability of *matK* and *rbcL* across seven angiosperm lineages. The relative genetic variability *Matk* and *rbcL data* were then compared to 34 previously tested noncoding cpDNA regions by Shaw et al. (2005; 2007) to get an idea of how well these two coding regions may identify closely related species.

The second part of my research was narrow in phylogenetic scope to better assess the ability of five cpDNA gene regions to identify species. Sequence data were generated for the following five gene regions: *matK* and *rbcL* (the ‗agreed upon' barcoding loci) along with *trnHpsbA*, *trnL*-*trnL*-*trnF,* and *trnS*-*trnG*-*trnG*, three noncoding regions that were promoted in earlier studies as potential barcoding regions (Kress et al., 2005; Taberlet et al., 2006; Nitta 2008). Figure 1 shows the location of these five gene regions within the chloroplast genome. Direct species discrimination was performed using 203 accessions of 54 species of *Prunus* L. (Rosaceae). *Prunus* is a large, complex genus with a global distribution and it is known to contain both closely and distantly related species (see part two for further discussion). Each region was tested alone and in all possible multilocus combinations to determine which region or regions delimited the most *Prunus* species.

Materials and Methods

Taxon Sampling

Comparing genetic variability of **matK** *and* **rbcL** *to 34 noncoding gene regions using seven angiosperm lineages—*This dataset contained 27 samples representing 27 species from seven lineages of angiosperms (magnoliids, monocots, caryophyllids, eurosids I and II, and euasterids I and II), based on APG II (2003). These are the same accessions that were previously used by

Shaw et al. (2005; 2007) to determine the relative utility of 34 noncoding chloroplast regions for low-level systematics, except for *Minuartia cumberlandensis* (B.E. Wofford and Krall) McNeill because all extracted DNA had been exhausted for previous studies. This sample was replaced by an accession that was shown in an unpublished study to contain no sequence divergence from the original accession (pers. comm. R. Small, University of Tennessee, Knoxville, TN, USA). A list of the taxa used and GenBank accession numbers are provided in Table 1 in Appendix I.

Prunus *species and outgroup selection for testing barcoding efficacy of* **matK** *and* **rbcL** *in comparison to three noncoding cpDNA regions*—*Prunus* L. is an excellent model taxon for DNA barcoding because it is a complex, hyperdiverse genus that includes 1) many widespread species (e.g., *P. serotina* Ehrh. and *P. arborea* (Blume) Kalkman, are found throughout eastern North America or Southeast Asia, respectively), 2) several narrow endemic species (e.g., *P. geniculata* Harper and *P. ramburii* Boiss., with the first restricted to the area around Orlando, FL, USA, and the latter Spain), 3) many closely related species (see Shaw and Small, 2004; 2005), and 4) distantly related species as recent molecular work shows that it also includes the previously recognized genera *Pygeum* Gaertn. and *Maddenia* Hooker and Thompson (Potter et al., 2007; Wen et al., 2008). Using *Prunus* s.l. will enable us to better gauge how well the five potential DNA barcodes will work at discriminating species in a large, taxonomically challenging group.

Prunus also includes the agriculturally important fruits like plums, apricots, peaches, cherries, and almonds. According to Boriss et al., (2009), the value of the above mentioned fruits to the U.S. economy in 2007 ranged from \$101.1 million for plums to more than \$2 billion for almonds. Based on data from the Food and Agricultural Organization (FAO) of the United Nations website (FAOSTAT TradeSTAT website, faostat.fao.org), the collective global

import/export value for fruits in this genus, exceeded \$13 billion in 2006. The genus also contains one commercially important medicinal species, *P. africana* (Hook. f.) Kalkm., which is used to treat benign prostatic hyperplasia (BHP) in men over 50 (Stewart, 2003). The estimated annual trade value of *P. africana* is around \$220 million and growing rapidly (Cunningham et al., 1997). Because of its global agricultural and medicinal value, the National Clonal Germplasm Repository (NCGR) at Davis, CA has been tasked with the responsibility of collecting, preserving, evaluating, and distributing the genetic resources to ensure that crop diversity in these species is available for future generations (NCGR at Davis website, http://www.ars.usda.gov/main/site_main.htm?modecode=53-06-20-00). NCGR currently houses over 1000 accessions of various *Prunus* species, which are preserved through vegetative propagation either by root cuttings or budding onto rootstocks (NCGR at Davis website).

Today, *Prunus* is thought to include ~ 150 species (pers. comm. Joey Shaw, The University of Tennessee at Chattanooga, Chattanooga, TN, USA) found mainly in the temperate regions of the Northern Hemisphere, with scattered distributions found throughout the tropical latitudes in southeast Asia, sub-Saharan Africa, and South America (Lee and Wen, 2001). According to Lee and Wen (2001), the generic delimitation of *Prunus* has been controversial starting with de Tournefort's work in which he used fruit morphology as the primary means for sorting the various taxa into six distinct genera. Since the publication of de Tournefort's work, the classification scheme for *Prunus* species has been amended at least ten times with as few as two genera recognized within *Prunus* s.l. and as many as seven (Lee and Wen, 2001). It was not until 1865 that Bentham and Hooker united the six genera of de Tournefort into *Prunus* s.l., which they further subdivided into seven sections (Lee and Wen, 2001).

Rheder's (1940) classification scheme, in which he included roughly 120 species or lesser taxa, is the most widely accepted treatment for the genus. Rheder (1940) subdivided the genus into five subgenera (*Prunus* L., *Amygdalus* L., *Cerasus* Pers., *Padus* (Moench) Koehne, and *Laurocerasus* Koehne.) and 12 sections. Krussman (1978) followed up on this work and included the same five subgenera, but went a step further and split section *Microcerasus* out of subgenus *Cerasus* Pers. creating a sixth subgenus, *Lithoceraus* Ingram*.* While *Prunus* taxonomy has been controversial, recent molecular work has largely supported the classification schemes of Rheder (1940) and Krussman (1978) (see Bortiri et al., 2001; 2002; and 2006; Shaw and Small 2004; 2005; Wen et al., 2008).

Our dataset contained 203 accessions representing 54 species of *Prunus* collected from around the world. All five subgenera recognized by Rheder (1940), *Prunus* s.s. L., *Amygdalus* L., *Cerasus* Mill., *Padus* Mill, and *Laurocerasus* M. Roem., were included along with four accessions from the previously recognized genus *Pygeum.* Of the 54 species, 48.1% were represented by two samples, 42.6% were represented by three to six samples, and 9.3% were represented by seven or more samples. *Prunus armeniaca* L. and *Prunus serotina* Ehrh. were sampled the heaviest (12 samples each). *Physocarpus opulifolius* (L.) Maxim. served as the outgroup taxon for analyses based on phylogenetic monophyly and was collected from a natural population in Tennessee. It was used as our outgroup taxon because none of the taxa within *Prunus* has unequivocally been shown to be sister to the rest of the genus (Mowrey and Werner, 1990; Bortiri et al., 2001; 2002; Lee and Wen, 2001).

Ingroup sampling of *Prunus* was from wild-collected populations (collected by J. Shaw, D. Potter, J. Wen, or S.-W. Chin) or leaf material sent to us by Clay Weeks from the National Clonal Germplasm Repository at Davis, California, U.S.A, with several exceptions. *Prunus*

maritima Marshall var. *gravesii* (Small) Watson, which has likely been extirpated from the wild, was obtained from the University of Connecticut, Storrs, Connecticut, USA. Samples of *P*. *subcordata* Benth., were sent to us by colleagues at Oregon State University, Corvallis, Oregon USA. Positive identification of wild-collected samples were made by experts in this genus including: D. Potter, J. Shaw, S-W Chin, or J. Wen. Wild-collected specimens were identified using the following works: Small (1933), Rehder (1940), Bailey and Bailey (1941), Fernald (1950), Blackburn (1952), Gleason (1952), Steyermark (1963), Radford et al. (1968), Correll and Johnston (1970), Duncan and Duncan (1988), Godfrey (1988), Wunderlin (1988), Gleason and Cronquist (1991), and Flora of North America (1993+, Draft Copy), Smith (1994), Wofford and Chester (2002), and Lingdi et al. (2003). A list of species used in this study, along with collector information and GenBank accession numbers, is provided in Appendix II, Table 1**.**

Marker Selection

Comparing genetic variability of **matK** *and* **rbcL** *to 34 noncoding gene regions using seven angiosperm lineages*—Sequence data for *matK* and *rbcL* were generated for 27 angiosperm species to determine how variable these two regions are. Results were subsequently compared to 34 noncoding regions, previously tested by Shaw et al (2005; 2007), to determine which regions contained the highest levels of genetic variability and the greatest potential to resolve relationships between closely related taxa. Laboratory and data analysis protocols are provided below.

Prunus *species and outgroup selection for testing the efficacy of* **matK** *and* **rbcL** *in comparison to three noncoding cpDNA regions*—The 203 samples used in this investigation were successfully amplified for the following regions: $trnG^{UUC}$ and $trnL^{UAA}$ introns, $trnS^{GCU}$

*trnG*UUC , *trnL*UAA -*trnF*GAA , *trnH*GUG -*psbA* intergenic spacers, and the portions of the *matK* and *rbcL* genes that were identified by PWG. *trnH*^{GUG}-psbA, *trnL*^{UAA}-*trnL*^{UAA}-*trnF*^{GAA}, and *trnS*^{GCU}*trnG*^{UUC}-*trnG*^{UUC}, were selected for study prior to the PWG announcement (2009), and are based on past DNA barcoding studies (Kress et al., 2005; Taberlet et al., 2006; Nitta, 2008), as well as, the work of Shaw et al. (2005; 2007).

*trnH***GUG -***psbA* **(***trnH***-***psbA***)**—While Shaw et al. (2005) ranked this region in the bottom third of noncoding cpDNA regions due to its short average length (465 bp, Shaw et al., 2005), they did note that it was the second most variable region based on percentage and that it amplified and sequenced easily across all phanerogam lineages. According to the PWG (2009), several members strongly advocated for its inclusion into any multilocus barcoding approach as it had been shown to contain high levels of genetic variability between species and it was found to be very informative in that study and previous ones (Kress et al., 2005; Kress et al., 2009). Several issues have come up regarding the inclusion of *trnH*-*psbA* as a barcode region including: sequence length variability and mononucleotide repeats. It has been shown that *trnH-psbA* ranges in size from \sim 400 bp to > 1000 bp (Shaw et al., 2005; Kress et al., 2005; PWG, 2009). Current sequencing technology is able to successfully read 750-800 bp of DNA, so any barcode region should be within this range to avoid issues with obtaining quality bidirectional sequence data. Also, *trnH-psbA* contains mononucleotide repeats, which can make PCR amplification and sequencing more difficult since these repeat regions tend to cause slip-strand mispairing (Devey et al., 2009). This disrupts normal PCR amplification by causing the *Taq* polymerase to slip and incorrectly amplify the target sequence (Devey et al., 2009) and can make it costlier to obtain quality DNA sequence data since automated sequencers end up stuttering and misreading the bases downstream of these repeat regions (Devey et al., 2009). While cost is not the only issue or factor in identifying a plant barcode, it is an important one to consider since any barcoding system needs to be affordable to be successful. We decided to include *trnH*-*psbA* in this study because it has been touted as a potential plant DNA barcode region early on (Kress et al., 2005) and it has been tested in 13/20 studies (see Table 1, Part 1).

*trnL***UAA -***trnL***UAA -***trnF***GAA (***trnLLF***)**—The *trnL* intron and the *trnL-F* intergenic spacer, alone or in combination, have been two of the more widely used regions for early phylogenetic studies (Shaw et al., 2005), and a search on GenBank (accessed 1/4/2010) yielded over 44,000 accessions. This existing framework may make it tempting to establish a barcoding database centered on this region. However, Shaw et al. (2005) showed that *trnLLF* is a relatively slowly evolving region and a poor choice for low level taxonomic studies. Taberlet et al. (2006) looked at the utility of using the *trnL* intron as a barcode region and found species discrimination was poor. We decided to include this region along with the flanking *trnL*-*trnF* intergenic spacer, which is easily amplified and sequenced together with the *trnL* intron, to further evaluate the utility of this region as a plant barcode because the intergenic spacer typically tends to be more variable than the *trnL* intron.

trnSGCU **-***trnG***UUC -***trnG***UUC (***trnSGG***)**—Shaw et al. (2005) showed this region to be among the top tier for low-level phylogenetic studies. Currently, only Nitta (2008) has tested this region in a barcoding context. His results suggest that this is a good potential barcode region since it had the lowest level of intraspecific variability and the highest level of interspecific variability of the three regions he tested (Nita, 2008). Low intraspecific variation and high interspecific variation are important components of a barcode region. Low levels of genetic variability present in a single species means fewer individuals will be needed to capture intraspecific genetic variation,

while a high interspecific variation means that it will be more likely that closely related species can be delimited. However, Nitta (2008) did not outright advocate the use of *trnSGG* due to difficulties associated with DNA amplification and aligned sequence length (> 1400 bp in Hymenophyllaceae), which is well over the proposed 750-800 bp limit for DNA barcodes. Like *trnH*-*psbA*, this region contains mono- and polynucleotide repeats, which can make obtaining quality sequence data difficult. This has required the development of internal primers to ensure quality bidirectional sequencing, which were used in this study for several problematic taxa. Despite these issues, this region was included because of its discriminatory power at low taxonomic levels (Shaw et al., 2005; Nitta, 2008), and because of the low degree of intraspecific variation shown by Nitta (2008).

PWG **matK and rbcL** *Barcode Regions*—Recently proposed by the PWG, the utility of *matK* has been of much debate (PWG, 2009). According to the PWG (2009), this region is among one of the most rapidly evolving protein coding regions in the chloroplast genome and it has consistently showed high levels of species discrimination in angiosperms. *MatK* was among the first regions suggested as potential plant DNA barcode region (Chase et al., 2007), and this was one of six loci that Ford et al. (2009) strongly suggested for further study due to its ease of amplification and overall high rate of variability. In the PWG (2009) study, only 66% of taxa were correctly identified using *matK* alone and species identification success only improved to 72% in multilocus tests.

Like *trnLLF* and *matK*, *rbcL* was utilized early on in phylogenetic studies and proved useful at the family level (Newmaster et al., 2006). The PWG (2009) noted that *rbcL* was not a highly variable site (alone it identified 61% of taxa), and Ford et al. (2009) noted that the inclusion of *rbcL* in any barcoding scheme had more to do with its historical use in deep-level

phylogenetics and not empirical DNA barcode testing. Its inclusion by the PWG (2009) had more to do with its performance in multilocus tests (72% of species were delimited when combined with *matK*)*. MatK* and *rbcL* were included in this study to directly compare the two proposed plant barcode regions to three noncoding regions at species identification.

Laboratory procedures

DNA was extracted from leaves using the DNeasy Plant Mini Kit (Qiagen, Valencia, California, USA). Samples were amplified by polymerase chain reaction (PCR) using the primer pairs listed in Table 2 for *matK*, *trnSGG*, *trnH-psabA*, *trnLLF*, and *rbcL*. PCR was performed using Eppendorf Mastercycler gradient or Mastercycler personal thermal cyclers in 50 μL volumes with the following reaction components: 2 μL template DNA (10–100 ng), 13 *ExTaq* buffer (PanVera/TaKaRa, Madison, Wisconsin, USA), 200 μmol/L each dNTP, 3.0 μmol/L MgCl2, 0.1 μmol/L each primer, and 1.25 units *ExTaq* (PanVera/TaKaRa). Reactions included bovine serum albumin at a final concentration of 0.2 mg/mL, which improved amplification of difficult templates.

All PCR protocols described below were preceded by template DNA denaturation at 80°C for 5 min except for *matK* and *rbcL*, which had an initial denaturation at 94°C for 1 min and 95°C for 4 min respectively. The PCR cycling conditions for *trnLLF* were: 30 cycles of denaturation at 94°C for 1 min, primer annealing at 50°C for 1 min, primer extension at 72°C for 2 min A final extension step consisted of 5 min at 72°C. The PCR cycling conditions for *trnHpsbA* were: 30 cycles of denaturing at 94°C for 30 s, annealing at 50°C for 30 s, and extension at 72°C for 1 min The PCR cycling conditions for *trnSGG* were by touchdown method: 15 cycles of denaturation at 96 \degree C for 1 min, primer annealing at 76 \degree C (20.4 \degree C/cycle) for 45 s, primer extension at 72°C for 2 min, and then 30 cycles of denaturation at 96°C for 1 min, primer

annealing at 69.5°C for 45 s, primer extension at 72°C for 2 min The PCR cycling conditions for *matK* were: 35 cycles of denaturation at 94 °C for 30 s, primer annealing at 52 °C for 20 s, primer extension at 72 $^{\circ}$ C for 50 s, and a final extension step for 5 min at 72 $^{\circ}$ C. The PCR cycling conditions for *rbcL* were: 35 cycles of denaturation at 94°C for 30 s, primer annealing at 55°C for 1 min, primer extension at 72° C for 1 min, and a final extension step for 10 min at 72° C. PCR products were checked on 1% agarose gels before being cleaned with ExoSAP-IT (USB, Cleveland, Ohio, USA). DNA sequencing was performed in house using the same primers used in amplification (Table 2) with the ABI Prism BigDye Terminator Cycle Sequencing Ready Reaction Kit, v. 2.0 or 3.1 (Perkin-Elmer/Applied Biosystems, Foster City, California, USA) and sent to the Molecular Biological Research Facility (Knoxville, TN, USA) for sequence reads on an ABI Prism 3100 automated sequencer except for *matK* and *rbcL*, which were sequenced commercially using Macrogen (Seoul, South Korea). Sequencher 4.7 (Gene Codes, 2007) was used to edit and assemble complementary DNA strands and check for agreement between them. In no cases, did the adjoined complementary strands disagree.

Data analysis

Due to the number of gene regions tested and the number of analyses performed an outline and explanation of electronic files created and used can be found in Appendix D. For all data sets, alignment of DNA sequences was done by eye in MacClade v. 4.06 (Maddison, 2001; Sinauer, Sunderland, Massachusetts, USA) or Mesquite v. 2.72 (Maddison and Maddison, 2009). Variable positions in the data matrices were double checked against the original chromatogram files to make sure that all base calls were correctly called. In all cases, alignment of potentially informative positions was unambiguous. Because indels have been shown to provide

approximately one-third of the potentially phylogenetically informative information in a cpDNA data set (Gielly and Taberlet 1994), they were coded using FastGap (Borchsenius, 2009).

In the *trnSGG* and the *trnH*-*psbA* datasets, 150 and 11 bases were omitted from analysis respectively, due to polynucleotide runs, since these maybe PCR artifact and not reflective of the phylogenetic history of the group. Also, in the *trnH*-*psbA* dataset, an additional 177 bases were omitted due to a large center portion of this intergenic spacer being unalignable. *Prunus* taxa used in a previous study by Shaw and Small (2004), were missing the 3′ *trnL* exon, but since exons are generally highly conserved regions, and likely would not add any information, resequencing of those taxa was not attempted.

Comparing genetic variability of **matK** *and* **rbcL** *to 34 noncoding gene regions using seven angiosperm lineages*—*MatK* and *rbcL* were compared directly to the previous regions tested by Shaw et al. (2005, 2007) to determine their relative genetic variability. *MatK* and *rbcL s*equence data were generated for three additional lineages used by Shaw et al. (2005) (gymnosperm, *Eupatorium*, *Solanum*). *MatK* sequences for two of the three gymnosperm samples did not produce quality sequence data, but since only data from the *Tortoise and the Hare II* were available for these three lineages (Shaw et al., 2005); we omitted them from analyses in order to maintain a parallel dataset with Shaw et al. (2007). The relative performance of each region was measured by counting the number of indels and substitutions between the two ingroup taxa, as well as between the ingroup taxa and a third species, which served as an outgroup taxon (see Shaw et al., 2005, 2007 for more detailed explanation of this methodology). The number of indels and substitutions were then summed to give a raw potentially informative character (PIC) value, which was then normalized for each region/lineage combination by dividing the number of PICs found within that region/lineage combination by the sum total of PICs found within a

given lineage. For example, in the magnoliid lineage the 21 PICs tallied for *rpl*16 were divided by the 1050 PICs found within that lineage across 34 noncoding regions plus *matK* and *rbcL*. By doing this, we reduced the influence of differing evolutionary rates or distances among the different taxa.

Assessment of species identification using **Prunus***—*Following CBOL guidelines, uncorrected pairwise genetic distances (UpD) were calculated using PAUP v. 4.0b (Swofford, 2002) for taxa with multiple accessions. Our data was made up of 203 samples representing 53 species or lesser taxa and three unknown samples, which were placed in the dataset to try and determine their respective identities. Additional matrices were created in Mesquite 2.72 for all possible multilocus combinations for a total of 31 datasets (including single locus). This was done to determine which multilocus combinations discriminated the most species. Each distance matrix was then imported into Microsoft[®] Excel and following PWG (2009) standards, species were considered successfully identified if the highest intraspecific distance was lower than any other interspecific distance (e.g. *Prunus africana* had an ingroup distance between 0 and 0.0017426 in the *rbcL* dataset, which means that all interspecific values must be > 0.0017426 in order for it to be considered positively identified).

A second set of analyses were performed on the same 31 datasets using Bayesian Analysis (BA) to determine the number of species or lesser taxa in the dataset that could be recovered as monophyletic (see Fazekas et al., 2008). Only species with multiple accessions were analyzed since single accession taxa could not be resolved as monophyletic. Due to the size of our dataset, the use of online servers was required to reduce the overall computation time of generating data. Bayesian analyses were performed using Parallel MrBayes @ BioHPC v. 3.1.2 (Ronquist and Huelsenbeck, 2003) (available at: http://cbsuapps.tc.cornell.edu/mrbayes.aspx) to

generate posterior probability distribution using Markov chain Monte Carlo (MCMC) methods. No a priori assumptions about tree topology were made. Models of DNA substitution were estimated using MultiPhyl Online v 1.0.6 (Keane et al., 2007; available at:

http://distributed.cs.nuim.ie/multiphyl.php); the General Time Reversible model (GTR, Tavaré, 1986) was used for the *rbcL*, *matK* and *trnLLF* datasets, while the Hasegawa, Kishino and Yano 1985 model (HKY85, Hasagawa et al., 1985) was selected for the *trnH-psbA* and *trnSGG* datasets. Four of the five gene regions tested had a gamma rate. Only *trnSGG* dataset had an I+G rate. MCMC process was set to run for 8 000 000 generations with four chains. This number of generations was chosen because it was the greatest number of generations that Parallel MrBayes would run before timing out on the largest datasets. We wanted to keep all parameters equal between datasets to prevent biased results towards smaller datasets, i.e. those with fewer total nucleotide characters, since these datasets may have been able to run for more generations, thus producing more resolved phylogenies compared to larger datasets. Burn-in was estimated visually by plotting log-likelihood values in Microsoft Excel to determine the number of generations that had run before likelihood values reached an asymptote. To calculate the posterior probability of each bipartition a 50% majority-rule consensus tree was constructed from the remaining trees using PAUP v. 4.0b (Swofford, 2002). Species were considered successfully identified if they formed a monophyletic clade with no other species present and had a posterior probability $> 50\%$.

Results

Amplification, Sequencing, and Alignability

For the seven angiosperm lineages, *matK* and *rbcL* were easily amplified by PCR. Amplification and sequencing of three gymnosperm taxa (*Cryptomeria japonica* (L.f.) Don.,

Taxodium distichum (Nutt.) Croom, and *Glyptostrobus pensilis* (Staunton) K. Koch) was attempted but quality *matK* sequence data could not be obtained for *Taxodium distichum* and *Glyptostrobus pensilis*; however, these three gymnosperm taxa and the 27 angiosperms samples were successfully amplified and sequenced for *rbcL*. The three gymnosperms were omitted from *rbcL* analyses to keep datasets parallel.

Within the *Prunus* dataset, the 203 samples and *Physocarpus opulifolius* were successfully amplified and sequenced for the five potential barcoding regions (*rbcL*, *matK*, *trnSGG*, *trnLLF*, *and trnH*-*psbA*). For *trnSGG*, about 33% of the samples required additional sequencing with internal primers to ensure that quality sequences were obtained. M*atK* and *rbcL* were the easiest regions to align because there were no poly A/T runs or hard to align indel regions. Alignment of the *trnLLF* dataset was also relatively easy and no data was considered unalignable. In the *trnSGG* and the *trnH*-*psbA* datasets, 150 and 11 bases were omitted from analysis respectively, due to polynucleotide runs. More importantly, in the *trnH*-*psbA* dataset an additional 177 bases (more than half of the raw sequence data) were omitted due to a large center portion being unalignable.

Comparing matK and rbcL to 34 noncoding gene regions using seven angiosperm lineages

The average aligned length of *matK* across the seven lineages was 847 bp, and the aligned length of this region ranged from 823 bp (monocots) to 862 bp (*Gratiola*). For *rbcL*, the average aligned length was 577 bp, and ranged from 573 bp (caryophyllids) to 589 bp (*Gratiola*).

MatK averaged 2.5 times more PICs than *rbcL* across the lineages (average PICs = 25.3; 10, respectively). In *matK*, the magnoliids had the fewest number of PICs (13) while the caryophyllids ranked the highest (45). The number of PICs in *rbcL* ranged from four (monocots)

to 23 (*Gratiola*). In *matK*, the normalized PIC value (which is the percentage that each region contributes to the total number of PICs within a lineage) ranged from 1.2 (magnoliids) to 3.3 (*Carphephorus*). In *rbcL*, the normalized PIC value ranged from 0.4 (monocots) to 1.5 (magnoliids). The average normalized PIC value across the seven lineages was skewed heavily in favor of *matK* (2.27; 0.88, *matK* and *rbcL*, respectively).

In order to determine which of the two gene regions were more variable, percent variability was calculated by dividing the number of parsimoniously informative characters by the aligned length (Table 3). Percent variability for *matK* ranged from 1.6 (magnoliids) to 5.3 (*Gratiola*), and for *rbcL*, the range was 0.7 (monocots) to 3.9 (*Gratiola*). *MatK* was on average more variable across the seven lineages than *rbcL* (average percent variability = 2.97; 1.72, respectively).

In Appendix B, Table 8, the number of substitutions, indels, and inversions can be found for *matK* and *rbcL* in each lineage (for the results of the 34 noncoding gene region/lineage combinations see Shaw et al., 2007). *MatK*, substitutions for ingroup sampling ranged from three in both the monocots and *Prunus* to 22 in *Gratiola*, while the number of substitutions between ingroup and outgroup samples ranged from eight (*Hibiscus*) to 35 (caryophyllids). In *matK*, indels for ingroup sampling ranged from zero in the magnoliids to five in *Gratiola*, and the number of indels between ingroup and outgroup taxa ranged from zero (magnoliids, monocots, and caryophyllids) to four (*Hibiscus*). With respect to *rbcL*, there were three lineages in which no substitutions were counted between ingroup samples (magnoliids, caryophyllids, and *Prunus*), while *Gratiola* had the highest number, nine*.* Between ingroup and outgroup samples in *rbcL* the number of substitutions ranged from three (monocots) to 16 (magnoliids). Indel characters in

rbcL were only counted in *Gratiola*, which had one between the ingroup taxa and one between the ingroup and outgroup taxa.

Figure 2 shows the average normalized PIC values across the gene region/lineage combinations, and in no lineage was *matK* or *rbcL* the most variable locus. *RbcL* was two to three times less variable than *matK* across all lineages, except in the magnoliids. Figure 3 shows the 34 previously tested regions along with the addition of *matK* and *rbcL*. The highest normalized PIC value was observed in *rpl32*-*trnL*, which had a normalized PIC value of 5.80 and averaged 63.4 PICs across the seven lineages tested. The *trnS*-*rps4* intergenic spacer was the least variable, with a normalized PIC value of 0.70 and an average PIC value of 7.7 across the seven lineages tested. A closer look at how much potential information *matK* and *rbcL* might yield to DNA barcoding (or low-level phylogenetic studies) shows that *matK* ranked $25th$ out of 36 regions tested, based on average normalized PIC value across the seven lineages (2.27); This is less than half as potentially informative as *rpl32*-*trnL*, and there are five gene regions with at least twice the genetic variability found in *matK. RbcL* is the third least variable region tested $(34th$ out of 36; avg. PIC value of 0.88 Fig. 3) and is only potentially better than *psbA*- $3'$ *trnK* (0.77) and *trnS*-*rps4* (0.7), which are both relatively short regions and are approximately half the size of *rbcL* (261 bp; 273 bp; 577 bp, respectively). There are 27 gene regions with more than twice the genetic variability found in *rbcL* (Fig. 4). What follows is a closer look at the performance of *matK* and *rbcL* compared to the other regions within each of the seven plant lineages (refer to Appendix 1, Table 2; Figs. 1 and 2 and Shaw et al., 2007).

*Magnoliids—*The raw values within this lineage ranged from seven in *trnS*-*rps4* to 86 in *trnQ*-*5'rps16*, and the normalized PIC value ranged from 0.667 to 8.190 in these same gene regions. *MatK* ranked 30/36 gene regions tested (normalized PIC value = 1.238), and *rbcL* 27/36 gene
regions tested (normalized PIC value = 1.524). This was the only lineage in which *rbcL* outperformed *matK* (16 and 13 raw PICs respectively), but both regions were close to the bottom of regions tested in this lineage and 13 regions were at least twice as variable than either region.

*Monocots—*Two regions, *trnS*-*trnG* and *5'rps12*-*rpl20*, are absent from monocots, so only 34 regions could be compared. The raw PIC value within this lineage ranged from four in *rbcL* to 78 in *rpl32*-*trnL* and the normalized PIC value ranged from 0.45 to 8.72 in these same gene regions. *MatK* was 25/34 gene regions tested and tied with *psbB*-*psbH* (normalized PIC value =1.676) *RbcL* ranked 34/34 gene regions tested (normalized PIC value = 0.45). *MatK* was about five times less variable than *rpl32*-*trnL*, while *rbcL* was nearly 19.5 times less variable.

*Minuartia—*The raw PIC value within this lineage ranged from two in *trnH*-*psbA* to 89 in *trnQ-5'rps16*, and the normalized PIC value ranged from 0.13 to 5.82. *MatK* ranked 15/36 gene regions tested (normalized PIC value = 2.94), and *rbcL* 35/36 gene regions tested (normalized PIC value = 0.46). With the exception of *rbcL*, this is one of two lineages in which PIC values were generally higher across all the regions tested than in the other lineages (see Shaw et al., 2007). *MatK* had two to three times the number of raw PICs (45) than it did in five of the six other lineages, but *matK* was still half as variable as the top performing region (*trnQ-5'rps16*) in this lineage, while *rbcL* was only better than *trnH-psbA*.

*Prunus—*The raw PIC value within this lineage ranged from two in *psbA*-*3'trnK* and *matK*-*5'trnK* to 62 in *rpl32*-*trnL*, and the normalized PIC value ranged from 0.24 to 7.36. *MatK* ranked 22/36 gene regions tested, tied with *trnS*-*trnM* (normalized PIC value = 2.14), and *rbcL* ranked 32/36 gene regions tested, tied with the *trnL* intron and *3'trnK*-*matK* (normalized PIC value =

0.71). *MatK* was nearly 3.5 times less variable than *rpl32*-*trnL*, while *rbcL* was more than 10 times less variable.

*Hibiscus—*The raw PIC value within this lineage ranged from two in *psbB*-*psbH* to 81 in *3' trnV-ndhC* and the normalized PIC value ranged from 8.80 to 0.22. *MatK* ranked 17/36 gene regions tested, tied with *trnT*-*trnL* and the *rps16* intron (normalized PIC value = 2.40), and *rbcL* 33/36 gene regions tested, tied with *3'trnK*-*matK* (normalized PIC value = 0.76). *MatK* was more than 3.5 times less variable than *3'trnV*-*ndhC*, while *rbcL* was more than 11.5 times less variable.

*Gratiola—*The raw PIC value within this lineage ranged from 10 in *trnL*-*trnF* to 125 in *rpl32 trnL* and the normalized PIC value ranged from 0.53 to 6.62. *MatK* ranked 23/36 gene regions tested (normalized PIC value = 2.23), and *rbcL* 29/36 gene regions tested (normalized PIC value $= 1.22$). This was the other lineage in which PIC values were generally high across the regions tested when compared to the other lineages. *MatK* was about three times less variable than the top-performing region, *rpl32-trnL*, while *rbcL* was about five times less variable.

*Carphephorus—*Four of the gene regions previously tested are absent in this lineage (*rpoB*-*trnC*, *trnS*-*trnG*, *trnD*-*trnT*, and *psbM*-*trnD*) due to rearrangements within the genome. The raw PIC value within this lineage ranged from three in *psbA*-*3'trnK* to 41 in *atpI*-*atpH*, and the normalized PIC value from 0.45 to 6.15. *MatK* ranked 18/32 gene regions tested and tied with *ycf6*-*psbM* (normalized PIC value = 3.3) *RbcL* tied with *ndhJ*-*trnF*, *psbB*-*psbH*, *trnL* intron, and *trnL*-*trnL* for 26/32 (normalized PIC value = 1.05). Of the 32 regions tested in this lineage, *matK* was about half as robust as *atpI*-*atpH*, while *rbcL* was nearly 6 times less variable.

Assessment of species identification using Prunus

To assess species identification in the model taxon *Prunus* we generated sequence data for *rbcL*, *matK*, *trnSGG*, *trnLLF*, and *trnH*-*psbA*. Table 4 shows that *trnH*-*psbA* was the shortest gene region with an aligned length of 520 bp, followed by *rbcL* (574 bp), *matK* (815 bp), *trnLLF* (974 bp), and *trnSGG* (1938 bp). *RbcL* had the fewest total parsimoniously informative characters and indel characters (33) across the five gene regions tested, followed by *trnH*-*psbA* (94), *matK* (62), *trnLLF* (163), and *trnSGG* (301) (see Table 4). Genetic variability was measured by taking the number of parsimoniously informative characters and dividing by the aligned length. *TrnLLF* was the most variable (9.24%), and *rbcL* was the least variable (5.75%), with *trnSGG*, *trnH*-*psbA*, and *matK* in the middle of these two regions (7.38%; 7.31%; 6.75%, respectively).

Using 203 samples representing 54 *Prunus* species, species delimitation was tested using each gene region individually and in all possible multilocus combinations with the other four gene regions for a total of 31 separate aligned and coded sequence data matrices (Table 5). Each dataset was analyzed using UpD and BA, which have been two of the more popular metrics used to date. The results are broken down by metric below for easier reading.

*Uncorrected p-Distance—*Table 5 and Figure 4 show the number of species positively identified for each of the 31 possible gene region combinations. *RbcL* identified the fewest species in single locus tests (7/54 species correctly identified; 12.96%), followed by *matK* (13; 24.07%), *trnHpsbA* (16; 29.63%), and *trnLLF* and *trnSGG* (18; 33.33%). In the single locus tests, only three species were identified by all five gene regions (*P*. *caroliniana*, *P*. *maritima*, and *P*. *subcordata*). A full list of species identified by each single locus is provided in Table 6. The PWG (2009) two

loci combination of *matK*+*rbcL* identified five more species than *matK* alone and the same number as *trnSGG* alone (18). The three-gene region combination of *matK*+*trnLLF*+*trnH*-*psbA* discriminated the most species (27; 50%); interestingly, all five-gene regions combined identified fewer species (23; 27, respectively) than the above mentioned three-gene region combination.

Monophyly using Bayesian Analysis—Table 5 and Figure 4 show the number of species positively identified for each of the 31 possible single and multilocus combinations. Figure 5 shows the five single locus trees with each monophyletic clade highlighted. No assessment of topological differences between the trees generated was made and only the five single locus trees are shown since there was little improvement on species identification as more data were added. Again, *rbcL* identified the fewest species in single locus tests (7/54 species correctly identified, 12.96%), followed by *trnH-psbA* and *trnLLF* (14; 25.93%), *matK* (16; 29.63%), and *trnSGG* (22; 40.74). In the single locus tests, only one species was identified by all five gene regions (*P*. *subcordata*). A full list of species identified by each single locus is provided in Table 6. The PWG (2009) two loci combination of *matK*+*rbcL* identified two more species than *matK* alone but four fewer species than *trnSGG* did alone. The four-gene region combination of *matK*+*rbcL*+*trnSGG*+*trnH*-*psbA* and the five-gene region combination discriminated the most species (26; 48.15%).

Discussion

It has been difficult to reach a consensus on which gene regions to use for plant barcoding due to their complex and often intertwined evolutionary histories. CBOL accepted the proposal of the PWG (2009) in late 2009, but called for more testing of *matK* and *rbcL* by the

plant barcoding community, as well as other potential DNA barcode regions to determine if these gene regions are indeed the best for plant DNA barcoding. The focus of our research was to test these two touted gene regions by comparing their relative genetic variability to 34 noncoding cpDNA gene regions previously studied by Shaw et al. (2005; 2007), and also to compare the performance of these same two regions with respect to species level identification. What follows is a discussion on the utility of *matK* and *rbcL* as 'universal' plant barcodes based on amplification and sequencing success, alignment issues, genetic variability, and direct species identification.

Amplification, Sequencing and Alignability

Much debate has been focused on identifying gene regions that easily amplify, sequence, and easily align using computerized alignment programs across the various land plant lineages (Kress et al., 2005; PWG, 2009). For these reasons, many in the plant barcoding community have advocated the use of coding over noncoding cpDNA regions to ensure amplification with a single set of primers, quality bidirectional sequencing, and easy alignment using global alignment algorithms without having to adjust sequence data by hand (Newmaster et al., 2006; Ford et al., 2009; PWG, 2009).

Using the PWG (2009) primer sets and PCR protocols for *matK* and *rbcL*, we found no amplification issues across the seven angiosperms lineages or within the 203 accessions within *Prunus* plus *Physocarpus opulifolius* and quality sequence data was obtained for all of the abovementioned accessions. As noted earlier, *matK* sequence data could not be generated for two of the three gymnosperms from the *Tortoise and the Hare* dataset (Shaw et al., 2005). The PWG (2009) noted their own difficulties with obtaining quality sequences for gymnosperms without using taxa specific *matK* primer sets. Why gymnosperms have been difficult to sequence using

‗universal' primers is likely due to differing evolutionary histories than angiosperms. It is clear that further evaluation of how to barcode this group is needed before the barcoding community can proclaim that a barcode for land plants has been found.

The use of noncoding regions, such as *trnH*-*psbA*, have been complicated by the presence of mononucleotide runs (poly A/T runs), which cause slippage of the DNA polymerase during PCR causing a 'stutter' effect (Devey et al. 2009). This makes obtaining quality sequences difficult. We found few difficulties related to PCR amplification and obtaining quality sequences was generally not an issue for the three noncoding regions tested using *Prunus*. Despite several mononucleotide repeat regions in the *trnH*-*psbA* and *trnSGG* datasets, sequencing was successful because of the short length of $trnH-psba$ (\sim 300 bp unaligned) and the use of internal primers to ensure quality sequence data for *trnSGG*. Recently, Fazekas et al., (2010), found that two DNA polymerases, Phusion (Finnzymes, Espoo, Finland) and Herculase II fusion (Agilent, Santa Clara, California, USA) regularly improved quality sequence reads through mononucleotide repeats up to 13 bp. Based on these findings, the use of noncoding regions as barcodes should still be considered since they are generally more variable than coding regions, more robust at infrageneric levels (Gielly and Taberlet, 1994, Kelchner 2000), and obtaining quality sequences is becoming easier through technological advances (Fazekas et al., 2010).

Sequence alignment has garnered the attention of many in the barcoding community because automated alignment programs can align coding regions more accurately and faster than noncoding regions since they generally contain very few indel characters (Ford et al., 2009; PWG, 2009). We attempted to align *trnH-psbA* using Clustal X v. 2.0 (Larkin et al., 2007), but computing processing time was slow $($ \sim four days). More importantly, results could not be relied on since they still needed to be adjusted by hand. The use of coding regions comes at a cost since

there are potentially fewer informative characters to facilitate the identification of closely related species. Noncoding regions generally have a higher rate of nucleotide substitutions and it is has been shown that indel characters can help resolve relationships between closely related taxa (Gielly and Taberlet, 1994; Kress, 2007). Alignment of the three noncoding regions tested using *Prunus* was generally straight forward. The only exception was the *trnH-psbA* dataset, in which 177 bases were omitted (more than half of the total base pairs sequenced), in order to align the data. Despite this omission of data, *trnH-psbA* was still as robust as *matK*, more robust than *rbcL*, and nearly as robust as *matK*+*rbcL* at species identification. According to Erickson et al. (2008), current multiple alignment algorithms are insufficient to handle large amounts of data from a noncoding region and the alternative use of pairwise alignments are too slow for large scale database application. While outside the scope of this paper, improvements to sequence alignment algorithms clearly pose a challenge for the bioinformatics community to solve in order to make analysis of data faster and reduce search times on databases like BOLD or GenBank.

The plant DNA barcoding community has focused on identifying gene regions that are easy to PCR amplify, sequence, and align. Based on these criteria, our data supports the selection of *matK* and *rbcL*. Importantly, we had no PCR amplifying issues and almost no difficulties obtaining quality sequence data for the three noncoding regions used in this study. Our biggest obstacle was the alignment of the three noncoding gene regions, which proved to be much more time consuming than *matK* and *rbcL*. However, Kress et al. (2005) argue that technology will continue to advance and solve this problem, thus noncoding regions should continue to be considered as potential plant DNA barcodes rather than being disregarded by the barcoding community for the sake of simple sequencing and alignment.

Comparing matK and rbcL to 34 noncoding gene regions using seven plant lineages

The focus of *The Tortoise and the Hare* series (Shaw et al., 2005; 2007) was to find the most variable cpDNA regions for low-level phylogenetic research. This work continues today with a number of studies focused on finding coding and noncoding cpDNA region that meet the criteria to be considered a DNA barcode (Haider, 2003; Shaw et al., 2005; 2007; Ford et al., 2008). With respect to phylogenetic studies, chloroplast noncoding regions have been shown to be more useful below the genus level since they accumulate mutations faster than coding regions (Kelchner, 2000).

MatK and *rbcL* were suggested early on in the search for a plant DNA barcoding region (Newmaster et al., 2006; Chase et al., 2007) and have been two of the more widely tested coding regions (Newmaster et al., 2006; Newmaster et al., 2007; Chase et al, 2007; Lahaye et al., 2008; Ford et al., 2009; Hollingsworth et al., 2009; PWG, 2009). Comparing these two regions to 34 previously tested noncoding cpDNA regions, we show that five noncoding gene regions are at least twice as variable as *matK* and 27 noncoding gene regions are more variable than *rbcL* (Fig. 3). Based on these results, more attention should be paid to testing noncoding regions despite the potential obstacles in alignment and search algorithms since they overwhelmingly outperformed *matK* and *rbcL* in this part of our study in every measure (Figs. 2 and 3).

DNA barcoding rests on finding regions with sufficient genetic variability between species, but that variability needs to be low within a species (intraspecific variation) so that, if possible, multiple haplotypes per species do not have to be described. With respect to this criterion, *matK* and *rbcL* are good choices as there was very little genetic distance within a species (intraspecific variation); however, they are poor choices since they also contained very

low levels of genetic variation between species (interspecific variation). Figure 3 shows that there are four regions that are more than twice as variable as *matK* and 28 regions compared to *rbcL*, which suggests that the plant barcoding community should continue to test other regions before settling on these two. In light of our results, we wonder if a plant DNA barcode region(s) should be primarily chosen on the basis of alignability at the expense of species discrimination.

Assessment of species identification using Prunus

For direct species identification tests using *Prunus* L., we were unable to achieve the 70% mark touted by the PWG (2009). Using their combination of *matK*+*rbcL* (18/54 species were positively identified). The three noncoding regions did not achieve the 70% mark alone or in combination, but they did identify more species than *matK* (when analyzed using UpD) or *rbcL* (regardless of metric used) in single locus tests (Table 5). Curiously, *trnLLF* and *trnH*-*psbA* identified fewer species using BA than UpD, while *matK* and *trnSGG* identified more in locus tests (Tables 5 and 6). *RbcL* was the least informative region at species discrimination in single locus tests, regardless of metric used, and Ford et al. (2009) point out that the inclusion of *rbcL* in any barcoding scenario is due to its historical popularity in phylogenetics not empirical testing. However, the PWG (2009) noted that species resolution improved when *rbcL* was included in multilocus tests, but we did not find this to be the case. In fact, we found that *matK* performed as well or better than *rbcL* in all the various multilocus tests in which the other was excluded and that results were mixed when both regions were present (Table 5).

We found little to no species resolution in many of the subgeneric groups of *Prunus*, irrespective of gene region combinations or metric used, due to zero genetic distance or overlaps in intraspecific and interspecific genetic distances between species. For example, there was no genetic distance between six species of North American plums, and the same six species formed

a polytomy using BA in all 31 consensus trees. Our results are in line with others who found poor species identification due to little genetic variation in cpDNA between species (Edwards et al., 2008; Spooner, 2009; Steel et al., 2010). According to Edwards et al. (2008), at least three gene regions will be needed for species identification for *Aspalathus* L. (Fabaceae), a genus endemic to the fynbos of South Africa. Not surprisingly, our results suggest that it will take between eight and ten gene regions to identify 70% of species in *Prunus*, which has a much broader range than *Aspalathus*.

It is noteworthy that species resolution reached an asymptote as the number of gene regions concatenated increased from two to three (Fig. 6) and species resolution only increased marginally as the number of loci concatenated increased from three to four and in several instances it decreased (Table 5, Fig. 6). According to Fazekas et al, (2009), stringing regions together will do little to improve the overall success of plant barcoding and our results concur. It may be that plants are just harder to discriminate than animals due to lower levels of genetic variability in chloroplast markers (Fazekas et al., 2009). This suggests that species identification in *Prunus*, and very likely in other large genera with closely related species, will continue to be difficult regardless of the gene regions used (see Spooner, 2009; Steel et al., 2010). As Meyer and Paulay (2005) point out, DNA barcoding will work well in thoroughly sampled groups with solid taxonomic foundations, but science has only described 10% of the flora and fauna on Earth (Wilson, 2003), which means there are more poorly understood groups than well understood ones.

Our results, coupled with others (Spooner, 2009; Steel et al., 2010), strongly suggests that developing a universal plant barcoding scheme will continue to be a difficult task. Our results do not support the use of the PWG (2009) *matK*+*rbcL* regions as barcodes for *Prunus* since it only

identified 1/3 of the species in our dataset. Even though the three noncoding regions were at least as variable as *matK* (Table 4), they had their own shortcomings. *TrnLLF* was easy to amplify, sequence, and align, but it did not enhance species resolution in multilocus tests, so its utility as a barcode region is probably limited at best. *TrnSGG* was the best performing region, but it is about 50% longer than *matK+rbcL*, harder to amplify, sequence, and align than either of the coding regions. Despite the drawbacks, *trnSGG* should not be ruled out of any barcoding scheme since it contains a high rate of genetic variability and technology will continue to improve making sequencing and alignment difficulties non-issues. *TrnH-psbA* was the most unusual region, in that it contained a high level of genetic variability despite the omission of nearly 50% of the raw sequence data due to alignability issues. Just like with *trnSGG*, if automated sequence alignment algorithms can be improved, then based our results and those of Kress et al. (2005; 2007) we support its inclusion in any barcoding scheme.

We strongly encourage the plant barcoding community to continue to investigate other gene regions, especially those in the small single copy region, which have received little attention from the plant barcoding community, before settling on *matK*+*rbcL*. We also urge the community to more robustly test the multigene-tiered approach put forth by Newmaster et al. (2006). The tiered approach starts with a coding region that can provide resolution to family or genus followed by one or more noncoding regions, which can be taxa specific, to resolve samples to species (Newmaster et al., 2006). If the criteria are easy amplification and sequencing across the various land plant lineages (including gymnosperms, bryophytes, liverworts, and pteridophytes), then *rbcL* makes the most sense to be the core gene region since it amplified across all 30 samples in our dataset, sequenced without issue (including gymnosperms), and did not require any manual editing or alignment. If, however, the plant barcoding community wants

to focus on angiosperms, which make up approximately 90% of the world's flora, *matK* would be the better choice since it contained a higher level of genetic variation than *rbcL* (Figs. 1 and 2).

Conclusions

Plant DNA barcoding has been anything but easy. Since Kress et al. (2005), a number of studies have been published that have looked at the utility of a number of coding and noncoding cpDNA regions as potential DNA barcodes across the different land plant lineages. Despite all of this work, results from one study cannot be easily compared to one another due differing study designs. For example, many studies have tested a suite of gene regions across a broad phylogenetic set of taxa. It is well known that genera within and between families of plants are phylogenetically nonequivalent, i.e. genera across families may have very different genetic divergence rates depending on the life histories of the various species included in those genera or families. This suggests that results regarding the utility of gene regions from these studies may be overstated since interspecific variation may not have been fully accounted for within the various genera and families they selected (see Fazekas, 2008; PWG, 2009). On the other hand, studies like Newmaster et al., 2006 may also not have accounted for genetic variability due geographic restrictions and small sample size. This means that while a gene region worked well in a particular study the results cannot be assumed to be a universal truth without more robust testing.

In 2009, in an attempt to standardize plant barcoding, CBOL established the combination of *matK* and *rbcL* as the "universal" plant barcodes based on their performance in previous studies, most notably the PWG (2009). From our study, it is clear that *matK* and *rbcL* lack enough genetic variability to identify closely related species in *Prunus* and therefore be

considered the ‗universal' plant barcodes. Our results, coupled with previous studies (Kress and Erickson, 2007; Edwards et al., 2008, Spooner, 2009; Newmaster et al. 2008), suggests that it is time to refocus our attention to other regions in the chloroplast genome, such as *ndhf*-*rpl32* and *rpl32*-*trnL*, which Shaw et al., (2007) found to be two of the most variable noncoding regions. Many more inquiries similar to the ones performed by Edwards et al. (2008), Farrington et al. (2009), Spooner (2009), and Steel (2010), are needed before we settle on a plant DNA barcode system. Better to go slow than learn later on that it could have been better.

Therefore, we conclude that it would be a mistake to settle on *matK* and *rbcL* in a rush to begin building a plant DNA barcode database unless we are willing to settle for $\leq 70\%$ species resolution. We have shown that a number of chloroplast gene regions display greater levels of genetic variation than *matK* or *rbcL*, as we all as amplify and sequence just as easily. We have also shown that species identification, regardless of metric or gene regions used, will continue to be difficult within a large genus that contains many closely related species, and that there is a need for more testing at the generic level to determine which gene regions will best delimit species. Ultimately, plant barcoding may never be as successful as it is in animals due to the unique and challenging evolutionary histories of plants.

Table 1. A sample of DNA Barcoding Studies. For cpDNA regions, coding regions are listed first followed by noncoding regions in alphabetical order. $nDNA = nuclear$ genome; $cpDNA = chloroplast$ genome.

Table 2. Primers used—Includes primer name, forward and reverse sequences, aligned length of the gene region in *Prunus* L., and author. $cpDNA = Chloroplast$ genome; bp = base pairs.

Table 3. Comparison of the PWG (2009) Gene Regions—Comparison of the relative utility of *matK* and *rbcL* to one another across seven angiosperm lineages. Avg. = Average; PIC $=$ Potentially informative characters. Avg. $=$ average; L. $=$ length; PIC $=$ potentially informative characters.

Table 4. Comparison of five potential DNA barcoding regions—The relative utility of each potential chloroplast gene region based on parsimoniously informative characters and indels, in 203 samples of *Prunus* L. and *Physocarpus opulifolius* (L.) Maxim. Calculations: % informative characters = parsimoniously informative characters/Aligned Length*100; % Indels = Indels/Aligned length*100; bp = base pairs.

Table 5. Assessment of species identification using uncorrected pairwise distance and percent monophyly—54 *Prunus* L. species were used to compare species identification success of five chloroplast gene regions and multilocus combinations of those loci using two common barcoding metrics. This table includes number of species identified and corresponding percentages. Information in bold represents the gene region(s) that identified the most species for each metric. Numbers in bold and italics refer to the PWG (2009) combination of $\text{matK} + \text{rbcL}$. bp = base pairs; ID = Identification; UpD = Uncorrected pairwise distance; BA = Bayesian analysis

Table 6. Species Identified—*Prunus* L. species identified by each gene region alone. Species in bold were identified by all five regions independently. For species identified using BA, species are listed in the order they appear on the trees in Figure 4. $BA = Bayesian$ analysis; UpD = Uncorrected pairwise distance.

Figure 1. Chloroplast Genome—*Gossypium hirsutum* L. chloroplast genome (Lee et al., 2006) is presented here as a reference map to show where the five gene regions tested are located. Gene regions tested are marked with an *****.

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Figure 2. Normalized PIC values across seven angiosperm lineages—Relative genetic variability of 36 chloroplast gene regions across seven angiosperm lineages. Figure recreated using data from Shaw et al. (2007) with *matK* and *rbcL* data added from this study. Avg. $=$ Average; PIC $=$ potentially informative characters

Figure 3. Comparison of *matK* and *rbcL* to 34 noncoding chloroplast regions—Average Normalized PIC values were calculated for the regions below to determine which regions contain the highest levels of genetic variation across seven angiosperm lineages. Figure recreated using data from Shaw et al. (2007). Numbers in parentheses above gene regions represent the average aligned length across the seven angiosperm lineages for that particular gene region. Average aligned lengths may differ from Shaw et al (2005; 2007) due to rounding differences. cpDNA = Chloroplast genome; PIC =

 \blacksquare Number of spp. Mononphyletic (BA) \blacksquare Number of spp. ID (UpD)

Figure 4. Species Identification of *Prunus*—Percentage of 54 species correctly identified using two common barcoding metrics. spp. = species; $ID =$ identified; $UpD =$ Uncorrected pairwise distance; $BA = Bayesian$ analysis. No. = Number; spp. = species.

Figure 5. Consensus trees—Species discrimination for the five single loci tests using Bayesian analysis to determine the number of species resolved as monophyletic. Clades highlighted in black were considered correctly identified following Fazekas et al. (2008). Species appear on tress as listed in Table 5 for Bayesian analysis.

Figure 5. Continued

 30.0

Figure 5. Continued

 $\bf{30.0}$

Figure 5. Continued

Figure 5. Continued

Figure 6. Relationship between sequence length and number of *Prunus* L. species identified— Graph shows that species identification begins to decelerate as the number of loci concatenated increases from two to three and hit an asymptote as the number of nucleotide characters increases to 3000 bp. Data is for 31 separate analyses using Bayesian analyses. bp = base pairs; spp.= species; No. = Number.

Literature Cited

- Alfred P. Sloan Foundation. Barcode of Life. <http://www.sloan.org/program/7>. Accessed 27 August 2009.
- APG II. 2003. An update of the angiosperm phylogeny group classification for the orders and families of flowering plants: APG II. *Botanical Journal of the Linnean Society* 141: 399- 436.
- Asahina, H., J. Shinozaki, K. Masuda, Y. Morimitsu, and M. Satake. 2010. Identification of medicinal *Dendrobium* species by phylogenetic analyses using *matK* and *rbcL* sequences. *Journal of Natural Medicines* 64: 133-138.
- Bailey, L.H. and E.Z. Bailey. 1941. Hortus second: a concise dictionary of gardening, general horticulture, and cultivated plants in North America. Macmillan, New York, New York, USA.
- Blackburn, B. 1952. Trees and shrubs in eastern North America: keys to the wild and cultivated woody plants in the temperate regions exclusive of conifers. Oxford University Press, New York, New York, USA.
- Blaxter, M.L., J. Mann, T. Chapman, F. Thomas, C. Whitton, R. Floyd, and E. Abebe. 2005. Defining operational taxonomic units using DNA barcode data. P*hilosophical Transactions of the Royal Society B* 360: 1935-1943.
- Borchsenius, F. 2009. Department of Biological Sciences, University of Aarhus, Denmark. Available online at: http://192.38.46.42.aubot/fb/FastGap_home.htm
- Borris, H., H. Brunke, and M. Keith. 2006. Almond profile: updated 2009 by D.Huntrods. *Agricultural Marketing Resources Center*.
- Bortiri, E., S. Oh, J. Jiang, S. Baggett, A. Granger, C. Weeks, M. Buckingham, D. Potter, and D.E. Parfitt. 2001. Phylogenetic and systematics of *Prunus* (Rosaceae) as determined by sequence analysis of ITS and the chloroplast *trnL*-*trnF* spacer DNA. *American Society of Plant Taxonomists* 26: 797-807.
- Bortiri, E., S. Oh, F. Gao, and D. Potter. 2002. The phylogenetic utility of nucleotide sequences of sorbitol 6-phosphate dehydrogenase in *Prunus* (Rosaceae). *American Journal of Botany* 89: 1697-1708.
- Bortiri, E., B. Vanden Heuvel, and D. Potter. 2006. Phylogenetic analysis of morphology in *Prunus* reveals extensive homoplasy. *Plant systematics and Evolution* 259: 53-71.
- CBOL Plant Working Group. 2009. A DNA barcode for land plants. *Proceedings of the National Academy of Sciences*, *USA* 106: 12794-12797.
- Chase, M.W., N. Salamin, M. Wilkinson, J.M. Dunwell, R.P. Kesanakurthi, N. Haidar, and V. Savolainen. 2005. Land plants and DNA barcodes: short-term and long-term goals. P*hilosophical Transactions of the Royal Society B* 360: 1889-1895.
- Chase, M.W., R.S. Cowan, P.M. Hollingsworth, C. van den Berg, S. Madriñán, G. Petersen, O. Seberg, T. Jørgsensen, K.M. Cameron, M. Carine, N. Pedersen, T.A.J. Hedderson, F. Conrad, G.A. Salazar, J.E. Richardson, M.L. Hollingsworth, T.G. Barraclough, L. Kelly, and M. Wilkinson. 2007. A proposal for a standardised protocol to barcode all land plants. *Taxon* 56: 295-299.
- Chen S., H. Yao, J. Han, C. Liu, J. Song, L. Shi, Y. Zhu, X. Ma, T. Gao, X. Pang, K. Luo, Y. Li, X. Li, X. Jia, Y. Lin, and C. Leon. 2010. Validation of the ITS2 Region as a Novel DNA Barcode for Identifying Medicinal Plant Species. *Plos One* 5: e8613. doi:10.1371/journal.pone.0008613.
- Clare, E.L., K.C.R. Kerr, T.E. von Königslöw, J.J. Wilson, and P.D.N. Hebert. 2008. Diagnosing Mitochondrial DNA diversity: applications of a sentinel gene approach. *Journal of Molecular Evolution* 66: 362-367.
- Correll, D.S. and M.C. Johnston. 1970. Manual of the vascular plants of Texas. Texas Research Foundation, Renner, Texas, USA.
- Cunningham, A.B. and U. Schippman. 1997. Trade in *Prunus afrciana* and the implementation of CITES. German Federal Agency for Nature Conservation.
- Devey, D.S., M.W. Chase, and J.J. Clarkson. 2009. A stuttering start to plant DNA barcoding: microsatellites present a previously overlooked problem in noncoding plastid regions. *Taxon* 58: 7-15.
- Duncan, W.H., and M.B. Duncan. 1988. Tress of the southeastern United States. University of Georgia Press, Athens, Georgia, USA.
- Edwards, D., A. Horn, D. Taylor, V. Savolaninen, and J.A. Hawkins. 2008. DNA barcoding of a large genus, *Aspalathus* L. (Fabaceae). *Taxon* 57: 1317-1327.
- Erickson, D.L., J. Spouge, A. Resch, L.A. Weigt, and J. Kress. 2008. DNA barcoding in land plants: developing standards to quantify and maximize success. *Taxon* 57: 1304-1316.
- Farrington, L., P. MacGillivray, R. Faast, and A. Austin. 2009 Investigating DNA barcoding options for the identification of *Caladenia* (Orchidaceae) species. *Australian Journal of Botany* 57: 276-286.
- Fazekas, A.J., K.S. Burgess, P.R. Kesanakurti, S.W. Graham, S.G. Newmaster, B.C. Husband, D.M. Percy, M. Hajibabaei, and S.C.H. Barrett. 2008. Multiple Multilocus DNA barcodes from the plastid genome discriminate plant species equally well. *PlosOne* 3: e2802. doi: 10.1371/journal.pone.0002802.
- Fazekas, A.J., P.R. Kesanakurti, K.S. Burgess, D.M. Percy, S.W. Graham, S.C.H. Barrett, S.G. Newmaster, M. Hajibabaei, and B.C. Husband. 2009. Are plant species inherently harder to discriminate than animal species using DNA barcode markers?. *Molecular Ecology Resources* 9: 130-139.
- Fazekas, A.J., R. Steeves, S.G. Newmaster, and P.M. Hollingsworth. 2010. Stopping the stutter: Improvements in sequence quality from regions with mononucleotide repeats can increase the usefulness of non-coding regions for DNA barcoding. *Taxon* 59: 694-697.
- Fernald, M.L. 1950. Gray's manual of botany, 8th ed. American Book, New York, New York, USA
- Flora of North America Editorial Committee, eds. 1993+. Flora of North America North of Mexico. 16+ vols. New York and Oxford.
- Food and Agriculture Organization of the United Nations Website. TradeSTAT: Crops and livestock products. <http://faostat.fao.org> Accessed 27 August 2009.
- Ford, C.S., K.L. Ayers, N. Toomey, N. Haider, J. Van Alphen Stahl, L.J. Kelly, N. Wikstrom, P.M. Hollingsworth, R.J. Duff, S.B. Hoot, R.S. Cowan, M.W. Chase, and D M.J. Wilkinson. 2009. Selection of candidate coding DNA barcoding regions for use on land plants. *Botanical Journal of the Linnean Society* 159: 1-11.
- Gielly, L., and P. Taberlet. 1994. The use of chloroplast DNA to resolve plant phylogenies: noncoding versus *rbcL* sequences. *Molecular Biology and Evolution* 11: 769-777.
- Gene Codes. 2007. Sequencher v. 4.7. Gene Codes, Ann Arbor, Michigan, USA.
- Gleason, H.A. 1952. The new Britton and Brown illustrated flora of the United States and Canada, vol. 2. Lancaster Press, Lancaster, Pennsylvania, USA.
- Gleason, H.A., and A. Cronquist. 1991. Manual of vascular plants of northeastern United States and adjacent Canada, 2nd ed. New York Botanical Garden, Bronx, New York, USA.
- Godfrey, R.K. 1988. Trees, shrubs, and woody vines of northeastern Florida and adjacent Georgia and Alabama. University of Georgia Press, Athens, Georgia, USA
- Haider, N. 2003. Developments and use of universal primers in plants. PhD Thesis, University of Reading, United Kingdom.
- Hajibabaei, M., G.A.C. Singer, P.D.N. Hebert, and D.A. Hickey. 2007. DNA barcoding: how it complements taxonomy, molecular phylogenetics and population genetics. *Trends in Genetics* 23: 167-172.
- Hasagawa, M., K. Kishino, and T. Yano. 1985. Dating the human-ape splitting by a molecular clock of mitochondrial DNA. *Journal of Molecular Evolution* 22: 160-174.
- Hebert, P.D.N., A. Cywinska, S.L. Ball, and J.R. De Ward. 2003. Biological identification through DNA barcodes. *Proceedings of the Royal Society of London B* 270: 313-321.
- Hebert, P.D.N., M.Y. Stoeckle, T.S. Zemlack, and C.M. Frances. 2004. Identification of birds through DNA barcodes. *PlosOne* 2: 16571663.
- Hilu, K., D.T. Borsch, K. Müller, D.E. Soltis, P.S. Soltis, V. Savolainen, M.W. Chase, M.P. Powell, L.A. Alice, Evans R., H. Sauquet, C. Neinhuis, T.A.B. Slotta, J.G. Rohwer, C.S. Campbell, and L.W. Chatrou. 2003. Angiosperm phylogeny based on *matK* sequence information. *American Journal of Botany* 90: 1758–1776.
- Hollingsworth, M.L., A.A. Clark, L.L. Forrest, J. Richardson, R.T. Pennington, D.G.
- Long, R.S. Cowan, M.W. Chase, M. Gaudeul, and P.M. Hollingsworth. 2009. Selecting barcoding loci for plants: evaluation of seven candidate loci with species-level sampling in three divergent groups of land plants. *Molecular Ecology Resources* 9: 439-457.
- Keane, T.M., T.J. Naughton, and J.O. McInerney. 2007. MultiPhyl: A high-throughput phylogenomics webserver using distributed computing. *Nucleic Acids Research* 35: W33- W37.
- Kelchner, S. 2000. The evolution of non-coding chloroplast DNA and its application in plant systematics. *Annals of Missouri Botanical Garden* 87: 482-498.
- Kelly, L.J., G.K. Ameka, and M.W. Chase. 2010. DNA barcoding of African Podostemaceae (river-weeds): A test of proposed barcode regions. *Taxon* 59: 251-260.
- Kress, W.J., K.J. Wurdack, E.A. Zimmer, L.A. Weigt, and D.H. Janzen. 2005. Use of DNA barcodes to identify flowering plants. *Proceedings of the National Academy of Sciences*, *USA* 102: 8369--8374.
- Kress, W.J., and D.L. Erickson. 2007. A two-locus global DNA barcode for land plants: the coding rbcL gene complements the noncoding trnH-psbA spacer region. *PlosOne* 1: 1 - 10.
- Kress. W.J., and D.L. Erickson. 2008. DNA barcodes: genes, genomics, and bioinformatics. *Proceedings of the National Academy of Sciences*, *USA* 105: 2761-2762.
- Kress, W.J., D.L. Erickson, F.A. Jones, N.G. Swenson, R. Perez, O. Sanjur, and E. Bermingham. 2009. Plant DNA barcodes and a community phylogeny of a tropical forest dynamics plot in Panama. *Proceedings of the National Academy of Sciences of the*, *USA* 106(44): 18627-18632.
- Krussman, G. 1978. Manual of cultivatedbroad-leaved trees and shrubs. Vol. Itt, Pru-Z(translated 1986). Timber Press, Portland,Ore.
- Lahaye, R., M. van der Bank, D. Bogarin, J. Warner, F. Pupulin, G. Gigot, O. Maurin, S. Duthoit, T.G. Barraclough, and V. Savolainen. 2008. DNA barcoding the floras of biodiversity hotspots. *Proceeding of the National Academy of Sciences*, *USA* 105: 2923- 2928.
- Larkin M.A., G. Blackshields, N.P. Brown, R. Chenna, P.A. McGettigan, H. McWilliam, F. Valentini, I.M. Wallace, A. Wilm, R. Lopez, J.D. Thompson, T.J. Gibson, and D.G. Higgins. 2007. Clustal W and Clustal X version 2.0. *Bioinformatics* 23: 2947-2948.
- Lee, S. and J. Wen. 2001. A phylogenetic analysis of *Prunus* and the Amygdaloideae (Rosaceae) using ITS sequences of nuclear ribosomal DNA. *American Journal of Botany* 88:1150- 1160.
- Lee, S-B., C. Kaittanis, R.K. Jansen, J.B. Hostetler, L.J. Tallon, C.D. Town, and H. Daniell. 2006. The complete chloroplast genome sequence of *Gossypium hirsutum*: organization and phylogenetic relationships to other angiosperms. *BMC Genomics* 7: doi: 10.1186/1471-2164/7/61.
- Lingdi, L., G. Cuizhi, L. Chaoluan, C. Alexander, B. Bartholomew, A.R. Brach, D.E. Boufford, H. Ikeda, H. Ohba, K.R. Robertson, and S.A. Spongberg. 2003. Rosaceae In: Z. Y. Wu, P. H. Raven, and D. Y. Hong, eds. 2003. Flora of China. Vol. 9 (Pittosporaceae through Connaraceae). Science Press, Beijing, and Missouri Botanical Garden Press, St. Louis, USA.
- Maddison, D.R., W.P. Maddison. 2001. MacClade 4.06. Sinauer, Sunderland Massachusetts.
- Maddison, D.R., and K.-S. Schulz (eds.) 2007. The Tree of Life Web Project. <http://tolweb.org> Accessed 27 August.
- Maddison, W.P. and D.R. Maddison. 2009. Mesquite: a modular system for evolutionary analysis. Version 2.72 Available at: http://mesquiteproject.org
- Meier, R., K. Shiyang, G. Vaidya, and P.K.L Ng. 2006. DNA barcoding and taxonomy in Diptera: a tale of high intraspecific variability and low identification success. *Systematic Biology* 55: 715-728.
- Meyer, C.P. and G. Paulay. 2005.DNA barcoding: Error rates based on comprehensive sampling. *PLoS Biol*o*lgy* 3: e422
- Mitchell, A. 2008. DNA barcoding demystified. *Australian Journal of Entomology* 47: 169-173.
- Mowrey B. D. and D. J. Werner. 1990. Phylogenetic relationships among species of *Prunus* as inferred by isozyme markers. *Theoretical and Applied Genetics* 80: 129-133
- National Clonal Germplasm Repository for fruit and nut crops, Davis, CA. < http://www.ars.usda.gov/main/site_main.htm?modecode=53-06-20-00> Accessed 27 August 2009.
- Newmaster, S.G., A.J. Fazekas, and S. Ragupathy. 2006. DNA barcoding in land plants: evaluation of *rbcL* in a multigene tiered approach. *Canadian Journal of Botany* 84: 335- 341.
- Newmaster, S.G., A.J. Fazekas, R.A.D. Steeves, and J. Janovec. 2008. Testing candidate plant barcode regions in the Myristicaceae. *Molecular Ecology Resources* 8: 480-490.
- Newmaster, S.G., and S. Ragupathy. 2009. Testing plant barcoding in a sister species complex of pantropical Acacia (Mimosoideae, Fabaceae). *Molecular Ecology Resources* 9: 172–180.
- Nitta, J.H. 2008. Exploring the utility of three plastid loci for biocoding the filmy ferns (Hymenophyllaceae) of Moorea. *Taxon* 57:3 725-736.
- Ragupathy, S., S.G. Newmaster, M. Murugesan, and V. Balasubramaniam. 2009. DNA barcoding discriminates a new cryptic grass species revealed in an ethnobotany study by the hill tribes of the Western Ghats in southern India. *Molecular Ecology Resources* 9: 164-171.
- Rieseberg, L.H. 1997. Hybrid origins of plant species. *Annual Review of Ecology and Systematics* 28: 359-389.
- Palmer, J.D., K.L. Adams, Y. Cho, C.L. Parkinson, Y-L. Qiu, and K. Song. 2000. Dynamic evolution of plant mitochondrial genomes: Mobile genes and introns and highly variable mutation rates. *Proceedings of the National Academy of Sciences*, *USA* 97: 6960-6966.
- Potter, D., T. Eriksson, R.C. Evans, S. Oh, J.E.E. Smedmark, D.R. Morgan, M. Kerr, K.R. Robertson, M. Arsenault, T.A. Dickinson, and C.S. Campbell. 2007. Phylogeny and classification of Rosaceae. *Plant Systematics and Evolution* 266: 5-43.
- Radford, A.E., H.E. Ahles, and C.R. Bell. 1968. Manual of the vascular flora of the Carolinas. University of North Carolina Press, Chapel Hill, North Carolina, USA.
- Rheder, A. 1940 Manual of cultivated trees and shrubs hardy in North America, exclusive of the subtropical and warmer temperate regions, $2nd$ revised and enlarged edition. Macmillan, New York, USA.
- Ronquist, F., and J.P. Huelsenbeck. 2003. MRBAYES 3: Bayesian phylogenetic inference under mixed models. *Bioinformatics* 19: 1572-1574.
- Sass, C., D.P. Little, D.W. Stevenson, and C.D. Specht. 2007. DNA Barcoding in the Cycadales: Testing the Potential of Proposed Barcoding Markers for Species Identification of Cycads. *Plos One* 2: e1154 doi: 10.1371/journal.pone.0001154.
- Seberg, O., and G. Petersen. 2009. How many loci does it take to DNA barcode a Crocus? *PlosOne* 4: e4598. doi: 10.1371/journal.pone.0004598.
- Shaw, J. and R.L. Small. 2004. Addressing the "hardest puzzle in American pomology:" Phylogeny of *Prunus* section *Prunocerasus* (Rosaceae) based on seven noncoding chloroplast DNA regions. *American Journal of Botany* 91: 985-996.
- Shaw, J. and R.L. Small. 2005. Chloroplast DNA phylogeny and phylogeography of the North American plums (*Prunus* subgenus *Prunus* section *Prunocerasus*; Rosaceae). *American Journal of Botany* 92: 2011-2030.
- Shaw, J., E.B. Lickey, J.T. Beck, S.B. Farmer, W. Liu, J. Miller, K.C. Siripun, C.T. Winder, E.E. Schilling, and R.L. Small. 2005. The tortoise and the hare II: relative utility of 21 noncoding chloroplast DNA sequences for phylogenetic analysis. *American Journal of Botany* 92: 142–166.
- Shaw, J., E.B. Lickey, E.E. Schilling, and R.L. Small. 2007. Comparison of whole chloroplast genome sequences to choose noncoding regions for phylogenetic studies in angiosperms: the tortoise and the hare III. *American Journal of Botany* 94: 275–288.
- Small, J.K. 1933. Manual of the southeastern flora. Science Press Printing, Lancaster, Pennsylvania, USA.
- Smith, E.B. 1994. Keys to the flora of Arkansas. University of Arkansas Press, Fayetteville, Arkansas, USA.
- Spooner, D.M. 2009. DNA barcoding will frequently fail in complicated groups: an example in wild potatoes. *American Journal of Botany* 96: 1177-1189.
- Starr, J.R., R.F.C. Naczi, and B.N. Chouinard. Plant DNA barcodes and species resolution in sedges (*Carex*, Cyperaceae). *Molecular Ecology Resources* 9: 151-163.
- Steele, P.R., L.M. Friar, L.E. Gilbert, and R.K. Jansen. 2010. Molecular systematics of the neotropical genus *Psiguria* (Cucurbitaceae): Implications for phylogeny and species identification. *American Journal of Botany*. 97: 156-173.
- Stewart, K.M. 2003. The African Cherry (*Prunus Africana*): from hoe-handles to the international herb market. *Economic Botany* 57: 559-569.
- Steyermark, J.A. 1963. Flora of Missouri. Iowa State University Press, Ames, Iowa, USA.
- Swofford, D.L. 2001. PAUP 4.0: Phylogenetic analysis using parsimony (and other methods). Sinauer Associates, Sunderland, MA.
- Taberlet, P., E. Coissac, F. Pompanon, L. Gielly, C. Miquel, A. Valentini, T. Veramt, G. Corthier, C. Brochmann, and E. Willerslev. 2006. Power and limitations of the chloroplast trnL (UAA) intron for plant DNA barcoding. *Nucleic Acids Research* 35: e14.
- Tavaré, S. 1986. Some probabilistic and statistical problems on the analysis of DNA sequences. Lect. Math. Life Sci. 17:57-86.
- Tautz, D., P. Arctander, A. Minelli, R.H. Thomas, and A.P. Vogler. 2003. A plea for DNA taxonomy. *Trends in Ecology and Evolution* 18: 70-74.
- Valentini, A., F. Pompanon, and P. Taberlet. 2009. DNA barcoding for ecologists. *Trends in Ecology and Evolution* 24:2 110-117.
- Ward, R.D., T.S. Zemlak, B.H. Innes, P.R. Last, and P.D.N. Hebert. 2005. DNA barcoding Australia's fish species. *Philosophical Transactions of the Royal Society B* 360: 1847- 1857.
- Wen, J., S.T. Berggren, C. Lee, S. Ickert-Bond, T. Yi, K. Yoo, L. Xie, J. Shaw, and D. Potter. 2008. Phylogenetic inferences in *Prunus* (Rosaceae) using chloroplast *ndhF* and nuclear ribosomal ITS sequences. *Journal of Systematics and Evolution* 46: 322-332.
- Will, K.W., and D. Rubinoff. 2004. Myth of the molecule: DNA barcodes for species cannot replace morphology for identification and classification. *Cladistics* 20: 47-55.
- Wilson, E.O. 2000. A global map of biodiversity. *Science*. 289:5488 2279.
- Wofford, B.E., and E.W. Chester. 2002. Guide to the trees, shrubs, and woody vines of Tennessee. University of Tennessee Press, Knoxville, Tennessee, USA.
- Wolfe, K.H., W-H. Li, and P.M. Sharp. 1987. Rates of nucleotide substitution vary greatly among plant mitochondrial, chloroplast, and nuclear DNAs. *Proceedings of the National Academy of Sciences*, *USA* 84: 9054-9058.
- Wunderlin, R.P. 1988. Guide to the vascular plants of Florida. University of Florida Press, Gainesville, Florida, USA.
- Zhang, W., J.F. Wendel, and L.G. Clarke. 1997. Bamboozled again! Inadvertent isolation of fungal rDNA sequences from Bamboos (Poaceae: Bambusoideae). *Molecular Phylogenetics and Evolution* 8: 205-217.
- Ziegenhagen, B., B. Fady, V. Kuhlenkamp, and S. Liepelt. 2005. Differentiating groups of *Abies* species with a simple molecular marker. *Silvae Genetica* 54: 123-126.

CHAPTER 2

FUTURE DIRECTIONS OF DNA BARCODING

Introduction

DNA barcoding is a burgeoning field that has the potential to help better identify species and better understand evolutionary dynamics at the species level. The PWG proposed in late 2009 that two coding chloroplast regions, *rbcL* and *matK*, serve as the barcodes for all plants with the goal of standardizing and unifying the plant barcoding community. Despite the call for further evaluation of *matK* and *rbcL* by CBOL PWG (2009), BOLD has begun accepting *matK*+*rbcL* data for plant barcoding. My thesis research has shown that *matK*+*rbcL* contain low levels of genetic variability when compared to 34 noncoding cpDNA regions that were previously tested by Shaw et al. (2005; 2007). I have also shown that this combination is a poor species barcode for *Prunus* L. A direct test of species identification showed the < 50% of *Prunus* species could be identified using *matK*+*rbcL*. My data show that there are other cpDNA regions that should be tested before the plant barcoding community settles on these two regions.

Identifying a particular region(s) is not the only concern facing the plant barcoding community. Within the constructs of DNA barcoding, issues surrounding species sampling and analytics need to be evaluated by the barcoding community in order to improve species identification. These key aspects should be improved before the barcoding community goes any further. Below is a brief discussion on why these two issues are important to the future of DNA barcoding as a whole and suggestions for future projects.

Discussion

Species Sampling

Species sampling has long been a problem in molecular phylogenetic studies, due to the inherent costs associated with generating sequence data for each specimen, but costs continue to decrease and improved automated sequencing has decreased the time to obtain sequence data. However, species identification using DNA barcodes relies on measuring the amount of genetic variability within and between species. To date, there is no consensus as to how many specimens are needed to build a species identification database. Taberlet et al. (2006) suggested that at least ten individuals per species be sequenced to capture genetic variation within a species, while Matz and Nielsen (2005) proposed 12 individual per species. To resolve this problem, researchers should sample heavily within populations to adequately capture the amount of genetic diversity within a species before any database should be considered reliable. Currently there are just over 88,000 formally described species with a barcode (see BOLD: http://www.boldsystems.org/views/login.php) and most species in BOLD are represented by \leq 10 sequences (Hajibabaei et al., 2007) and the chances of assigning an unknown sample to species could be great since intra- and interspecific genetic variation may not been adequately captured in the database.

Recently, Zhang et al. (2010) looked more closely at how many samples per animal species are needed to sufficiently measure genetic variability within a species. Using *COI* sequence data from a species of butterflies (data obtained from original study by Hebert et al., 2004) and simulated data, they found that sampling 5-10 individuals per species was far from adequate to capture most genetic diversity within a species and further argue that most investigations have under sampled by at least 400% (Zhang et al., 2010). More importantly,

Zhang et al., (2010) note that there is no magic number for sample size, i.e. it is speciesdependent, so it is up to individual researchers to determine what the appropriate sample size is for each species they want to sample. Future plant barcoding research should look at how many species are needed to capture all possible haplotypes in plants since Zhang et al. (2010) focused on animal species. Considering the unique evolutionary histories of plants it would be interesting to see if the sampling needs in plants would be markedly different than those noted by Zhang et al. (2010).

Another potential study based on Zhang et al. (2010), could look at sampling species across their ranges to measure the amount of genetic variability within a single species complex in order to correctly capture all possible haplotypes. For example, studies could look at the genetic variability contained within species found in the following three types of ranges: narrow endemic species, a broad ranging species with a semi-restricted habitat, and a species with a cosmopolitan distribution. Intraspecific and interspecific variation could be markedly different for these three species depending on range overlap with sister species, since spatial closeness promotes haplotype sharing between closely related species (Shaw et al., 2004; 2005).One also has to consider time since divergence from most common recent ancestor, since reproductive barriers may not be fully developed allowing for hybridization. Following a sampling scheme like this, could help researchers better estimate how many samples per species are needed in order to capture genetic diversity within a species.

Until species sampling requirements are better understood, the assignment of unknown specimens to species will continue to be unreliable. This endeavor will not come without increased costs but it is imperative to building a better infrastructure. Ironically, increased sampling means that DNA barcoding bioinformatics and search algorithms, used by GenBank

and BOLD, are going to have to be significantly improved. Add to the fact that plant barcoding is going to require more than a single locus and this could be a difficult bioinformatics challenge.

Analytics

Analysis of species identification is a burgeoning field that is adapting quickly to the unique needs of DNA barcoding analytics (see Sarkar, 2009). An understanding of DNA barcoding bioinformatics is one of the most pressing needs for the barcoding community since it is the backbone of species identification. There are currently no agreed upon analytical protocols in place, so it is difficult to determine which markers will identify the most species since differing analytical methods may produce different results, as I show in the preceding part.

Currently, CBOL recommends the use of uncorrected pairwise distances for species identification; however, my research found tree building analyses to be slightly more robust than distance based metrics in some gene region combinations. Unfortunately, tree building analyses are slower to run and require the use of multiple samples per species to accurately depict relationships, so they may not be as useful as distance metrics are for search databases such as BOLD and GenBank. Distance metrics, on the other, rely on selecting an arbitrary threshold at which point one considers two samples separate species. Right now, that threshold has been set at 3% but there is no empirical data to suggest that this value has any validity. Future endeavors should test whether or not the 3% threshold is valid for species identification purposes.

Erickson et al. (2008) argue that bioinformatics should influence the selection of a plant barcode since analysis of data places requirements on the feasibility of database and algorithm design. They further point out that noncoding gene regions present a challenge to database design since they increase search times due to the frequent presence of indel characters (Erickson et al. 2008). Following this assertion, future projects could utilize the sequence data generated

for the 256 *Prunus* L. samples to test search times on BOLD or GenBank's BLAST for coding and noncoding gene regions to see if there is a significant difference between submitting a query and obtaining results, as well as how often the database returns a correct or incorrect identification. Future projects could also focus on new species identification programs (as they become available) and compare the results from this study to determine if these new programs improve species identification.

With the speed at which barcoding bioinformatics is changing it is going to be difficult to unify the field in the short term. Developing search databases and the analytics behind them requires that species sampling be adequate to capture all possible haplotypes as well as understanding evolutionary rates of change at the DNA level. Without these pieces the chances of assigning a sample to species incorrectly increases. Simply put, the DNA barcode library is still being constructed, and it is clear that the analytics will continue to improve species identification results.

Conclusions

DNA barcoding continues to garner more and more attention, but it is clear that there is a need to establish more rigorous standards and improve upon current ones. My research suggests that there are other cpDNA regions that may be more suitable barcodes than *matK* and *rbcL* and that the search for a plant barcode should not be over, unless we are satisfied with identifying 70% of plants. There are a number of other areas that the barcoding community must still address, most notably species sampling and bioinformatics.

The barcoding community needs to develop sampling standards and strategies as the field moves forward in order to ensure that within and between species genetic variation is accurately accounted. Without these improved sampling efforts it will be difficult to improve species

identification bioinformatics or search databases that are used to identify unknown specimens. Improvement to species sampling will undoubtedly enhance and improve the quality of queries from search databases.

The unique evolutionary histories of plants pose many hurdles for the plant barcoding to overcome, most notably incomplete lineage sorting, chloroplast sharing between closely related species, and the lack of genetic divergence between the chloroplasts of closely related species. There is still a lot of work to be done before DNA barcoding can be considered as successful in plants as it is in animals, generally speaking, but success is not inevitable. Despite the obstacles, DNA barcoding of the world's flora continues to move forward.

Literature Cited

- CBOL Plant Working Group. 2009. A DNA barcode for land plants. *Proceeding of the National Academy of Sciences* 106:31 12794-12797.
- Erickson, D.L., J. Spouge, A. Resch, L.A. Weigt, and J. Kress. 2008. DNA barcoding in land plants: developing standards to quantify and maximize success. *Taxon* 57:4 1304-1316.
- Hajibabaei, M., G.A. Singer, E.L. Clare, P.D.N. Hebert. 2007. Design and applicability of DNA arrays and DNA barcodes in biodiversity monitoring. *BMC Biology* 5: 24.
- Hebert, P.D.N., E.H. Penton, J.M. Burns, D.H. Janzen, and W. Hallwachs. 2004. Ten species in one: DNA barcoding reveals cryptic species in the neotropical skipper butterfly Astraptes fulgerator. *Proceedings of the National Academy of Sciences*, *USA* 101: 14812–14817.
- Matz, M.V., and R. Nielsen. 2005. A likelihood ratio test for species membership based on DNA sequence data. *Philosophical Transactions of the Royal Society B* 360: 1969–1974.
- Sarkar, I.N. 2009. Biodiversity Informatics: the emergence of a field. *BMC Bioinformatics*. 10: S1
- Shaw, J. and R.L. Small. 2004. Addressing the "hardest puzzle in American pomology:" Phylogeny of *Prunus* section *Prunocerasus* (Rosaceae) based on seven noncoding chloroplast DNA regions. *American Journal of Botany* 91: 985-996.
- Shaw, J. and R.L. Small. 2005. Chloroplast DNA phylogeny and phylogeography of the North American plums (*Prunus* subgenus *Prunus* section *Prunocerasus*; Rosaceae). *American Journal of Botany* 92: 2011-2030.
- Shaw, J., E.B. Lickey, J.T. Beck, S.B. Farmer, W. Liu, J. Miller, K.C. Siripun, C.T. Winder, E.E. Schilling, and R.L. Small. 2005. The tortoise and the hare II: relative utility of 21 noncoding chloroplast DNA sequences for phylogenetic analysis. *American Journal of Botany* 92: 142–166.
- Shaw, J., E.B. Lickey, E.E. Schilling, and R.L. Small. 2007. Comparison of whole chloroplast genome sequences to choose noncoding regions for phylogenetic studies in angiosperms: The tortoise and the hare III. *American Journal of Botany* 94: 275–288.
- Taberlet, P., E. Coissac, F. Pompanon, L. Gielly, C. Miquel, A. Valentini, T. Veramt, G. Corthier, C. Brochmann, and E. Willerslev. 2006. Power and limitations of the chloroplast *trnL* (UAA) intron for plant DNA barcoding. *Nucleic Acids Research* 35: e14.
- Zhang, A.B., L.J. He, R.H. Crozier, C. Muster, and C.-D. Zhu. 2010. Estimating sample sizes for DNA barcoding. *Molecular Phylogenetics and Evolution* 54:2010 1035-1039.

APPENDIX A

GENBANK ACCESSION NUMBERS FOR SEVEN ANGIOSPERM LINEAGES USED IN THIS STUDY

Table 7. Lineages from Shaw et al. (2005; 2007) used in this study, cpDNA regions tested and GenBank Accession Numbers. For voucher locations see Shaw et al. (2005). OG = Outgroup

APPENDIX B

GENETIC VARIABILITY DATA FOR MATK AND RBCL ACROSS SEVEN ANGIOSPERM LINEAGES USED IN THIS STUDY

Table 8. Quantitative data collected for *matK* and *rbcL* in this study (see Shaw et al., 2005; 2007 for raw data for the 34 noncoding cpDNA regions). Each cell (cpDNA region/threespecies survey) contains data regarding: the aligned length of the three-species surveyed; the number of indels (between the ingroup taxa and the ingroup and the outgroup taxon); the number of nucleotide substitutions (between the ingroup taxa and the ingroup and the outgroup taxon); $PICs = total indels + nucleotide substitutions + inversions; the$ normalized PIC value; and the percent variability. Appendix abbreviations: L. = aligned length of three-species group, Subst. $=$ nucleotide substitutions, PIC $=$ potentially informative character. $bp = base pairs$.

APPENDIX C

PRUNUS L. ACCESSIONS USED IN THIS STUDY

Table 9. Taxa used in assessment of species identification analyses. cpDNA regions tested, source, voucher and GenBank accession numbers provided. Source and voucher information could not be found at this time for a handful of samples and have been left blank. In some cases, pieces of source and voucher information were known and have been placed in the table but bolded to show that they are not complete.

Species	Source & Voucher	$psbA-trnH$	trnL Intron	$trnL-trnF$	$trnS-trnG-trnG$	matK	rbcL
Maddenia hypoleuca396 Koehne	Wen8062	HQ188700	HQ243708	HQ243945	HQ244182	HQ235063	HQ235345
Physocarpus opulifolius084 (L.) Maxim.	JSh1015; TENN	AY500637	HQ243944	HQ244181	AY500742	HQ235318	HQ235600
Prunus africana202 (Hook. f) Kalkman	T. Eriksson 1010	HQ188701	HQ243709	HQ243946	HQ244183	HQ235064	HQ235346
Prunus africana217	Q. Luke 11470	HQ188702	HQ243710	HQ243947	HQ244184	HQ235065	HQ235347
Prunus africana260	DPRU 2557.1	HQ188703	HQ243711	HQ243948	HQ244185	HQ235066	HQ235348
Prunus africana390	DPRU 2557.2	HQ188704	HQ243712	HQ243949	HQ244186	HQ235067	HQ235349
Prunus alleghaniensis Porter var. davisii005 Wight	G. Schmidt; MI: TENN	AY500606	AY500754,	AY500773	AY500711	HQ235072	HQ235351
Prunus alleghaniensis001 Porter	JSh834	AY500607	AY500755,	AY500774	AY500717	HQ235069	HQ235352
Prunus alleghaniensis006	JSh837	HQ188706	AY500755	AY500774	AY500712	HQ235070	HQ235353
Prunus alleghaniensis373		HQ188707	HQ243715	HQ243952	HQ244189	HQ235071	HQ235354
Prunus americana Marshall var. lanata047 Sudw.	J. Beck 49955; TN: TENN	AY500596	AY500744	AY500763	AY500701	HQ235074	HQ235355
Prunus americana038 Marshall	JSh038, TN: TENN	AY500595	AY500743	AY500762	AY500700	HQ235073	HQ235356
Prunus amplifolia213 Pilg.	PPM- 3489; Santa Cruz, Bolivia	HQ188708	HQ243716	HQ243953	HQ244190	HQ235075	HQ235357
Prunus amygdalus073 Batsch	DPRU 1463.5: TENN	AY500625	HQ243717	HQ243954	AY500730	HQ235076	HQ235358
Prunus amygdalus277	DPRU 2330.5	HQ188709	HQ243718	HQ243955	HQ244191	HQ235077	HQ235359
Prunus amygdalus283	DPRU 2434	HQ188710	HQ243719	HQ243956	HQ244192	HQ235078	HQ235360
Prunus amygdalus292	DPRU 2406.14	HQ188711	HQ243720	HQ243957	HQ244193	HQ235079	HQ235361
Prunus amygdalus296	DPRU 1128	HQ188712	HQ243721	HQ243958	HQ244194	HQ235080	HQ235362
Prunus amygdalus297	DPRU 1463.4	HQ188713	HQ243722	HQ243959	HQ244195	HQ235081	HQ235363

Table 9 Continued

Prunus wilsonii428 (Diels ex Schneid.) Koehne	Wen 9344	HO188905	HO243939	HO244176	HO244390	HO235313	HQ235595
Prunus zippeliana2025 Miq.	Wen 10889	HO188906	HO243940	HO244177	HO244391	HO235314	HO235596
Prunus zippeliana2026	Wen 10902	HO188907	HO243941	HO244178	HO244392	HO235315	HQ235597
Prunus zippeliana2027	Wen 10914	HO188908	HO243942	HO244179	HO244393	HO235316	HO235598
Prunus zippeliana2031	Wen 10583	HO188909	HO243943	HO244180	HO244394	HO235317	HO235599

APPENDIX D

OUTLINE AND EXPLAINATION OF ELECTRONIC FILES FOR THIS STUDY

Due to the size and scope of this thesis, a number of files and folders were created to organize the data. Below is an outline of all the electronic files created and used for this thesis. The layout of this appendix is arranged in a hierarchical manner, which mirrors how the files are arranged on the computer. In preparing this appendix there are several broad pieces of information that need to be understood in order for this appendix to be used correctly.

The original *Prunus* dataset consisted of 256 samples however, in order to measure species identification success, only those taxa with two or more samples could be used. This resulted in species without a conspecific pair being removed from the dataset. This left 203 samples of *Prunus*; *Physocarpus opulifolius* was included as an outgroup taxon for Bayesian analyses. It is also present in many of the Uncorrected pairwise distance (UpD) Excel files since the same file created for Bayesian analyses could be used to generate genetic distances in PAUP. This reduced the amount of time spent on creating matrices for Bayesian and UpD analyses. Datasets that have been reduced to 203 or 204 taxa contain 'DeleTaxa' within their file name. There are no analyses files for matrices containing 256 samples. Originally, both *trnLLF* and *trnSGG* were considered single gene regions, respectively, but for publication purposes both were broken apart to accurately measure species identification, i.e. *trnL* intron, *trnL-F* IGS, *trnG* Intron, and *trnG-S* IGS. This separation was only done for the deleted taxa files.

In many cases, there are redundant matrix files owing to the need to create an initial matrix and then create a new file with Bayes block added to the end using Mesquite. These files are the same files with the exception of a Bayes block added to the end and contain 'MrB' within their file name. Importantly, all MrBayes output files contain the original matrix uploaded to Parallel MrBayes Online. This ensures that one can always find the correct file to run future analyses or double-check results. All files have been named using a number and lettering shorthand (ex $m \alpha t K = A2$) and the gene regions being tested in a particular matrix, like A2, can be found in the file 'DNABC_Analyses Combinations.xls'.

All files are contained in the master file labeled 'Ian Cohen Final DNABC Files'. This folder is located on the hardrive of the Macintosh computer located by the window in the Plant Lab (Holt 115). Within this master file, there are nine subfolders and two word documents. Below is a detailed explanation of the files that are contained within each of these subfolders to help a user navigate through each folder and find the information that they may need.

0. Appendix III—This Microsoft Word file contains a list and explanation of all the files that were created and used for this thesis. It is the electronic version of this document.

1. All Barcode Papers—This folder contains 112 peer-reviewed papers on DNA barcoding, chloroplast genome evolution, and *Prunus* L. (Rosaceae). Not all papers were referenced in this thesis.

2. CBOLInfo&Data—This folder contains four files that correspond to PCR reaction conditions, Plant Working Group taxa used, and data analysis recommendations from CBOL and the PWG.

3. PowerPoints—This folder contains all Microsoft PowerPoint files created for DNABC talks at academic conferences.

4. SEQS-GENBANK—This folder contains seven MacClade files with all the samples that were submitted to GenBank for this thesis project. Files labeled CUT had the front end trimmed before submission to GenBank so that all sequences started at the same nucleotide position.

5. Sequencher Files—This folder contains two subfolders that are organized by the taxonomic scope of the thesis.

- Prunus Sequencher Files—Contains one Excel file, 'A. Prunus-SequencherChecklist.xls,' that corresponds to the samples sent to Macrogen for sequencing. Samples sent to Macrogen were labeled with numbers and in the Excel file, the species name and corresponding number are present to match to the Macrogen files (ex. 133 refers to *Prunus bucharica (Korsh.) Hand.-Mazz.*). There are five subfolders that correspond to the sequencher files, and unaligned MacClade files for each gene region tested. The subfolder 'RAW Prunus' Sequences' contains seven subfolders that contain all the original sequences obtained from the Molecular Biological Research Facility, Knoxville, TN or Macrogen Inc., Seoul, Korea. Files are separated by gene region, sequences from samples that did not belong to the genus *Prunus*, and non sequencher files from Macrogen (ex. PDF tracer files or phrd files).
- T&H_DNABC Files—Contains all sequences, sequencher files, and unaligned MacClade files for *matK* and *rbcL*, which were added to the original *Tortoise and the Hare* dataset to compare the utility of these two regions to that of 34 noncoding cpDNA regions.

6. Single_Locus_Aligned_Matrices—Contains the single locus matrices for each gene region tested using samples from the genus *Prunus*. Matrices have been aligned and are either coded using Fast Gap (FG) or not coded at all.

- 204 Samples
	- Singlesw/ofg—This subfolder contains nine files representing the seven single locus gene regions plus *trnLLF* and *trnSGG* concatenated; Gaps are uncoded, hence the file name includes " w/ofg " = without Fast Gap coding.
	- Singlesw/fg— This subfolder contains nine files representing the seven single locus gene regions plus trnLLF and trnGGS concatenated; Gaps have been coded using Fast Gap Program, hence the file name includes " wfg " = with Fast Gap coding. Note: *rbcL* does not contain any gaps but to make it easier for future users to find the correct data file it was placed in this folder.
- 256 Samples
	- Singlesw/ofg—This subfolder contains five files representing each of the gene regions tested. In this folder, *trnLLF* and *trnGGS* were not separated into the *trnL* intron, *trnL-F* IGS, *trnG* Intron, or *trnG-S* IGS since there were no analyses done for the thesis or publication using the 256 dataset; Gaps are uncoded.
	- Singlesw/fg—This subfolder contains four files representing four of the five single locus gene regions. *RbcL* was excluded because it does not contain any gaps to code. This file can be found in the above subfolder. In this folder, *trnLLF* and *trnGGS* were not separated into *trnL* intron, *trnL-F*

IGS, *trnG* Intron, or *trnG-S* IGS since there were no analyses done for the thesis or publication using the 256 dataset; Gaps are uncoded ($FG = fast$) gap coding.; Gaps are coded using Fast Gap Program. Note: *rbcL* does not contain any gaps but to make it easier for future users to find the correct file that I used for my thesis work it was placed in this folder.

7. Prunus Matrix combinations based on 204 samples—This folder contains one Excel file that shows which gene regions are specified in each matrix (ex. A3 corresponds to *rbcL*). There are also six numbered subfolders. This represents all the possible combinations of gene regions. Listed below is the number of files in each folder. There are 127 matrices, which represents all the possible gene region combinations. These files are a separated by the number of regions concatenated (e.g., The subfolder labeled 2 contains matrices with two gene regions combined, like *rbcL*+*matK*).

- 1: Seven files are contained in this folder.
- 2: 21 files are contained in this folder.
- 3: 35 files are contained in this folder.
- 4: 35 files are contained in this folder.
- 5: 21 files are contained in this folder.
- 6&7: Eight files are contained in this folder.

8. Prunus Analyses_Files—These are the data files used to determine species identification success for each gene region alone and in combination. The use of ‗DeleTaxa' refers to files with 203 samples or 204 samples depending on if *Physocarpus opulifolius* was included as an outgroup taxon for Bayesian analyses. In some distance matrices, it was not included.

- A. DNABC_Analyses Combinations.xls—This file is a list of all 127 possible combinations. Files in the subfolders below are labeled based on position in chart $(e.g., matK = A2).$
- B. Coded, Fused, and Bayes Block Matrices—This subfolder contains six folders. Each folder contains the coded and concatenated gene regions for all of the possible gene region combinations (ex. 6 equals 6 gene regions Refer to DNABC Analyses Combinations.xls file for which gene regions are contained in each file.
- C. Multiphyl_DeleTaxa_MrBayes—This subfolder contains the files that were created using the online program Multiphyl to determine which evolutionary model to use for MrBayes. In all, there are nine files representing the seven single locus gene regions plus *trnLLF* and *trnSGG* concatenated, respectively).
- D. MrBayes Output Files—This subfolder contains six folders. Each folder contains the coded and concatenated gene regions for all of the possible gene region combinations (ex. 6 equals 6 gene regions combined) and Parallel MrBayes output for Bayesian analysis portion of study. Refer to DNABC_Analyses Combinations.xls file for which gene regions are contained in each file.
- E. UpD_Excel_Files—This subfolder contains six folders. Each folder contains the coded and concatenated gene regions for all of the possible gene region combinations (ex. 6 equals 6 gene regions combined). Each folder contains Excel files with genetic distances that were calculated using PAUP for the genetic

distance analysis portion of study. Refer to DNABC_Analyses Combinations.xls file for which gene regions are contained in each file.

9. DNABC_Tables&Figures—This folder contains five subfolders corresponding to tables created for the thesis and publication.

- Bayes Trees—This subfolder contains six Bayesian Tree graphics produced in FigTree v1.3.1 in PDF form.
- MiscTables—This subfolder contains two Excel files, Prunus spp. ID_Checks.xls and PrunusTable1MAY-2010.xls. The first file contains a list of *Prunus* samples with questionable identifications based on DNA barcoding results. The second file contains key information on all of the *Prunus* samples used in this study (ex. Collector, collection number, UTC ID).
- Shaw Files—This subfolder contains the original *Tortoise and the Hare* tables. These tables were not altered in any way but were saved as different versions and subsequently expanded to add *matK* and *rbcL* data. Only the file 'Shaw et al TH3 Appendix S2.xls' was used in my study.
- Tables_For_DNABC_Pub—This subfolder contains three tables to be used for the publishable unit.
- Thesis Tables—This subfolder contains 10 Excel files that were used in this thesis. All files are up to date and correct.

10. IC_Thesis_Rewrite_Figs1-18_JSh.doc—This is the completed thesis file

11. Joey'sPrunusMacCladeFiles—This subfolder contains eight MacClade files from Joey's PhD work. Files were used to determine which taxa sequence data had already been obtained for.

APPENDIX E

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EDUCATION *University of Tennessee at Chattanooga*, *Chattanooga*, *TN* **M.S. Environmental Science** *Plant DNA barcoding: Testing the utility of the Consortium for the barcoding of Life's Two 'Agreed Upon' Loci* **2008-2010** *University of Tennessee at Chattanooga*, *Chattanooga*, *TN* **B.S. Biology** *Cum Laude* **2008** AWARDS & HONORS Sigma Xi Graduate Student Research Award Association of Southeastern Biologists Graduate Student Support Award University of Tennessee at Chattanooga Graduate Student Association Travel Award UTC Provost Student Research Grant Association of Southeastern Biologists Graduate Student Support Award University of Tennessee at Chattanooga Graduate Student Association Travel Award National Science Foundation- REU Internship UTC Cooperative Education & International Programs Travel Award Geospatial Information & Technology Association Grant **2010 2010 2010 2009 2009 2009 2007 2007 2006** CLASSROOM EXPERIENCE **Adjunct Faculty** Instructor for Introductory Biology I lab in the Dept. of Biological and Environmental Sciences at the University of Tennessee at Chattanooga. Responsible for developing lesson plans for the various lab activities each week, as well as, quizzes and exams. **Fall 2009 Spring 2010 Teaching Assistant** Teaching assistant for General Chemistry I and II labs at the University of Tennessee at Chattanooga. Assisted students with experiments and ensured they maintained and followed safe working conditions in lab; also responsible for cleaning up and restocking materials used by students. Worked under, lab instructors Tracy Bingham and C.K. Reynolds. **Fall 2005 Fall 2006** RESEARCH EXPERIENCE *University of Tennessee at Chattanooga*, *Chattanooga*, *TN* **Thesis** *Plant DNA barcoding: Testing the utility of the Consortium for the barcoding of Life's Two 'Agreed Upon' Loci*. Worked with Dr. Joey Shaw on determining the efficacy of DNA barcoding. Generated data **2008-2010**

for three noncoding chloroplast regions (*trnS-trnG* IGS, *trnG* intron, *trnL-trnF* IGS, *trnL* Intron, and *psbA-trnH* IGS), and portions of two

coding chloroplast regions (*matK* and *rbcL*)

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