Running title: Domain analysis in oomycete plant pathogens

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A domain-centric analysis of oomycete plant pathogen genomes reveals unique protein organization

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ABSTRACT

Oomycetes comprise a diverse group of organisms that morphologically resemble fungi but belong to the stramenopile lineage within the supergroup of chromalveolates. Recent studies have shown that plant pathogenic oomycetes have expanded gene families that are possibly linked to their pathogenic lifestyle. We analyzed the protein domain organization of 67 eukaryotic species including four oomycete and five fungal plant pathogens. We detected 246 expanded domains in fungal and oomycete plant pathogens. The analysis of genes differentially expressed during infection revealed a significant enrichment of genes encoding expanded domains as well as signal peptides linking a substantial part of these genes to pathogenicity. Overrepresentation and clustering of domain abundance profiles revealed domains that might have important roles in host-pathogen interactions but, as yet, have not been linked to pathogenicity. The number of distinct domain combinations (bigrams) in oomycetes was significantly higher than in fungi. We identified 773 oomycete-specific bigrams, with the majority composed of domains common to eukaryotes. The analyses enabled us to link domain content to biological processes such as host-pathogen interaction, nutrient uptake, or suppression and elicitation of plant immune responses. Taken together, this study represents a comprehensive overview of the domain repertoire of fungal and oomycete plant pathogens and points to novel features like domain expansion and species-specific bigram types that could, at least partially, explain why oomycetes are such remarkable plant pathogens.
INTRODUCTION

Oomycetes are a diverse group of organisms that live as saprophytes or as pathogens of plants, insects, fish, vertebrates, and microbes (Govers and Gijzen, 2006). The numerous plant pathogenic oomycete species cause devastating diseases on many different host plants and have a huge impact on agriculture. A prominent example is *Phytophthora infestans*, the causal agent of late blight of potato and tomato, and responsible for the Irish potato famine in the 19th century. Plant pathogenic oomycetes include a large number of different species that vary in their lifestyle, from obligate biotrophic, hemi-biotrophic to necrotrophic. In addition, they show great differences in host selectivity ranging from broad to very narrow (Erwin and Ribeiro, 1996; Agrios, 2005). Oomycetes have morphological features similar to filamentous fungi and the two groups exploit common infection structures and mechanisms (Latijnhouwers et al., 2003). Together with diatoms, brown algae and golden-brown algae, oomycetes are classified as stramenopiles, a lineage that is united with alveolates in the supergroup of chromalveolates (Baldauf et al., 2000; Yoon et al., 2002). The monophyly of this supergroup however is under debate (Baurain et al., 2010). The genomes of oomycetes sequenced so far are variable in size and content, ranging from 65 Mb in *Phytophthora ramorum* to 240 Mb in *P. infestans* (Haas et al., 2009) and only include plant pathogenic species. Analysis of these genomes revealed that several gene families facilitating the infection process are expanded (Martens et al., 2008). Extreme examples are gene families encoding cytoplasmic effector proteins such as RXLR effectors that share the host cell-targeting motif RXLR and suppress defense responses in the host, and the necrosis inducing proteins classified as Crinklers (Crn) (Haas et al., 2009). To date, a few oomycete genomes have been sequenced and this enables a comprehensive comparison of genomic features present in oomycetes, fungi and other eukaryotic species such as gene families and protein domains. Experimentally derived functional knowledge of the majority of gene products in oomycetes in a comparable depth as for model species like *Saccharomyces cerevisiae* and *Arabidopsis thaliana*, will likely not be accessible in the near future. Hence, comparative genomics provides an important framework to functionally characterize oomycete gene products and generate hypotheses on the basic cellular functions as well as the complex interactions of these plant pathogens with their hosts and environment.

In this study we focus on protein domains because these are the basic functional, evolutionary and structural units that shape proteins (Rossmann et al., 1974; Orengo et al., 1997; Vogel et al., 2004). Domains function independently in single domain proteins or synergistically in multi-domain proteins (Doolittle, 1995; Vogel et al., 2004; Bashton and
Chothia, 2007). Accordingly, some domains always occur with a defined set of functional partners, whereas others are highly versatile and form combinations of two consecutively occurring domains (also called bigrams) with different N- or C-terminal partners (Marcotte et al., 1999; Basu et al., 2008). Here we analyzed the domain repertoire predicted from the genome sequences of 67 eukaryotic species and compared filamentous plant pathogens to other eukaryotes with a special emphasis on oomycetes. We show how differences in the domain repertoire of oomycetes, especially in the expansion of certain domain families and the formation of species-specific bigram types, can be linked to the biology of this group of organisms. This allowed the generation of candidate sets of proteins and domains that are likely to play roles in the lifestyle of oomycetes or their interaction with plants.
RESULTS

The domain repertoire of oomycete plant pathogens and its comparison to other eukaryotes

We analyzed the domain architecture of the predicted proteomes in 67 eukaryotes covering all major groups of the eukaryotic tree of life with the exception of the supergroup Rhizaria (Fig. 1a and Tab. S1). We included seven stramenopiles, four of which are plant pathogenic oomycetes, namely the obligate biotrophic downy mildew *Hyaloperonospora arabidopsidis*, and three hemi-biotrophic *Phytophthora* spp. The selection also contained five fungal plant pathogens including rice blast fungus (*Magnaporthe grisea*) and corn smut (*Ustilago maydis*), both species with a (hemi-)biotrophic lifestyle comparable to the oomycete plant pathogens used in the analysis (Fig. 1b).

The domain architecture of all 1,250,996 predicted proteins in the 67 eukaryotic genomes was analyzed using HMMER (Eddy, 1998) and a local Pfam-A database (Finn et al., 2008). Overall, 59% (737,851) of all proteins have one or more predicted domain. We detected a total of 1,464,807 domains in all species, 80,180 within the stramenopiles and 51,030 in oomycetes.

In order to characterize the domain repertoire of eukaryotes we used two metrics: the number of domain types and the number of different combinations of adjacent domains also called bigrams (Fig. 2). In total, 13,994 bigram types were identified in the 67 eukaryotic genomes, consisting of 6,356 different domain types. As described by Basu et al. (2008) the number of bigram types increases super-linearly relative to the number of domain types with the highest numbers in multicellular organisms (Fig. 3). We observed separate clusters for metazoans, fungi and plants (including land plants and mosses). Oomycetes and fungi have similar numbers of domain types ranging from 2,000 to 2,500, however oomycetes, in particular *Phytophthora* spp., contain significantly more bigram types. The three analyzed *Phytophthora* spp. appeared to have ~50% more bigram types compared to other organisms that have similar numbers of domain types (Fig. 3, p-value = 0.00019, one sided Wilcoxon rank sum test). This even holds when we apply a more conservative approach by discarding all domain- and bigram types that occur once in each predicted proteome (Fig. S1a). We observed that the number of domain types as well as the number of bigram types increases with proteome size and reaches saturation for larger proteomes (Fig S1b & Fig S1c) (Cosentino Lagomarsino et al., 2009). Although oomycetes and in particular *Phytophthora* spp. contain a similar number of domain types as fungi, they have a larger predicted proteome (Fig S1b). However, they contain more bigram types than fungi, but less than
other species with predicted proteomes of similar size, e.g. *Drosophila melanogaster* (Fig. S1c).

**Domain overrepresentation provides a snapshot of pathogen-host interaction**

Apart from a wide and abundant repertoire of domains related to transposable elements (Haas et al., 2009), the most abundant domain types in oomycetes are similar to those in other eukaryotes (Tab. S2). Hence, absolute domain abundance alone is not indicative enough to correlate domains to the lifestyle of both fungal and oomycete plant pathogens. Instead, we identified domains that are overrepresented in plant pathogens relative to other eukaryotes (Fig. 1b).

Our analysis inferred 246 overrepresented domains in plant pathogens that are observed in 24,970 proteins. (Tab. S3, Fisher exact test, p-value <0.001, a selection of well-described overrepresented domains is depicted in Fig. 4a). Since we analyzed the expansion in plant pathogens at the level of a group rather than an individual species, domains that are reported as being expanded in the group are not necessarily expanded in all species of the group or may even be absent (Tab. S3). For example, secreted proteins encoding carbohydrate-binding family 25 domains (IPR005085) are only found in *Phytophthora* spp. and not in fungal plant pathogens whereas secreted proteins containing the cysteine rich domain (CFEM, IPR008427) are only observed in fungal pathogens (Kulkarni et al., 2003).

Many proteins involved in host-pathogen interaction are secreted in the apoplast, or, like the RXLR effector proteins, translocated into host cells following their secretion from the pathogen (Haas et al., 2009). Hence, we also predicted the presence of potential N-terminal signal peptide sequences in the whole proteomes of the analyzed species. The combined secretome encompasses 100,521 potentially secreted proteins of which 11,352 are predicted in plant pathogens (Fig. S2). Approximately 20% (2,478) of these proteins contain overrepresented domains and hence proteins containing overrepresented domains are 1.85-fold enriched in the predicted secretome of the analyzed plant pathogens (Fisher exact test, p-value $2.57 \times 10^{-231}$).

Oomycete proteins with significantly expanded domains are prime candidates for being pathogenicity-associated. To assess this hypothesis we tested if *P. infestans* genes that are differentially expressed during infection of the potato host are enriched for the aforementioned expanded domains. For this, we utilized NimbleGen microarray data that includes genome-wide expression levels of *P. infestans* genes at different days post inoculation (dpi) of potato leaves, as well as from mycelium grown *in vitro* on different media (Haas et al., 2009). We identified in total 1,584 genes that are significantly induced or
repressed in *P. infestans* during infection (differentially expressed at least one of the time points, 2-5 dpi) compared to grown in vitro (three different growth media) (Tab. S4a & Methods, T-test, p-value <0.05, q-value <0.05). Of the 1,584 differentially expressed genes, 259 encode proteins containing significantly expanded domains (Tab. S4b), which is 1.2-fold more than expected (Fisher exact test, p-value 8.8 x 10^{-5}). Moreover, 44 of these 259 genes also encode proteins with a predicted signal peptide, which is a significant enrichment (Fisher exact test, p-value 4.38 x 10^{-5}, 1.8-fold). The majority (41) of these 44 genes are differentially expressed early in infection (2 dpi) (Fig. 5a). All genes differentially expressed at 3 dpi are also differently expressed at 2 dpi (Fig. 5a & 5b). Consequently, the 44 differentially expressed genes coding for proteins with both predicted signal peptides as well as overrepresented domains are promising candidates for pathogenicity-associated proteins of which several will be discussed in detail later.

For several groups of overrepresented domains a direct or indirect role in host-pathogen interaction and/or plant pathogen lifestyle has already been hypothesized or demonstrated (e.g. (Dean et al., 2005; Tyler et al., 2006; Haas et al., 2009)). Nearly 18% of the 246 overrepresented domains belong to three groups of domains: (i) hydrolase domains; (ii) domains involved in substrate transport over membranes, e.g. the general ABC transporter-like domain (IPR003439), but also more specialized transporters of sulphate (IPR011547) and amino acids (IPR004841/ IPR013057) and (iii) domains present in peptidases, e.g. the metalloprotease type M28 domain (IPR007484) found in many secreted proteins. Of the (i) hydrolases, which encompass 9% of the overrepresented domains, the majority is present in enzymes that hydrolyse glycosidic bonds. An example is the glycoside hydrolase (GH) family 12 domain (IPR002594). This domain is observed 34 times in plant pathogens, which overall contain 91,747 domains, and 43 times in all eukaryotes that have a total of 1,464,807 domains and hence is 12.62-fold (3.66 log_{2}-fold) enriched in the plant pathogens. This domain is mainly observed in secreted proteins (27 out of 34, SignalP prediction). The majority (79%) of the GH-12 domains are found in oomycete plant pathogens and the expression of two of these hydrolase genes in *P. infestans* (PITG_08944 and PITG_16991) is significantly induced during infection of potato (Tab. S4 & Fig. 5). In total, 33 differentially expressed genes during plant infection in *P. infestans* encode proteins that contain glycoside hydrolase domains including GH-17 (IPR000490) in endo-1,3-beta-glucosidase and GH-81 (IPR005200) in beta-1,3-glucanases, as well as several members of GH-28 (IPR00743), a domain involved in soft-rotting of host tissues and described in both fungal and bacterial plant pathogens (He and Collmer, 1990; Ruttkowski et al., 1990). Twenty-eight *P. infestans* genes coding for domains involved in transmembrane transport are differentially expressed during plant infection (data in Tab. S4). Examples of genes encoding (ii) domains involved in
substrate transport over the membrane are PITG_04307 that encodes an ABC-2 type transporter (IPR013525), PITG_12808 that encodes an amino acid transporter (IPR013057) as well as PITG_22087, a gene encoding both ABC-like (IPR003439) and ABC-2 type domains (Tab S4). Extracellular degrading enzymes like cutinases contain an overrepresented domain (IPR000675) (p-value: 3.72 x 10^{-61}). This domain is observed 65 times in plant pathogenic species corresponding to a 13.3-fold (3.73 log_{2}-fold) enrichment (Fig. 4a). In total 61 proteins in plant pathogens predicted to possess this domain are potentially secreted (Methods). Another overrepresented domain that is present in secreted proteins and involved in maceration and soft-rotting of plant tissue is the pectate lyase (IPR004898). This domain is 15.34-fold (3.94 log_{2}-fold) enriched in plant pathogens and mainly found in oomycetes. Five genes in *P. infestans* encode this domain as well as a predicted N-terminal signal peptide and are differentially expressed (Fig. 5).

**Novel candidate domains significantly expanded in plant pathogens**

Next to domains that were already directly or indirectly implied in host-pathogen interaction, we identified novel candidates that are also expanded in plant pathogens, several of which are encoded in *P. infestans* genes differentially expressed during infection of the host. E.g. genes encoding the significantly expanded alcohol dehydrogenase (zinc-binding, IPR013149) as well as a GroES-like alcohol dehydrogenase (IPR013154) domains are ubiquitous in all analyzed eukaryotes and also the combination of these two domains is present in all species with only few exceptions. Nine of these genes in *P. infestans* are induced during infection (Tab. S4). Sixty-five genes in plant pathogens encode proteins with FAD-linked oxidase (IPR006094) and Berberine/berberine-like (BBE) domains (IPR012951) of which three out of six in *P. infestans* are induced during infection (PITG_02928, PITG_02930 and PITG_20764). The BBE domain is involved in the biosynthesis of the alkaloid berberine (Facchini et al., 1996). The genes encode a predicted N-terminal signal peptide, however molecular analysis of proteins containing these domains in plants indicated that at least some of these are not secreted but instead targeted to specialized vesicles (Amann et al., 1986; Kutchan and Dittrich, 1995; Facchini et al., 1996). Moreover, Moy and colleagues observed induced expression of a soybean gene (BE584185) shortly after infection with *P. sojae* containing these two domains (Moy et al., 2004). A recent analysis from Raffaele and colleagues focusing solely on the secretome in *P. infestans* corroborate ours results and also concludes that proteins with BBE and FAD-linked oxidase domains are candidate virulence factors (Raffaele et al., 2010). Three genes encoding secreted metallo-phosphoesterases (IPR004843, PITG_20454, PITG_07720, PITG_10322) show induced genes expression. These metallo-phosphoesterase domains are found in phosphatases and...
hence are involved in regulation of protein activity since they work as antagonists of kinase activity.

For approximately 6% of all overrepresented domains no or limited functional information is available in Pfam. These are the so-called DUFs: domains of unidentified function. Given their expansion in plant pathogens and the fact that other overrepresented domains are known to function in diverse aspects of plant-pathogen interactions, these DUFs are also likely to play a role in the lifestyle of plant pathogens and are hence promising targets for further experimental validation (Tab. S3). Secreted proteins containing a combination of two overrepresented DUFs, DUF2403 (IPR018807) and DUF2401 (IPR018805), are exclusively found in fungi and in oomycetes, with the majority (~75%) in oomycetes. The N-terminal DUF2403 contains a glycine rich region without further functional annotation whereas five highly conserved cysteine residues characterize the C-terminal DUF2401. Proteins containing both DUFs have been characterized in *S. cerevisiae* and in *Candida albicans* as being covalently linked to the cell wall (Terashima et al., 2002; Yin et al., 2005; Klis et al., 2009). Another overrepresented DUF within plant pathogens and mainly found in oomycetes is DUF953 (IPR010357). This domain is present in several eukaryotic proteins with thioredoxin-like function and two genes in *P. infestans* containing this domain are differentially expressed during infection (PITG_07008 and PITG_07010). DUF590 (IPR007632), which is ubiquitous in nearly all eukaryotes, is observed in proteins containing eight putative transmembrane helixes. These proteins exhibit calcium-activated ion channel activity and are involved in diverse biological processes (Yang et al., 2008). The *P. infestans* gene PITG_06653 that contains DUF590 domain is differentially expressed during infection and this provides further support for a role in host-pathogen interaction. The exemplified DUFs as well as other overrepresented domains with less or no functional annotation are interesting candidates for further functional studies to decipher their precise role in plant pathogens.

**Domain overrepresentation in oomycete plant pathogens**

Since the previous analysis grouped both fungal and oomycete plant pathogens, domains specifically enriched in oomycetes were not directly discernible. Hence, we compared the relative domain abundance predicted in plant pathogens (Fig. 1b) with the aim to identify domains specifically enriched in oomycetes. Of the 75 domains that are overrepresented in oomycetes, 20 are not observed in any fungal plant pathogen, and can therefore be considered oomycete-specific within plant pathogens (Tab. S5). In general the abundance of expanded domains in *Phytophthora* spp. is higher than in *H. arabidopsidis*. A well-described example is the NPP1 domain (IPR008701) that is present in secreted (SignalP: 122)
necrosis-inducing proteins. It shows a significant overrepresentation in oomycetes (1.68-fold (0.75 log$_2$-fold) enriched), in particular in *Phytophthora* spp., but is also observed ten times in fungal plant pathogens as well as in a few cases in non-pathogenic fungi as noted before (Gijzen and Nürnberg, 2006). *Four P. infestans* genes encoding this domain are induced early during infection (2-3 dpi) whereas a single gene (PITG_18453) is induced late (5 dpi). Several peptidases, e.g. containing the peptidase S1/S6 and C1A domains, are overrepresented compared to other plant pathogens. S1/S6 (IPR001254, 1.6-fold (0.74 log$_2$-fold)) is predicted in 91 proteins of which 67 have a predicted secretion signal, while C1A (IPR000668, 1.79-fold (0.85 log$_2$-fold)) is predicted in 78 proteins of which 31 are potentially secreted. C1A is present in several eukaryotic species but within the plant pathogenic group it is exclusively found in oomycetes. Several secreted protease inhibitors of the Kazal family containing the Kazal I1 (IPR002350) and Kazal-type (IPR011497) domain are significantly expanded in oomycetes and are within the group of analyzed plant pathogens specific to oomycetes. This suggests that they provide an increased level of protection of the pathogen against host-encoded defense-related proteases (Tyler et al., 2006). Another domain that is oomycete-specific within the plant pathogens is the Na/Pi co-transporter (IPR003841) involved in uptake of phosphate. Several other transporters that have already been described as being overrepresented in plant pathogens, e.g. the ABC-2 type transporters, are significantly expanded within oomycete plant pathogens since these species are the major contributors to the overall abundance of this domain in plant pathogens. The abundance of predicted serine/threonine-like kinase domains (IPR017442) compared to other plant pathogenic species is surprisingly high and this domain is specifically expanded in the *Phytophthora* spp. Even if several expanded domains are observed in both oomycete as well as fungal plant pathogens, the exploration of domains primarily expanded in oomycetes, e.g. certain transporter families, defense and signaling related domains, highlights functional entities that discriminate between these groups of plant pathogens.

**Clustering of abundance profiles reveals additional potential pathogenicity factors**

We extended the set of candidate domains that might be important for host-pathogen interaction beyond overrepresented domains by searching for additional domains that show presence, absence and expansion profiles similar to overrepresented domains since these domains are likely to be functionally linked or involved in similar biological processes (Pellegrini et al., 1999). We calculated a normalized profile of domain abundance and clustered similar abundance profiles using hierarchical clustering (File S1). Several clusters contained a mix of significantly overrepresented domains and domains whose expansions in plant pathogens are not significant. We exemplify this with three clusters that contain 20% of all overrepresented domains in plant pathogens (Fig. 6).
In the first cluster (Fig. 6) domains are mainly expanded in oomycete plant pathogens. The abundance of some domains in plant pathogens is too low to be identified as being overrepresented. For example, the PcF domain (IPR018570), which is present in a small ~50 amino acid-long necrosis-inducing protein found in various *Phytophthora* spp. (Orsomando et al., 2001; Liu et al., 2005), was not identified in the initial overrepresentation analysis. Also in this cluster is the sugar fermentation stimulation domain (IPR005224) that is mainly found in bacteria and involved in regulation of maltose metabolism (Kawamukai et al., 1991). In this first cluster we observed a high number (~40%) of domains without functional characterization that are mainly present in bacteria. An example is the DUF1949 (IPR015269), a domain that is only found in the three analyzed *Phytophthora* spp. This domain is observed in functional uncharacterized bacterial proteins like YIGZ in *Escherichia coli* K12 and adopts a ferredoxin-like fold (Park et al., 2004). The *Phytophthora* and bacterial proteins containing DUF1949 also contain a second, N-terminal uncharacterized protein family UPF (UPF00029, IPR001498). This domain is also found in the human protein Impact and conserved from bacteria to eukaryotes (Okamura et al., 2000). The *P. infestans* gene (PITG_00027) containing both domains is induced early in infection (Tab. S4b). Since these DUFs cluster with overrepresented domains they are promising candidates for further study.

The domains in the second cluster mainly show an expansion of the abundance in both fungal and oomycete plant pathogens. This cluster contains for example cell wall-degrading domains like cutinases, pectate lysases and other hydrolases and also the NPP1 domain that is found in necrosis-inducing proteins. The glycosyl hydrolase family 88 comprises unsaturated glucoronyl hydrolases thought to be involved in biofilm degradation and is mainly found in bacteria and fungi (Itoh et al., 2006). Interestingly, homologs are also observed in plant pathogenic bacteria e.g. *Pectobacterium atrosepticum*, in fungi, e.g. *M. grisea*, and in all three *Phytophthora* spp.

The third cluster contains domains that are not exclusively found in plant pathogens but have a broader abundance profile. This cluster includes a variety of overrepresented hydrolases, epimerases and the ABC-2 type transporter domain (IPR013525) that is observed nearly 500 times in the plant pathogenic species. Another domain that is found in this cluster is the dienelactone hydrolase domain (IPR002925), observed in all plant pathogens and also in other eukaryotic species with a high abundance in plants as well as in fungi. This domain hydrolyses denelactone to maleylacetate in bacteria (Pathak et al., 1991) and is also detected in a putative 1,3:1,4-b-glucanase from *P. infestans* that is proposed to be involved in cell wall metabolism (McLeod et al., 2003).

**Quantification of oomycete-specific bigrams**
Domains generally do not act as single entities in proteins but rather synergistically with other domains in the same protein or with domains in interacting proteins (Park et al., 2001; Vogel et al., 2004). Domains involved in signaling, sensing and generic interactions are versatile and form combinations with several different partner domains (Tab. S6a-c). As described by others, we observed that the versatility of domains is proportional to their abundance (Fig. S3) (Vogel et al., 2005). Hence we applied a weighted bigram frequency that corrects for abundance to detect domains that are promiscuous, or prone to form combinations with different partners (Basu et al., 2008). The average number of promiscuous domains in oomycetes is 424 and in Phytophthora 464. This is higher than the average number of promiscuous domains (357) over all other species (Tab. S7).

We observed that oomycetes have a higher number of bigram types than species with a comparable number of domain types (Fig. 3). We identified in total 13,994 different bigram types throughout the 67 analyzed species. The majority of these bigram types, i.e. 7,724 or 55.2%, are predicted in only a single species. In oomycetes bigram types formed by domains that are associated with transposable elements showed a high abundance (Tab. S8 & Tab. S9). We identified 1,107 bigram types occurring exclusively in plant pathogens, the majority of which (773) is only observed in the analyzed oomycetes (Tab. S10). These oomycete-specific bigram types are identified in total 1,511 times in 1,375 predicted proteins. Of the 773 oomycete-specific bigram types, 53 are present in all oomycetes (Fig. 7a). The biggest overlap in oomycete-specific domain types is observed between the Phytophthora spp., especially between P. ramorum and P. sojae. A recent analysis of domain combination in P. ramorum and P. sojae already revealed several proteins involved in metabolism and regulatory networks containing novel bigrams (Morris et al., 2009). We additionally observed in total 43 bigram types that are shared either between P. infestans and P. sojae or P. infestans and P. ramorum. However, the majority of oomycete-specific bigram types (467) is specific for a single species. The number of oomycete-specific bigram types highly exceeds the number of oomycete-specific domain types (41). Interestingly, only six of the oomycete-specific domains participate in forming the specific bigrams. Therefore, common domain types form the majority of the observed species-specific domain combinations, emphasizing the importance of novel domain combinations rather than novel domain types as a source for species-specific functionality. Even when we selectively look at the bigrams that occur at least twice in the same proteome or once in at least two different proteomes we still observe 320 bigram types that are specific to oomycetes and occur in 982 predicted proteins.

Approximately 8% of the proteins containing an oomycete-specific bigram have a predicted secretion signal (9.2% of all oomycete proteins contain a predicted secretion signal). An example that is observed in a secreted putative cysteine protease present in all analyzed
oomycetes is the combination of the Peptidase C1A domain (IPR000668) and the ML domain (IPR003172). The ML domain is known to be involved in lipid binding and innate immunity, and has been observed in plants, fungi and animals (Inohara and Nunez, 2002). The proteins containing this bigram also have a N-terminal cathepsin inhibitory domain (IPR013201) that is often found next to the Peptidase C1A domain and prevents access of the substrate to the binding cleft (Groves et al., 1996). Another bigram that is found in secreted proteins predicted in the analyzed Phytophthora spp. is the combination of the carbohydrate binding domain family 25 (IPR005085, CBM25) with a GH-31 domain (IPR000322) as well as the tandem combination of CBM25 domains N-terminal to the glycosyl hydrolase domain. The presence of the secreted CBM25 and GH-31 combination has recently been noted in Pythium ultimum (Levesque et al., 2010). We further tried to elucidate the presence of RXLR or Crn motifs in proteins containing oomycete-specific bigrams. We predicted the presence of one of these motifs using individual HMMER models for both the RXLR and the Crn motif (detailed in the methods section). We overall predicted 746 proteins containing an RXLR and 99 proteins with a Crn motif. None of these proteins are predicted to contain an oomycete-specific bigram type.

The most abundant oomycete-specific bigram type that occurs in 64 proteins is a combination of the PI3P binding zinc finger (FYVE type) and the GAF domain. The presence of this oomycete specific bigram in P. ramorum and P. sojae has been noted before (Morris et al., 2009). The GAF domain is described as one of the most abundant domains in small-molecule binding regulatory proteins (Zoraghi et al., 2004). It is present in a large number of different proteins with a wide range of cellular functions, such as gene regulation (Aravind and Ponting, 1997), light detection and signaling (Sharrock and Quail, 1989; Montgomery and Lagarias, 2002). A typical eukaryotic domain composition involving the GAF domain is N-terminal to the 3’5’-cyclic phosphodiesterase domain found in phosphodiesterases that regulate pathways with cNMPs as second messengers (Sharrock and Quail, 1989; Martinez et al., 2002). This organization is observed in total 111 times and five times in oomycetes (Fig. 7b). The GAF-FYVE bigram is either observed as a single bigram (in 53 proteins) or in combination with other domains (in 11 proteins), for example with myosin (Richards and Cavalier-Smith, 2005). In P. infestans, two genes (PITG_07627 & PITG_09293) encoding proteins with this combination are induced early during infection of the plant (Tab. S4b). A phylogenetic analysis of the GAF domain in eukaryotes and prokaryotes showed that all GAF domains in oomycetes that are involved in the fusion with FYVE exclusively cluster with prokaryotic GAF domains, whereas other GAFs also cluster with eukaryotes. Hence, this suggests a horizontal gene transfer (HGT) from bacteria to oomycetes of those GAF domains that are involved in the fusion with FYVE (Fig. 7c, Methods). The FYVE type zinc
finger is not identified in prokaryotic species, hence we suggest two independent events, namely a HGT of the GAF domain from bacteria to oomycetes and subsequently a fusion to the zinc finger domain. Horizontal gene transfer seems to play an important role in the evolution of eukaryotes (Keeling and Palmer, 2008) and recent evidence points that these events also have a significant contribution to the genome content of protists and oomycetes as they received genetic material form different sources (Richards and Talbot, 2007; Martens et al., 2008; Morris et al., 2009). Because GAF domains are known to be involved in many different cellular processes, we can only speculate about the biological function of proteins harboring the GAF – FYVE bigram. A possible function is the targeting of proteins to lipid layers by the zinc finger domain in response to second messengers sensed by the GAF domain.

Several domains involved in phospholipid signaling domain were found to be overrepresented in the filamentous plant pathogens and in particular in oomycetes. These included the phosphatidylinositol 3-/4-kinase, PIK (IPR000403), the phosphatidylinositol 4-phosphate 5-kinase domain, PIPK (IPR002498), as well as the PI3P binding FYVE. Novel domain compositions in proteins involved in phospholipid signaling and metabolism in Phytophthora spp. have been reported previously (Meijer and Govers, 2006). Signaling domains like the FYVE and the PIK, as well as domains like the IQ-calmodulin-binding domain (IPR000048) and the phox-like domain (IPR001683), form highly abundant oomycete-specific bigram types (Tab. S10). Moreover, other domains like the serine/threonin protein kinase-like (IPR017442), pleckstrin homology (IPR001849) and the DEP (IPR000591) domains are involved in several oomycete-specific bigram types, e.g. the DEP – serine/threonin protein kinase-like domain fusion is predicted in the proteomes of all analyzed oomycetes. Additionally, domains that are components of the histone acetylation-based regulatory system form oomycete-specific bigrams, e.g. the AP2 (IPR001471) and the histone deacetylase (HDAC, IPR000286) domain combination (Iyer et al., 2008) which is observed in P. ramorum as well as in P. sojae.
DISCUSSION

We predicted the domain repertoire encoded in the genomes of four oomycete plant pathogens and compared it to a broad variety of eukaryotes spanning all major groups, including several fungal plant pathogens that have a similar morphology, lifestyle and ecological niche as oomycete plant pathogens. We quantified and examined domain properties observed in oomycetes and especially emphasized differences and common themes within fungal- and oomycete plant pathogens and their probable contribution to a pathogenic lifestyle.

We observed that oomycete plant pathogens, in particular *Phytophthora* spp., have significantly higher numbers of unique bigram types compared to species with a similar number of domain types (Fig. 2a). However, oomycetes also have on average 50% more predicted genes than most of the analyzed fungi, but at the same time encode a comparable number of domain types and hence exhibit similar domain diversity (Fig. S1b). The high number of genes observed in oomycetes suggests enlarged complexity compared to fungi, which is not directly obvious from the domain diversity, but instead from the number of unique bigram types (Fig. S1c). This observation has two possible explanations: (i) the larger number of genes predicted from oomycete genomes provides the flexibility to form new domain combinations based on a limited set of already existing domains that are in quantities similar to fungi; (ii) the domain models that cover specific domains are incomplete and therefore do not provide the required sensitivity for oomycetes genomes. Hence, we would underestimate the number of observable domain types (and to a certain level the number of predicted bigram types). Additionally, oomycetes, especially *Phytophthora* spp., are no longer following the observed trend that organisms with higher number of genes (proteins) contain a larger number of domain types. Consequently they are shifted when comparing the number of predicted domain- and bigram types. Nevertheless, both possible explanations and the observed numbers allow us to conclude that oomycete genomes, especially *Phytophthora* spp., harbor a large repertoire of genes encoding different bigram types compared to species of comparable complexity and, in the case of filamentous fungi, even similar morphology.

Oomycetes and fungal plant pathogens seem to be very similar to other eukaryotes with respect to absolute domain abundance (Tab. S2) and this metric is hence not sufficiently indicative to correlate domains directly or indirectly with the pathogenic lifestyle. Therefore, we predicted overrepresented domains in plant pathogens and identified 246 domains that are significantly expanded (Tab. S3). Proteins containing overrepresented domains are significantly enriched in the predicted secretome of the analyzed plant pathogens.
corroborating the idea that expanded domain families are involved in host-pathogen interaction and these proteins are mainly acting in the extracellular space. It has to be noted that the presence of a predicted signal peptide does not necessarily mean that these proteins are found extracellularly, since some proteins are retained in the ER/Golgi and hence not secreted (Bendtsen et al., 2004).

Since we anticipate that proteins that are directly involved in host-pathogen interaction are differentially regulated upon infection, we utilized the NimbleGen microarray data of *P. infestans* (Haas et al., 2009) and identified 259 induced/repressed genes encoding proteins containing overrepresented domains. Genes containing overrepresented domains are significantly enriched within the set of differentially expressed genes containing a predicted domain. Moreover, this subset contains a significant higher abundance of genes with a predicted N-terminal signal peptide than expected. These observations highlight and corroborate the initially emerging link between domain expansion and host-pathogen interaction.

The majority of the 246 expanded domains is present in proteins that are involved in general carbohydrate metabolism, nutrient uptake, signaling networks and suppression of host responses and hence might contribute to establishing and maintaining pathogenesis (Fig. 4). The variety of overrepresented domains involved in substrate transport over membranes is of special interest. Filamentous plant pathogens and especially oomycetes exhibit a complex and expanded repertoire of these domains enabling them to absorb nutrients from their environment and host. E.g. the expression of *P. infestans* genes encoding ABC-2 like transporters, amino-acid transporters and Na/Pi co-transporter is induced early in infection of the plant, suggesting these proteins act during the biotrophic phase of infection. Several other genes encoding proteins with a predicted extracellular localization are induced during infection and contain overrepresented domains. For example, three *P. infestans* genes encoding predicted N-terminal signal peptide as well as FAD-linked oxidase and Berberine/berberine-like (BBE) domains are induced during infection. The BBE domain is involved in the biosynthesis of the alkaloid berberine (Facchini et al., 1996). Moy and colleagues showed that a soybean homolog of this gene is inducing after infection with *P. sojae* (Moy et al., 2004). Molecular studies of proteins containing BBE domains in plants have indicated that several proteins containing these domains are in fact not secreted but instead targeted to specific alkaloid biosynthetic vesicle where the proteins accumulate (Amann et al., 1986; Kutchan and Dittrich, 1995; Facchini et al., 1996). The expansion of domain families with potential direct or indirect roles in host-pathogen interaction in filamentous plant pathogens strongly suggests adaptation to their lifestyle at the genomic level.
In addition to known domains, the set of overrepresented domains also revealed domains that, as yet, have not been implicated in pathogenicity nor are functionally characterized. An example is DUF953 domain that, within plant pathogens, is mainly found in oomycetes. This domain is observed in eukaryotic proteins with a thioredoxin like function and *P. infestans* genes encoding these domains are differentially expressed during infection. The significant expansion of these domains in plant pathogens and the fact that other well-described domains with a function in plant pathogenicity are also overrepresented, make proteins encoding poorly described but expanded domains interesting candidates to decipher their role in filamentous plant pathogens in general, or oomycetes in particular.

We determined domain overrepresentation on the base of species groups (plant pathogens and oomycetes) rather than on the level of individual species. We are aware that as a consequence of this approach we might have identified domains as being overrepresented in one group even if they do not need to be present or expanded in all the members (Tab. S3 & Tab. S5). Hence we might falsely extrapolate the functional role of a domain in a subset of species to the whole group, e.g. a domain that is exclusively found in plant pathogenic fungi and not in oomycete would still be overrepresented in the plant pathogenic group. Especially when comparing oomycete to fungal plant pathogens the dominant expansion of domains families within *Phytophthora* spp. over families in *H. arabidopsidis* might bias the inferred overrepresented domain (Tab. S5). Since we in general want to identify candidate domains that might be directly or indirectly involved in host-pathogen interaction, either at the level of filamentous plant pathogens or oomycetes, we think our group based approach is appropriate to establish a set of candidate proteins and domains.

Moreover, the clustering of presence, absence and expansion pattern of domains known or implicated to be involved in a plant pathogenic lifestyle with domains that have no known or direct connection to host-pathogen interactions, aids in expanding this set of novel candidate domains (Fig. 5). E.g. DUF1949 is within our species selection exclusively found in *Phytophthora* spp. and adopts a ferredoxin-like fold. The N-terminal region of proteins containing this domain shows similarity to another domain (UPF00029) that has been found in the human Impact protein. The *P. infestans* gene containing both domains is induced early during infection of the plant providing additional, independent evidence for the possible role of genes encoding this uncharacterized domain in host-pathogen interaction. However, domains that are also abundant in non-pathogenic species, e.g. other stramenopiles, might not be related or only indirectly involved to pathogenicity. Hence, the exact nature of the contribution of these domains to pathogenesis or to general life style requires more in depth experimental studies of the candidate domains and genes predicted to contain these functional entities.
Protein domains generally do not act as single entities but in synergy with other domains in the same protein or with other domains in interacting proteins. We identified 773 oomycete-specific bigrams of which 53 are observed in all analyzed oomycetes (Fig. 7a & Tab. S10). Based on our species selection we cannot conclude that the oomycete-specific bigrams are common to all oomycetes since they might only be specific for plant pathogenic oomycetes or even for the selected oomycetes analyzed in this study. The majority of the 773 bigrams, however, is specific for a subset of the tested oomycete species or even a single species. The 320 bigrams types that are observed in more than a single species or twice in the same proteome are observed in 982 predicted proteins. These bigrams are less likely to be the result of a wrong gene annotation and include already well described examples of oomycete-specific domain combinations, e.g. the FYVE-PIK bigram observed in *Phytophthora* phosphatidylinositol kinases (Meijer and Govers, 2006), the AP2-HDAC bigram that is specifically found in *P. ramorum* and *P. infestans* (Iyer et al., 2008), and the myosin head domain – FYVE bigram as well as the FYVE-GAF bigram found in myosin proteins in all analyzed oomycetes (Richards and Cavalier-Smith, 2005). Still, some of the bigrams could be artificial due to false negatives or false positives in the domain predictions. The remaining, species-specific bigrams could be the result of artificial fusion of genes due to wrong gene annotation, or an actual biological signal in one of the analyzed oomycete species. The derived results are not only dependent on the quality of the genome sequences of the analyzed oomycetes, but also on that of the other eukaryotes. Wrong predictions of bigrams in these species would lead to false negatives in oomycetes. Hence, the number of derived oomycete-specific bigrams is only an approximation and the true set of oomycete-specific bigrams needs to be further analyzed. Recent analysis of the underlying molecular mechanisms of domain gain in animals have shown that in fact gene fusion, tightly linked with gene duplication, is the major mechanism that shaped novel protein architecture (Buljan et al., 2010; Marsh and Teichmann, 2010). The contributions of this mechanism in forming lineage- or even species-specific bigrams in oomycetes and the probable role of the flexible genomes have to be further analyzed. The bigrams presented here form a comprehensive starting point for in depth bioinformatic and experimental analysis of promising gene families coding novel domain combinations.

Common domain types form the majority of the observed oomycete-specific bigrams, emphasizing the importance of novel combinations rather than novel domain types as a source for species-specific functionality. Only a minority of proteins containing oomycete-specific bigrams is secreted and none of these proteins is predicted to contain a RXLR and Crn motif. We are aware that the total number of predicted proteins containing RXLR or Crn motifs is lower than reported in other studies and that they were predicted using multiple
complementary methods (Haas et al., 2009). However, when directly comparing the number of proteins predicted to contain the RXLR motif by HMMER alone, the reported numbers are similar to our predictions. Together with the observation that RXLR proteins do not contain known Pfam-A domains in the C-terminal domain (Haas et al., 2009) our data is not in conflict with RXLR protein predictions from previous studies. Of the known Crn genes in *P. infestans*, 40% do not encode a secretion signal (Haas et al., 2009) and hence these sequences are not considered in the prediction of Crn motifs in our analysis and explain the discrepancy between the previously reported numbers and our predictions. Haas BJ and colleagues have reported a huge number of different C-terminal structures in *P. infestans* Crns that contained up to 36 different domains of which 33 are not described in Pfam. Several of these domains induce necrosis in plants. Since we focused in our analysis exclusively on Pfam domains we did not expect to find these proteins containing specific bigrams.

The majority of proteins containing oomycete-specific bigrams seem to be functional in the pathogen cytoplasm. Moreover, domains involved in mediation between macromolecules or lipids, e.g. the FYYE or the phox-like domain, as well as signaling domains, e.g. serine/threonine kinases-like or the DEP domain, are highly abundant in oomycete-specific bigrams. The serine/threonine kinase domain-like is overrepresented in oomycetes compared to fungal plant pathogens and particularly expanded within the *Phytophthora* spp. (Tab. S5). This expanded repertoire together with the high abundance of this domain in oomycete-specific bigrams strongly suggests that oomycetes have the capacity to recombine existing signaling pathways in a novel and complicated network that is distinct from other eukaryotes. This might also be true for other interaction networks since several domains mediating interactions between macromolecules (e.g. DNA binding zinc finger (IPR007087) or protein-protein interaction like WW/Rsp5/WWP (IPR001202)) are also highly abundant in oomycete-specific bigrams. Whether this reflects a general phenomenon in all oomycetes, specific for the plant pathogenic species analyzed in this study or only for *Phytophthora* spp. can only be answered when more oomycetes, including saprophytes and pathogens with different hosts have been sequenced.

We outlined a complex but comprehensive picture of the domain repertoire of filamentous plant pathogens focusing on oomycetes and showed how differences compared to other eukaryotes are reflecting the biology of these groups of organisms. Especially the expansion of certain domain families is directly linked with the lifestyle of oomycete plant pathogens and allowed the generation of a set of candidate domains likely to play important roles in the interaction with the plant host. Proteins containing overrepresented domains are enriched in the predicted secretome of the analyzed species. Moreover, the gene expression analysis of
genes encoding overrepresented domains during infection of the plant revealed a significant enrichment of genes encoding overrepresented domains within the differentially expressed genes. Furthermore, we observed a significant higher than expected abundance of genes encoding a signal peptide within the set of differentially expressed genes containing expanded domains. This added additional, independent evidence for the biological significance of our observations. Furthermore, oomycete genomes encode a set of proteins containing oomycete-specific domain combinations that are formed by common domain types and include several domains involved in signaling and/or mediation of interactions between macromolecules. Oomycetes might therefore possess altered regulatory and signaling networks that differ from other eukaryotes. If the described and discussed differences in the domain repertoire of oomycetes have a direct influence in plant pathogenicity or are of generally useful in these organisms has to be analyzed further. Nevertheless, they provide promising starting points that will aid our understanding of the biology of oomycetes in general and plant pathogens in particular.
METHODS

Species used in the analysis

In the performed analysis, 67 eukaryotic species representing four of the five eukaryotic supergroups (excluding Rhizaria) were considered (Fig. 1a and Tab. S1 for species abbreviations used in the manuscript). We used the predicted best model proteomes for all subsequent analysis.

Identification of domain composition

We predicted the domain repertoire of all proteins encoded in the diverse genomes using hmmpfam (HMMER package version 2.3.2) and a local Pfam-A database (v23). We applied a domain-model specific gathering cutoff and used HMM-models that are optimized to search for full-length entities (ls) in the query sequence.

In order to obtain the non-overlapping domain architecture of multi-domain proteins, we resolved overlapping domains according to certain rules. We defined two domains as overlapping if more than 10% of the predicted domain locations were overlapping (based on the relative length of the domains). If, in the case of overlapping domains, the e-value difference was larger than 5 (on a \(\log_{10}\) scale) we kept the domain with the highest e-value. In cases where the difference was smaller, we kept the longest model. If both overlapping models had the same length we considered differences in e-value and bit-score. In the case of the Pfam-based predictions for 15 proteins, the applied rules did not resolve overlapping entities. Therefore, we considered the conserved domain database (CDD, version 2.16) superfamily annotation, which automatically clusters domain entities that resemble evolutionary related domains. If both domains corresponded to the same family, we choose one entity.

Based on the non-overlapping domain architecture we derived different metrics for each proteome. We counted the abundance for every domain and the resulting number of different domain types per analyzed proteome. We defined domain bigrams as two consecutively located domains in a single protein. We discriminated between reciprocal domain pairs, so that the bigram (A|B) is not identical to (B|A), and took repeating domains into account, e.g. (A|A). Based on the set of bigrams we also determined the versatility of all individual domains in a given proteome, which is defined by the number of different direct N- and C-terminal partner, also including reciprocal and self-repeated pairs.
Prediction of secreted proteins

Secreted proteins were predicted using SignalP (version 3.0) (Bendtsen et al., 2004) in combination with TMHMM (version 2.0) (Krogh et al., 2001). We restricted the analysis to the first 70 amino acids of the protein and accepted signal peptide predictions if both the neural network and the HMM implemented in SignalP predicted the presence of a signal peptide under default parameters. Moreover, we declined predicted signal peptides if TMHMM predicted more than one transmembrane region in the protein. If only a single transmembrane helix was predicted and the predicted region was overlapping with the SignalP prediction for more than 10 amino acids and positioned within the first 35 amino acids from the start we included the protein in the set of secreted proteins.

Domain overrepresentation

Domain overrepresentation was calculated using a one-sided Fisher-Exact test. The derived p-values were Bonferroni-corrected for multiple testing by multiplying the p-value with the number of conducted tests. The corrected p-values were compared to an alpha = 0.001 to infer domain overrepresentation. For the overrepresented domains in oomycete plant pathogens compared to fungal plant pathogens, we considered domains that occur at least once in a single plant pathogen but nevertheless could also occur in other eukaryotic species.

Gene expression analysis of *P. infestans*

We extracted NimbleGen expression data of *P. infestans* during infection of potato 2-5 days post inoculation from GEO (http://www.ncbi.nlm.nih.gov/geo/). The setup and initial analysis of the NimbleGen data is describe by Haas BJ and colleagues (Haas et al., 2009). The log2 transformed and mean centered array intensities were analyzed for differentially expression using Multiexperiment Viewer (MeV) (Saeed et al., 2006). T-tests were conducted between two groups (group A: different media types, group B: replicates for one of the days post infection). The test was applied for each day after inoculation and significant up-/down-regulated genes were reported applying a p-value cutoff of 0.05. False discovery rates were addressed using R and the ‘qvalue’ package by computing q-values for each of the comparisons and subsequently applying a q-value cutoff of 0.05 (Storey and Tibshirani, 2003; R Development Core Team, 2010). Visualization of the heatmaps was done using R and the Bioconductor package utilizing Spearman correlation as a distance measurement and hierarchical clustering (average linkage) (Gentleman et al., 2004). Gene expression intensities relative to the average expression intensities in media types (V8, RS, Pea) were computed in R.
Clustering of domain profiles

We created abundance profiles for each domain based on the abundance in each individual proteome. We excluded domains that were only identified in a single species. The rows (domains) were multiplied by a scaling factor so that the sum of squares is one and subsequently the columns (species) were normalized in the same way. We performed a hierarchical clustering (average linkage) of the profiles using Spearman correlation matrix as a distance measurement. The normalization and clustering was performed using Cluster (Eisen et al., 1998) and the visualization was done using TreeView (M. Eisen, http://rana.lbl.gov/EisenSoftware.htm).

Domain promiscuity

We calculated the domain promiscuity for every domain in the analyzed species based on weighted bigram frequency (Basu et al., 2008). We took a relative moderate cutoff for determining promiscuous domains; every domain with a higher promiscuity score than a domain that is only present once in the genome and is participating in one bigram type is called promiscuous.

Prediction of the RXLR and Crn motif in oomycetes

We identified the presence of the RXLR motif in all predicted proteins in the analyzed oomycetes using three different HMMER models (Jiang, RHY, personal communication). The first model was created using *P. ramorum* and *P. sojae* RXLRs and included the RXLR motif itself and 10 amino acids down- and upstream of the motif. The two other models were based separately on RXLRs from *P. infestans* and *H. arabidopsidis* and include 10 amino acids upstream from the RXLR motif and five amino acids downstream of the DEER motif. We used HMMER (hmmsearch) with an e-value cutoff of 10 and subsequently combined all predictions. Furthermore, we demanded the presence of a predicted signal peptide (SignalP, s.a.) cleavage site within the first 30 amino acids of the protein, the gap between cleavage site and the start of the motif to be $\leq 30$, the start of the motif to be within the first 100 amino acids of the protein and the starting position of the RXLR motif to be downstream of the cleavage site. For the identification of the Crn LFLAK motif we used a HMMER model of that region (Haas, BJ, personal communication) and the same sequence demands as for the RXLRs. Secreted stramenopiles-specific families were defined by the presence of at least a single sequence with a predicted secretion signal within the cluster.

Phylogenetic analysis of the GAF domain
We derived all sequences containing a GAF domain from the selected proteomes and extracted the amino acid sequence of the domain based on the start and end points of the domain model. We conducted a similarity search with the extracted domains using blastp (version 2.2.20) with an e-value cutoff of $1 \times 10^{-5}$ and a low complexity filter against a set of 295 bacterial predicted proteomes (downloaded from the NCBI ftp server (27/01/2009)). In the homologs that were obtained domains were predicted using hmmpfam as described above. Subsequently, prokaryotic GAF domains were extracted and aligned together with the eukaryotic domains using maft (v.6.713b) with the local alignment strategy (Katoh et al., 2002). A phylogenetic tree was constructed with RAxML (v7.0.4) using the GAMMA model of rate heterogeneity and the WAG amino acid substitution matrix (Stamatakis, 2006).

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Figure Legends

Figure 1 – (a) The major eukaryotic groups considered in the analysis and the number of species represented in every group. For the exact species used in the analysis see Supplementary Tab. 1. The tree is adapted from Simpson and Rogers (Simpson and Roger, 2004) and incorporates the phylogeny for the stramenopiles based on Blair and colleagues (Blair et al., 2008) (b) Fungal and oomycete plant-pathogenic species used within this analysis. The plant pathogens include species with different lifestyles, indicated by the symbol following the species name. The phylogeny for the fungi is based on James and colleagues (James et al., 2006).

Figure 2 – Description of different metrics used in this study. In the example shown we observe five different domain types. The abundance of a domain type is defined as the number of occurrences of the individual entity within the species, e.g. domain type B has an abundance of two. The versatility is defined as the number of different direct adjacent N- or C-terminal neighbors. We distinguish between N- and C-terminal partners, e.g. the versatility of domain type C is three. A bigram is a set of two directly adjacent domains and we also consider two entities of the same domain a bigram, e.g. we observe six different bigram types in the proteome of which two have an abundance of two (see right panel).

Figure 3 - Dependence of the number of domain and bigram types observed in the analyzed species. The average number of different bigrams of species that have between 2,000 and 2,500 different domain types is indicated with the lower horizontal red bar. The upper horizontal red bar indicates the average number of different bigrams for Phytophthora spp. The full species names corresponding to the abbreviations can be found in Supplementary Table 1. A magnification of the area encompassing the oomycete and fungal plant pathogens is shown; the species of interest are highlighted. The dots are colored according to the major eukaryotic groups as indicated in the text box.

Figure 4 – Overrepresentation of selected, well-described domains involved in plant-pathogen interaction and establishing or maintaining infection. (a) The log_{2}-fold overrepresentation of the domains in plant pathogens is shown in the bar chart. The absolute number of occurrences in plant pathogens is displayed in the bar, the percentage of all predicted domains in plant pathogens and the corrected p-value at the tip of the bars. The fold overrepresentation and the p-value for the Kazal protease inhibitor domain were based on the overrepresentation in oomycetes compared to plant pathogens (indicated by white bar and *). (b) The overrepresented domains described in (a) are depicted in their possible cellular role during infection of the plant host.
**Figure 5** – Gene expression analysis of *P. infestans* genes encoding overrepresented domains and a predicted N-terminal signal peptide. Genes with significant gene expression changes at different time points after infection (2-5 dpi) relative to the expression intensities of different growth media are displayed (T-test, p-value <0.05, q-value <0.05). Heatmaps of the significantly differentially expressed genes at different time points relative to growth media. Genes were clustered using Spearman rank correlation and average linkage clustering. Gene identifiers as well as domain description are displayed. Gene expression profiles are displayed for the expression intensities relative to the average intensities of the growth media for each time point after infection. Panels a-d display the heatmaps and expression profiles of the significantly differentially expressed genes relative to the growth media at the individual time points (a: 2 dpi, b: 3 dpi, c: 4 dpi, d: 5dpi).

**Figure 6** – Average linkage clustering of normalized domain profiles using Spearman rank correlation as distance measurement. The species tree for all eukaryotic species is depicted on top with the color code of their supergroup as introduced in Fig. 1. Plant pathogens are marked with a star and the arrowheads highlight domains identified as overrepresented in plant pathogens.

**Figure 7** – (a) A Venn diagram depicting the presence of oomycete-specific bigram types in the analyzed oomycete proteomes and indicating the number of shared bigram types between different proteomes. The total number of oomycete-specific bigram types in each proteome is shown in brackets. The Venn diagram was produced using Venny (Oliveros, 2007). (b) Domain architecture of example proteins containing a GAF domain. The two upper architectures resemble common protein architectures, the cGMP dependent 3',5'-cyclic phosphodiesterase (observed 111 times in eukaryotes and 5 times in oomycetes) and phytochrome A (observed 21 times in eukaryotes). The lower two architectures depict oomycete-specific architectures. The FYVE-GAF fusion is observed 53 times independent of other domains and the myosin motor head in combination with the FYVE-GAF fusion is observed 4 times, a single copy in each of the oomycetes included in this study. (c) Simplified evolutionary tree based on the phylogenetic analysis of the GAF domain in prokaryotes and eukaryotes. GAF domains from proteins with a FYVE-GAF fusion are exclusively found to be close to bacterial GAF domains. Other oomycete proteins that only contain the GAF without the FYVE domain also cluster with other eukaryotic sequences.
Supplementary Data

Supplementary Figure 1 – (a) Dependence of the number of domain and bigram types excluding singletons observed in the species analyzed. (b) Relationship between the proteome size and the number of distinct domain types as well as (c) the number of bigram types. The species names corresponding to the abbreviations can be found in Supplementary Table 1.

Supplementary Figure 2 – The predicted secretome of the 67 analyzed eukaryotic species. The absolute size as well as the percentage of the predicted secretome is displayed. The analyzed plant pathogens are indicated with a star and species abbreviations are shown in Supplementary Table 1.

Supplementary Figure 3 - The average abundance and versatility of all observed domains in a log₂-log₂ plot. The graph displays a linear positive correlation between the abundance and versatility of different domains (regression line in red). Domains that are highly abundant and do not have a high number of different N- or C-terminal partners are shown in the lower right sector of the plot. Domains that show an uneven distribution of versatility in the examined species might have a low average versatility, even if they have many different partners in some species.

Supplementary Table 1 - Overview of species used in the analysis. The species name, the used species abbreviation, the version as well as the download source, the NCBI taxon ID, the size of the predicted proteome and the eukaryotic group of the species are depicted. The citable reference, if available, is shown, too.

Supplementary Table 2 - Domain abundance of all predicted Pfam domains. The InterproID, the PfamID as well as the description for each domain is indicated. The rank (based on the overall abundance), abundance over all analyzed species and the average number per species is shown. The numbers for all species (a), plant pathogens (b) and oomycetes (c) are shown in different sheet including rank relative to all species.

Supplementary Table 3 - Table of overrepresented domains in plant pathogens. Overrepresentation was inferred using Fisher Exact Test (p-value corrected for multiple testing). The InterproID, PfamID and description for each domain are given, as well as the abundance of this domain in all species, in plant pathogens and in oomycetes, respectively. The fraction of the domain in all species and in plant pathogens is calculated based on the abundance of all domains in the respective group. The contribution of fungal plant
pathogens to the domain abundance in plant pathogens is shown in blue (>80%) and of oomycetes in orange (> 80%) or in red (100%). The number of proteins in plant pathogens, the contribution of the different plant pathogenic oomycete species, the number of proteins in plant pathogens containing a predicted secretion signal (Methods) and the database IDs for the proteins in plant pathogens are listed.

**Supplementary Table 4** – Differentially expressed *P. infestans* genes at different time points post inoculation (days post inoculation dpi). (a) Differentially expressed genes are presented as well as (b) differentially expressed genes encoding significantly expanded domains. The gene identifier, the p-value and the q-value for the different time points are reported (Methods). The presence of a predicted signal peptide as well as the predicted domains is noted.

**Supplementary Table 5** - Table of overrepresented domains in oomycete plant pathogens. Overrepresentation was inferred using Fisher Exact Test (p-value corrected for multiple testing). The InterproID, PfamID and description for each domain is given as well as the abundance of this domain in plant pathogens and in oomycetes, respectively. The fraction of the domain in plant pathogens and oomycetes is calculated based on the abundance of all domains in the respective group. The contribution of oomycete plant pathogens to the domain abundance in plant pathogens is shown in orange (> 80%) or in red (100%). The number of proteins in oomycetes, the contribution of each oomycete species to the overall number, the number of proteins containing a predicted secretion signal (Methods) and the database IDs for the proteins are listed.

**Supplementary Table 6** - Domain versatility is calculated for each individual domain. Versatility is defined as the number of different adjacent N- or C-terminal domains. The rank, the versatility and the average versatility within the group are reported for (a) all species, (b) plant pathogens (c) oomycete plant pathogens. For (b) and (c), the relative rank of the domain compared to the rank in (a) is noted.

**Supplementary Table 7** - Averaged promiscuity is calculated for each individual domain. The rank of the domain and the average promiscuity is reported for (a) all species, (b) plant pathogen and (c) oomycete plant pathogens. In (b) and (c), the relative rank of the domain in relation to (a) is noted.

**Supplementary Table 8** - The abundance of bigrams (directly adjacent domains A and B, including self-repeated domains) is reported for (a) all species, (b) plant pathogens and (c) plant pathogenic oomycetes. The rank, the summed abundance per bigram and the average
number per group are reported. For (b) and (c), the relative rank of the bigram to (a) is noted.

**Supplementary Table 9** - The abundance of bigrams (directly adjacent domains A and B, excluding self-repeated domains) is reported for (a) all species, (b) plant pathogens and (c) plant pathogenic oomycetes. The rank, the summed abundance per bigram and the average number per group are reported. For (b) and (c), the relative rank of the bigram to (a) is noted.

**Supplementary Table 10** - Oomycete-specific bigrams (i.e. two adjacent domains A and B), the abundance of the bigrams in oomycete plant pathogens, the number of proteins containing the bigram type in oomycetes and the species distribution is shown. Database IDs for proteins containing the predicted bigrams for plant-pathogenic oomycete are reported.

**Supplementary Files 1** – Clustering of the domain abundance profiles. The clustering files can be viewed with Treeview (M. Eisen, http://rana.lbl.gov/EisenSoftware.htm).
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Figure 1 – (a) The major eukaryotic groups considered in the analysis and the number of species represented in every group. For the exact species used in the analysis see Supplementary Table. The tree is adapted from Simpson and Rogers (Simpson and Roger, 2004) and incorporates the phylogeny for the stramenopiles based on James and colleagues (Blair et al., 2008) (b) Fungal and oomycete plant-pathogenic species used within this analysis. The plant pathogens include species with different lifestyles, indicated by the symbol following the species name. The phylogeny for the fungi is based on James and colleagues (James et al., 2006).
Proteome species X

Figure 2 – Description of different metrics used in this study. In the example shown we observe five different domain types. The abundance of a domain type is defined as the number of occurrences of the individual entity within the species, e.g. domain type B has an abundance of two. The versatility is defined as the number of different direct adjacent N- or C-terminal neighbors. We distinguish between N- and C-terminal partners, e.g. the versatility of domain type C is three. A bigram is a set of two directly adjacent domains and we also consider two entities of the same domain a bigram, e.g. we observe six different bigram types in the proteome of which two have an abundance of two (see right panel).
Figure 3 - Dependence of the number of domain and bigram types observed in the analyzed species. The average number of different bigrams of species that have between 2,000 and 2,500 different domain types is indicated with the lower horizontal red bar. The upper horizontal red bar indicates the average number of different bigrams of Phytophthora spp. The full species names corresponding to the abbreviations can be found in Supplementary Table 1. A magnification of the area encompassing the oomycete and fungal plant pathogens is shown; the species of interest are highlighted. The dots are colored according to the major eukaryotic groups as indicated in the text box.
Figure 4 – Overrepresentation of selected, well-described domains involved in plant-pathogen interaction and establishing or maintaining infection. (a) The log2-fold overrepresentation of the domains in plant pathogens is shown in the bar chart. The absolute number of occurrences in plant pathogens is displayed in the bar, the percentage of all predicted domains in plant pathogens and the corrected p-value at the tip of the bars. The fold overrepresentation and the p-value for the Kazal protease inhibitor domain were based on the overrepresentation in oomycetes compared to plant pathogens (indicated by white bar and *). (b) The overrepresented domains described in (a) are depicted in their possible cellular role during infection of the plant host.
Figure 5 – Gene expression analysis of P. infestans genes encoding overrepresented domains and a predicted N-terminal signal peptide. Genes with significant gene expression changes at different time points after infection (2-5 dpi) relative to the expression intensities of different growth media are displayed (T-test, p-value < 0.05, q-value < 0.05). Heatmaps of the significantly differentially expressed genes at different time points relative to growth media. Genes were clustered using Spearman rank correlation and average linkage clustering. Gene identifiers as well as domain description are displayed. Gene expression profiles are displayed for the expression intensities relative to the average intensities of the growth media for each time point after infection. Panels a-d display the heatmaps and expression profiles of the significantly differentially expressed genes relative to the growth media at the individual time points (a: 2 dpi, b: 3 dpi, c: 4 dpi, d: 5 dpi).
Figure 6 – Average linkage clustering of normalized domain profiles for a random sample of eukaryotic species as introduced in Fig. 1. Plant pathogens are marked with a star and the arrowheads highlight domains identified as overrepresented in plant pathogens.
Figure 7 – (a) A Venn diagram depicting the presence of oomycete-specific bigram types in the analyzed oomycete proteomes and indicating the number of shared bigram types between different proteomes. The total number of oomycete-specific bigram types in each proteome is shown in brackets. The Venn diagram was produced using Venny (Oliveros, 2007). (b) Domain architecture of example proteins containing a GAF domain. The two upper architectures resemble common protein architectures, the cGMP dependent 3',5'-cyclic phosphodiesterase (observed 111 times in eukaryotes and 5 times in oomycetes) and phytochrome A (observed 21 times in eukaryotes). The lower two architectures depict oomycete-specific architectures. The FYVE-GAF fusion is observed 53 times independent of other domains and the myosin motor head in combination with the FYVE-GAF fusion is observed 4 times, a single copy in each of the oomycetes included in this study. (c) Simplified evolutionary tree based on the phylogenetic analysis of the GAF domain in prokaryotes and eukaryotes. GAF domains within oomycetes and the FYVE-GAF fusion are exclusively found to be close to bacterial GAF domains. Other oomycete proteomes that only contain the GAF without the FYVE domain also cluster with other eukaryotic sequences.