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1	Comparative genomic analysis of 31 Phytophthora genomes reveal genome plasticity and horizontal	
2	gene transfer	
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44	Abstract			
45	Phytophthora species are oomycete plant pathogens that cause great economic and ecological impacts.			
46	The Phytophthora genus includes over 180 known species, infecting a wide range of plant hosts including			
47	crops, trees, and ornamentals. We sequenced 31 individual Phytophthora species genomes and 2			

48 individual transcriptomes to study genetic relationships across the genus. *De nov*o genome assemblies
49 revealed variation in genome sizes, numbers of predicted genes, and in repetitive element content across

50 the *Phytophthora* genus. A genus-wide comparison evaluated orthologous groups of genes. Predicted

51 effector gene counts varied across *Phytophthora* species by effector family, genome size, as well as plant

52 host range. Predicted numbers of apoplastic effectors increased as the host range of *Phytophthora* species

53 increased. Predicted numbers of cytoplasmic effectors also increased with host range but leveled off or 54 decreased in *Phytophthora* species that have enormous host ranges. With extensive sequencing across 55 the *Phytophthora* genus we now have the genomic resources to evaluate horizontal gene transfer events 56 across the oomycetes. Using a machine learning approach to identify horizontally transferred genes with 57 bacterial or fungal origin we identified 44 candidates over 36 Phytophthora species genomes. Phylogenetic 58 reconstruction indicates that the transfers of most of these 44 candidates happened in parallel to major 59 advances in the evolution of the oomycetes and Phytophthoras. We conclude that the 31 genomes 60 presented here are essential for investigating genus-wide genomic associations in Phytophthora.

61 Introduction

Members of the *Phytophthora* genus are oomycete plant pathogens that collectively infect a wide range of plants (1, 2) and cause great economic, environmental, and societal impact (3). Oomycetes are morphologically similar to filamentous fungi (4–7) but are classified as stramenopiles, a group that also includes diatoms and brown algae (4, 8). Oomycetes include many plant pathogenic species besides *Phytophthora* including numerous *Pythium* species that cause seed, seedling, root and fruit rots and a broad diversity of obligate biotrophs that cause downy mildew.

68 Phytophthora species infect numerous plants including crops, trees, and ornamentals in managed and 69 natural ecosystems. The agent responsible for potato blight, *Phytophthora infestans* triggered the 1840s 70 potato famine (9, 10), and *Phytophthora ramorum* is responsible for sudden oak death in North America 71 and sudden larch death in the UK, which have destroyed millions of trees, in addition to infecting hundreds 72 of additional tree and ornamental species (11-14). Some Phytophthora species are relatively host-specific, 73 such as the soybean pathogen Phytophthora sojae (15), the strawberry pathogen Phytophthora fragariae 74 (16), and the lychee pathogen Phytophthora litchi (17). In contrast, others including Phytophthora 75 cinnamomi (18), Phytophthora palmivora (1), and Phytophthora parasitica (1), can infect a vast assortment 76 of plant hosts. The mechanistic basis of this large variation in apparent host specificity is currently unknown 77 (10, 19).

There are over 180 known *Phytophthora* species (20, 21) phylogenetically falling into 12 or more clades (21–23) and further divided into numerous sub-clades (20). Previous genome sequencing studies have

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examined species within the *Phytophthora* genus (9, 19, 20, 22, 24–28) with *P. sojae* and *P. ramorum* (29), *P. infestans* (10, 30), and *Phythophthora capsici* (31) serving as key models for the genus. Overall, these
studies revealed highly dynamic genomes containing both rapidly evolving and conserved regions.

83 Phytophthora species infect plant hosts through the use of two broad classes of secreted effector proteins 84 (5, 32-35). Apoplastic effectors act outside the plant cells and include glycoside hydrolases, necrosis 85 inducing proteins (NLPs) (36), proteases, lipases, lipid-binding proteins, and protease inhibitors (37). Roles 86 of apoplastic effectors include weakening of plant physical and chemical defenses, and a source of nutrition 87 early in infection. In contrast, cytoplasmic effectors enter plant cells, often through the differentiation of 88 specialized structures called haustoria, and include RxLR effectors (29, 38, 39), crinkler effectors (CRN) 89 (40–44) and non-conventionally secreted effectors (45). In oomycetes, as well as other pathogens, 90 cytoplasmic effectors manipulate numerous aspects of host physiology and morphology to promote 91 susceptibility, including suppression of host immunity and programmed cell death (46, 47), and stimulating 92 and inhibiting the release of nutrients (5, 48-51). The genomes of oomycete pathogens sequenced to date 93 include large rapidly evolving gene families encoding these effectors (10, 29, 31). Many of these effector 94 genes, especially those encoding RxLR effectors, display evidence of accelerated evolution due to host-95 pathogen co-evolutionary conflict (38, 52, 53).

Examination of previously sequenced genomes has identified two distinct partitions namely gene-dense, repeat-poor regions and gene-sparse, repeat-rich regions (10, 29–31). Highly conserved housekeeping genes are typically found in gene-dense regions, while rapidly evolving gene families associated with infection are typically found in gene-sparse regions that are transposon-rich (10, 29–31, 54). This arrangement has been labeled "the two-speed genome" (19, 24). It has been hypothesized that transposons in the gene-sparse, transposon-rich regions may contribute to the genomic diversity and possibly to epigenetic variability of expression of genes in those regions including infection-associated genes (55).

To investigate phylogenetic relationships, horizontal gene transfer (HGT), effector genomics, and possible mechanisms underlying host ranges across members of the genus *Phytophthora*, we sequenced 31 genomes using Illumina short read technology. Our newly sequenced genomes include most species in clade 7 as well as representative species from nine of the phylogenetic clades (56). Several genomes have 107 already been published individually as a result of this project: *P. litchii* (25), *Phytophthora megakarya* and 108 *Phytophthora palmivora* (26), *Phytophthora fragariae* and *Phytophthora rubi* (27), and *Phytophthora* 109 *cactorum* (28, 57). Here we present a combined analysis of the genome sequences of 37 *Phytophthora* 110 species, including the 31 newly sequenced species by our sequencing consortium, resulting in the first 111 large scale comparative genomic study including species from nine *Phytophthora* clades. This work 112 provides insights into genome architecture and evolution in the genus *Phytophthora* as well as novel 113 genomic resources of broad interest.

114 **RESULTS**

115 Sequencing and assembly of the 31 Phytophthora genomes

116 The 31 Phytophthora genomes sequenced by our consortium produced between 16.6M and 44.7M raw reads per genome (Table 1 and Table S1). The sizes of the Phytophthora genome assemblies varied 117 greatly, ranging from 37.3 to 107.8 Mb (mean 61.3 ±17.6Mb). Due to variations in the numbers of reads 118 119 per genome and differences in genome sizes, the quality of assembled genomes varied as well. The 120 number of contigs per assembly ranged from 2,131 to 28,263 (mean 11380 ±7919.2). The more completely 121 assembled genomes benefited from deeper coverage. For example, the Phytophthora boehmeriae 122 assembly was composed of 2,866 contigs with an N_{50} of 41,917 bp; this assembly benefitted from 87X 123 coverage as a result of receiving 34.7M reads to cover a 39.7 Mb assembly. In contrast, the Phytophthora 124 lateralis assembly had only 45X coverage (23.2M reads across a 50.5 Mb assembly size), resulting in 125 28,263 contigs and an N_{50} of 2,396 bp (Table 1). Sequence read length did not seem to make a difference 126 in assembly quality. Two genomes (P. cactorum and P. idaei) with 250 bp reads (compared to most 127 assemblies with 50 bp reads) had an N₅₀ and number of contigs that were not better than the other genomes 128 when compared to genome size (*P. cactorum*: contigs 7,888, N₅₀ 15,053 size 56 Mb; *P. idaei*: contigs 7,163, 129 N₅₀ 14,461, size 53.5 Mb) (Table 1).

To aid with gene calling, and identify active genes, RNA sequencing was conducted for 25 *Phytophthora* species on V8-grown mycelia and either germinated cysts or Plich-grown mycelia for those species that did not readily yield zoospores (58). RNA sequencing produced 12.5M to 36.0M paired-end sequence reads per sample, with resulting transcriptome assemblies averaging 36,189 contigs per assembly with an average N_{50} of 1,822 bp. Most species' transcriptome assemblies had between 20K and 40K contigs, with the exception of *Phytophthora parvispora* that produced more than 109K contigs. TransDecoder, which reduces duplication in *de novo* transcriptome assemblies by identifying candidate coding regions and removing duplicates, was run on the Trinity assemblies and lowered average transcriptome assembly to 33,200 contigs and raised average N₅₀ to 2,753bp. Transcriptome size differences between species are not correlated with genome size differences with the two largest genome assemblies (*P. megakarya* and *P. palmivora*) both resulting in mid-range transcriptome assembly sizes (Table S1).

141 De novo repeat identification and analysis

Repeats were identified, classified, and masked to prepare genomes for gene prediction. *De novo* repeat prediction identified between 27 and 295 different repeat sub-families per species. The percentage repeat content of the assemblies varied greatly across the *Phytophthora* genomes sequenced. The genome assemblies of *Phytophthora kernoviae*, *P. litchii*, and *Phytophthora agathidicida* contained very low repeat content of 4.15%, 5.76%, and 5.98%, respectively (Fig. 1). On the other end, several genome assemblies had high repeat content: *P. megakarya*, *Phytophthora hibernalis*, and *Phytophthora pinifolia* contained 32.94%, 32.46, and 33.00%, respectively.

Repeat annotations were classified into types (Class I Retrotransposons, Class II DNA transposons, and other), families, and sub-families, (Fig. 1). Fifteen different DNA transposons were identified across all 31 species. Long terminal repeat (LTR) retrotransposons were more diverse across *Phytophthora* species. As many as 57 Copia LTR retrotransposons and 144 Gypsy LTR retrotransposon types were found in each genome assembly.

154 *Phytophthora* gene prediction and annotation

Gene predictions ranged from 12,391 to 37,283 (mean 19,943, ±5,864) across the newly sequenced genomes. These gene counts per species are within the general range of previously published *Phytophthora* genomes (*P. sojae*, *P. ramorum*, *P. infestans*, *P. capsici*, *P. cinnamomi*, and *P. parasitica*) ranging from 16,066 to 26,584 genes per species, considering that we observed higher gene counts in genomes with genome duplications (*P. palmivora* and *P. megakarya* (26)). Some of the previously published *Phytophthora* genome sequences were annotated with the MAKER gene prediction process outlined here to validate the methods. The *P. capsici* genome was reported to contain 19,805 predicted genes (31), while we obtained 18,917 predicted genes. The *P. sojae* genome v3.0 was reported to contain 26,584 predicted genes (29), whereas we identified 21,447 MAKER-predicted genes. Therefore, our MAKER pipeline may slightly undercount the gene content compared to other methods.

165 The completeness of the gene sets predicted from the genomes and transcriptomes was assessed by 166 identifying single-copy core orthologs using the BUSCO pipeline with the Alveolata Stramenopiles database (234 genes) as the reference. BUSCO analysis of the 31 genome assemblies identified 147-231 167 168 complete genes (mean 204 ±27) of the 234 single copy genes in the database (Fig. 2A). Analysis of the 169 24 transcriptome assemblies identified 18 to 228 complete genes (mean 187 ±54) (Fig. 2B). The P. 170 kernoviae and P. lateralis transcriptomes were outliers, with only 20 and 18 complete genes identified, 171 respectively. When P. kernoviae and P. lateralis were removed, transcriptome assemblies ranged from 172 142 to 228 complete (single and duplicated) genes (mean 203 ±18). Results of BUSCO analyses run on 173 predicted proteins ranged from 139-222 complete (single and duplicated) genes (mean 192, ±25) of the 234 conserved orthologous proteins in the database (Fig. 2C). 174

175 Predicted proteins from the MAKER gene prediction were functionally annotated by matching to published 176 Phytophthora, stramenopile, and fungal proteins. Across the 614,862 proteins predicted in the 31 177 Phytophthora species, when aligned to the NCBI and UniProt TrEMBL 294,146 Phytophthora proteins 178 produced alignments that passed the cutoff filter. Removing 'Uncharacterized Protein' or similarly 179 uninformative functional annotations yielded 196,652 proteins with functional classifications (mean 6,343.6 180 \pm 3,191.7). When aligned to the *Phytophthora*, stramenopile, and fungal sequence databases, 445,458 181 proteins passed the alignment cutoffs, with 304,951 proteins (mean 9,837.1 ± 2523.3) that had informative 182 functional annotations.

InterProScan was used to identify domains and motifs in all predicted proteins. Of the 614,862 proteins,
 173,727 had domains identified. GO terms were assigned from BLASTX alignments between the UniProt
 BLASTX alignments and the InterProScan predictions, identifying 321,953 proteins with GO terms
 assigned.

187 Effector protein identification in *Phytophthora*

The cytoplasmic effector identification process predicted a total of 10,354 RxLR effector proteins and 4,415 CRN effector proteins from the genomes and transcriptomes of the species sequenced in this study (Fig. 3). The numbers of predicted RxLR effector genes differed greatly across genomes. *P. pinifolia* exhibited the lowest number, with 46 predicted RxLR effectors, while *P. megakarya* exhibited the highest number, with 1,183 predicted proteins. We also observed great variation in the number of predicted CRN effectors, with *P. litchii* showing the lowest number of CRN effectors, 27, while *Phytophthora cajani* showed the highest, 274.

Our search for apoplastic effectors identified 6,671 glycoside hydrolases, 1,191 NLPs, and 1,046 protease inhibitors across the 31 *Phytophthora* genomes (Fig. 3). The predicted glycoside hydrolase genes range from 139 (*Phytophthora brassicae*) to 386 (*P. palmivora*). The numbers of NLP genes range from 5 (*P. pinifolia*) to 78 (*Phytophthora niederhauserii*). The counts of protease inhibitor genes range from 23 (*Phytophthora brassicae*) to 67 (*P. palmivora*).

200 Orthology clustering of *Phytophthora* proteins

The 715,980 predicted genes from the 31 genome assemblies along with those of *P. sojae, P. ramorum, P. infestans, P. capsici, P. cinnamomi,* and *P. parasitica* were subjected to orthology analysis. In the first step, the *Phytophthora* genes were matched against the pre-computed publicly available orthologous groups. From this analysis, 560,201 genes were assigned to 7,829 unique clusters, leaving 155,779 genes unassigned. In step two, these remaining genes were clustered using OrthoMCL yielding 13,474 additional unique clusters. In total, the 715,980 genes were assigned to 21,303 orthologous clusters.

Orthologous groups were assigned functional annotations based on the proteins that composed the group.
Of the 21,303 orthologous clusters, 9,806 could be assigned informative functional annotations as defined
in the gene annotation section.

The numbers of genes present in orthologous groups encompassing all 37 *Phytophthora* species shows a bimodal frequency distribution (Fig. 4), with peaks observed at 1 to 6 genes per group, and 34 to 41 genes per group. We hypothesize this first peak represents rapidly evolving genes that are conserved in only a few of the 37 *Phytophthora* species in this study. This first peak was much smaller when only genes with meaningful annotations were considered, suggesting an enrichment for genes that have previously uncharacterized functions. This peak included both small orthologous groups in which all genes were from the same *Phytophthora* clade, as well as groups consisting of genes from multiple clades. The second large peak was centered at 37 genes per ortholog group, thus representing ortholog groups that have one gene per species. Additional groups include one or a few species missing the orthologous gene or one or a few species with a second copy of an orthologous gene.

220 Phylogenetic relationship across Phytophthora species

We reconstructed the phylogenetic relationships of the sequenced species of the genus *Phytophthora* (Fig. 5) based on 61 single-copy core orthologous genes shared across 37 species. Predicted genes and amino acid protein sequences used to build the phylogeny are found in Table S2. The RAxML phylogenetic tree clustered these species into phylogenetic clades consistent with previous studies (20, 21, 56, 59, 60), with the exception of *Phytophthora* taxon totara placed into clade 5. The separation of *P*. taxon totara from clade 3 containing *P. pluvialis* has been reported previously (59).

227 Relationship of *Phytophthora* effector gene numbers to plant host range.

To examine the assemblies for clues as to genomic basis for the diverse host ranges of the sequenced 228 229 species, Phytophthora species were categorized into plant host ranges as follows. Thirteen species were 230 defined as having 'Narrow', 12 as 'Multiple', six as 'Wide', and six as 'Huge' host range (Fig. 5). The 231 numbers of genes predicted to encode various families of effectors were plotted for each host range class. 232 Paralogous effectors with greater than 95% nucleotide identity over the full sequence length were counted 233 only once. Counts were plotted for each of two cytoplasmic effector categories (RxLR and CRN) and three 234 apoplastic effector categories (glycoside hydrolases, protease inhibitors, and NLPs) (Fig. 6). Subcategories 235 of the apoplastic effectors glycoside hydrolases and protease inhibitors were individually plotted (Fig. S1 A 236 and B). In each case, numbers of genes predicted to encode each effector sub-category for each species 237 were plotted against host range.

238 Apoplastic effectors show a distinct overall pattern; Phytophthora species with smaller host ranges had 239 fewer predicted effector genes, with numbers of predicted effector genes increasing with increased host 240 range (Fig. 6: A, B, C). Predicted cytoplasmic effector genes show a similar pattern, starting with low 241 numbers of predicted genes in Phytophthora species with 'Narrow' host ranges and increasing in those with 242 'Multiple' and 'Wide' host ranges. However, for both the RxLR and CRN effector categories, the numbers 243 of predicted genes per species decreas from the 'Wide' to the 'Huge' host range species (Fig. 6: D and E). 244 Two species were outliers with respect to the number of apoplastic NLP effector genes, P. pistaciae in the 245 'Narrow' host category and P. sojae in the 'Multiple' host category had many more predicted apoplastic 246 effector genes than the other species in each of their host range categories, respectively. Four species 247 were outliers with respect to the numbers of predicted RxLR effector genes; P. megakarya and P. pistaciae 248 in the 'Narrow' host category, P. parvispora, in the 'Multiple' host category, and P. palmivora in the 'Huge' 249 host range category all had many more predicted RxLR genes than the other species in those categories. 250 One outlier was observed in the CRN effector genes; P. infestans in the 'Wide' host range category had 251 many more genes than the other species in that category.

252 Horizontal Gene Transfer

253 We evaluated all 31 genomes for evidence of HGT. We used machine learning to identify HGT candidates 254 and phylogenetic approaches to validate candidate HGT genes.

255 Support Vector Machine classifier predicted HGT candidates. Analysis of the 722,232 transcripts with 256 our support vector machine classifier (SVM) over the 31 genome assemblies and the six previously 257 published Phytophthora genomes identified 35,246 HGT candidates. A total of 28,791 of these transcripts 258 that could be regrouped in orthology groups encoding putative transposable elements were discarded, 259 resulting in 6,455 non-TE HGT candidates. The number of candidates predicted ranged from 91 in P. 260 agathidicida to 233 in P. megakarya and 262 in P. palmivora (mean 160.13 ±36.68). P. agathidicida has 261 one of the lowest number of transcripts annotated (12,923 transcripts), while P. palmivora and P. megakarya have the highest gene content with 37,283 and 33,614 transcripts, respectively. Overall, we 262 263 identified a significant linear correlation between gene space in each genome and the number of HGT candidates predicted with the SVM classifier ($r^2 = 0.45$; p < 0.0001). 264

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265 Phylogenetic filtering of HGT candidates. The 6,455 non-TE HGT candidates predicted with the SVM 266 classifier were subject to a two-step filtering process to discard false positives. In the first step, we searched 267 for homologs among a database of sequences built from seven clades (including putative fungal and 268 bacterial donors), followed by phylogenetic tree reconstruction with bootstrap analysis. The phylogenetic 269 filter retained 2,214 candidates, among which 1,113 (50.3%) showed a strong phylogenetic discordance 270 and were seen nested within a distantly related clade (Fungi, Bacteria or Amoebozoa) in direct contradiction 271 to the expected phylogenetic relationships of the respective organisms. The 1,110 other candidates left 272 also branched within a clade of fungal or bacterial genes; however, in these cases, placement of the 273 Phytophthora transcript with Fungi or Bacteria was caused by the absence of homologs in the intermediate 274 clades Viridiplantae, Alveolata and/or Amoebozoa. Numbers of candidates that passed this filter ranged 275 from 36 in Phytophthora europea to 127 in P. palmivora; There was a significant correlation between the 276 total number of genes and the number of HGT candidates across the analyzed species ($r^2 = 0.49$; p < 277 0.0001) (Fig. S2).

278 Sequence identity filtering of HGT candidates. The 2,214 HGT candidates retained after the 279 phylogenetic filtering were submitted to a sequence identity discrepancy filter. A total of 1,688 candidates 280 were rejected after the first "identity test", resulting in a "relaxed" set of 526 HGT candidates for which the 281 sequence identity between the candidate HGT sequence in *Phytophthora* and its closest homolog 282 sequences in the putative "donor" species was shorter than the average identity between the two species; 283 an average of 14.6 (±6.2) candidates were retained per Phytophthora genome with a maximum of 33 for P. 284 palmivora and P. niederhauserii (Fig. S2). A gene enrichment analysis of this candidate set showed a 285 significant enrichment for GO terms related to oxidoreductase activity and hydrolysis and metabolism of 286 carbohydrates (cutinase activity, carbohydrate metabolic process) and proteins (Table S3).

Among the 526 candidates of the relaxed set, 44 passed the second "identity test", constituting a "strict" set of HGT candidates. For 28 of them (56.0%), BLAST search results indicated a strong homology with the clade of the putative donor where the BLAST e-values with species of the putative donor clade were lower than the e-values observed with species from non-donor clades. We then looked at their physical location on their respective scaffold to eliminate potential contaminants; All the candidates were found on scaffolds that had at least two gene models predicted on them. GO term enrichment analysis of this "strict" set indicated significant enrichment for GO terms related to oxidoreductase activity (GO:0055114; GO:0016491; GO: 0008670), and carbohydrates activity and cell wall modification (GO:0000272; GO:0045490; GO: 0042545; GO:0045493; GO:0031176; GO:0030599) (Table 2).

296 Phylogenetic reevaluation of the strict set of HGT candidates. The 44 candidates of the "strict" set of 297 HGT candidates were subjected to reevaluation by sampling additional taxa within the oomycetes. Their 298 amino-acid sequences were first clustered into closely related groups of sequences by assigning them to 299 the 21,303 orthologous clusters previously defined (see Orthology Clustering of Phytophthora Proteins). 300 This process reduced the set of 44 candidates into 28 orthologous clusters, that were then searched against 301 the sequences of the 31 genome assemblies, five *Phytophthora* species sequenced in previous studies, 302 and 36 oomycete genomes (Table S4). Protein members of six of these clusters had homologs (BLASTp 303 e-values \leq 1E-025; Table S4) in the set of 21 strongly supported HGT candidates identified in the genome 304 of P. ramorum, P. infestans and P. sojae by Richards et al. (61). Following these searches, we 305 reconstructed maximum likelihood phylogenies for 19 of these candidates that had putative functions 306 related to the modification of compounds of the plant cell wall (e.g., pectin esterase, xylulose reductase, 307 tannase and endo-1,4-beta-xylanase), peptidases, oxidoreductases and putative elicitors such as a NPP1 308 protein and an ABC transporter (Fig. S3). In 15 cases, the HGT candidate was found nested within a group 309 of fungi or bacteria, as expected under the hypothesis of a transfer from one of these groups through a 310 horizontal transfer event; comparative topology analysis of alternative tree hypotheses (expected 311 phylogeny and transfer from an oomycete donor to a fungus or bacterium) using the Shimodaira-Hasegawa 312 test (SH-test) provided support for this observation in 14 cases (Table 3). In four other cases, the topology 313 test was significant for the opposite relationship where a transfer occurred from the oomycetes to fungi or 314 bacteria. For HGT9, taxon sampling was not sufficient to accurately infer with confidence the putative HGT 315 donor and enable tree topology testing (Table 3).

The distribution of sequence homologs of the "strict" set of HGT candidates among the oomycete phylogeny was strongly variable (Fig. 7; Table S4). The majority of transfer events to oomycetes appear to have occurred relatively recently; three candidates had strong statistical support for transfer from bacteria or 319 fungi to a common ancestor of the Phytophthora genus (HGT5, HGT7 and HGT10), two candidates to the 320 Phytophthora and Peronosporales with hemibiotrophic lifestyle (HGT2 and HGT15) and two to 321 Peronosporales with a hemibiotrophic or an obligate biotrophic lifestyle (HGT12 and HGT20) (Fig. 7A). Four 322 of these transfers reached close to gene fixation within the *Phytophthora* genus as they were found in more 323 than 80% of the species surveyed and in the nine Phytophthora phylogenetic clades considered (Fig. 7A). 324 However, fixation was not the general rule accompanying recent transfers. For instance, two HGT 325 candidates, with functions related to plant pathogenicity (NPP1 protein and peptidase S9) were unique to 326 clade 8 in Phytophthora and did not have a homolog in any other Phytophthora clade or oomycete species 327 (HGT5 and HGT7). Several HGT events with strong statistical support (Fig. 7A) appear to have occurred 328 following major lifestyle transitions within the oomycetes, i.e., necrotrophy in the Pythiales (HGT6, HGT13 329 and HGT14) to obligate biotrophy and hemibiotrophy in the Peronosporales (HGT2, HGT12, HGT15 and 330 HGT20) and transition to parasitism with three events trackable to a common ancestor of the 331 Saprolegniales, Pythiales and Peronosporales (HGT1, HGT4 and HGT8). Eight out of these ten genes had 332 homologs (BLASTp e-values from 1E-137 to 1E-012) with pathogenicity, virulence, and effector genes of 333 the Pathogen Host Interaction database (PHI-base; (62)). Transition to the necrotrophic lifestyle involved 334 transfers of genes encoding enzymes potentially involved in redox activity and toxin production (2,4-dienoyl-335 CoA reductase and phenol acid decarboxylase) while two out of the four genes transfer at the transition to 336 hemibiotrophic lifestyle comprehended have putative functions related to the degradation of the cell wall 337 (xylulose reductase and endo-1,4-beta-xylanase).

Finally, the four significant transfers for the opposite relationship (oomycetes to bacteria or fungi) were all for genes fixed in the *Phytophthora* genus and mapped within the Peronosporales (three candidates) or the Peronosporales and Pythiales (one candidate), suggesting relatively recent transfer events (Fig. 7B). Annotation of these genes indicates that they are potentially involved in the plant-pathogen interaction as they encode proteins involved in protection against plant defensive molecules (tannase and ABC transporter) and the oxidative stress occurring during the plant defense response (quinone oxidoreductase) and remodeling of the plant cell wall (pectinesterase) (Table 3; Table S5).

345 **DISCUSSION**

346 In this comparative genome study of 37 Phytophthora spp., we sequenced and assembled 31 genomes de 347 novo. We investigated these genomes for evidence of horizontal gene transfer, phylogenetic relationships of genome structure and effectors, and association of host ranges. Horizontal gene transfer has been 348 349 identified as a significant source of variation in connection with the evolution of pathogenicity in Phytophthora. So far, genome-wide analyses of HGT impact on oomycete and Phytophthora genome 350 351 evolution has identified putative transfers from fungi (61, 63) and bacteria (64), many of which involve 352 functions related to carbohydrate metabolism and pathogenicity. Our analysis supports these findings, 353 identifying a set of 44 HGT candidates in *Phytophthora* species, associated with enzymes putatively 354 involved in the deconstruction of plant cell wall components, evasion and protection against host defenses 355 (Table 3; Table S5). Our machine learning approach to identify HGT candidates aimed to identify genes 356 likely inherited from bacteria or fungi; validation of these candidates with classical methods based on the 357 identification of topological incongruence in phylogenies and the detection of discrepancies between gene 358 and species distances resulted in a more conservative list of candidates than those previously proposed 359 for *Phytophthora* species (61, 63, 64). Such a stringent approach had the power of rejecting the alternative 360 evolutionary scenario where the gene was present in the last common ancestor of the donor and recipient 361 and was lost in intermediate lineages. Despite such a conservative approach, 30% of HGT candidates 362 identified in a previous study (61) that included only three Phytophthora genomes were retrieved in our analysis. 363

364 An underlying hypothesis related to laterally transferred genes is that they may have functional or ecological 365 roles, allowing the recipient to adapt to a novel lifestyle or to exploit a new ecological niche (65). Using the 366 comprehensive collection of *Phytophthora* genomes sequenced in this study and the oomycetes for which 367 genome assemblies were available, we have been able to assess the extent of distribution of homologs of 368 these candidates across the oomycete phylum. We confirmed that most of the HGTs into the oomycetes 369 have occurred coincident with the emergence of major lifestyle innovations, such as the acquisition of plant 370 parasitism, or biotrophy (obligate or hemibiotrophy). Many candidates were detected in a large majority of 371 the *Phytophthora* genomes sequenced, for example seven candidates (HGT6, HGT11, HGT17, HGT18, 372 HGT19, HGT21 and HGT22) were found in 94% or more of the 36 genomes surveyed. In some instances, 373 homologs have been retained in distinct genetic lineages among the oomycetes, suggesting that these

374 candidate genes may confer a function conserved across lineages with different lifestyles; for example, the 375 endonuclease encoding gene HGT1 was likely transferred before the radiation of the oomycetes and 376 retained in the five oomycete orders surveyed in this study. On the other hand, in some cases the acquisition 377 and/or retention of specific key pathogenicity genes appears to be restricted to some specific clades within 378 Phytophthora (e.g., HGT5 and 7, NPP1; Fig. 7), suggesting recent transfers following divergence of 379 Phytophthora clades. In cases where a putative HGT gene is present in a limited number of Phytophthora 380 species from diverse clades, e.g., HGT13, gene loss by drift in species where there was little benefit may 381 be an explanation. Rapid diversification of the HGT gene under positive selection might make the gene 382 undetectable to our algorithm in some species.

383 Oomycete-derived transfers to other kingdoms have been identified in a few rare instances, usually with 384 limited statistical support (61). With the comprehensive genome sampling of our study, we found strong 385 support for four transfers (out of 44) from oomycetes to either bacteria or fungi, indicating bi-directional 386 exchanges across kingdoms. By providing a source of novel genetic material that can increase the fitness 387 of micro-organisms to their environments or their hosts (66–68), genes transferred horizontally have the 388 potential to be traded back and forth across kingdoms. In the context of an ecological system in which a 389 host plant interacts with a multitude of micro-organisms (microbiota), we can hypothesize that some of the 390 evolutionary innovations that are generated during the co-evolutionary arms race between a pathogen and 391 the host could be shared within the microbiota. For host-associated micro-organisms sharing the same 392 ecological niche, the transfer of genetic material from those that are fit to the shared environment should 393 represent a straightforward mechanism that will drive rapid adaptation of others to this environment (69).

We investigated multiple aspects of *Phytophthora* genome structure and how this relates to the genus phylogeny. Genome size, gene amounts, and counts of orthologous genes varied within phylogenetic clades highlighting the great diversity within the *Phytophthora* genus and likely reflect the large observed differences in repeat content, some of which resulted from genome duplication. While some variation in repeat content may be due to differences in repeats collapsed in the genome assembly, repeat types and lengths identified in this study and sequences used for assembly are generally consistent across the sequenced genomes and therefore should collapse in assembly at similar rates. Interestingly, we noted

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401 that none of the genomes sequenced in this study approaches the 73% repeat content reported in the 402 Sanger-assembled P. infestans genome (10) and is possibly due to the differences in sequencing 403 technologies including sequence longer sequence lengths. Greater repeat content was not necessarily an 404 indicator of large genome assembly size. While the five species with reported repeat content of greater 405 than 30% (P. pinifolia, P. megakarya, P. palmivora, P. fragariae and P. hibernalis) were within the nine 406 largest assembly sizes, other species with large assembly sizes had less repeat content. For example, P. 407 niederhauserii had an assembly size of 90Mb but only 20.9% repeat content which was considerably lower 408 than expected when compared to even moderately repetitive Phytophthora genomes such as P. sojae 40% 409 (29). Genome size estimation using K-mer analysis also shows assemblies are shorter than expected 410 (Table S1). These observations suggest that repeat content may be underestimated in short-read genome 411 assemblies and would expand with improved assembly and may also indicate missing repeat sequences 412 from the *de novo* repeat identification process.

413 The numbers of predicted genes per species, genes per orthologous groups, and effector genes per 414 species were consistent with those previously reported for Phytophthora species. Both the numbers of 415 genes and the average sizes of genes were well within the ranges of the six previously sequenced 416 Phytophthora species, for example P. sojae was shown to have 26K genes with an average size of 1,181 417 bp (29). This supports our observation that the smaller assembly sizes of the *Phytophthora* genomes 418 presented in this study were mainly associated with an overall reduction in repetitive regions while the gene-419 containing sequences are relatively consistent in size. BUSCO analysis of the sequenced core ortholog 420 content also showed similar results to previous Phytophthora studies. Some genome assemblies, including 421 P. palmivora that underwent whole genome duplication, had lower single copy gene numbers due to 422 duplications. But overall, censuses of single-copy orthologs showed that both the genome assemblies and 423 gene predictions were quite complete and comprised the majority of genes in each individual species 424 sequenced. This suggests, that while genomes were small due to collapsed repeat regions, the majority 425 of core orthologs were captured in the assemblies and it can be extrapolated that the majority of gene 426 regions are assembled.

427 Effectors are proteins produced by pathogens that assist in host infection. Effectors are considered rapidly 428 evolving genes that are usually conserved among few closely related species and guickly diverge along the 429 phylogeny. Two types of ortholog groups were identified in our *Phytophthora* genus analysis supporting this 430 hypothesis: one group corresponds to well-conserved genes among all Phytophthora species responsible 431 for core cellular functions, while a second group includes rapidly evolving gene families likely responsible 432 for host infection and adaptation. We also investigated how the amounts of predicted apoplastic and 433 cytoplasmic effectors related to the host range of each *Phytophthora* species. There were large differences 434 in the numbers of effector genes identified per species. We did not observe a correlation between the 435 number of predicted effectors with phylogenetic relationships or genome size. However, a clear relationship 436 with host range was observed. Species with smaller host ranges had on average fewer predicted effectors 437 than those with larger host ranges. When the Phytophthora species in this study were separated into 4 438 host range categories, the distribution of apoplastic effectors increased as the range of infected hosts 439 increased, from narrow to multiple, multiple to wide, and wide to huge. Cytoplasmic effectors showed a 440 similar pattern, however, both the RxLR and CRN effector numbers dropped from wide to huge host ranges. 441 Our study does not include a detailed measurement of gene expression levels of these effector genes 442 during infection of numerous hosts, so there are limitations in how these correlations can be interpreted. In 443 the absence of that information, we speculate that a large diversity of apoplastic effectors may be important for successfully overcoming the apoplastic defenses of a large diversity of host plants. 444

445 There may be a similar requirement for larger numbers of cytoplasmic effectors, but expression of very 446 large numbers of cytoplasmic effectors may limit host range due to plant immune surveillance mechanisms. 447 Detection of a single cytoplasmic effector by an NLR resistance protein may be sufficient to prevent 448 infection, therefore a limited number of cytoplasmic effectors may result in a greatly expanded host range. 449 Our observation of reduced cytoplasmic effector compliments in huge host range species may also be 450 indicative of cryptic host-specialization within these Phytophthora species. Recent work in P. cactorum, 451 commonly considered a broad host-range pathogen, has shown genomic signatures of host specificity (57). 452 In this case, high resolution phylogenetics demonstrated that host adaptation was associated with effector 453 gene gain/loss between strawberry and apple infecting clades. Where such cryptic host-adaptation is

454 present, pangenomic analysis may be a useful tool to infer broad or narrow host range and provide insight
455 to associations of effector diversity across *Phytophthora*.

456 **METHODS**

457 Collection and isolation of a genus-wide *Phytophthora* collection

Mycelium samples were isolated for all *Phytophthora* species in this study, as well as germinated cyst samples for RNA sequencing of a subset of the species. For mycelium tissue, plugs of mycelium grown on standard V8 agar plates were added to a flask with 20% liquid V8 media clarified with calcium carbonate and incubated with shaking at 45 rpm at 25°C for one week. Agar plugs were removed from the mycelium mass and the tissue was ground to a powder with liquid nitrogen, followed by DNA extraction using the methods in Moller et al. (1992) (70) except using 1% CTAB and phenol/chloroform treatment, or total RNA extraction (71) using TRIzol (Invitrogen, Carlsbad, CA) following manufacture's instructions.

465 For Phytophthora robiniae and Phytophthora vignae, similar to the protocol for P. sojae (29) zoospores 466 were produced by repeated washing of 11 day-old V8-200 plates of mycelium with sterile double distilled 467 wather followed by overnight incubation at 14°C. Germinated cysts were produced by exposing collected 468 zoospores to cleared V8 broth for 1 hour. For P. parvispora, mycelium mats were grown in liquid V8 for 5 469 days, then the liquid V8 was changed out for soil extract (soil collected with stream water, mixed, and filter-470 sterilized) and zoospores were collected after three days, followed by germinated cyst induction as above. 471 For P. cajani, P. europaea, Phytophthora foliorum, P. hibernalis, P. pistaciae, and Phytophthora uliginosa 472 species, that did not readily yield zoospores, mycelium was grown in Plich medium (72) for RNA sequencing 473 to compare against V8 medium growth. Known ITS and CoxII sequences for each species were used to 474 confirm species identification before high-throughput sequencing. DNA and RNA guality were checked 475 with electrophoresis (DNA), Bioanalyzer (RNA), and NanoDrop.

Isolate P414 of the strawberry crown rot pathogen *P. cactorum* and isolate SCRP371 of the raspberry root
rot pathogen *P. idaei* were sequenced at the National Institute of Agricultural Botany at East Malling
Research (NIAB EMR). P414 and SCRP371 were isolated from symptomatic strawberry and raspberry
plants, respectively. DNA extraction was performed on freeze dried mycelium using a GenElute Plant

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Genomic DNA Miniprep Kit (Sigma), following the manufacturers protocol with the following modifications:
the RNase A digestion step was not performed and samples were eluted using 2x 100 µL elution buffer for
P414 and 2x 75 µL for SCRP371. Genomic libraries were prepared using a Nextera XT Library Preparation
Kit (Illumina) or TruSeq DNA LT Kit (Illumina) for *P. cactorum* and *P. idaei*, respectively.

484 *P. kernoviae*, *P. lateralis*, *P. cryptogea*, and *P. pinifolia* were collected and isolated as described (73). *P.*

agathidicida, P. multivora, P. pluvialis, and P. taxon totara were collected and isolated as described (74).

Genome sequencing and assembly

487 Thirty-one Phytophthora species were sequenced by our consortium. Genomes of 21 Phytophthora species were sequenced by BGI Genomics (Shenzhen, China) (P. boehmeriae, P. brassicae, P. cajani, P. 488 489 pini, P. europaea, P. foliorum, P. fragariae, P. hibernalis, P. litchii, P. megakarya, Phytophthora melonis, P. 490 niederhauserii, P. palmivora, P. parvispora, P. pisi, P. pistaciae, P. robiniae, P. rubi, P. syringae, P. 491 uliginosa, and P. vignae) using 90-bp paired-end reads produced on the Illumina HiSeg2000 platform. Four 492 Phytophthora species were sequenced by the University of British Columbia (P. kernoviae, P. lateralis, P. 493 pinifolia, and P. crypogea), using Illumina HiSeg 2000 100 bp paired-end reads (73). Genomes from four 494 Phytophthora species isolated from New Zealand (P. agathidicida, P. multivora, P. pluvialis, and Phytophthora taxon totara) were sequenced by Scion (New Zealand Forest Research Institute, Ltd.), using 495 496 primarily Illumina HiSeq 100 bp paired-end reads (74). Two Phytophthora genomes were sequenced by 497 NIAB EMR (P. idaei and P. cactorum), using 250-bp paired-end reads produced on a MiSeg Benchtop 498 Analyser (Illumina, San Diego, CA, USA). BGI-sequenced genomes were adapter trimmed to remove 499 Illumina adapters and quality trimmed to remove Phred scores of less than Q20 from the ends of reads 500 (75). Genome sequences were assembled with SOAPdenovo2 (76). Several initial assemblies were done 501 to identify an optimal K-mer length for each genome. Gap filling and single base proofreading were 502 conducted with SOAPAligner (77). The University of British Columbia (UBC) genomes were quality trimmed 503 and assembled using ABySS (78) and a range of k-values from 32 to 96 (73). Scion genomes were 504 assembled using SPAdes (79) and contigs were extended using SSPACE (74, 80). Genomes sequenced 505 at NIAB EMR were trimmed and adapters removed using fastq-mcf (81) prior to de novo assembly of the 506 data using Velvet (82), at K-mer lengths of 61 and 41 bp for P. cactorum and P. idaei, respectively.

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507 Genome size was estimated using K-mer counts of the raw Illumina sequence. K-mers were counted using 508 Jellyfish count (version 2.2.6, -m 32) (83) Histograms created using Jellyfish hist were plotted using R (84) 509 to identify the apex and boundaries of the single copy K-mer peak. Genome size was calculated by dividing 510 the total of unique K-mers by the mean coverage (peak K-mer frequency).

In order to separate mitochondrial genome contigs from the nuclear genome assembly, full length mitochondrial genome sequences were collected from GenBank (85) for the following 9 *Phytophthora* species; *P. andina*, *P. infestans*, *P. ipomoeae*, *P. mirabilis*, *P. parasitica*, *P. phaseoli*, *P. polonica*, *P. ramorum*, and *P. sojae*. The consortium's 31 assembled *Phytophthora* genomes were aligned with Blat (86) to identify mitochondrial contigs. Blat alignments were filtered to return alignments greater than 50% of the aligned contig length, greater than 100 bp, and with gaps less than 50% of the contig length. Contigs identified as mitochondrial were removed from the genome assembly and are a part of a different study.

518 Transcriptome sequencing and assembly

519 Twenty-four of the genome-sequenced Phytophthora species underwent RNA sequencing (P. brassicae, 520 P. cactorum, P. cajani, P. pini, P. europaea, P. foliorum, P. fragariae, P. hibernalis, P. kernoviae, P. lateralis, 521 P. litchii, P. megakarya, P. melonis, P. niederhauserii, P. palmivora, P. parvispora, P. pinifolia, P. pisi, P. 522 pistaciae, P. robiniae, P. rubi, P. syringae, P. uliginosa, and P. vignae). Two samples, V8-grown mycelia 523 and either Plich-grown mycelia or germinated cysts, were sequenced for each. Twenty-one species were 524 sequenced by the BGI using custom library construction protocol: random hexamer-primer were used to 525 synthesize the first-strand cDNA; second-strand cDNA was synthesized using buffer, dNTPs, RNase H and 526 DNA polymerase I; short fragments were purified with QiaQuick PCR extraction kit resolved with EB buffer 527 and connected with sequencing adaptors. Each Phytophthora transcriptome received 90 bp paired end 528 reads. Two of the above species were sequenced by UBC (P. kernoviae, and P. lateralis). RNA of P. 529 cactorum was sequenced by NIAB EMR.

To create a transcriptome reference for gene predictions, Trinity assemblies (87) were made for each *Phytophthora* species using both RNA sequence samples (--seqType fq --min_contig_length 200). The *de novo* transcriptome assemblies were cleaned using the 3-step TransDecoder process (88). A set of predicted protein sequences was made by combining *Phytophthora* protein sequences from GenBank (85) 534 with protein sequences from the six previously sequenced and annotated Phytophthora species (P. sojae (29). 535 Ρ. ramorum (29), Ρ. infestans (10),Ρ. capsici (31),Ρ. cinnanomi Ρ. 536 (http://genome.jgi.doe.gov/Phyci1/Phyci1.home.html), and parasitica 537 (https://www.ncbi.nlm.nih.gov/genome/11752?genome assembly id=49439)). Transdecoder.LongOrfs (88) was used to identify the longest open reading frames in the Trinity assembly. BLASTP (89) was used 538 539 to align the longest ORFs to the set of constructed proteins identified from GenBank (parameters: max target segs 1 -evalue 1e-5). Finally, Transdecoder.Predict was used to predict the gene structure 540 541 from the transcriptome assembly. The resulting cleaned transcriptome assemblies were used in the 542 subsequent gene prediction methods.

543 **De novo repeat identification**

Each genome was repeat-masked to create a genome assembly ready for gene prediction as described below. Repeat elements were *de novo* identified separately by species. *De novo* predictions were combined along with previously identified *Phytophthora* repeats for species-specific repeat identification.

547 To identify de novo discover LTR retrotransposons, LTR harvest (90) and LTR_(91) finder were run on each 548 genome assembly. By species, LTR retrotransposon predictions from both LTR harvest and LTR finder were condensed by coordinates and reduced by Blat alignments. LTR retrotransposons, non-LTR 549 550 retrotransposons, DNA transposons, and other repeat elements were identified following the MAKER 551 'Repeat Library Construction-Advanced' (92) method 552 (http://weatherby.genetics.utah.edu/MAKER/wiki/index.php/Repeat Library Construction-Advanced). 553 This process utilizes the following programs: MITE-Hunter (with default parameters) (93); GenomeTools

554 suffixorator, LTRharvest, LTRdigest (run with 99% and 85% identity) (94); RepeatModeler (95) which calls 555 RECON (96), RepeatScout (97). TRF (98), NSEG (99) and RMBlast 556 (http://www.repeatmasker.org/RMBlast.html); and sequence databases provided by the MAKER 'Repeat 557 Library Construction-Advanced' method.

558 De novo repeat identification was further supplemented by LTR_retriever (100) using results from 559 LTRharvest and LTRfinder. LTR_retriever retrotransposons were classified into sub-families by species. For each assembled genome, predicted repetitive elements identified in the above methods were combined with GIRI RepBase (volume 18, issue 9) (101) *Phytophthora* repeats. Genomes were repeat-masked using RepeatMasker (v 4.0.6, run with the described combined custom *Phytophthora* library and default parameters) (102) and the combined repeat database to create gene prediction-ready genome assemblies.

564 Gene prediction and annotation in 31 *Phytophthora* species

565 For each species sequenced, gene training models were made with both Augustus (103) and SNAP (104). 566 Augustus was trained using the genome assembly and the set of previously sequenced Phytophthora 567 proteins described in the transcriptome sequencing and assembly process in the text above. To train 568 SNAP, for each species BUSCO (105) was run using the Alveolata Stramenopiles database in genome 569 mode on each genome assembly to identify core orthologs. BUSCO gff files were converted to zff using 570 maker2zff (92), SNAP tools fathom (-categorize 1000, -export 1000), forge, and hmm-assembler were used 571 to create a training HMM. Genes identified as single copy core orthologs were combined and used as the 572 SNAP training set.

573 Several supplementary files were created to run the MAKER gene prediction pipeline. For the 'EST 574 Evidence' section of MAKER the transcriptome result of the three-step TransDecoder process (88) was 575 used as the EST field. In the seven cases where RNA was not sequenced and therefore the TransDecoder 576 transcriptome was not created, the phylogenetically closest species with RNA sequences was used. A concatenation of the gene sequences of previously sequenced Phytophthora species and TransDecoder 577 578 transcriptome assemblies was used for the alt-est field. For each genome assembly, P. sojae, P. infestans, 579 and P. ramorum gene predictions were combined with five representative TransDecoder-cleaned 580 transcriptome assemblies. To create species-specific alt-est sets the five representative species were 581 selected from within the *Phytophthora* phylogenetic clade (excluding the species of interest). If less than 582 five species received RNA sequencing, nearby clades were used until five transcriptome assemblies were 583 combined. This set of eight gene predictions and transcriptome assemblies was used in the alt-est field. For the 'Protein Homology' section of MAKER all previously identified Phytophthora proteins were combined 584 585 from GenBank together with the six previously sequenced *Phytophthora* species.

To predict genes, MAKER was run on the 31 genomes. The repeat-masked genome for each assembly was run using the Augustus and SNAP training models described above and the EST and protein evidence sequence sets as described above.

For validation purposes, three sets of BUSCO analyses were run. BUSCO 3.02 was run on the 31 genomic
assemblies, on the 24 transcriptome assemblies, and on the 31 sets of predicted proteins. In all cases, the
Alveolata_Stramenopiles BUSCO database of 234 single copy orthologs was used.

592 Predicted genes were functionally annotated using BLASTX (89) to align against known proteins. First, all 593 predicted proteins were aligned against all Phytophthora species' proteins obtained from the RefSeq non-594 redundant protein database in NCBI (85, 106) and the UniProt TrEMBL (107) database. Second, all 595 predicted proteins were aligned against all stramenopile proteins from NCBI RefSeg and UniProt. Third, 596 all predicted proteins were aligned against all Fungi proteins from NCBI RefSeq and UniProt. BLASTX 597 alignments were generated with the following parameter settings: -evalue 1e-5, -max_target-seqs 50 and 598 were further filtered to return only hits that were at least 50% identical for 50% of the length of the subject 599 protein. Protein functional annotation was made from the consensus of the top five protein alignments for 600 each taxonomic classification. All results per query protein were screened to return proteins with 601 informative functional annotations ranked by the two rounds of alignments.

Predicted proteins from MAKER were screened using InterProScan version 5.20-59.0 (108) to identify functional domains. Gene Ontology (GO) (109, 110) terms were obtained from UniProt BLASTX alignments and the InterProScan runs.

605 Identification of Effector Proteins

The prediction of cytoplasmic effectors of the RxLR and crinkler (CRN) families was performed on the sixframe translations of the *Phytophthora* whole-genome assemblies using the application getorf (EMBOSS suite) (111). We searched for evidence of the presence of the motifs of interest (RxLR+EER motif for RxLR effectors (112), and LxLAK for CRN effectors (10, 42)) in each ORF translation by using a combination of regular expressions using effectR (113).

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611 To identify additional effector proteins that may not include one the canonical motifs, thus not recognized 612 by the RxLR or CRN regular expression, we searched against a profile-HMM (114). We built the HMM 613 profile using an intersect of each set of candidate effectors predicted using regular expressions for each 614 sequenced species in the consortium and the previously predicted effectors from the reference genomes 615 of P. infestans, P. ramorum, and P. sojae (10, 29). We searched for additional effectors in all ORF 616 translations against each of HMM profile using the hmmsearch program in HMMER (115) using default 617 threshold parameters. Predicted effectors from the motif method and the HMM method were examined for 618 signal peptides using SignalP 3.0 (116).

Putative apoplastic protease inhibitors were annotated by batch BLASTP (E-value < 1e-30) against MEROPS database (117). The glycoside hydrolase proteins were annotated using the carbohydrate-active enzyme database (cazy) annotation web server dbCAN (118). The hidden Markov model (HMM) profile of NLP family (PF05630) was downloaded from Pfam database (119). The hmmsearch program (115) (with default threshold parameters) was used to search for NLP proteins in each genome assembly.

624 Classification into Orthologous Groups

625 Predicted proteins from the 31 sequenced Phytophthora and from the six previously sequenced 626 Phytophthora species were combined into orthologous groups using OrthoMCL (120). Due to the large 627 data set of 37 full genomes, a two-stage process was used. In stage one, proteins were assigned to the 628 online pre-constructed OrthoMCL orthology groups (121). Predicted proteins from the MAKER process 629 were uploaded to the OrthoMCL web site (www.orthomcl.org). This returned a file of proteins assigned to 630 OrthoMCL groups. In stage two, all unassigned proteins were assigned to groups using the stand-alone 631 version of OrthoMCL. All unassigned proteins from all 37 species were combined into a single FASTA 632 sequence file. The protein FASTA file was aligned against itself using BLASTP. The BLASTP output was 633 converted for input into OrthoMCL which was run in 'Mode 4'.

634 Phylogenetic analysis of single copy orthologs across *Phytophthora* species

To estimate phylogenetic relationships across the 31 genomes and the six previously sequenced *Phytophthora* species, we first identified the single-copy, core orthologous genes shared across all 637 sequenced species. We selected each of the orthology groups that contain exactly one gene from each of638 the 37 genomes in the orthology construction.

We constructed a phylogenetic tree using 61 genes from each species. Each set of orthologous proteins
were multiply aligned using MAFFT ver. 7.271 (122, 123). The phylogenetic tree was reconstructed using
RAxML (124), using each gene as an independent partition with its own substitution model, bootstrapped
1,000 times. Only one tree was calculated using all partitions.

643 Effector distribution across plant host ranges

The USDA fungal database (https://nt.ars-grin.gov/fungaldatabases) was used to define the number of known plant hosts infected by each *Phytophthora* species considered in this study. With this information, host ranges were classified into four categories defined by the number of host genera containing known hosts: 'Narrow', host species confined to 1 plant host genus (1-3 host species total); 'Multiple', host species confined within 2 to 9 host genera (2 to 32 host species total); 'Wide', host species spanning 16-55 host genera (22 to 119 host species total); and 'Huge', host species spanning 107-327 host genera (163-718 host species total).

651 Numbers of predicted cytoplasmic and apoplastic effector genes were plotted for each host range. In order 652 to reduce errors caused by genome assembly artifacts, and to limit counts of functionally identical effector 653 genes, near-identical paralogs were removed from the counts. To identify near-identical paralogs, effector 654 amino acid sequences were aligned to one another using the Smith-Waterman local aligner from the 655 EMBOSS package (111) to identify similarity. Effectors with greater than 95% amino acid identity over the 656 full sequence length were reduced to one representative effector sequence. The resulting reduced set of 657 effector predictions was used for analysis of the relationship of effector repertoires to host range. Host 658 ranges were plotted for each effector type using ggplot2 (125) in R (84).

659 Horizontal Gene Transfer

We used a two-step process to identify HGT gene candidates in *Phytophthora* genomes. In the first step, we used a SVM classifier to predict HGT candidates. In the second step, we applied two filters to screen out false positive candidates and assess the likelihood that the candidates were acquired through HGT.

663 Support Vector Machine classifier for prediction of Horizontal Gene Transfer candidates

664 We hypothesized that DNA sequence-composition features such as G + C content, codon bias and codon 665 usage frequency (126) can be used to identify genes of recent bacterial or fungal origin in Phytophthora 666 genomes. We constructed a multiclass SVM; (127) for composition-based analysis of Phytophthora 667 protein-coding genes and classification as either Phytophthora, bacterial or fungal origin. SVM is well suited 668 for sequence-composition classification because of the availability of SVM libraries that perform well with 669 large data sets with numerous variables and the ability of SVM to minimize unimportant features (128). The 670 SVM algorithm was implemented in a custom Python script using the SVC function, available from Scikit-671 learn Python library (128).

672 Training sets consisted of 15,000 each of ascomycete, *Phytophthora* and bacterial transcripts, for a total of 673 45,000 transcripts. Ascomycete transcripts were selected by submitting a collection of complete transcript 674 sets predicted from the genomes of representative species of eight main ascomycete classes: Tuber 675 melanosporum (Pezizomycetes; GCA 000151645) (129), Arthrobotrys oligosporus (Orbiliomycetes; 676 GCA 000225545) (130), Penicillium chrysogenum (Eurotiomycetes; GCA 000226395) (131), 677 Leptosphaeria maculans (Dothideomycetes; GCA_000230375) (132), Cladonia grayi (Lecanoromycetes) 678 (133), Sclerotinia sclerotiorum (Leotiomycetes; GCA_000146945) (134), Fusarium graminearum 679 (Sordariomycetes; GCA 000240135.3) (135), and Xylona heveae (Xylonomycetes; GCA 001619985) 680 (136). Potential genes that underwent HGT were discarded from each transcript set by applying the 681 following protocol: 1, transcripts were translated into proteins and clustered using OrthoMCL (coverage and 682 identity of at least 50%, E-value cut-off of 1e-05, inflation parameter = 2.5) (120); 2, one protein from each 683 cluster was then gueried against the NCBI nr database (max target sequences = 500); 3, clusters with at 684 least one hit in any other organisms than a fungal taxon were discarded; 4, for each remaining cluster, each 685 protein was queried against nr (max target sequences = 500) and step 3 was re-applied. Phytophthora genes were selected by the same process, using transcripts from the following species: P. syringae (this 686 687 study), P. sojae (GCA 000149755) (29), P. ramorum (GCA 000149735) (29), P. lateralis 688 (GCA 000500205) (73), P. pinifolia (GCA 000500225) (73), P. cryptogea (GCA 000468175) (73), P. 689 infestans (GCA_000142945) (10), P. brassicae (this study), and P. kernoviae (GCA_000448265) (73) and

eliminating clusters with any protein match other than with an oomycete taxon. Bacterial transcripts were
selected following the same filtering approach on 21,096 transcripts retrieved from GenBank (representing
bacterial classes).

693 Sequence-composition features were used as input vectors to an SVM classifier and the curated training 694 sets (see above) were used as model data. Following a preliminary analysis, codon usage frequency and 695 GC content were selected as the sequence features as they resulted in a higher prediction accuracy than 696 codon bias (0.976 ± 0.002 vs. 0.973 ± 0.004 , t = 10.4, p <0.0001; data not shown). This is consistent with 697 the point that codon use frequency is inherently the fusion of both codon usage bias and amino acid 698 composition signals (137). To choose the best kernel for the SVM, we first used principal component 699 analysis (PCA) to explore the relationships among the three different classes (Fig. S4). Radial basis function 700 (rbf) kernel parameters (C and gamma) were systematically varied to optimize prediction accuracy using a two-dimensional grid where both parameters were chosen from the set {10⁻³, 10⁻², ..., 10⁶}. All these 701 702 optimizations were performed with fivefold cross-validation of the training set (randomly withholding one-703 fifth of the training data as a testing data set; 100 random draws for each pair of parameters tested) (Fig. 704 S4). Accuracy as defined by (TP + TN) / (TP + TN + FP + FN) was used as a measure of the quality of the 705 classification. Best classification accuracy (98.3%) was obtained with rbf kernel parameters of C = 1000.0706 and *gamma* = 1.0 (Fig. S4).

707 Phytophthora transcript classification for Horizontal Gene Transfer

708 The 618,240 transcripts predicted from the 31 genomes and 103,992 transcripts predicted from five 709 previously sequenced Phytophthora species (i.e. P. sojae, P. ramorum, P. infestans, P. capsici and P. 710 cinnamomi [see Transcriptome sequencing and assembly]) were submitted to the classifier and sorted into 711 Phytophthora-origin, bacterial-origin or fungal-origin classes depending on the probability returned by the 712 classifier for each of these classes. To generate a confidence score, we repeated the training of the 713 classifier 100 times before running the classification on each genome. To maximize the training process of 714 the classifier without increasing computing time and overloading memory we used a random subsample of 715 45,000 transcripts (15,000 genes in each of the three classes) as a training set each time. Preliminarily, we 716 determined the minimum threshold number of bootstrap replicates in which an HGT candidate was found

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717 that would minimize the probability that this candidate was a false positive (e.g., misclassifying a 718 Phytophthora or a bacteria sequence as deriving from a fungal donor via HGT). This was done by submitting 719 a subsample of 1,500 sequences randomly picked in the training set (500 transcripts in each class) to the 720 classifier with the bootstrap procedure; then, false positive and true positive rates were calculated for 721 incremental values of bootstrap replicates. Based on this test, we determined that the chance of 722 misclassifying a fungal transcript as a *Phytophthora* or a bacterial sequence (i.e. a false positive) was < 723 0.2% (1/500) if it was classified as fungal in at least 89/100 bootstrap replicates; in such case the true 724 positive rate (recall; TP/(TP + FN)) would be 92.3% (Fig. S5A). For the bacterial sequences, this value was 725 ≥79/100 bootstrap replicates; this corresponded to a true positive rate (recall) of 97.5% (Fig. S5B). These 726 two bootstrap replicate thresholds were then used for the identification of HGT candidates in *Phytophthora* 727 species.

728 Horizontal Gene Transfer candidate false positive filtering

729 HGT candidates predicted with the SVM classifier were submitted to a phylogenetic filtering step by 730 assessing the congruence of the gene phylogeny with the organism phylogeny. Each candidate transcript 731 was translated into a protein sequence and searched using DIAMOND BLASTp (minimum BLASTP E-value 732 of 1e-03, sequence subject coverage of 50% and sequence query coverage of 50%) (138) for closest 733 homologs against protein-coding sequences downloaded from the NCBI Reference Sequence collection 734 (RefSeq) (106) for Phytophthora species. (72,639 sequences) and the following clades: heterokonts 735 (excluding Phytophthora; 164,619 sequences), Alveolata (1,527,928 sequences), Amoebozoa (113,408 736 sequences), Viridiplantae (5,556,940 sequences), Fungi (2,912,973 sequences), Archaea (1,830,006 737 sequences) and Bacteria (131,971,793 sequences) clades. Candidates with no hits in the Bacteria or Fungi 738 clades were directly rejected. Protein sequences for the top three DIAMOND BLASTp hits within each of 739 the above clade were retrieved and aligned with the query protein using MAFFT ver. 7.271 (123). Amino-740 acid sites with a gap in more than one third of the sequences were removed. IQ-TREE was used to 741 determine the best-fitting substitution model and reconstruct a maximum-likelihood tree for each of the 742 alignments (139, 140) with node support assessed by using the ultrafast bootstrap approximation method 743 (141). Each phylogenetic tree was exported in Newick format, automatically rooted with sequences from

744 the Archaea or Bacteria clades and exported into a .png file using the BioPython package Phylo. For 745 facilitating visual examination of trees, png files were gathered into one single pdf catalog with the Python library PyFPDF. Tree nodes were visually inspected to identify phylogenetic discordance (142). Two types 746 747 of discordance were examined: 1, "complete incongruence", when the HGT candidate sequence clusters 748 with the fungi or bacteria clade with bootstrap support ≥50%, resulting in a phylogenetic tree completely 749 discordant with the expected organism phylogeny i.e. ((((((Phytophthora, Heterokonta), Alveolata), 750 Viridiplantae), (Fungi, Amoebozoa)), Archaea), Bacteria) (143, 144) 2, "missing clades", when the HGT 751 candidate sequence clusters with the fungi or bacteria clade because other clades are missing (i.e. the 752 HGT candidate sequence didn't have orthologs in intermediate clades such as Viriplantae, Amoebozoa and 753 Alveolata).

754 HGT candidates that passed phylogenetic filtering were submitted to a sequence identity discrepancy filter. 755 Assuming a molecular clock, the sequence identity between a pair of orthologous genes should be in the 756 same range as the average sequence identity between their respective species. However, for a pair of 757 sequences related through an HGT event between two species, the proportionality should be broken, 758 leading to an identity discrepancy when compared to the pairwise species identity (145). To identify such 759 discrepancies, we performed a first "identity test": We calculated the nucleotide sequence identity between 760 the candidate HGT sequence in *Phytophthora* and its closest homolog sequences in the putative "donor" 761 species in Bacteria or Fungi. Then, the full transcriptome of the putative "donor" species was downloaded 762 and searched with BLASTN for 1,000 random transcripts from the Phytophthora species to identify one-to-763 one orthologs and plot a distribution of expected nucleotide identity values. In a first "identity test", 764 discrepancies were identified by comparing the observed nucleotide sequence identity found between the 765 HGT candidate in *Phytophthora* and its homolog sequence in the putative donor to the expected distribution 766 using a Wilcoxon sign-rank test; Candidates were rejected if the difference between the observed value 767 (sequence identity between the HGT candidate in *Phytophthora* and its homolog sequence in the putative 768 donor) and the average of the expected distribution was not significant and lower than an arbitrary 769 "discrepancy cutoff" of 8.56 (corresponding to the top guartile of the distribution of the difference between 770 the observed values and the expected values). Proteins retained at this step were included into a "relaxed" 771 list of candidates. To ascertain that the discrepancy was not caused by a high conservation of the gene

772 among the different clades, we validated this list with a second "identity test" that consisted of examining if 773 the nucleotide sequence identity between the HGT candidate and the closest species in the non-donor 774 clades was not significantly higher than the average identity expected between the two species. Only 775 candidates for which the difference in nucleotide sequence identity between the HGT candidate in 776 Phytophthora and its homolog sequence in the putative donor was lower than 52% (corresponding to the 777 uppermost quartile of the distribution of the difference between the observed values and the expected 778 values of nucleotide sequence identity between pairs of homologs from the two species) were retained after 779 this stage. When the second "identity test" could not be performed due to the absence of sequence 780 homologs in non-donor clades (i.e. Viriplantae, Amoebozoa and Alveolata), the "discrepancy cutoff" i.e. the 781 difference between the nucleotide sequence identity for the HGT candidate in *Phytophthora* and its homolog 782 sequence in the putative donor and the average of the expected distribution was raised to 80.83% identity 783 corresponding to the 5% upper quantile of the of the distribution of the difference between the observed 784 values and the expected values. Proteins that passed this second filter were kept in a "strict" list of 785 candidates.

We assessed if the HGT candidate could have been a bacterial or fungal contaminant mistakenly sequenced and assembled with a genome assembly generated in this study. HGT candidates were considered as a putative "contaminant" candidates if they were the only coding sequences to map to a given scaffold.

790 Data Availability

Genome assemblies and genome and transcriptome sequences that were created in this study have been deposited in NCBI under BioProjects PRJNA746351, PRJNA714689, and PRJNA702516. *P. cactorum* and *P. idaei* data are available under BioProjects PRJNA383548 and PRJNA391273. Other genomic resources including assembly files, gene/protein predictions and annotations, differential expression analysis, and orthology analysis can be accessed at https://phyto-seq.cqls.oregonstate.edu.

796 Acknowledgments

NJG was supported by USDA Agricultural Research Service project 2072-22000-041-000-D, National
Institute of Food and Agriculture grant 2018- 67013-27823, and the J Frank Schmidt Foundation.

799 Genome sequencing of *P. pini* was supported by the California Walnut Board and the United States

800 Department of Agriculture Agriculture Research Service, CRIS Project # 5306-22000-014-00D to TK. We

- thank Gregory Browne and Daniel Kluepfel for *P. pini* isolate and valuable inputs.
- This work was supported in part by grants to YW from China National Funds for Distinguished Young Scientists (31225022).

804 RJH, ADA and CFN were supported by grants from the UK Biotechnology and Biological Sciences

805 Research Council (BBSRC; BB/K017071/1, BB/K017071/2 and BB/N006682/1).

- 806 RM, PP and NW were funded by the New Zealand Ministry of Business, Innovation and Employment
- 807 (grant number CO4X1305), the Forest Growers Levy Trust (administered by the New Zealand Forest
- 808 Owners Association) and the Radiata Pine Breeding Company under the "Healthy trees, Healthy future"
- 809 research programme at Scion (NZFRI, Ltd).
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			Genor	ne	Anno	tations
Species	Clade	Contigs	N ₅₀	Assembly Length	Repeat Percent	Predicted Genes
P. cactorum	1	7888	15053	56443298	19.96	18027
P. idaei	1	7163	14461	53468943	16.19	18038
P. pini	2	2131	42987	38730000	7.31	14019
P. multivora	2	2844	46133	40059192	10.86	13682
P. pluvialis	3	4340	30816	53616150	16.04	16285
P. litchii	4	2543	34546	38200938	5.98	12391
P. palmivora	4	24815	6694	107798747	29.62	37283
P. megakarya	4	24073	7093	101609312	31.94	33614
P. agathidicida	5	3754	19544	37337699	5.76	12923
P. taxon totara	5	4425	30809	55576372	16.56	17619
P. parvispora	7	9906	6820	46825958	8.75	15642
P. pisi	7	7667	15253	58856683	16.80	18953
P. robiniae	7	14865	8754	69938814	25.58	23128
P. niederhauseri	7	26463	4805	90270009	20.96	29587
P. cajani	7	18255	5113	64854085	20.65	19840
P. vignae	7	10330	8363	56137732	17.45	18535
P. melonis	7	11353	15342	73416743	25.93	21276
P. pistaciae	7	10414	10302	63209321	18.56	19423
P. uliginosa	7	8955	10095	57072031	24.16	17226
P. europaea	7	8301	11551	58787065	23.33	17117
P. fragariae	7	8544	20362	76969737	30.81	20448
P. rubi	7	9434	17808	74863594	29.48	23476
P. pinifolia	6	22610	6021	74478861	33.00	23717
P. lateralis	8	28263	2396	50496828	23.39	19503
P. hibernalis	8	6587	21408	71256216	32.46	23578
P. foliorum	8	5320	15800	48973082	19.26	16083
P. brassicae	8	12447	12337	72849437	28.39	26010
P. syringae	8	6572	15987	57045526	21.71	18234
P. cryptogea	8	25944	4730	69446343	17.65	24936
P. boehmeriae	10	2866	41917	39747814	7.83	13325
P. kernoviae	10	13710	5225	42698878	4.15	14322

 Table 1. Sequencing and assembly statistics for 31 Phytophthora species grouped by clade.

GO	Term	# terms in full set ^a	# terms in HGT set ^b	Pr(<i>X=k</i>) ^c
Biological proc	ess			
GO:0055114	obsolete oxidation-reduction process	1543	5	<0.001
GO:0034079	butanediol biosynthetic process	28	5	<0.001
GO:0045493	xylan catabolic process	151	1	<0.01
GO:0008152	Metabolism	3663	3	<0.01
GO:0042545	cell wall modification	493	1	<0.02
GO:0000272	polysaccharide catabolism	635	1	<0.02
GO:0002084	protein depalmitoylation	374	1	<0.02
GO:0006118	electron transport	1389	1	<0.05
GO:0045490	pectin catabolic process	865	1	<0.05
Molecular funct	ion	10991	11	<0.001
GO:0016491		10881	11	< 0.001
GO:0000721	(R,R)-butanediol dehydrogenase activity	28	5	< 0.001
GO:0003939	L-iditol 2-dehydrogenase activity	11	1	<0.001
GO:0016831	carboxy-lyase activity	201	6	< 0.001
GO:0008080	N-acetyltransferase activity	1082	3	<0.001
GO:0008270	zinc ion binding	26982	8	<0.001
GO:0005488	Binding	2200	2	<0.01
GO:0031176	endo-1,4-beta-xylanase activity	111	1	<0.01
GO:0008670	2,4-dienoyl-CoA reductase (NADPH) activity	15	1	<0.01
GO:0004022	alcohol dehydrogenase activity	102	1	<0.01
GO:0030599	pectinesterase activity	498	1	<0.02
GO:0008474	palmitoyl-(protein) hydrolase activity	457	1	<0.02
GO:0051213	dioxygenase activity	460	1	<0.02
GO:0045330	aspartyl esterase activity	423	1	<0.02

Table 2. Over-represented GO terms for a set of 44 candidate HGT transcripts found in 36 Phytophthora genomes.

GO:0015267 channel activity 742 1 < 0.05 Cellular component GO:0005576 extracellular region 4624 2 < 0.02 ^aTranscriptome of 35 *Phytophthora* genomes.

GO:0045330 aspartyl esterase activity

GO:0030570 pectate lyase activity

^b44 HGT candidates.

^cProbability (q-value) of obtaining the same number of transcripts (k) or more by chance as given by a hypergeometric probability distribution.

1

< 0.02

657

Table 3. Summary of 19 HGT candidates with strong phylogenetic support identified among *Phytophthora* spp.

	Putative function	Best hit on PHI-base (e-value)ª	HGT identification ²	# of <i>Phyto.</i> species	# of <i>Phyto.</i> clades	Closest clade ³
Transf	er from other groups to a	oomycetes				
HGT_1	Endonuclease	PHI:5754, endonuclease, <i>Fusarium graminearum</i> (3.0e-024) PHI:2256, xylitol	SVM, IP, D, T	29	8	Fungi
HGT_2	Xylulose reductase	dehydrogenase, Parastagonospora nodorum (1.0e-128)	SVM, IP, D, T	12	5	Fungi
HGT_15	Zinc-binding dehydrogenase, Polyketide synthase, enoylreductase domain	PHI:8321, gluconate 5- dehydrogenase, <i>Salmonella</i> <i>enterica</i> (1.0e-016)	SVM, IP, D, T	21	8	Bacteria
HGT_4	Aquaporin	PHI:7047, water channel protein aquaporin, <i>Cryptococcus neoformans</i> (4.0e-012)	SVM, IP, D, T	30	8	Bacteria
HGT_5	NPP1	-	SVM, IP, D, T	4	1	Bacteria
HGT_6	Phenol acid carboxylase	-	SVM, MC, D, T	35	9	Fungi
HGT_7	Peptidase S9	-	SVM, IP, D, T	8	1	Fungi
HGT_8	UDP-N- acetylglucosamine- peptide N- acetylglucosaminyl- transferase	PHI:4921, flagellin glycosyltransferase, <i>Burkholderia cenocepacia</i> (1.0e-021)	SVM, IP, D, T	2	2	Bacteria
HGT_9	Alternative oxidase	-	SVM, IP, D	10	4	Fungi
HGT_10	Dioxygenase	-	SVM, MC, D, T	33	9	Fungi
HGT_11	Thioesterase	PHI:4988, sfp-type 4'- phosphopantetheinyl transferase, <i>Bipolaris</i> <i>maydis</i> (3.0e-005)	SVM, IP, D, T	35	8	Amoebozoa
HGT_12	Endo-1,4-beta- xylanase GH10	PHI:7912, endo-beta-1,4- xylanase <i>Phytophthora</i> <i>parasitica</i> (1.0e-137)	SVM, IP, D, T	33	9	Fungi
HGT_13	4-coumarate CoA ligase	fatty-acid–Co Aligase, <i>Pseudomonas aeruginosa</i> (1.0e-032)	SVM, IP, D, T	4	4	Fungi/ Bacteria
HGT_14	2,4-dienoyl-CoA reductase	PHI:8134, 3-Oxoacyl-[acyl- carrier-protein] reductase, <i>Salmonella enterica</i> (1.0e- 014)	SVM, IP, D	28	7	Bacteria/ Archaea
HGT_20	Ribosomal-protein- alanine acetyltransferase	-	SVM, IP, D	29	9	Bacteria

Transfer from oomycetes to other groups						
HGT_3	Quinone oxidoreductase	-	SVM, IP, D, T	23	9	Fungi
HGT_16	Putative tannase	PHI:10222, feruloyl esterase, <i>Valsa mali</i> (4.0e- 030)	SVM, IP, D, T	33	8	Bacteria
HGT_17	Putative pectinesterase CE8	PHI:278, pectin methylesterase, <i>Botrytis</i> <i>cinerea</i> (4.0e-077)	SVM, IP, D, T	34	9	Fungi
HGT_18	ATP-binding Cassette (ABC)	-	SVM, IP, D,	34	9	Fungi

^a Best BLASTp hit on the Pathogen Host Interaction database PHI-base (Urban et al. 2019); ²SVM, support vector machine; IP, incongruent phylogeny; MC, missing clades; D, distance; T, Alternate topology test (Shimodaira-Hasegawa-test) significant; ³as reported in BLASTp analysis; ³HB, only hemibiotrophic Peronosporales i.e. *Phytophthora*, *Phytopythium* and *Nothophytophthora*.



Figure 1. Analysis of 31 *Phytophthora* species shows an abundance of repetitive elements. Species are shown in phylogenetic clade order; clade designations are shown on the left. Repeat content is displayed as percentage of the total genome content. Repeat classifications are shown as colored bar segments.

705x481mm (90 x 90 DPI)



Figure 2. BUSCO analysis demonstrates completeness of the 31 *Phytophthora* species in this study. A) genomic assembly, B) transcriptome assembly, and C) predicted proteins. Species are shown in phylogenetic clade order and clade designations are shown on the left. For each BUSCO analysis, results from searching 234 single copy orthologs in the Alveolata_Stramenopiles dataset are shown.

987x518mm (90 x 90 DPI)



Figure 3. The number of predicted effectors varied across the *Phytophthora* genomes. Bar chart representing amounts of effector genes found in 31 *Phytophthora* species for crinkler (CRN), RxLR, glycoside hydrolases, necrosis inducing proteins (NLPs), and protease inhibitors. Species are shown in phylogenetic clade order, clade designations are shown on the left.

705x481mm (90 x 90 DPI)





Figure 4. Distribution of genes in *Phytophthora* orthology shows a bimodal frequency distribution,
highlighting genes that are conserved in only a few genomes and orthology groups that have one gene per species. Amounts of genes assigned to orthology cluster with OrthoMCL (118) are shown. The 31 sequenced *Phytophthora* and six additional previously sequenced *Phytophthora* genomes (*P. capsici, P. cinnanomi, P. infestans, P. parasitica, P. ramorum, P. sojae*) are included. 'All Orthology Groups' (orange) show all genes assigned into orthology groups. 'Informative Functional Annotations' show genes assigned into orthology groups that have useful functional definitions and exclude genes labeled as 'uncharacterized', 'hypothetical', or similar.

846x623mm (90 x 90 DPI)



Figure 5. Phylogenetic relationships of the 31 sequenced *Phytophthora* spp. and six additional previously sequenced *Phytophthora* spp. Sixty-one single-copy core orthologous proteins shared across 37 species were used to create a RAxML (122) phylogenetic tree using each gene as an independent partition with its own substitution model and bootstrapped 1,000 times. Ranges of infected hosts are shown next to the phylogenetic tree species defined as 'Narrow', host species confined to 1 plant host genus; 'Multiple', host species confined within 2 to 9 host genera; 'Wide', host species spanning 16-55 host genera; and 'Huge', host species spanning 107-327 host genera. Clade assignments are shown on the right.

812x762mm (96 x 96 DPI)



Figure 6. Generally, *Phytophthora* with larger host range showed a greater predicted number of effector genes. Box plots showing amounts of effectors found per *Phytophthora* species, categorized into 'Narrow' (1 plant genera), 'Multiple' (2-9 plant genus), 'Wide' (16-55 host genera), and 'Huge' (107-327 host genera) host ranges. Effectors glycoside hydrolases (A), protease inhibitors (B), NLPs (C), RxLRs (D), and CRNs (E) are shown. Near identical paralogs were removed; proteins with greater than 95% amino acid identity over the full sequence length were reduced to one representative effector sequence. Statistically significant classifications were seen between the Narrow-Wide and Narrow-Huge comparisons in glycoside hydrolases (A).

1016x571mm (90 x 90 DPI)



Figure 7. Conservation level of the 44 *Phytophthora* HGT candidates in oomycetes. A and B, set of 19 HGT candidates for which a maximum likelihood phylogeny was reconstructed and alternate tree topologies were tested with the Shimodaira-Hasegawa test (an asterisk indicate significant topological difference [P < 0.05] between the constrained alternate topology and the observed topology; A, HGT to oomycetes; B, HGT for the opposite relationships i.e. transfers from oomycetes to fungi or bacteria; C, HGT candidates with no maximum likelihood phylogeny support. For each HGT candidate the number of sequence homologs identified among 37 *Phytophthora* and 30 oomycetes transcriptomes (identified by reciprocal DIAMOND BLASTp, minimum E-value of 1e-03, sequence subject coverage of 50% and sequence query coverage of 50%; a dash indicates the absence of a one-to-one ortholog) is reported. The filamentous brown alga *Ectocarpus siliculosus* (Ectocarpales, Ectocarpaceae) was used as an outgroup. Putative functions are indicated on the right. Top rows: Pyth., Pythiales; Lag., Lagenidiales; Alb., Albuginales; HB, hemibiotrophic lifestyle; OB, obligate biotrophic; S, saprotrophic; N, necrotrophic. Species names, *Phytophthora* clade names and group of species names are indicated on the bottom; numbers between brackets indicate the number of species considered in a group.

711x558mm (96 x 96 DPI)



Figure S1. Box plots showing amounts of glycoside hydrolases, sorted by substrate, in *Phytophthora* species and categorized into host ranges 'Narrow' (1 plant genera), 'Multiple' (2-9 plant genus), 'Wide' (16-55 host genera), and 'Huge' (107-327 host genera). Glycoside hydrolases pollysaccharides, pectins, cellulose, and hemicellulose are shown (A). For protease inhibitors, I01, I25B, I43, I51, I63, and I87 are shown (B). Near identical paralogs were removed; proteins with greater than 95% amino acid identity over the full sequence length were reduced to one representative effector sequence.

762x609mm (96 x 96 DPI)



Figure S2. Number of HGT candidates (transcripts) retained for 37 *Phytophthora* genomes after false positive filtering of candidates initially identified with a Support Vector Machine classifier with two "phylogenetic" and "identity" tests. The total number of transcripts predicted for each Phytophthora genome is indicated above bars.

855x481mm (38 x 38 DPI)

Figure S3. Horizontal gene transfer for 19 candidate genes in *Phytophthora* spp. A, Maximum likelihood phylogeny showing relationships between candidate HGT gene(s) (indicated by three asterisks) and closest relative sequences found within seven clades (Heterokonta, Alveolata, Viridiplantae, Fungi, Amoebozoa, Archaea and Bacteria); B. expected phylogeny according to Burki et al. (2020) and Keeling et al. (2019). C and D, sequence identity identity between the candidate HGT sequence in *Phytophthora* and its closest homolog sequences in the putative "donor" species; the distribution represent the expected nucleotide identity values for the two same species calculated from a random sample of 1000 sequences; D, the arrow indicates the sequence identity between the candidate HGT sequence and the closest species in the non-donor clades; the distribution represents the expected nucleotide identity values for the two same from a random sample of 1000 sequences. Bottom table, Shimodaira-Hasegawa test (SH-test) between the observed ML tree and alternative tree topologies.

Burki, Fabien et al. 2020. The New Tree of Eukaryotes. Trends in Ecology & Evolution, 35: 43 – 55.

Keeling, Patrick J. et al. 2019 Progress towards the Tree of Eukaryotes. Current Biology, 29: R808 - R817.



Tīcapadiggy	-Ln	D(-Ln)	Pacelice
Closebseedettptbodbgy	242545533339	0	ns
5-Epeqteadtoftipagjagy	25200004948	55447/4488	≪000011
Fungi bFicendrbedawahheindHweiheindHoentelao(kHoentea(kHoented)abna)	2429103031616	66777722	≪000011

Fig. S3_1 - HGT_1 (Endonuclease)



Třapadigy	-Ln	D(-Ln)	PaedLee
ChOsebseendettiptingtygy	8801338966	0	ns
5 Epeqteakt/thaj ag/	88900444	140002557	≪000011
Fungi bFuendhedenvalheind-keenteka (k-bentad(k-heat)dona)	8877888800	77/40041	≪000011
Haladi bartaladikaanat badawat hara Fwiltigi (Fugia (Folomgoal) donor)	8811831.133	1442277	ns

Fig. S3_2 - HGT_2 (Xylulose reductase)





0

Nucleotide identity (%)

Ticquodiggy	-Ln	D(-Ln)	Paællæ
CliCeebseedettptbogbygy	1312352657474	0	ns
EEpeqtecatedptpagyagy	1-314384787222	2222114477	≪000011
eack-battadkoanatbadavndthandFwittgi (Fungia (Fo lomgoai)donor)	1313637679292	1400221177	≪000011

Fig. S3_3 - HGT_3 (Quinone oxidoreductase)



Ticquadiggy	-Ln	D(-Ln)	Paaillae
ChOadaseadatiptingbgy	7712233955	0	ns
5-Epeqtecelohttpacijog/	7-72200003	90601077	≪000011
BateiBlocalandirladawidferd-Welhirdk-bertead(-koertadi(-hoet)donor)	771199227244	00007/08	≪000011
H ttd/-bettadizondhtdzwitterBvätteitBB(Zztata(Braat) donor)	7711224/239	-0.33	ns

Fig. S3_4 - HGT_4 (Aquaporin)



Nucleotide identity (%)

Ттаранду	-Ln	D(-Ln)	PAedice
CiOstosenietițtingbg	101606668222	0	ns
EEpeqteadotttaajag	10170472423338	-74.1	≪000011
BadeiBibadentidhebidewittierd-Welthind-bertead(-koentad)(-hoet)(donor)	10170473455555	7653366	≪000011

Fig. S3_5 - HGT_5(NPP1)



Trappatiggy	-Ln	D(-Ln)	Preduce
CbCechraeed.etchtboch.gy	892255221133	0	ns
BadBeziadaenidal-laredi-hrethompohnydephicyleic	8886009986	440999933	≪000011
Fungi bFuandhedawatheindHwaiheindHoentelao(kHoentad(dHeot);bno)	88339911.1155	1-133990022	≪000011
eack-bartackcarca/backavrdthairdFwittgi(Fungia(Foluorngcail)donor)	882265772283	1455115522	ns

Fig. S3_6 - HGT_6 (Phenol acid carboxylase)



Ticquettgy	-Ln	D(-Ln)	Predite
CtOstosendetistinging	8811822850	0	ns
BacBasiadamidal-lated-methompothydapticyletc	8921611239	77884400	≪000011
Fungi bFuardbadavdheird-Ivalheird-Ivarteta(I-herta(I-hert),bro)	822310316	5633996	≪000011
Haladk-bartaladkaanal-badavndthaindFwithgi(Fungia(Foluorngoal)donor)	8822558333	4422993	≪000011

Fig. S3_7 - HGT_7 (Peptidase S9)





TTaqquatiqgy	-Ln	D(-Ln)	Paællæ
CtOwnswedettiptingtygy	16116317370608	0	ns
5Epeqtexdbdtfacg/gy	18188756757373	4339655	≪000011
Bate Block nicht der Weltrick bertauf (Kertauf (Kertauf (Kertauf)	1813823/2/005	1880006	≪000011
Head beten b	18118410402536	-3.30	ns

Fig. S3_8 - HGT_8 (UDP-N-acetylglucosamine-peptide N-acetylglucosaminyltransferase)



Nucleotide identity (%)

0

Tiaquattagy	-Ln	D(-Ln)	Preduce
ChOadasendettyttingtygy	14946009833	0	ns
54 page and the second se	14996040077	1-(32214	ns
Fungi bFitandhedanvalheind-kaetetan(k-bentad(k-hedt)boo)	1-1998622	1455339	ns
HaadHoantaadkoancalbadavnatheindFvuthgijFurgia[Fduarngoal)donor)	1499996006	-5.23	ns

Fig. S3_9 - HGT_9 (Alternative oxidase)



Nucleotide identity (%)

Ticappottbggy	-Ln	D(-Ln)	Pacellee
CbCeabsead.ettptbogb.gy	1218278773636	0	ns
EFEpeqteattoptproglagy	1312352553434	3377779988	≪000011
FungibFuandhedanvahheindHwalheindHoentetao((Hbentaa((Hheat)ubno))	1219239337575	5663388	≪000011
ead+batadkcanahadavndhandFwthgi(Fungia(Fotcangcal)donor)	1218281818484	-4.47	ns

Fig. S3_10 - HGT_10 (Dioxygenase)



Tī сīpatīgy	-Ln	D(-Ln)	Pacellee
ChCarbsersiethttingbgy	220899738	0	ns
E-Epeqteadottingjag/	2020595500	5660144	≪000011
AmoebAam baarbaahdaavadrid-Webrok tertead(-toet ad)-toetad(-toetad)	20205863091	6622011	≪000011
H adil betalakundi laukundi laukundi laukundi laukundi laukundu la	2000897689	000000011	ns

Fig. S3_11 - HGT_11 (Thioesterase)



-Ln	D(-Ln)	Paæliæ
2421421323636	0	ns
2423123201747	1499771100	≪000011
2426405078484	448844488	≪000011
24234531617979	223389422	≪000011
	-Ln 24242023666 242342302447 24264600494 2423463667979	Ln D(-Ln) 242420233666 0 242542320447 149977100 24260507884 448944486 24236336799 223884422

Fig. S3_12 - HGT_12 (Putative xylanase)





Nucleotide identity (%)

Trappediggy	-Ln	D(-Ln)	Pvæilæ
CiOxinserdetiptinghgy	7-722853344	0	ns
BEpaqteadkhittpagay	746814333366	901752	≪000011
CtjCBrdycBasictweithwHttHosikanctHota	7461122538	901 5224	≪000011
OtjOFnJrgFlwrgfiwHittHasteardeana	7-788222211	9 87119 7711	≪00011
FurgEFaurtybBaachabdavtifterd-Webbiok-berta(d-kastitac(t-bat).donor)	7-777558900	40000	≪00011
H#ddHettbookondtedavattitteeFuvigtBeFatrigtBegBEfFatrigtBrazidonor)	7-722891159	-2.84	ns

Fig. S3_13 - HGT_13 (4-coumarate CoA ligase)



TRappotingay		D(-Ln)	Pwallue
CiCateseridittetagigy	1411415165464	0	ns
BEpeqtectbrittingjog	14(26(3))/2/2	14444008	≪000011
BatelBAardreakeeAurainaarheadhwaittiined-lawkindid-hekketideardeade/beitdeanor)	14162251414	3349975	≪000011
Händel Haakanoor Handwall hai helta valitään Persona taasia (Persona Handhandhandhandhandhandhandhandhandhandh	1411011312227	1446233	ns

Fig. S3_14 - HGT_14 (2,4-dienoyl-CoA reductase)



Nucleotide identity (%)

Ticappatitggy	-Ln	D(-Ln)	Paallae
CiOabseadettptingbgy	120250004	0	ns
ExEpeqterationation	1217297091212	4441088	≪000011
BateiBlacencinedawitherd-weblind-beneac(-foentab(-hoen)donor)	121333366252	337.448 6	≪000011
H ttol betelakandadakuttiin:Bvättiin:B/Aztoliaj(Btartai) adonor)	1213233551116	-1.55	ns

Fig. S3_15 - HGT_15 (Alcohol dehydrogenase)



Třappatiggy	-Ln	D(-Ln)	Paailae
CiOabseadatttiagigy	33/199/15511	0	ns
EEpeqteceltatiog/gg	3210210392225	142230055	≪000011
BadeieBaderwichebalewicherd-Weltrick-bertead(-koertadi(-tras)/donor)	32321014/4/44	90000077	≪000011
H add be teadkaon di ladawiti in Bwithi in BaBatala (Braat) donor)	323624647676	33442000	≪000011

Fig. S3_16 - HGT_16 (Putative tannase)


-Ln	D(-Ln)	Paællæ
2208054099	0	ns
292292995553	2233554444	<90001
2920907874949	33334400	<90001
229007/5583	22117759	<90065
	-Ln 2806054009 20239290555 290097558	-Ln D(-Ln) 20195054009 0 20295095553 23354444 202095096090 3333400 2020975556 2217799

Fig. S3_17 - HGT_17 (Putative pectinesterase)



Trappatiggy	-Ln	D(-Ln)	Predice
ChCesbaeedettptbocbgy	16186162727	0	ns
BEpeqtexteloptiong/og/	161933959222	144223986	<90001
H aad-koentaalakoemta bada witheind Fwithgi (Frumggi Founggi donor)	1711731230202	3311557755	<00001
Fungi bFuandbadanwatheindHwatheindHocenteac(kHoentad(xHeot)dono)	1618285264545	14001188	ns

Fig. S3_18 - HGT_18 (ATP-binding Cassette [ABC])



Trappatiggy	-Ln	D(-Ln)	Pvzłilize
ChOstoseadettytingby	4488377552	0	ns
5 6psqbadkittjag igy	44 E 9722894	3315322	<90065
Hand headadaan da adaan da harina ayaa ayaa ayaa ayaa ayaa ayaa ayaa a	44994339533	-6.31	ns
BateiBatean circle dawid ferd-wethick tenteral (-treated (-treat) donor)	44897660097	33005066	<90001

Fig. S3_19 - HGT_20 (Ribosomal-protein-alanine acetyltransferase)



Figure S4. Identification of the radial basis function (rbf) kernel parameters (*C* and *gamma*) resulting in the highest accuracy in classifying Phytophthora, bacteria and fungal transcripts with a Support Vector Machine (SVM) classifier based on Sequence-composition features (GC-content, codon usage). Accuracy values were obtained with a 5-fold cross-validation of the SVM classifier. The value in white represents the highest accuracy obtained.

855x481mm (38 x 38 DPI)



Figure S5. Relationships between the false positive rate (FP) and the true positive rate (TP) for HGT candidate discovery with a Support Vector Machine classifier and the number of bootstrap replicate datasets in which the putative HGT candidate was found. The grey area corresponds to the number of bootstrap replicates where the false positive rate reaches the null value.

452x193mm (236 x 236 DPI)

Supplementary Table 1. Full assembly and annotation statistics for 31 Phytophthora genomes.

				Gen	ome			
Species	Group	Clade	Sequence	Contigs	N50	Assembly	K-mer size	Sequence
	•		Reads			Length	estimate	Reads
P. cactorum	EMR	1	NA ^a	7,888	15,053	56,443,298	NA ^a	28,693,892
P. idaei	EMR	1	NA ^a	7,163	14,461	53,468,943	NA ^a	NA^{b}
P. pini	USDA-D	2	29,016,981	2,131	42,987	38,730,000	53,447,651	24,861,861
P. multivora	Scion	2	NA ^a	2,844	46,133	40,059,192	NA ^a	NA^{b}
P. pluvialis	Scion	3	NA ^a	4,340	30,816	53,616,150	NA ^a	NA ^b
P. litchii	NJAU	4	17,809,306	2,543	34,546	38,200,938	58,822,528	20,209,114
P. palmivora	USDA-B	4	44,740,938	24,815	6,694	107,798,747	178,336,461	21,955,557
P. megakarya	USDA-B	4	29,074,100	24,073	7,093	101,609,312	170,257,105	19,683,391
P. agathidicida	Scion	5	NA ^a	3,754	19,544	37,337,699	NA ^a	NA^{b}
P. taxon totara	Scion	5	NA ^a	4,425	30,809	55,576,372	NA ^a	NA ^b
P. parvispora	OSU	7	21,793,750	9,906	6,820	46,825,958	72,690,845	12,813,391
P. pisi	Fribourg	7	34,621,167	7,667	15,253	58,856,683	105,905,508	18,243,326
P. niederhauseri	Fribourg	7	39,377,584	26,463	4,805	90,270,009	126,036,921	20,314,715
P. robiniae	NJAU	7	34,689,250	14,865	8,754	69,938,814	107,714,597	12,626,895
P. cajani	OSU	7	26,013,994	18,255	5,113	64,854,085	108,588,397	12,589,952
P. vignae	OSU	7	25,708,667	10,330	8,363	56,137,732	85,443,060	12,598,323
P. melonis	NJAU	7	31,728,667	11,353	15,342	73,416,743	106,514,211	22,052,362
P. pistaciae	OSU	7	41,390,989	10,414	10,302	63,209,321	80,941,143	12,493,625
P. europaea	OSU	7	22,034,445	8,301	11,551	58,787,065	71,680,138	12,832,217
P. uliginosa	OSU	7	21,317,084	8,955	10,095	57,072,031	72,763,188	12,712,782
P. fragariae	USDA-G	7	32,666,667	8,544	20,362	76,969,737	100,354,493	13,274,875
P. rubi	USDA-G	7	38,666,667	9,434	17,808	74,863,594	119,201,141	13,035,132
P. pinifolia	UBC	6	NA ^a	22,610	6,021	74,478,861	NA ^a	36,047,200
P. lateralis	UBC	8	23,158,308	28,263	2,396	50,496,828	NA ^a	23,158,308
P. hibernalis	OSU	8	26,010,981	6,587	21,408	71,256,216	86,155,007	12,973,692
P. foliorum	OSU	8	16,677,112	5,320	15,800	48,973,082	56,233,540	12,894,827
P. brassicae	Fribourg	8	31,477,833	12,447	12,337	72,849,437	105,501,807	20,290,544
P. syringae	USDA-G	8	19,410,417	6,572	15,987	57,045,526	66,661,136	12,799,213
P. cryptogea	UBC	8	NA ^a	25,944	4,730	69,446,343	NA ^a	NA^b
P. boehmeriae	NJAU	10	34,703,473	2,866	41,917	39,747,814	NA ^a	NA^{b}
P. kernoviae	UBC	10	21,178,252	13,710	5,225	42,698,878	NA ^a	21,178,252

^aGenome assemblies were completed in a different study.

^bRNA was not sequenced.

Transcriptome					Anno	tations	
Assemble	Assembl	Contigs after	N50 after	Assembly	Repeat	Predicted	
d Contigs	ed N50	TransDecoder	TransDecoder	Length	Percent	Genes	Strain
28,465	1,890	25,690	2,595	47,486,039	19.96	18,027	P414
NA^{b}	NA^{b}	NA ^b	NA ^b	NA ^b	16.19	18,038	SCRP371
20,548	2,695	22,766	3,818	59,507,934	7.31	14,019	Unknown
NA^{b}	NA^{b}	NA ^b	NA ^b	NA ^b	10.86	13,682	NZFS 3378
NA^b	NA^{b}	NA ^b	NA ^b	NA ^b	16.04	16,285	LC9-1
28,496	2,139	28,596	3,087	60,642,271	5.98	12,391	SHS3
30,410	1,745	26,805	2,443	45,279,452	29.62	37,283	sbr112.9
31,088	1,802	28,293	2,614	49,914,387	31.94	33,614	zdho120
NA^{b}	NA^{b}	NA ^b	NA ^b	NA ^b	5.76	12,923	NZFS 3772
NA^b	NA^{b}	NA ^b	NA ^b	NA ^b	16.56	17,619	NZFS 3642
108,991	833	65,896	1,572	70,613,891	8.75	15,642	P8494;23
44,946	885	32,997	1,179	31,197,744	16.80	18,953	Unknown
75,089	1,177	52,635	1,476	59,497,972	20.96	29,587	Unknown
38,510	1,432	33,462	2,346	52,628,916	25.58	23,128	Unknown
33,359	2,139	36,595	2,878	76,345,822	20.65	19,840	P3105;27
41,393	1,596	37,102	2,552	62,795,853	17.45	18,535	P3019;24
28,272	1,873	26,842	2,618	48,278,380	25.93	21,276	Unknown
34,064	2,320	39,208	3,366	91,860,426	18.56	19,423	P6196;25
38,785	2,066	36,606	3,233	76,609,174	23.33	17,117	P10324;31
31,138	2,115	34,305	3,184	74,176,007	24.16	17,226	P10413;26
29,266	1,848	28,012	2,725	52,169,825	30.81	20,448	CBS 209.46
34,446	1,774	31,999	2,677	57,725,263	29.48	23,476	pd0101050015038
38,654	2,461	47,683	3,693	124,451,965	33.00	23,717	CBS 122922
13,645	670	8,094	1,690	10,528,234	23.39	19,503	CBS 168.42
36,774	2,899	49,849	4,759	150,938,535	32.46	23,578	P0647;28
27,112	2,358	32,393	3,607	80,179,397	19.26	16,083	P10974;29
35,153	1,485	27,252	2,207	42,057,210	28.39	26,010	Unknown
29,481	2,735	35,799	4,253	99,905,260	21.71	18,234	PSY-09-046
NA ^b	NA^b	NA ^b	NA ^b	NA ^b	17.65	24,936	CBS 418.71
NA^b	NA^{b}	NA ^b	NA ^b	NA ^b	7.83	13,325	Unknown
10,461	791	7,931	1,499	9,532,089	4.15	14,322	CBS 122049

Host Strawberry Unknown Unknown Idesia polycarpa Unknown litchi Theobroma cacao Unknown Agathis australis Podocarpus totara Unknown Strawberry Raspberry Unknown Unknown Unknown Unknown Unknown Unknown Unknown Unknown Unknown

GO	Term		# terms in HGT set ²	Pr(<i>X=k</i>) ³
Biological proc	ess			
GO:0034079	butanediol biosynthetic process	28	6	<0.001
GO:0009405	pathogenesis	1538	6	<0.001
GO:0055114	obsolete oxidation-reduction process	1543	4	<0.001
GO:0046210	nitric oxide catabolic process	34	1	<0.05
GO:0051301	cell division	817	2	<0.05
GO:0000462	maturation of SSU-rRNA from tricistronic rRNA transcript	431	2	<0.01
GO:0071500	cellular response to nitrosative stress	54	1	<0.05
GO:0048285	organelle fission	64	1	<0.05
GO:0009245	lipid A biosynthesis	116	1	<0.05
GO:0051409	response to nitrosative stress	102	1	<0.05
GO:0045493	xylan catabolic process	151	1	<0.05
GO:0046514	ceramide catabolic process	174	1	<0.05
GO:0016024	CDP-diacylglycerol biosynthetic process	162	1	<0.05
GO:0009072	aromatic amino acid family metabolic process	198	1	<0.05

Supplementary Table 3. Over-represented GO terms for a set of 551 candidate HGT transcripts found in 35 *Phytophthora* genomes.

Molecular function

GO:0000721	(R,R)-butanediol dehydrogenase activity	28	6	<0.001
GO:0016831	carboxy-lyase activity	201	18	<0.001
GO:0016491	oxidoreductase activity	10,881	14	<0.001
GO:0008080	N-acetyltransferase activity	1,082	4	<0.001
GO:0050525	cutinase activity	110	2	<0.001
GO:0015267	channel activity	742	3	<0.001
GO:0016787	hydrolase activity	8,099	6	<0.01
GO:0003724	RNA helicase activity	505	2	<0.01
GO:0008759	UDP-3-O-[3-hydroxymyristoyl] N-acetylglucosamine deacetylase activity	61	1	<0.05
GO:0005344	oxygen transporter activity	61	1	<0.05
GO:0004838	tyrosine transaminase activity	72	1	<0.05
GO:0097573	glutathione oxidoreductase activity	82	1	<0.05
GO:0031176	endo-1,4-beta-xylanase activity	111	1	<0.05
GO:0008941	nitric oxide dioxygenase activity	109	1	<0.05
GO:0004731	purine-nucleoside phosphorylase activity	110	1	<0.05
GO:0019825	oxygen binding	118	1	<0.05
GO:0004300	enoyl-CoA hydratase activity	151	1	<0.05
GO:0010181	FMN binding	1595	2	<0.05
GO:0102121	ceramidase activity	162	1	<0.05
GO:0017040	N-acylsphingosine amidohydrolase activity	177	1	<0.05

GO:0004605	phosphatidate cytidylyltransferase activity	244	1	<0.05
GO:0004318	enoyl-[acyl-carrier protein] reductase (NADH) activity	289	1	<0.05

Cellular component

 GO:0005576	extracellular region	4624	6	<0.001
GO:0005835	fatty acid synthase complex	271	1	<0.05
GO:0032040	small-subunit processome	704	2	<0.05
GO:0032153	cell division site	94	1	<0.05
 GO:0005730	nucleolus	2223	2	<0.05

¹Transcriptome of 35 *Phytophthora* genomes.

²551 HGT candidates.

³Probability (Q-value) of obtaining the same number of transcripts (k) or more by chance as given by a hypergeometric probability distribution.

Table S4. Putative pathogenicity functions inferred for 28 HGT candidates identified among *Phytophthora* spp.

	Ortho group	Putative function	Best hit on PHI-base (E-value) - Reference	Notes - Other reference(s)
HGT_1	OG5_134265	Endonuclease	PHI:5754, endonuclease, <i>Fusarium</i> graminearum (3.0E-024) - Yun et al. (2015)	-
HGT_2	OG5_126928_1	Xylulose reductase	PHI:2256, xylitol dehydrogenase, <i>Parastagonospora nodorum</i> (1.0E-128) - Lowe et al. (2008)	Oxidation of xylitol to xylulose; alters plant defense responses against aphids - MacWilliams et al. (2020) L-arabinose metabolism, degradation of hemicellulose - Dimarogona and Topakas (2016)
HGT_3	OG5_130201	Quinone oxidoreductase	No hit	Protection against oxidative stress in entomopathogenic fungi (Pedrini et al. 2015); Fungal protection against destructive host-produced quinones (Petrasch et al. 2019)
HGT_4	OG5_126615	Aquaporin	PHI:7047, water channel protein aquaporin, <i>Cryptococcus neoformans</i> (4.0E-012) - Meyers et al. (2017)	-
HGT_5	OG5_133138	NPP1	No hit	Necrosis-inducing <i>Phytophthora</i> protein - Seidl and Van den Ackerveken (2019)
HGT_6	OG5_177086	Phenol acid carboxylase	No hit	Upregulated expression during <i>Phytophthora cactorum</i> x strawberry interaction - Chen et al. (2011)
HGT_7	OG5_152798	Peptidase S9	No hit	Abundant in fungal plant pathogens - Muszewska et al. (2017)
HGT_8	OG5_144202	UDP-N-acetylglucosamine- peptide N- acetylglucosaminyltransferase	PHI:4921, flagellin glycosyltransferase, <i>Burkholderia cenocepacia</i> (1.0E-021) - Khodai-Kalaki et al. (2015)	
HGT_9	OG5_140249	Alternative oxidase	No hit	-
HGT_10	OG5_139476	Dioxygenase	No hit	-

HGT_11	OG5_130394	Thioesterase	PHI:4988, sfp-type 4'-	-
			phosphopantetheinyl transferase,	
			Bipolaris maydis (3.0E-005) - Zainudin et	
			al. (2015)	
HGT_12	OG5_135408	Endo-1,4-beta-xylanase <u>GH10</u>	PHI:7912, endo-beta-1,4-xylanase	-
			Phytophthora parasitica (1.0E-137) - Lai	
			and Liou, (2018)	
HGT_13	OG5_126609	4-coumarate CoA ligase	PHI:10606, long-chain-fatty-acid–Co	-
			Aligase, Pseudomonas aeruginosa (1.0E-	
			032) - Pan et al. (2020)	
HGT_14	OG5_135743	2,4-dienoyl-CoA reductase	PHI:8134, 3-Oxoacyl-[acyl-carrier-	-
			protein] reductase, Salmonella enterica	
			(1.0e-014) - Kwan et al. (2018)	
HGT_15	OG5_126928_2	Zinc-binding dehydrogenase,	PHI:8321, gluconate 5-dehydrogenase,	-
		Polyketide synthase,	Salmonella enterica (1.0E-016) - Alves	
		enoylreductase domain	Batista et al. (2018)	
HGT_16	OG5_135419	Putative tannase	PHI:10222, feruloyl esterase, Valsa mali	-
			(4.0E-030) - Xu et al. (2018)	
HGT_17	OG5_133550	Putative pectinesterase <u>CE8</u>	PHI:278, pectin methylesterase, Botrytis	-
			cinerea (4.0E-077) - Valette-Collet et al.	
			(2003)	
HGT_18	OG5_223601	ATP-binding Cassette (ABC)	No hit	Putative virulence factor - Stergiopoulos et al. (2003);
				Zeng and Charkowski, (2021)
HGT_19	OG5_161673	Antibiotic biosynthesis	No hit	-
		monooxygenase		
HGT_20	OG5_132142	Ribosomal-protein-alanine	No hit	-
		acetyltransferase		
HGT_21	OG5_136477	Endoribonuclease	No hit	-
HGT_22	OG5_132299	Pectate lyase PL3	PHI :4620, pectate lyase, Phytophthora	-
			<i>capsici</i> (1.0E-029) – Fu et al. (2015)	
HGT_23	OG5_203245	Csa-calmodulin	No hit	-

HGT_24	OG5_126661	Mannitol-1-phosphate	PHI:9524, Alcohol dehydrogenase,	-
		dehydrogenase	Candida albicans (4.0E-021)- Song et al.	
			(2019)	
HGT_25	OG5_137179	hypothetical protein	No hit	-
HGT_26	OG5_129399	2-nitropropane dioxygenase	No hit	Triggers expression of type III secretion system in
				Ralstonia solanacearum to induce pathogenicity on
				tobacco - Zhang et al. (2017)
HGT_27	ORTHOMCL13162	hypothetical protein	No hit	-
HGT_28	OG5_126721	type I methionyl	PHI:9334, effector protein, Citrobacter	-
		aminopeptidase	rodentium (4.0E-020) - Xia et al. (2019)	

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Valette-Collet O, Cimerman A, Reignault P, Levis C, Boccara M. 2003. Disruption of *Botrytis cinerea* pectin methylesterase gene bcpme1 reduces virulence on several host plants. *Molecular Plant-Microbe Interactions* **16**: 360-367.

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Supplementary Table 5. Putative pathogenicity functions inferred for 28 HGT candidates identified among *Phytophthora* spp.

	Ortho group	Putative function	Best hit on PHI-base (E-value) - Reference	Notes - Other reference(s)
HGT_1	OG5_134265	Endonuclease	PHI:5754, endonuclease, <i>Fusarium graminearum</i> (3.0E-	-
HGT_2	OG5_126928_1	Xylulose reductase	024) - Yun et al. (2015) PHI:2256, xylitol dehydrogenase <i>, Parastagonospora</i> <i>nodorum</i> (1.0E-128) - Lowe et al. (2008)	Oxidation of xylitol to xylulose; alters plant defense responses against aphids - MacWilliams et al. (2020) L-arabinose metabolism, degradation of hemicellulose - Dimarogona
HGT_3	OG5_130201	Quinone oxidoreductase	No hit	and Topakas (2016) Protection against oxidative stress in entomopathogenic fungi (Pedrini et al. 2015); Fungal protection against destructive host- produced quinones (Petrasch et al. 2019)
HGT_4	OG5_126615	Aquaporin	PHI:7047, water channel protein aquaporin, <i>Cryptococcus neoformans</i> (4.0E-012) - Meyers et al. (2017)	-
HGT_5	OG5_133138	NPP1	No hit	Necrosis-inducing <i>Phytophthora</i> protein - Seidl and Van den Ackerveken (2019)
HGT_6	OG5_177086	Phenol acid carboxylase	No hit	Upregulated expression during <i>Phytophthora cactorum</i> x strawberry interaction - Chen et al. (2011)
HGT_7	OG5_152798	Peptidase S9	No hit	Abundant in fungal plant pathogens - Muszewska et al. (2017)
HGT_8	OG5_144202	UDP-N-acetylglucosamine-	PHI:4921, flagellin glycosyltransferase, Burkholderia	-
_	_	peptide N- acetylglucosaminyltransferase	cenocepacia (1.0E-021) - Khodai-Kalaki et al. (2015)	
HGT_9	OG5_140249	Alternative oxidase	No hit	-
HGT_10	OG5_139476	Dioxygenase	No hit	-
HGT_11	OG5_130394	Thioesterase	PHI:4988, sfp-type 4'-phosphopantetheinyl transferase, Bipolaris maydis (3.0E-005) - Zaipudin et al. (2015)	-
HGT_12	OG5_135408	Endo-1,4-beta-xylanase	PHI:7912, endo-beta-1,4-xylanase <i>Phytophthora parasitica</i> (1.0E-137) - Lai and Liou, (2018)	-
HGT_13	OG5_126609	4-coumarate CoA ligase	PHI:10606, long-chain-fatty-acid–Co Aligase, <i>Pseudomonas</i> <i>aeruginosa</i> (1.0E-032) - Pan et al. (2020)	-
HGT_14	OG5_135743	2,4-dienoyl-CoA reductase	PHI:8134, 3-Oxoacyl-[acyl-carrier-protein] reductase, Salmonella enterica (1.0e-014) - Kwan et al. (2018)	-
HGT_15	OG5_126928_2	Zinc-binding dehydrogenase, Polyketide synthase, enoylreductase domain	PHI:8321, gluconate 5-dehydrogenase, <i>Salmonella enterica</i> (1.0E-016) - Alves Batista et al. (2018)	-
HGT_16	OG5_135419	Putative tannase	PHI:10222, feruloyl esterase <i>, Valsa mali</i> (4.0E-030) - Xu et al. (2018)	-
HGT_17	OG5_133550	Putative pectinesterase	PHI:278, pectin methylesterase, <i>Botrytis cinerea</i> (4.0E-077) - Valette-Collet et al. (2003)	-
HGT_18	OG5_223601	ATP-binding Cassette (ABC)	No hit	Putative virulence factor - Stergiopoulos et al. (2003); Zeng and Charkowski, (2021)
HGT_19	OG5_161673	Antibiotic biosynthesis monooxygenase	No hit	-
HGT_20	OG5_132142	Ribosomal-protein-alanine acetyltransferase	No hit	-
HGT_21	OG5_136477	Endoribonuclease	No hit	-
HGT_22	OG5_132299	Pectate lyase	PHI :4620, pectate lyase, <i>Phytophthora capsici</i> (1.0E-029) – Fu et al. (2015)	-
HGT_23	OG5_203245	Csa-calmodulin	No hit	-
HGT_24	OG5_126661	Mannitol-1-phosphate dehydrogenase	PHI:9524, Alcohol dehydrogenase, <i>Candida albicans</i> (4.0E-021)- Song et al. (2019)	-
HGT_25	OG5_137179	hypothetical protein	No hit	-
HGT_26	OG5_129399	2-nitropropane dioxygenase	No hit	Triggers expression of type III secretion system in <i>Ralstonia</i> <i>solanacearum</i> to induce pathogenicity on tobacco - Zhang et al. (2017)
HGT_27	ORTHOMCL13162	hypothetical protein	No hit	-
HGT_28	OG5_126721	type I methionyl aminopeptidase	PHI:9334, effector protein, <i>Citrobacter rodentium</i> (4.0E-020) - Xia et al. (2019)	-

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