Genomic signatures accompanying the dietary shift to phytophagy in polyphagous beetles

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Abstract

The diversity and evolutionary success of beetles (Coleoptera) are proposed to have arisen from

millions of years of specialized trophic interactions with land plants. In particular, ingestion of toxic plant

allelochemicals may impose selective pressures that drive genomic diversification and speciation in

phytophagous beetles. However, evidence of changes in beetle gene repertoires driven by these

interactions remains largely anecdotal and without explicit hypothesis testing. To address this, we

explored the genomic consequences of beetle-plant trophic interactions by performing comparative

gene family analyses across 18 species representing the two most speciose beetle suborders. By

contrasting gene content of species from the phytophagous-rich suborder Polyphaga with those of the

mainly predatory Adephaga, we identified families of detoxification enzymes that underwent adaptive

expansions in Polyphaga. These genomic signatures that accompany the dietary shift to phytophagy in

polyphagous beetles suggest a key role for interactions with plant chemical defenses in driving beetle

diversification.

Introduction

Species richness among eukaryotes varies substantially, with some clades having only a few representatives and others comprising hundreds of thousands of extant species. In particular, the class Insecta outnumbers all other classes with more than half of all described extant species (Farrell, 1998; Grimaldi and Engel, 2005). Beetles (Coleoptera) encompass approximately 380,000 described species, representing ca. 40% of described insect diversity (Slipinski et al., 2011). Several hypotheses have been proposed to explain this richness, notably their complex interactions with flowering plants (Farrell, 1998; Leschen and Buckley, 2007; McKenna et al., 2009, 2015; Zhang et al., 2018) and a high lineage survival rate (Hunt et al., 2007). Nevertheless, detailed supporting evidence from molecular genetic studies remains sparse, making it difficult to assess the relative importance of these and other potentially important contributing factors (Barraclough et al., 1998; Suchan and Alvarez, 2015).

The remarkable evolutionary success of beetles may have been driven by the interplay between their trophic niche and their genomic content and architecture. This is based on the premise that environmental and ecological conditions are likely to be predominant factors influencing the fate of genetic variation in populations under natural selection (Barrick and Lenski, 2013), eventually driving divergence into distinct species (Seehausen et al., 2014). Among all components of the biotic environment, the trophic niche (principal source of nourishment) of an organism plays a crucial role in shaping the evolution of phenotypic innovations and their underlying genomic changes, e.g. feeding modes in cichlid fishes (Parsons et al., 2016), mouth development in *Pristionchus* nematodes (Ragsdale et al., 2013), and bitter taste receptors in vertebrates (Li and Zhang, 2014). Among several hypotheses explaining the tremendous diversity among beetles, a shift from an ancestral diet as saprophages (detritus-feeding) or mycophages (fungi-feeding) (Betz et al., 2003) to phytophagy (feeding on living plant material in a broad sense) is often evoked (Farrell, 1998; Leschen and Buckley, 2007; McKenna et al., 2009). While the suborder Adephaga (~45,000 species) comprises mostly predatory species, including ground beetles and diving beetles, the largest beetle suborder, Polyphaga (~315,000), is predominantly comprised of phytophagous clades, among which the most species-rich families are weevils (Curculionidae, ~51,000), longhorn beetles (Cerambycidae, ~30,000), and leaf beetles (Chrysomelidae, ~32,000) (Slipinski et al., 2011). Phytophagy appeared approximately 425 million

years ago, quickly after terrestrial life was established (Labandeira, 2002). It progressively diversified

to target most plant tissues (Labandeira, 2013), shortly before the radiation of flowering plants 120-100

million years ago (Grimaldi, 1999). In response, plants have evolved diverse strategies to protect

themselves, which in turn impose selective pressures on the animals that feed on them.

While many biological processes are likely to play a role in this evolutionary battle, a key weapon in the

arsenal of phytophagous insects' adaptations is their ability to neutralize or minimize the effects of plant

secondary compounds. Protein families known to be crucial for eliminating harmful plant toxins are

cytochrome P450 monooxygenases (P450s), carboxylesterases (CEs), UDP-glycosyltransferases

(UGTs), and glutathione S-transferases (GSTs) (Voelckel and Jander, 2014). While P450s and CEs

modify residues to make compounds more hydrophilic, UGTs and GSTs conjugate xenobiotic

compounds to hydrophilic molecules. Detoxification is completed by membrane transporters, such as

ATP-binding cassette (ABCs) transporters, which move xenobiotic compounds to where they can either

be excreted, or less frequently sequestered in order to be reused as a defense mechanism (Voelckel

and Jander, 2014). Additionally, to prevent phytophagous insects from digesting their tissues, plants

produce enzyme inhibitors that block catalytic sites or compete with the substrates of enzymes involved

in digestion. The major families affected are endopeptidases, such as cysteine, and serine proteases,

as well as more specific enzymes such as glycoside hydrolases (GHs), certain types of which are able

to break down polysaccharide molecules, including cellulose, hemicellulose and pectin in plant cell walls

(McKenna et al., 2016; Pauchet et al., 2010). Other adaptations to phytophagy include repertoires of

chemoreceptors that are crucial for finding appropriate food sources (Goldman-Huertas et al., 2015),

and the specialization of mouthparts in response to plant mechanical barriers, which are highly

diversified in insects (Labandeira, 1997).

As lineages diverge, their genomes accumulate changes, some of which are expected to be directly

linked to functional adaptations. Identifying such genomic features and linking them to phenotypic

differences, whilst robustly distinguishing between the effects of stochastic changes and natural

selection (Hurst, 2009), is critical to deciphering the genomic drivers of species radiations (Shaw and

Lesnick, 2009). Changes include point substitutions, which may affect existing functional elements, but

also larger-scale changes such as duplications, from individual genes to entire genomes, which by

adding new members to the repertoires of key gene families may constitute an ideal mechanism to

facilitate the emergence of novel functions leading to successful phytophagy (Kondrashov, 2012).

Whereas newly generated gene copies are usually redundant or deleterious and pseudogenized,

rendering the gene copy non-functional (Innan and Kondrashov, 2010), they are sometimes maintained.

Particularly interesting cases of gene family expansions are the ones restricted to specific lineages,

resulting in lineage specific expansion (LSE). Evolutionary mechanisms causing LSE are numerous

and not all adaptive (see Innan and Kondrashov, 2010 for a comprehensive review). However,

duplicated gene copies may provide an immediate selective advantage and be maintained by selection.

This can be due to an increased dosage of the gene, or to changes following the duplication being

selected in one gene copy but not the other, which might allow evolution towards a different function in

so-called neo-functionalization processes. Enzymes are considered particularly relevant candidates for

such evolutionary processes as they could expand their range of substrates (Francino, 2005).

Here we apply a comparative genomics approach to examine the evolution of genes putatively involved

in plant-insect interactions by sampling from the two largest beetle suborders, which, generally-

speaking, present contrasting trophic niches. We contrast exemplars from the characteristically

predaceous Adephaga with exemplars from Polyphaga and we hypothesize that plant-insect

interactions during the dietary shift to phytophagy should be accompanied by genomic evolutionary

signatures visible at the subordinal scale. Using genomic and transcriptomic data from 18 beetle

species, we estimate ancestral gene family content, taking into account gene gains and losses across

the species phylogeny, to identify significant LSEs of gene families related to phytophagy and

signatures of adaptive expansions in these families. Ignoring sensory receptors, as their evolution might

be driven by agents other than those related strictly to trophic niche (Brito et al., 2016), and

morphological genes, as their inferred association with diet is less robust, we focus on genes coding

for enzymes, for which adaptive LSE specific to Polyphaga would suggest a role for detoxification and

digestive pathways in driving adaptation and speciation.

Results

A representative sampling of the two major coleopteran suborders

Reliable estimation of gene gain and loss events requires a robust evolutionary framework, i.e. a phylogeny that includes the species studied, as well as the characterization of gene families across complete gene sets from these same species. To study adaptation to phytophagy, we sampled from both Adephaga (mostly predaceous) and Polyphaga (with diverse trophic habits, including a very large number of phytophagous species). A balanced sampling of each suborder was achieved comprising twelve transcriptomes and six genomes, with Benchmarking Universal Single-Copy Ortholog (BUSCO) completeness estimates (Simão et al., 2015; Waterhouse et al., 2018) ranging from 71.9% to 97% (Figure 1, Table 1). The species phylogeny was estimated using 405 BUSCO genes found to be complete in all species and the strepsipteran outgroup, *Stylops melittae* (Figure 1). Protein-coding sequence predictions ranged from 9,844 to 24,671 genes per beetle species. These sequences matched 14,908 Arthropoda orthologous groups (OGs) containing at least one species of Coleoptera in the OrthoDB v8 catalogue (Kriventseva et al., 2015). This represented a minimum of 6,742 and a maximum of 11,149 OGs for *Carabus frigidus* and *Leptinotarsa decemlineata,* respectively. OGs containing genes from only one of the two sampled suborders were excluded, resulting in a total of 9,720 OGs for the analysis that have evolutionary histories traceable to the last common ancestor of beetles. Functional annotations of the sequences within these OGs were used to identify and assign several of them to enzyme families relevant to the tested hypothesis. These candidate OGs comprised four UGTs, 22 P450s, 19 CEs, six GSTs, four SERs, seven CYSs, 28 ABCs, and one GH, for a total of 91 candidate OGs from eight families of genes (i.e. functional categories) (Table 2).

Table 1. Beetle genomes and transcriptomes included in the study. Taxonomic classifications are listed with data sources, as well as completeness (Benchmarking Universal Single-Copy Ortholog, BUSCO, score, C=Complete, S=Complete Single-Copy, D=Complete Duplicated, F=Fragmented, M=Missing), number of predicted proteins, and number of orthologous groups (OGs) with genes from each species. The outgroup species used in the phylogeny, *Stylops melittae*, belongs to the order Strepsiptera, which is the sister group of Coleoptera (Niehuis et al., 2012).

Species	Short form	Suborder	Family	Type of assembly	Accession	Bioproject	Source	BUSCO score (1658 genes)	Predicted proteins	OrthoDB groups (OG)
Cicindela hybrida	CHYBR	Adephaga	Carabidae	Transcriptome	GDMH01000000	PRJNA286505	1KITE, this study	C:90.4[S:84.9%,D:5.5%],F:3.4%,M:6.2%	13,916	8,111
Calosoma frigidum	CFRIG	Adephaga	Carabidae	Transcriptome	GDLF01000000	PRJNA286499	1KITE, this study	C:78.3%[S:73.5%,D:4.8%],F:9.8%,M:11.9%	9,844	6,742
Elaphrus aureus	EAURE	Adephaga	Carabidae	Transcriptome	GDPI01000000	PRJNA286520	1KITE, this study	C:92.4%[S:85.9%,D:6.5%],F:2.2%,M:5.4%	12,808	8,020
Noterus clavicornis	NCLAV	Adephaga	Noteridae	Transcriptome	GDNA01000000	PRJNA286561	1KITE, this study	C:90.9%[S:84.1%,D:6.8%],F:4.4%,M:4.7%	12,981	7,918
Haliplus fluviatilis	HFLUV	Adephaga	Haliplidae	Transcriptome	GDMW01000000	PRJNA286525	1KITE, this study	C:91.8%[S:76.3%,D:15.5%],F:4.3%,M:3.9%	19,408	8,528
Cybister lateralimarginalis	CLATE	Adephaga	Dytiscidae	Transcriptome	GDLH01000000	PRJNA286512	1KITE, this study	C:88.2%[S:83.1%,D:5.1%],F:3.7%,M:8.1%	13,916	7,256
Sinaspidytes wrasei	SWRAS	Adephaga	Aspidytidae	Transcriptome	GDNH01000000	PRJNA286492	1KITE, this study	C:87.0%[S:76.7%,D:10.3%],F:4.0%,M:9.0%	13,392	7,721
Dineutes sp.	DINEU	Adephaga	Gyrinidae	Transcriptome	GDNB01000000	PRJNA286516	1KITE, this study	C:71.9%[S:51.6%,D:20.3%],F:13.6%,M:14.5%	14,644	7,089
Gyrinus marinus	GMARI	Adephaga	Gyrinidae	Transcriptome	GAUY01000000	PRJNA219564	1KITE, Misof et al., 2014	C:81.5%[S:79.0%,D:2.5%],F:8.1%,M:10.4%	13,867	7,663
Aleochara curtula	ACURT	Polyphaga	Staphylinidae	Transcriptome	GATW01000000	PRJNA219522	1KITE, Misof et al., 2014	C:91.4%[S:87.5%,D:3.9%],F:4.0%,M:4.6%	20,280	8,513
Anoplophora glabripennis	AGLAB	Polyphaga	Cerambycidae	Genome	GCF_000390285	PRJNA167479	I5k, McKenna et al., 2016	C:96.9%[S:95.8%,D:1.1%],F:2.7%,M:0.4%	22,035	10,959
Agrilus planipennis	APLAN	Polyphaga	Buprestidae	Genome	GCF_000699045	PRJNA230921	I5k, unpublished	C:92.5%[S:91.2%,D:1.3%],F:4.5%,M:3.0%	15,497	9,089
Dendroctonus ponderosae	DPOND	Polyphaga	Curculionidae	Genome	GCF_000355655	PRJNA162621	Keeling et al., 2013	C:91.2%[S:86.0%,D:5.2%],F:4.1%,M:4.7%	13,457	8,518
Leptinotarsa decemlineata	LDECE	Polyphaga	Chrysomelidae	Genome	GCF_000500325	PRJNA171749	I5k, Schoville et al., 2018	C:88.9%[S:87.5%,D:1.4%],F:9.9%,M:1.2%	24,671	11,149
Laparocerus tessellatus	LTESS	Polyphaga	Curculionidae	Transcriptome	10.5281/zenodo.1336288	N/A	This study	C:93.8%[S:91.9%,D:1.9%],F:1.4%,M:4.8%	18,448	8,616
Meloe violaceus	MVIOL	Polyphaga	Meloidae	Transcriptome	GATA01000000	PRNJA219578	1KITE, Misof et al., 2014	C:90.3%[S:85.6%,D:4.7%],F:5.9%,M:3.8%	14,295	8,480
Onthophagus taurus	OTAUR	Polyphaga	Scarabaeidae	Genome	GCF_000648695	PRNJA167478	I5k, unpublished	C:96.2%[S:93.9%,D:2.3%],F:2.5%,M:1.3%	17,483	9,315
Tribolium castaneum	TCAST	Polyphaga	Tenebrionidae	Genome	GCF_000002335	PRJNA12540	Tribolium Genome Sequencing Consortium et al., 2008	C:97.0%[S:96.5%,D:0.5%],F:1.6%,M:1.4%	16,645	9,429
Stylops melittae	SMELI	Outgroup	Stylopidae	Transcriptome	GAZM02000000	PRNJA219603	1KITE, Misof et al., 2014	C:76.5%[S:55.0%,D:21.5%],F:7.1%,M:16.4%	13,026	6,104

181 **Table 2.** Candidate gene categories with the keywords and identifiers used to select them from the full

182 sets of sequences annotated with InterProScan. To be included as candidate orthologous groups (OGs)

183 in the category, OGs were required to have at least one sequence matching both a UniRef and an

184 InterProScan entry, and an additional gene ontology term in the case of serine proteases.

Category	InterProScan (Pfam or InterPro identifiers) or gene ontology	UnifRef KeyWord	Number of OGs
UDP-glycosyltransferases (UGTs)	PF00201	name:"cluster UDP glucuronosyltransferase" OR name:"cluster UDP glycosyltransferase"	4
Cytochrome P450 oxidases (P450s)	PF00067	name:"cluster Cytochrome P450"	22
Carboxylesterases (CEs)	PF02230, PF00135	name:"cluster carboxylesterase" OR name:"carboxylic ester hydrolase"	19
Glutathione S-transferases (GSTs)	PF00043, PF02798	name:"cluster Glutathione S-transferase"	6
Serine proteases (SERs)	PF00450,PF12146, PF05577,GO:0008236	name:"cluster Serine protease" OR name:"cluster Serine peptidase"	4
Cysteine proteases (CYSs)	PF00112	name:"cluster cysteine protease" OR name:"cluster cystein protease" OR name:"cluster Papain"	7
ABC transporters (ABCs)	PPF00005, PF00664	name:"cluster ABC"	28
Glycoside hydrolases (GHs)	IPR000334, IPR000743, IPR001360, IPR001547	name:"cluster Glycoside hydrolase"	1
Total			91

Polyphaga exhibit more frequent gains across a larger set of OGs

Analysis of per-species gene counts of the complete set of 9,720 OGs was performed with the

Computational Analysis of gene Family Evolution (CAFE v3) (Han et al., 2013) tool. The mode

considering distinct gene gain (λ=0.0019 gain/gene/million years) and gene loss (μ=0.0018

loss/gene/million years) was preferred over a single value for λ and μ, having a significantly greater

maximum likelihood score (see Methods). The λ (gain) and μ (loss) values predicted when CAFE was

run on each suborder separately were λ=0.0020 and μ=0.0027 for Adephaga, versus λ=0.0023 and

μ=0.0021 for Polyphaga, showing a tendency for Adephaga to lose genes and for Polyphaga to gain

genes. Among the 9,720 OGs were 21 with reported expansions originating at the Adephaga root and

126 at the Polyphaga root (see Figure 1 to locate the nodes). Conversely, 240 OGs showed gene losses

for Adephaga and 354 for Polyphaga. Two expansions and 21 losses affected the candidate OGs for

Adephaga, and nine expansions and six losses for Polyphaga. Other polyphagan nodes leading to

polyphagous-rich clades (i.e., Chrysomeloidea and Curculionidae) also exhibited more candidate OGs

expanding than contracting (Figure 1). All counts of gene gains and losses per node are presented in

Supplementary Figures 1 and 2. Additionally, CAFE assigned individual OG p-values of < 0.01 to a

subset of 910 (9.3%) OGs, which, according to De Bie et al., 2006, indicates gene families likely to have

experienced accelerated rates of gain and loss. These are interesting to investigate further as they may

represent large OGs of potentially unequal size between the suborders. Among these were 26 of the

91 candidate OGs (28.6%), a significantly larger proportion (2-sample test for equality of proportions,

chi-square test, p-value < 0.0001) compared with just 9.3% of non-candidate OGs.

Figure 1. The ultrametric species phylogeny with gene family expansions and contractions quantified for nodes of interest and bar charts showing completeness of the genomic and transcriptomic datasets studied. The species tree was built from 405 single-copy orthologs and constrained to have Geadephaga (*C. frigidum*, *E. aureus*, *C. hybridia*) and Hydradephaga (the six other Adephaga) as monophyletic sister clades (e.g., following McKenna et al., 2015). Branch lengths are scaled in millions of years. Maximum likelihood bootstrap support was 99 or 100% for all branches. [G] symbol indicates data from species with sequenced genomes with the remaining species being from transcriptomes. The numbers of orthologous groups (OGs) with expansions (+) and contractions (-) are displayed at the root node of each suborder. Pie charts show proportions of OGs with gene losses (black) and gene gains (green) with respect to OGs with no significant losses or gains for all considered OGs (grey) and only the candidate OGs (blue). While gains constitute only a small subset of all OGs in both suborders, the proportion of gains is much larger among candidate OGs in Polyphaga. The nodes indicated by blue circles in the Polyphaga sub-tree lead to species-rich clades containing species that are largely phytophagous (e.g., Chrysomelidae and Curculionidae, respectively Chrys. and Curc.) and experienced larger proportions of gains among the candidate OGs. The Benchmarking Universal Single-Copy Ortholog (BUSCO) scores indicate the relative levels of completeness and putative gene duplications for the genome-based and transcriptome-based datasets in terms of 1,658 BUSCOs from the insecta_odb9 assessment dataset.

Signatures of adaptive expansion are more prevalent in Polyphaga

All 910 OGs with significant variations in their gene content were tested for signatures of adaptive

expansion in each suborder, by comparing Brownian motion (BM, neutral) to Ornstein-Uhlenbeck (OU,

selective pressure) evolutionary models (Beaulieu et al., 2012). As mentioned previously, these

included 26 OGs that belong to one of the functional categories listed in Table 2 ("candidate" OGs). The

models consider per-species gene count as a trait that can evolve towards a value, which may or may

not differ between the two suborders and may or may not be guided by selective pressure; we call this

the "optimum" value in models integrating selection. In total, 21 OGs displayed a higher optimum for

Adephaga (0.2% of the initial 9,720 OGs) and 88 for Polyphaga (0.9%). Eight of these 88 polyphagan

OGs (Table 3 and gene trees in Figure 2 and Supplementary Figures 3-9) are candidate OGs belonging

to one of the candidate gene families of Table 2, while none of the 21 adephagan OGs belong to any

of the candidate gene families. The proportion of OGs with expansions and higher optima in the

background (all "candidate" and remaining "control" OGs) was significantly larger for Polyphaga

compared to Adephaga (2-sample test for equality of proportions, chi-squared, 88/9720 vs. 21/9720, p-

value < 1e-09), indicating that Polyphaga have experienced globally more LSE under selection on

protein-coding genes. Furthermore, a test for enrichment (see Methods) of OGs with LSE under

selection from the candidate families (Table 2) compared to the background was significant for

Polyphaga (8/91 vs. 88/9720, p-value < 1e-09). The same test applied individually on each candidate

gene family within the candidate dataset demonstrated that categories enriched for LSE under selection

in Polyphaga were GSTs (3/6 positive tests, fdr-corrected p-value < 1e-09) and CEs (3/19, fdr-corrected

p-value < 0.005), as shown in detail in Table 4.

Figure 2. Molecular phylogeny from the largest glutathione S-transferase (GST) orthologous group among those exhibiting lineage-specific expansions driven by selection. Red labels indicate genes belonging to species of Polyphaga, accounting for 98 out of 152 genes (their Ornstein-Uhlenbeck per-species optimum is 11.69 vs. 6.85 for Adephaga (blue labels), see Table 3). The presence of several clades of polyphagan and adephagan genes delineates duplication events following the divergence of the two suborders. Encircling the gene labels are red bars that highlight polyphagan clades with bootstrap support of >50% and yellow bars that highlight intra-specific duplications with bootstrap support of >50%. Corresponding full names of species are given in Table 1. Branch lengths represent substitutions per site and bootstrap support below 50% is not displayed.

Table 3. Candidate orthologous groups (OGs) with CAFE overall p-values p<0.01 for which a model

favoring selection for larger sizes in Polyphaga showed a greater likelihood. OG identifiers for functional

category cytochrome P450s (P450), carboxylesterases (CE), glutathione S-transferases (GST), and

cysteine proteases (CYS) are from OrthoDB v8 (ODB8 ID). Small-sample-size corrected Akaike

Information Criterion (AICc) values are reported for all tested models. BM1 (Brownian motion with a

single rate for the whole tree), BMS (Brownian motion with different rates for each regime), OU1

(selection towards the same optimum for both regimes), all representing the null hypothesis. OUM

(selection towards two optima, same variance) and OUMV (selection towards two optima, two

variances), representing the alternative hypotheses. The mean values in each suborder (Adephaga

versus Polyphaga) are presented in the last two columns. Values in bold font indicate the preferred

(maximum likelihood) model. A delta AICc > 2 is required for H1 to be retained.

Category	ODB8 ID	BM1 AICc H0.1	BMS AICc H0.2	OU1 AICc H0.3	OUM AICc H1.1	OUMV AICc H1.2	Mean Adephaga	Mean Polyphaga
P450	EOG805VG7	148.37	153.21	143.35	148.10	**137.95**	34.13	34.47
CE	EOG87DCWX	143.23	143.77	143.63	**138.91**	141.15	6.55	18.78
CE	EOG8KD911	87.08	91.23	82.90	89.10	**79.74**	0.89	2.86
CE	EOG876NDC	80.64	85.67	80.08	87.42	**77.23**	1.72	3.48
GST	EOG87WR3Z	86.24	87.76	76.05	74.40	**72.12**	1.71	3.16
GST	EOG81RS7Z	108.77	114.44	109.19	113.76	**103.79**	6.85	11.69
GST	EOG85F05D	117.62	115.88	111.53	107.62	**106.32**	5.69	9.16
CYS	EOG8JDKNM	91.85	91.62	89.74	**85.66**	88.25	1.80	3.78

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Table 4. Gene family category and candidate orthologous group (OG) enrichments among positive

results. The top panel presents the statistical significance of each test for enrichment of candidate gene

families among the positive results when compared to the background, for Polyphaga. The lower panel

indicates the number of positive results in both suborders, for candidate OGs and background.

Significant values at the 0.05 threshold are shown in bold.

Category	Positive/Total OGs	Category enrichment FDR
P450	1/22	0.36268
CE	3/19	**0.00252**
GST	3/6	**0.00016**
CYS	1/7	0.16627
UGT	0/4	1.00000
SER	0/4	1.00000
ABC	0/28	1.00000
GH	0/1	1.00000
Category	Positive/Total OGs	Candidate vs. background enrichment p-value
Background (Polyphaga)	88/9,720	
Candidates (Polyphaga)	8/91	**0**
Background (Adephaga)	21/9,720	
Candidates (Adephaga)	0/91	1

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Discussion

Comparative genomic analyses often highlight expanded gene families and link these expansions to

biological functions peculiar to, or of special interest in, their focal organism(s). However, these analyses

usually do not explicitly test for any hypothesized evolutionary model that might support such links. Here

we test a specific hypothesis of adaptation to a phytophagous diet, by comparing candidate gene family

repertoires from nine adephagan (a mostly predaceous suborder) and nine polyphagan (a highly

phytophagous suborder) beetle species. These candidate families are putatively involved in

detoxification of plant allelochemicals and digestion of plant tissues. Specifically, we identify evidence

for potentially adaptive gene family expansions in the species rich Polyphaga. This result is robust to

potentially confounding factors that could arise from combining genomic and transcriptomic datasets,

conservative definitions of candidate gene families, or the greater species richness of the Polyphaga

(see discussion points below and Supplementary Information). Through explicitly testing for adaptive

LSEs, these results offer conclusive supporting evidence for the key evolutionary role of the

phytophagous trophic niche in driving gene family expansions in Coleoptera (specifically Polyphaga), a

feature that likely facilitated adaptation of polyphagan beetles to specialized plant feeding.

Dataset heterogeneity

For the comparison of gene repertoires between the two groups to be unbiased, the gene content of all

analyzed species should be of similar accuracy and completeness. The number of predicted proteins

for the genomic resources for each beetle species (Table 1, mean 15,977 and standard deviation 3,748)

was within the range expected of insects (see Waterhouse, 2015). The average total gene count for

Adephaga species (all transcriptomes) was about 4,200 fewer than for Polyphaga, which include two

genomes with more than 22,000 genes. This difference in average gene counts is reduced to just 1,384

when considering only genes assigned to the 9,720 OGs selected for the analysis. Our conservative

orthology filtering therefore ensured that the comparisons focused on gene families with reliably

traceable evolutionary histories that span both groups of beetles. Secondly, assessments of

completeness showed that the majority of the datasets contained more than 90% of complete BUSCOs

(Figure 1, Table 1). While the dynamically evolving families that are the focus of this study are clearly

not universal single-copy orthologs, the high levels of BUSCO completeness support the assumption

that the datasets represent good coverage of the species' gene content. Re-analyses of our data that

exclude the two adephagan beetle species with fewer than 80% complete BUSCOs reduced the power

of the model tests but nevertheless still identified the three GST OGs that favor a model with a higher

optima for Polyphaga (see Supplementary Results). Three of the adephagan transcriptomes showed

more than 10% of duplicated BUSCOs, which could have arisen from suboptimal filtering of the

transcriptomes, i.e. failure to remove alternative transcripts of the same gene. While such potentially

inflated gene counts for these adephagans might prevent the identification of some true expansions in

Polyphaga, they do not invalidate those that were identified. Finally, half of the OGs representing

positive results showed a higher mean value for polyphagan transcriptomes than genomes, including

the three GST OGs (Supplementary Table 1), and explicitly testing for effects due to using both genome

and transcriptome data for the species of Polyphaga, by performing a modified OUwie analysis with

data type as the regime under selection, identified only one CE (EOG8KD911) for which the favored

model linked gene family expansion to species with genomes (see Supplementary Results).

Candidate OG identification

The annotation strategy was designed to link OGs to candidate gene families based on manually

selected keywords used to filter sequence search results, as well as Pfam and InterPro identifiers (Table

2), with the aim of excluding false positives (see Methods). This conservative strategy may not have

fully captured all possible candidate OGs, which would therefore have remained in the background set

of OGs that were used as controls. For example, we identified nine GST OGs (six were retained as

candidates after filtering) while ten subclasses have been identified in arthropods (Roncalli et al., 2015).

While the strict (conservative) strategy we employed to identify candidate OGs may have resulted in an

underestimate of the extent of the observed effects, this does not invalidate those that were identified.

In addition, filtering the OGs to retain only those with genes from both Adephaga and Polyphaga

excluded from the analyses any genes that were specific to either suborder. These might include genes

with key roles in phytophagy, e.g. enzymes acquired by horizontal gene transfer identified from the *A.*

planipennis, *A. glabripennis*, and *D. ponderosae* genomes (McKenna et al., 2016). While

acknowledging their importance, here we explicitly tested for adaptive LSE in one lineage versus the

other so gene evolutionary histories were required to span the two suborders and thus be traceable to

their last common ancestor.

The more speciose Polyphaga exhibit more dynamic gene repertoire evolution

The μ and λ values reported by CAFE on all 9,720 OGs are consistent with assessments of other insect

clades (Hahn et al., 2007; Neafsey et al., 2015). Although the overall gain rate is slightly higher than

the loss rate, the number of OGs losing genes reported by CAFE at each individual node is generally

larger than the number of OGs with gains. This is reconciled by considering that across Coleoptera

many OGs lost a few genes while few families gained many genes. As most OGs display a low number

of genes per species, i.e. they are evolving under 'single-copy control' (Waterhouse et al., 2011), losing

more than one ortholog per species is understandably rare, while there is no theoretical limit for an OG

to gain new members. Comparing the two clades, Polyphaga has a higher rate of gene gain and six

times more OGs with gains, and while the Adephaga rate of gene loss is higher, Polyphaga have 1.5

times more OGs that have experienced gene losses. Hence, the gene repertoires of Polyphaga exhibit

a more dynamic evolutionary history with more gains (rate) in more OGs (counts) and fewer losses

(rate) spread out over more OGs (counts). It is possible that this greater dynamism may be generally

linked to the greater species richness of Polyphaga, with no specific role for phytophagy underpinning

this trend. However, among the candidate OGs for detoxification and digestion there are also more

gains in Polyphaga, and, in contrast to the background, fewer losses. Thus, both gain and maintenance

are higher for candidates in Polyphaga, which is consistent with a key role for phytophagy in driving

dynamic gene repertoire evolution, and particularly LSEs.

Evidence for adaptive expansions of gene families involved in detoxification in polyphagan beetles

In addition to observing more expansions among candidate OGs in the suborder Polyphaga, the positive

results from the OUwie analysis support the hypothesis that selective pressures drive detoxification

enzymes towards larger gene family sizes. This is especially pronounced for GSTs, for which half of

the OGs tested positive and for which a significant enrichment compared to the background was found.

The importance of GSTs in dietary shifts to phytophagy has been noted in mustard-feeding flies, where

duplicated GSTs involved in the mercapturic acid pathway showed signatures of positive selection

(Gloss et al., 2014). Our results therefore suggest that comparable phenomena have been acting at the

level of polyphagan beetles. The CEs also show a statistically significant enrichment for positive results

compared to the background, further supporting the diet detoxification hypothesis. The other positive

results include a P450 OG and a CYS OG, neither of which led to a category enrichment compared to

the background. The P450 OG is by far the largest among the positive results (Supplementary Figure

3), highlighting the importance of P450s in beetle (and generally insect) physiology with diverse roles

beyond detoxification, e.g. hormone biosynthesis (Kong et al., 2014). However, while considered as

significantly expanded and under selection by our model, the actual mean values in the suborders are

not dramatically different. Importantly, the enrichment of positive results among candidates still holds if

this OG is excluded (see Supplementary Results). The involvement of P450s in many other processes

may explain why a broader difference between the suborders was not identified. Apart from one positive

result among the cysteine proteases (no significant category enrichment), our study did not highlight

additional expansions in other digestive enzymes or in transporters within a suborder. The lack of

evidence for expansion in Polyphaga with respect to ABC transporters, which is the candidate functional

category encompassing the highest number of OGs, may indicate that the ancestral diversity of

transporters was sufficient for maintaining the excretion of toxins, despite variations in the substrates

imposing a selective pressure on early stages of the detoxification pathway. Alternatively, if such

pressure were acting on later stages of the pathway, i.e. transporters, its strength could have been too

low for the detection power of our methods and data, unlike for GSTs or CEs.

Conclusions

By comparing the degree of expansion among gene families involved in detoxification of plant

secondary compounds in two suborders of beetles characterized by generally contrasting trophic niches

(i.e., Polyphaga contain a high proportion of phytophagous species while Adephaga encompass mostly

predacious species), we provide molecular genetic evidence supporting the popular hypothesis that

Coleoptera species richness may be in part explained by their interaction with land plants. Candidate

OGs of GSTs, CEs, P450s, and CYSs tested positive for adaptive LSEs in the phytophagous

polyphagan beetle lineage, and categories of GSTs and CEs in particular, were enriched for OGs with

such adaptive LSEs. Moreover, across all OGs tested, Polyphaga exhibited significantly more adaptive

LSEs than Adephaga. This indicates that genes other than the candidate detoxification and digestion

enzymes, which could include genes with functions less obviously related or unrelated to phytophagy,

are also likely to have played a role in the adaptive success of Polyphaga. While this suggests that

additional functional categories remain to be explored, contrasting gene family evolution across the two

major suborders of beetles demonstrates a role for interactions with plant secondary compounds, and

supports a role for phytophagy in general, as important drivers of the remarkable radiation of

polyphagan beetles.

Methods

A chart summarizing the main steps of the analysis is available as Supplementary Figure 10.

Data sources

This study included six genomes and 13 transcriptomes representing a balanced sampling of polyphagan and adephagan beetles, along with one representative of the sister group to Coleoptera, Strepsiptera, to root the species phylogeny. Annotated gene sets from four genomes were sourced from the i5k pilot project datasets (Robinson et al., 2011) (*Anoplophora glabripennis* v0.5.3 (McKenna et al., 2016), *Leptinotarsa decemlineata* v0.5.3 *(Schoville et al., 2018), Onthophagus taurus* v0.5.3*, Agrilus planipennis* v0.5.3) and two were independently published *Dendroctonus ponderosae* Ensembl Metazoa v1.0 (Keeling et al., 2013) and *Tribolium castaneum* Ensembl Metazoa v3.22 (Tribolium Genome Sequencing Consortium et al., 2008). One Polyphaga transcriptome, *Laparocerus tessellatus* (Supplementary Methods), was sequenced for this project and the others were provided by the 1KITE project (Supplementary Methods, http://www.1kite.org, Misof et al., 2014, Peters et al., 2017). A detailed list is presented in Table 1.

Coding sequence predictions, transcriptome and genome quality assessments

Coding sequences and peptide sequences were predicted from all transcriptomes using TransDecoder (v2.0.1 https://transdecoder.github.io [last accessed May 4th, 2017]) along with a custom python script to retain the best-scoring entry among overlapping predictions. The coding sequences and peptide sequences from the genomes were retrieved from their official annotated gene sets. All genome and transcriptome gene sets were assessed using BUSCO (v2.0, python 3.4.1, dataset insecta_odb9/2016-10-21, mode proteins) (Waterhouse et al., 2018). This tool identifies near-universal single-copy orthologs by using hidden Markov model profiles from amino acid alignments. CD-HIT-EST v4.6.1 (Li and Godzik, 2006) was run on the protein sequences with a 97.5 percent identity threshold to ensure that all species datasets were filtered to select a single isoform per gene.

Orthology delineation

The OrthoDB (Kriventseva et al., 2015) hierarchical orthology delineation procedure was employed to predict orthologous protein groups (OGs). Briefly, protein sequence alignments are assessed to identify

all best reciprocal hits (BRHs) between genes from each pair of species, which are then clustered into

OGs following a graph-based approach that starts with BRH triangulation. The annotated proteins from

the genomes of *A. planipennis*, *O. taurus* and all transcriptomes were mapped to OrthoDB v8 at the

Arthropoda level (with 87 species including four of the beetles with sequenced genomes). Mapping

uses the same BRH-based clustering procedure but only allows genes from mapped species to join

existing OGs. These OGs were then filtered to identify the 9,720 OGs with representatives from both

Polyphaga and Adephaga to focus the study on OGs with evolutionary histories traceable to the last

common ancestor of all the beetles, i.e. 5,188 OGs with genes from only one of the two suborders were

removed.

Species phylogeny

To build an ultrametric phylogeny required for the CAFE analyses, the maximum likelihood molecular

species phylogeny was first estimated based on the concatenated superalignment of orthologous amino

acid sequences from each of the datasets. Protein sequences of single-copy BUSCO genes and the

best-scoring duplicated genes present in all species were individually aligned for each set of BUSCO-

identified orthologs using MAFFT with the --auto parameter (Katoh and Standley, 2013) and each result

was manually reviewed to exclude poor-quality alignments. Four hundred and five alignments were

retained out of 436 and concatenated into a superalignment, partitioned according to the best model

for each set of orthologs using aminosan 1.0.2015.01.23 (Tanabe, 2011). RAxML v8.1.2 (-f a -m

PROTGAMMA -N 1000) (Stamatakis, 2014), and used to compute the maximum likelihood tree. The

monophyly of Geadephaga and Hydradephaga was constrained to match the generally accepted

resolution of Adephaga (as in McKenna et al., 2015). The chronos function of the R package ape (v3.4

on R 3.2.1, relaxed model) (Paradis et al., 2004) was used to obtain an ultrametric tree and the tip to

root length was adjusted to match the approximately 250 million year evolutionary history of crown

group Coleoptera (McKenna et al., 2015).

Functional annotation and definition of candidate genes

InterProScan was run on all species protein sets (-appl Pfam --goterms, 5.16.55) (Jones et al., 2014)

to identify protein families. Additionally, blastp 2.3.0 (Altschul et al., 1997; Camacho et al., 2009) was

run against uniref50 (version Jun 22, 2016; Suzek et al., 2015) with an e-value cut-off of 1e-20. An OG

was included in the set of candidate OGs when it had a match to both the uniref50 clusters and Pfam

families (Finn et al., 2016) or gene ontologies (The Gene Ontology Consortium, 2017) as detailed in

Table 2.

CAFE analysis

The number of genes in OGs for each species were counted. All candidates and remaining (control)

OGs were pooled together and processed with CAFE 3.1 (Han et al., 2013), to infer gene family

evolution in terms of gene gains and losses. First, the python script provided by CAFE was used to

estimate the error in our dataset. The CAFE software was then run using the mode in which the gain

and loss rates are estimated together (λ) and a second mode in which they are estimated separately

(gains=λ, losses=μ). The more complex model was retained as it reached a significantly better score (-

199,989 for a single estimated parameter and -199,981 for two distinct estimated parameters, 2x delta

log-likelihood = 16, chi-squared distribution, df=1). For the entire analysis, the CAFE overall p-value

threshold was kept at its default value (0.01). To run CAFE on each suborder separately, the newick

file was pruned to retain only required species using newick utils 1.1.0 (Junier and Zdobnov, 2010).

Evolutionary models

To evaluate adaptive OG expansion, the likelihood of the count data was tested by optimizing

parameters considering two methods provided by the OUwie R package v1.51 (Beaulieu and O'Meara,

2016; Beaulieu et al., 2012). First, a Brownian motion (BM) approach was used, which assumes no

selection and thus differences result from a stochastic process whose rate is estimated. Second,

Ornstein-Uhlenbeck (OU) models were used. They take into account an optimal family size which is

obtained by selective pressure. Two groups were defined in the phylogeny, namely Polyphaga and

Adephaga, to which the two different regimes to consider were assigned, plus a third regime to the root.

This represents a simplified scenario allowing for the comparison of gene contents between one group

and the other rather than attempting to estimate 'levels' of phytophagy or zoophagy across the

phylogeny. The models BM1 (Brownian motion with a single rate for the whole tree), BMS (Brownian

motion with different rates for each group), OU1 (selection towards the same optimum for both groups)

were optimized as null hypotheses (H0) and compared to OUM (selection towards two optima, same

variance) and OUMV (selection towards two optima, two variances) models as alternative hypotheses

(H1). The Akaike information criterion corrected for sample-size (AICc) (Hurvich and Tsai, 1989) was

used to compare models and an AICc>2 between the best H0 and the best H1 model was considered

as significant to prefer the H1 model.

Statistical enrichment

All results for candidates and controls were pooled together to obtain a background distribution of

positive and negative results. Positive results are those OGs that passed the OUwie analysis, and

negative results are all of the 9,720 OGs that did not obtain a significant overall CAFE p-value or did

not pass the OUwie analysis. Then, 100,000 random draws (using the R function sample, without

replacement) having the sample size of the candidate category to test for enrichment were taken from

the background and the significant outcomes for Polyphaga and Adephaga were counted. A p-value

was calculated for each group as follows: the number of random draws reaching the amount of

significant outcomes found for the candidate category, or more, divided by 100,000. Additionally, the

multiple tests conducted on each individual candidate category were corrected for false discovery rate

(FDR) using the R p.adjust function (method BH, Benjamini Hochberg).

Gene trees

The alignments for the gene trees were produced using MAFFT with the --auto parameter. The gene

trees were computed with RAxML v8.1.2 (-f a -m PROTGAMMALGF -N 100) and plotted with EvolView

(He et al., 2016; Zhang et al., 2012).

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Author Contributions

MS and NA conceived the study. PI delineated orthology. CP prepared the libraries for *L. tessellatus* transcriptome sequencing. HEE, DDM, BM, SS, and XZ provided access to 1KITE transcriptome data. MS conducted the analyses. BCE, MRR, JR, RMW and NA supervised the analyses. MS, RMW, and NA wrote the manuscript, with input from all authors.

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