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13 Contemporary evolution of a Lepidopteran species, *Heliothis virescens*, in response to modern 14 agricultural practices

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26 Abstract

27 Adaptation to human-induced environmental change has the potential to profoundly influence the genomic architecture of affected species. This is particularly true in agricultural ecosystems, where 28 29 anthropogenic selection pressure is strong. *Heliothis virescens* feeds on cotton in its larval stages and 30 US populations have been declining since the widespread planting of transgenic cotton, which 31 endogenously express proteins derived Bacillus thuringiensis (Bt). No physiological adaptation to Bt 32 toxin has been found in the field, so adaptation to this altered environment could involve: 1) shifts in 33 host plant selection mechanisms to avoid cotton, 2) changes in detoxification mechanisms required for 34 cotton-feeding versus feeding on other host plants, or 3) loss of resistance to previously used 35 management practices including insecticides. Here we begin to address the question of whether such 36 changes occurred in *H. virescens* populations between the years 1997 and 2012. As part of our study, 37 we produced an *H. virescens* genome assembly and used this in concert with a ddRAD-seq enabled 38 genome scan to identify loci with significant allele frequency changes over the 15 year period after Bt 39 crops became widespread in the agricultural landscape. Confirmation that a known genetic change at a 40 previously described *H. virescens* target of selection was detectable in our genome scan increased our confidence in this methodology. We quantified the strength of selection required to elicit the observed 41 allele frequency changes at loci under selection. Potential contributions of genes near the loci under 42 43 selection and their to adaptive phenotypes in the *H. virescens* cotton system are discussed.

44

45 Key Words

46 Heliothis virescens, tobacco budworm, Bacillus thuringiensis, cotton, selective sweep

47 Introduction

48 Human-induced change in the natural landscape places strong selective pressure on populations to adapt over relatively short evolutionary timescales (Palumbi 2001). These changes shape the 49 50 genomes of local species, providing insight into contemporary evolution and it's implications for 51 affected species. Cultivation of the natural landscape for agricultural purposes is one of the most ubiquitous examples of human-induced environmental change. Modern agricultural practices often 52 53 involve sweeping changes to the composition of plant species across broad geographic regions, re-54 sculpting of the physical terrain and chemical inputs into the environment (Tilman 2001). The strong 55 selective pressure placed on species that inhabit agricultural ecosystems make them ideal for examining 56 genetic responses to anthropogenic forces (Taylor *et al.* 1995).

57 One such major change in recent agricultural history is the commercialization of transgenic 58 crops that themselves produce proteins for the management of key insect species. The tobacco 59 budworm, *Heliothis virescens*, feeds on cotton in its larval stages and populations in the Southern 60 United States have been declining since the widespread planting of transgenic cotton (Supplementary 61 Figure 1). These cotton cultivars endogenously express insecticidal proteins derived from the bacterium 62 Bacillus thuringiensis (Bt), which are lethal to H. virescens. In the Southern United States, Bt-63 expressing cotton was rapidly adopted after it became commercially available for management of *H*. 64 virescens in 1996 (James 2015; Supplementary Figure 2). Prior to the widespread use of Bt-expressing 65 cotton, populations of *H. virescens* had evolved resistance to every insecticide used for their management (Blanco 2012), including pyrethroid insectides (Luttrell et al. 1987, Campanhola and 66 67 Plapp 1989). Concerns over the possibility that *H. virescens* and other insect targets of Bt crops would 68 evolve resistance to the endogenously expressed proteins spawned an entire field of research related to 69 Bt resistance and associated genetic mechanisms (Reviewed in Heckel et al. 2007, Tabashnik et al.

2013). Of primary concern was the loss of efficacy of toxic Bt proteins (USEPA 1998, 2001, 2006). In
the case of *H. virescens*, no physiological adaptation to the Bt toxin in cotton fields has been detected
(Tabashnik *et al.* 2013). Yet widespread adoption of Bt-expressing crops likely placed selective
pressure on *H. virescens* in other ways.

74 As one example, widespread planting of Bt cotton cultivars led to an overall decline in 75 insecticide use on cotton in the United States (NASEM 2016, Benbrook 2012), including the use of 76 pyrethroids (personal comm. D. Reisig). In 1999, North Carolina extension entomologists stopped 77 recommending pyrethroids for the damaging generations of *H. virescens* in cotton (Bacheler 1999) and 78 still do not recommend pyrethroids. Similarly, in Louisiana, pyrethroids are no longer recommended 79 for *H. virescens* (LSU Ag Center 2016). In the state of California, one of the only states that makes 80 records of pesticide sales and applications publicly available, the pounds per acre of cotton for three 81 commonly used pyrethroids (deltamethrin, cypermethrin and cyfluthrin) declined from 0.05 to 0.03 82 between the years 2000 and 2012 (http://calpip.cdpr.ca.gov), and it is not clear if any of these 83 pyrethroids sprays are currently used to target *H. virescens*. Prior to Bt cotton adoption, when 84 pyrethroids were heavily used to suppress *H. virescens* populations, the voltage-gated sodium channel 85 gene (Vasc) was identified as one gene target of selection wherein pyrethroid resistance alleles rose to high frequency (Park and Taylor 1997, Park et al. 1997). Yet in H. virescens and other insect species, 86 87 *Vqsc* mutations often result in an overall loss of fitness for individuals carrying them (Zhao *et al.* 2000, 88 Foster et al. 2005, Kliot and Ghanim 2012, Brito et al. 2013). Under these conditions, stability in the 89 frequency of insecticide resistance alleles depends upon whether or not populations are continually 90 exposed to insecticidal pressure. Therefore, one possible effect of Bt adoption in *H. virescens* is a 91 reversion to susceptibility at their pyrethroid resistance locus.

92

Additional inadvertent targets of selection by Bt-expressing cotton could include loci involved

93 in feeding and oviposition behaviors. *H. virescens* are well known for the damage they cause to 94 cultivated cotton (reviewed in Blanco 2012), but their host plant range includes tobacco, soybean, 95 garbanzo bean (Fitt 1989), and a number of wild hosts (Sudbrink and Grant 1995). Heritable, intra-96 specific variation in host choice has been observed for *H. virescens* (Sheck and Gould 1993, Sheck and 97 Gould 1995, Karpinski et al. 2014), as well as other closely-related Lepidopteran species (Jallow and 98 Zalucki 1996, Jallow et al. 2004, Oppenheim et al. 2012). It is possible that the introduction of Bt-99 expressing cotton has selected against *H. virescens* individuals that preferentially oviposit and feed 100 upon cotton hosts plants. Allele frequency changes in genes associated with chemosensation, central 101 nervous system function, and metabolism may have occurred as *H. virescens* was driven off of its 102 primary cotton host plant (Blanco 2012).

103 In recent years, identifying genomic change in response to selective forces has been enabled by 104 the development of next-generation sequencing (NGS) technologies. A variety of NGS-enabled marker 105 development techniques are used to generate novel, high density marker sets for model and non-model 106 organisms, including Restriction-site Associated DNA sequencing (RAD-seq; Baird et al. 2008), 107 Genotyping-by-Sequencing (GBS; Elshire *et al.* 2011), double-digest RAD-seq (ddRAD-seq; Peterson et al. 2012) and others (reviewed in Andrews et al. 2016). These marker sets enable scientists to scan 108 109 the genomes of field-collected organisms in search of the gene targets of selection, particularly where 110 selection is strong and a reference genome assembly for read alignment is available (Lowry *et al.* 2017, 111 Catchen *et al.* 2017). Strong selection for advantageous alleles at target genes can influence allelic 112 composition at physically linked neutral sequences, including nearby marker sites, resulting in a 113 genomic footprint of selection that is much broader than the target gene alone (Nielsen 2005). The 114 breadth of this genomic footprint is influenced by several factors, including the strength of selection, 115 the initial frequency of the advantageous allele, effective pest population size and recombination rate

116 (Charlesworth and Charlesworth 2010).

117 Here we scanned the genomes of two *H. virescens* field populations collected in the Southern United States between the years 1997 and 2012 to detect loci that have changed over time. Given that 118 119 pesticides and transgenic crops impose very strong selection on their target pest species (Onstad 2014), 120 we initially focused on a genomic region known to be associated with insecticide resistance as a confirmation that ddRAD-seq could be used identify genes responsible for adaptive phenotypes under 121 122 strong selection. To achieve this goal, we produced an annotated draft assembly of the *H. virescens* 123 genome and used it for alignment of ddRAD-seq reads from barcoded individuals collected across 124 space and time. We then tested the hypothesis that changes in a candidate pyrethroid resistance gene, 125 *Vqsc*, could be detected through our ddRAD-enabled genome scanning techniques. Furthermore, we 126 identified additional ddRAD-seq loci with strongly diverging marker allele frequencies, and quantified 127 the strength of selection required to produced the observed changes at these sites. Some of the ddRAD-128 seq loci identified as under seleciton were linked to genes involved in toxin metabolism and 129 chemosensation. We concluded with a discussion of the adaptive phenotypes that these newly 130 identified gene targets of selection might represent in a field environment. 131

132 Methods

133 Insect Material

For all population genomic analyses, adult male moths were collected by pheromone-baited trap from Bossier Parish, LA, and Burleson County, TX. Collections took place in LA from May through September, and in TX from May through October, in the years 1997, 2002, 2007, and 2012. GPS coordinates from trapping locations can be found in Supplementary Table 1. Moths from each collection date and location were immediately placed together in bottles of 95% ethanol for long-term 139 storage. Bottles from 2002, 2007 and 2012 were always held at -20°C until specimens were used, 140 while those from 1997 were initially held at room temperature and then transferred to -20°C. To 141 develop our *H. virescens* genome assembly, individuals from a long-standing colony strain (Gould *et* 142 al. 1995) were sib-mated for 10 generations to produce inbred material for sequencing (Fritz et al. 143 2016). Siblings from a single inbred family were used for sequencing and analysis. Five sibling pupae 144 were stored at -80°C prior to DNA isolation and library preparation. For all insect samples, DNA was 145 isolated with a Qiagen Blood and Tissue Kit (Qiagen, Inc., Valencia, CA, U.S.A.) using the mouse tail 146 protocol.

147

148 H. virescens Candidate Gene Approach

149 A polymerase chain reaction (PCR) based upon the methods of Park and Taylor (1997) was 150 used to amplify a 432 bp region in the alpha subunit of the *Vgsc*. The primer pair Nhp3304+ (5' 151 ATGTG GGACT GIATG TTGGT) and Nhp3448- (5' CTGTT GAAGG CCTCT GCTAT) flanked a mutation known as L1029H. In this targeted region of the *Vqsc*, a single nucleotide polymorphism 152 153 (SNP) caused a Leucine to Histidine amino acid substitution and thereby pyrethroid resistance. 154 Additional mutations associated with pyrethroid resistance have been detected in *H. virescens*, 155 including D1561V + E1565G and V421M (Rinkevich *et al.* 2013). We specifically targeted L1029H 156 for our research because D1561V+ E1565G has not yet been functionally confirmed using ectopic 157 expression assays (Rinkevich et al. 2013), and V421M was not found in our study populations, even in 158 1997 when phenotypic pyrethroid resistance was at its peak. 159 Amplicons from PCRs targeting the L1029H mutation were digested by restriction enzyme Nla-

160 III, which cut in the presence of the resistance allele (Supplementary Figure 3). Genotypes were scored

161 by visualizing the digested PCR products on a 3.5% agarose gel (90 to 120 min at 120 V). We

162 examined the genotypes at this pyrethroid resistance locus for *H. virescens* individuals collected from 163 1997 (n = 194), 2002 (n = 204), 2007 (n = 268), and 2012 (n = 194) in LA, and 1997 (n = 142), 2007 (n 164 = 120), and 2012 (n = 196) in TX. Changes in pyrethroid resistance allele frequencies over time and 165 space were examined using a series of nested generalized linear regression models with binomial error 166 structures in R version 3.1.2 (R Core Team 2014; used here and throughout). The following full model 167 was used to examine the frequency of individual pyrethroid resistance alleles (i):

168
$$\Pr(y_i = 1) = \operatorname{logit} (\beta_{0i} + \beta_{1 \operatorname{Yeari}} + \beta_{2 \operatorname{Seasoni}} + \beta_{3 \operatorname{Locationi}} + \beta_{4 \operatorname{Year \times Seasoni}}),$$

170 where Year represented collection year (e.g. 1997, 2002, 2007, or 2012), Season represented whether 171 the collections were made early (May or June) or late (August through October) in the cotton growing 172 season, and Location represented the collection location of the samples. We identified a model term as 173 statistically significant ($\alpha = 0.05$) when a comparison of nested models by analysis of deviance 174 indicated that removal of that term significantly influenced model deviance.

for *i* = 1,...., n

175

176 Strength of Selection against the Vgsc Resistance Allele

177 Following the discovery of a decline in frequency of the pyrethroid resistance allele, we quantified the strength of selection associated with the decline in pyrethroid pressure. We used the following 178 179 equation to calculate the selection coefficient against the recessive resistance allele for our field 180 populations from TX and LA over 15 years after the introduction of Bt cotton into the landscape:

- 181
- $\triangle p = \frac{s p q^2}{1 s q^2}$ 182

183 where p was the frequency of the susceptible allele, q was the frequency of the resistance allele, and *s* 184 was the selection coefficient. To calculate the change in p over 1 generation, we took the difference in bioRxiv preprint doi: https://doi.org/10.1101/103382; this version posted July 26, 2017. The copyright holder for this preprint (which was not certified by peer review) is the author/funder. All rights reserved. No reuse allowed without permission.

allele frequency over the 15 year period, and divided it by the total number of generations. For this, we
assumed 4 generations (Barber 1937) per year for each of the 15 years examined.

187

188 Illumina WGS Library Preparation and Sequencing

189 Genomic DNA from one pupa was submitted to the North Carolina State Genomic Sciences 190 Laboratory (Raleigh, NC, USA) for Illumina paired-end (PE) library construction and sequencing. 191 Prior to library preparation, the DNA template was quantified by a Qubit 2.0 Fluorometer (Invitrogen, 192 USA). The PE library with an 800bp insert size was constructed using an Illumina TruSeq Nano 193 Library Kit (Illumina, Inc. San Diego, CA) according to standard protocol. Following enrichment by 194 PCR, the library was checked for quality and final concentration using an Agilent 2100 Bioanalyzer 195 (Agilent Technologies, USA) with a High Sensitivity DNA chip before sequencing on an Illumina 196 HiSeq 2500 (100x2 paired end, rapid run).

197 Genomic DNA from a second pupa was used for mate-pair (MP) sequencing. Prior to library 198 preparation, whole genomic DNA was run out on a 0.5% agarose gel at 130v for 2 hours. Fragments 199 8kb or larger, as compared with Hyperladder I (Bioline USA Inc. Tauton, MA, U.S.A), were excised 200 from the gel and purified using a Zymoclean large fragment recovery kit (Zymo Research Corp. Irvine, 201 CA, U.S.A.). The DNA sample was submitted to the Michigan State University Research and 202 Technology Support Facility (East Lansing, MI, USA) for 8kb MP library preparation and sequencing. The DNA library was prepared using an Illumina Nextera Mate Pair Sample Preparation Kit according 203 204 to standard protocol. The library was validated using a Qubit dsDNA assay, Caliper LabChipGX 205 (Perkin Elmer, Waltham, MA, U.S.A.) and Kapa Library Quantification gPCR for Illumina Libraries. 206 The library was loaded on one lane of an Illumina HiSeq 2500 High Output flow cell and sequenced in 207 a 2x125bp paired-end format using HiSeq SBS version 4 reagents. For both PE and MP libraries, base

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calling was done by Illumina Real Time Analysis v1.18.64, the output of which was converted to FastQ
format with Illumina Bcl2fastq v1.8.4.

210

211 PacBio Library Preparation and Sequencing

212 Genomic DNA from 4 pupae, one of which was also used for Illumina PE sequencing, were 213 prepared into two libraries for PacBio sequencing. For each library, the SMRTbell Template 214 Preparation Kit version 1.0 (Pacific Biosciences, Menlo Park, CA, U.S.A.) was used for gDNA 215 preparation, but shearing and size-selection steps differed. For the first library, shearing was minimal 216 and no size selection was performed. For the second library, shearing prior to DNA concentration was 217 avoided to maximize fragment length, and a BluePippin (Sage Science Inc., Beverly, MA, U.S.A.) was 218 used to select fragments that were at least 7kb long. This produced sufficient prepared library material 219 for 17 and 5 SMRT cells, respectively. Prior to sequencing, the library concentration and fragment 220 length profiles were checked on a Qubit 2.0 and an Agilent Tapestation 2200 (Agilent Technologies, 221 USA) with a high molecular weight tape. Both libraries were sequenced at the University of North 222 Carolina Chapel Hill Sequencing facility on a PacBio RS II.

223

224 H. virescens Genome Assembly

Read quality was checked for all Illumina data using FastQC (Babraham Bioinformatics,
Cambridge, UK). Low quality ends were trimmed from both PE and MP reads using trimmomatic (v.
0.32; Bolger *et al.* 2014) and cutadapt (v. 1.9.1; Martin 2011), respectively. Any remaining Illumina
adapter sequences and Nextera transposon sequences were also removed. Reads were filtered for
potential microbial contaminants and *H. virescens* mitochondrial DNA (Supplementary Data File 1)
using BBmap (version 35.10; Bushnell B. - sourceforge.net/projects/bbmap/). For the full list of the

screened contaminants, see Supplementary Table 2. SOAPdenovo2 (v. 2.04) was used for assembly,
scaffolding and gap closure (Luo *et al.* 2012) with a k-mer length set to 63. Contigs and scaffolds over
2kb were used for further analysis.

RepeatScout (version 1.0.5; Price *et al.* 2005) was used to find *de novo*, species-specific repeats in the K63 assembly, while RepeatMasker (version open-4.0; Smit *et al.* 2013) was used to identify other common insect repeats available from Repbase (version 20150807; Jurka *et al.* 2005). We softmasked both repeat classes using BEDTools (version 2.25.0; Quinlan and Hall 2010) and collapsed redundant haplotypes using the default settings in Haplomerger2 (version 3.1; Huang *et al.* 2012). To fill intra-scaffold gaps, we applied PacBio reads over 5kb in length to our Illumina assembly using PBsuite (version 14.9.9; English *et al.* 2012).

241 As an assembly quality check, BlastStation (TM Software, Inc., Arcadia, CA, U.S.A.) was used to align 654 mapped ddRAD-seq marker sequences from the F1 parent used to produce an *H. virescens* 242 243 linkage map (Fritz *et al.* 2016; Dryad digital repository http://dx.doi.org/10.5061/dryad.567v8) to our 244 scaffolds. All top hits were exported and markers with alignment hit lengths greater than 150bp (of 350 245 bp total), identities greater than 80%, and e-values below 0.001 were further examined. This enabled 246 us to check for potential misassemblies, and provide additional information about which short scaffolds likely belong together on individual chromosomes (Supplementary Table 3). BlastStation was also 247 248 used to identify the scaffold to which the alpha-subunit of the *Vqsc* (GenBank Accession: AH006308.2) 249 aligned.

250

251 Structural annotation

252 The Just_Annotate_My_Genome (JAMg; https://github.com/genomecuration/JAMg) platform 253 was used to generate putative gene models. First, the genome was masked using RepeatMasker (Smit

254 et al. 2013) and RepeatModeler (Smit et al. 2013). Subsequently, RNA-Seq data was obtained from 255 NCBI for H. subflexa and H. virescens (SRA accessions: ERR738599, ERR738600, ERR738601, 256 ERR738602, ERR738603, ERR738604, ERR738605, SRR1021613), preprocessed using 257 "justpreprocessmyreads" (http://justpreprocessmyreads.sourceforge.net), and assembled with Trinity RNA-Seq 2.1.1 (Haas et al. 2013) using both the 'de-novo' and 'genome-guided' options as 258 259 implemented in JAMg. The platform made use of multiple lines of evidence to support each gene 260 model: the two Trinity RNA-Seq assemblies integrated with 63,504 publicly available Sanger-261 sequenced Expressed Sequence Tags using our new version of PASA (Haas et al. 2003); protein 262 domain annotation of putative exons via HHblits (Remmert et al. 2012); the de-novo gene predictors 263 GeneMark.HMM-ET (Lomsadze et al. 2014) and Augustus (Stanke et al. 2006) using the assembled 264 and raw RNA-seq and protein domain data as external evidence. These evidence tracks were condensed 265 to an Official Gene Set (OGS) using Evidence Modeler (Haas et al. 2008). A quantitative assessment 266 of our assembly and annotation completeness was conducted using BUSCO software using the 267 metazoan lineage setting (version 2.0.1; Simao *et al.* 2015).

268

269 H. virescens ddRAD-seq library preparation

DdRAD-seq libraries were prepared according to Fritz *et al.* (2016) with minor modifications.
Briefly, 200 ng of genomic DNA from the thorax of each field-collected specimen was digested with
EcoRI and MspI. Overhang sites from each specimen were ligated to Truseq Universal adapters
(Illumina, Inc. San Diego, CA) modified to contain a unique barcode (Elshire *et al.* 2011, Fritz *et al.*2016). Adapter-ligated DNA fragments from each individual were combined into pools of no more
than 24 individuals. A Blue Pippin (Sage Science, Inc., Beverly, MA) was used to select adapterligated DNA fragments ranging from 450-650 bp from each pool, and size-selected DNA pools were

277 amplified in a Peltier PTC200 thermalcycler under the following reaction conditions: 72 °C for 5min, 278 18 cycles of 98 °C for 30 sec, 65 °C for 20 sec, 72 °C for 30 sec, followed by 72 °C for 5 min. For 279 each pool, 1 of 4 Illumina indices (1,2,6, or 12) was added via PCR to the MspI adapter. Amplified 280 pools were combined, cleaned with a Qiaquick PCR Purification Kit (Qiagen, Inc., Valencia, CA, 281 U.S.A.), and diluted to 4nM prior to sequencing. Prepared genomic DNA libraries constructed from a 282 total of 177 H. virescens individuals were spread across four 2x300 paired-end Illumina MiSeq runs. 283 Individuals from each year and collection location were spread evenly across each MiSeq run to 284 minimize sequencing run bias in our downstream analysis.

285

286 Demultiplexing and Genome Alignment of DdRAD-seq Markers

287 Illumina-generated read 1 and 2 files were merged using FLASH version 1.2.7 (Magoc and 288 Salzburg 2011), then demultiplexed and filtered for quality using the process_radtags script from 289 Stacks version 1.09 (Catchen *et al.* 2011, 2013). Quality filtering entailed removal of reads when: 1) 290 they did not have an intact EcoRI cut site, 2) had a quality score < 30, or 3) were smaller than 350 bp. 291 We disabled the rescue reads feature in the process_radtag script, and therefore no read containing 292 errors in the barcode sequence was used for downstream analysis. All remaining merged reads were 293 truncated to a maximum length of 350 bp. Filtered demultiplexed reads were aligned to our *H*. 294 virescens genome assembly using Bowtie 2 (version 2.2.4; Langmead and Salzberg 2012). All reads 295 were aligned in end-to-end mode using the preset parameters with the highest sensitivity (--very-296 sensitive).

297

298 Association of DdRAD-seq Marker Genotypes with the Pyrethroid Resistance Allele

299 We first identified whether any raw ddRAD sequencing reads aligned to the scaffold containing

300 the *Vasc* using Integrative Genomic Viewer (IGV; Robinson *et al.* 2011). Following identification of 301 potential ddRAD-seq markers near the *Vqsc*, we inspected stacks of ddRAD-seq reads for individuals 302 with genotypic data at the pyrethroid resistance locus. Particular attention was paid to individuals that 303 were homozygous for the pyrethroid resistance allele. Through an IGV visual inspection of ddRAD-304 seq raw reads, we identified one 350bp ddRAD-seq locus (hereafter Hv_11322), for which a single 305 350bp sequence (hereafter Hv_11322_hap1) was commonly associated with the L1029H mutation at 306 the *Vqsc*. Filtered, genome-aligned reads from all specimens were then fed into the Stacks v. 1.09 307 (Catchen *et al.* 2011, 2013) pipeline for read clustering. Custom R and python scripts were used to call 308 350bp ddRAD-seq genotypes at Hv_11322 for all field-collected individuals, which were then 309 manually inspected and edited to include any insertions and deletions that were omitted by the Stacks 310 software. For purposes of genotype calling at Hv_11322, individuals with a read count of 6 or higher 311 for a single 350bp sequence were considered homozygotes, with two copies of that allele. Where 312 individuals carried fewer than 6 reads for a single 350bp sequence, their genotypes were scored as a 313 single copy of that observed allele plus one null allele. This threshold was chosen because individuals 314 with 6 or more reads can be called homozygotes with greater than 95% certainty (Buerkle and Gompert 315 2013).

We postulated that if the breadth of the "selective sweep" surrounding the *Vgsc* resistance allele included Hv_11322, such that Hv_11322_hap1 was associated with the L1029H mutation, the rates of their decline in frequency should be similar, if indeed there was a decline. We therefore examined whether the frequencies of the L1029H mutation and Hv_11322_hap1 differed in their rate of decline over time. Specifically, we used a series of nested generalized linear models with binomial error structures to examine whether locus and collection year interacted to influence individual allele (i). In the case of the Hv_11322 response, Hv_11322_hap1 was scored as a 1 and all other alleles were scored

323 as a zero. Our full statistical model was as follows:

324
$$\Pr(y_i = 1) = \operatorname{logit} {}^{-1}(\beta_{0i} + \beta_{1 \operatorname{Yeari}} + \beta_{2 \operatorname{Locusi}} + \beta_{3 \operatorname{Year} \times \operatorname{Locusi}}),$$

where Year represented the years during which the moths were collected and Locus indicated either the *Vgsc* or ddRAD-seq marker Hv_11322. As before, we identified a model term as statistically significant when a comparison of nested models by analysis of deviance indicated that removal of that term significantly influenced model deviance. No significant difference between a model with and without the interaction term indicated that the slope of the decline in the L1029H mutation was similar to that of Hv_11322_hap1.

332 We also analyzed the distribution of Hv 11322 hap1 for groups of individuals that were 333 homozygous for either the resistant or susceptible alleles at the Vgsc. In total, 32 individuals were 334 homozygous in our target region of the *Vqsc* and contained sufficient ddRAD-seq data at nearby locus 335 Hv_11322 to call at least one allele. Of these 32 individuals, two Hv_11322 alleles could be called for 336 26 individuals, whereas only a single allele could be confidently called for 6 of the individuals due to 337 their lower than 6X depth of coverage. In total, 58 haplotypes (from 32 individuals), which contained 338 genotypic information for both the *Vgsc* and the nearby ddRAD-seq marker were examined. A Fisher's 339 exact test of independence was used to determine whether there was an association between the 340 frequencies of Hv_11322_hap1 and the L1029H mutation.

341

342 H. virescens ddRAD-seq Enabled Genome Scan

Previous work by Groot *et al.* showed that genetic differentiation among North American
populations of *H. virescens* was low (Groot *et al.* 2008). Given these already documented high levels
of gene flow between collection sites, and our goal to detect genomic change over time, we specifically

346 focused on analyzing allele frequency changes between years. Samtools (version 0.1.18; Li *et al.* 347 2009, Li 2011) view was used to convert SAM files output by Bowtie 2 to BAM files, and SNPs were 348 called using mpileup. BCFtools was used to generate SNP and indel genotypes, as well as genotype 349 likelihoods in a Variant Call Formatted (VCF) file. This VCF file was filtered by VCFtools (version 350 0.1.15, Danacek et al. 2011, <u>https://vcftools.github.io</u>) prior to downstream population genomic 351 analysis. The filtered dataset included loci that: 1) were sequenced to a depth of 3 or more reads, 2) 352 had a minor allele frequency of 0.1 or greater, 3) were represented in at least 50% of individuals, and 4) 353 included only SNP variant sites (indels were excluded). The number of SNPs was thinned such that no 354 more than 2 were examined per ddRAD-seq locus. This thinned SNP dataset was transformed from 355 VCF format to genepop format using PGDSpider (version 2.1.1.0, Lischer and Excoffier 2012). 356 Pairwise SNP outlier analyses of these thinned SNP datasets were made using Lositan (Antao et al. 357 2008) with the following parameter settings: 1) "neutral" and "forced" mean FST settings were 358 engaged, 2) the Infinite Alleles model was assumed, 3) the false discovery rate was set to 0.1, and 4) 359 the type I error was set to 0.01.

360 Scaffolds containing SNP markers with statistically significant pairwise-genetic divergence 361 were identified for further analysis (Supplementary Data File 2). Using the physical distance between 362 the ddRAD-seq locus Hv_11322 and the *Vqsc* as our guide, the predicted structural genes on each 363 scaffold identified in our Supplementary Data File 2 and found within 36 kb of each SNP outlier were 364 examined. Protein sequences corresponding to each annotation along a scaffold where divergent 365 markers were present were aligned to the NCBI Arthropod database (taxid: 6656) via blastp using 366 Blast2GO software. Gene ontology (GO) assignments for each predicted gene sequence within 36 kb 367 of an outlier locus are also provided in Supplementary Data File 3.

368 Given the large number of outlier loci (Figure 3), and that the potential traits under selection in

369 this agricultural system (e.g. metabolic detoxification of insecticides and/or host plant defensive 370 compounds, host volatile detection) are often quantitative, we reasoned that certain gene families may 371 be over-represented near our outliers with respect to their overall distribution throughout the *H*. 372 virescens genome. Therefore, we examined the distribution of GO categorizations (54 level two 373 categorizations in each of three GO domains: "Biological Process", "Molecular Function", and 374 "Cellular Component") for the subset of predicted genes found near outlying SNPs for each by-year 375 comparison. We compared those distributions to the numbers of genes found in each of these same 376 categories in the overall *H. virescens* genome using a series of Fisher's Exact tests. Due to the large 377 number of comparisons (n = 54), we used a Bonferroni-corrected alpha value of 0.0009 to establish 378 statistically significant over-representation in any one GO category. For each by-year comparison, we 379 examined the subsets of genes within 2 different intervals from SNP outliers: 10 kb ("moderate" 380 linkage according to Lowry et al. 2017) and 36 kb (extended linkage, based upon the distance between 381 Hv_11322 and the *Vgsc* target of selection).

382

383 Strength of Selection on Outlier Loci

384 Finally, we examined what the strength of selection must have been to produce the observed 385 allele frequency changes at all identified outlier loci for each by-year comparison. We calculated the 386 coefficient of selection (s) against q, the declining allele frequency, assuming either dominance of p or 387 co-dominance of p and q according to Falconer and Mackay (1996). Custom scripts written in R were used to calculate *s* for all outliers with the exception of those with initial values of *p* that were very low 388 389 (p > 0.05). These were excluded because they are most susceptible to under-sampling due to our small 390 sample sizes, and small biases in these values have the potential to significantly influence the selection coefficient. 391

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392

- 393 Results
- 394 H. virescens Candidate Gene Analysis

In 1997, the frequency of the L1029H mutation was 0.66 in LA and 0.63 in TX. By the year

396 2012, the frequency of this resistance allele declined to 0.44 in LA and 0.36 in TX (Figure 1). This

397 decline in the resistance allele frequency over our 15 year sampling period was statistically significant

398 (p < 0.001). Neither the interaction between year and season (p = 0.36), season itself (p = 0.21), nor

399 sampling location (p = 0.25) significantly influenced the frequency of the resistance allele.

400

401 Strength of Selection against the Vgsc Resistance Allele

In LA, the frequency of the resistance allele declined by 0.22 over the 15 years between 1997 and 2012. In TX, the decline in the resistance allele frequency was 0.27. Therefore, the increase in p per generation was 0.004 and 0.005 for LA and TX, respectively. With this information, we were able to calculate a selection coefficient of 0.03 for each of the LA and TX populations.

406

407 Genome Sequencing and Assembly

In total, the Illumina sequencing runs produced 122,433,923 and 232,607,659 reads for PE and MP libraries, respectively. After read trimming and filtering, 115,374,414 and 227,857,423 reads from the PE and MP libraries were used for assembly. An additional 482,464 PacBio reads with an average length of 7560 (s.d. = 2663) bp were applied to our Illumina assembly using PBsuite software for gap filling. The final *H. virescens* genome assembly was comprised of 8,826 scaffolds with a total length of 403,154,421 bp, similar to the previously estimated *H. virescens* genome size of 401 Mbp (Gregory and Herbert 2003). The scaffold N50 was 102,214 bp (mean size = 45,678 bp; range = 659 – 628,964 415 bp). A BUSCO analysis of our final assembly indicated that 865 (88%) of the 978 core conserved 416 eukaryotic genes were complete. Further specifications for our genome assembly can be found in 417 Supplementary Table 4. When we examined our previously mapped *H. virescens* ddRAD-seq markers 418 (Fritz *et al.* 2016), a total of 562 out of 654 met the aforementioned alignment criteria relative to the 419 reference genome and were used to examine and group scaffolds into chromosomes. Of these 562 420 markers, 557 aligned uniquely to a single scaffold, while 5 markers (4851, 5891, 13906, 22644, 29612) 421 aligned well to multiple scaffolds. This suggested that either those scaffolds were allelic, or that the 422 marker sequences contain repetitive DNA. Four-hundred eighty three of the 8826 scaffolds present in 423 our assembly were aligned to at least 1 mapped marker. In most cases (n = 421 scaffolds), a single 424 scaffold was associated with a single mapped marker. However, 62 scaffolds could be aligned to 425 multiple mapped markers, which enabled us to check the quality of our assembly against our linkage 426 map. Of these 62 scaffolds, only 5% (3 scaffolds) aligned to markers that originally mapped to 427 different linkage groups. A summary of the scaffold names and groupings by linkage group can be 428 found in Supplementary Table 3. One scaffold, numbered 4600, contained the entire *Vqsc* sequence 429 available from GenBank accession AH006308.2.

430

431 Association of DdRAD-seq Marker Genotypes with the Pyrethroid Resistance Allele

We located one ddRAD-seq marker, called Hv_11322, spanning bp 11,397 through 11,747 of Scaffold 4600 which is *ca*. 37 kb upstream from the pyrethroid resistance locus. In total, 55 unique Hv_11322 sequences (alleles) could be identified from 138 individuals using Stacks. Fifty of these 55 alleles were found fewer than three times in our field-collected populations. The remaining five most common alleles (Supplementary Figure 4) were found 5, 5, 6, 12, and 161 times, respectively. According to an NCBI blast, all of these sequences aligned well with an *Helicoverpa armigera*

438 sequence (GenBank accession DQ458470.1) that also contained the *Vgsc*.

439 No statistically significant difference existed between the slope of the decline in the L1029H mutation and Hv_11322_hap1 (deviance = -0.355, df = 1, p = 0.551). These allele frequency declines 440 441 are plotted in Figure 1. When we examined the full haplotypes (e.g. containing both the Hv_11322 442 locus and the pyrethroid resisance locus), specifically in homozygotes at the pyrethroid resistance 443 locus, we identified 20 unique Hv_11322 alleles in the 32 total individuals (or N = 58 chromosomes). 444 When chromosomes containing the L1029H mutation (pyrethroid resistance-conferring *Vqsc* allele) 445 were examined, 85% (29 of 34) also carried the ddRAD-seq allele Hv_11322_hap1 (Figure 2). The 446 remaining 5 chromosomes bearing the L1029H mutation contained 4 unique Hv_11322 alleles, none of 447 which were the 5 most common Hv_11322 alleles. Only 5% (1 of 24) of the chromosomes bearing the 448 wild-type *Vgsc* allele also carried Hv_11322_hap1. Fifteen unique Hv_11322 alleles of the 20 alleles 449 found in homozygotes were associated with the wild-type *Vgsc* allele. A Fisher's Exact test indicated 450 that there was a statistically significant association between the presence of Hv_11322_hap1 and the 451 L1029H mutation (p < 0.001).

452

453 H. virescens ddRAD-seq Enabled Genome Scan

Of the 1,682,114 SNPs in our ddRAD-seq marker dataset, the total number of filtered SNPs
included in the analyzed dataset was 8,963. Based upon this filtered dataset, overall population
genomic divergence between years was low. Pairwise Weir and Cockerham's overall FST values were
0.005, 0.004, and 0.013 for the comparisons 1997-2007, 1997-2012, and 2007-2012, respectively.
We first examined SNPs along Scaffold 4600, where the *Vgsc* was located, for evidence of
genomic divergence between years. Between the years 1997 and 2012, one SNP at position 11706 on
Scaffold 4600, which was part of Hv_11322, showed signs of statistically significant genetic

divergence with a Weir and Cockerham's FST value of 0.205 (p < 0.001). This SNP, a cytosine to
thymine transition, went from a cytosine allele frequency of 0.875 (n = 44 diploid individuals) in 1997,
fell to a frequency of 0.724 in 2007 (n = 49) and further declined to a frequency of 0.548 in 2012 (n =
52; Fig. 1).

465 In addition to this FST outlier on Scaffold 4600, our genome scan revealed a number of additional diverging allele frequencies for each by-year comparison. In total, we detected a total of 351 466 467 SNPs on 314 scaffolds (3.6% of the 8,826 total scaffolds) as outliers in at least one by-year 468 comparison. Table 1 shows the number of genomic outliers for each comparison, as well as the number 469 of unique scaffolds on which these unique outliers were found. Between the years 1997 and 2007, 201 470 SNPs (2.2% of the 8,963 total SNPs examined) showed signs of significant allele frequency 471 divergence, whereas only 35 SNPs (0.4% of the 8,963 total SNPs) significantly diverged between the years 2007 and 2012. When a comparison was made over the total time period, between the years 1997 472 473 and 2012, 184 SNPs showed signs of statistically significant allelic divergence. 474 In some cases, the same SNPs showed signs of divergence in two different by-year 475 comparisons, but no SNP outliers were shared among all 3 comparisons (Supplementary Figure 5). Fifty-three SNPs were considered outliers between the years 1997 and 2007, and again between 1997 476 477 and 2012. This indicated that significant allele frequency changes occurred between the years 1997 478 and 2007, followed by stability or small, non-significant allele frequency changes through the year 479 2012. Likewise, 5 SNPs showed significant allele frequency divergence between the years 1997 and 2012, and again between the years 2007 and 2012, indicating that allele frequencies were stable, or 480 481 underwent modest changes between the years 1997 and 2007 followed by a significant change in the 482 year 2012.

483

When we examined the 36 kb of DNA flanking the SNP outliers on these 314 scaffolds (72 kb

total), the mean number of putative genes identified in these broader genomic regions was fewer than 3,
and the maximum was 12 (see Table 1 for a complete breakdown by comparison). A complete list of
outlier SNPs and their genome positions for each by-year comparison, as well as the putative genes and
their GO assignments within the 36 kb window surrounding these SNPs can be found in Supplementary
Data files 2 and 3, respectively.

Some outliers were found near putative genes with functions related to either insecticide 489 490 resistance or changes in host use. For example, one predicted gene sequence within 10kb of two 491 outliers, JAMg_model_7841, aligned to a cytochrome p450 protein sequence (*Cyp6AE12*) from 492 *Helicoverpa armigera* (GenBank Accession AID54888.1) with a 100% query cover and 83% identity. 493 It was identified on Scaffold 3424 and the associated SNPs were detected as outliers between 1997 and 494 2012, and also between 2007 and 2012. As a second example, two predicted gene sequences, 495 JAMg_model_4652 and JAMg_model_4653 aligned with at least 70% query cover and 78% identity to 496 Athetis lepigone olfactory receptor (OR60; GenBank Accession KT588155.1) and Helicoverpa assulta 497 olfactory receptor (OR33; GenBank Accession KJ542684.1) sequences, respectively. These gene 498 sequences were found within 10kb of a SNP outlier from 1997-2007 and 1997-2012 comparisons and 499 was located on Scaffold 2173. We examined the ontology of genes near SNP outliers, with an eye 500 toward those predicted to be important for detoxification and behavior, to identify whether any GO 501 category was over-represented near outlier SNPs relative to the overall genome.

GO assignments by Blast2GO were only achieved for a subset of predicted genes. The
proportions of putative genes within 36 kb of outliers for which a function could be predicted were
0.37, 0.38, and 0.32 for the by-year comparisons 1997-2007, 1997-2012, and 2007-2012, respectively.
While low, these were similar to the proportion of gene predictions assigned a function by Blast2GO in
the overall genome assembly (0.40). When Fisher's Exact tests were applied to identify whether any

level 2 GO term was over-represented in either 10 or 36kb windows near outlier SNPs from each byyear comparison, no significant difference was found relative to their distribution in the overall
genome.

510

511 Strength of Selection on Outlier Loci

512 Selection coefficients (*s*) were calculated for 164, 150, and 31 outliers for the following by-year 513 comparisons, respectively: 1997 and 2007, 1997 and 2012, and 2007 and 2012. For selection against q, 514 the declining allele, s ranged from 0.009 - 0.301 across all by-year comparisons when dominance of p 515 was assumed. When co-dominance of p and q was assumed, selection coefficients ranged from 0.012 516 to 0.623. The mean selection coefficients (± standard deviation) calculated across outliers were 517 0.047(0.020), 0.077 (0.029) and 0.165 (0.062) assuming dominance of p for the 1997 - 2012, 1997 -518 2007, and 2007 - 2012 by-year comparisons, respectively. Mean selection coefficients assuming 519 codominance of p and q were 0.058 (0.037), 0.088 (0.058), and 0.200 (0.150) for these same time 520 periods. In general, SNPs with higher selection coefficients were associated with greater rates of 521 change in the frequency of q, as well as higher initial starting values of q (Figure 4).

522

523 Discussion

Double-digest RAD-seq and other NGS marker-development methods have been used to detect signatures of local adaptation in a number of non-model plant and animal species (e.g. Hohenlohe *et al.* 2010, Nadeau *et al.* 2014, Pujolar *et al.* 2014, Ruegg *et al.* 2014, Pais *et al.* 2016). Here we have demonstrated the power of ddRAD-seq to identify genomic regions that have diverged over short evolutionary time scales (fewer than 2 decades) in a landscape characterized by human-induced environmental change. We postulated that widespread adoption and cultivation of Bt cotton in the 530 Southern United States would likely impose strong selection on Lepidopteran herbivore and cotton 531 pest, *H. virescens*, through shifts in host plant composition and insecticide use. We first identified 532 allele frequency changes at a likely gene target of selection, the *Vqsc*, in field-collected populations of 533 *H. virescens.* We then demonstrated that this change could be detected using a nearby ddRAD-seq 534 marker. Allele frequencies at many other regions of the *H. virescens* genome also diverged over time, 535 likely in response to selection pressures imposed by widespread adoption of Bt cotton. We calculated 536 selection coefficients for SNPs that were detected as having changed significantly over time, to 537 demonstrate the strength of selection encountered by organisms found in agricultural ecosystems. 538 Furthermore, we sequenced and assembled an *H. virescens* genome to help us identify potential 539 structural genes involved in adaptation to agricultural inputs, and made it publicly available at NCBI. 540 Our initial examination of the *Vgsc*, a candidate gene likely to be impacted by the decline in 541 pyrethroid use that followed Bt cotton adoption, demonstrated that the resistance-conferring L1029H 542 mutation declined in frequency over time. Indeed, examination of allele frequency changes at the Vasc 543 locus yielded a selection coefficient of 0.03. This indicated that the resistance allele was moderately to 544 strongly deleterious in field populations where pyrethroid pressure was low (Eyre-Walker and Keightley 2007, Kim et al. 2017). A selection coefficient of this magnitude seemed reasonable given 545 546 the previously reported fitness cost associated with carrying this resistance allele (Zhao *et al.* 2000). 547 We were surprised to see the frequency of the resistance allele plateau in the year 2007 and remain at 548 *ca.* 0.4 through the year 2012, however. There are several possible explanations for this. One 549 explanation is that pyrethroid pressure has declined but remains sufficiently high such that maintenance 550 of the resistance allele in field populations is advantageous, in spite of the fitness cost to individuals 551 that carry it (Zhao *et al.* 2000). Alternatively, as the resistance allele frequency declines there are relatively fewer homozyogous resistant genotypes. If the fitness cost is only associated with 552

homozygotes then the decline in resistance allele frequency could level off, even in the absence ofpyrethroid selection.

555 Using a ddRAD-seq dataset, we identified one marker that aligned to our reference genome 37 556 kb upstream of the *Vqsc*. One allele of this 350bp marker, called Hv_11322_hap1, was associated with 557 the L1029H mutation that confers pyrethroid resistance. This suggested that Hv_11322_hap1 was in 558 linkage disequilibrium with the L1029H mutation. Furthermore, the breadth of the selective sweep in 559 this genomic region extended at least 36 kb on one side of the *Vgsc*. Upon further examination of 560 Scaffold 4600, which contains this region under selection, we identified three cytochrome p450s that 561 are found between Hv_11322 and the *Vqsc*. This confirmed previous reports of tight physical linkage 562 between the Vgsc and Cyp6B10 in H. virescens (Park and Brown 2002). It is possible that these 563 cytochrome p450s could also be targets of selection by pyrethroid insecticides, and future work could 564 be directed at whether or not they play any roles in the expression of pyrethroid resistance phenotypes. 565 Work in another closely-related Lepidopteran species, *H. armigera*, suggests that this particular 566 cytochrome p450 is not involved in pyrethroid resistance, however (Grubor et al. 2007). 567 SNP data from our ddRAD-seq marker Hv_11322 enabled us to rediscover changes at the *Vgsc* associated with the L1029H mutation over time. While the SNP outlier in the Hv_11322 marker 568 569 demonstrated significant allelic divergence relative to the genome-wide average FST value, additional 570 SNP outliers on other *H. virescens* scaffolds were also detected. In spite of the fact that we applied a 571 correction to reduce detection of false positives, it is possible that up to 35 of our total 351 SNP outliers are false positives given our false discovery rate of 0.1. This correction threshold was selected to 572 573 minimize false positive detection, while retaining true positives (Verhoeven *et al.* 2005). To safeguard 574 against pursuing potential false positives, further research could initially focus on genes near outliers with significant allele frequency changes in multiple by-year comparisons. Fifty-three SNPs showed 575

576 significant allelic divergence across multiple by-year comparisons, where major allele frequency shifts 577 took place between the years 1997 and 2007. Between 2007 and 2012, 5 SNPs showed significant 578 shifts in allele frequency and these changes were detected in multiple by-year comparisons as well. It 579 is likely that these SNP markers are in linkage disequilibrium with gene targets of selection as 580 management of cotton ecosystems has led to the replacement of conventional cotton cultivars with Bt-581 expressing varieties. Interestingly, two of these SNPs were found on scaffolds containing genes with 582 plausible roles in detoxification (of insecticides and plant defensive compounds) or host plant 583 detection.

584 Scaffold 3424 contained SNPs that diverged significantly over time in our field-collected 585 populations of *H. virescens*. Divergence was strongest in by-year comparisons from 1997-2007, and 586 1997-2012. This suggests that most genomic change occurred between the years 1997 and 2007, and that allele frequencies remained stable between the years 2007 and 2012. Blast results for predicted 587 588 gene sequences found on this scaffold revealed homology with the cytochrome p450 superfamily. The 589 predicted sequence aligned well with an *H. armigera Cyp6AE* gene, which is a cytochrome p450 590 family known to be involved in detoxification (Zhou *et al.* 2010). The best alignment was to an *H*. 591 *armigera* Cyp6AE12, and expression levels of this gene in *H. armigera* are modified in response to 592 pyrethroid insecticides (Yue *et al.* 2007, Zhou *et al.* 2010). It is possible that allelic changes on this 593 scaffold are a response to reduced pyrethroid use in the Southern United States as a result of Bt cotton 594 deployment in the agricultural landscape.

An alternative explanation exists for changes in allele frequencies at this candidate gene, however. *Cyp6AE12* expression is also modified in response to the plant compound xanthotoxin (Zhou *et al.* 2010). Another *Cyp6AE* monoxygenase, *Cyp6AE14*, has 68% protein sequence similarity to *H*. *armigera Cyp6AE12*, is highly expressed in the midgut of *H. armigera*, and is likely involved in

599 detoxification of the cotton defensive compound gossypol (Mao *et al.* 2007). Indeed, RNAi silencing 600 of *Cyp6AE14* in *H. armiqera* led to a decline in larval growth when gossypol was present in their diet 601 (Mao *et al.* 2007). If this *H. virescens* cytochrome p450 is indeed the target of selection, it is possible 602 that the divergence in allele frequencies between the years 2007 and 2012 was the result of selection for 603 improved larval performance in *H. virescens* populations that feed on alternative hosts. When 604 widespread planting of Bt-expressing cotton drove *H. virescens* off of their previously abundant cotton 605 host to alternative host plants, *H. virescens* would have been exposed to new plant defensive 606 compounds. Genes associated with larval performance in one host plant genera may or may not be 607 associated with improved larval performance on other host plant genera (Sheck and Gould 1993). 608 Therefore, allelic changes near our *H. virescens* SNP outlier may be caused by relaxation of selection 609 for gossypol detoxification as they moved out of cotton, or adaptation to new host plants and their 610 antifeedant chemicals. To determine whether phenotypes resulting from these molecular shifts are 611 directly associated with changing *H. virescens* management practices (e.g. pyrethoid or Bt toxin use) or 612 shifts in host plant use, further work could involve measuring associations between Bt or pyrethoid 613 response phenotypes and genotypes at target genes on these scaffolds.

614 Equally as important to host shifts is the ability of adult females to find a suitable host for 615 oviposition. Chemosensation is important for host plant identification in phytophagous insects, and 616 genes involved in olfaction and gustation are often targets of selection when host shifts occur (Dworkin 617 and Jones 2009, Smajda et al. 2012). An outlier SNP was detected on H. virescens Scaffold 2173 that 618 underwent allele frequency changes between the years 1997 and 2007, then remained stable between 619 the years 2007-2012. Within 10 kb of this outlier was a pair of olfactory receptor genes. It is plausible 620 that allele frequency changes at these olfactory receptor genes may reflect changes in the olfactory 621 percept, enabling females to identify non-cotton host plants. Future work directed at elucidating the

phenotypic effects of changes at these genes will be critical to determining their role, if any, in hostplant detection.

624 Both toxin metabolism and chemosensation can be complex traits with multiple loci involved. 625 Arrays of duplicated cytochrome p450s (Li *et al.* 2002, David *et al.* 2013) and carboxylesterases (Field 626 et al. 1988, Guillemaud et al. 1997) have been implicated in metabolic detoxification of chemicals in 627 other insect species. Likewise, clusters of olfactory and gustatory receptors are thought to contribute to 628 host-plant utilization in some insect species (Smajda et al. 2012). Therefore, we examined GO level 2 629 categorizations, particularly those related to detoxification and chemosensation, for the putative genes 630 surrounding our SNP outliers to determine if any GO categories were over-represented near our outliers 631 relative to the rest of the genome. Our results did not suggest there was any statistically significant 632 over-representation of any GO category near our outliers for any of our by-year comparisons. There 633 are at least two possible explanations for this: 1.) The targets of selection linked to our SNP outliers do 634 not necessarily involve duplicated, amplified, or arrayed genes of similar function. Instead these 635 targets may involve single copy genes or gene regulatory regions associated with adaptive phenotypes 636 in *H. virescens*. 2.) Only 30-40% of the putative genes in our genome were assigned a function in our 637 annotation pipeline. While our *H. virescens* draft assembly was instrumental to identifying selection at 638 the *Vasc*, as well as identifying novel genomic regions under selection in field-collected populations, 639 there is room for improvement. Increasing the contiguity of our assembly will enable chromosome-640 level identification of runs of homozygosity, a hallmark of selection. Likewise, manual curation efforts 641 will improve official gene set and the likelihood of identifying the gene targets near our SNP outliers. 642 While the gene targets of selection during these time periods require further validation, the 643 strength of selection near these targets, as measured by *s*, ranged from 0.009 and 0.623, depending upon by-vear comparison and degree of dominance assumed. Following theoretical expectations, 644

645 selection against q required higher selection coefficients when initial frequencies of q were high 646 (Figure 4). Furthermore, degree of dominance impacted s, where values ranged from 14-23% higher 647 when co-dominance was assumed, rather than dominance of p. Based upon theoretical expectations, 648 the coefficient of selection against q should be greater under assumptions of a recessive p and a 649 dominant q. Indeed, examples of selection against a dominant susceptible allele (q), resulting in the 650 increase in frequency of a recessive resistance allele (p) have been described in the insecticide 651 resistance literature (for example, see Ferré and Van Rie 2002). Across all outlier SNPs discovered 652 within each by year comparison, 1997-2012, 1997-2007, and 2007-2002, the average selection 653 coefficients were 0.047 and 0.058, 0.077 and 0.088, and 0.165 and 0.200, respectively, depending on 654 degree of dominance, and these are considered to be moderate to high values of *s* according to Eyre-655 Walker and Keightley (2007), and Kim *et al.* (2017). Ultimately, these values of *s* demonstrate the 656 strong selection imposed on species found in agricultural ecosystems.

657 In conclusion, we demonstrated that ddRAD-seq enabled genomic scanning can be used to identify organismal responses to anthropogenic changes in agricultural ecosystems, even on short, 15 658 659 year time scales. We then identified additional genomic regions in this Lepidopteran species that are 660 likely changing in response to shifts from conventional cotton planting to widespread Bt-cotton adoption. From an applied perspective, our results suggest that ddRAD-seq genome scans may be 661 662 useful for monitoring pest populations for real-time changes in allele frequencies at loci responding to 663 the very strong selection imposed by insect management practices. We conclude that this technology 664 could be useful for identifying strong selection for resistance alleles, and then acting to mitigate 665 widespread phenotypic resistance to management practices across plant and insect species in 666 agricultural ecosystems.

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667 Data Availability

- 668 Scripts and configuration files used for genome assembly can be found at:
- 669 https://github.com/mcadamme/Hv_Genome_Assembly_Draft1

670

- 671 Scripts used for population genomic analysis can be found at:
- 672 https://github.com/mcadamme/FieldHv_Pop_Genomics

673

674 Raw sequence ddRAD-seq data have been deposited in the NCBI SRA as: (updated following

675 acceptance of the manuscript)

676

677 Our *H. virescens* draft 1 assembly and associated read sequences have been deposited in the NCBI
678 database under BioProject number PRJNA379496.

679

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1013 **Tables**

- 1014 **Table 1** Number of SNPs with pairwise FST values deemed as significantly divergent according to
- 1015 Lositan analysis for each by-year comparison. The number of unique scaffolds (of 8826) containing at
- 1016 least 1 significantly diverged SNP, as well as the average and maximum numbers of putative genes
- 1017 within 36 kb of the outlier SNP is also included.

Pairwise Comparisons	Total SNPs	Number Outlier SNPs	Number Scaffolds Containing Outliers	Mean Number Putative Genes within 36 kb of Outlier SNPs	Max Number Putative Genes within 36 kb of Outlier SNPs
1997-2007	8963	201	190	2.6	12
2007-2012	8963	35	33	2.6	10
1997-2012	8963	184	170	2.7	10

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1020 Figures

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Figure 1 – The decline in the frequency of the pyrethroid resistance allele in *H. virescens* (pooled LA and TX samples), represented by the dashed red line, was statistically significant over the course of our 15 year sampling period (n = 659). A unique ddRAD-seq haplotype, represented by the solid black line, was found *ca*. 36Kb upstream from the alpha subunit of the *Vgsc* and also declined in frequency in the subset of individuals (n = 141) sequenced for our genome scan. Error bars represent bootstrapped 95% confidence intervals (N = 5000) around the mean of each year.

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1030 **Figure 2** – Significantly greater diversity was observed from ddRAD-seq haplotypes linked to the 1031 susceptible *Vqsc* allele relative to those linked to the resistance allele. (A) is a visualization of the 1032 relationship of Hv 11322 to the L1029H *Vasc* SNP along an 81kb genome scaffold. The different 1033 colored bars at 11kb represent unique alleles at the Hv_11322 locus, showing a greater diversity of Hv_11322 alleles were found to be associated with the *Vqsc* wild-type allele, relative to the resistance 1034 1035 allele as would be indicative of a selective sweep. Due to the number of unique Hv_11322 alleles 1036 found to be associated with each *Vasc* allele (n = 5 and n = 15 for resistant and wild-type, respectively), 1037 full representation of this diversity could not be incorporated into (A). However, the unique colors in 1038 (B) depict the number and proportion of Hv_{11322} alleles linked to the *Vqsc* resistant (n = 34) and 1039 susceptible (n =24) SNP alleles. Each unique color represents a different Hv_11322 allele. The dark 1040 blue wedges always represent Hv_11322_hap 1.



1041 **Figure 3** – Pairwise genetic divergence according to Weir and Cockerham's FST for populations of *H*. 1042 *virescens* collected in the years A) 1997 and 2012, B) 1997 and 2007, C) 2007 and 2012. Each black 1043 point represents one SNP of 8,963 along the *H. virescens* genome. Points above the red line represent 1044 loci with pairwise genetic divergence that is statistically significant at the α = 0.01 level following 1045 correction for false discovery. Pairwise genetic divergence at the SNP near the *Vgsc* on Scaffold 4600 1046 is represented by the red point on panel A.



Figure 4 – Selection coefficients associated with SNP outliers where the initial frequency of p was
greater than 0.05. Selection against q is plotted against the rate of allele frequency change and the
initial starting frequency of q. Colors for each plotted point help to visualize the initial starting
frequency of q, where red is low and black is high. Plots in rows A, B, and C represent selection
coefficients calculated assuming dominance and codominance of p for each of the by-year comparisons
1997-2007, 2007-2012, and 1997-2012, respectively.