Gene Structure-Based Homology Search Identifies Highly Divergent Putative Effector Gene Family

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Abstract

Homology of highly divergent genes often cannot be determined from sequence similarity alone. For example, we recently identified in the aphid *Hormaphis cornu* a family of rapidly evolving *bicycle* genes, which encode novel proteins implicated as plant gall effectors, and sequence similarity search methods yielded few putative *bicycle* homologs in other species. Coding sequence-independent features of genes, such as intron-exon boundaries, often evolve more slowly than coding sequences, however, and can provide complementary evidence for homology. We found that a linear logistic regression classifier using only structural features of *bicycle* genes identified many putative *bicycle* homologs in other species. Independent evidence from sequence features and intron locations supported homology assignments. To test the potential roles of *bicycle* genes in other aphids, we sequenced the genome of a second gall-forming aphid, *Tetraneura nigriabdominalis* and found that many *bicycle* genes are strongly expressed in the salivary glands of the gall forming foundress. In addition, *bicycle* genes are strongly overexpressed in the salivary glands of a non-gall forming aphid, *Acyrthosiphon pisum*, and in the non-gall forming generations of *H. cornu*. These observations suggest that Bicycle proteins may be used by multiple aphid species to manipulate plants in diverse ways. Incorporation of gene structural features into sequence search algorithms may aid identification of deeply divergent homologs, especially of rapidly evolving genes involved in host-parasite interactions.

Key words: gene homology, gene structure, genome evolution, rapidly evolving genes, effector genes, aphid gall.

Significance

Many related genes, and especially genes involved in host-parasite interactions, have changed so extensively between species that similarities in the primary amino acid sequence are no longer detectable by standard search methods. The structure of such genes, however, such as the lengths of exons and the positions of introns, often evolves more slowly than the gene sequences themselves. We have exploited this relatively slow evolution of gene structure to identify a large family of rapidly evolving genes that are expressed in the salivary glands of plant sap-sucking insects and that may facilitate their ability to parasitize plants.

Introduction

One challenge that is faced by evolutionary and functional genetic studies is that many genes have been identified as “orphan” or lineage-specific genes because homologous sequences cannot be identified outside of a limited taxonomic range. Many lineage-specific genes may not be truly novel, however, since sequence divergence can cause homologs to become undetectable by sequence-search methods (Vakirlis et al., 2020; Weisman et al., 2020). Identifying such extremely divergent homologs remains a significant bioinformatic challenge and limits functional inferences derived from homology (Loewenstein et al., 2009). Genes involved in host-parasite systems often evolve extremely divergent sequences as a result of genetic conflict (Paterson et al., 2010; Eizaguirre et al., 2012). Aphids and
their host plants represent one such antagonistic pair. Aphids are small insects that feed by inserting their thin mouthparts (stylets) into the phloem vessels of plants to extract nutrients (Dixon, 1978). Like many herbivorous insects, aphids introduce effector molecules into plant tissues to manipulate the physiology and development of plants to the insects’ advantage (Mutti et al., 2008; Hogenhout and Bos, 2011; Elzinga and Jander, 2013; Rodriguez et al., 2017). For example, aphids introduce calcium binding proteins that prevent the plant’s ability to block phloem cell transport in response to phloem vessel damage (Will et al., 2007). It is believed that aphids introduce a wide range of effector proteins and that these molecules contribute to the debilitating effects of multiple aphid species on plants, which imposes significant financial damage on most major agricultural crops worldwide (Schaible et al., 2008).

In an extreme form of aphid manipulation of plant physiology and development, some aphid species induce novel plant organs, called galls, which provide the aphids with a ready food source and with protection from the elements and from natural enemies (Shorthouse et al., 2005; Giron et al., 2016). Aphid galls can be significant nutrient sinks (Burstein et al., 1994), demonstrating that galling aphids induce both local and long-range changes to plant physiology. Many arthropods induce galls on a wide diversity of plant species (Mani, 1964), but the complex life cycle of galling aphids provides unique advantages for studying the mechanisms of gall induction that are not found in other galling arthropods. In most galling aphid species, only one generation each year, the gall foundress or fundatrix, appears to be capable of inducing galls. Individuals of subsequent generations in the life cycle do not induce galls, even though they may feed on the same host plant.

Comparisons of genes transcribed specifically in salivary glands of individuals across generations provided some of the first clues to the putative effector proteins contributing to gall induction (Korgaonkar et al., 2021).

We reported recently that the genome of one species of galling aphid encodes 476 related genes encoding diverse presumptive effector proteins with no sequence homology to previously described proteins and that many of these proteins contain two cysteine-tyrosine-cysteine (CYC) motifs (Korgaonkar et al., 2021). These genes were therefore named bicycle (bi-CYC-like) genes. A genome-wide association study revealed that variation in one bicycle gene, called determinant of gall color (dgc), was strongly associated with a red versus green gall color polymorphism and this genetic polymorphism was associated with strong downregulation of dgc in aphids and strong upregulation of a small number of genes involved in anthocyanin production in plant galls. In addition, the majority of bicycle genes are strongly upregulated in the salivary glands specifically of the aphid generation that induces galls (the fundatrix). Together with genetic evidence that dgc regulates the gall phenotype, the enrichment of bicycle genes expressed in fundatrix salivary glands suggested that many bicycle genes contribute to gall development or physiology.

The primary amino-acid sequences of Bicycle proteins have evolved rapidly, apparently in response to extremely strong positive selection (Korgaonkar et al., 2021). In preliminary studies, we attempted to identify bicycle homologs in other insect species using sequence similarity algorithms such as Basic Local Alignment Search Tool (BLAST) (Altschul et al., 1990) and hmmer (Johnson et al., 2010; Eddy, 2011), but identified few putative homologs. It was therefore unclear if bicycle genes are a recently evolved family of genes or an ancient family that has evolved highly divergent coding sequences.

Extremely divergent homologs have been identified previously using features of gene structures that evolve more slowly than the coding sequence, such as exon sizes and intron positions (Bazan, 1991; Brown et al., 1995; Betts, 2001), and these characteristics of genes have also proven to be valuable for phylogeny reconstruction (Rokas et al., 1999; Venkatesh et al., 1999; Rokas and Holland, 2000; Telford and Copley, 2011). We noted previously that bicycle genes have unusual gene structures containing many microexons (Korgaonkar et al., 2021), and we considered the possibility that bicycle gene structure may be more conserved than bicycle gene sequences. To explore the evolutionary history of bicycle genes, we therefore sought a method that would allow identification of highly divergent bicycle gene homologs that did not rely on sequence similarity. To accomplish this, we built a logistic regression classifier based only on structural features of bicycle genes. We found that this classifier was very accurate, despite the fact that it does not include any sequence information. This classifier identified many highly-divergent bicycle homologs in Hormaphis cornu that could not be identified by sequence similarity search methods, including genes that encode proteins that do not possess the previously canonical CYC motif. In addition, the classifier identified many highly divergent candidate bicycle homologs in all aphids, phylloxerans, and scale insects we studied. We did not detect any putative homologs in three progressively basal outgroups. Multiple sequence alignment of these putative homologs revealed that many contain N-terminal signal sequences and CYC motifs, consistent with their assignment as bicycle homologs. In addition, gene-structure aware sequence alignment revealed multiple apparently shared intron boundaries between putative homologs that share little obvious sequence similarity. In addition, putative bicycle homologs are highly enriched in the salivary gland mRNA of two gall forming and one non-gall forming aphid. All of this evidence supports the hypothesis that bicycle gene homologs encode effector proteins in gall-forming and non-gall-forming aphids and also in phylloxerans and scale insects.
Results

Sequence-Based Homology Searches Find Few Bicycle Genes Outside of *H. cornu*

*Bicycle* genes in *H. cornu* were identified originally as a subset of previously unannotated genes that were enriched in the salivary gland mRNA of the gall-inducing foundress (Korgaonkar et al., 2021). *Bicycle* genes were found to be the single largest group of such genes that shared sequence similarity. These 476 genes are extremely divergent from one another at the amino-acid sequence level and appear to have evolved rapidly due to positive natural selection. Using sequence-based homology search methods *BLAST* and *hmmer*, we identified 34 additional candidate *bicycle* paralogs in *H. cornu* (using *BLAST*) and some putative homologs in other aphid species (in total 15 with *BLAST* and 81 with *hmmer*). We did not detect any candidate *bicycle* orthologs in genomes from insects of the families Phylloxeridae, Coccoidea, Psylloidea, Aleyrodoidea, and Fulgoroidea (fig. 1).

Gene synteny conservation can be used to identify highly divergent orthologs (Tekaia, 2016). We observed that within the *H. cornu* genome many closely related *bicycle* genes were located on separate chromosomes (supplementary fig. S1A, Supplementary Material online), suggesting that *bicycle* genes undergo transposition at a high rate and that *bicycle* genes are unlikely to display extensive conserved synteny. To more directly test whether conserved gene synteny may allow identification of *bicycle* homologs across the species studied here, we searched for conserved gene synteny near orthologous *bicycle* genes that were identified between *H. cornu* and *Tetraneura nigriabdominalis* (fig. 1) as reciprocal best *hmmer* hits. Bicycle orthologs shared many fewer homologous genes in their flanking regions than randomly selected subsets of genes (supplementary fig. S1B, Supplementary Material online), further suggesting that conserved synteny is unlikely to be a useful measure to identify *bicycle* homologs across the species studied here. We therefore sought a different method to identify highly divergent *bicycle* homologs.

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**Fig. 1.**—The gene structure-based classifier detects *bicycle* genes in genomes of aphids, phylloxerids, and coccids. A whole-proteome phylogeny of the species studied is shown on the left. Family membership is shown in color beside the species names. The number of candidate *bicycle* homologs detected by *BLAST*, *hmmer*, and the gene structure-based classifier is shown to the right of the phylogeny. A second *hmmer* search was performed starting with the candidate *bicycle* homologs identified by the gene structure-based classifier within each species, and the results of this search are shown in the far right column. Scale bar represents 0.1 substitutions per residue.

| Aphididae | Phylloxeridae | Coccoidea | Psylloidea | Aleyrodoidea | Fulgoroidea |
|-----------|--------------|-----------|------------|-------------|-------------|
| BLAST     | HMMER        | Classifier | *bicycle* genes |
| 0         | 0            | 15        | 12         |
| 2         | 1            | 117       | 59         |
| 1         | 3            | 121       | 69         |
| 1         | 0            | 64        | 31         |
| 3         | 2            | 29        | 15         |
| 2         | 3            | 65        | 32         |
| 1         | 0            | 27        | 40         |
| 0         | 1            | 3         | 8          |
| 1         | 53           | 279       | 69         |
| 0         | 3            | 27        | 23         |
| 4         | 15           | 70        | 41         |
| 510       | 458          | 653       | 446        |

| Laodelphax striatellae |
|------------------------|
| 0 0 0                    |

| Nilaparvata lugens      |
|-------------------------|
| 0 0 0                    |
A Classifier Based on Only Gene Structural Features Can Identify Bicycle Genes with High Probability

Sequence-based search methods suggested that bicycle genes may be found in other species, but the low number of genes identified, and the incongruence between results from BLAST and hmmre suggested that sequence-based search may provide low sensitivity to detect bicycle homologs (Vakirlis et al., 2020; Weisman et al., 2020). We noticed, however, that while sequence divergence was high at the amino-acid level among H. cornu bicycle genes, non-sequence specific features of bicycle genes appeared to be relatively well conserved. For example, bicycle genes contain a large number of unusually small internal exons (Kargaonkar et al., 2021), making them clear outliers to most genes in the H. cornu genome. We also observed that almost all internal exons of bicycle genes start with codon position 2, that is, they exhibit exon phase 2 (supplementary fig. S1C, Supplementary Material online), which is an extremely different distribution of exon phases compared with other genes in the genome (supplementary fig. S1D, Supplementary Material online). Genes containing many exons of similar size all with the same phase are rare in most genomes (Ruivinsky and Watson, 2007).

We hypothesized that some aspects of bicycle gene structure may evolve more slowly than the primary protein sequences and thus provide an evolutionary signal to detect distantly related bicycle homologs (Aguilar et al., 2015). For example, the pattern of exon phases can allow identification of highly divergent homologous genes (Roy and Gilbert, 2005; Ruivinsky and Watson, 2007). We therefore explored whether a linear logistic regression classifier could identify bicycle homologs based on the following structural features of each gene: length of gene in genome, from start to stop codons; first and last exon length; internal exon mean length; length of the sum of codons; and number of internal exons that start in phase 0, 1, or 2 (supplementary fig. S2A, Supplementary Material online). Since almost all internal bicycle exons are of phase 2, there is no additional information from the order of exon phases in bicycle genes, and we therefore simply counted the number of exons of each phase.

Characterized bicycle genes have an average of approximately 17 and a minimum of 5 exons (supplementary fig. S2E, Supplementary Material online), which is strongly different from the genome-wide background (supplementary fig. S2F, Supplementary Material online) (two-sample Kolmogorov–Smirnov test P-value < 2.2 × 10$^{-16}$), and Bicycle protein lengths are well explained by total exon number (c.f. supplementary fig. S1G and S1H, Supplementary Material online). We searched predicted genes containing at least four exons so that we could accurately summarize internal exon mean lengths and number of internal exons that start in each of the three phases. Thus, if any bicycle genes have evolved to have three or fewer exons, then they could not be detected with this classifier.

The classifier exhibited extremely good performance with all data partitions (supplementary fig. S2B, Supplementary Material online) and, therefore, a full model was constructed using all bicycle and annotated genes. (Non-annotated genes were excluded from model training and validation because we hypothesized that this set may include additional bicycle homologs.) The classifier categorized the vast majority of genes as either non-bicycle or bicycle genes with high probability (supplementary fig. S2C, Supplementary Material online). To decide on a probability cutoff to identify putative bicycle homologs, we considered the trade-off between precision (the proportion of true positives among both true and false positives) and recall (the proportion of true positives among true positives and false negatives). We chose a model probability cutoff that equalized precision and recall (P/R = 1), which yielded precision and recall values of 0.99 (supplementary fig. S2D,E, Supplementary Material online).

To determine whether all gene features were required for accurate classification, we removed one factor at a time, retrained the model, and calculated precision and recall values using the same threshold as the full model. We found that no single factor was solely necessary for model performance (supplementary fig. S2F, Supplementary Material online), although removal of the number of mode 2 exons resulted in the strongest decrement in model performance, resulting in precision and recall values of 0.98 and 0.96, respectively.

To determine whether any single predictor was sufficient to identify bicycle genes with high confidence, we calculated precision and recall values for linear models using one predictor at a time (supplementary fig. S2G, Supplementary Material online). Mean internal exon length alone performed best (supplementary fig. S2G, Supplementary Material online), consistent with the observation that bicycle genes contain an unusually large number of small exons (supplementary fig. S3, Supplementary Material online). The number of internal exons in mode 2 also displayed some predictive power for discriminating bicycle genes from the background (supplementary fig. S2G, Supplementary Material online). The observation that no single predictor was solely necessary for performance but two predictors were individually sufficient to provide good predictive power indicates some correlation between the predictor variables (supplementary fig. S2H–J, Supplementary Material online). Since no predictors were scalar multiples of each other, we employed the full model using all predictors to search for new bicycle genes.

Additional Candidate Bicycle Genes Can Be Found in Many Other Species

To search for candidate bicycle genes in other species, we applied the classifier to all predicted genes from the
Candidate Bicycle Genes in *H. cornu* are Highly Expressed and Share Structural Homology

When we applied the classifier to all genes in the *H. cornu* genome, we identified 653 candidate bicycle genes, compared with the 476 bicycle genes we described in our original study (Korgaonkar et al., 2021). Four originally defined bicycle genes were not detected and are thus false negatives. In addition, 181 newly detected bicycle genes were detected and are thus “false positives.” Since we chose a probability cutoff that should report an approximately equal number of false negatives and false positives, we hypothesized that the vast majority (approximately 181 - 4) of “false positives” are newly discovered bicycle homologs.

We clustered these newly discovered bicycle genes together with the original bicycle genes based on sequence similarity (fig. 2A) and found that 563 represented a single group of related genes (Cluster 2, fig. 2A) that all share sequence similarity with, and included, all but one of the originally defined 476 bicycle genes (fig. 2C).

We also identified a second cluster of 94 genes that encode proteins that do not share strong sequence similarity with the originally defined Bicycle proteins and contain degenerated CYC motifs (Cluster 1, fig. 2A and 2B). Approximately 20 of these genes were strongly enriched in the salivary glands of gall-inducing foundresses, but many of these Cluster 1 genes were strongly enriched in salivary glands of other life stages, especially generations that feed on *Betula nigra* (River Birch) (fig. 2D–F). In addition, approximately 20 bicycle genes containing the CYC motif (Cluster 2) were weakly expressed in fundatrix salivary gland, but strongly expressed at other life stages (fig. 2E–G).

Like the original 476 bicycle genes, the new putative bicycle homologs likely experienced increased levels of positive selection. We found that the new putative bicycle homologs displayed a similar pattern to the original bicycle genes of elevated ratios of non-synonymous (dS) to synonymous (dS) substitutions between *H. cornu* and its closely related sister species *H. hamamelidis* compared to the genome background (supplementary fig. S4A–C, Supplementary Material online). In addition, we performed a genome-wide selection scan using a site frequency spectrum-based composite likelihood ratio test that calculates the likelihood ratio of a sweep at a certain location in the genome to the neutral model. We found that signals of selective sweeps are enriched near both the original bicycle genes and the newly discovered putative bicycle homologs (supplementary fig. S4D, Supplementary Material online).

Since some of these new putative bicycle homologs shared little apparent sequence similarity with the originally defined bicycle genes, we sought additional evidence for homology. We hypothesized that other details of gene structure not employed in the classifier, such as intron positions, might provide additional evidence of shared ancestry. We therefore performed gene-structure-aware multiple sequence alignment of the original bicycle genes and the new putative bicycle genes identified by the classifier (Gotoh, 2021).

Initial attempts to align all 657 candidate proteins were uninformative, and we hypothesized that this was because these proteins display a large range of lengths and exon numbers (supplementary fig. S4E, Supplementary Material online). We therefore divided Bicycle proteins of Cluster 2 into four groups based on protein length (blue dotted lines in fig. 3A). Logo plots of these four groups revealed a striking pattern, proteins contain one, two, three, or four CYC motifs (fig. 3B–E), and we therefore call these unicycle, bicycle, tricycle, and tetracycle genes, respectively. This evolutionary comparison suggests that the CYC motif is a functional unit, which can be multimerized within individual proteins.

We then divided *H. cornu* bicycle homologs into four length categories and performed gene-structure aware protein sequence alignments to search for shared intron position between the originally defined bicycle genes and the newly discovered bicycle putative homologs of
The gene-structure based classifier identifies additional CYC and non-CYC motif containing genes in the *H. cornu* genome that may contribute to the effector protein repertoire of multiple stages of the *H. cornu* life cycle. (A) Hierarchical clustering of candidate *H. cornu* bicycle homologs detected by the gene-structure based on amino-acid sequence similarity measure reveals two clusters of genes with similar sequences. Some of the genes in Cluster 1 appear to have some sequence similarity to Cluster 2 genes. (B) Genes in Cluster 1 encode proteins with N-terminal signal sequences and weak similarity to bicycle genes. (C) Genes in Cluster 2 encode proteins with sequence similarity to the previously described bicycle genes. (D) Diagram of life cycle of *H. cornu*. Individuals from different generations exhibit phenotypes that are specialized for each stage of the complex life cycle. The fundatrix (G1) is the only generation that induces a gall. This panel was published previously as Figure 1J in Korgaonkar et al. (2001) and is available under Creative Commons Attribution License (CC BY 4.0). (E) A volcano plot of the strength of evidence for differential expression (-log10(P value)) versus the fold-change differential expression of salivary glands isolated from fundatrices versus sexuals reveals that some of the Cluster1 (black) and 2 (orange) genes identified by the gene structure based classifier are as strongly enriched in fundatrix salivary glands as the originally described bicycle genes (blue). In addition, several Cluster 1 and 2 genes are more strongly enriched in salivary glands of sexuals. (F–G) Plots of gene expression levels for Cluster 1 (F) and 2 (G) genes across four generations of the life cycle reveals that while most bicycle homologs are most strongly expressed in fundatrix salivary glands (e.g. purple lines), some are weakly expressed in fundatrix salivary glands and then strongly expressed in salivary glands of individuals from other stages of the life cycle (e.g. red lines).
Cluster 1. We identified many evolutionarily shared intron positions between the two classes of putative homologs (supplementary fig. S5, Supplementary Material online).

To determine whether more introns are shared between the original bicycle genes and newly identified putative bicycle genes than expected by chance—for example given an alignment of two groups of unrelated genes—we developed a metric of intron concordance and estimated the null distribution by resampling intron concordance between bicycle genes and three unrelated gene families.

Fig. 3.—H. comu bicycle genes belong to four major categories, unicycle, bicycle, tricycle, and tetracycle genes. (A) The originally identified bicycle genes encode proteins that exhibit a wide size range, with representatives of approximately four size classes, indicated by the vertical dotted blue lines. (B–E) Logo plots of the proteins from each of the four size classes of Bicycle proteins identifies unicycle (B), bicycle (C), tricycle (D), and tetracycle (E) proteins. The CYC motifs in each protein are identified by thick black lines below each Logo plot.
Intron concordance was measured as the correlation coefficient between the number of genes sharing an intron at each alignment position for one gene family versus the bicycle gene family (supplementary fig. S6A–D, Supplementary Material online). Unrelated gene families were found to have intron positions distributed approximately randomly relative to bicycle genes with mean R close to 0 (supplementary fig. S6C,E, Supplementary Material online). In contrast, newly discovered bicycle genes, both from *H. cornu* and from other species, displayed positive R much greater than 0 and significantly different from distributions of R values found for pairs of unrelated gene families (supplementary fig. S6E, Supplementary Material online). This analysis provides statistically significant support for the conclusion that newly discovered putative bicycle genes share introns in the same locations more often than expected by chance between unrelated gene families and thus that they are likely bicycle gene homologs.

These results provide evidence for several conclusions. First, measures of gene structure alone can reliably identify many, but perhaps not all, divergent bicycle genes. Second, some bicycle homologs have evolved highly divergent sequences, raising the possibility that highly divergent homologs may be present in other species but are undetectable with sequence-similarity search algorithms. Third, some bicycle homologs are highly expressed in salivary glands of non-gall forming life stages, suggesting that some of these proteins may contribute to the effector-protein repertoire outside of the context of gall development. Given these observations, we next tested whether the classifier could identify bicycle homologs in a diverging gall ingesting aphid species.

Many Candidate Gall Effector Genes in Tetraneura nigriabdominalis Are Highly Divergent Bicycle Genes

There are two major clades of gall forming aphids, the Hormaphididae, to which *H. cornu* belongs, and the Pemphigidae. These two aphid families are thought to have shared a common ancestor that induced galls. Therefore, to determine whether bicycle genes were present in the common ancestor of gall forming aphids, we assembled and annotated the genome of *T. nigriabdominalis* (supplementary fig. S6A, Supplementary Material online), a gall forming aphid belonging to the Pemphigidae (Álvarez et al., 2013). Details of the genome assembly can be found in Supplementary Methods, Supplementary Material online. We annotated the *T. nigriabdominalis* genome using mRNA sequencing reads generated from salivary glands and carcasses of the fundatrix (G1) and G2 generations (supplementary fig. S6B, Supplementary Material online).

We applied the *H. cornu* bicycle gene classifier to predicted genes of the *T. nigriabdominalis* genome and identified 279 candidate bicycle homologs (fig. 1). In contrast, sequence-based search using BLAST and *hmmer* identified 1 and 53 candidate homologs at E-value < 0.01, respectively. We clustered the 279 *T. nigriabdominalis* genes discovered by the gene-structure based classifier based on their predicted amino acid sequences (fig. 4A) and identified three clusters that included proteins with clear evidence for CYC motifs (fig. 4B–D). One cluster of 94 genes does not exhibit clear CYC motifs (fig. 4E). The extreme sequence divergence of these genes within *T. nigriabdominalis* is supported by the fact that profile *hmmer* search recovered only 69 of these original 279 genes (fig. 1).

To test whether these genes likely shared a common ancestor, we performed gene-structure aware alignments of proteins with and without CYC motifs and identified more shared introns between putative homologs with and without CYC motifs (supplementary fig. S6C, Supplementary Material online) than expected by chance (supplementary fig. S6E, Supplementary Material online). Thus, the *T. nigriabdominalis* genome contains apparent bicycle genes that have become essentially unrecognizable as bicycle homologs at the sequence level.

The discovery of bicycle genes in *T. nigriabdominalis* using the gene-structure based classifier provides evidence that the classifier can identify putative bicycle homologs even when the genes cannot be identified using sequence-based homology methods and that bicycle genes were likely present in the common ancestor of gall-forming aphids. We therefore next tested whether the classifier can detect putative bicycle homologs in an aphid species that does not induce galls.

The Pea Aphid (*Acyrthosiphon pisum*) Genome Includes Many Bicycle Homologs That Are Candidate Effector Proteins

The gene-structure based classifier detected 121 putative bicycle homologs in the genome of *A. pisum*, a species that does not induce galls (Richards et al., 2010; Li et al., 2019). In contrast, sequence-based search using BLAST and *hmmer* identified 1 and 3 candidate hits at E-value < 0.01, respectively. Multiple sequence alignment of the 121 *A. pisum* putative bicycle homologs revealed that most of these genes encode proteins with N-terminal secretion signal sequences and a single CYC motif (fig. 5A).
Fig. 4.—The *T. nigriabdominalis* bicycle homologs detected by the gene structure-based classifier display CYC motifs and highly divergent protein sequences. (A) Hierarchical clustering of candidate *T. nigriabdominalis* bicycle homologs based on amino-acid sequence similarity measure reveals four clusters of genes. (B–E) The four clusters of *T. nigriabdominalis* bicycle homologs display two (B, C) or one (D) CYC motif, or no clear CYC motifs (E). The cluster identities are indicated in each subplot. (F) Volcano plot of genes differentially expressed in fundatrix salivary glands versus carcass with bicycle genes labelled. Most of the *T. nigriabdominalis* candidate bicycle homologs are strongly over-expressed in the fundatrix salivary glands.
introns between *H. cornu bicycle* genes and *A. pisum* putative *bicycle* homologs (supplementary fig. S8, Supplementary Material online) than expected by chance (supplementary fig. S6E, Supplementary Material online), providing independent confirmation that these genes shared a common ancestor.

To explore whether these *A. pisum bicycle* homologs may contribute to the effector gene repertoire of *A. pisum*, we performed RNA-seq of salivary glands and carcasses. We found that 55% of the putative *bicycle* genes are among the most strongly overexpressed genes in salivary glands, along with a similar number of additional unannotated genes with signal peptides (fig. S5B). In addition, 17 proteins encoded by genes with multiple microexons were identified by a previously-published proteomic study of proteins secreted by *A. pisum* into an artificial food medium (Dommel et al., 2020), and we found that at least 13 of these proteins are *bicycle* homologs (supplementary table S1, Supplementary Material online). In summary, the *A. pisum* genome encodes approximately 121 *bicycle* homologs that may contribute to the *A. pisum* effector protein repertoire.

**Bicycle Genes Are Present in Genomes of Aphids, Phylloxerids, and Coccids**

The presence of *bicycle* homologs in three divergent aphid species, *H. cornu*, *T. nigriglandula*, and *A. pisum* implies that *bicycle* genes were present in the common ancestor of aphids, which lived approximately 280 MYA. To further explore the origins and evolution of *bicycle* genes, we downloaded genomes and RNA-seq data for nine additional aphid species and ten outgroup species from NCBI. We annotated predicted genes in all genomes using the same bio-informatic pipeline that we have used previously to discover *bicycle* homologs near the *bicycle* gene. While these homologs are present on multiple contigs, which would lead to over-counting *bicycle* gene homologs. While these problems with current annotations are expected to increase variance of estimates of the number of *bicycle* homologs, we did not detect a dependence of number of *bicycle* homologs detected on genome assembly quality (supplementary fig. S9A, Supplementary Material online) or on the number of genes annotated for each genome (supplementary fig. S9B, Supplementary Material online). In addition, we employed chromosome-level genome assemblies for two outgroup species where no predicted *bicycle* genes were found, *P. venusta* (Li et al., 2020) and *T. vaporariaum* (Xie et al., 2020), suggesting that the failure to identify *bicycle* genes in these species did not result from poor genome assemblies.

We estimated the phylogeny for these 22 species using whole-genome proteomic predictions (Emms, 2019), and this phylogeny is in general agreement with earlier studies based on a small number of genes (von Dohlen and Moran, 1995; Nováková et al., 2013; Johnson et al., 2018) (fig. 1). We detected putative *bicycle* homologs in all aphid species studied here, supporting the inference that *bicycle* genes were present in the common ancestor of aphids (fig. 1). We found extensive variation in the number of *bicycle* genes between species. In all aphid species with a sufficient number of *bicycle* homologs, multiple sequence alignment revealed the presence of CYC domains (supplementary fig. S9C–F, Supplementary Material online). In most Aphidini species (e.g., *M. persicae, A. pisum, P. nigriceps*), most homologs appear to be *unicycle* genes (supplementary fig. S9C,D, Supplementary Material online). In *Cinara cedri*, many homologs appear to be *tetracycles* (supplementary fig. S9F, Supplementary Material online).

We found 69 putative *bicycle* homologs in *Daktulosphaira vitifoliae* indicating that *bicycle* genes were present in the common ancestor of the Aphidomorpha (Phyloidea+Phylloxeridae). We also found 74, 26 and 2 putative *bicycle* homologs in the coccid species *Macconellicoccus hirsutus*, *Phenacoccus solenopsis*, and *Ericerus pela*, respectively. In *M. hirsutus* and *E. pela* profile hmmer search with the classifier identified genes identified 17 and 12 additional candidate *bicycle* genes, respectively. The putative *bicycle* homologs from these non-aphid species do not contain obvious CYC motifs (supplementary fig. S9A–C, Supplementary Material online). However, gene structure aware alignment with the *H. cornu bicycle* genes revealed that the putative homologs from *D. vitifoliae*, *M. hirsutus*, and *P. solenopsis* share more introns in the same locations with *H. cornu bicycle* genes (supplementary fig. S10D–F, Supplementary Material online) than expected by chance (supplementary fig. S10E, Supplementary Material online), supporting the inference that these are *bicycle* homologs.

We did not detect *bicycle* homologs in any of the species sampled from the Psyllioidea, Aleurodoidae, or Fulgoroidae. These results indicate that *bicycle* genes were present in the common ancestor of coccids and aphids and may have evolved in the common ancestor of these lineages. However, for at least two reasons we cannot rule out the possibility that *bicycle* genes originated earlier. First, we observed extensive variation in the number of *bicycle* homologs across lineages, so *bicycle* genes may have been lost in the specific outgroup species studied here. Second, ancestral *bicycle* genes may not display the specific gene structure detected by our classifier. Deeper taxonomic
sampling and other homology search approaches may reveal an older origin of bicycle genes.

**Aphid Genomes Contain a Conserved Megacycle Gene**

In 11 of the 12 aphid genomes we studied, our classifier identified a single extremely large gene containing more than 100 microexons and encoding a protein of approximately 2000 amino acids as a candidate bicycle homolog (fig. 6A). Since this gene was so large and most existing aphid genomes are incompletely assembled, we observed that many of these gene models were incomplete. We therefore performed de novo transcript assembly for all species studied using Trinity (Grabherr et al., 2011) and manually assembled consensus transcripts of these long candidate bicycle genes.

Since we had previously found that bicycle gene length is correlated with the number of encoded CYC motifs (fig. 3), we explored whether these genes encoded proteins with multiple repeating units. Using Rapid Automatic Detection and Alignment of Repeats (Heger and Holm, 2000) via EMBL’s European Bioinformatics Institute analysis tool (EMBL-EBI) (Madeira et al., 2019), we found that these large proteins contain an N-terminal secretion signal and many repeating motifs and each motif is approximately the size of two CYC motifs. This structure was most obvious for the Melanaphis sacchari gene (fig. 6A). Sequence-based alignment of 15 predicted repeats of the M. sacchari protein, which did not use information of intron positions, revealed multiple shared intron locations across repeats (fig. 6B). Thus, this protein appears to have evolved through multiple internal tandem duplications. This large protein exhibited little obvious sequence similarity to bicycle genes, but gene-structure aware alignment revealed multiple apparently conserved intron positions shared between the internal duplications of the large gene from M. sacchari and H. cornu bicycle genes (supplementary fig. S11A, Supplementary Material online), suggesting that they share a common ancestor.

These large genes appear to be divergent Bicycle proteins that have duplicated the core motif many times. We...
therefore call these megacycle genes. Compared to most bicycle genes, megacycle orthologs are relatively well conserved across aphids and a phylogenetic tree reconstructed from the inferred Megacycle proteins is similar to the aphid phylogeny estimated from the whole proteome (supplementary fig. S11B, Supplementary Material online). Most aphid genomes we examined contain a single copy of this gene; we did not detect this gene in Diuraphis noxia, and we found three paralogs in both C. cedri and T. nigriabdominalis. No megacycle orthologs were found in any of the non-aphid species.

For the three species for which we have good salivary gland transcriptomic data, H. cornu, T. nigriabdominalis, and A. pisum, the megacycle genes are strongly expressed in salivary glands (fig. 6C–E). Thus, megacycle genes may contribute to the effector protein repertoire in aphids. The relatively conserved lengths and sequences of megacycle homologs suggests that they may share a relatively conserved role across aphids.

**Discussion**

Some gene features, such as the structure of the encoded protein or the intron-exon structure of the gene, can provide evidence for homology that is independent of sequence (Bazan, 1991; Brown et al., 1995; Betts, 2001). This source of information is becoming increasingly available with the recent availability of many well-assembled,
annotated genomes. It is possible to conceptualize a model that jointly utilizes both sequence and exon-intron structure in homology search, but we realized that bicycle genes are encoded with extremely unusual intron-exon structures and that a model incorporating only general features of exons may allow detection of distant homologs.

We found that a linear classifier using information on gene length, exon sizes, exon numbers, and exon phases provided a highly accurate predictor of bicycle homologs that could not be identified using sequence similarity searches. This predictor allowed identification of bicycle homologs across aphids and in several outgroup taxa (fig. 1). Homology was supported by post hoc observation of CYC motifs and of an excess of shared introns across many homologs. Thus, sequence independent features provide substantial evidence of homology and integration of even a few sequence independent features into existing sequence-based search methods may substantially increase power to detect homology relationships between genes. All gene structure features that we used contributed to classifier performance, but no single feature was critically important for classifier performance. Therefore, in future work it would be ideal to incorporate multiple features of gene structure into homology search models.

For several reasons, our classifier may have underestimated the number of bicycle homologs in each species. First, this approach is sensitive to the quality of genome annotation, which is itself dependent on genome assembly quality. We found that most automatically annotated gene models of bicycle genes required manual re-annotation. Thus, we may have overlooked additional bicycle genes in these genomes because of genome fragmentation and inaccurate automated gene annotation. Second, bicycle gene annotation in most species is likely hampered by the lack of salivary gland RNA-seq samples. We observed considerable variation in bicycle gene expression levels among samples, and deep sequencing of diverse samples is likely required to provide sufficient RNAseq evidence to build accurate bicycle gene models. Finally, it is possible that some bicycle homologs have evolved both divergent sequences and novel gene structures, preventing identification with any existing method.

Bicycle Genes are Not Strictly Associated with Gall Formation

Bicycle genes were detected originally as putatively secreted proteins strongly expressed in the salivary glands of the gall-forming generation of the aphid H. cornu and in several outgroup taxa (fig. 1). Homology was supported by post hoc observation of CYC motifs and of an excess of shared introns across many homologs. Thus, sequence independent features provide substantial evidence of homology and integration of even a few sequence independent features into existing sequence-based search methods may substantially increase power to detect homology relationships between genes. All gene structure features that we used contributed to classifier performance, but no single feature was critically important for classifier performance. Therefore, in future work it would be ideal to incorporate multiple features of gene structure into homology search models.

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First, we found that many bicycle genes are most strongly expressed in salivary glands of generations of H. cornu that do not induce galls (fig. 2F–G). Second, two of the aphids studied here, H. cornu and T. nigrabdominalis, induce complex galls and their genomes include many more bicycle genes than other species we examined, but the genomes of two other gall-forming species, E. lanigerum and D. vitifolii, have only 27 and 69 bicycle genes, respectively. In addition, the genomes of some non-galling aphids have many bicycle genes, such as Myzus persicae and A. pisum, with 117 and 121, respectively. Thus, across species there does not appear to be a strong correlation between the number of bicycle genes and the gall-forming habit.

Given these new observations, we hypothesize that if Bicycle proteins have conserved molecular functions, then they likely have different targets in different plant species. In some cases, the targets may induce patterned cell proliferation, resulting in galls, and in others the targets may alter plant physiology to confer benefits on aphids.

New Experimental Opportunities Provided by the Presence of Bicycle Genes in Many Species

One of the challenges with studying the function of bicycle genes is that initially they were thought to be restricted to the non-model gall-forming insect H. cornu. H. cornu displays a complex life cycle, exhibiting many polyphenic forms and migration between two trees every year (Korgaonkar et al., 2021), and prospects for laboratory rearing and subsequent experimental manipulations of this species were poor.

The discovery of bicycle genes in other insects opens up the possibility that the function of bicycle genes can be studied in species more amenable to laboratory manipulation. For example, RNAi allows gene knockdown in M. persicae (Pitino et al., 2011; Coleman et al., 2015; Mathers et al., 2017; Niu et al., 2019; Chen et al., 2020; Gao et al., 2021;) and A. pisum (Han et al., 2006; Mutti et al., 2006, 2008; Shakesby et al., 2009; Will and Vilcinskas, 2015; Vellichirammal et al., 2017; Niu et al., 2019) and CRISPR-Cas9 mutagenesis allows gene knockout in A. pisum (Le Trionnaire et al., 2019). Thus, the discovery of
many bicycle genes in these species presents new experimental opportunities to uncover the organismal and molecular functions of bicycle genes.

Materials and Methods

Gene Structure Based Bicycle Gene Classifier

The bicycle gene classifier was built as a generalized linear model (GLM) with the following predictor variables: total gene length, total length of coding exons, first coding exon length, last coding exon length, mean internal exon length, and number of internal exons in phase 0, 1, and 2. All length measurements are in base pairs. All genes with zero or one internal exon were removed from the training and test datasets because mean internal exon lengths and number of internal exons starting in each of the three phases cannot be accurately estimated; hence, our classifier was unable to detect any potential bicycle genes with only one to three exons. The dependent variable from the GLM is the predictor of bicycle gene classification. The GLM was implemented with the glm function as binomial with logit link, and response variables for the whole-genome gene set were predicted using predict, both in the R base package stats v.3.6.1.

In the training set, the dependent variable of the GLM is whether the transcript was previously classified as a bicycle gene (Korgaonkar et al., 2021), where bicycle genes were coded as 1 and non-bicycle genes were coded as 0. We used only genes classified as either bicycle genes or previously annotated genes in the training set due to the possibility that unannotated genes include additional previously undetected bicycle genes. Previously annotated genes were identified as those having at least one significant match at E < 0.01 after Bonferroni correction for multiple searches with blastp (blast v.2.9.0), blastx (blast v.2.9.0), or hmmscan (HMMER v3.2.1) against the Pfam database.

We evaluated the performance of the classifier using a precision-recall curve where precision = TP/(TP + FP) and recall = TP/(TP + FN); where TP, FP, and FN are the number of true positives, false positives, and false negatives, respectively. To determine a cutoff value on the dependent variable of the GLM for bicycle gene classification, we tested all possible cutoffs from 0 to 0.98 in increments of 0.02 to find the cutoff resulting in a precision to recall ratio closest to 1. We first validated our model by repeatedly sampling and training on 70% of the data and testing on the remaining 30% of the data for 100 replicates and found no substantial change to the trained classifier or the resulting cutoff value (mean 0.7338, s.d. 0.1578) amongst replicates. We therefore trained the final classifier on all H. cornu bicycle genes and annotated non-bicycle genes without subsampling and determined a cutoff value of 0.72 (precision = 0.9925, recall = 0.9925). The coefficients to the independent variables in the GLM are as follows: intercept = 6.26, total exon length = −0.0206, total gene length = −0.0000642, first exon length = −0.00632, last exon length = 0.0103, mean exon length = −0.0655, number of exons in phase 0 = −3.32, phase 1 = −1.12, and phase 2 = 1.21. We evaluated the contribution of each independent variable by training the H. cornu dataset using GLMs with each term on its own and each term removed one at a time from the full model. We applied the GLM trained by the H. cornu dataset to twenty-one other species belonging to the families of Aphididae, Phylloxeridae, Coccoidea, Psyllioidea, Aleyroidea, and Fulgoroidea.

Aphid Collections and Dissections

Aphids of T. nigrriabdominalis were collected from galls collected on Ulmus americana on the grounds of Janelia Research Campus (supplementary tables S2 and S3, Supplementary Material online). Species identification was confirmed using both morphological characteristics (Walczak et al., 2019) and by comparing the mitochondrial DNA of the genome, we assembled with the mitochondrial sequences at NCBI available for many species of Tetraneura (Lee and Akimoto, 2015; Hebert et al., 2016). Aphids of A. pisum LSR1 were kindly provided by Greg Davis and grown on broad bean plants in the laboratory. Salivary glands were dissected out of multiple instars of fundatrices and their offspring for T. nigrriabdominalis and of virginoparae for A. pisum and stored in 3 ul of Smart-seq2 lysis buffer (0.2% Triton X-100, 0.1 U/ml RNasin Ribonuclease Inhibitor) at −80°C for later mRNA isolation. Carcasses of both species were ground in PicoPure extraction buffer, and mRNA was prepared using the PicoPure mRNA extraction kit. RNAseq libraries were prepared as described previously (Korgaonkar et al., 2021).

Genome Sequencing, Assembly, and Annotation of Tetraneura nigrriabdominalis

HMV gDNA was prepared from T. nigrriabdominalis aphids isolated from a single gall following the Salting Out Method provided by 10X Genomics (https://support.10xgenomics.com/genome-exome/sample-prep/doc/demonstrated-protocol-salting-out-method-for-dna-extraction-from-cells). 10X linked-read libraries were prepared and Chromium sequencing was performed by HudsonAlpha Institute for Biotechnology. Genomes were assembled with Supernova v.2.1.0 commands run and mkoutput (Weisenfeld et al., 2017).

We used 133,280,000 reads for an estimated raw genome coverage of 56.18X. The genome size of the assembled scaffolds was 344.312 Mb, and the scaffold N50 was 10 Mb. Using BUSCO completeness analysis with gVolante version 1.2.0 and BUSCO_v2A3 (Nishimura
et al., 2017), we found that the genome contains 1035 of 1066 (97.1%) completely detected core arthropod genes and 1043 of 1066 (97.8%) partially detected core genes. The genome was annotated for protein-coding genes using BRAKER (Hoff et al., 2019) with multiple RNA-seq libraries (supplementary table S3, Supplementary Material online). This genome has been deposited at GenBank under the accession JAUCT00000000.

**Differential Expression Analysis**

Candidate bicycle genes from *H. cornu*, *T. nigrabdominalis*, and *A. pism* identified by the classifier were manually re-annotated in Apollo (Dunn et al., 2019) guided by RNAseq data viewed in Integrative Genomics Viewer (IGV) (Thorvaldsdóttir et al., 2013). Samples used for analysis are described in supplementary table S3–S6, Supplementary Material online. Differential expression analyses were performed as described previously (Korgaonkar et al., 2021) using edgeR and BLAST Searches

The phylogeny of all species studied here was estimated using Orthofinder v 2.3.1 (Emms and Kelly, 2018) on the complete gene sets predicted by BRAKER for all genomes using the following settings: -M msa -S diamond -A mafft -T fasttree. The phylogeny was plotted using ggtree v. 2.0.4 in R.

We searched for bicycle homologs in all aphid species and outgroups using Position-Specific Interactive Basic Local Alignment Search Tool (PSI-BLAST) (Altschul, 1997) and hmmmer (Eddy, 2011). We ran PSI-BLAST v. 2.9.0 + using -max_target_seqs 5 and all other parameters as default. We used a cutoff of E < 0.01, Bonferroni corrected for 476 genes used as query searches. We ran profile-based hmmsearch v.3.2.1 using all default parameters and cutoff of E < 0.01 without multiple testing correction.

**Selection Tests**

DnDs between *H. cornu* and *H. hamamelidis* for all transcripts in the transcriptome was calculated using the codeml function from the PAML package v4.9 (Yang, 2007). The genome-wide selection scan was performed using Sweed-P v.3.1 (Pavlidis et al., 2013). Details of both analyses can be found in the Methods section of Korgaonkar et al., 2021.

**Gene-Structure Aware Sequence Alignment**

We performed gene structure aware alignment using prrn5 v 5.2.0 (Gotoh, 2021) with default weighting parameters. We prepared gene-structure-informed extended fasta files that were suitable as input to prrn5 using the ANA2GSi-seq.pl utility provided at https://github.com/ogotoh/prrn_aln/blob/master/perl/anno2gsi-seq.pl. We summarized the results of these multiple sequence alignments by generating histograms with a bin size of 1 to illustrate the fraction of sequences in the alignment that had an intron at each position in the alignment.

To determine whether intron positions were shared between genes more often than expected by chance, we computed the concordance in intron positions resulting from prrn5 alignment of groups of unrelated genes. We identified paralog groups for the whole *H. cornu* transcriptome using Orthofinder v 2.3.3 (Emms and Kelly, 2018) and identified three paralog groups containing at least 10 genes and having at least 10 exons for comparison with bicycle genes: SLC33A1, abcG23, and nrf-6. Intron concordance was measured as the correlation coefficient $\rho$ between the number of genes from the test gene family and the bicycle gene family containing an intron at each location in the alignment (supplementary fig. S16, Supplementary Material online).

**Re-Annotation of Previously Sequenced Genomes**

The *P. venusta* genome is an approximately full-chromosome genome assembly that was kindly provided to us by Yiyuan Li and Nancy Moran (Li et al., 2020) prior to publication. All other genomes, except for *H. cornu* and *T. nigrabdominalis*, were downloaded from NCBI (supplementary table S5, Supplementary Material online). We downloaded RNAseq data from NCBI (supplementary table S4, Supplementary Material online) and predicted coding genes using BRAKER (Hoff et al., 2019) with the same workflow we had used previously that had allowed discovery of bicycle genes (Korgaonkar et al., 2021).

**Supplementary Material**

Supplementary data are available at Genome Biology and Evolution online.

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HHMI’s Open Access to Publications policy. HHMI lab heads have previously granted a nonexclusive CC BY 4.0 license to the public and a sublicensable license to HHMI in their research articles. Pursuant to those licenses, the author-accepted manuscript of this article can be made freely available under a CC BY 4.0 license immediately upon publication.

Author Contributions

D.L.S. conceptualized the project, collected aphids, performed dissections, made DNA and most RNA samples, performed initial analyses, and wrote the paper. C.H. developed the bicycle gene classifier, independently verified all of the computational analyses, and assisted with preparations of figures and manuscript.

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Ethics Declarations

HHMI has filed a provisional patent, number 63/243,904, for the inventors D.L.S. and C.H. covering a gene structure, coding sequence-independent method of identifying bicycle genes.

Data Availability

All new sequence data generated during this study is available at NCBI at the accession numbers provided in supplementary tables S2–S4, Supplementary Material online. All analysis scripts, which allow reproduction of all analyses and reproduction of all figures, are available on Figshare with DOIs provided in supplementary table S7, Supplementary Material online. Whole transcriptome annotation files, classifier-identified bicycle genes lists, and multi-sequence alignment (MSA) files for all analysis where Logo plots were generated are also available on Figshare at DOIs provided in supplementary table S7, Supplementary Material online. The code for training and applying the gene-structure classifier has been instantiated as Google Colabs freely available to the public at the following URLs: https://colab.research.google.com/drive/1GiSkUDC6US1HlW1flk3WcJUPgKI-RY8, https://colab.research.google.com/drive/1oJAHehrX4ytqT9OZtLKPVNXQAljixcx.

Literature Cited

Aguiar ERGR, et al. 2015. Sequence-independent characterization of viruses based on the pattern of viral small RNAs produced by the host. Nucleic Acids Res. 43:6191–6206.

Altschul S. 1997. Gapped BLAST and PSI-BLAST: a new generation of protein database search programs. Nucleic Acids Res. 25:3389–3402.

Altschul SF, Gish W, Miller W, Myers EW, Lipman DJ. 1990. Basic local alignment search tool. J Mol Biol. 215:403–410.

Alvarez RA, González-Sierra S, Candelas A, Martinez J-JI. 2013. Histological study of galls induced by aphids on leaves of Ulmus minor: Tetraneura ulmi induces globose galls and Eriosoma ulmi induces pseudogalls. Arthropod Plant Interact. 7:643–650.

Bazan JF. 1991. Neuroepoietic cytokines in the hematopoietic fold. Neuron 7:197–208.

Beardsley JW, Gonzalez RH. 1975. The biology and ecology of armored scales. Annu Rev Entomol. 20:47–73.

Betts MJ. 2001. Exon structure conservation despite low sequence similarity: a relic of dramatic events in evolution? EMBO J 20:5354–5360.

Brown NP, Whittaker AJ, Newell WR, Rawlings CJ, Beck S. 1995. Identification and analysis of multigene families by comparison of exon fingerprints. J Mol Biol. 249:342–359.

Burstein M, Wool D, Eshel A. 1994. Sink strength and clone size of sympatric, gall-forming aphids. Eur J Entomol. 91:57–61.

Chen Y, et al. 2020. An aphid RNA transcript migrates systemically within plants and is a virulence factor. Proc Natl Acad Sci USA 117:12763–12771.

Coleman AD, Wouters RHM, Mugford ST, Hogenhout SA. 2015. Persistence and transgenerational effect of plant-mediated RNAi in aphids. J Exp Botany. 66:541–548.

Cook LG, Gullan PJ. 2004. The gall-inducing habit has evolved multiple times among the eriococcid scale insects (Stenorrhyncha: Coccioidea: Eriococcidae). Biol J Linn Soc. 83:441–452.

Dixon AFG. 1978. Biology of aphids, studies in biology. London (UK): Edward Arnold.

Dommel M, et al. 2020. Big genes, small effectors: pea aphid cassette effector families composed from mini- exons. Front Plant Sci. 11, 1230.

Dunn NA, et al. 2019. Apollo: democratizing genome annotation. PLoS Comput Biol. 15:e1006790.

Eddy SR. 2011. Accelerated profile HMM searches. PLoS Comput Biol. 7:e1002195.

Eizaguirre C, Lenz TL, Kalbe M, Milinski M. 2012. Rapid and adaptive evolution of MHC genes under parasite selection in experimental vertebrate populations. Nat Commun. 3:1–6.

Elzinga DA, Jander G. 2013. The role of protein effectors in plant-aphid interactions. Curr Opin Plant Biol. 16:451–456.

Emms DM. 2019. OrthoFinder: phylogenetic orthology inference for comparative genomics 14.

Emms DM, Kelly S. 2018. STAG: species tree inference from all genes. bioRxiv 1–29. https://doi.org/10.1101/267914.

Gao Y, et al. 2021. The gustavus gene can regulate the fecundity of the green peach aphid, Myzus persicae (Sulzer). Front Physiol. 12:596392.

Giron D, Huguet E, Stone GN, Body MJA. 2016. Insect-induced effects on plants and possible effectors used by gall-making and leaf-mining insects to manipulate their host-plant. J Insect Physiol. 84:70–89.

Gottoh O. 2021. Cooperation of Spaln and Prrn5 for construction of gene-structure-aware multiple sequence alignment. In: Katoh K, editors. Multiple sequence alignment, methods in molecular biology. New York (NY): Springer US. p. 71–88.

Grabherr MG, et al. 2011. Full-length transcriptome assembly from RNA-Seq data without a reference genome. Nat Biotechnol. 29:644–652.

Han J, et al. 2006. Molecular basis for the recognition of primary microRNAs by the Drosha-DGCR8 complex. Cell 125:897–901.

Hebert PDN, et al. 2016. Counting animal species with DNA barcodes: Canadian insects. Phil Trans R Soc B. 371:20150333.
Heger A, Holm L. 2000. Rapid automatic detection and alignment of repeats in protein sequences. Proteins Struct Funct Bioinform. 41:224–237.

Hoff KJ, Lomsadze A, Borodovsky M, Stanke M. 2019. Whole-genome annotation with broker. Method Mol Biol. 1962:65–95.

Hogenhout SA, Bos JB. 2011. Effector proteins that modulate plant-insect interactions. Curr Opin Plant Biol. 14:422–428.

Johnson KP, et al. 2018. Phylogenomics and the evolution of hemipteroid insects. Proc Natl Acad Sci USA 201815820.

Johnson LS, Eddy SR, Portugal E. 2010. Hidden Markov model speed heuristic and iterative HMM search procedure. BMC Bioinform. 11.

Korgaonkar A, et al. 2021. A novel family of secreted insect proteins linked to plant gall development. Curr Biol. 31:1836–1849.

Lee W, Akimoto S-I. 2015. Development of new barcoding loci in gall-forming aphids (Eriosomatinae: Eriosomatini): comparing three mitochondrial genes, ATP6, ATP8, and COI. J Asia Pac Entomol. 18:267–275.

Le Trionnaire G, et al. 2019. An integrated protocol for targeted mutagenesis with CRISPR-Cas9 system in the pea aphid. Insect Biochem Mol Biol. 110:34–44.

Li Y, Park H, Smith TE, Moran NA, Singh N. 2019. Gene family evolution in the pea aphid based on chromosome-level genome assembly. Mol Biol Evol. 36:2143–2156.

Li Y, Zhang B, Moran NA. 2020. The aphid X chromosome is a dangerous place for functionally important genes: diversity evolution of hemipteran genomes based on chromosome-level assemblies. Mol Biol Evol. 37:2357–2368.

Loewenstein Y, et al. 2009. Protein function annotation by homology-based inference. Genome Biol. 10:207.

Madeira F, et al. 2019. The EMBL-EBI search and sequence analysis tools APIs in 2019. Nucleic Acids Res. 47:W636–W641.

Mani MS. 1964. Ecology of plant gallers. The Hague, The Netherlands: Springer-Science + Business Media, B.V.

Mathers TC, et al. 2017. Rapid transcriptional plasticity of duplicated gene clusters enables a clonally reproducing aphid to colonise diverse plant species. Genome Biol. 18:1–20.

Mutti NS, et al. 2008. A protein from the salivary glands of the pea aphid, *Acyrthosiphon pism*., is essential in feeding on a host plant. Proc Natl Acad Sci USA 105:9965–9969.

Mutti NS, Park Y, Reek GR. 2006. RNAi knockdown of a salivary transcript leading to lethality in the pea aphid, *Acyrthosiphon pism*. J Insect Sci. 6, 1–7.

Nishimura O, Hara Y, Kuraku S. 2017. GVolante for standardizing genomes. Bioinform. 33:3635–3637.

Ni J, et al. 2019. Topical dsRNA delivery induces gene silencing and mortality in the pea aphid. Pest Manag Sci. 75:2873–2881.

Nováková E, et al. 2013. Reconstructing the phylogeny of aphids (Hemiptera: Aphiidae) using DNA of the obligate symbiont *Buchnera aphidica*. Mol Phylogenetic Evol. 68:42–54.

Parr T. 1940. Asterolecanium variolosum ratzeburg, a gall-forming coccid, and its effect upon the host trees. Yale School of the Environment Bulletin Series, No. 46: 90 pp.

Paton S, et al. 2010. Antagonistic coevolution accelerates molecular evolution. Nature 464:275–278.

Pavlídis P, Zivkovic D, Stamatakis A, Alachiotis N. 2013. SweeD: likelihood-based detection of selective sweeps in thousands of genomes. Mol Biol Evol. 30:2224–2234.

Pitino M, Coleman AD, Maffei ME, Ridout CJ, Hogenhout SA. 2011. Silencing of aphid genes by dsRNA feeding from plants. PLOS ONE 6:e25709.

Richards S, et al. 2010. Genome sequence of the pea aphid *Acyrthosiphon pism*. PloS Biol. 8:e1000313.

Rodriguez PA, Escudero-Martinez C, Bos JB. 2017. An aphid effector targets trafficking protein VP552 in a host-specific manner to promote virulence. Plant Physiol. 173:1892–1903.

Rokas A, Holland PWH. 2000. Rare genomic changes as a tool for phylogenetics. Trends Ecol Evol. 15:454–459.

Rokas A, Kathirithamby J, Holland PWH. 1999. Intron insertion as a phylogenetic character: the engrafted homeobox of Strepsiptera does not indicate affinity with Diptera. Insect Mol Biol. 8:527–530.

Roy SW, Gilbert W. 2005. Rates of intron loss and gain: implications for early eukaryotic evolution. Proc Natl Acad Sci USA 102:5773–5778.

Ruvinsky A, Watson C. 2007. Intron phase patterns in genes: preservation and evolutionary changes. TOEVL1 1:1–14.

Schable G, et al. 2008. Economic impacts of the U. S. soybean aphid infestation: a multi-regional competitive dynamic analysis. Agricult Res Econ Rev. 37:227–242.

Shakesby AJ, et al. 2009. A water-specific aquaporin involved in aphid osmoregulation. Insect Biochem Mol Biol. 39:1–10.

Shorthouse JD, Wool D, Raman A. 2005. Gall-inducing insects - nature’s most sophisticated herbivores. Basic Appl Ecol. 6:407–411.

Tekiaa F. 2016. Inferring orthologs: open questions and perspectives. Genomics Insights 9:GEI.S37925.

Telford MJ, Copley RR. 2011. Improving animal phylogenies with genomic data. Trends Genetic. 27:186–195.

Thorvaldsdottir H, Robinson JT, Mesirov JP. 2013. Integrative genomics exploration. Brief Bioinform. 14:178–192.

Vakiris N, Carvunis A-R, McLyshaght A. 2020. Synteny-based analyses indicate that sequence divergence is not the main source of orphan genes. elife 9:e53500.

Vellichirammal NN, Gupta P, Hall TA, Brisson JA. 2017. Ecdyson signaling underlies the pea aphid transgenerational wing polyphenism. Proc Natl Acad Sci USA 114:1419–1423.

Venkatesh B, Ning Y, Brenner S. 1999. Late changes in spliceosomal introns define clades in vertebrate evolution. Proc Natl Acad Sci USA 96:10267–10271.

von Dohlen NA, Moran CD. 1995. Molecular phylogeny of the homoptera: a paraphyletic taxon. J Mol Evol. 41:211–223.

Walczak U, Borowiak-Sobkowiak B, Wilkaniec B. 2019. Tetraneura (Tetraneurella) nigriabdominalis (Hemiptera: Aphidoidea) – a species extending its range in Europe, and morphological comparison with Tetraneura (Tetraneura) ulmi. Entomol Fenn. 28:21–26.

Weisenfeld NI, Kumar V, Shah P, Church DM, Jaffe DB. 2017. Direct determination of diploid genome sequences. Genome Res. 27:757–767.

Weisman CM, Murray AW, Eddy SR. 2020. Many, but not all, lineage-specific genes can be explained by homology detection failure. PloS Biol. 18:e3000862.

Will T, Tjallingii WF, Thönnessen A, van Bel AJE. 2007. Molecular sabotage of plant defense by aphid saliva. Proc Natl Acad Sci USA 104:10536–10541.

Will T, Vilcinskas A. 2015. The structural sheath protein of aphids is required for phloem feeding. Insect Biochem Mol Biol. 57:34–40.

Xie W, He C, Fei Z, Zhang Y. 2020. Chromosome-level genome assembly of the greenhouse whitefly (Trialeurodes vaporariorum Westwood). Mol Ecol Res. 20:995–1006.

Yang Z. 2007. PAML 4: phylogenetic analysis by maximum likelihood. Mol Biol Evol. 24:1586–1591.