A novel family of secreted insect proteins linked to plant gall development

Highlights

- Novel aphid bicycle genes contribute to plant gall development
- Variation in a bicycle gene alters plant gene expression and a gall phenotype
- bicycle genes encode a large family of diverse, secreted, cysteine-rich proteins
- Many bicycle genes have experienced repeated diversifying selection

Authors

Aishwarya Korgaonkar, Clair Han, Andrew L. Lemire, ..., Peter Andolfatto, Shuji Shigenobu, David L. Stern

Correspondence

sternd@hhmi.org

In brief

Korgaonkar et al. report on novel secreted aphid proteins encoded by bicycle genes. Variation in the bicycle gene determinant of gall color alters expression of targeted plant genes, suggesting that BICYCLE proteins modulate gall development.

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A novel family of secreted insect proteins linked to plant gall development

Aishwarya Korgaonkar,1 Clair Han,1 Andrew L. Lemire,1 Igor Siwanowicz,1 Djawed Bennouna,2 Rachel E. Kopec,2,3 Peter Andolfatto,4 Shuji Shigenobu,5,6,7 and David L. Stern1,8

1Janelia Research Campus of the Howard Hughes Medical Institute, 19700 Helix Drive, Ashburn, VA 20147, USA
2Human Nutrition Program, Department of Human Sciences, The Ohio State University, 262G Campbell Hall, 1787 Neil Avenue, Columbus, OH 43210, USA
3Ohio State University’s Foods for Health Discovery Theme, The Ohio State University, 262G Campbell Hall, 1787 Neil Avenue, Columbus, OH 43210, USA
4Department of Biology, Columbia University, 600 Fairchild Center, New York, NY 10027, USA
5Laboratory of Evolutionary Genomics, Center for the Development of New Model Organism, National Institute for Basic Biology, Okazaki 444-8585, Japan
6NIBB Research Core Facilities, National Institute for Basic Biology, Okazaki 444-8585, Japan
7Department of Basic Biology, School of Life Science, SOKENDAI (The Graduate University for Advanced Studies), 38 Nishigonaka, Myodaiji, Okazaki 444-8585, Japan
8Lead contact
*Correspondence: sternd@hhmi.org
https://doi.org/10.1016/j.cub.2021.01.104

SUMMARY

In an elaborate form of inter-species exploitation, many insects hijack plant development to induce novel plant organs called galls that provide the insect with a source of nutrition and a temporary home. Galls result from dramatic reprogramming of plant cell biology driven by insect molecules, but the roles of specific insect molecules in gall development have not yet been determined. Here, we study the aphid Hormaphis cornu, which makes distinctive “cone” galls on leaves of witch hazel Hamamelis virginiana. We found that derived genetic variants in the aphid gene determinant of gall color (dgc) are associated with strong downregulation of dgc transcription in aphid salivary glands, upregulation in galls of seven genes involved in anthocyanin synthesis, and deposition of two red anthocyanins in galls. We hypothesize that aphids inject DGC protein into galls and that this results in differential expression of a small number of plant genes. dgc is a member of a large, diverse family of novel predicted secreted proteins characterized by a pair of widely spaced cysteine-tyrosine-cysteine (CYC) residues, which we named BICYCLE proteins. bicycle genes are most strongly expressed in the salivary glands specifically of galling aphid generations, suggesting that they may regulate many aspects of gall development. bicycle genes have experienced unusually frequent diversifying selection, consistent with their potential role controlling gall development in a molecular arms race between aphids and their host plants.

INTRODUCTION

Organisms often exploit individuals of other species, for example, through predation or parasitism. Parasites sometimes utilize molecular weapons against hosts, which themselves respond with molecular defenses, and the genes that encode or synthesize these molecular weapons may evolve rapidly in a continuous “arms race.” Some of the most elaborate molecular defenses—such as adaptive immune systems, restriction modification systems, and CRISPR—have resulted from such host-parasite conflicts. In many less-well-studied systems, parasites not only extract nutrients from their hosts, but they also alter host behavior, physiology, or development to the parasite’s advantage. Insect galls represent one of the most extreme forms of such inter-species manipulation.

Insect-induced galls are intricately patterned homes that provide insects with protection from environmental vicissitude and from some predators and parasites. Galls are also resource sinks, drawing nutrients from distant plant organs and providing insects with abundant food. Insect galls are atypical plant growths that do not result simply from unpatterned cellular over-proliferation, as observed for microbial galls, like the crown gall induced by Agrobacterium tumefaciens. Instead, each galling insect species appears to induce a distinctive gall, even when related insect species attack the same plant, implying that each species provides unique instructions to reprogram latent plant developmental networks.

At least some gall-inducing insects produce phytohormones, although it is not yet clear whether insects introduce these hormones into plants to support gall development. However, injection of phytohormones alone probably cannot generate the large diversity of species-specific insect galls. In addition, galling insects induce plant transcriptional changes independently of phytohormone activity. Thus, given the
complex cellular changes required for gall development, insects probably introduce many molecules into plant tissue to induce galls.

In addition to the potential role of phytohormones in promoting gall growth, candidate gall effectors have been identified in several gall-forming insects. However, none of these candidate effectors have yet been shown to contribute to gall development or physiology. In addition, although many herbivorous insects introduce effector molecules into plants to influence plant physiology, there is no evidence that any previously described effectors contribute to gall development. Because there are currently no galling insect model systems that would facilitate a genetic approach to this problem, we turned to natural variation to identify insect genes that contribute to gall development.

We studied the aphid, Hormaphis cornu, which induces galls on the leaves of witch hazel, Hamamelis virginiana, in the Eastern United States (Figures 1A–1F and 1J). In early spring, each H. cornu gall foundress (fundatrix) probes an expanding leaf with her microscopic mouthparts (stylets; Figures 1A and 1B; Video S1) and pierces individual mesophyll cells with her stylets (Figures 1G and 1H). We found that plant cells near injection sites, revealed by the persistent stylet sheaths, over-proliferate through periclinal cell divisions (Figure 1H). This pattern of cytoteliosis is not otherwise observed in leaves at this stage of development (Figure 1I) and contributes to the thickening and expansion of leaf tissue that generates the gall (Figures 1D–1G). The increased proliferation of cells near the tips of stylet sheaths suggests that secreted effector molecules produced in the salivary glands are deposited into the plant via the stylets.

After several days, the basal side of the gall encloses the fundatrix and the gall continues to grow apically and laterally, providing the fundatrix and her offspring with protection and abundant food. After several weeks, the basal side of the gall opens to allow aphids to remove excreta (honeydew) and molted nymphal skins from the gall and, eventually, to allow winged migrants to depart. Continued gall growth requires the constant ingestion of leaf tissue that generates the gall (Figures 1D–1G). The formation of cells near the tips of stylet sheaths suggests that secreted effector molecules produced in the salivary glands are deposited into the plant via the stylets.

RESULTS

A natural gall color polymorphism is linked to regulatory variation in a novel aphid gene, determinant of gall color

We found that populations of H. cornu galls include approximately 4% red galls and 96% green galls (Figure 1F). We inferred that this gall color polymorphism results from differences among aphids, rather than from differences associated with leaves or the location of galls on leaves, because red and green galls are located randomly on leaves and often adjacent to each other on a single leaf (Figure 1F). We sequenced and annotated the genome of H. cornu (Figures S1A and S1B; STAR methods) and performed a genome-wide association study (GWAS) on fundatrices isolated from 43 green galls and 47 red galls by sequencing their genomes to approximately 3x coverage. There is no evidence for genome-wide differentiation of samples from red and green galls, suggesting that individuals making red and green galls were sampled from a single interbreeding population (Figures S1C–S1F). We identified SNPs near 40.5 Mbp on chromosome 1 that were strongly associated with gall color (Figure 2A). We resequenced approximately 800 kbp flanking these SNPs to approximately 60x coverage and identified 11 SNPs within the introns and upstream of gene g16073 that were strongly associated with gall color (Figures 2B–2D). There is no evidence that large-scale chromosomal aberrations are associated with gall color (Figures S1G–S1K; STAR methods).

Because GWASs can sometimes produce spurious associations, we performed an independent replication study and found that all 11 SNPs were highly significantly associated with gall color in fundatrices isolated from 435 green and 431 red galls (LOD = 191–236; Figure 2E). All fundatrices from green galls were homozygous for the ancestral allele at 9 or more of these SNPs (Figure 2E). In contrast, 98% of fundatrices from red galls were heterozygous or homozygous for derived alleles at 9 or more SNPs (Figure 2E). This pattern suggests that alleles contributing to red gall color are genetically dominant to alleles that generate green galls. Two percent of fundatrices that induce red galls were homozygous for ancestral alleles at these SNPs and likely carry genetic variants elsewhere in the genome that confer red color to galls (Figure S1L; STAR methods).

Based on these genetic associations and further evidence presented below, we assigned the name determinant of gall color (dgc) to g16073. dgc encodes a predicted protein of 23 kDa with an N-terminal secretion signal sequence (Figure S1M). The putatively secreted portion of the protein shares no detectable sequence homology with any previously reported proteins.

Most SNPs associated with green or red galls were found in one of two predominant haplotypes (Figure 2E) and exhibited strong linkage disequilibrium (LD) (Figures S2A–S2D). LD can result from suppressed recombination. However, these 11 SNPs are in linkage equilibrium with many other intervening and flanking SNPs (Figure S2). Also, multiple observed genotypes are consistent with recombination between these 11 SNPs (Figure 2E), and we found no evidence for chromosomal aberrations that could suppress recombination (Figures S1G–S1K; STAR methods). Thus, LD among the 11 dgc SNPs associated with gall color cannot be explained by suppressed recombination. It is more likely that the non-random association of the 11 dgc alleles has been maintained by natural selection, suggesting that the combined action of all 11 SNPs may have a stronger effect on gene function than any single SNP alone.

Regulatory variants at dgc dominantly silence dgc expression

Because all 11 dgc polymorphisms associated with gall color occur outside of dgc exons (Figure 2D), we tested whether these polymorphisms influence expression of dgc or of any other genes in the genome. We first determined that dgc is expressed highly and specifically in fundatrix salivary glands and lowly or not at all in other tissues or other life cycle stages (Figure 3A). We then performed RNA sequencing (RNA-seq) on salivary glands from fundatrices with dgcGreen/dgcGreen or dgcRed/dgcGreen genotypes. dgc stands out as the most strongly differentially expressed gene between these genotypes (Figure 3B). Because dgcRed alleles appeared to be dominant to the dgcGreen
alleles for gall color, we expected that \textit{dgc} transcripts would be upregulated in animals with \textit{dgc}^{Red} alleles. In contrast, \textit{dgc} transcripts were almost absent in fundatrices carrying \textit{dgc}^{Red} alleles.

\textbf{Figure 1. \textit{Hormaphis cornu} aphids drive patterned over-proliferation of plant cells to produce galls on leaves of \textit{Hamamelis virginiana}}

\textbf{(A–C)} Photographs of the abaxial surface of \textit{H. virginiana} leaves being attacked by first-instar fundatrices of \textit{H. cornu}. Nymphs gather on the unopened leaf buds (A), and soon after bud break, the fundatrices gather near leaf veins (B) and inject material that begins to transform the leaf into a gall (C). Magnified region in blue rectangle of (A) shows three fundatrices waiting on unopened bud (A').

\textbf{(D–F)} Photographs of the adaxial surface of leaves of \textit{H. virginiana}, showing galls at early (D), middle (E), and late (F) growth stages.

\textbf{(G–I)} Confocal micrographs of sections through a \textit{H. cornu} gall (G and H) and ungaled \textit{H. virginiana} leaf (I) stained with Congo red and calcofluor white. Extensive hypertrophy is observed in the mesophyll (yellow arrows) at a considerable distance from the location of the aphid’s stylets (blue arrows; G). The tips of aphid stylets (pink arrow) can be observed within cells of \textit{H. virginiana}, and plant tissue shows evidence of hyperproliferation and periclinal divisions (gray arrows) close to the tips of stylets and the termini of stylet sheaths (H). Periclinal divisions are observed in both spongy and palisade mesophyll cells during gall development but never in ungaled leaf tissue (I). Plant tissue is presented in the aphid’s frame of reference, with abaxial leaf surface up.

\textbf{(J)} Diagram of life cycle of \textit{H. cornu}. \textit{H. cornu} migrates annually between \textit{H. virginiana} (blue line) and \textit{Betula nigra} (pink line), and the gall is produced only in the spring on \textit{H. virginiana} (brown line). Each nymph of the first generation (G1, the fundatrix) hatches from an overwintering egg and initiates development of a single gall. Her offspring (G2) feed and grow within the gall and develop with wings, which allows them to fly to \textit{B. nigra} in late spring. For three subsequent generations (G3–G5), the aphids develop as small, coccidiform morphs on \textit{B. nigra}. In the fall, aphids develop with wings (G6), fly to \textit{H. virginiana} plants, and deposit male and female sexuals (G7), the only generation possessing meiotic cells. The sexuals feed and complete development on the senescing leaves of \textit{H. virginiana}. As adults, they mate and the females deposit eggs that overwinter and give rise to fundatrices the following spring.

See also Video S1.

\textbf{(Figure 3C). That is, red galls are associated with strongly reduced \textit{dgc} expression in fundatric salivary glands.} \textit{dgc} expression is reduced approximately 20-fold in fundatric salivary glands with \textit{dgc}^{Red}/\textit{dgc}^{Green} (27 ± 22.6 cpm; mean ± SD) versus \textit{dgc}^{Green}/\textit{dgc}^{Green} genotypes (536 ± 352.3 cpm; mean ± SD). This result suggested that \textit{dgc}^{Red} alleles downregulate both the \textit{dgc}^{Red} and \textit{dgc}^{Green} alleles in heterozygotes. To confirm whether the \textit{dgc}^{Red} allele downregulates the \textit{dgc}^{Green} allele in trans, we identified exonic SNPs that were specific to each allele and could be

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identified in the RNA-seq data. We found that both \textit{dgc}^{\text{Red}} \text{ and } \textit{dgc}^{\text{Green}} alleles were strongly downregulated in heterozygotes, confirming the trans activity of the \textit{dgc}^{\text{Red}} allele (Figure S3A). We observed no systematic transcriptional changes in neighboring genes (Figure 3C), most of which are \textit{dgc} paralogs, indicating that \textit{dgc}^{\text{Red}} alleles exhibit a perhaps unique example of locus-specific repressive transvection.\textsuperscript{10}

High levels of \textit{dgc} transcription are associated with downregulation specifically of plant anthocyanin genes and two anthocyanins

Because red galls are associated with strong differential expression of only \textit{dgc}, we wondered how the plant responds to changes in this single putative effector. To examine this question, we sequenced and annotated the genome of the host plant \textit{Hamamelis virginiana} and then performed whole-genome differential expression on plant mRNA isolated from galls induced by aphids with \textit{dgc}^{\text{Red}}/\textit{dgc}^{\text{Green}} versus \textit{dgc}^{\text{Green}}/\textit{dgc}^{\text{Green}} genotypes (STAR methods). We did not observe genome-wide differentiation between red and green galls (Figures S3B and S3C), only eight plant genes were differentially expressed between red and green galls, and all eight genes were downregulated in green galls (Figures 4A and 4B). That is, high levels of \textit{dgc} are associated with downregulation of only eight plant genes in galls.

Red pigmented galls could result from production of carotenoids,\textsuperscript{41} anthocyanins,\textsuperscript{42,43} or betacyanins. However, in red galls induced by \textit{H. cornu}, the seven most strongly upregulated plant
The pigments present in red galls are products of enzymes in the anthocyanin biosynthetic pathway (Figure 4C). One gene encodes an enzyme (ACCA) that irreversibly converts acetyl-coenzyme A (CoA) to malonyl CoA, a biosynthetic precursor of multiple anthocyanins. Two genes encode anthocyanidin 3-0-glucosyltransferases (UGT75C1 and UFGT), which glycosylate unstable anthocyanins into more stable forms. Two genes encode anthocyanidin 3-0-glucosyltransferases. The three pairs of glucosyltransferases, which methylate anthocyanin derivatives. Finally, two genes encode phi class glutathione S-transferases (GSTF11 and GSTF12), which conjugate glutathione to anthocyanins, facilitating their transport and stable accumulation in vacuoles.

Six of the enzymes upregulated in red galls are required for final steps of anthocyanin production and deposition (Figure 4C), and their upregulation in red galls may account for the accumulation of pigments in red galls. To test this hypothesis, we extracted and analyzed pigments from galls (Figure 4D) and identified high levels of two pigments only in red galls (Figure 4E), the anthocyanins malvidin-3,5-diglucoside and peonidin-3,5-diglucoside (Figures 4F and S3F–S3J). Thus, the pigments present in red galls are products of enzymes in the anthocyanin biosynthetic pathway, such as those encoded by genes that are upregulated in red galls. The two abundant anthocyanins are produced from distinct intermediate precursor molecules (Figure 4C), three of which were also detected in red galls (Figures 4F, S3F, S3G, and S3J), and synthesis of these two anthocyanins likely requires activity of different methyltransferases and glucosyltransferases. The three pairs of glucosyltransferases, methyltransferases, and glutathione transferases upregulated in red galls may provide the specific activities required for production of these two anthocyanins.

Taken together, these observations suggest that dgc represses transcription of seven anthocyanin biosynthetic enzymes. It is not clear how dgc induces specific transcriptional changes in seven plant genes; it may act by altering activity of an upstream regulator of these plant genes.

**Aphids induce widespread transcriptomic changes in galls**

Gall color represents only one aspect of the gall phenotype, apparently mediated by changes in expression of seven plant genes, and the full complement of cell biological events during gall development presumably requires changes in many more plant genes. To estimate how many plant genes are differentially expressed during development of the H. cornu gall on H. virginiana, we performed differential expression analysis of H. cornu fundatrix gall transcripts across all paralogs of the paralogous genes are shown at the top, and expression levels normalized by mean expression across all paralogs are shown below. Only dgc is strongly downregulated in this gene cluster between individuals carrying dgcRed/dgcGreen (red) versus dgcGreen/dgcGreen (green) genotypes. See also Figures S1 and S3.

**Figure 3. dgc is the most strongly differentially expressed gene in salivary glands of fundatrices collected from red versus green galls.**

(A) Expression of dgc in salivary glands, whole bodies, or carcasses (body minus salivary glands) throughout the H. cornu life cycle. Salivary glands were dissected from multiple nymphal stages and adults of four generations, representing major morphs of the life cycle, G1 (fundatrices), G2, G3, and sexuals. dgc is expressed at highest levels in salivary glands of fundatrices. dgc expression declines in salivary glands through the instars of G2 animals and later generations and was not detected in salivary glands of sexuals. Expression observed in full bodies of G1 animals (fundatrices) probably reflects expression in the salivary glands, and expression was not observed in carcasses.

(B) Genome-wide differential expression analysis of H. cornu fundatrix salivary gland transcripts from individuals heterozygous for dgcRed/dgcGreen (red; n = 15) versus homozygous for dgcGreen (green; n = 7), illustrated as a volcano plot. dgc is strongly downregulated in dgcRed/dgcGreen fundatrices.

(C) Boxplots of salivary gland expression levels of genes in the paralog cluster that includes dgc. Gene models of the paralogous genes are shown at the top, and expression levels normalized by mean expression across all paralogs are shown below. Only dgc is strongly downregulated in this gene cluster between individuals carrying dgcRed/dgcGreen (red) versus dgcGreen/dgcGreen (green) genotypes. See also Figures S1 and S3.
Thus, approximately 15,000 plant genes are differentially expressed in galls, representing a system-wide reprogramming of plant cell biology. If other aphid effector molecules act in ways similar to \textit{dgc}, which is associated with differential expression of only eight plant genes, then gall development may require injection of hundreds or thousands of effector molecules.

Figure 4. Genome-wide differential expression analysis of \textit{H. virginiana} red versus green galls and galls versus leaves

(A) Genome-wide differential expression analysis of \textit{H. virginiana} transcripts isolated from galls made by fundatrices heterozygous for \textit{dgc\textsuperscript{Red}/dgc\textsuperscript{Green}} (red; \textit{n} = 17) versus homozygous for \textit{dgc\textsuperscript{Green}} (green; \textit{n} = 23), illustrated as a volcano plot. Only eight genes are differentially expressed at FDR < 0.05, and all are overexpressed in red galls. The seven most strongly differentially expressed genes encode anthocyanin biosynthetic enzymes (FAOMT-1 = g23591; FAOMT-2 = g7147; GSTF11 = g134919; GSTF12 = g109682; UGT75C1 = g14194; UFGT = g22774; ACCA = g97071).

(B) Boxplots of expression levels, in counts per million reads, of the seven anthocyanin biosynthetic genes overexpressed in red galls, in green (green) and red (red) galls and ungalled leaves (blue). Each data point within each gene is from a separate genome-wide RNA-seq sample.

(C) Simplified diagram of the anthocyanin biosynthetic pathway. Enzyme classes upregulated in red galls are shown in purple font. The two terminal anthocyanins that generate the red color in galls, peonidin-3,5-diglucoside and malvidin-3,5-diglucoside, are shown in bold font, and three precursor molecules found in red galls are shown in bold italic font. Anthocyanin names are abbreviated (dg = 3,5-diglucoside).

(D) Photos of cross-sections of green (top left) and red (top right) galls and the pigments extracted from green and red galls (below).

(E) UHPLC-DAD chromatograms at 520 nm of extract from red (red line) and green (green line) galls.

(F) Overlaid UHPLC-mass spectrometry (MS) chromatograms of green (top) and red (middle) gall extracts and authentic standards (bottom). Each pigment is indicated with a different color: green, delphinidin-3,5-diglucoside (\textit{m/z} = 627.1551); black, cyanidin-3,5-diglucoside (\textit{m/z} = 611.1602); blue, petunidin-3,5-diglucoside (\textit{m/z} = 641.1709); purple, peonidin-3,5-diglucoside (\textit{m/z} = 625.1768); and red, malvidin-3,5-diglucoside (\textit{m/z} = 655.1870). Anthocyanin names are abbreviated (dg = 3,5-diglucoside). Peonidin-3,5-diglucoside and malvidin-3,5-diglucoside together account for 87% of pigment detected in red galls.

(G) Genome-wide differential expression analysis of \textit{H. virginiana} transcripts isolated from galls (\textit{n} = 36) versus leaves (\textit{n} = 17). Approximately 60% of expressed genes are differentially expressed between gall and leaf tissue at FDR < 0.05.

(H) Gene Ontology analysis of GO terms down- (left) and upregulated (right) in galls, presented as volcano plots. Genes involved in cell division and morphogenesis were strongly upregulated in galls, and genes involved in photosynthesis were strongly downregulated in galls. See also Figure S3.
**dgc** is a member of a large class of novel **bicycle** genes expressed specifically in the salivary glands of gall-inducing aphids

To identify additional proteins that aphids may inject into plants to contribute to gall development, we exploited the fact that only some individuals in the complex life cycle of *H. cornu* induce galls (Figure 1J). Only the fundatrix generation induces galls, and only her immediate offspring live alongside her in the developing gall. In contrast, individuals of generations that live on river birch (*Betula nigra*) through the summer and the sexual generation that feed on *H. virginiana* leaves in the autumn do not induce any leaf malformations. Thus, probably only the salivary glands of the generations that induce galls (the fundatrix [G1] and possibly also her immediate offspring [G2]) produce gall-effector molecules. We identified 3,048 genes upregulated in fundatrix salivary glands versus the fundatrix body (Figure S4A) and 3,427 genes upregulated in salivary glands of fundatrices, which induce galls, versus sexuals, which do not induce galls, although they feed on the same host plant (Figure S4B). Intersection of these gene sets identified 1,482 genes specifically enriched in the salivary glands of fundatrices (Figure S4C).

Half of these genes (744) were homologous to previously identified genes, many of which had functional annotations (Figure 5A). Gene Ontology analysis of the “annotated” genes suggests that they contribute mostly to the demands for high levels of protein secretion in fundatrix salivary glands (Figure S4D). Most do not encode proteins with secretion signals (671; Figure S4E) and are thus unlikely to be injected into plants. We searched for homologs of genes that have been proposed as candidate gall-effector genes in other insects but found little evidence that these classes of genes contribute to aphid gall development (Figures S4F–S4H). We therefore focused on the remaining 738 unannotated genes, which included 459 genes encoding proteins with predicted secretion signals (Figure S4E). Hierarchical clustering of the unannotated genes by sequence similarity identified one large (476 genes) and one small (43 genes) cluster of related genes and 222 genes sharing few or no homologs among the unannotated genes (Figure 5B). Genes in both the large and small clusters encode proteins with N-terminal secretion signals, as expected for effector proteins that might be injected into plants. The small cluster encodes a divergent set of proteins containing several conserved cysteines (C) and a well-conserved tryptophan (W) and glycine (G), and we named these CWG genes (Figures S5 A–S5C).

Proteins encoded by the large cluster display conservation mainly of a pair of widely spaced cysteine-tyrosine-cysteine

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**Figure 5.** **bicycle** gene transcripts are enriched in salivary-glands of gall-associated *H. cornu* generations

(A) Differential expression of fundatrix versus sexual salivary glands with only genes significantly upregulated in fundatrix salivary glands marked with colors, shown as a volcano plot. Genes with and without homologs in public databases are labeled as “annotated” (blue) and “not annotated” (brown), respectively.

(B) Hierarchical clustering of unannotated salivary-gland-specific genes reveals one large cluster of **bicycle** genes, one small cluster of CWG genes, and many largely unique, unclustered genes.

(C) Amino-acid logo for predicted **BICYCLE** proteins. Sequence alignment used for logo was filtered to highlight conserved positions. Most genes encode proteins with an N-terminal signal sequence and a pair of conserved cysteine-tyrosine-cysteine motifs (CYC).

(D) **bicycle** (green) and remaining unannotated (brown) genes labeled on a differential expression volcano plot illustrate that, on average, **bicycle** genes are the most strongly differentially expressed genes expressed specifically in fundatrix salivary glands.

(E) **bicycle** gene expression levels in salivary glands of aphids from four generations. **bicycle** genes are expressed at highest levels in the fundatrix and mostly decline in expression during subsequent generations. Sample sizes: G1 (n = 20); G2 (n = 4); G5 (n = 6); and G7 (n = 6). See also Figures S1, S4, and S5 and Table S2.
(CYC) motifs and spacing between the C, Y, and C residues of each motif are not well conserved (Figure 5C). This pair of CYC motifs led us to name these bicycle (bi-CYC-like) genes. The bicycle genes were the most strongly upregulated class of genes in fundatrix salivary glands (Figures 5A and 5B) and were expressed specifically in the salivary glands of the two generations associated with galls (G1 and G2; Figure 5E). Many bicycle genes are found in paralog clusters throughout the H. cornu genome (Figures 6A and 6B), and each bicycle gene contains approximately 5-25 microexons (Figures 6A and 6C)—a large excess relative to the genomic background (Figure 6C)—interrupted by long introns (Figure 6A).

We found that dgc shares many features with bicycle genes—it is strongly expressed specifically in fundatrix salivary glands, and it exhibits many microexons and a pair of CYC motifs (Figures 2D and S1M)—and that it is evolutionarily related to other bicycle genes. Thus, dgc is a member of a diverse family of genes encoding secreted proteins expressed specifically in the salivary glands of gall-forming generations. bicycle genes are therefore good candidates to encode many of the molecules required to generate the extensive transcriptional changes observed in galls.

**bicycle genes experienced intense diversifying selection, consistent with a potential arms race between aphids and plants**

bicycle genes are extremely diverse at the amino acid sequence level, as has been observed for other candidate insect gall effector genes. Each BICYCLE protein has accumulated approximately two substitutions per amino acid site since divergence from paralogs (Figure 6D). To explore whether this diversity resulted from natural selection rather than genetic drift, we compared rates of non-synonymous (dN) versus synonymous (dS) substitutions between the sister species H. cornu and H. hamamelidis (which also induces galls) in bicycle versus non-bicycle genes, because dN/dS values greater than one provide evidence for positive selection. To calculate polymorphism of orthologous genes in each species, we mapped sequencing reads from individuals of each species to the H. cornu genome. For divergence estimates, we estimated the H. hamamelidis genome by mapping H. hamamelidis sequencing reads to the H. cornu genome (STAR methods) and compared this genome with the original H. cornu genome. A large excess of bicycle genes displayed dN/dS significantly greater than 1 relative to the genomic background (Figures 7A and 7B; Table S1; p < 2.2e−16), revealing recurrent adaptive amino acid substitution at many bicycle genes since these species diverged.

We then quantified the frequency of adaptive amino acid substitutions at bicycle genes and other categories of genes overexpressed in fundatrix salivary glands by calculating z, the proportion of non-synonymous substitutions fixed by positive selection. We estimate that, for all bicycle genes, z = 0.33 (95% confidence interval [CI] 0.24–0.41) and that, for the subset of bicycle genes displaying dN/dS significantly greater than 1, z = 0.62 (95% CI 0.45–0.73; Table S2). Other categories of genes overexpressed in fundatrix salivary glands display values of z in the range of 0.27–0.38; however, bicycle genes display a considerably higher ratio of non-synonymous to synonymous substitutions than other categories of genes (Table S2). Most strikingly, as a fraction of protein length, bicycle genes display a considerably greater fraction of adaptive amino acid substitutions than other categories of genes (Figure 7C). Since speciation between H. cornu and H. hamamelidis, positive selection...
Figure 7. Genome-wide signals of selective sweeps are enriched near bicycle genes

(A) The majority of bicycle genes display dN/dS values larger than 1, with few showing strong sequence conservation (dN/dS << 1). Dashed vertical red line indicates dN/dS = 1.

(B) Non-bicycle genes are more conserved, on average, than bicycle genes (Mann-Whitney U test p = 2.6e-76). Dashed vertical red line indicates dN/dS = 1.

(C) Mean number of adaptive non-synonymous substitutions scaled by protein length for different categories of genes overexpressed in fundatrix salivary glands. As a proportion of protein length, bicycle genes display the fastest rate of adaptive evolution of any category of these genes. Error bars represent 95% confidence intervals. Note that the four categories on the right include all genes shown on the left but categorized by whether genes were annotated and included a signal peptide. Thus, for example, the category “no annotation with signal peptide” is composed mostly of bicycle and CWG genes.
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has resulted, on average, in fixation of approximately 2 to 3 substitutions in each bicycle gene and approximately 10 substitutions in the most rapidly evolving bicycle genes. This represents a considerable fraction of the average length of these proteins (~200 residues) and reveals intense selection on bicycle genes, presumably for novel functions that require multiple amino acid substitutions.

The previous tests cannot detect selection on non-coding regions and do not discriminate between selection acting in the deep past versus more recently. To search for recent selection in bicycle gene regions, we examined patterns of polymorphism and divergence within and between H. cornu and H. hamamelidis. Polymorphism was strongly reduced relative to divergence in bicycle gene regions compared with the genomic background (Figures 7D–7K and S6), and patterns of reduced polymorphism were strikingly similar in both species. This pattern is suggestive of recent selective sweeps in bicycle gene regions in both species, which we tested by performing genome-wide scans for positive selection.51 Genome-wide signals of selective sweeps were enriched near bicycle genes, and multiple signals fell within bicycle gene clusters in both species (Figures 7L–7O and S7). Thus, in addition to long-term adaptive protein evolution of BICYCLE proteins, it appears that strong positive selection has acted recently and presumably frequently near many bicycle genes throughout the genome.

In summary, we find evidence for widespread, strong, recent, and frequent positive selection on bicycle genes. Because bicycle genes are likely secreted from salivary glands specifically in gall-forming aphids, these observations are consistent with the hypothesis that bicycle genes encode proteins that are intimately involved in reciprocal molecular evolution between the aphid and their host plant.

DISCUSSION

We presented multiple lines of evidence that suggest that bicycle gene products provide instructive molecules required for aphid gall development. The strongest evidence is that eleven derived regulatory polymorphisms at dgc are associated with red galls (Figure 2), with almost complete silencing of dgc in aphid salivary glands (Figure 3) and with upregulation of seven anthocyanin biosynthetic genes and two red-purple anthocyanins in galls (Figure 4). Gall color is one small, but convenient, aspect of the panoply of cell biological changes required for gall development. We hypothesize that the product of each bicycle gene has its own unique set of targets in the plant and that the combined action of all bicycle gene products regulates many aspects of gall development. Testing this hypothesis will require the development of new methods to explore and manipulate this aphid-plant system.

BICYCLE protein functions
dgc likely encodes a protein that is deposited by aphids into gall tissue, and current evidence suggests that this protein specifically and dramatically results in the downregulation of seven anthocyanin biosynthetic genes. The mechanisms by which this novel aphid protein could alter plant transcription remains to be determined. The primary sequences of DGC and other BICYCLE proteins provide few clues to their molecular mode of action. Outside of the N-terminal secretion signal, BICYCLE proteins possess no similarity with previously reported proteins and display no conserved domains that might guide functional studies. The relatively well-conserved C-Y-C motif appears to define a pair of ~50–80 amino acid (aa) domains in each protein, and the paired cysteines may form disulfide bonds, which is commonly observed for secreted proteins. Secondary structure prediction methods provide little evidence for structural conservation across BICYCLE proteins, and the extensive variation in the spacing between conserved cysteines further suggests that BICYCLE proteins may display structural heterogeneity. Identification of their molecular mode of action will require identification of BICYCLE protein binding targets.

Evolution of bicycle genes
We were unable to detect any sequence homology between bicycle genes and previously reported genes, which is one reason that it is difficult to infer the molecular function of bicycle genes from sequence alone. It is possible that bicycle genes evolved de novo in an ancestor of gall-forming aphids, perhaps through capture of a 5’ exon encoding an N-terminal signal sequence. However, if bicycle genes experienced strong diversifying selection since their origin, perhaps in an ancestor of gall-forming aphids about 280 mya, then the rate of amino-acid substitution that we detected between two closely related species would likely be sufficient to have eliminated sequence similarity that could be detected by homology-detection algorithms. Identifying the evolutionary antecedents of bicycle genes will likely require tracing their evolutionary history across genomes of related species. It may also be more fruitful to use the unusual gene structure of bicycle genes to search for the antecedents of bicycle genes.

STAR*METHODS

Detailed methods are provided in the online version of this paper and include the following:

(D–G) Gene models and population genomic statistics for the 800-kb dgc bicycle gene cluster (D) and for three additional genomic regions containing bicycle gene clusters (E–G). Divergence between (black line) and polymorphism within H. cornu (blue line) and H. hamamelidis (pink line) in 3,000-bp windows is shown below gene models.

(H and J) Ratio of Pi to Dxy for bicycle and non-bicycle gene regions in H. cornu (H) and H. hamamelidis (J).

(I and K) The observed difference in Pi/Dxy between non-bicycle and bicycle genes (dashed red line) is much larger than the expectation generated by permuting the locations of Pi/Dxy values relative to gene locations for both H. cornu (I) and H. hamamelidis (K).

(L and N) Distance from each bicycle gene to the closest significant selective sweep signal is shown as red histogram, and dashed blue line indicates the median of this distribution for H. cornu (L) and H. hamamelidis (N).

(M and O) The median distance from each bicycle gene to the closest significant selective sweep signal from (L) for H. cornu (M) and from (N) for H. hamamelidis (O) is shown with dashed blue line, and the values after 1,000 permutations of sweep signals relative to gene locations are shown as gray histograms. The observed sweep signals are closer to bicycle genes than expected by chance. See also Figures S6 and S7 and Tables S1 and S2.
We thank Jim Truman for performing the initial aphid salivary gland dissections and for teaching us the tricks, which enabled the entire project; Erika Gajda and Henry Horn for help collecting galls; Patrick Reilly for his genotyping pipeline and helpful conversations; Goran Ceric for getting all software packages to run on the Janelia compute cluster; Tom Dolai for maintaining the Apollo genome annotation service; Mountain Lakes Biological Station for permission to collect specimens; and Jim Truman, Nicolas Frankel, Richard Mann, Brett Mensh, and Vanessa Ruta for helpful comments on the manuscript. Gall pigment analyses were partially supported by the USDA National Institute of Food and Agriculture (Hatch project W4122), the US National Institutes of Health (NIH) Award P30 CA016058, OSU, and OSUCCC. HHMI has filed a provisional patent, number 63/092,942, for the inventors A.K. and D.L.S., covering unique aphid polypeptides for use in modifying plants.
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### STAR METHODS

#### KEY RESOURCES TABLE

| REAGENT or RESOURCE | SOURCE | IDENTIFIER |
|---------------------|--------|------------|
| **Biological samples** |        |            |
| *Hormaphis cornu*    | This study | Methods S1 |
| *Hormaphis hamamelidis* | This study | Methods S1 |
| *Hamamelis virginiana* | This study | Methods S1 |
| **Chemicals, peptides, and recombinant proteins** |        |            |
| Calcofluor White     | Sigma-Aldrich | F3543 |
| Congo Red            | Sigma-Aldrich | C6767 |
| Escin                | Sigma-Aldrich | E1378 |
| Collagenase/Dispase   | Roche | 10269638001 |
| Hyaluronidase        | Sigma-Aldrich | H3884 |
| Methyl salicylate    | Sigma-Aldrich | M6752 |
| Phenol:Chloroform:Isoamyl alcohol | Thermo Fisher | AM9730 |
| Proteinase K         | Sigma-Aldrich | 0311587001 |
| RNasin Ribonuclease Inhibitor | Lucigen | 30281 |
| Hexadecytrimethylammonium bromide | Sigma-Aldrich | 52365 |
| Polyvinylpyrrolidone | Sigma-Aldrich | PVP40 |
| β-mercaptoethanol    | Sigma-Aldrich | M6250 |
| EDTA                 | Sigma-Aldrich | 324506 |
| Malvidin 3,5-diglucoside chloride | Sigma-Aldrich | PHL89727 |
| Peonidin-3,5-diglucoside chloride | Carbosynth | FP65437 |
| **Critical commercial assays** |        |            |
| Zymo ZR-96 Quick gDNA kit | Zymo Research | D3012 |
| Nextera XT DNA Library Preparation Kit | Illumina | FC-131-1096 |
| Quick-RNA Plant Miniprep Kit | Zymo Research | R2024 |
| PicoPure RNA Isolation Kit | Thermo Fisher | KIT0204 |
| MyBaits              | Arbor Bioscience | N/A |
| **Deposited data**   |        |            |
| Raw and analyzed data | This paper | Methods S1A–S1J |
| **Oligonucleotides** |        |            |
| WGB library sequencing primers | IDT, Methods S1K |
| WGB_RTr1:TCGTCCGGAGCGTCAAGATGTGATAAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGAGA

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## REAGENT or RESOURCE SOURCE IDENTIFIER

| REAGENT or RESOURCE | SOURCE | IDENTIFIER |
|---------------------|--------|------------|
| VIC: CCTATACCCACCTGCAGCTTAA | This paper | IDT, Methods S1M |
| VIC: TCTATCTGATGGACCTAA | This paper | IDT, Methods S1M |
| context_sequence: TTCGCTTAGTCATTCCTATCTGATG | This paper | IDT, Methods S1M |
| FWD: CCTATCTGATGGACCTAA | This paper | IDT, Methods S1M |
| REV: ATGGGTGACAAATGAAACACCGAACC gwas-8 | This paper | IDT, Methods S1M |
| VIC: ACATGTACCAAAATTTAA | This paper | IDT, Methods S1M |
| context_sequence: TTAGCTATGGCTTGGTGGAAATTTAATTTAATTTTTCATTCAAAATA | This paper | IDT, Methods S1M |
| FWD: ACATGTACCAAAATTTAA | This paper | IDT, Methods S1M |
| REV: TGGAAAACGTATAATGTATTATTTTAATTTTGAATTTGGT | This paper | IDT, Methods S1M |
| FWD: ACATGTACCAAAATTTAA | This paper | IDT, Methods S1M |
| REV: TGGAAAACGTATAATGTATTATTTTAATTTTGAATTTGGT | This paper | IDT, Methods S1M |
| gwas-other | This paper | IDT, Methods S1M |

### Software and algorithms

| Software and algorithms | Source | Website |
|-------------------------|--------|---------|
| FIJI | Chicago and HIC genome scaffolding | https://imagej.net/Fiji |
| Supernova version 2.1.1 | Dovetail Genomics | https://support.10xgenomics.com/de-novo-assembly/software/pipelines/latest/using/running |
| RepeatMasker version 4.0.9 | BUSCO | http://www.repeatmasker.org |
| BUSCO | gVolante web interface | https://busco.ezlab.org |
| BRAKER version 2.1.4 | BRAKER | https://github.com/Gaius-Augustus/BRAKER |
| UTRme | APOLLO | https://github.com/sradiouy/UTRme |
| Integrative Genomics Reviewer | Integrative Genomics Reviewer | https://github.com/sradiouy/UTRme |
| BWA version 0.7.17-r1188 | BWA | https://github.com/lh3/bwa |
| SAMtools version 0.1.19 / 1.3 / 1.7 | SAMtools | http://www.htslib.org/download/ |
| BCFtools version 0.1.19 / 1.7 / 1.9 | BCFtools | http://www.htslib.org/download/ |
| VCFtools version 0.1.15 / 0.1.16 | VCFtools | https://vcftools.github.io |
| R version 3.6.1 | R | https://www.r-project.org |
| SNPRelate version 1.20.1 (R package) | SNPRelate | https://bioconductor.org/packages/release/bioc/html/SNPRelate.html |
| ggplot2 3.2.2 (R package) | ggplot2 | https://ggplot2.tidyverse.org |

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### REAGENT or RESOURCE SOURCE IDENTIFIER

| REAGENT or RESOURCE | SOURCE | IDENTIFIER |
|---------------------|--------|------------|
| Picard version 2.18.0 | Broad Institute | http://broadinstitute.github.io/picard/ |
| GATK version 3.4 | https://gatk.broadinstitute.org/hc/en-us |
| PLINK version 1.90 | http://zzz.bwh.harvard.edu/plink/ |
| Sushi version 1.24.0 (R package) | https://github.com/phinanst/Sushi |
| BEDTools version 2.29.2 | https://bedtools.readthedocs.io/en/latest/ |
| vcR version 1.10.0 (R package) | https://github.com/knausb/vcR |
| snpStats version 1.36.0 (R package) | http://bioconductor.org/packages/release/bioc/html/snpStats.html |
| LDheatmap version 0.99.8 (R package) | https://sfustatgen.github.io/LDheatmap/index.html |
| Trim Galore! version 0.6.5 | http://www.bioinformatics.babraham.ac.uk/projects/trim_galore/ |
| cutadapt version 2.7 | https://cutadapt.readthedocs.io/en/stable/ |
| RepBase | https://www.girinst.org/repbase/ |
| PseudoreferencePipeline | P.F. Reilly (personal communication) | https://github.com/YourePrettyGood/PseudoreferencePipeline |
| HISAT version 2.1.0 | http://daehwankimlab.github.io/hisat2/ |
| HTSeq version 0.12.4 | https://htseq.readthedocs.io/en/master/ |
| Glimma version 1.14.0 (R package) | https://bioconductor.org/packages/release/bioc/html/Glimma.html |
| edgeR version 3.28.1 (R package) | http://bioconductor.org/packages/release/bioc/html/edgeR.html |
| EnhancedVolcano version 1.4.0 (R package) | https://github.com/kevinbielge/EnhancedVolcano |
| BLAST version 2.7.1 | https://blast.ncbi.nlm.nih.gov/Blast.cgi?PAGE_TYPE=BlatDocs&DOC_TYPE=Download |
| HMMER version 3.1b2 | http://hmmer.org |
| ReSpect database | https://rdr.io/github/WMBEdmands/compMS2Miner/man/ReSpect.html |
| UniProt/Swiss-Prot database | https://www.uniprot.org |
| WebGestalt 2019 | http://www.webgestalt.org |
| PFAM database | http://pfam.xfam.org |
| SignalP-5.0 | http://www.cbs.dtu.dk/services/SignalP/abstract.php |
| tmhmm version 2.0 | http://www.cbs.dtu.dk/services/TMHMM/ |
| MAFFFT version 7.419 | https://mafft.cbrc.jp/alignment/software/ |
| stats (R package) | https://www.rdocumentation.org/packages/stats/versions/3.6.2 |
| trimAl version 1.4 | http://trimal.cgenomics.org/downloads |
| Seqinr version 3.6-1 (R package) | http://seqinr.r-forge.r-project.org |
| Ggseqlogo version 0.1 (R package) | https://github.com/omarwagih/ggseqlogo |
| seqtk version 1.3 | https://github.com/lh3/seqtk |
| FastTree version 2.1.11 | http://www.microbesonline.org/fasttree/ |
| ggtree version 2.2.2 (R package) | https://bioconductor.org/packages/release/bioc/html/ggtree.html |
| PAML version 4.9j | http://abacus.gene.ucl.ac.uk/software/paml.html |
| Polymorphorama version 6 | https://ib.berkeley.edu/labs/bachtrog/data/poly/MORPHOrama/polyMORPHOrama.html |
| SweeD-P version 3.1 | https://cme.h-its.org/exelixis/web/software/sweed/ |
| MaCS version 0.4f | https://github.com/gchen98/macs |

### Other

| H. cornu RNA sequencing | This paper | Methods S1B |
| H. cornu red-green GWAS | This paper | Methods S1C |

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RESOURCE AVAILABILITY

Lead contact
Further information and requests for resources and reagents should be directed to and will be fulfilled by the Lead Contact, David L. Stern (sternd@hhmi.org).

Materials availability
This study did not generate new unique reagents.

Data and code availability
All sequencing data generated during this study are available at the NCBI Short Read Archive and the accession numbers are provided for each sample in Methods S1.

The genomes are available at GenBank and the genomes and gene annotations (GFF files) are available at FigShare (Methods S1J).

All of the analysis scripts we produced for this study are freely available at FigShare (Methods S1J).

EXPERIMENTAL MODEL AND SUBJECT DETAILS

Aphids and plants used in this study are listed in Methods S1.

METHOD DETAILS

Imaging of leaves and fundatrices inside developing galls
Young *Hamamelis virginiana* (witch hazel) leaves or leaves with early stage galls of *Hormaphis cornu* were fixed in Phosphate Buffered Saline (PBS: 137 mM NaCl, 2.7 mM KCl, 10 mM Na₂HPO₄, 1.8 mM KH₂PO₄, pH 7.4) containing 0.1% Triton X-100, 2% paraformaldehyde and 0.5% glutaraldehyde (paraformaldehyde and glutaraldehyde were EM grade from Electron Microscopy Services) at room temperature for two hours without agitation to prevent the disruption of the aphid stylet inserted into leaf tissue. Fixed leaves or galls were washed in PBS containing 0.1% Triton X-100, hand cut into small sections (~10 mm²), and embedded in 7% agarose for subsequent sectioning into 0.3 mm thick sections using a Leica Vibratome (VT1000s). Sectioned plant tissue was stained with 0.1 mg/mL Calcofluor White (Sigma-Aldrich, F3543) and 0.25 mg/mL Congo Red (Sigma-Aldrich, C6767) in PBS containing 0.1% Triton X-100, 0.5% DMSO and 0.05 mg/mL Escin (Sigma-Aldrich, E1378) at room temperature with gentle agitation for 2 days. Stained sections were washed with PBS containing 0.1% Triton X-100.

Soft tissues were digested and cleared to reduce light scattering during subsequent imaging using a mixture of 0.25 mg/mL collagenase/dispase (Roche #10269638001) and 0.25 mg/mL hyaluronidase (Sigma Aldrich #H3884) in PBS containing 0.1% Triton X-100 with 0.5% DMSO and 0.05 mg/mL Escin (Sigma-Aldrich, E1378) at room temperature with gentle agitation for 2 days. Stained sections were washed with PBS containing 0.1% Triton X-100. Soft tissues were digested and cleared to reduce light scattering during subsequent imaging using a mixture of 0.25 mg/mL collagenase/dispase (Roche #10269638001) and 0.25 mg/mL hyaluronidase (Sigma Aldrich #H3884) in PBS containing 0.1% Triton X-100 for 5 hours at 37°C. To avoid artifacts and warping caused by osmotic shrinkage of soft tissue and agarose, samples were gradually dehydrated in glycerol (2% to 80%) and then ethanol (20% to 100%) and mounted in methyl salicylate (Sigma-Aldrich, M6752) for imaging. Serial optical sections were obtained at 2 μm intervals on a Zeiss 880 confocal microscope with a Plan-Apochromat 10x/0.45 NA objective, at 1 μm with a LD-LCI 25x/0.8 NA objective or at 0.5 μm with a Plan-Apochromat 40x/0.8 NA objective. Maximum projections of confocal stacks or rotation of images were carried out using FIJI.

Hormaphis cornu genome sequencing, assembly, and annotation
We collected *H. cornu* aphids from a single gall for genome sequencing (Methods S1A). All aphids within the gall were presumed to be clonal offspring of a single fundatrix, because all *H. cornu* galls we have ever examined have contained only a single fundatrix and the ostiole of the galls was closed at the time we collected this gall, so there is little chance of inter-gall migration. High molecular weight (HMW) DNA was prepared by gently grinding aphids with a plastic pestle against the inside wall of a 2 mL Eppendorf tube in 1 mL of 0.5% SDS, 200 mM Tris-HCl pH 8, 25 mM EDTA, 250 mM NaCl with 10 μL of 1 mg/mL RNase A. Sample was incubated at 37°C for 1 hour and then 30μL of 10 mg/mL Proteinase K was added and the sample was incubated for an additional 1 hr at 50°C with gentle
agitation at 300 RPM. One mL of Phenol:Chloroform:Isoamyl alcohol (25:24:1) was added and the sample was centrifuged at 16,000 RCF for 10 min. The supernatant was removed to a new 2mL Eppendorf tube and the Phenol:Chloroform:Isoamyl alcohol extraction was repeated. The supernatant was removed to a new 2 mL tube and 2.5 X volumes of absolute ethanol were added. The sample was centrifuged at 16,000 RCF for 15 min and then washed with fresh 70% ethanol. All ethanol was removed with a pipette and the sample was air-dried for approximately 15 minutes and DNA was resuspended in 50 μL TE. This sample was sent to HudsonAlpha Institute for Biotechnology for genome sequencing.

DNA quality control, library preparation, and Chromium 10X linked read sequencing were performed by HudsonAlpha Institute for Biotechnology. Most of the mass of the HMW DNA appeared greater than approximately 50 kb on a pulsed field gel and paired end sequencing on an Illumina HiSeq X10 yielded 816M reads. The genome was assembled using Supernova using 175M reads, which generated the best genome N50 of a range of values tested. This 10X genome consisted of 21,072 scaffolds of total length 319.352 MB. The genome scaffold N50 was 839.101 KB and the maximum scaffold length was 3.495 MB.

We then contracted with Dovetail Genomics to apply Chicago (in vitro proximity ligation) and HiC (in vivo proximity ligation) to generate larger scaffolds (https://dovetailgenomics.com/ga_tech_overview/). We submitted HMW gDNA from the same sample used for 10X genome sequencing for Chicago and a separate sample of frozen aphids for HiC (Methods S1A). The Dovetail genome consisted of 11,244 scaffolds of total length 320.34 MB with a scaffold N50 of 36.084 Mb. This genome, named hormaphis_cornu_26Sep2017_PoQx8, contains 9 main scaffolds, each longer than 17.934 Mb, which appear to represent the expected 9 chromosomes of H. cornu. This assembly also includes the circular genome of the bacterial endosymbiont Buchnera aphidicola of 643,259 bp.

Assembly statistics

| Assembly statistics                      | hormaphis_cornu_26Sep2017_PoQx8 |
|------------------------------------------|----------------------------------|
| # of scaffolds:                          | 11242                            |
| Total length:                            | 320,336,030                      |
| Longest sequence:                        | 60,222,264                       |
| N50 sequence length:                     | 36,083,769                       |
| Sum length of sequences > 1M            | 297,831,669 (93.0% of total length) |
| Sum length of sequences > 10M           | 294,347,208 (91.9% of total length) |

BUSCO Analysis

| BUSCO Analysis                          | hormaphis_cornu_26Sep2017_PoQx8 |
|-----------------------------------------|----------------------------------|
| Total # of core genes queried           | 1066                             |
| # of complete core genes detected       | 1026 (96.25%)                    |
| # of complete and partial core genes detected | 1038 (97.37%)            |
| # of missing core genes:                | 28 (2.63%)                       |
| Average # of orthologs per core genes  | 1.02                             |
| % of detected core genes that have more than 1 ortholog | 2.34                             |

As further checks on the quality of this genome assembly, we examined the K-mer spectra (Figure S1A) and the HiC contact map (Figure S1B).

We annotated this genome for protein-coding genes using RNA-seq data collected from salivary glands and carcasses of many stages of the H. cornu life cycle (Methods S1B) using BRAKER. To increase the efficiency of mapping RNA-seq reads for differential expression analysis, we predicted 3’ UTRs using UTRme. We found that UTRme sometimes predicted UTRs within introns. We therefore applied a custom R script to remove UTRs located within introns. Later, after discovering the bicycle genes, we manually annotated all predicted bicycle genes, including 5’ and 3’ UTRs, in APOLLO by examining evidence from RNA-seq reads aligned to the genome with the Integrative Genomics Viewer. We found that the start sites of many bicycle genes were incorrectly annotated by BRAKER at a downstream methionine, inadvertently excluding predicted putative signal peptides from these genes. RNA-seq evidence often supported transcription start sites that preceded an upstream methionine and these exons were corrected in APOLLO. The combined collection of 18,895 automated and 667 manually curated gene models (19,582 total) were used for all subsequent analyses of H. cornu genomic data. The genome assembly (Genbank: JABAAO000000000) and sequence reads (BioProject: PRJNA614456) are available from NCBI. The genome assembly and our annotations are also available on FigShare (Methods S1J).

**Genome-wide association study of aphids inducing red and green galls**

Galls produced by H. cornu were collected in the early summer (Methods S1C) and dissected by making a single vertical cut down the side of each gall with a razor blade to expose the aphids inside. DNA was extracted using the Zymo ZR-96 Quick gDNA kit from the foundress of each gall. We performed tagmentation of genomic DNA derived from 47 individuals from red galls and 43 from green
galls using barcoded adaptors compatible with the Illumina sequencing platform. Tagmented samples were pooled without normalization, PCR amplified for 14 cycles, and sequenced on an Illumina NextSeq 500 to generated paired end 150 bp reads to an average depth of 2.9X genomic coverage. The average sequencing depth before filtering was calculated by multiplying the number of read pairs generated by SAMtools flagstat version 1.3 by the read length of 150bp, then dividing by the total genome size (323,839,449bp).

We performed principle component analysis on the genome-wide polymorphism data to detect any potential population structure that might confound a GWAS. Reads were mapped using bwa mem version 0.7.17-r1188 and joint genotyped using SAMtools mpileup version 1.3, with the flag -u/gf, followed by BCFtools call version 1.9, with the flag -m. Genotype calls were then filtered for quality and missingness using BCFtools filter and view version 1.9, where only SNPs with MAF > 0.05, QUAL > 20, and genotyped in at least 80% of the individuals were kept. To limit the number of SNPs for computational efficiency, the SNPs were additionally thinned using VCFtools –thin version 0.1.15 to exclude any SNPs within 1000 bp of each other. PCA was performed using the spngdsPCA function from the R package SNPRelate version 1.20.1 in R version 3.6.1.

We performed a GWAS with these low coverage data by mapping reads with bwa mem version 0.7.17-r1188 and calculating the likelihood of association with gall color with SAMtools mpileup version 0.1.19 and BCFtools view -vcs version 0.1.19 using BAM files as the input. Association for each SNP was measured by the likelihood-ratio test (LRT) value in the INFO field of the output VCF file, which is a one-degree of freedom association test p value. This method calculates association likelihoods using genotype likelihoods rather than hard genotype calls, ameliorating the issue of low-confidence genotype calls resulting from low-coverage data. The false discovery rate was set as the Bonferroni corrected value for 0.05, which was calculated as 0.05 / 50,957,130 (the total number of SNPs in the genome-wide association mapping).

**Enrichment and sequencing of the genomic region containing highly significant GWAS hits**

The low coverage GWAS identified multiple linked SNPS on chromosome 1 that were strongly associated with gall color (Figure 2A). To identify all candidate SNPs in this genomic region and to generate higher-confidence GWAS calls, we enriched this genomic region from a library of pooled tagmented samples of fundatrix DNA from red and green galls using custom designed Arbor Bioscience MyBaits for a 800,290 bp region on chromosome 1 spanning the highest scoring GWAS SNP (40,092,625 - 40,892,915 bp). This enriched library was sequenced on an Illumina NextSeq 550 generating paired-end 150 bp reads and resulted in usable re-sequencing data for 48 red gall-producing individuals and 42 green gall-producing individuals, with average pre-filtered sequencing depth of 58.2X.

We mapped reads with bwa mem version 0.7.17-r1188 and sorted bam files with SAMtools sort version 1.7, marked duplicate reads with Picard MarkDuplicates version 2.18.0 (http://broadinstitute.github.io/picard/), re-aligned indels using GATK IndelRealigner version 3.4, and called variants using SAMtools mpileup version 1.7 and BCFtools call version 1.7 (https://github.com/SAMtools/bcftools). This genotyping pipeline is available at https://github.com/YourePrettyGood/PseudoreferencePipeline (thereafter referred to as PseudoreferencePipeline). SNPs were quality filtered from the VCF file using BCFTools view version 1.7 at DP > 10 and MQ > 40 and merged using BCFTools merge.

For PCA analysis, the joint genotype calls were filtered for quality and missingness using BCFTools filter and view version 1.9, where only SNPs with MAF > 0.05 and genotyped in at least 80% of the individuals were retained. PCA was performed using the spngdsPCA function from the R package SNPRelate version 1.20.1 in Rstudio version 3.6.1.

Association testing was performed using PLINK version 1.90 with minor allele frequency filtered at MAF > 0.2. We did not apply a Hardy-Weinberg equilibrium filter because the samples were not randomly collected from nature. Red galls are rare in our population and we oversampled fundatrices from red galls to roughly match the number of fundatrices sampled from green galls. Results of the GWAS were plotted using the plotManhattan function of Sushi version 1.24.0.

To calculate LD for the 45kbp region surrounding the 11 most strongly associated SNPs, we extracted positions 40,475,000 – 40,520,000 of chromosome 1 from the merged VCF and retained SNPs that both exhibited MAF > 0.05 and were genotyped in at least 80% of the samples using BCFTools view and filter version 1.9.

To plot LD for the entire target enrichment region, we filtered the VCF for only the SNPs with MAF > 0.2 and that were genotyped in at least 80% of the samples using bcftools view and filter version 1.9, and thinned the resulting SNP set using VCFTools–thin version 0.1.15 to exclude any SNPs within 500bp of each other. We further merged back the 11 significant GWAS SNPs using bcftools concat, since the thinning process could have removed one or more of these SNPs. We also removed SNPs in regions where the H. cornu reference genome did not align with the genome of the sister species H. hamamelidis using BEDTools intersect version 2.29.2.

The LD heatmaps were generated using the R packages vcfR version 1.10.0, snpStats version 1.36.0, and LDheatmap version 0.99.8 in Rstudio version 3.6.1. The R code used to generate the LD heatmap figure was adapted from code provided at https://sfstatgen.github.io/LDheatmap/articles/vcfOnLDheatmap.html. The gene models were plotted using the plotGenes function from the R package Sushi version 1.24.0.

**Lack of evidence for chromosomal aberrations**

To identify possible chromosomal rearrangements or transposable elements that might be linked to the GWAS SNPs, we first trimmed adapters from the H. cornu target enrichment data using Trim Galore! version 0.6.5 and cutadapt version 2.7. The trimming pipeline is available at https://github.com/YourePrettyGood/ParsingPipeline. We then mapped the reads to the H. cornu reference genome with bwa mem version 0.7.17, sorted BAM files with SAMtools sort version 1.9, and marked duplicate reads with Picard MarkDuplicates.
We selected the subset of the BAM file for each individual that contained only the target enrichment region on chromosome 1 from 40,092,625 to 40,892,915 bp using SAMTools view version 1.9 and generated a merged BAM file for each color using SAMTools merge. The discordant reads were then extracted from each BAM file using SAMTools view with flag -F 1286 and the percentage of discordant reads was calculated as the ratio of the number of discordant reads over the total number of mapped reads for each 5000 bp window.

To further explore the possibility that chromosomal aberrations near the GWAS signal might differ between red- and green-gall producing individuals, we plotted the mapping locations of discordant reads in the 100 kbp region near the 11 GWAS hits using bedtools bamtostfq version 2.29.2 and used bwa mem to map these reads as single-end reads for read 1 and read 2 separately to a merged reference containing the *H. cornu* genome and the 343 *Acyrthosiphon pisum* transposable elements annotated in RepBase.79 We then removed duplicates and sorted the BAM file using SAMTools mdup and sort and determined the mapping location of all discordant reads using SAMTools view. We masked the windows on chromosome 1 from 40,400,000–40,599,999 bp in the genome-wide scatterplot of discordant reads mapping because the majority of the discordant reads are expected to map to these regions and displaying their counts would obscure potential signals in the rest of the genome.

Large scale survey of 11 dgc SNPs associated with gall color
Aphids were collected from red and green galls as described above for the GWAS study directly into Zymo DNA extraction buffer and ground with a plastic pestle. DNA was prepared using the ZR-96 Quick gDNA kit. We developed qPCR assays and amplicon-seq assays to genotype all individuals at all 11 SNPs (Methods S1L and S1M). PCR amplicon products were barcoded and samples were pooled for sequencing on an Illumina platform.

Adaptors were trimmed from amplicon reads using Trim Galore! version 0.6.5 and cutadapt version 2.7. The wrapper pipeline is available at https://github.com/YourePrettyGood/ParsingPipeline. We mapped reads to a 34 kbp region of chromosome 1 of the *H. cornu* genome that includes the amplicon SNPs (40,477,000 – 40,511,000 bp) with bwa mem version 0.7.17, sorted BAM files with SAMTools sort version 1.9, and re-aligned indels using GATK IndelRealigner version 3.4. No marking of duplicates was done given the nature of amplicon sequencing data. To maximize genotyping efficiency and improve accuracy, we performed variant calling with two distinct pipelines: SAMTools mpileup version 1.7 plus BCFtools call version 1.7, and GATK HaplotypeCaller version 3.4. The mapping and indel re-alignment pipelines are available as part of the MAP (with flag only_bwa) and IR functions of the PseudoreferencePipeline. Using the same PseudoreferencePipeline, variant calling was performed using the Mpileup function of BCFtools and HC function of GATK. FASTA sequences for each individual, where the genotyped SNPs were updated in the reference genome, were then generated for both BCFtools and GATK variant calls using the above PseudoreferencePipeline’s PSEUDOFASTA function, with flags “Mpileup, no_markdup” and “HC, no_markdup” respectively. The BCFTools SNP updating pipeline used bcftools filter, query, and consensus version 1.9, and we masked all sites where MQ ≤ 20 or QUAL ≤ 26 or DP ≤ 5. The HC SNP updating pipeline used GATK SelectVariants and FastaAlternateReferenceMaker version 3.4, and we masked all sites where MQ < 50, DP ≤ 5, GG < 90 or RQG < 90.

We then merged the variant calls from BCFtools and GATK, as well as the qPCR genotyping results, and manually identified all missing or discrepant genotypes. We manually curated these missing or discrepant genotype calls from the indel realigned BAM files using the following criteria: for heterozygous calls, the site had at least two reads supporting each allele, and for homozygous calls, the site had at least ten reads supporting the allele and no reads supporting alternative alleles.

RNA-seq of salivary glands from aphids inducing red and green galls
We dissected salivary glands from fundatrices isolated from green and red galls in PBS, gently pipetted the salivary glands from the dissection tray in < 0.5µL volume of PBS, and deposited glands into 3 µL of Smart-seq2 lysis buffer (0.2% Triton X-100, 0.1 U/µL RNAsin Ribonuclease Inhibitor). RNA-seq libraries were prepared with a single-cell RNA-seq method developed by the Janelia Quantitative Genomics core facility and described previously.52

RNA-seq libraries were prepared as described above for red and green gall samples except that the entire 3µL sample of salivary glands in lysis buffer was provided as input. Barcoded samples were pooled and sequenced on an Illumina NextSeq 550. We detected 9.0 million reads per sample on average. We replaced the original oligonucleotides with modified oligonucleotides to generate unstranded reads from the entire transcript (Methods S1K). Samples were PCR amplified for 18 cycles and the library was prepared using ¼ of the standard Nextera XT sample size and 150 pg of cDNA.

Differential expression analyses of fundatrix salivary glands from red and green galls
All differential expression analyses for plant and aphid samples were performed in R version 3.6.1 18 and all R Notebooks are provided on FigShare (Methods S1J). Adapters were trimmed using cutadapt version 2.7 and read counts per transcript were calculated by mapping reads to the genome with hisat2 version 2.1.0 101 and counting reads per gene with htseq-count version 0.12.4.81 In R,
technical replicates were examined and pooled, since all replicates were very similar to each other. We performed exploratory data analysis using interactive multidimensional scaling plots, using the command glMDSPlot from the package Glimma,82 and outlier samples were excluded from subsequent analyses. Differentially expressed genes were identified using the glmQLFTest and associated functions of the package edgeR.83 Volcano plots were generated using the EnhancedVolcano command from the package EnhancedVolcano version 1.4.0.84

**Hamamelis virginiana genome sequencing, assembly, and annotation**

Leaves from a single tree of *Hamamelis virginiana* were sampled from the Janelia Research Campus forest as follows. Branches containing leaves that were less than 50% expanded were wrapped with aluminum foil and harvested after 40 hours. Leaves were cleared of obvious contamination, including aphids and other insects, and then plunged into liquid N₂. Samples were stored at −80°C and sent to the Arizona Genomics Institute, University of Arizona on dry ice, which prepared HMW DNA from nuclei isolated from the frozen leaves. The Janelia Quantitative Genomics core facility generated a 10X Chromium linked-read library from this DNA and sequenced the library on an Illumina NextSeq 550 to generate 608M linked reads.

The *H. virginiana* genome was assembled with the supernova commands run and mkoutput version 2.1.1, with options minsize = 1000 and style = pseudohap.54 We used 332M reads in the assembly to achieve raw coverage of 56X as recommended by the supernova instruction manual. BUSCO completeness analysis56 was performed using the gVolante Web interface57 using the plants database. Genome assembly and BUSCO statistics are reported below.

| Assembly statistics                      | Hvir_nuclei_sn_run2_2_pseudohap |
|------------------------------------------|---------------------------------|
| # of scaffolds:                          | 84,975                          |
| Total length:                            | 907,642,797                     |
| Longest sequence:                       | 7,097,227                       |
| N50 sequence length:                    | 167,515                         |
| Sum length of sequences > 1M            | 142                             |
| Sum length of sequences > 1M (nt)       | 303,391,067 (33.4% of total length) |

| BUSCO Analysis                           | Hvir_nuclei_sn_run2_2_pseudohap |
|------------------------------------------|---------------------------------|
| Total # of core genes queried           | 1440                            |
| # of complete core genes detected       | 1309 (90.90%)                   |
| # of complete and partial core genes detected | 1365 (94.79%)               |
| # of missing core genes:                | 75 (5.21%)                      |
| Average # of orthologs per core genes: | 1.05                            |
| % of detected core genes that have more than 1 ortholog: | 3.90 |

The assembled genome reference was repeat masked with soft masking using RepeatMasker version 4.0.9.55 Twenty-five RNA-seq libraries from galls and leaves were used for genome annotation. RNA-seq reads were adaptor trimmed using cutadapt version 2.7 and sent to the genome using HISAT2 version 2.1.0. Genome annotation was performed with BRAKER version 2.1.4 using the RNA-seq data to provide intron hints58,59,60,86,106–108,110,111 and 3’ UTRs were predicted using UTRme.60 The genome assembly (Genbank: JAESVK000000000) and sequence reads (BioProject: PRJNA614456) are available from NCBI. The genome assembly and our annotations are also available on FigShare [Methods S1 J].

**H. virginiana RNA extraction and RNA-seq library preparation**

RNA was extracted from frozen *H. virginiana* leaf or gall tissue as follows. Plant tissue frozen at −80°C was placed into ZR Bashing-Bead Lysis Tubes (pre-chilled in liquid N₂) and pulverized to a fine powder in a Talboys High Throughput homogenizer (Troemer) with minimal thawing. Powdered plant tissue was suspended in extraction buffer (100 mM Tris-HCl, pH 7.5, 25 mM EDTA, 1.5 M NaCl, 2% (w/v) Hexadecytrimethylammonium bromide, 10% Polyvinylpyrrolidone (w/v) and 0.3% (v/v) β-mercaptoethanol) and heated to 55°C for 8 min followed by centrifugation at 13,000 x g for 5 minutes at room temperature to remove insoluble debris.712 Total RNA was extracted from the supernatant using the Quick-RNA Plant Miniprep Kit (Zymo Research) with the inclusion of in-column DNase I treatment. RNA-seq libraries were prepared with the Universal Plus mRNA-Seq kit (Nugen).

**Differential expression analysis of red versus green galls**

We performed differential expression analysis on red and green galls by collecting paired red and green gall samples from the same leaves. In total, we collected 17 red galls and 23 green galls from 17 leaves. RNA was prepared as described above for plant material and RNA-seq libraries were prepared with the single-cell RNA-seq method described above.
These RNA-seq data contained on average 4.4 million mapped reads per sample. Reads were quality trimmed and mapped to the transcriptome as described above. Only genes with greater than 1 count per million in at least 15 samples were included in subsequent analyses. Red and green samples clustered together in a principal components analysis (Figures S3B and S3C) and no samples were identified as outliers. The expression analysis model included the effect of leaf blocking.

**Gall pigment extraction and analysis**

Frozen gall tissue was ground to a powder under liquid nitrogen. We first tested for the presence of carotenoids by dehydrating gall tissue with methanol and extracting with hexane/acetone. However, the lipophilic extract was colorless and all color remained in the polar phase and pellet, indicating that carotenoids do not contribute to red gall color.

We therefore next tested for presence of anthocyanins. Approximately 20 mg of ground gall tissue was suspended in 100 μL methanol (Optima grade, Fisher Scientific) and 400 μL of 5% aqueous formic acid (Optima grade, Fisher Scientific), vortexed for 30 s and centrifuged at 8000 × g for 2 min at 10 °C. The supernatant containing pigment was filtered using a 0.2 μm diameter PTFE syringe filter to remove debris. Colorless pellet was discarded. Authentic anthocyanin pigment standards for malvidin 3,5-diglucoside chloride (Sigma Aldrich, St. Louis, MO, USA) and peonidin-3,5-diglucoside chloride (Carbosynth LLC, San Diego, CA, USA) were prepared at 1 mg/mL in 5% aqueous formic acid.

Pigment separation and identification alongside standards was performed on a reverse phase C18 column (Acquity Plus BEH, 50 mm × 2.1 mm, 1.7 μm particle size, Waters, Milford, MA) using an Agilent 1290 UHPLC coupled to an Agilent 6545 quadrupole time-of-flight mass spectrometer (Agilent Technologies, Santa Clara, CA, USA) using an ESI probe in positive ion mode. Five μL of filtered pigment extract or a 1:100 dilution of anthocyanin standard was injected. Solvent (A) consisted of 5% aqueous formic acid and Solvent (B) 1:99 water/acetonitrile acidified with 5% aqueous formic acid (v/v). The gradient conditions were as follows: 1 min hold at 0% B, 4 min linear increase to 20% B, 5 min linear increase to 40% B, ramp up to 95% B in 0.1 min, hold at 95% B for 5 min, return to 0% B and hold for 0.9 min. The column flow rate was 0.3 mL/min, and the column temperature 30 °C. The MS source parameters for initial anthocyanin detection were as follows: capillary = 4000 V, nozzle = 2000 V, gas temperature = 350 °C, gas flow = 13 L/min, nebulizer = 30 psi, sheath gas temperature = 400 °C, sheath gas flow = 12 L/min. DAD detection at 300 nm and 520 nm, and MS scanning from 50-1700 m/z at a rate of 2 spectra per second. Iterative fragmentation, followed by targeted MS/MS experiments, were performed using a collision energy = 35. Authentic standards confirmed the presence of peonidin-3,5-diglucoside and malvidin 3,5-diglucoside. The remaining anthocyanin species were identified using UV-Vis spectra, retention time relative to the other species in the sample, [M]+ precursor ions, and aglycone fragment ions matching the respective entries in the RIKEN database.

**Differential expression analysis of galls versus leaves**

We performed RNA seq on 36 gall samples and 17 adjacent leaf samples. These gall samples did not overlap with the gall samples used in red versus green gall comparison described earlier. For larger galls, RNA was isolated separately from basal, medial, and apical gall regions (Methods S1H). Libraries were sequenced on an Illumina NextSeq 550 to generate 150 bp paired-end reads with an average of 8.1 million mapped reads per sample. Only genes expressed at greater than 1 count per million in at least 18 samples were included in subsequent analysis. Gall and leaf samples clustered separately in a principal components analysis (Figures S3D and S3E) and no samples were excluded as outliers. We included only samples for which paired gall and leaf samples were available from the same leaf and potential leaf effects were modeled in the different expression analysis.

To facilitate Gene Ontology (GO) analysis, the UniProt IDs of the differentially expressed genes were obtained by mapping the coding sequences of the *H. virginana* genome to the UniProt/Swiss-Prot database using Protein-Protein BLAST 2.7.115,16 and by extracting the differentially expressed genes (Figure 4G). We used the WebGestalt 2019 webtool to perform GO analysis on the differentially expressed genes.

**Differential expression analysis of *H. cornu* organs and life stages**

RNA-seq libraries were generated for fundatrix salivary glands (n = 20) and whole bodies (n = 8), G2 salivary glands (n = 6) and carcasses (n = 3), G5 salivary glands (n = 6) and carcasses (n = 2), and G7 salivary glands (n = 5). Libraries were generated as described above for salivary glands except that RNA samples of carcasses and whole bodies were prepared using the Arcturus PicoPure RNA Isolation Kit (Applied Biosystems). Only genes expressed with at least 1 count per million in at least 29 samples were included in subsequent analyses.

**Bioinformatic identification of bicycle genes in *H. cornu***

Genes that were upregulated specifically in the salivary glands of the fundatrix generation were prime candidates for inducing galls. We therefore identified genes that were upregulated both in salivary glands of fundatrices versus sexuals and in fundatrix salivary glands versus fundatrix body. These differentially expressed genes were then separated into genes with and without homologs containing some functional annotation. Homologs with previous functional annotations were identified using three methods: we performed (1) translated query-protein (*blastx*) and (2) protein-protein (*blastp*) based homology searches using BLAST 2.7.115,16 against the UniProt/Swiss-Prot database and (3) Hidden-Markov based searches with the predicted proteins using hmmscan in HMMER version 3.1.0 against the pfam database. For all predicted proteins, we also searched for secretion signal peptides using SignalP-5.0 and for transmembrane domains using tmhmm version 2.0. Gene Ontology analysis of genes with annotations...
that were enriched in fundatrix salivary glands was performed by searching for Drosophila melanogaster homologs of differentially expressed genes and using these D. melanogaster homologs as input into gene ontology analysis.

To determine whether any of the differentially expressed genes without detectable homologs in existing protein databases were homologous to one another, we performed sensitive homology searches of all-against-all of these genes using jackhmmer in HMMER version 3.1b2. We performed hierarchical clustering on the quantitative results of the jackhmmer analysis by first calculating distances among genes with the dist function using method canberra and clustering using the hclust function with method ward.D2, both from the library stats in R. We aligned sequences of the clustered homologs using MAFFT version 7.419 with default parameters, trimmed aligned sequences using trimAI with parameters -gt 0.50, and generated sequence logos by importing alignments using the functions read.alignment and ggseqlogo in the R packages seqinr and ggseqlogo. After identification of the bicycle genes, we searched for additional bicycle genes in the entire H. cornu genome, which might not have been enriched in fundatrix salivary glands, using jackhmmer followed by hierarchical clustering to identify additional putative homologs. As described above, we manually annotated all of these candidate bicycle genes.

RNA-seq analysis of a single H. cornu fundatrix with a dgc\textsuperscript{Green} genotype inducing a red gall

Approximately 2.1% of fundatrices inducing red galls were homozygous for ancestral “green” alleles at all of the 11 dgc SNPs (Figure 2E). Five such individuals were found in our original GWAS study and we did not observe any variants in the dgc gene region that were associated specifically with these individuals, suggesting that they carried variants elsewhere in the genome that caused them to generate red galls. Since isolating salivary glands from fundatrices is challenging and time consuming, we were unable to systematically examine transcriptome changes in the salivary glands of the rare individuals homozygous for dgc\textsuperscript{Green} that induced red galls. However, we fortuitously isolated salivary glands from one fundatrix from a red gall that was homozygous for dgc\textsuperscript{Green}. We performed whole-transcriptome sequencing of the salivary glands from this one individual and compared expression levels of all genes in the bicycle gene paralog group to which dgc belongs. We also examined the dgc transcripts produced by this individual and found no exonic SNPs in the dgc transcripts produced by this individual, indicating that this individual probably expressed a functional dgc transcript.

H. cornu and H. hamamelidis polymorphism and divergence measurements in GWAS region

To summarize the polymorphism and divergence patterns in H. cornu and H. hamamelidis in the target enrichment region, we generated SNP updated FASTA sequences for each individual by mapping the H. cornu reads to the H. cornu genome reference and the H. hamamelidis reads to a H. hamamelidis SNP-updated reference genome in H. cornu genome coordinate space. We used the same genotyping pipeline (PseudoreferencePipeline) as described above for the enrichment region GWAS and sites with DP ≤ 10, MQ ≤ 20.0, or QUAL ≤ 29.5 were masked. High coverage individuals of H. cornu (n = 90) and H. hamamelidis (N = 92) were included in this analysis, including both of the color phenotypes. Polymorphism within each species and divergence (Dxy) between species were calculated using a custom script calculateDiversity (available at https://github.com/YourePrettyGood/DyakInversions/tree/master/tools). Windowed measurements of each of these statistics were generated with the custom script nonOverlappingWindows.cpp, using window size 3000bp (script available at https://github.com/YourePrettyGood/RandomScripts). Sites with missing genotypes due to masking in 50% or more of the samples were not included in the windowed average and windows where 50% or more of the sites were missing were excluded from the plots in Figures 7D–7G.

Whole-genome polymorphism and divergence measurements in H. cornu and H. hamamelidis

The sister species H. cornu and H. hamamelidis produce similar looking galls in adjacent geographic areas and were long confused as populations of a single species. However, H. cornu exhibits a life cycle where aphids alternate between Hamamelis virginiana and Betula nigra (Figure 1J), whereas H. hamamelidis does not host alternate to B. nigra and displays a truncated life cycle, where the offspring of the fundatrix develop as sexparapae (the equivalent of G6 in Figure 1J) and deposit sexuals on H. virginiana in the autumn. Populations of H. cornu tend to live in the lowlands with prolonged summers and H. hamamelidis tend to live in the highlands and more northern latitudes, which experience shorter summers. In some locations, both species can be found making galls on the same trees.

Previous mtDNA sequencing supported the hypothesis that these are distinct species. We estimate genome-wide divergence (Dxy) between these species to be 0.0189 and polymorphism to be 0.0132 and 0.0125 for H. cornu and H. hamamelidis, respectively, providing further support that these are reproductively separate species. Assuming a mutation rate of $\mu = 2.8e-9$, we estimated effective population sizes using $\theta = 4N_e\mu$ for H. cornu and H. hamamelidis as 1,174,297 and 1,115,822, respectively.

To examine genome-wide patterns of polymorphism and divergence, we used the same whole-genome sequencing data used in the genome-wide association study for gall color. These data included samples from fundatrices isolated from 43 green galls and 47 red galls for H. cornu and 48 galls for H. hamamelidis. Reads were mapped using bwa mem version 0.7.17-r1188. The H. cornu reads were mapped to the H. cornu reference genome and the H. hamamelidis reads were mapped to the H. hamamelidis SNP-updated reference genome. We joint genotyped each species separately using bcftools version 1.9 mpileup and call -m to generate raw, multisample VCF for each species. Sites filtering for QUAL $> 20$ was then done using bcftools filter, and insertions and deletions sites were removed using VCFtools version 0.1.16--remove-indels. To mask sites lacking genotype calls for each sample individually, we used bcftools query to extract the positions of all sites, bcftools view with flag -g ^miss to extract the positions of genotyped sites for
each sample, and bedtools version 2.29.2 complement to generate the positions of non-genotyped sites, which were then masked. Consensus sequences were generated for each sample using bcftools consensus, with the flag –iupac-codes.

Polymorphism within each species and divergence (Dxy) between species was calculated using a custom script calculateDiversity (available at https://github.com/YourePrettyGood/DyakInversions/tree/master/tools). Windowed measurements of each of these statistics were generated with the custom script nonOverlappingWindows.cpp, using window size 1000bp (script available at https://github.com/YourePrettyGood/RandomScripts). Windows were designated as “bicyle” if they overlapped in coordinates with any bicycle genes; otherwise they were designated as “non-bicycle.” Sites masked in 80% or more of the samples were excluded in the windowed average and windows with 80% or more of the sites missing were excluded from genome-wide summary statistic calculations. This genotyping rate filter is more lenient than the enrichment region genotyping to accommodate the shallower sequencing depth of the whole-genome data, which resulted in a higher rate of missing data in the whole-genome data.

Bioinformatic analysis of bicycle gene structure and evolution

The H. cornu median exon size and number of exons per bicycle gene were calculated from the GFF annotation file Augustus.updated_w_annots.21Aug20.gff3. The bicycle genes alignment for the divergence tree was generated using FastTree version 2.1.11 and plotted using ggtree version 2.2.2 in R.

To calculate DnDs between H. cornu and H. hamamelidis, we first generated a masked H. hamamelidis reference genome by updating the H. cornu reference genome with H. hamamelidis SNPs. We selected a random subset of H. hamamelidis reads from the 150 bp PE 10X linked reads sequencing data to approximately 65X coverage using seqtk version 1.3 sample -s100 (https://github.com/YourePrettyGood/RandomScripts) and trimmed adapters using Long Ranger version 2.2.2 basic (10X Genomics). We then mapped the H. hamamelidis reads to the H. cornu reference and updated the SNPs using the PseudoreferencePipeline. Sites where MQ ≤ 20, QUAL ≤ 26 or DP ≤ 5 were masked, resulting in 21.0% of the genome being masked. This H. hamamelidis reference genome was used in the DnDs calculation.

DnDs between all orthologous genes in H. cornu and H. hamamelidis was calculated using the codeml function from the PAML package, version 4.9.1. The CDS sequences were extracted from both reference genomes for each gene using constructCD SesFromGFF3.pl and we split all degenerate bases into the two alleles to generate two pseudo-haplotypes using fakeHaplotype.pl -s 42 (both scripts available at https://github.com/YourePrettyGood/RandomScripts). Haplotype 1 was used as input into codeml for both species. The settings used in the codeml control file were: runmode = –2, seqtype = 1, CodonFreq = 0, clock = 0, model = 0, NSeites = 0, fix_kappa = 0, kappa = 2, fix_omega = 0, omega = 1, Small_Diff = 0.5e-6, method = 0, fix_blength = 0. Then, to evaluate whether DnDs is significantly different from 1 for each gene, we additionally ran a null model with the same control file, except with fix_omega = 1. We then calculated the likelihood scores between the two models as 2(lnL1-lnL2) and compared it to the 95th percentile of the Chi-square distribution with 1 degree of freedom. We performed a Mann-Whitney U test comparing DnDs of bicycle and non-bicycle genes using the wilcox.test function in R version 3.6.1.

To perform McDonald-Kreitman tests for bicycle genes, we generated a population sample of the bicycle gene coding regions for H. cornu by mapping RNA-seq data from 21 fundatrix salivary gland samples, which provided high coverage. We performed genotyping using the STAR and IRRNA, and HC functions in the PseudoreferencePipeline. This pipeline used STAR Version 2.7.3a for mapping and IndelRealigner and HaplotypeCaller in GATK version 3.4 for indel realignment and variant calling. SNP-updated FASTA sequences were generated for each individual from the genotype VCF using the PSEUDOFASTA function of the Pseudoreference Pipeline without any additional masking. Then, for each individual, we split all degenerate bases into the two alleles to generate two pseudo-haplotypes using fakeHaplotype.pl -s 42. We calculated the number of synonymous and nonsynonymous polymorphisms (Ps and Pn, respectively) and divergent substitutions (Ds and Dn, respectively) using Polymorphorama version 6 (https://ib.berkeley.edu/labs/bachtrog/data/polyMORPHOrama/polyMORPHOrama.html) for multiple categories of genes significantly overexpressed in the fundatrix salivary gland. The fraction of non-synonymous divergent substitutions that were fixed by positive selection was estimated using the Cochran-Mantel-Haenszel test framework. Using the mantelhaen.test function in R version 3.6.1, we estimated a common odds ratio \( \frac{P_s}{P_n} \) across a series of McDonald-Kreitman tables, and estimated \( \alpha = 1 - \frac{D_s}{D_n} \). A MAF filter of 0.03 was applied to the polymorphism counts to reduce the downward bias in the estimator due to segregating weakly deleterious amino acid polymorphisms.

Genome-wide tests of selection

To scan for genome-wide signatures of positive selection in H. cornu and H. hamamelidis, we used the joint-genotyped VCF for each species as described above. We filtered for QUAL ≥ 20 using bcftools filter, and 80% maximum missing genotype and biallelic sites using VCFTools version 0.1.16 -max-missing 0.2, -max-alleles 2, and -min-alleles 2. We then performed the selection scan using SweeD-P version 3.1, with one grid point per kilobase and -folded flag.

To determine the significance cutoff for the composite likelihood ratio (CLR) output by SweeD, we simulated 10 Mbp regions under neutrality for 100 haplotypes for each species using MacCS version 0.4f. The effective population size was set to 1,174,297 and 1,115,822 for H. cornu and H. hamamelidis, respectively. The mutation and recombination rates were set to 2.8e-9 and 1.1e-8, respectively, per bp per generation. The simulation output was converted to multi-sample VCF format using a custom script and run through SweeD-P using the same parameters as for the observed data. We ran ten iterations of the simulation for each species.
and combined the results of each iteration within each species to account for stochasticity of the simulations. The CLR cutoffs for the observed data were chosen as the 99th quantiles of the CLR values from the neutral simulations of their respective species.

**Identification of H. cornu homologs of genes previously proposed as gall effector genes in other species**

The SSGP-71 s effector gene family was described from the Hessian fly genome and it was suggested that these encode E3-ligase-mimicking effector proteins. We identified four homologs of the SSGP-71 s effector gene family in *H. cornu* by searching *H. cornu* protein sequences using a Hidden Markov Model search with a protein profile generated using the 426 SSGP-71 genes from Document S2, SSGP-71 s tab, from Zhao et al. (hmmsearch with HMMER v3.2.1). Two of these genes were not expressed at sufficient levels to be detected in our differential expression studies and neither of the remaining two were among the 1,482 genes specifically enriched in the salivary glands of fundatrices (Figure S4F). Thus, it is unlikely that SSGP homologs contribute to gall development in *H. cornu*.

We also identified eight ubiquitin E3-ligase genes in *H. cornu* based on protein similarity to proteins in the PFAM database that include “E3” in the description. Seven of these genes had sufficient expression to be detected in the differential expression studies, and only two were among the 1,482 genes enriched in fundatrix salivary glands (Figure S4G). In addition, none of these seven genes contain N-terminal secretion signals, making it unlikely that they could be injected into plants. Thus, secreted ubiquitin E3-ligase genes are unlikely to contribute to *H. cornu* gall development.

Ring domain proteins usually mediate E2 ubiquitin ligase activity, which can modify protein function or target proteins for proteosomal degradation. The genome of a galling phylloxeran, a species from the sister family to aphids, has been shown to encode secreted RING domain proteins, although it is not yet clear how the phylloxeran secreted RING domain proteins act and whether they are involved in gall-specific processes. We identified two *H. cornu* secreted RING domain proteins based on their similarity to proteins in the PFAM database that include “Ring finger domain” in the description, and only one was among the 1,482 genes specifically upregulated in fundatrix salivary glands (Figure S4H). Thus, secreted RING domain proteins are unlikely to contribute to *H. cornu* gall development.

**QUANTIFICATION AND STATISTICAL ANALYSIS**

All quantitative methods and statistical analyses were explained in the Method details section.
Correction

A novel family of secreted insect proteins linked to plant gall development

Aishwarya Korgaonkar, Clair Han, Andrew L. Lemire, Igor Siwanowicz, Djawed Bennouna, Rachel E. Kopec, Peter Andolfatto, Shuji Shigenobu, and David L. Stern

*Correspondence: sternd@hhmi.org

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In the published version of this article, an error was identified in Figure 3A. In the original figure, the salivary gland data for G1 and G2 nymphal stage 1 were plotted as raw read counts, rather than counts per million reads. The corrected figure shows these values as counts per million reads in the online version of the paper. In addition, the article contained a typographical error in one sentence that incorrectly reported that anthocyanin biosynthetic genes were downregulated in red galls. As shown in Figure 4A, these genes are upregulated in red galls. The text has been corrected online and all other sentences that discuss these data are correct. No conclusions or statements in the text are altered by this correction. We apologize to readers for any confusion this may have caused.

Figure 3A. *dgc* is the most strongly differentially expressed gene in salivary glands of fundatrices collected from red versus green galls (corrected)

Figure 3A. *dgc* is the most strongly differentially expressed gene in salivary glands of fundatrices collected from red versus green galls (original)