Supplementary Methods

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Customizing the training sets

4 The format for the training files are described in the original RDP classifier 5 version 2.5 sample data folder that comes with the distribution available from 6 http://sourceforge.net/projects/rdp-classifier/ (Wang et al. 2007). For each of our training 7 sets, the two files we used to train the classifier are provided so that they can be used to 8 train the classifier or be modified to suit a user's needs if necessary. Using our files, the 9 classifier can be trained to the genus or family rank using GenBank taxonomy and once 10 the assignment is made to the genus/family rank the associated lineage (family, order, class, phylum, kingdom) is retrieved. As insect taxonomy develops, or users' preferred 12 taxonomic scheme varies, changes to the associated taxonomic lineage for any sequence 13 in the training file can be accommodated. Taxonomic changes can be made directly to 14 the all of the affected Fasta headers in the sequence Fasta file. Sequence additions or 15 deletions can also be made directly to the sequence Fasta file. We provide a Perl script in 16 the supplementary material that can then be used to generate the corresponding taxonomy 17 file used to train the classifier to accommodate the taxonomic changes. Instructions for 18 re-training the classifier are clearly described in the original RDP classifier version 2.5 19 distribution in the sample data folder 'README' file. For simplicity, the only 20 taxonomic ranks supported are kingdom, phylum, class, order, family, and genus for sets trained to the genus rank; and kingdom, phylum, class, order, and family for sets trained 22 to the family rank.

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Supplementary Discussion

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Sequence annotation in publically available sequences

The impact of insufficiently identified and mis-annotated sequences that may result in the reduced quality of sequences in public databases has been previously evaluated (Nilsson et al. 2006; Bidartondo et al. 2008; Costa et al. 2012; Kwong et al. 2012). The result is that many insufficiently identified sequences cannot be utilized for high-resolution taxonomic assignments to species rank. As previously noted, many iBOL sequences are not present in GenBank (Kwong et al. 2012) and many COI sequences in GenBank do not contain the 'barcode' keyword. The result is that our GenBank-barcode training set was the smallest and least representative set. It was useful here for comparison purposes, but should not be used for making taxonomic assignments. Additionally, the iBOL data release package 3.75 v1 does not provide corresponding GenBank numbers so cross-referencing sequences from each database can be problematic. Although taxonomic assignments for insects in the iBOL data release package are limited to the order rank, taxonomic assignments to BINs (barcode index numbers) are provided (Ratnasingham & Hebert 2013). For instances where the BIN unit is not sufficient or appropriate, we have shown that in some cases our insect classifier may be able to refine taxonomic assignments to lower (more specific) ranks.

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Single marker bias

Problems using a single arbitrary marker, in this case COI, largely focus on its use as the sole piece of evidence for species delimitation and new species discovery (Cognato

2006). The 'barcoding gap' refers to a separation between the within species and among species sequence variability for COI sequences (Meyer & Paulay 2005; Meier et al. 2008). It has been suggested that this barcoding gap, sometimes used to explain the inability of a barcode assignment method to correctly assign unknown queries, may actually be an artifact of insufficient sampling (Wiemers & Fiedler 2007). We recognize these issues, but agree with the view that a DNA barcode can still be used as a tool to assign unknown specimens to predefined groups (Nielsen & Matz 2006), in this case, genus or family level taxa defined by NCBI taxonomy. Considering that the classifier was trained only to the genus or the family ranks, and that assignments can be summarized to even more inclusive ranks as needed to obtain a minimum level of bootstrap support, taxonomic assignments made with this tool are relatively conservative compared with a BLAST top hit approach (a commonly used method in environmental sequencing studies). Use of this method in conjunction with other taxonomic assignment methods (such as phylogeny, indicator vector, segregating sites, LCA, or coalescent based approaches), and other genes when available, are not mutually exclusive and is encouraged when there is a need to visualize, corroborate, or test taxonomic assignments flagged to be of special interest (Munch et al. 2008a,b; Sirovich et al. 2009, 2010; Lou & Golding 2010; Yang & Rannala 2010; Huson et al. 2011).

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Choosing meaningful bootstrap support cutoffs

It has been recently recognized that choosing a suitable bootstrap support cutoff for taxonomic assignments will vary with query length and assignment rank (Mizrahi-Man *et al.* 2013). The most commonly used cutoff is 80% bootstrap proportion for the

Ribosomal Database Project naïve Bayesian classifier with the 16S training sets (Wang *et al.* 2007). Alternatively, a 50% cutoff has been recommended for 16S query sequences shorter than 250 bp (Liu *et al.* 2008; Claesson *et al.* 2009). In this study, we used LOOCV testing to propose bootstrap support cutoffs appropriate for each trained version of the classifier for a variety of query sequence lengths and taxonomic ranks. Note that these values worked well for query sequences known to be present in the reference dataset. We have shown that unknown insect queries from tropical field studies and poorly represented insect orders remain problematic and may require bootstrap support cutoffs higher than those listed in Table 3 (and Tables S5 and S6) to avoid making incorrect taxonomic assignments.

Impact of sequence error on taxonomic assignments

Previous work has also addressed the impact of very low frequency variants due to sequencing error on the taxonomic assignment of COI barcode sequences (Stoeckle & Kerr 2012). A previous study has suggested that the naïve Bayesian classifier that uses kmer frequencies for classification may be particularly sensitive to sequence error compared with BLAST that uses local sequence alignments (Porter & Golding 2012). This makes sequence quality filtering of sequences especially important prior to using this tool.

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141 **Supporting Figure Legends** 142 143 Figure S1 Abundance of taxonomic orders in the iBOL data release package 3.75 – v1 144 (N=86,306). Each insect order is followed by the number of sequences represented in the 145 iBOL dataset. 146 147 Figure S2 Proportion of correctly classified queries during 'leave one out cross 148 validation' (LOOCV) testing of the GenBank-barcode trained classifier. No bootstrap 149 support cutoff was used to filter results. 150 151 Figure S3 Proportion of correctly classified queries during 'leave one out cross 152 validation' (LOOCV) testing of the GenBank-family trained classifier. No bootstrap 153 support cutoff was used to filter results. 154 155 Figure S4 Country of origin for sequences in the iBOL data release package 3.75 – v1. 156 The top ten most abundant countries are shown for a) Lepidoptera (N=8,647) and b) 157 Diptera (N=46,233). Each country is followed by the number of sequences represented in 158 the iBOL dataset.

Figure S1

Insect taxonomic orders represented in iBOL data release 3.75 v1

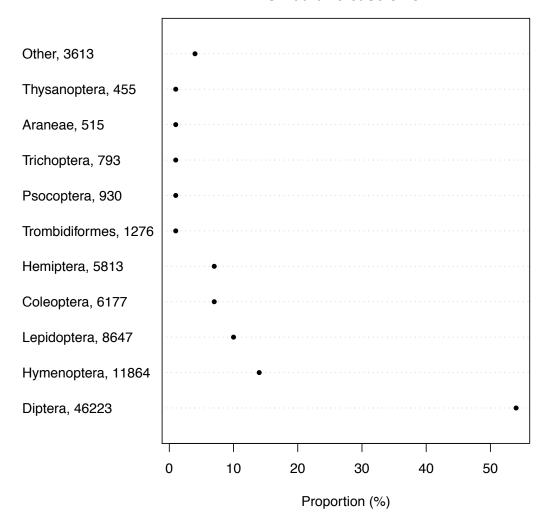


Figure S2

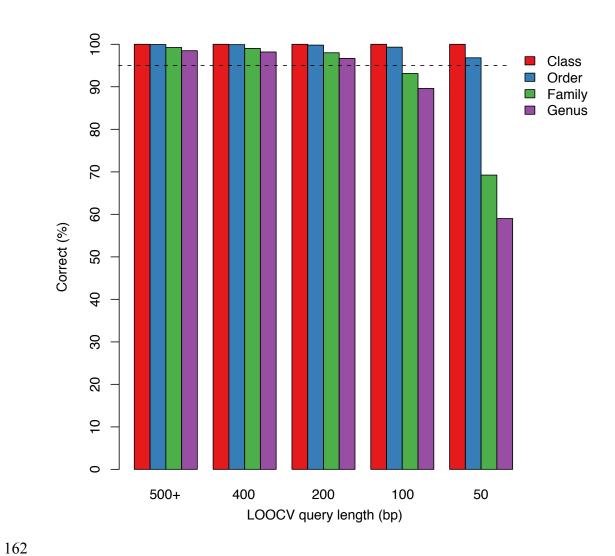


Figure S3

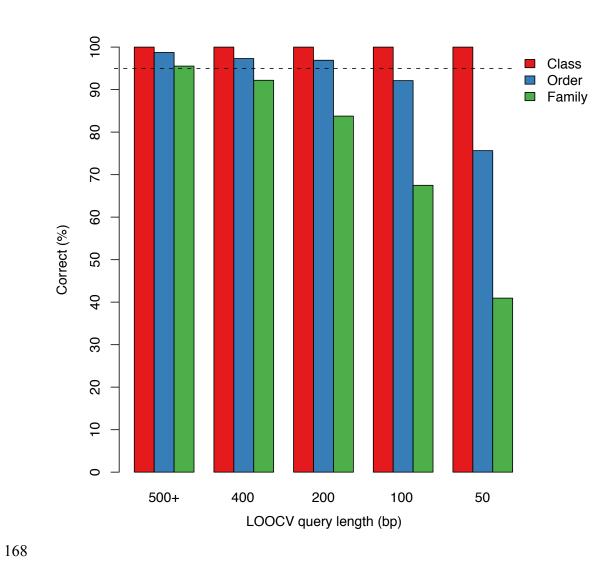
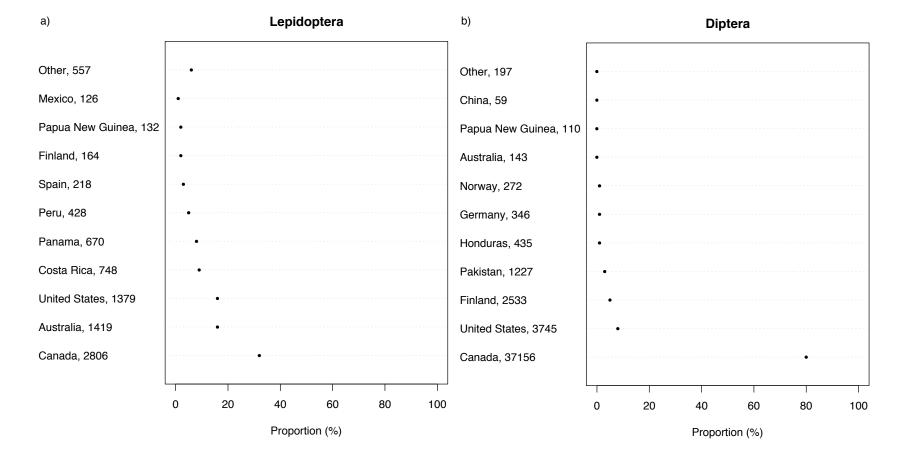


Figure S4



172 Table S1 – List of Mantodea genera included during testing and their GenBank

accessions.

Family	Genus	GenBank Accession	
Acanthopidae	Raptrix	FJ802836	
	Cliomantis	FJ802777	
Amorphoscelidae	Gyromantis	FJ802776	
	Paraoxypilus	FJ802755	
Empusidae	Empusa	FJ802806	
	Acromantis	FJ802908	
		FJ802783	
		FJ802888	
	Amorphoscelis	FJ802906	
		FJ802821	
		FJ802843	
	Ceratomantis	FJ802896	
Hymenopodidae	Creobroter	FJ802787	
	Ephippiomantis	FJ802899	
	Euantissa	FJ802834	
	Hestiasula	FJ802909	
		FJ802815	
	Oxypilus	FJ802845	
	Oxyphus	FJ802862	
		FJ802819	
	Bolbena	FJ802852	
	Hapalomantis	FJ802858	
	Hapalopeza	FJ802886	
Iridopterygidae	Miromantis	FJ802907	
muopterygidae		FJ802910	
	Tropidomantis	FJ802838	
		FJ802792	
	Xanthomantis	FJ802904	
Liturgusidae	Liturgusa	FJ802797	
Liturgusidae	Stenomantis	FJ802770	
	Bolbe	FJ802771	
	Bolbe	FJ802759	
	Coptopteryx	FJ802778	
Mantidae		FJ802917	
	Deroplatys	FJ802918	
		FJ802895	
	Elmantis FJ802853		

Family	Genus	GenBank Accession
		FJ802848
	Entella	FJ802824
	Euchomenella	FJ802786
	Gimantis	FJ802901
	Gonypeta	FJ802894
	Gonypetyllis	FJ802785
	Ligaria	FJ802809
	Ligariella	FJ802772
	Mantis	FJ802846
		FJ802844
	Miomantis	FJ802829
	Miomantis	FJ802823
		FJ802831
	Myrcinus	FJ802851
	Orthoderella	FJ802804
	Parastagmatoptera	FJ802762
	Phyllovates	FJ802839
	Polyspilota	FJ802847
	Pseudomantis	FJ802884
	Rhombodera	FJ802913
	Sphodromantis	FJ802756
		FJ802856
		FJ802857
	Stagmomantis	FJ802761
		FJ802813
		FJ802760
		FJ802774
	Statilia	FJ802781
	Statilla	FJ802915
		FJ802849
	Tamolanica	NC_007702
		FJ802803
	Vates	FJ802798
		FJ802799
		FJ802822
Mantoididae	Mantoida	FJ802793
		FJ802794
Sibyllidae	Sibylla	FJ802808
Tarachodidae	Dysaules	FJ802835
i ai aciiodidae	Leptomantella	FJ802832

Family	Genus	GenBank Accession
	Pyrgomantis	FJ802825
	Tarachodes	FJ802842
Thespidae	Hoplocorypha	FJ802767
	Musoniella	FJ802800
	Oligonicella	FJ802775

Table S2 Number of unique taxa in three insect COI training sets. The number of species included in the training sets is shown although the classifier was only trained to the genus or family ranks.

		GenBank-	GenBank-
Rank	GenBank	barcode	family
Kingdom	1	1	1
Phylum	1	1	1
Class	1	1	1
Order	29	16	29
Family	637	256	718
Genus	8,679	3,565	10,242
Species	27,370	9,991	46,815*

^{*}includes insufficiently identified species assignments including the terms sp., aff., cf.,

181 etc.

Table S3 Number of taxa represented by a single sequence (singletons) in three insect COI training sets (and proportion of total taxa from Table S2 in parentheses).

		GenBank-	GenBank-
Rank	GenBank	barcode	family
Kingdom	0	0	0
Phylum	0	0	0
Class	0	0	0
Order	1	1	0
Family	76	46	90 (13%)
Genus	2808 (32%)	699 (20%)	N/A

187 N/A - not applicable since the GenBank-family training set was trained to the family

188 rank

Table S4 Proportion of sequences misclassified at the genus rank for each insect order after performing 'leave one out cross validation' (LOOCV) testing with the Genbank trained classifier.

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	GenBank				
Insect Orders	Number of queries tested	% Misclassified ¹			
Archaeognatha	84	0.0			
Blattodea	150	5.3			
Coleoptera	26,455	6.1			
Dermaptera	30	6.7			
Diptera	21,538	6.7			
Embioptera	24	0.0			
Ephemeroptera	3,281	1.6			
Grylloblattodea	0	-			
Hemiptera	10,230	8.0			
Hymenoptera	14,801	4.5			
Isoptera	386	2.1			
Lepidoptera	94,750	3.2			
Mantodea	29	27.6			
Mantophasmatodea	30	0.0			
Mecoptera	32	28.1			
Megaloptera	358	0.6			
Neuroptera	389	9.3			
Odonata	1,288	1.7			
Orthoptera	3,217	3.7			
Phasmatodea	649	4.5			
Phthiraptera	179	1.7			
Plecoptera	1,012	1.1			
Psocoptera	6	16.7			
Raphidioptera	2	0.0			
Siphonaptera	6	0.0			
Strepsiptera	5	40.0			
Thysanoptera	385	8.1			
Trichoptera	8,200	1.9			
undef_Insecta	9	22.2			

^{*}Taxonomic assignments were not filtered by any bootstrap support cutoff. Results for

singletons not shown.

Table S5 Bootstrap support cutoffs that result in at least 99% correctly classified queries during leave-one-out cross-validation of the GenBank-barcode trained insect COI classifier.

	% Bootstrap support cutoff required to obtain 99% correct classifications						
Rank	50 bp	100 bp	200 bp	400 bp	FULL (500 bp+)		
Genus	N/A	50	30	20	20		
Family	95	50	20	0	0		
Order	80	0	0	0	0		
	% Queries classified at the appropriate bootstrap support cutoff						
	(from above)						
Genus	0%	68%	95%	98%	99%		
Family	4%	79%	99%	~100%	~100%		
Order	85%	~100%	~100%	~100%	~100%		

^{*}Results for singletons are not summarized.

N/A = Not available.

Table S6 Bootstrap cutoffs that result in at least 99% correctly classified queries during leave-one-out cross-validation of the GenBank-family trained insect COI classifier.

% Bootstrap support cutoff required to obtain 99% correct classifications					ect				
Rank	50 bp 100 bp 200 bp 400 bp (500 bp+)								
Family	N/A	80	80	80	80				
Order	80	60	60	70	40				
	% Queries classified with the appropriate bootstrap support cutoff (from above)								
Family	0%	20%	47%	76%	89%				
Order	17%	74%	95%	96%	99%				

^{*}Results for singletons are not summarized.

N/A = Not available.

Table S7 Comparison of order rank taxonomic assignments using three versions of the classifier versus known taxonomic assignments based on morphological characters. Bootstrap support cutoffs were chosen to minimize the number of taxonomic assignments that did not match (Non-match category below) known taxonomic assignments for the classifier with the smallest training set (GenBank-barcode). This table is directly comparable to Table 4.

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		Naïve Bayesian classifier version:				
Dataset		GenBank	GenBank-barcode	GenBank-family		
Mantodea ¹						
	Match	79 (96%)	0 (0%)	81 (99%)		
	Non-match	0 (0%)	0 (0%)	0 (0%)		
	Not classified	3 (4%)	82 (100%)	1 (1%)		
Lepidoptera ²						
	Match	8,605 (~100%)	8,608 (~100%)	8,554 (99%)		
	Non-match	2 (~0%)	0 (0%)	4 (~0%)		
	Not classified	40 (~0%)	39 (~0%)	89 (1%)		
Diptera ³						
	Match	24,115 (52%)	22,125 (48%)	34,208 (74%)		
	Non-match	5 (~0%)	0 (0%)	6 (~0%)		
	Not classified	22,103 (48%)	24,098 (52%)	12,009 (26%)		
Malaise ⁴						
	Match	297 (31%)	225 (24%)	443 (47%)		
	Non-match	2 (~0%)	0 (0%)	2 (~0%)		
	Not classified	650 (68%)	724 (76%)	504 (53%)		

¹Mantodea (N=82, average length = 871 bp) with bootstrap cutoffs: 96%, 96%, 96% for

217 GenBank, GenBank-barcode, and GenBank-family trained classifiers

⁴Malaise (N=949, average length = 310 bp) with bootstrap cutoffs: 79%, 79%, 79%

²Lepidoptera (N=8,647, average length = 637 bp) with bootstrap cutoffs: 60%, 60%, 60%

³Didptera (N=46,223, average length = 627 bp) with bootstrap cutoffs: 83%, 83%, 83%

Table S8 Naïve Bayesian classifier automated taxonomic assignments of Mantodea sequences (N=82) verified to match known taxonomic assignments from the iBOL data release package 3.75 - v1 using bootstrap support cutoffs from Tables 3, S5, and S6.

	GenBank (80% cutoff)			GenBank-barcode (20% cutoff)		GenBank-family (80% cutoff)			
	Order	Family	Genus	Order	Family	Genus	Order	Family	Genus
Match	82 (100%)	61 (74%)	54 (66%)	0 (0%)	0 (0%)	0 (0%)	82 (100%)	42 (51%)	N/A
Non-match	0 (0%)	3 (4%)	4 (5%)	82 (100%)	82 (100%)	81 (99%)	0 (0%)	28 (34%)	N/A
Not classified	0 (0%)	18 (22%)	24 (29%)	0 (0%)	0 (0%)	1 (1%)	0 (0%)	12 (15%)	N/A

N/A = Not applicable.

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Table S9 Lepidoptera (N=8,647) and Diptera (N=46,233) sequences from iBOL data release package 3.75 - v1 that were originally taxonomically assigned to the order rank but are putatively refined to the family and genus ranks using the naïve Bayesian classifier using bootstrap support cutoffs from Table 3, S5, and S6.

	GenBank (80% cutoff)		GenBank-barcode (20% cutoff)		GenBank-family (80% cutoff)	
Lepidoptera	Family	Genus	Family	Genus	Family	Genus
	4,079 (47%)	3,511 (41%)	7,296 (84%)	5,284 (61%)	5,158 (60%)	N/A
Diptera						
	10.348 (22%)	7.748 (17%)	22.807 (49%)	19.546 (42%)	17,462 (38%)	N/A

N/A = Not applicable.