

## SHORT COMMUNICATION

## Selection of primers for optimal taxonomic classification of environmental 16S rRNA gene sequences

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Microbial community profiling using 16S rRNA gene sequences requires accurate taxonomy assignments. 'Universal' primers target conserved sequences and amplify sequences from many taxa, but they provide variable coverage of different environments, and regions of the rRNA gene differ in taxonomic informativeness—especially when high-throughput short-read sequencing technologies (for example, 454 and Illumina) are used. We introduce a new evaluation procedure that provides an improved measure of expected taxonomic precision when classifying environmental sequence reads from a given primer. Applying this measure to thousands of combinations of primers and read lengths, simulating single-ended and paired-end sequencing, reveals that these choices greatly affect taxonomic informativeness. The most informative sequence region may differ by environment, partly due to variable coverage of different environments in reference databases. Using our Rtax method of classifying paired-end reads, we found that paired-end sequencing provides substantial benefit in some environments including human gut, but not in others. Optimal primer choice for short reads totaling 96 nt provides 82-100% of the confident genus classifications available from longer reads. The ISME Journal (2012) 6, 1440-1444; doi:10.1038/ismej.2011.208; published online 12 January 2012 Subject Category: microbial population and community ecology

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Variation in 16S ribosomal gene sequences has been used since the mid-1980's to characterize microbial diversity (Stahl et al., 1984). Interest in sequencebased surveys of environmental microbes has exploded in recent years with the availability of sequencing technologies that produce ever-larger data sets at ever-decreasing cost; in particular, the Illumina platform is attractive because of throughput, despite its short reads (Sogin et al., 2006; Lazarevic et al., 2009; Claesson et al., 2010; Caporaso et al., 2011; Degnan and Ochman, 2012). Here, we examine the reliability of assignment of novel sequences to known taxa under thousands of simulated scenarios, varying primer choice, read length and environment.

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Previous work on taxonomic classification of environmental 16S rRNA gene sequences has focused on whether reference sequences matching given query share taxonomic annotations (Jonasson et al., 2002; Desantis et al., 2006; Sogin et al., 2006; Wang et al., 2007). Validations of taxonomic classifiers have typically compared a limited range of primers, read lengths and environments (Sundquist et al., 2007; Huse et al., 2008; Liu et al., 2008; Wu et al., 2008; Hamp et al., 2009). Reference databases contain many sequences derived from some environments and few associated with others, however (Supplementary Figure S1), leading to substantial variation in classification quality. In addition, the use of 'leave-one-out' cross-validation at the sequence level (Sundquist et al., 2007; Wang et al., 2007; Liu et al., 2008; Wu et al., 2008)—where a single sequence with a known annotation is held out from a reference database and classified using the remainder—is problematic: reference sequences matching held-out query sequences are likely to originate from the same sample, because natural environments contain 'microdiverse' clusters of closely related strains (Acinas et al., 2004).

We addressed these issues by simulating truncated reads from eight large environmental data

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sets of near-full-length 16S rRNA gene sequences extracted from GreenGenes (Supplementary Table S1), using pairs of 44 'universal' primers commonly found in the literature (Supplementary Tables S2 and S3). These were selected from an initial set of 94 primers by the criterion that each primer had to match at least 40% of the sequences in at least one of the chosen environmental samples. Single-end reads were tested from each primer with all viable amplification partners (794 combinations), and paired-end reads were tested using all 374 viable pairings of the 22 forward and 22 reverse primers. Simulations using 11 read lengths (32 nt, 48 nt, 64 nt, 80 nt, 96 nt, 112 nt, 128 nt, 260 nt, 400 nt, 800 nt and full-length), with the constraint that read length could not exceed amplicon length for each primer pair, produced 6617 single-end and 3061 paired-end datasets per environment.

Reference databases were constructed by holding out each entire study in turn from GreenGenes, clustering the remainder at 99% using UCLUST (Edgar, 2010), and selecting one representative sequence per cluster (see Supplementary Methods for details). Each query fragment was then matched against remaining representative sequences using USEARCH (Edgar, 2010), configured to penalize

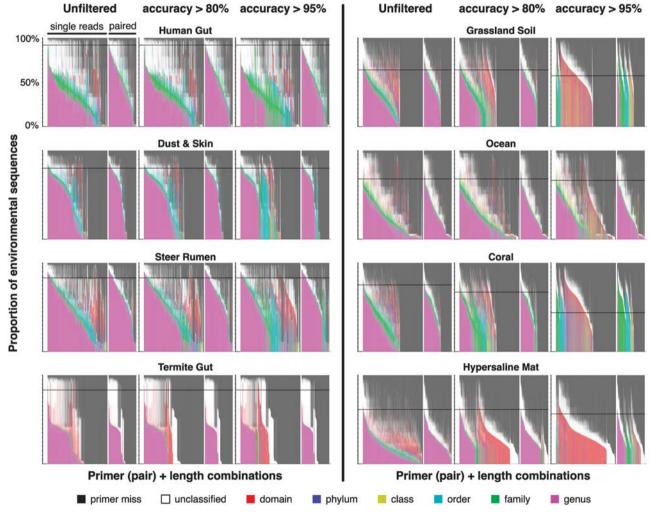


Figure 1 Classification performance, at three levels of estimated accuracy (Supplementary Methods), of 6617 possible choices of amplification primer, sequencing primer and read length for single-ended reads from different environments (left portion of each panel) and 3061 possible choices of primer pair and read length for paired-end reads (right portion). Combinations of primers and read lengths are sorted on the x axis according to a measure of overall classification performance (Supplementary Methods). Stacked bars show the proportion of non-chimeric, non-unique sequences from each sample—not the proportion of the total sample—that can be classified to each taxonomic level for each combination. See Supplementary Figure S1 and Supplementary Table S1 for the excluded proportion of novel (and thus a priori unclassifiable) sequences in each sample. The top of each colored section indicates how much of the sample can be classified to the given level or better. 'Primer miss' (black) indicates sequences that did not match a given primer and so would not be amplified. Classifications more specific than the genus level are exceedingly rare and so are not visible here. Horizontal lines indicate the maximum proportion of each sample classifiable to the genus level using 96 nt or less of sequence (i.e., with an optimal choice of primer or primer pair; see also Supplementary Tables S4 and S5), showing that short reads from the best primers frequently—but not always provide taxonomic information nearly matching that obtained from longer read lengths. Full-size versions of these panels are available in the supplementary data.

Table 1 Genus classification rates for optimal choices of primers, grouped by total read length

Total nucleotides	Read length	Forward primer	Reverse primer	Ocean	Coral	Hypersaline Mat	Grassland Soil	Steer Rumen	Human Gut	Dust & Skin	Termite Gut	
32 48	single 32	(E786F end)	E826R	0	5	2	5	<i>3</i> ovement	25	42	83	
	single 64	(end)	E533Ra	53	38	38	34	64	39	76	83	
64	pair 32	E969F	E1492R	17		24	01	74	90	78	- 00	
80	single 80	E341F	(E1406R E533Ra end)	63	52	58	49	60	41	66	45	
96	pair 48	E341F	E926Ra	68	55	50	55	81	48	73	64	
	single 96	E517F U515F	(end)	48	75	61	64	83	57	65	32	
	pair 48	E341F	E1064R	55	51	46	54	81	73	80	42	
	single 96	E341F	(end)	60	62	55	58	64	46	80	45	
	single 96	E343F U341F	(end)	55	59	51	47	75	48	80	46	
	pair 48	E517F U515F	E926Ra	52	51	56	57	65	40	70	45	
	pair 48	E969F	E1492R	16	77	16		76	92	65	40	1
112	single 112	E517F U515F	(end)	52 50	77 56	57 50	69	87 82	55 42	66	42	
	single 112 pair 64	(end) E341F	E926Ra E1406R	75	63	50 <b>56</b>	62 66	85	65	80 79	40 46	
128	pair 64	E517F U515F	E1406R	68	76	57	63	82	70	72	47	
	single 128	E341F	(E1406R E533Ra end)	58	60	62	63	83	57	73	71	percentage of sample classified to genus level
	pair 64	E517F U515F	E1407R	69	74	54	63	80	65	72	47	93,
	pair 64	E343F	E1406R U1406R	70	62	55	54	86	66	80	46	ň
	pair 64	E343F	E926Ra	59	60	49	57	87	68	86	42	ge
	single 128	E517F	(end)	47	76	58	70	85	51	62	43	<u>o</u>
	single 128	(end)	E1406R	43	68	57	72	78	42	85	44	S
	pair 64	U519F	E1406R	70	72	35	58	83	55	71	13	<u> </u>
	single 128	(end)	E357R	39	53	53	53	88	51	71	42	рle
	pair 64	E341F E343F	E1492R	39		32		78	92	82		<u>C</u>
	pair 64	E517F U515F	E1492R	36		34		78	94	78		ass
160	pair 80	E517F U515F	E1406R E1407R U1406R	73	76	63	74	92	82	80	45	iii
	pair 80	E341F	E1406R E1407R	75	64	57	73	90	80	86	46	be
	pair 80	E517F U515F	E926Ra	57	71	59	75	88	65	76	43	ō
	pair 80	E517F U515F	E1492R	39		36		79	95	81		ge
	pair 80	E341F E517F	E1492R E1407R	39 74	76	36 <b>59</b>	78	77 93	91 81	85 80	45	Ĕ
192	pair 96 pair 96	E517F E517F	E1407R E1406R	74	76 76	62	77	90	82	79	45	<u>s</u>
	pair 96	E517F	E926Ra	56	76	61	78	88	65	76	42	¥e
	pair 96	E341F	E926Ra	53	64	55	74	89	77	88	41	_
224	pair 112	E341F	E1406R E1407R U1406R	78	68	55	75	91	89	87	45	
	pair 112	E517F	E1406R	72	79	60	77	93	84	80	43	
	pair 112	E517F	E926Ra	56	77	63	77	89	63	74	42	<u>c</u>
	pair 112	U519F	E1407R	69	74	36	70	92	64	81	13	estimated classification accuracy:
256	pair 128	E517F U515F	E1406R U1406R	69	77	61	77	93	84	79	44	estimated ssification accuracy:
	pair 128	E341F	E1406R E1407R	75	65	53	73	92	80	85	47	nat ati
	pair 128	E517F U515F	E926Ra	55	76	64	74	90	62	76	43	ed Sy:
	pair 128	U519F	E1406R	69	73	37	69	93	65	79	13	
260	single 260	(end)	E1406R	41	76	54	71	91	57	86	44	ita noi <b>bo</b>
	single 260	(end)	E926Ra	37	72	62	69	88	61	84	41	<b>G:</b> //c:
400			=					ovement				italic: < 80% normal: 80% - 95% bold: >= 95%
520	pair 260	E517F U515F	E1406R	58	81	61	76 76	92	81	78	43	30% 80% - 19 <b>5%</b>
	pair 260	E341F	E1406R E1407R E926Ra	62	68	48	76	93	80	86	44	% ^ ^
	pair 260	E341F	E926Ha E1492R	50	70	60	74	89	64 90	83 <b>89</b>	43	95%
	pair 260 pair 400	E341F E341F	E1492R E1406R	36 61	71	33 56	73	80 92	79	87	43	%
800	pair 400	E517F U515F	E926Ra	43	78	62	72	90	67	70	43	
1600	pail 400	20171 00101	Lozona	-70	- 70	32		ovement		.0	72	
				1								į
			Maximum	78	81	64	78	93	95	89	83	

Thousands of combinations that produce suboptimal results are not shown (see Supplementary Methods). Cells are colored on a gradient from worst (red) to best (green) per column. Estimated classification accuracy (Supplementary Methods) is indicated by bold or italic font. Primers in parentheses are used in single-ended experiments for amplification but not sequencing. Primers appearing together perform equivalently; that is, for a given row, any choice among the given sequencing and amplification primers will produce the same result. 'End' indicates an end primer such as E8F, E1406R, U1406R, E1407R, E1492R or E1506R. Primer E1492R could not be tested in three datasets because sequences were not long and the contraction of the enough; the corresponding cells remain blank.

indels and mismatches equally. Clusters were then selected that matched within 0.5% identity of the best hit (hits <80% identity were disregarded). For paired-end query sequences, our Rtax procedure (Supplementary Methods) selected those reference clusters that matched both reads simultaneously with an average percent identity within 0.5% identity of the maximum. Taxonomic classifications were made at each level by retaining annotations agreeing among >50% of the clusters (including those with no annotation in the denominator); these generally extended at best to the genus level, because the reference database provides few species-level annotations.

Sequences from novel taxa (or sequences that appear novel due to sequencing error or chimerism) clearly cannot be correctly classified; however, such sequences may constitute a substantial proportion of a given sample (Supplementary Figure S1 and Supplementary Table S1). The version of Green-Genes that we used excluded taxa (defined by 97% identity) that were unique to a single sample, as one of the several strategies to remove chimeras. These unique sequences were therefore excluded from our query sets. Thus the classification rates we report represent the proportion of non-chimeric, nonunique sequences that can be classified to each rank. If an environmental sample is not similarly filtered before classification, then the classifiable proportion (that is, taken with respect to the total sample) will be correspondingly lower.

Classification rate and accuracy vary widely among environments and sequence regions, for several reasons: (1) the reference database provides different levels of coverage of each environment, (2) no primer is truly 'universal' and different primers (and pairs) hit different proportions of sequences in each environment and (3) the targeted regions are variably informative. Figure 1 shows proportions of sequences from each environment classified to each rank, for all 9678 single-ended and paired-end primer and read length combinations. Horizontal panels compare unfiltered results to classifications passing 80% and 95% estimated accuracy filters (see Supplementary Methods), showing that most classifications can be made with high accuracy when optimal primers are chosen. Remarkably, only 96 nt of sequence (taken as a single read or as a pair of 48 nt reads) can provide 82–100% of the 80% accurate genus classifications available from any read length (Supplementary Table S4). Paired-end sequencing can provide substantial gains in classification rate for some—but not all environments and read lengths. Paired-end classifications are typically more accurate than those made from single reads, and so are more likely to pass the 95% estimated accuracy filter (Supplementary Table S5). Another surprise is that hypervariable regions need not be specifically targeted, as there is no obvious relationship between taxonomic informativeness of a region and the

extent to which it overlaps any of the classical 'V-regions'.

No one combination of primers and read length works best in all environments, but near-optimal performance in six out of the eight environments is available using paired-end 80 nt reads from primers such as E517F, U515F or E341F paired with £1406R or closely related primers (Table 1). However, practical considerations such as ability to amplify low-biomass samples will sometimes influence which primers are used. For instance, short amplicons may be preferred because these are less subject to length heterogeneity biases and chimera formation. Similarly, short single-ended sequences are less subject to errors due to chimeras, simply because they are less likely to span a breakpoint. Classification performance for experimental choices matching such constraints can be found in the supplementary data.

The choice of reference database and taxonomy can have a dramatic impact on the resulting classification accuracy. In this study, we used the current GreenGenes taxonomy, which has been filtered to remove chimeras and where the taxonomic annotations are comprehensive and consistent with the phylogenetic tree (McDonald et al., 2011). Experiments using a previous version of the GreenGenes taxonomy lacking these features yielded far poorer accuracy (data not shown). In addition, bolstering areas of low coverage in reference databases will substantially improve classifier performance. For instance, taxa in the hypersaline mat, coral and grassland soil samples were underrepresented in the reference database (Supplementary Figure S1), and—presumably as a consequence—classifications of sequences from those samples were less likely to prove correct (Figure 1). Additional data sets from poorly sampled environments will also help to distinguish chimeric from legitimate but novel sequences.

In combination, these results indicate that taxonomic classifications of short reads—especially genus-level classifications—should be treated with skepticism, unless the specific combination of primer, read length, environmental source, reference database and assignment method has been thoroughly validated. At the same time, optimal choices of these parameters allow high classification rates and high accuracy. Thus, large-scale projects such as the Earth Microbiome Project (Gilbert et al., 2010), which aims to collect and analyze samples from tens of thousands of microbial habitats around the globe, may reasonably proceed with standardized primer choices and short reads.

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