Title: The collision of two genomes threatens global food security

Authors: Henry L. North *[1,2], Gabriela Montejo-Kovacevich *[1,3,4], Douglas Amado *[1,5], Ian

A. Warren [1], Marek Kucka [6,7], Abby Williams [1,8], Yingguang Frank Chan [6,7], Tom Walsh [9],

Alberto Soares Corrêa [1,5], Celso Omoto [5], Chris D. Jiggins ‡[1]

* These authors contributed equally to this work

‡ Materials and correspondence: cj107@cam.ac.uk

Affiliations:

1 Department of Zoology, University of Cambridge, Downing Street, Cambridge CB2

3EJ, UK

2 Girton College, University of Cambridge, Huntingdon Road, Cambridge CB3 0JG UK

3 Evolutionary Biology Centre, Department of Ecology and Genetics, Science for Life

Laboratory, Uppsala University, Uppsala, Sweden

4 King's College, University of Cambridge, King's Parade, Cambridge, CB2 1ST UK

5 Department of Entomology and Acarology, Escola Superior de Agricultura Luiz de

Queiroz, Universidade de São Paulo, Av Padua Dias 11, 13418-900 Piracicaba, SP,

Brazil

6 Friedrich Miescher Laboratory of the Max Planck Society, 72076 Tübingen, Germany

7 Groningen Institute for Evolutionary Life Sciences (GELIFES), University of Groningen

9747AG Groningen, The Netherlands

8 Present Address: Department of Biology, University of Oxford, South Parks Road,

Oxford OX1 3EL

9 CSIRO Environment, Black Mountain Innovation Precinct, Clunies Ross St., Acton,

2601, ACT, Australia.

Summary paragraph

Human activity alters selection pressures and redistributes biodiversity, creating

opportunities for hybridization when different species come into secondary contact. It

is now well-established that ancient hybridization plays an important role in adaptive

radiation [1], but its role in recent anthropogenic adaptation remains unclear. We

investigated this in a study of hybridizing native and invasive *Helicoverpa* moths in

Brazil, which are among the most economically damaging crop pests globally. Native *H.*

zea has recently evolved Bt resistance and is better adapted to maize as a host plant,

while invasive *H. armigera* is resistant to pyrethroid pesticides and is better adapted to

soybean. Using a decade-long time series of 975 genomes from both species, we

demonstrate rapid, bidirectional adaptive introgression of structural variants that cause

resistance to two distinct pesticides. In one direction, a pyrethroid resistance gene from

H. armigera (*CYP337B3*) is now almost fixed in native *H. zea*. In the other direction,

~30% of recently-sampled invasive *H. armigera* carry a *H. zea*-derived trypsin repeat

cluster. Using controlled crosses, we show that this trypsin repeat cluster confers

resistance to Bt (Cry1Ac) soy in the *H. armigera* genomic background. Thus,

hybridization has combined *H. armigera* adaptation to soy with H. zea Cry1Ac

resistance to facilitate exploitation of Brazilian Bt soy monoculture — a vast

anthropogenic habitat exceeding the area of Germany. Our results show that the

combinatorial mechanisms that produce biodiversity can also drive rapid

anthropogenic adaptation on ecological timescales, with consequences for global food

security.

Main text

Introduction

Human activity imposes natural selection at unprecedented scales through changes in land use, climate, and chemical exposure. At the same time, humans are redistributing biodiversity across the globe, resulting in the introduction of non-native species to novel ecological conditions. As a consequence, biological invasion events — where ecological opportunists establish and spread in new environments — are occurring at an accelerating rate [2]. As species' distributions shift, once-isolated lineages can come into secondary contact and may hybridize. This confluence of strong natural selection and admixture can create conditions for evolutionary rescue, with far-reaching implications from conservation to food security [3,4].

It is now well-established that introgression between previously isolated lineages can produce novel hybrid trait combinations that drive adaptive diversification over thousands or millions of years [1]. Central to this 'combinatorial mechanism' is the amalgamation of modular genetic architectures into new configurations that facilitate the occupation of new niche space. Since admixture introduces adaptive variants at high frequency, which have already been optimised by selection in parental linages, adaptation can proceed rapidly. For example, hybridization brought together two wing colour developmental switches selected in independent *Heliconius* butterfly linages, resulting in a novel recombinant phenotype underlying adaptation and reproductive isolation [5]. Similarly, novel traits required for ecological divergence in *Helianthus* sunflowers resulted from a combination of alleles that arose in distinct ancestral

lineages [6]. A handful of studies have shown that anthropogenic activity can lead to

unidirectional adaptive introgression [7–10]. Given the role of combinatorial evolution in

ancient adaptive diversification, this begs the question: could similar mechanisms drive

adaptation at even shorter timescales relevant to anthropogenic change? Cases of

secondary contact between native and invasive species are particularly tractable

systems in which to investigate this question.

Secondary contact between previously isolated noctuid moth species of the genus

Helicoverpa present a remarkable natural experiment in rapid adaptation. In 2013

Helicoverpa armigera was detected as an invasive species in Brazil, where it

encountered its sister species, *H. zea*. While *H. armigera* has a broad distribution

spanning most of the eastern hemisphere, *H. zea* — which diverged in allopatry 1.5-

2MYA — is native to the Americas [11]. The two species are among the world's most

destructive agricultural pests, and the establishment of *H. armigera* in South America

resulted in annual economic damage of around $US 10^7 [12]. Highly polyphagous larvae

feed on the reproductive structures of major crops including soy, maize, and cotton [13].

The species' broad-scale pesticide resistance, short generation times (up to 11/yr), high

fecundity (>1,000 eggs/female), and long-distance dispersal (up to 40 km/night) enable

explosive population growth [14–17]. Despite being difficult to distinguish morphologically,

the two species are separated by strong prezygotic behavioural and mechanical

barriers, which makes them difficult to cross in captivity [18]. It was therefore a surprise

when genomic data revealed occasional interspecific hybridization in Brazil [19,20]. This

resulted in adaptive introgression of the *de novo* chimeric pyrethroid pesticide

resistance gene *CYP337B3* from invasive *H. armigera* to native *H. zea* populations in

Brazil and subsequently North America [21,22]. Previous work hinted at a single bout of

locus-specific introgression [21]. However, the two species display complementary

ecological adaptations, leading to speculation that further introgression could drive

ongoing adaptation to control measures [20,23].

Invasive *H. armigera* primarily affects soybean and cotton — the former being the most

abundant crop in Brazil, comprising 47.3×10^6 hectares in the 2024-25 crop season, an

area exceeding the total size of Germany [24,25]. However, following its outbreak in 2013,

H. armigera damage to cotton and soy has been effectively suppressed through the

widespread use of soy crops genetically modified to express *Bacillus thuringiensis* (Bt)

toxins such as Cry1Ac [26,27]. In contrast, native *H. zea* cannot complete its lifecycle on

soy; it primarily affects the second most-abundant Brazilian crop — maize — and is

tolerant to Cry toxins, which have been expressed in North American maize and cotton

since the late 1990s [28]. The coexistence of these species raises the question of whether

hybridization could combine the Bt tolerance of native *H. zea* with the soy adaptation of

invasive *H. armigera* to facilitate adaptation to Bt soybean monoculture in Brazil.

Here, we study the recent adaptive consequences of hybridization using a decade-long

whole-genome time series of 975 field-collected moths from across Brazil, combined

with the genomes of a further 149 individual offspring from a cross between Bt-resistant

and susceptible *H. armigera*. This unique dataset captures the complete time course of

the invasion and its aftermath. We reveal ongoing hybridisation and bidirectional gene

flow, identify a trypsin gene cluster introgressed from *H. zea* into *H. armigera*, and show

experimentally that it confers resistance to Cry1Ac-expressing soy in the *H. armigera*

genomic background. These architectures are produced by the sorting of alleles at

major-effect loci independently optimised by anthropogenic selection in native and

invasive pests. These results provide a rare insight into rapid adaptation over ecological

timescales through selection on novel genomic architectures for pesticide resistance.

Results

Outcome of secondary contact

We first sought to characterise the outcome of secondary contact between these

species over time by quantifying individual-level ancestry across the full set of Brazilian

Helicoverpa samples (*n*=975, Extended Data Table 1). There are various possible

outcomes to hybridization. At one extreme, the two species could collapse into a single

hybrid swarm, predicting a pattern of increased intermediate-ancestry individuals over

time. At the other extreme, reinforcing selection would predict declining intermediate-

ancestry individuals and declining ancestry heterozygosity over time. To distinguish

between these patterns, we used non-admixed samples of both species to assign

ancestry (Fig. 1A)

Most Brazilian samples clustered with reference *H. armigera* and *H. zea* samples in

principal component space, with a minority of intermediate individuals (Fig. 1B).

Ancestry proportions estimated using diagnostic loci (Fig. 1C) and genotype likelihoods

(Fig. 1D, Extended Data Fig 1) confirm that most samples primarily carry ancestry from

one species or the other. Early-generation hybrids were rare. Nevertheless, the vast

majority of samples were advanced-generation back-crosses carrying a small

proportion of ancestry from the alternative species. Patterns of ancestry across space

and time indicate an established mosaic hybrid zone throughout Brazil (Extended Data

Figs. 2A,B) consistent with previous reports [29]. The dominant ancestry of samples was

predicted by the crop on which they were sampled: *H. armigera* on soy and cotton, and

H. zea on maize (Extended Data Fig. 3). Proportionally more early-generation hybrids

(those with <95% ancestry of either species) were found where the two species

overlapped spatially and temporarily (Extended Data Fig. 2C).

The frequency of advanced-generation back-crosses in early samples, combined with

the lack of spatial or temporal change in ancestry, confirms that *H. armigera* had

established in South America well before its official discovery [30]. Although the ancestry

average heterozygosity decreased over time, early-generation hybrids were still found in

recent years, indicating rare but ongoing hybridization (Fig. 1D, Extended Data Fig. S4).

In summary, the data show stable coexistence of the two species, but with extensive

interspecific gene flow: most individuals carry a small proportion of alleles from the

alternate species. We next sought to identify those introgressed alleles.

Figure 1) (A) Sample sites of individual genomes across all years (*N*=975). Circle size

corresponds to per-site sample size, colour corresponds to ancestry group. **(B)** Principal

components analysis (PCA) of reference non-admixed samples and Brazilian samples.

(C) Ancestry proportion and heterozygosity of Brazilian samples in Call Set 2. **(D)** *H.*

armigera ancestry proportion of Brazilian samples in the genotype likelihood dataset for

each year.

Genome-wide patterns of introgression

To identify loci with disproportionate representation of the alternate species' ancestry,

we assigned individuals to species using an ancestry proportion threshold of 0.5 (Fig.

1C) and measured excess allele sharing with reference non-admixed heterospecific

samples using the statistic f_d. We used two test topologies, which allowed us to

distinguish between introgression from South American *H. armigera* to South American

H. zea and introgression in the opposite direction (see Methods). Readers familiar with

earlier *Helicoverpa* reference assemblies should note that the assembly used here,

generated by Jin *et al*. [15], numbers chromosomal scaffolds by size (see Extended Data

for synteny comparisons showing the corresponding chromosomes). Aggregating

samples across the time-series, the genome-wide average signature of introgression

was similar for both directions (Extended Data Fig. 5), but the chromosomal distribution

of the introgression signal differed significantly between the two species (Extended

Data Figs. 6,7). Local reductions in F_{ST} and d_{XY} in a sympatric comparison of the two

species (in Brazil), relative to the allopatric comparison (non-admixed native-range

populations), confirm introgression in Brazil (Extended Data Figs. 8, 9). From invasive *H.*

armigera to native *H. zea*, the introgression signal is concentrated at the *CYP337B3*

locus on chromosome 5, confirming previous results (Extended Data Fig. 6). Nucleotide

diversity decreased over time in both donor and recipient species around *CYP337B3*,

clearly demonstrating adaptive introgression (Extended Data Fig. 10). By contrast,

introgression in the other direction is distributed across most chromosomes, with

notable 'plateau' patterns consistent with locally reduced recombination on

chromosomes 21, 24, and 26 (Extended Data Fig. 7). Although synteny between the two

species is high, comparisons within and between both species confirm the

introgression of inversion polymorphisms on these chromosomes (Extended Data Figs.

11, 12, 13, 14, 15, 16). A subset of samples sequenced using haplotagging confirm the

presence of these inversions, and reveal other structural variants (SVs) that were not

introgressed (Extended Data Fig. 17). Intriguingly the haploblock introgressed from *H.*

zea represents the ancestral orientation in all cases (Extended Data Fig. 18). PCA-based

inversion genotyping showed that all *H. armigera* individuals carrying the *H. zea*

inversion allele were heterozygous (Extended Data Figs. 19, 20), suggesting a lack of

strong selection. To narrow down potential instances of adaptive introgression, we next

identified introgression signals that increased consistently over time.

Temporal patterns of introgression

We divided samples into three timepoints comprising approximately equal individual

sample sizes (Extended Data Fig. 21). For each we calculated f_d as above, as well as the

positive change in the introgression signal over the two intervals: $\left(f_{d_{t3}} - f_{d_{t2}}\right) +$

$\left(f_{d_{t2}} - f_{d_{t1}}\right)$. As expected, gene flow from *H. armigera* to *H. zea* increased over time at

loci surrounding *CYP337B3* on chromosome 5 (Fig. 2A, Extended Data Figs. 22, 23, 24).

At this locus, *H. armigera* ancestry increased rapidly among *H. zea* over the first time

interval, reaching near-fixation in recently collected samples. On the remainder of

chromosome 5, *H. armigera* ancestry declined over time in *H. zea* (Extended Data Fig.

24). This result is consistent with decoupling of *CYP337B3* from linked *H. armigera*

alleles over successive generations, declining in frequency either through drift or

selection against interspecific incompatibilities. This phenomenon is predicted in

theory and can be inferred retrospectively, but is rarely observable directly.

From native *H. zea* to invasive *H. armigera*, there was no detectable change in the

frequency of *H. zea* inversion haploblocks on chromosomes 21, 24, and 26, suggesting

that these structural variants are not under directional selection (Extended Data Fig.

25). Instead, introgression primarily increased over time in a narrow region of

chromosome 12 in samples collected after 2018 (Fig. 2B, Extended Data Figs. 26, 27,

28). A cluster of tandem repeat trypsin genes (digestive proteases) mapped to the

chromosome 12 locus showing the greatest increase in introgression from *H. zea* to *H.*

armigera across the whole genome (Extended Data Fig. 27). In North American *H. zea*,

this cluster has previously identified and associated with field-evolved resistance to the

Bt toxin Cry1Ac [28,31]. More generally, trypsins have been repeatedly implicated in Bt

resistance in lepidopterans [32–34]. We confirmed the presence of this introgressed

structural variant by using a subset of linked-read haplotagged samples collected early

in the time series: *H. zea* samples show a pattern of tag-sharing diagnostic of tandem

duplication at the trypsin locus, while *H. armigera* samples did not (Fig. 2C). Although

recently-collected *H. armigera* samples were not haplotagged, the introgressed repeat

cluster could be inferred through changes in local depth of coverage over the time

course (Fig. 2D). At the *Trypsin* locus, normalised depth of coverage in *H. zea* was

elevated and constant between 2012 and 2022, but increased over time in *H. armigera*

coincident with increased use of Bt soy in Brazil (Fig. 2D). Notably, 4 of 8 Brazilian *H. zea*

samples collected in 2005 displayed the diagnostic peak in read depth, indicating

polymorphism prior to the deployment of Bt in Brazil. Together these results point to

selection for a structural variant conferring Bt resistance in Brazilian *H. zea* between

2005 and 2012, followed by recent introgression into *H. armigera*. We next aimed to test

the hypothesis that this introgressed structural variant confers a novel resistance

phenotype in Brazilian *H. armigera* exposed to Bt soy.

Figure 2. (A) Left panel: f_d using the topology (((ZEA, ZEA_BRAZIL), ARM)), PUN),

indicative of introgression from *H. armigera* to *H. zea*, shown for chromosome 5 for

individuals split into three temporal sets. Right panel: f_d using the topology (((ARM,

ARM_BRAZIL), ZEA)), PUN), indicative of introgression from *H. zea* to *H. armigera*, shown

for chromosome 12 for individuals split into three temporal sets. **(B)** Haplotag barcode

sharing in *H. zea* samples collected early in the time-series (bottom right corner) and *H.*

armigera collected early in the time-series (top left corner). A triangular shape

consistent with a tandem duplication is evident at the introgressed trypsin locus in *H.*

zea samples. **(C)** Normalised depth of coverage on chromosome 12 for *H. zea* samples

(top panel) and *H. armigera* samples (bottom panel), where each individual is one line.

Increases in relative depth of coverage in *H. armigera* appear over time, consistent with

introgression. **(D)** Detection of the repeat cluster using depth of coverage. Circle size

indicates individual sample size for *H. armigera* (orange), early-generation hybrids (grey)

and *H. zea* (blue). Dashed lines: Bt adoption in Brazil; percentage of total Brazilian soy

area with Bt, as reported by de Freitas Bueno *et al.* [35] and percentage of Brazilian maize

area with Bt, as reported by Schuster *et al.* [36].

A *H. zea* trypsin repeat cluster confers *Bt* resistance in wild *H.*

armigera

To identify the loci underlying Bt resistance in Brazilian *H. armigera*, we collected wild

individuals in 2018 and artificially selected those that survived on leaf discs of Bt

soybean expressing Cry1Ac (see Methods). Tolerant and susceptible lines were crossed

to produce an F2 population segregating for Cry1Ac resistance (Fig. 3A, 3B). Using

whole-genome resequencing data for data for 149 F_2 individuals plus their parents and

grandparents, we confirmed that all samples were *H. armigera* (minimum *H.* armigera

ancestry proportion = 0.966, mean = 0.975; Extended Data Fig. 29) and carried out

quantitative trait locus (QTL) mapping for Cry1Ac resistance (Extended Data Fig. 30,

Extended Data Table 2). This revealed two major effect loci on chromosomes 12 and 5,

which showed no evidence of epistatic interactions (Fig. 3D, Extended Data Figs. 31, 32,

Extended Data Tables 2, 3, 4, 5).

We next quantified introgression signals as above for the *H. armigera* QTL samples.

Remarkably, both major-effect loci showed a localised enrichment of *H. zea* ancestry

among individuals that survived Cry1Ac treatment (Fig. 3C, Extended Data Fig. 33). This

confirms predictions that locally adapted *H. zea* variants would enhance *H. armigera*

fitness in Brazil [20,23]. However, only one of the two QTL showed evidence of introgression

in field-collected samples. At the chromosome 5 QTL (LOD= 6.48, *p*=3.29×10^{-7}) there

were no obvious signs of interspecific introgression in wild-caught *H. armigera*

(Extended Data Fig. 33) despite its additive effect on Cry1Ac resistance. In contrast,

confidence intervals for the chromosome 12 QTL (LOD= 8.80, *p*=1.57×10^{-9}) overlapped

with the *Trypsin* repeat cluster (Fig. 3C), which we identified as the locus with the

strongest increase in introgression through time in the wild (Fig. 2A). *Trypsin* CNV

presence/absence genotyping based on depth of coverage at this locus showed that

individuals with the CNV were significantly more likely to be Bt-resistant (Fig. 3D). In

wild *H. armigera*, this same locus showed extreme genetic differentiation over time

relative to the genome-wide average (Fig. 3E, Extended Data Fig. 10). Therefore,

laboratory selection for Cry1Ac resistance in Brazilian *H. armigera* had the effect of

enriching for existing introgressed *H. zea* structural variant alleles, thereby replicating

the adaptive introgression observed in wild populations. Interestingly, this QTL showed

signs of underdominance in the presence of Bt, where fitness is lowest in heterozygotes

(Extended Data Fig. 32). Despite this apparent barrier to introgression at low

frequencies, the introgressed trypsin allele frequency in *H. armigera* closely tracks the

rollout of Bt soy in Brazil [35] (Fig. 3D). Although the CNV was fixed in *H. zea* collected from

2012 onwards, samples from 2005 were polymorphic. This suggest that *H. zea* evolved

resistance relatively recently, likely through gene flow from North American *H. zea*

populations exposed to Cry-expressing maize and cotton since the late 1990s, followed

by adoption in Brazilian maize since 2007 (Fig. 3D).

Figure 3: (A) Logarithm of odds (LOD) score for quantitative trait locus (QTL) mapping of Cry1Ac resistance in Brazilian *H. armigera*. Two major effect loci are identified on chromosomes 12 and 5 **(B)**. Intercross design for the linkage mapping experiment presented in 3A. **(C)** Difference in the signal of *H. zea* allele sharing between alive and dead individuals. In both cases f_d was calculated using the topology (((ARM, QTL_SAMPLE), ZEA)), PUN). Dashed line indicates the trypsin locus, with grey shaded regions indicating the QTL confidence intervals. *H. armigera* samples that survived when exposed to Cry1Ac carried an excess of *H. zea* alleles on a haplotype directly overlapping with the QTL confidence interval. **(D)** Count of individuals in the intercross that survived or died when exposed to Cry1Ac, split by genotype and detection of the trypsin CNV based on excess normalised depth of coverage. The A allele, associated with survival in homozygotes, corresponds to the trypsin CNV. **(E)** Genetic differentiation (F_{ST} calculated using genotype likelihoods) on chromosome 12 between samples collected early and late in the time-series. *H. armigera* is shown in orange; *H. zea* is shown in blue. *H. armigera* samples show excess differentiation at the trypsin locus (dashed line; QTL confidence intervals shaded in grey), consistent with introgression over time. *H. zea* shows reduced differentiation at this locus compared to the rest of the chromosome, consistent with a selective sweep.

Discussion

Together these findings indicate persistent but permeable species boundaries upon secondary contact between native and invasive agricultural pests, resulting in bidirectional adaptive introgression of structural variants conferring resistance. In addition to adaptive introgression of *CYP337B3* from invasive *H. armigera* to native *H. zea*, more recent adaptive introgression in the other direction has introduced Bt (Cry1Ac) resistance alleles to ~30% of the invasive *H. armigera* population. Thus, hybridization has generated a novel combination of traits — soy host plant adaptation in *H. armigera* and Bt resistance selected in *H. zea* — which together enable colonisation of an expansive anthropogenic habitat: Brazilian Bt soybean monoculture.

Just as ancient hybridization can combine old adaptive variants to facilitate adaptive diversification over thousands of years, these results demonstrate how contemporary hybridization can assemble independently selected alleles into a single admixed genome, resulting in adaptation over ecological timescales. Notably, the recombined adaptive variants do not need to be 'ancient': the trypsin allele underlying Cry1Ac resistance was polymorphic in Brazilian *H. zea* as recently as 2005, and has only recently been selected — likely through exposure to Cry-expressing maize and cotton in North America [28]. Thus, anthropogenic change not only drives rapid adaptation through selection on standing variation, but can also promote hybridization, which can further accelerate adaptation. Consequently, in this system, resistance to control measures has evolved in two major agricultural pests through bidirectional adaptive introgression over the course of less than a decade.

The practical outcome of hybridization between *H. armigera* and *H. zea* is the 'worst of

both worlds': two distinct agricultural pests are now resistant to control measures, and

show greater adaptive potential. In the case of *H. zea*, the pyrethroid resistance gene

CYP337B3 had reached near-fixation in the most recently collected samples. This large

pyrethroid-resistant Brazilian source population of *H. zea* is a biosecurity threat to the

two largest maize producing agro-economies: the United States, where the spread *H.*

armigera-derived *CYP337B3* in *H. zea* is already causing an increase in pyrethroid

control failures [37], and China, where *H. zea* is non-native. In the case of *H. armigera*,

approximately 30% of samples collected between 2019 and 2022 carried Cry1Ac

resistance alleles. An additional *H. zea*-derived Cry1Ac resistance QTL on chromosome

5 was present in wild-caught *H. armigera*, and is expected to spread in the population

given its additive phenotypic effect. These results provide an early warning of Cry1Ac-

resistant *H. armigera* in Brazil. More broadly, Brazilian *H. armigera* pose an increasingly

imminent biosecurity threat to the U.S., where the species has not yet spread. Transport

of Brazilian *H. armigera* back into its native range is also a biosecurity concern. For

example, Australian *H. armigera* populations harbour monogenic resistance mutations

for the Bt toxin Vip3Aa [38]. This raises the possibility of recombinant trans-Pacific *H.*

armigera genotypes conferring resistance to multiple Bt toxins, potentially threatening

'pyramiding' strategy of multi-toxin Bt crop use.

Overall, a combination of serial-sampled genomic data and linkage mapping has

revealed a combinatorial mechanism driving rapid adaptation. The outcome of

interspecific hybridization can range from rapid adaptation to demographic collapse,

ultimately depending on the genetic architecture and ecological contingency of hybrid

fitness [39]. In systems such as this, admixture has enabled the sharing of adaptive

variation between distinct and competing species. For pests and pathogens, such rapid

evolutionary rescue presents a challenge to food security and multidrug resistance. For

some threatened taxa, the same process could facilitate persistence. Recognising

hybridization as a driver for adaptation on anthropogenic timescales will be essential for

understanding evolutionary responses to global change.

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Methods

Sample collection, DNA extraction, library preparation, and

sequencing

For samples in 'batch 1', *Helicoverpa* adults were collected throughout Brazil as part of routine monitoring. For all other sample batches, *Helicoverpa* larvae were collected from agricultural regions across Brazil and sent to the Arthropod Resistance Laboratory at the Luiz de Queiroz College of Agriculture, University of São Paulo, as part of routine monitoring of susceptibility to chemical insecticides and Bt proteins during the 2012–2022 crop seasons (Fig. 1A, Extended Data Table 1; SISBIO numbers: 65052, 75582 and 61824; SisGen numbers: A792B04 and R1EA05A; TTM/ATM number: 1014273). For each field-collected sample, we performed both molecular identification to distinguish *H. armigera* from the native *H. zea*, following the PCR–RFLP protocol described by Behere *et al.*[1]. Field-collected larvae were reared for at least one generation under controlled laboratory conditions (25 ± 1°C, 70 ± 10% RH, and a 14h photophase). Between three and seven days after adult emergence, individuals from each population in each season were individually placed in 1.5mL Eppendorf tubes containing 98% ethanol and stored at –20 °C.

Genomic DNA was extracted from 2mm wide segments of the thorax using Kučka and Chan's high molecular weight magnetic bead DNA extraction protocol[2]. DNA concentration was quantified using a Qubit™ fluorometer (Invitrogen, USA), and low-concentration samples were re-extracted or excluded. For a subset of samples, DNA was extracted either from the thorax or leg using the CTAB extraction protocol and

resuspended in 50 µl of Milli-Q water [3]. For this subset, DNA quality control was initially

assessed by visual inspection on a 0.8% (w/v) agarose gel, followed by

spectrophotometric and fluorometric analyses, and each sample was evaluated using a

NanoDrop spectrophotometer to detect potential contaminants and an agarose gel to

assess DNA integrity, and subsequently quantified using a Qubit fluorometer.

For all samples, sequencing libraries were prepared via a tagmentation based approach

using a homemade tn5 transposase as per Picelli *et al*. [4]. Prior to pooling, DNA

concentration and insert size were assessed using a Qubit™ fluorometer and

TapeStation (Agilent Technologies) respectively. Equal amounts of each library were

pooled together, and then sequenced (150bp paired-end) using an Illumina NovaSeq

6000 (Novogene UK).

Haplotagging Library Construction and Sequencing

Haplotagging libraries were prepared following the protocol of Meier *et al*. [5]. In total, we

analysed 93 individuals: 63 *H. armigera*, 16 *H. zea*, and 14 putative hybrids collected in

Brazil between 2012–2018. Genomic DNA was extracted from thorax tissue and

normalised to 0.15 ng µl^{-1} in 10 mM Tris (pH 8). Libraries were constructed in 96-well

plates. Each sample received 1.2 µl of haplotagging beads (~0.9 million beads carrying

one of 885K barcodes), 30 µl WASH buffer (20 mM Tris pH 8, 50 mM NaCl, 0.1 % Triton-

X100), 10 µl of 5× tagmentation buffer (50 mM TAPS pH 8.5, 25 mM $MgCl_2$, 50 % DMF),

and 25 µl of 0.6 % SDS to strip Tn5 after tagmentation.

For amplification, one-tenth of the bead–DNA complexes from each of the 96 samples

was pooled within eight-tube strips, and subsequently collapsed into four final pools.

After placing the bead pools on a magnetic rack, residual buffer was removed and

replaced with 20 µl of 1× Lambda Exonuclease buffer supplemented with 10 U

exonuclease I and 5 U Lambda exonuclease (NEB). Beads were incubated at 37 °C for

30 min, washed twice with 150 µl WASH buffer, and resuspended for PCR. Libraries

were amplified using Q5 High-Fidelity DNA polymerase (NEB) in four 25 µl reactions per

pool, using 10 µM TruSeq-F (AATGATACGGCGACCACCGAGATCTACAC) and TruSeq-R

primers (CAAGCAGAAGACGGCATACGAGAT). Cycling conditions were: 72 °C for 10

min; 98 °C for 30 s; followed by 10 cycles of 98 °C for 15 s, 65 °C for 30 s, and 72 °C for

60 s. PCR products were pooled, size-selected using AMPure XP beads, quantified by

Qubit, and normalised to 2.5 nM using 10 mM Tris (pH 8.0, 0.1 mM EDTA).

Pooled libraries were sequenced on an Illumina HiSeq 3000, using a 150 + 13 + 12 + 150

(R1 + i7 + i5 + R2) configuration. This design produces 13-bp i7 and 12-bp i5 index reads

corresponding to haplotag plate and well barcodes.

Bioinformatics

Read mapping

The mapping and variant-calling pipeline was executed using Snakemake [6]. We mapped

reads to the highest-quality available *H. armigera* reference assembly, generated from a

sample collected in north-western China [7]. After trimming fastq files using Fastp v0.23 [8],

and bwa mem v0.7.12 [9] was used to map reads. SAMtools v1.10 [10] was used to sort and

index the resulting bam files. Duplicates were marked and removed using Picard's

MarkDuplicates v2.9.2 [11].

Sequencing Depth Profiling and Sexing

To characterise genome-wide and chromosome-specific sequencing depth per

individual, we used mosdepth v0.3.3 and d4tools v0.3.4 to compute and summarise

base-level coverage from mapped reads. All analyses were conducted on BAM files

previously aligned to the reference genome. Depth was computed per base using

mosdepth with the following parameters: --mapq 20 to restrict to confidently mapped

reads, --d4 to output in compressed D4 format, and --fast-mode to skip insertions and

deletions. Each sample was processed in parallel using 4 threads. To obtain per-

chromosome depth summaries, we ran d4tools stat with the -s median and -s mean

options on the .d4 files. This produced one summary file per sample containing median

and mean depth values per chromosome. These values were used in downstream

analyses to infer sex chromosome copy number, normalise depth across the genome,

and examine population-level variation in coverage patterns.

Haplotagging Read Processing, Mapping and SV detection

Raw BCLs were converted to FASTQ using bcl2fastq v2.20 with: --use-bases-mask

Y150,I13,I12,Y150 --minimum-trimmed-read-length 1 --mask-short-adapter-reads 1 --

create-fastq-for-index-reads (Illumina). Haplotag barcodes consist of four 6-bp

sequences embedded in the FASTQ read name. Barcode strings were assigned to the

BX tag following Meier *et al.* [5], allowing one mismatch when an unambiguous closest

match existed. Adapter trimming and quality filtering (if required) were conducted prior

to alignment. Reads were mapped to the H. armigera reference genome (GenBank:

GCA_030705265.1) using BWA-MEM2, retaining barcode information using the –C flag.

Alignments were coordinate-sorted with samtools, and PCR duplicates were marked

with Picard MarkDuplicates using: READ_ONE_BARCODE_TAG=BX

CREATE_INDEX=true. This preserves molecule-level identity during duplicate marking.

All resulting BAMs were inspected to confirm correct propagation of BX:Z: tags and

uniform coverage across windows. These processed, barcode-aware BAM files were

used for downstream analyses.

We identified candidate structural variants (SVs) using Wrath v2024.03, a tool designed

to detect large-scale SVs from linked-read sequencing data by quantifying barcode

sharing between genomic windows [12]. Analyses were performed separately for each

chromosome using the BAM files with barcodes encoded in the BX tag. To run Wrath,

we provided a list of BAM files per group, the chromosome of interest and a fixed

window size of 10 kb. Wrath calculates the Jaccard index of shared barcodes between

each pair of non-overlapping windows, producing a barcode-sharing matrix for each

chromosome.

Variant calling for call sets 1 and 2

In addition to genotype-liklihood-based methods, some analyses made use of variant

calls. Variants were called using GATK HaplotypeCaller v4.3 [13]. To improve genotyping

calling performance, and to enable comparison with reference populations, a diverse

array of publicly available *Helicoevrpa* sp. was used, in addition to samples from

closely related species. Before filtering, a total of 1366 individuals were included for

genotyping. These included 1017 newly sequenced Brazilian *Helicoverpa* samples; 124

native-range *H. armigera* reported by Jin *et al*. [7], of which 20 were used as reference non-

admixed *H. armigera*; a further 31 *Helicoverpa* samples reported by Anderson *et al*. [14]

including 7 *H. punctigera* individuals as an outgroup; 13 *H. zea* sampled in 2002

reported by Taylor *et al.* [15] for use as reference non-admixed samples; 4 *H. zea* samples

collected in the US in 2019 by North *et al.* [16]; 62 *Helicoverpa* samples collected in North

America by North *et al.* (2026); 112 Brazilian *Helicoverpa* samples collected by Valencia

Montoya *et al.* [17] and Anderson *et al.* [14]; and three additional outgroups: *Chloridea*

subflexa and *C. virescens* generated by Guo *et al.* [18] and *H. assulta* generated by Xu *et al.*

[19].

Some newly sequenced samples previously sequenced by Valencia Montoya *et al.* [17] or

Anderson *et al.* [14]. Of these duplicates, we retained whichever had the highest depth.

After removing such duplicates, we retained a total of included 1128 Brazilian

Helicoverpa samples, including those used for linkage mapping.

GATK CombineVCFs was used to merge GVCF files across individuals and

GenotypeGVCFs was used to call genotypes jointly across samples. In order for

genotyping to be computationally feasible, each of the 31 chromosomes were divided

into 20 equally sized intervals. Genotyping was carried out in parallel per-interval before

concatenating intervals into a single VCF file.

Two filtering approaches were used to generate call set 1 and call set 2, respectively.

Filtering: call set 1

Call set 1 was used to quantify introgression, and was applied to Brazilian *Helicoverpa*

samples, and reference non-admixed groups (see Quantifying Introgression, below).

First, bcftools v1.19 was used to retain only biallelilc SNPs from Brazilian *Helicoverpa*

samples with an autosomal mean depth of coverage of at least 5X, as calculated using

vcftools v 0.1.17. Reference non-admixed samples were also required to meet this

criteria, with the exception of 5 *H. punctigera* samples (M0236, M0239, M0244. M0245,

and M0263) which fell marginally below an average of 5X. From here, all samples GATK

VariantFiltration and SelectVariants (v4.2.5.0) were used to flag and exclude variants

according to the following conditions: SOR > 3.0, FS > 60.0, MQRankSum < -20.0,

MQRankSum > 20.0, ReadPosRankSum < -10.0, ReadPosRankSum > 10.0. Invariant

sites were then excluded. Next, vcftools v 0.1.17 was used to apply the following

conditions: --max-missing 0.75 --minQ 20 --min-meanDP 5 --minDP 5. Next, individual-

level genotype-level filters were applied using a custom python script (filter1.py), which

retained genotypes with a depth in the range between $\frac{1}{5}x$ and $5x$ where x is the mean

autosomal depth of coverage for that individual calculated using vcftools, and

otherwise set genotype calls to missing data ('./.'). Finally, after the introduction of

missing data through the genotype-level filter, vcftools --max-missing 0.75 was re-

applied. This resulted in a VCF file of 628 individuals containing 25,319,153 sites across

all chromosomes.

Filtering: call set 2

Call set 2 consisted of more individuals and a less-stringent filter with respect to the

proportion of missing data and the per-individual mean depth of coverage. Call set 2

was only used for individual-level summary statistics (see Ancestry Proportion and

Ancestry Heterozygosity, below). Filter 2 was only to autosomal loci for Brazilian

Helicoverpa samples and reference non-admixed samples with a depth of at least 0.5X

using mosdepth (see above). First, GATK VariantFiltration and SelectVariants was used

to apply the following conditions: SOR > 3.0, FS > 60, MQRankSum < -20.0,

MQRankSum > 20.0, ReadPosRankSum < -10.0, ReadPosRankSum > 10.0. Next,

vcftools was used to apply the following filters: -max-missing 0.5 --minQ 20 --min-

meanDP 5 --minDP 5. Next, individual-level genotype filters were applied using a

custom python script (filter2_4_v2.py) using the same logic as that used for filter 1:

retaining genotypes with a depth in the range between $\frac{1}{5}x$ and $5x$ where x is the mean

autosomal depth of coverage for that individual. Finally, vcftools --max-missing 0.5 was

re-applied. This resulted in a VCF file of 1,014 individuals comprising 31,013,075

polymorphic autosomal loci.

Ancestry proportion ancestry heterozygosity using call set 2

The ancestry proportion (AP, equivalent to the hybrid index) and ancestry heterozygosity

(AH, equivalent to the interclass heterozygosity) were calculated using sites that were

fixed differences when comparing reference non-admixed *H. armigera* and reference

non-admixed *H. zea* by applying scripts adapted from Liu *et al.* [20] to call set 2.

Specifically, the AP is the per-sample proportion of alleles derived from *H. armigera*,

while the AH is the proportion of genotypes that are heterozygous. AH is therefore

equivalent to Fitzpatrick's interclass heterozygosity H_I and Turelli and Orr's p_H, while the

AP is directly comparable to the hybrid index calculated in genomic or geographic cline

analyses [21,22]. The AP correlated strongly with the ANGSD ancestry proportion (Extended

Data Fig. 1).

Quantifying introgression using call set 1

Given the availability of samples collected in locations and times when admixture had

never been reported (native-range *H. armigera* collected in North-Western China and *H.*

zea collected in North America in 2002, and *H. punctigera*), the following 'backbone'

phylogeny could be used: (ARM, ZEA), PUN. Using the ANGSD *H. armigera* ancestry

proportion to assign Brazilian *Helicoverpa* samples to *H. armigera* (>0.5) or *H. zea*

(<0.5), introgression could therefore be measured in both directions. We reasoned that

gene flow from invasive *H. armigera* to native Brazilian *H. zea* would result in an excess

of ABBA patterns relative to BABA in the following topology: (((ZEA, ZEA_BRAZIL), ARM)),

PUN). In the reverse direction, gene flow from native *H. zea* to invasive Brazilian *H.*

armigera would result in an excess of ABBA patterns relative to BABA in the following

topology: (((ARM, ARM_BRAZIL), ZEA)), PUN). Using this framework, the summary

statistic f_d [23] were calculated in 200kbp windows with at least 500 informative sites, and

20kbp windows with at least 200 usable sites. This was done for both test topologies

using Python scripts described in Martin *et al.* [23]. For window-based analyses, outlier

regions were defined as the upper 5th and 1st percentile of the f_d distribution.

Samples were divided into three timepoints comprising approximately equal individual

sample sizes (Extended Data Fig. 21). For each, the positive change in the introgression

signal over the two intervals was calculated as $\left(f_{d_{t3}} - f_{d_{t2}}\right) + \left(f_{d_{t2}} - f_{d_{t1}}\right)$.

Additionally, F_{ST} and d_{XY} were calculated for the same window sizes using the same

informative site threshold for two comparisons: an allopatric comparison (reference

non-admixed *H. armigera* and *H. zea*) and a sympatric comparison (Brazilian *H.*

armigera and *H. zea*). This was used to confirm the expectation that regions

introgressed in sympatry should show high f_d values and reduced F_{ST} and d_{XY} in

sympatry relative to allopatry.

Mapping and genotyping inversion polymorphisms using call set 1

DGENIES, which implements Minimap2, was used to generate genome-wide dot plot

synteny maps and MAF alignment files [24,25]. Only query and reference alignments of at

least 200kbp, and assembly scaffolds of at least 1Mbp, were retained. Synteny plots

(Extended Data Figs. 11, 12, 13, 14, 15, 16) were generating using R scripts described by

Kim *et al.* [26]. This was applied to intra- and inter-specific comparisons between *H.*

armigera and *H. zea* reference assemblies [7,27–30].

To assign inversion genotypes, PCA was calculated with PLINK using only loci within the

inversion regions of interest, as defined by the breakpoints mapped with Minimapper2.

Clusters were assigned to genotypes manually according to expected genotypic states

under the model proposed in Extended Data Fig. 18.

Population Structure Inference from Genotype Likelihoods

Population structure was inferred using genotype likelihoods computed in ANGSD

v0.935, followed by principal component analysis and admixture estimation using

PCAngsd v1.10. BAM files were grouped by population subset and filtered to retain sites

where ≥80% of individuals were genotyped. Genotype likelihoods were computed using

the GATK model (-GL 2) with base and mapping quality filters (-minQ 20, -minMapQ 30),

and with major alleles and allele frequencies estimated internally (-doMajorMinor 1, -

doMaf 1). Depth filters were scaled by the number of individuals: a minimum of 2×N and

a maximum of 5×mean depth×N. Genotype likelihoods were output in Beagle format (-

doGlf 2) and analysed with PCAngsd using the following non-default parameters:

number of ancestral populations fixed at K=2, shrinkage of the covariance matrix

enabled, and no LD pruning. PCAngsd was run with --admix, --beagle, and inferred

admixture proportions and principal components directly from the genotype

likelihoods.

Admixture proportions using genotype likelihoods

Population structure and ancestry proportions were inferred using PCAngsd v1.10,

which estimates covariance matrices and admixture components directly from

genotype likelihoods. Genotype likelihoods were computed in ANGSD using the filtering

scheme described above, but without downsampling sites; instead, we retained all

genomic positions that passed quality filters and for which ≥75% of individuals in the

population had at least one informative read (ANGSD option -minInd, set to 0.75 × N).

This approach preserves the full genomic signal while excluding sparse or poorly

covered sites that could bias covariance estimation. For each population, genotype

likelihoods were exported in Beagle format after applying minimum allele frequency (-

minMaf 0.05), site-quality, and depth filters, as well as a population-specific maximum-

depth threshold to avoid inflated likelihoods from read pile-ups. Concatenated per-

chromosome likelihood files were used as input to PCAngsd, which jointly estimates the

covariance matrix under a probabilistic framework and computes individual ancestry

coefficients without the need for called genotypes. The resulting admixture coefficients

were taken from the .*admix* output and used as continuous ancestry estimates for all

downstream analyses, including Fst contrasts and temporal structuring.

Detection of hybridisation hotspots across space and time

To test whether hybrids overlap in space and time with regions where both species (*H.*

armigera and *H. zea*) were sampled at a given timepoint, we clustered individuals by

geographic proximity within each year. Specifically, we applied hierarchical clustering

using average linkage on geographic coordinates (latitude and longitude), grouping

individuals sampled within the same year into spatial clusters with a maximum pairwise

distance of 100 km (using the Haversine distance). This approach accounted for the

high dispersal ability of Helicoverpa moths while ensuring sufficient sample sizes within

clusters. For each year-specific cluster, we recorded the number of individuals assigned

to each of three ancestry categories: H. armigera (>95% inferred ancestry), H. zea

(<5%), and early-generation hybrids (intermediate ancestry). A cluster was classified as

an area of species overlap if it contained at least three individuals of each parental

species (H. armigera and H. zea). We then quantified the number of clusters with and

without overlap, and with and without hybrids, and visualised their distributions as a

stacked barplot. To test whether overlap between the parental species was associated

with the presence of hybrids, we performed a Chi-squared test of independence with

Yates' continuity correction.

Genetic Differentiation Analyses using genotype likelihoods

Genetic differentiation between population pairs was estimated using the site allele

frequency (SAF)–based FST framework implemented in ANGSD v0.935 and realSFS.

Individuals were grouped into predefined groups by time ranges (12-15, 16-18, 19-22)

and species' ancestries (considered one of the two study species in their ancestry

proportion was >95%). Genotype likelihoods were computed using the SAMtools model

(-gl 1) with filters for read and mapping quality (-minQ 20, -minMapQ 30), and base

alignment quality recalculation enabled (-baq 2). Sites were further filtered to remove

low-quality reads (-remove_bads 1) and counts were generated using -doCounts 1.

SAFs were estimated per population using -dosaf 1, with ancestral and reference

sequences both specified (-anc, -ref). To avoid bias from over-represented sites, a

maximum depth threshold scaled by population size and average depth was applied (-

setMaxDepth), calculated as 5×mean depth×N. A joint two-dimensional site frequency

spectrum (2D-SFS) was estimated between population pairs using realSFS, based on

their .saf.idx likelihood files. This 2D-SFS was then used to compute FST indices with

realSFS fst index, and genome-wide sliding window FST values were calculated with

realSFS fst stats2 using a window size of 5 kb and step size of 1 kb for plotting.

Nucleotide Diversity and Neutrality Statistics using genotype likelihoods

Nucleotide diversity (θ statistics) was estimated for each predefined population using

the ANGSD–realSFS workflow. For each population, genotype likelihoods were

recalculated as described above and per-population site allele frequency (SAF)

likelihoods were generated using -dosaf 1, with the reference genome also specified as

the ancestral state. To avoid inflation of diversity estimates due to read pile-ups, a

population-specific maximum depth threshold was applied using -setMaxDepth,

calculated as 5 × mean depth × number of individuals in the population. For each

chromosome (1–31), an unfolded per-chromosome site frequency spectrum (SFS) was

first inferred with *realSFS*, restricted to the corresponding genomic interval. This SFS

was then used to convert SAF likelihoods into θ-estimators with *realSFS saf2theta*,

generating per-site estimates of pairwise nucleotide and Tajima's *D* among others.

Genome-wide summary statistics for each chromosome were obtained using thetaStat

do_stat. Sliding-window diversity profiles were then computed for each chromosome

with thetaStat do_stat -win 50000 -step 10000, producing 50 kb windows stepped every

10 kb.

Linkage mapping

Phenotype bioassay

799   A total of 395 F_2 intercross larvae were tested on fresh Bt soybean leaves expressing the

800   insecticidal protein Cry1Ac. Of these, 303 larvae survived (77%) and 92 larvae died (23%).

801   Following DNA extraction and library preparation, due to the low quality of some DNA

802   samples, we successfully sequenced 87 surviving larvae (58%) and 62 dead larvae (42%),

803   resulting in a total of 149 sequenced larvae from the F_2 intercross population.

Linkage map

The linkage map was constructed using Lep-MAP3 [31], utilizing genotypic data from 149 F_2

individuals along with their parents and grandparents, totalling 3,738 markers. The

"ParentCall2" module (with the option removeNonInformative=1) was employed to

estimate the most accurate posterior probabilities of parental genotypes. Additionally,

we used the "Filtering2" module (with the option dataTolerance=0.01) to remove markers

exhibiting significant segregation distortion. The modules "SeparateChromosomes2"

and "joinSingles2all" were not utilized because the reference genome is assembled at

the chromosome level; thus, markers were retained on their respective chromosomes.

We ran the "OrderMarker2" module for each of the 31 chromosomes 15 times and

selected the run with the highest likelihood score. In this step, specific options such as

"outputPhasedData=1", "recombination2=0", "usePhysical=1 0.1" and "useKosambi=1"

were applied. The "OrderMarker2" module orders markers within each linkage group by

maximizing the likelihood of the data given the marker order. Markers that did not show

strong linkage with their flanking markers were manually removed to avoid long gaps. The

final linkage map provided phased chromosomal marker data with missing genotypes

imputed. The phased data were converted for QTL mapping using Lep-MAP3's

map2genotypes.awk script, and the genotypes 1 1, 1 2, 2 1, and 2 2 were transformed to

AA, AB, AB, and BB, respectively.

The final linkage map consisted of 3,735 markers distributed across 31 linkage groups

(LGs), covering a total length of 3,661.8 centimorgans (cM). The number of markers per

LG ranged from 95 to 156, and LG lengths varied between 88.3 and 143.6 cM. The average

inter-marker spacing within each LG ranged from 1.4 to 14.2 cM (Extended Data Fig. 30

and Table S2).

QTL Mapping

QTL mapping was performed using the R/qtl package [32] in the R program [33], employing the

Expectation-Maximization (EM) algorithm implemented in the "scanone" function. The

phenotypes "survival" and "death" were coded as "1" and "0", respectively, and the

parameter model = "binary" was specified to accommodate the binary nature of the trait.

The functions "fitqtl" (with drop-one-term analysis) and, subsequently, "refineqtl" were

used to identify QTL positions with maximum likelihood under a fixed QTL model. The

significant LOD threshold (0.95 percentile) was determined through 10,000 permutation

tests. Confidence intervals for the detected QTLs were estimated using the "lodint"

function. Additionally, we used the "scantwo" function to perform a two-dimensional

genome scan with a two-QTL model to detect potential epistatic interactions between

QTLs.

The QTL analysis for the survival trait revealed suggestive peaks (LOD > 4.09) (Extended

Data Table 3 and Fig. 1). Two significant QTLs associated with survival were identified,

one on LG 5 (LOD = 6.48, $P < 0.05$) and another on LG 12 (LOD = 8.80, $P = 0.004$)

(Extended Data Fig. 31). Additionally, a peak was observed on LG 12 at position 104.9 cM.

However, the QTL analysis did not support this peak (LOD = 0.51, $P = 0.303$). The QTL on

LG 5 exhibited a confidence interval of 15.18 cM, while the QTL on LG 12 had a

confidence interval of 9.65 cM (Extended DataTable 3).

An epistatic interaction test between the QTLs on LGs 5 and 12 showed no significant

interaction (LOD = 1.42, $P = 1$), indicating that these QTLs do not exhibit epistatic effects

associated with the survival trait (Extended Data Table 4). Individuals carrying the

heterozygous genotype of the QTL on LG 5 displayed intermediate survival, indicative of

incomplete dominance. In contrast, the survival was lowest among those with the

heterozygous genotype of the QTL on LG 12, suggesting underdominance (Extended Data

Figure 32 and Extended Data Table 5).

Identification of Candidate genes

To investigate potential genes associated with *H. armigera* survival in soybeans that

express Cry1Ac protein, we used a 300 kb upstream and downstream of regions near

each significant QTL marker identified. This process utilized the genome annotation file

(GCF_030705265.1-RS_2024_03) from *H. armigera* genome assembly GCF_030705265.1

published in the NCBI.

Data Availability

All newly generated sequencing data have been uploaded to the ENA: ERP162059.

Sample metadata is available in Table S1 and on the Earthcape database:

https://heliconius.ecdb.io/

Code Availability

Custom scripts available at https://github.com/Insect-evolution-and-genetics-
lab/genome_collision_scripts

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Author contributions

HLN: Conceptualisation (computational analysis, sampling design), data curation, formal analysis (bioinformatics), methodology (bioinformatics), project administration, software implementation, investigation (lab work). **GMK**: Conceptualisation (computational analysis, sampling design), data curation, formal analysis (bioinformatics), methodology (bioinformatics), project administration, software implementation, supervision (of AW), investigation (lab work). **DA**: Conceptualisation (computational analysis, experimental design), data curation, formal analysis (bioinformatics, linkage map), methodology development and implementation (DNA extraction and library preparation), project administration, investigation (bioassay), software implementation (QTL mapping), resources (sample provision). **IAW:** Methodology (DNA extraction and library preparation), investigation (lab work), data curation. **MK:** Methodology (DNA extraction, library preparation, haplotagging), investigation (lab work). **AW:** formal analysis (spatial analysis). **YFC:** Conceptualisation, Resources, Funding. **TW:** Conceptualisation, Resources, Funding, Data Curation. **ASC:** Conceptualisation, Investigation (DNA extraction and Library preparation), Formal Analysis (preliminary analysis), Resources, Funding, Data

1005   Curation, Project Administration. **CO:** Conceptualisation, Resources, Funding, Data

1006   Curation, Project Administration. **CDJ:** Conceptualisation, Supervision (of HLN, GMK,

1007   DA, AW), Resources, Funding, Data Curation, Project Administration.

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1009   Competing Interests Declaration

The authors declare no competing interests.

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1027 # Extended Data

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1029 Citations used in Extended Data Tables and Figures correspond to those in the Methods
1030 References.

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1032 **Extended Data Table 2.** Linkage map summary.

LG	Nº Marker	Length	Ave. Spacing	Max. Spacing
1	119	141.5	1.2	6.9
2	127	122.8	1.0	4.8
3	133	96.6	0.7	2.1
4	119	112.4	1.0	3.5
5	124	88.3	0.7	1.4
6	127	93.8	0.7	1.4
7	141	138.0	1.0	6.9
8	136	125.6	0.9	8.4
9	119	109.7	0.9	4.1
10	128	110.4	0.9	3.5
11	112	93.8	0.8	4.1
12	130	111.8	0.9	3.5
13	122	114.5	0.9	3.5
14	117	91.7	0.8	3.5
15	130	143.6	1.1	6.2
16	125	135.3	1.1	5.5
17	116	97.3	0.8	2.8
18	117	115.9	1.0	4.8
19	156	141.4	0.9	2.8
20	137	142.8	1.1	3.5
21	121	127.0	1.1	5.5
22	121	111.1	0.9	3.5
23	110	95.9	0.9	3.5
24	110	137.4	1.3	5.5
25	101	120.8	1.2	6.2
26	112	104.9	0.9	4.8
27	98	133.3	1.4	5.5
28	119	125.6	1.1	4.8
29	100	123.6	1.2	6.2
30	95	130.1	1.4	14.2
31	113	125.0	1.1	7.6
Overall	**3,735**	**3,661.8**	-	-

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Extended Data 3. Localisation and LOD scores of markers in the QTL peak and confidence intervals for Cry1Ac resistance in the *Helicoverpa armigera* F2 intercross population.

QTL	LG	Pos (cM)	LOD	P	Low Pos (cM)	High Pos (cM)
Q1	5	62.76	6.48	3.29×10^{-07}	51.72	66.90
Q2	12	2.76	8.80	1.57×10^{-09}	0.69	10.34

Extended Data 4. Summary table of the two-QTL genome scan to identify QTL interactions between the QTL involved in survival trait in *H. armigera* resistant to Cry1Ac.

LG	Pos (cM)	LG	Pos (cM)	LOD Full[a]	P	LOD FV[b]	P	LOD Int.[c]	P	LOD Add.[d]	P	LOD AV[e]	P
5	62.76	12	2.76	14.77	< 0.05	7.90	< 0.05	1.42	1	13.35	< 0.05	6.48	< 0.05

[a]LOD Full: The LOD score of the full model encompasses both QTLs and their interaction.
[b]LOD FV: The LOD score of the full model compared to a model that includes only the second QTL (Pos2) without the first QTL (Pos1).
[c]LOD Int: The LOD score for the interaction between the two QTLs.
[d]LOD Add: The LOD score of the additive model that includes both QTLs without considering their interaction.
[e]LOD AV: The LOD score of the additive model compared to a model that includes only the second QTL (Pos2) without the first QTL (Pos1).

Extended Data 5. Genotypic frequencies of markers identified at QTL peaks associated with *Helicoverpa armigera* survival to the Cry1Ac protein.

Genotype	Survival	Dead	Total	% Survival	% Dead
QTL 1 - Marker rs5P9830222					
AA	38	12	50	76	24
AB	40	26	66	61	39
BB	7	22	29	24	76
Total	85	60	145		
QTL 2 - Marker rs12P7798483					
AA	76	31	107	71	29
AB	2	19	21	10	90
BB	7	10	17	41	59
Total	85	60	145		

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Extended Data Figure 1 *H. armigera* ancestry proportion (AP) calculated using
genotype calls and reference non-admixed *H. armigera* and *h. zea* (horizontal axis)
against the ancestry proportion calculated using genotype likelihoods (vertical axis),
with individual sample IDs labelled. Points and text are coloured by the ancestry
heterozygosity calculated using genotype calls, which is maximised in early-generation
hybrids (Figure 1).
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Extended Data Figure 2: Spatial and temporal dynamics of Helicoverpa ancestry and
early-generation hybrid occurrence in Brazil. **(A)** Proportion of individuals assigned to
>95% H. armigera ancestry (orange), >95% H. zea ancestry (blue), or early generation
hybrids (white) across Brazilian sampling sites grouped into three time periods: 2012–
2015, 2016–2018, and 2019–2022 and in clusters whereby individuals are at a maximum
of 100 km from each other. Pie chart size is proportional to sample size per locality. **(B)**
Kernel density distributions showing spatial clustering of H. armigera, early hybrids, and
H. zea across Brazil. Warmer colours indicate higher density of sampled individuals. **(C)**
Proportion of sampling clusters where early hybrids are present or absent, split by regions
with spatial overlap vs no spatial overlap of the two parental species. Hybrids are
significantly more common in regions where the distributions of H. armigera and H. zea
overlap (Chi–sq = 8.39, df = 1, p = 0.0038).

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1102 **Extended Data Figure 3** Individual ancestry proportion (AP) and ancestry
1103 heterozygosity (AH) across all timepoints and sample sites as for Figure 1C, for each
1104 crop from which individuals were sampled: maize, cotton, soybean (A-C respectively).
1105 Predominantly-*H. armigera* samples were collected on soybean and cotton, while
1106 predominantly-*H. zea* samples were collected on maize.
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1112 **Extended Data Figure 4** Individual ancestry heterozygosity vs collection date. Line
1113 shows a fitted linear model, which has a significantly negative gradient (Person's
1114 product moment correlation: *p*<0.01, t = -4.9656, df = 746; correlation coefficient -
1115 0.177). **A**: all data shown. **B**: data shown for AH values <0.2.
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Extended Data Figure 5: Distribution of f_d calculated in 200kbp windows (horizontal axis limited to values between 0 and 0.1). Orange: gene flow from *H. armigera* to *H. zea* using the topology (((ZEA, ZEA_BRAZIL), ARM)), PUN); blue: gene flow from *H. zea* to *H. armigera* using the topology (((ARM, ARM_BRAZIL), ZEA)), PUN). Dashed lines indicate genome-wide medians: 0.0136 (*H. armigera* → *H. zea*), 0.0178 (*H. zea* → *H. armigera*).

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1153 **Extended Data Figure 6: Genome-wide landscape of introgression from _H. armigera_**
1154 **to _H. zea_ across all timepoints and locations.** f_d calculated in 200kbp windows (lines)
1155 and 20kbp windows (points) using the topology (((ZEA, ZEA_BRAZIL), ARM), PUN).
1156 Highlighted windows indicate 20bkp windows in the upper first percentile of f_d values
1157 (those equal to or greater than 0.310).

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1160 **Extended Data Figure 7: Genome-wide landscape of introgression from *H. zea* to *H.***
1161 ***armigera* across all timepoints and locations.** f_d calculated in 200kbp windows (lines)
1162 and 20kbp windows (points) using the topology (((ZEA, ZEA_BRAZIL), ARM)), PUN).
1163 Highlighted windows indicate 20bkp windows in the upper first percentile of f_d values
1164 (those equal to or greater than 0.139).

Extended Data Figure 8: Local reductions in genetic differentiation in sympatry relative to allopatry, between *H. armigera* and *H. zea*. Genetic differentiation (F_{ST}) was calculated in 200kbp windows with at least 500 informative sites (black lines) and 20kbp windows with at least 200 usable sites (points). "Allopatric" refers to the comparison of reference non-admixed *H. armigera* and *H. zea* in China and the United States respectively. "Sympatric" refers to the comparison of predominantly-*H. armigera* and predominantly-*H. zea* samples in Brazil, as defined by their ancestry proportion. Shaded regions indicate 20kbp windows with f_d values in the upper first percentile (blue: *H. zea* → *H. armigera*; orange: *H. armigera* → *H. zea*).

Extended Data Figure 9: Local reductions in genetic divergence in sympatry relative to allopatry, between *H. armigera* and *H. zea*. Genetic divergence (d_{XY}) was calculated in 200kbp windows with at least 500 informative sites (black lines) and 20kbp windows with at least 200 usable sites (points). "Allopatric" refers to the comparison of reference non-admixed *H. armigera* and *H. zea* in China and the United States respectively. "Sympatric" refers to the comparison of predominantly-*H. armigera* and predominantly-*H. zea* samples in Brazil, as defined by their ancestry proportion. Shaded regions indicate 20kbp windows with f_d values in the upper first percentile (blue: *H. zea* → *H. armigera*; orange: *H. armigera* → *H. zea*).

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Extended Data Figure 10: Genome-wide and locus-specific shifts in diversity statistics
calculated with ANGSD between early and late collections of *Helicoverpa armigera* (A-
D) and *H. zea* (E-H). (A, B) Genome-wide distributions of ΔTajima's D (A, E) and Δπ (B, F)
for *H. armigera* and *H. zea*, respectively (2012–2015 → 2019–2022). Vertical lines
indicate the Δ values at CYP337B3 on chromosome 5 (solid) and the trypsin cluster on
chromosome 12 (dashed). Chromosome-specific ΔTajima's D (C, G) and Δπ (D, H)
across chromosomes 5 and 12 in *H. armigera* and *H. zea*, with the positions of
CYP337B3 and the trypsin cluster marked.

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Extended Data Figure 11: Synteny is generally highly conserved within and between *Helicoverpa armigera* and *Helicoverpa zea*. Axes show scaffold positions of all 31 homologous chromosomes, where "1" corresponds to the Z chromosome in the arm_XJ reference genome used here. Grid lines delineate scaffolds. Colour corresponds to genetic divergence (green=low, yellow=medium, orange=high). **(A)** Comparison of the reference genome used here, derived from a sample collected in north-western China (horizontal axis) and a reference genome of a sample collected from a distinct *H. armigera* population in southern China (vertical axis) [34] **(B)** Comparison of the *H. armigera* reference genome used here (horizontal axis) and a reference genome generated for *H. armigera conferta*, collected in Australia (vertical axis) [29]. **(C)** Comparison of the *H. armigera* reference genome used here (horizontal axis) and the *H. zea* reference genome generated using the GA-R strain (vertical axis) [27]. **(D)** Comparison of the *H. zea* GA-R reference genome (vertical axis) and the *H. zea* HzStark_Cry1AcR strain [28]. **(E)** Comparison of the *H. armigera* reference genome used here (horizontal axis) and a *H. armigera* reference genome generated using a sample collected in Europe (vertical axis) [30].

H. zea strain HzStark_Cry1AcR

H. zea strain GA-R

Extended Data Figure 12: No inversion polymorphism between *H. zea* reference assemblies at the three chromosomes of interest. GA-R chromosomes 26, 28 and 27 correspond to chromosomes 21, 24 and 26 in the reference genome used here. Shaded regions show nucleotide alignments between the GA-R reference genome (lower scaffolds) [27] and the *H. zea* HzStark_Cry1AcR strain (upper scaffolds) [28].

H. armigera armigera (South China)

H. armigera armigera (Xinjiang)
Reference genome used here

Extended Data Figure 13: Inversions polymorphic between *H. armigera* and *H. zea* are also polymorphic between distinct *H. armigera* lineages. Shaded regions show nucleotide alignments between the reference genome used here (lower scaffolds) and a reference assembly generated from a sample collected in South China (upper scaffolds) [34].

H. armigera conferta (Australia)

H. armigera armigera (Xinjiang)
Reference genome used here

Extended Data Figure 14: On chromosome 26, inversions polymorphic between *H. armigera* and *H. zea* are not polymorphic between *H. armigera armigera* and *H. armigera conferta*. The other two inversions of interest are polymorphic within *H. armigera*. Shaded regions show nucleotide alignments between the reference genome used here (lower scaffolds) and a *H. armigera conferta* reference assembly [29].

H. armigera armigera (Europe)

H. armigera armigera (Xinjiang)
Reference genome used here

Extended Data Figure 15: Inversions polymorphic between *H. armigera* and *H. zea* are also polymorphic between *H. armigera* populations. Shaded regions show nucleotide alignments between the reference genome used here (lower scaffolds) and a reference assembly generated from a sample collected in Europe (upper scaffolds) [30]

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Extended Data Figure 16: Introgression of three *H. zea* inversions into *H. armigera*. All
three inversions arose within *H. armigera*, so the introgressed *H. zea* haplotype is the
ancestral orientation. **(A-C)** Synteny maps comparing the *H. armigera* reference genome
used here (bottom scaffold) to the *H. zea* GA-R reference genome [27] for chromosomes
21, 24 and 26 respectively. Scaffold numbers indicate homologous chromosomes in the
two assemblies (upper: *H. zea*, lower: *H. armigera*). **(D-F)** f_d (*H. zea* → *H. armigera*)
calculated in non-overlapping 200kbp windows (lines) and 20kbp non-overlapping
windows (points). Dark coloured points indicate those in the upper first percentile of f_d
values genome-wide. Shaded regions indicate the bounds of the inversion identified in
A. **(G-I)** F_{ST} calculated between allopatric non-admixed *H. armigera* and *H. zea*
(maroon) and sympatric Brazilian *H. armigera* and *H. zea* (green). Lines: 200kbp
windows, points: 20kbp windows.

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Extended Data Figure 17: Genome-wide barcode-sharing heatmaps used to detect
structural variants in *H. zea* and *H. armigera*. Heatmaps show Wrath-derived barcode-
sharing matrices across all chromosomes, calculated from linked-read BAM files using
10-kb non-overlapping windows. For each chromosome, the upper triangular matrix
represents the *H. zea* dataset and the lower triangular matrix represents the *H. armigera*
dataset. Colours reflect the Jaccard index of shared barcodes between window pairs,
with darker regions indicating stronger barcode similarity. This method captures local
deviations from the expected decay of barcode sharing with distance and highlights
putative structural variants. Chromosome numbers are displayed in the upper right of
each panel. Heatmaps correspond to the same matrices used by Wrath's SV-calling
module (–l), which applies z-score outlier detection and a double exponential decay
model to identify structural rearrangements.

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Extended Data Figure 18: Proposed scenario for all three inversion polymorphisms on
chromosomes 21, 24 and 26, based on the most parsimonious explanation for inversion
polymorphisms. An inversion mutation occurred within *H. armigera*, resulting in a
derived inverted allele that remains polymorphic across divergent *H. armigera* lineages.
The mutation occurred in a population ancestral to multiple present-day *H. armigera*
lineages, or arose in one *H. armigera* lineage and spread to others. There are therefore
three possible genotypes in allopatric *H. armigera* and one possible genotype in
allopatric *H. zea*. Upon secondary sympatry in Brazil, there are six possible genotypes.
Based on the available data I cannot exclude the possibility that the inversion was
present in the ancestor to *H. armigera* and *H. zea*, but was lost in *H. zea*.
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Extended Data Figure 19: PCA in in inversion regions confirms introgression from *H. zea* to

H. armigera. Point colour corresponds to ancestry, as defined by the likelihood-based

ancestry proportion. Unlike the pattern outside of inverted regions in which individuals

cluster by ancestry proportion, in inversion regions individuals fall into three major

clusters with up three rarer intermediate clusters. This is consistent with expectations

under the proposed evolutionary scenario for all three inversions (Extended Data Fig.

18). Heterozygous individuals are overwhelmingly *H. armigera* samples, consistent with

the introgression signal.

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Extended Data Figure 20: Putative inversion genotypes assigned using PCA clusters.

Shape corresponds to ancestry, colour corresponds to genotype assignment. This

approach distinguishes between two possible *H. armigera* alleles, but does not assign

which is ancestral or derived. This is because of the significant degree of genetic

differentiation between the three putative homozygous genotypes. Unlike the pattern

outside of inverted regions in which individuals cluster by ancestry proportion, in

inversion regions individuals fall into three major clusters with up three rarer

intermediate clusters. This fits expectations under the proposed evolutionary scenario

for all three inversions. Individual genotype assignments are reported in Extended Data

Table 1. Recombinant genotype assignments fall between genotype categories,

representing possible cases of recombination or gene conversion.

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Extended Data Figure 21. Population ancestry composition among Brazilian
Helicoverpa samples across three genomic datasets used for these analyses. (A)
Genotype-likelihood dataset (n = 820 individuals), (B) variant-calling dataset (n = 796),
and (C) haplotagging dataset (n = 93). Bars show absolute counts of individuals per
year-group, coloured by ancestry group as determined by PCAngsd (armigera-like, zea-
like, and early generation hybrid categories). All datasets include only Brazilian and
newly sequenced individuals; additional published samples were excluded for
consistency across panels. All individuals in the variant-calling dataset (B) are included
in genotype-likelihood dataset (A).
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1329 **Extended Data Figure 22:** Positive change in f_d over three timepoints, calculated in
1330 200kbp windows with at least 500 usable sites per window for the topology (((ZEA,
1331 ZEA_BRAZIL), ARM)), PUN), *i.e.* gene flow from *H. armigera* to *H. zea*.
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Extended Data Figure 23: Vertical axis: Positive change in f_d over three timepoints,
calculated in 20kbp windows with at least 200 usable sites per window for the topology
(((ZEA, ZEA_BRAZIL), ARM)), PUN), *i.e.* gene flow from *H. armigera* to *H. zea*. Horizontal
axis: minimum number of sites used per window. Points are labelled with the
chromosome on which the window occurs. The window on which *CYP337B3* occurs is
highlighted in orange.
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 Extended Data Figure 24: f_d using the topology (((ZEA, ZEA_BRAZIL), ARM)), PUN),
 indicative of introgression from *H. armigera* to *H. zea*, shown for chromosome 5 for
 individuals split into three temporal sets. 200kbp windows with at least 500 usable sites
 shown as lines; 20kbp windows with at least 200 usable sites shown as points. Dasshed
 line indicates the position of *CYP337B3*.

 Extended Data Figure 25: f_d using the topology (((ARM, ARM_BRAZIL), ZEA)), PUN),
 indicative of introgression from *H. zea* to *H. armigera*, shown for chromosomes 21, 24,
 and 26 (A-C respectively), for individuals split into three temporal sets. f_d calculated in
 200kbp windows (lines) and 20kbp windows (points), with inversion breakpoints
 inferred based on synteny comparisons displayed in grey. There is no apparent change
 in the introgression signal over time in the inversion regions.

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1359 **Extended Data Figure 26:** Positive change in f_d over three timepoints, calculated in

1360 200kbp windows with at least 500 usable sites per window for the topology (((ARM,

1361 ARM_BRAZIL), ZEA)), PUN), *i.e.* gene flow from *H. zea* to *H. armigera*.

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Extended Data Figure 27: Vertical axis: Positive change in f_d over three timepoints,

calculated in 200kbp windows with at least 500 usable sites per window for the

topology (((ARM, ARM_BRAZIL), ZEA)), PUN), *i.e.* gene flow from *H. zea* to *H. armigera*.

Horizontal axis: minimum number of sites used per window. Points are labelled with

the chromosome on which the window occurs. The window on which the trypsin allele

cluster and the QTL for Bt resistance occurs is highlighted in blue.

Extended Data Figure 28: f_d using the topology (((ARM, ARM_BRAZIL), ZEA)), PUN), indicative of introgression from *H. zea* to *H. armigera*, shown for chromosome 12 for individuals split into three temporal sets. 200kbp windows with at least 500 usable sites shown as lines; 20kbp windows with at least 200 usable sites shown as points. Dashed line indicates the position of the trypsin cluster.

Extended Data Figure 29: Ancestry proportion and heterozygosity (as for Figure 1C) of QTL samples using Call Set 2; all individuals are *H. armigera* (minimum *H.* armigera ancestry proportion was 0.966, mean = 0.975).

 **Extended Data Figure 30: Linkage map of F2 the intercross population, derived from
 the cross between the *Helicoverpa armigera* susceptible and the Cry1Ac resistant
 strains.** Linkage groups are ordered along the horizontal axis. Each line in a linkage
 group denotes a marker, and the vertical axis shows the genetic position of markers in
 cM.

 **Extended Data Figure 31: The LOD profiles of linkage groups 5 and 12, where
 significant QTLs for the survival trait were identified.** The Y-axis shows the LOD ratio,
 and the X-axis shows the linkage groups 5 and 12. The lines at the bottom of the linkage
 group denote markers.

Extended Data Figure 32: Effect plot of the main QTL on linkage groups 5 and 12. The X-axis represents the genotypes homozygote resistant (AA), heterozygote (AB) and homozygote susceptible (BB). The Y-axis represents the phenotypic value (survival ratio) after a transformation using the logit function.

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Extended Data Figure 33 A: Difference in the signal of *H. zea* allele sharing between
alive and dead individuals. In both cases f_d was calculated using the topology (((ARM,
QTL_SAMPLE), ZEA)), PUN). Grey shaded regions indicate the QTL confidence intervals.
B: introgression from *H. zea* to *H. armigera* among wild samples divided into 3
timepoints, using the same test topology. Lines indicate 200kbp windows; points
indicate 20kbp windows with at least 200 usable sites.
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