Intersecting experimental evolution and CRISPR screens to identify novel toxin resistance loci

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Short title: Orthogonal selection screens for toxin resistance genes

Abstract

Understanding toxin resistance in insects is key to appreciate niche adaptations but remains challenging due to its often-polygenic basis. A well-known example is the specialized association of *Drosophila sechellia* with noni fruit (*Morinda citrifolia*), which is toxic to most other insects, including the closely-related drosophilids *D. simulans* and *D. melanogaster*. Noni toxicity is due to its high concentration of octanoic acid (OA), but the mechanisms that determine sensitivity or resistance to OA remain poorly understood. Here, we experimentally-evolved *D. simulans* with increased OA resistance, identifying multiple loci under selection. Cross-referencing these with a genome-wide, OA-resistance CRISPR screen in a *D. melanogaster* cell line highlighted two proteins: Kraken, a putative detoxification enzyme expressed in digestive and renal tissues, and Alkbh7, a mitochondrial protein linked to fatty acid metabolism. Both genes show elevated expression in *D. sechellia* and OA-resistant *D. simulans*. In *D. melanogaster*, *kraken* mutants are more OA-sensitive, while *Alkbh7* overexpression increased OA resistance. Importantly, mutation of these genes in *D. sechellia* reduced OA tolerance. Our identification of genes underlying OA resistance in laboratory and natural contexts demonstrates how complementary, cross-species selection approaches can provide insights into complex mechanisms of toxin susceptibility and adaptation, and have practical applications in the characterization of novel insecticides.

Introduction

How insects resist toxic chemicals in the environment is of both basic scientific interest – reflecting insects' arms race with plant food sources (Erb and Reymond 2019; Jones, et al. 2022) or venomous predators (Smith, et al. 2013) – and applied interest, because, throughout history, humans have used chemicals to combat insect disease vectors and agricultural pests (Umetsu and Shirai 2020). The best-understood insecticides are those that bind to and modulate neuronal ion channels, where resistance arises through changes in these target proteins (Sparks, et al. 2020; Raisch and Raunser 2023). For example, dichlorodiphenyltrichloroethane (DDT) and pyrethroids prevent closure of voltage-gated sodium channels, resulting in neuronal excitotoxicity and organismal paralysis; diverse insect species displaying resistance for these chemicals have independently-arising mutations in genes encoding these channels (Busvine 1951; Dong 2007). Similarly, neonicotinoids bind to and hyperactivate nicotinic acetylcholine receptors, leading to neuronal death; mutations in various receptor subunits can confer resistance (Matsuda, et al. 2020). Even in such well-studied examples, however, it is thought that other neuronal targets exist (Soderlund and Bloomquist 1989) and/or that the chemicals impact receptors in non-neuronal tissues (Di Prisco, et al. 2013). For many insecticides – as well as other environmental pollutants (Gandara, et al. 2024) – the mode of action is incompletely, or not at all, characterized (Sparks, et al. 2020).

Resistance observed in nature is only partially understood, reflecting the contribution of many genes beyond those encoding target proteins, which can underlie distinct defence mechanisms (Clarkson, et al. 2018; Catteruccia 2020; Lucas, et al. 2023). For example, thickening of the cuticle to reduce entry of insecticides into the body is widely considered an effective resistance mechanism (Balabanidou, et al. 2018; Balabanidou, et al. 2019). Cuticular changes have been linked to overexpression of proteins implicated in exoskeleton development (e.g.,

cuticle proteins, ABC transporters) but their precise roles, and how they are upregulated, remain unclear (Balabanidou, et al. 2018). Increased expression of metabolic enzymes that degrade insecticides is another commonly observed adaptation in resistant strains, sometimes resulting from copy-number increases in the corresponding genes (Weetman, et al. 2018). Duplicated metabolic genes can also undergo neofunctionalization to produce enzymes with altered specificity, as illustrated by Cytochrome P450s in strains of the brown planthopper that recently evolved resistance to the neonicotinoid imidacloprid (Zimmer, et al. 2018), as well as by more ancient evolution of glutathione S-transferases during transition of the *Scaptomyza* genus of Drosophilidae to herbivory to counter toxic isothiocyanates (Gloss, et al. 2014). However, the lack of experimental resources in many insect species renders it difficult to characterize the causal contribution of specific genetic changes.

A compelling natural model clade to study the evolution of toxin resistance is the trio of closely-related drosophilid species, *Drosophila melanogaster*, *Drosophila simulans* and *Drosophila sechellia* (Fig. 1A). *D. sechellia* is an extreme specialist on the toxic noni fruit of the *Morinda citrifolia* shrub (Jones 2005; Auer, et al. 2021). Amongst *D. sechellia*'s numerous adaptations (R'Kha, et al. 1991; Dekker, et al. 2006; Auer, et al. 2020; Auer, et al. 2021; Alvarez-Ocana, et al. 2023; Shahandeh, et al. 2024), this species has evolved high resistance to octanoic acid (OA), an abundant noni chemical that is toxic to other drosophilids and more divergent insects (Legal, et al. 1994; Jones 1998; Legal, et al. 1999; Markow 2019; Auer, et al. 2021; Kaczmarek, et al. 2022; Kaczmarek, et al. 2024). Previous pioneering studies of *D. sechellia*'s resistance to OA has been studied through genomic mapping-based approaches in *D. sechellia*/*D. simulans* hybrids, which revealed this trait to be polygenic, comprising at least five genomic regions (Amlou, et al. 1997; Amlou, et al. 1998; Jones 1998; Huang and Erezyilmaz 2015). One of the main effect loci encompasses a cluster of 18 genes including members of the *Osiris* (*Osi*) family, which are thought to function in cuticle patterning (Ando, et al. 2019; Sun, et al. 2024). RNA interference (RNAi) knockdown of some *Osi* genes in *D. melanogaster* led to reduced tolerance of OA (Andrade Lopez, et al. 2017; Lanno, et al. 2019). However, genetic evidence for the contribution of these genes to OA resistance in *D. sechellia* is lacking. Given the challenges of identifying candidate genes through genomic mapping-based strategies, in this work we took two complementary approaches to investigate the evolution of OA resistance in this drosophilid trio: experimental evolution in *D. sechellia*'s closest relative, *D. simulans*, through an Evolve-and-Resequence strategy (Schlotterer, et al. 2016), and CRISPR/Cas9-based genome-wide genetic screening in cultured *D. melanogaster* cells (Viswanatha, et al. 2018). We show how the intersection of these approaches can identify novel genes contributing to OA susceptibility/resistance in these drosophilids, establishing these methods and this species clade as a powerful paradigm to investigate ecotoxicology.

Results

Experimental evolution of OA resistance in *D. simulans*

A previous study demonstrated that *D. simulans* strains can be experimentally evolved for higher OA resistance, enabling the identification of genomic regions (through microsatellite markers) responding to selection (Colson 2004). This success

inspired us to take an Evolve-and-Resequence approach, which uses high-
throughput sequencing to analyse the entire genome of populations before and after
exposure to specific selective pressures, thereby facilitating the identification of
adaptive loci from standing genetic variation in controlled and replicable settings
(Turner and Miller 2012; Schlotterer, et al. 2016; Barghi, et al. 2019; Kelly and
Hughes 2019; Shahrestani, et al. 2021). For studies of insecticide resistance, it has
been predominantly used to examine changes in frequency of alleles of known
genes rather than as a means to identify novel loci (Zoh, et al. 2021; Sadia, et al.
2024).
The susceptibility/resistance of drosophilids to OA can be easily assessed by
exposing flies to the toxin in a culture tube and counting the survivors after 24 h (Fig.
1B). In line with previous observations (Amlou, et al. 1997; Colson 2004), we found
that a dose of 3 µl of pure OA (mixed with 1 ml glucose solution), corresponding to
roughly half the amount estimated to be present in 2 g of noni pulp (see Materials
and Methods), leads to death of ~60% of *D. simulans* but has no effect on *D.
sechellia* (Fig. 1C). By collecting the *D. simulans* survivors, we reasoned that we
could allow these to mate and establish the next generation (Fig. 1B), thus enriching
genetic variants conferring higher resistance, similarly to (Colson 2004). Repeating
the process over multiple generations and multiple replicates would allow us to track
phenotypic changes in resistance and investigate the associated trajectories in allele
frequencies.
Standing genetic variation should be present in the evolving populations for
natural selection to act upon (Teotonio, et al. 2009; Long, et al. 2015). As the
spontaneous mutation rate in drosophilids is too low for artificial selection (Keightley,
et al. 2014), we generated ten independent polymorphic base populations of *D.
simulans* from 93 isogenic strains (Signor, et al. 2018) by round-robin crossing (Fig.
1D). These lines contain millions of SNPs, but we note that they were sampled from
a single wild population in California, therefore reflecting the genetic variation
present in that locality. The base populations therefore might not include all variants
found across the species' range, including those in regions where *D. sechellia* co-
occurs. To minimize bottleneck effects, each population was kept at a size of ~1000
flies. At each generation, ~500 flies from each population were randomly collected
and selected for resistance to 3 µl OA. Over the course of 25 generations, all the
populations showed increased levels of survivability, suggesting parallel adaptive
evolution (Fig. 1E). Given the consistent responses across all populations, we also
analysed the phenotypic changes grouping all replicates, finding that the mean
survivability significantly increased from ~43% at generation 0 to ~78% at generation
25 (Fig. 1F).
As we kept the same amount of OA throughout the 25 generations of
selection, the selective pressure lowered as the populations' survivability increased,
evidenced as a larger fraction of flies surviving and reproducing. Viewing the
phenotype of populations both individually (Fig. 1E) and collectively (Fig. 1F), it was
apparent that the change in phenotype plateaued at around generation 15. For
continued evolution, we therefore gradually increased the amount of OA during
generations 26-31 (Fig. 1G), reaching a selection pressure comparable to the initial
phase of the experiment with 9 µl OA (~38% mean survivability across all
populations) (Fig. 1G). We then performed a second round of selection at this dose,
observing an increase in survivability across all populations, reaching ~70% by
generation 50 (Fig. 1H). As the populations adapted to the new condition, a second
plateau started around generation 40 (Fig. 1H-I). Over 50 generations we were

therefore able to evolve *D. simulans* to have approximately 5-fold higher survivability
to 9 µl OA compared to the unevolved populations (Fig. 1J), though still falling short
of the survival exhibited by *D. sechellia* (Fig. 1J). These plateaus can be explained
by the decreasing selective pressure exerted by fixed OA volumes as an ever-
greater proportion of flies adapt and survive the treatment. Determining whether an
upper limit exists that would constrain the evolution of a *D. sechellia*-like level of
phenotypic resistance would require additional selection rounds using progressively
higher OA concentrations over an extended (and unknown) number of generations.
OA is only one component of noni fruit, although probably the major toxic chemical
(Legal, et al. 1994). We therefore tested three of our evolved populations for
resistance to noni fruit pulp, finding that they exhibited much higher survival on this
substrate compared to the same original populations maintained under neutral
conditions (i.e., no OA selection) (Fig. 1K).
Specialization of *D. sechellia* on toxic noni is associated with a reduced egg-
laying capacity (R'Kha, et al. 1997; Markow, et al. 2009) and a shorter (albeit strain-
specific) lifespan (Watada, et al. 2020; Abe, et al. 2022; Shahandeh, et al. 2024).
Our success in evolving *D. simulans* with a partial *D. sechellia*-like resistance trait
begs the question as to whether these selected strains exhibit other phenotypic
changes as a trade-off for higher toxin resistance. We tested whether the evolved
resistant populations of *D. simulans* exhibit similar phenotypic changes in control
conditions. However, none of the three focal populations tested exhibited significant
decreases in the number of eggs laid (Fig. 1L), developmental viability (Fig. 1M) or
lifespan (Fig. 1N). Thus, at least within the timeframe of our experimental evolution
experiment, *D. simulans* did not display obvious phenotypic trade-offs for the gain in
OA resistance.
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Dynamic directional selection of allele frequencies
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We hypothesized that the observed phenotypic changes in *D. simulans* depended on
the directional selection of alleles of consistent loci across the ten populations, while
most of the genome evolves under neutrality. Allele frequencies at the selected loci
should therefore consistently change across all populations, allowing us to
statistically distinguish them from randomly evolving loci (Fig. 2A). We sequenced
the genomes of the populations of evolved (generation 25 and generation 50) and
unevolved (generation 0) flies, identifying 5,127,840 high-quality variants distributed
across all chromosomes. Pairwise identity-by-state analysis indicated that as the
number of generations increased, the proportions of shared alleles decreased, both
between time points and between pairs of populations (left-to-right and top-to-
bottom, respectively, in Fig. 2B).
To identify the selected genetic variants underlying the phenotypic adaptation,
we analysed the genome-wide allele frequency changes both between generations 0
and 25, and between generations 0 and 50, across all populations. The identification
of significant changes in frequency is non-trivial, and many methods have been
developed (Vlachos, et al. 2019; Kojima, et al. 2020; Barata, et al. 2023). We
therefore assessed the significance of frequency changes using two complementary
approaches: the Cochran–Mantel–Haenszel (CMH) test (Cochran 1954; Mantel and
Haenszel 1959), a widely-used method that tests for allele frequency changes
between two time points, and Bait-ER (Barata, et al. 2023), a Bayesian method that
tests for selection while explicitly modelling genetic drift. Both tests identified large
portions of the genome under positive selection at generation 25 and 50 (Fig. 2C,D).

The majority of prominent peaks were found by both methods and across time points
(Fig. 2E).
Based on the analysis of genome-wide linkage disequilibrium decay (Fig.
S1A), we grouped the significant SNPs identified by each of the two methods into
genomic blocks according to their proximity (i.e., <50 kb). We retrieved all the
candidate genes that fall within the block intervals, obtaining a total of 439 and 401
genes for the CMH and Bait-ER, respectively, at generation 25, and 1238 and 174
genes at generation 50 (Data S3). These genes likely include true targets of
selection, but also neighbouring genes affected by hitchhiking. The increase in CMH
candidate genes over time – in contrast to the decrease observed with Bait-ER –
might reflect the CMH test's tendency to overestimate significance and to be
insensitive to divergence among replicate populations.
The overlap of all significant genes between the two time points is <30%. This
might reflect that a large fraction of the candidate genes results from noise caused
by genetic drift and draft (hitchhiking) and/or reflect a shift in the phenotypic
contribution of genes under the stronger selection (i.e., the higher dose of OA) during
generations 27-50. Moreover, previous work on *Drosophila* using temporal sampling
showed that adaptation involving complex traits can lead to non-overlapping sets of
selected SNPs, in contrast with the simple expectation of continuous increases in
allele frequency until fixation (Orozco-terWengel, et al. 2012; Schlotterer, et al.
2016). Gene ontology analysis is complicated at this stage as most hits are likely
hitchhikers of the true causal genes. It is also hard to compare currently to previous
mapping studies in *D. simulans/D. sechellia* hybrids (Amlou, et al. 1997; Amlou, et al.
1998; Huang and Erezyilmaz 2015) or in the earlier *D. simulans* experimental
evolution experiment (Colson 2004), which only identified wide genomic regions. The
sole exception, a 170-kb window on 3R containing only 18 genes (including several
Osiris genes) (Hungate, et al. 2013) did not emerge in our analysis.
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278 **A genome-wide, pooled CRISPR screen for OA susceptibility and resistance**
279 **genes in *D. melanogaster* S2R+ cells**
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Given that these previous studies also contain extensive hitchhiking signals and
substantial noise across large genomic regions, comparing their results with our
analyses would likely yield numerous false positives and still leave a large pool of
putative loci, offering limited benefit for fine-mapping. We therefore sought an
orthogonal screening approach. Motivated by previous observations of OA toxicity
for insect Sf9 cultured cells (Kaczmarek, et al. 2022), we found that OA also leads to
reduced viability of *D. melanogaster* S2R+ cells (Fig. 3A). This phenotype prompted
us to perform a selection experiment in this cell line through exploitation of a
genome-wide, pooled CRISPR/Cas9 mutagenesis approach (Fig. 3B) (Viswanatha,
et al. 2018). In brief, a library of sgRNA-encoding plasmids (6-8 sgRNAs/gene) was
transfected into Cas9-expressing cells where site-specific plasmid integration
predominantly limits targeting to a single gene per cell. Two replicate cultures of this
pool of cells were either left untreated, or exposed to OA in the medium, at two
different concentrations (Fig. 3B). After 26 days in culture, DNA was extracted from
the cell pool, then sgRNA sequences were PCR amplified and sequenced to
determine those that were enriched after OA selection (reflecting genes whose loss
enhances cell viability, referred to hereafter as "susceptibility genes") or depleted
(reflecting genes whose loss leads to lower OA resistance, hereafter "resistance
genes").

To identify candidate OA susceptibility/resistance genes, we first filtered out
genes essential for general cell survival (Viswanatha, et al. 2018) and genes that
were not detectably expressed in S2R+ cells (using DGET (Hu, et al. 2017).
(Removal of such genes are inevitable limitations of this cell-based screening
approach; see Discussion). After comparing results from each condition and
replicate, we identified the top 5% susceptibility genes (Fig. 3C) and top 5%
resistance genes (Fig. 3D). Amongst the 87 susceptibility genes, there was no single
stand-out hit, suggesting that – at least in S2R+ cells – OA does not have a single
major cellular target (unlike, for example, insecticidal bacterial toxins that bind a
specific cell surface receptor (Xu, et al. 2022)). Gene-ontology analysis of these
susceptibility genes revealed a significant enrichment of genes involved membrane
trafficking (Fig. 3C). As OA has previously been hypothesized to disrupt cellular
membranes (Borrull, et al. 2015; Kaczmarek, et al. 2022), the recovery of these hits
suggests that general reduction in membrane transport reduces OA uptake into cells
and/or renders membranes less susceptible to OA treatment. 22 OA resistance
genes were identified, encoding diverse types of proteins (Fig. 3D), suggesting there
are multiple mechanisms by which cells can resist the toxic effects of OA.
We next examined which of the genes identified in the S2R+ cell selection
experiments might be relevant for evolution of OA resistance in whole animals by
determining those also present within significant blocks of the experimental evolution
analyses (Fig. 3E,F). Four susceptibility genes were identified: CG1620 (a predicted
SANT-MYB domain transcription factor), CG4291 (a putative component of the
splicing machinery), *Nph* (a chromatin remodelling factor (Emelyanov, et al. 2014))
and *Debcl* (a pro-apoptotic member of the Bcl-2 family (Galindo, et al. 2009)); these
are all likely to have indirect roles in conferring susceptibility to OA given their broad
functions. By contrast, two resistance genes have potentially direct roles in
conferring tolerance to OA, *kraken* and CG14130. *kraken* encodes a putative serine
hydrolase (Edwin Chan, et al. 1998) and stood out both because it was the top hit in
the resistance gene screen (Fig. 3D), and because of its location in one of the most
prominent peaks in the experimental evolution at generations 25 and 50 (Fig. 3F and
Fig. 2C-D). CG14130 encodes the orthologue of mammalian *Alkbh7*, which is
involved in regulating fatty acid metabolism (Solberg, et al. 2013; Zhang, et al. 2021),
and is located within a significant peak at generation 25 (Fig. 3F and Fig. 2C).
Therefore, intersecting the experimental evolution and CRISPR screening results
allowed us to narrow down tens or hundreds of potential candidates to two focal
genes for downstream investigations.
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337   **Sequence, expression and evolutionary analyses of candidate OA resistance**
338   **genes**
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To assess the candidacy of *kraken* and CG14130 (hereafter *Alkbh7*) in contributing
to OA resistance in animals, we first examined their sequence and expression in *D.*
melanogaster. Kraken belongs to a large family of serine hydrolases, which
encompass lipases, esterases and proteases that use a nucleophilic serine residue
in the active site for hydrolysis of substrates (Edwin Chan, et al. 1998). Consistent
with the conservation of this serine in Kraken (Fig. 4A and Fig. S2A), an activity-
based proteomic screen of the serine hydrolase superfamily in *D. melanogaster*
provided evidence for *in vivo* catalytic activity of Kraken (Kumar, et al. 2021). *kraken*
is broadly expressed in adults, with particular enrichment in the gut and Malpighian
tubules (Fig. 4B), similar to previous observations of *kraken* expression in the

embryo by RNA *in situ* hybridization (Edwin Chan, et al. 1998). Using the Fly Cell
Atlas (Li, et al. 2022), we examined sub-tissue adult expression of *kraken*: in the
Malpighian tubules, *kraken* is expressed at highest levels in the principal cells (Fig.
4C), which are involved in ion transport and detoxification. In the gut, *kraken*
transcripts are most robustly detected in various types of enterocytes, which secrete
digestive enzymes (Lemaitre and Miguel-Aliaga 2013) (Fig. 4C). However, Kraken
lacks an N-terminal signal sequence (Teufel, et al. 2022), suggesting that the protein
remains in the cytoplasm of these cells, unlike canonical digestive enzymes
(Lemaitre and Miguel-Aliaga 2013), and more similar to detoxification enzymes of the
Cytochrome P450 and Glutathione S-transferase families (Yang, et al. 2007). To
examine *kraken* expression *in vivo*, we performed RNA fluorescence *in situ*
hybridization, which detected transcripts in various region of the gut (crop, crop duct,
proventriculus, hindgut, ampulla) and Malpighian tubules (Fig. 4D).

Alkbh7 is a member of the alpha-ketoglutarate-dependent hydroxylase (Alkbh)
family (Fig. 4A), which are enzymes that act on diverse substrates in processes such
as biosynthesis, metabolism and epigenetic regulation (Xu, et al. 2021). As in
mammals (Solberg, et al. 2013), *D. melanogaster Alkbh7* displays broad tissue
expression (Fig. 4B). Murine Alkbh7 is localized to the mitochondrial matrix (Solberg,
et al. 2013). This property appears to be conserved in the *Drosophila* protein, which
has a mitochondrial targeting sequence (Fig. 4A and Fig. S2B) (Fukasawa, et al.
2015); consistently, we identified Alkbh7 within *D. melanogaster* proteomes of the
mitochondrial matrix (Chen, et al. 2015; Sen and Cox 2022).

Although the sequences of both Kraken and Alkbh7 are overall highly similar
across *D. melanogaster*, *D. simulans* and *D. sechellia* (Fig. S2A,B), we wondered
whether we could detect signs of selection at these loci in *D. sechellia* employing
population sequence datasets of *D. sechellia* and *D. simulans* (Schrider, et al. 2018).
Using the McDonald-Kreitman test (McDonald and Kreitman 1991) on the coding
sequences of both genes, we found evidence for gene-level positive selection on
kraken but not *Alkbh7* (Fig. S2C). Because selection might act pervasively at
individual sites rather than along the entire protein coding region, we also
implemented a codon-based analysis using FUBAR (Fast, Unconstrained Bayesian
AppRoximation) (Murrell, et al. 2013), which identified 1 and 4 codons under
diversifying positive selection in *kraken* and *Alkbh7,* respectively (Fig. S2D).
Although their functional relevance remains unclear, such mutations might reflect
underlying evolutionary dynamics at the protein level.

We next compared the expression levels of these genes both in laboratory
and wild-caught strains of *D. sechellia* and *D. simulans* (Fig. 4E), as well as in RNA-
seq datasets of various strains of *D. sechellia*, *D. simulans* and *D. melanogaster*
(Ma, et al. 2018; Watanabe, et al. 2019; Kalra, et al. 2024) (Fig. 4F, Fig. S3A,B).
While there were some exceptions, overall both *kraken* and *Alkbh7* show higher
expression levels in *D. sechellia* adults and larvae compared to the other species.
Importantly, in our experimentally-evolved *D. simulans* populations, we also found
that both *kraken* and *Alkbh7* were more highly expressed compared to unselected
populations (Fig. 4G). The expression differences between species, and between
unevolved and evolved *D. simulans*, reflect variation in basal levels of expression, as
prior exposure to OA did not lead to changes in expression in either *D. sechellia* or
D. simulans (Fig. S3C). RNA FISH of *kraken* in *D. sechellia* revealed similar spatial
expression as in *D. melanogaster*, with the addition of a robust signal in the midgut
(Fig. S3D). These observations suggest that elevated expression in this species is
due, at least in part, to a novel expression pattern in this digestive organ.

Functional contributions of *kraken* and *Alkbh7* to OA resistance

Given the expression differences observed in the naturally and experimentally evolved OA-resistant drosophilids, we asked whether artificially modulating gene expression would impact OA tolerance. We first performed ubiquitous RNAi of these genes in *D. melanogaster* (Fig. S4A) and assessed the effects on resistance to OA using a short-term, plate-based assay (Jones 1998) (see Materials and Methods). *kraken^RNAi* led to significantly diminished OA resistance compared to genetic controls, while *Alkbh7^RNAi* did not have a discernible effect (Fig. 5A). We next generated deletion mutants for these genes in *D. melanogaster* (Fig. S4B). Similar to the RNAi experiments, *kraken* mutants, but not *Alkbh7* mutants, displayed reduced resistance to OA (Fig. 5B). All genotypes remained fully viable when no OA was added (Fig. S5A).

Loss-of-function analyses might fail to uncover a role for *Alkbh7* in whole animals if there are redundant resistance mechanisms absent from S2R+ cells. We therefore took a complementary approach by overexpressing these genes in *D. melanogaster*. Strikingly, overexpression of *Alkbh7* resulted in survivability of *D. melanogaster* to levels close to those observed for *D. sechellia* under the same experimental conditions (Fig. 5C, Fig. S4C), indicating that this gene is sufficient to confer high-level OA resistance. By contrast, *kraken* overexpression did not have any detectable impact on OA resistance (Fig. 5C); a similar lack of effect was seen when overexpressing *D. sechellia kraken* (Fig. S2E). The absence of a phenotypic effect upon *kraken* overexpression is surprising, given its elevated expression in *D. sechellia* and in evolved *D. simulans* but this could reflect either insufficient overexpression levels (Extended Data Fig.□4c) and/or genotype-by-genotype interactions that are not recapitulated when manipulating a single locus in a *D. melanogaster* background.

A key question is whether these genes contribute to OA resistance of *D. sechellia*. Implementing CRISPR/Cas9-based genome-engineering in this species (Auer, et al. 2020; Auer, et al. 2021), we generated and validated null mutants in these genes in *D. sechellia* (Fig. S4B). These mutant flies were homozygous viable and fertile, with no obvious morphological or behavioural abnormalities, allowing us to test their resistance to OA in the plate assay. Although we did not detect any significant difference to genetically-matched wild-type flies (Fig. 5D), this was anticipated given the polygenic nature of OA resistance in *D. sechellia* (Amlou, et al. 1997; Amlou, et al. 1998; Jones 1998; Huang and Erezyilmaz 2015), and the modest or absent phenotypes of equivalent mutants in *D. melanogaster* (Fig. 5B). Likewise, rearing *D. sechellia* null mutants on noni pulp (Fig. S5B), we observed no significant differences between mutants and wild-type flies over >30 days, consistent with the conclusion that loss of either gene alone does not compromise *D. sechellia*'s capacity to persist in its natural host fruit.

We reasoned that more subtle differences might be detected by monitoring *D. sechellia* survivability at higher OA concentrations over several days. We therefore exposed flies to food containing 0.9% OA, approximately threefold higher than the ~0.3% OA reported for ripe noni fruit (3.06 g OA/kg fruit) (Pino, et al. 2009). Indeed, we observed higher mortality rates of both *kraken* and *Alkbh7* mutants compared to genetically matched wild-type flies over a period of 6 days (Fig. 5E). By contrast, both mutants survived equally well as controls in the absence of OA over this time period (Fig. 5E). To determine whether the observed OA sensitivity was attributable

to a generalised reduction in fitness, we examined survivorship under several
conditions of thermal stress (Fig. S5C): mutant and wild-type flies showed no
significant differences across all temperatures tested, indicating that the increased
mortality observed upon OA exposure does not result from an overall physiological
weakness of the knockout lines. These data provide the first evidence for the
contribution of specific genes to OA resistance in *D. sechellia*. However, future allelic
swap studies with *D. simulans* or *D. melanogaster* would be required to assess if and
how they contribute to species-specific differences in OA sensitivity (e.g., through
structural and/or regulatory differences in the genes).
459
460 **Discussion**
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Identifying specific loci underlying phenotypic evolution is challenging due to the
polygenicity of the vast majority of adaptive traits, with recent advocacy of new
experimental paradigms to bridge genotype-phenotype relationships (Tautz, et al.
2026). Investigating the evolution of chemical toxicity is an attractive model trait
because of the relatively easy phenotypic read-out. To gain insight into the genetic
basis of susceptibility and resistance to a natural toxin, OA, we have intersected the
results of selection over vastly different timescales: weeks in cell lines, years in
experimentally-evolved laboratory strains, and millennia during speciation.
Unlike insecticides that activate or antagonize broadly-expressed neuronal ion
channels to cause organismal death, the mode of action of OA appears to be
mechanistically very different, capable of inducing death at the level both of whole
organisms and of cultured cells. Beyond insect cell lines (our work and (Kaczmarek,
et al. 2022)), OA is also toxic to budding yeast (Mota, et al. 2024) and bacteria (Liu,
et al. 2013), suggesting that this toxin acts via a more fundamental cellular
mechanism. Indeed, a genome-wide, knock-out screen in yeast for genes required
for tolerance of OA highlighted contributions of several core cellular processes, such
as vesicle trafficking, mitochondrial function or chromatin regulation (Mota, et al.
2024), reminiscent of our CRISPR screen results. Consistent with this idea, the
results of our experimental evolution and CRISPR screening, as well as earlier
mapping efforts in *D. sechellia*/*D. simulans* hybrids (Amlou, et al. 1997; Amlou, et al.
1998; Jones 1998; Huang and Erezyilmaz 2015) revealed multiple genomic
regions/genes contributing to the resistance to OA.
Considered separately, the experimental evolution and cell-based CRISPR
screens identified many candidates. That there was very limited overlap between the
screen results is unsurprising given their very different designs. The loss-of-function,
cell-based screen of course has several limitations, as it cannot identify genes that
are not expressed in the cell line, those essential for cell viability in the absence of
OA, nor genes that function in biological processes only pertinent at a
tissue/organismal level (e.g., cuticle biogenesis, neuronal signalling). Nevertheless,
both lists are likely to contain multiple genes relevant to understanding OA
susceptibility and resistance, and hence mode(s) of action of OA. Here, we focused
on two resistance gene candidates, *kraken* and *Alkbh7*, which were common hits of
both screens. We note, however, that neither of these genes were identified in
previous studies of OA resistance (Amlou, et al. 1997; Amlou, et al. 1998; Jones
1998; Huang and Erezyilmaz 2015), which could reflect several factors, such as
differences in OA toxicity assay conditions, developmental stages, species and
molecular readouts (DNA or RNA), and the evident multilayered, polygenic nature of
this trait (discussed further in (Marconcini, et al. 2026)).

The (partial) necessity, but insufficiency, of *kraken*, and lack of necessity but
sufficiency of *Alkbh7* to confer OA resistance implies fundamentally different roles of
these proteins. The expression of *kraken* in the gut and renal system supports a
hypothesis of this protein as a detoxification enzyme, although future work will be
necessary to determine whether Kraken has a direct role in OA degradation. The
mild loss-of-function phenotype of *kraken* in both *D. melanogaster* and *D. sechellia*
likely reflects substantial redundancy of defence mechanisms, including other
potential detoxification mechanisms, especially given the plethora of enzymes
expressed in the fly gut and Malpighian tubules (Lemaitre and Miguel-Aliaga 2013;
Dow, et al. 2022). Indeed, extremely few associations between such enzymes and
substrates and/or physiological roles have previously been defined. One exception is
the Cytochrome P450 gene *Cyp6g1*, which has a key role in Malpighian tubules in
conferring resistance to DDT (Yang, et al. 2007). Our data on Kraken therefore
provide one of the first links between a natural toxic chemical and a putative
detoxification enzyme.
The mitochondrial Alkbh7 enzyme likely has a distinct role in OA resistance.
Insights from mammals are informative: mice lacking *Alkbh7* are obese, with higher
body fat levels, proposed to result from a defect in metabolism of fatty acids
(Solberg, et al. 2013). Molecularly, one study suggests that Alkbh7's metabolic role
stems from its regulation of RNA editing to control mitochondrial protein translation
(Zhang, et al. 2021), although it is unclear whether this is its only function. While the
Alkbh7 loss-of-function phenotype is mild in drosophilids, it is striking that artificially
elevated expression in *D. melanogaster* is sufficient to produce *D. sechellia*-like
levels of OA resistance. This result implies a potent effect of Alkbh7 on the
mitochondrial metabolic network to modulate toxin resistance, perhaps through
control of fatty acid metabolism. While such ideas can be tested in future studies, we
note that defects in lipid metabolism and mitochondrial function have been observed
upon exposure to other types of insecticide (Martelli, et al. 2020; Martelli, et al.
2022). Intriguingly, our implication of Alkbh7 in the niche adaptation of *D. sechellia*
has a potential parallel in bats, where this gene displays evidence of positive
selection during dietary diversification of different species (Gutierrez-Guerrero, et al.
2020).
Taken together, our work has demonstrated the power of orthogonal selection
approaches in the laboratory to understand toxin resistance, which might be relevant
to understand a historic paradigm of this phenomenon (R'Kha, et al. 1991), as well
as offering clues as to how the natural toxin OA exerts its lethal effects. Moreover,
the methodology – notably cell-based CRISPR screening, which can now be
extended to both gain-of-function approaches (Xia, et al. 2023) and mosquito cell
lines (Viswanatha, et al. 2021) – is applicable to explore the molecular mechanisms
of susceptibility and resistance of insects to the wide-range of artificial toxic
chemicals used in the environment (Gandara, et al. 2024). Such knowledge will have
application in the future characterization of novel types of insecticides.
542
543 **Author contributions**
544
545 M.M. and R.B. conceived the project. M.M. designed and performed the
546 experimental evolution experiment, generated and analysed all transgenic and
547 mutant flies, and performed the comparative gene expression and molecular
548 evolutionary analyses. S.C. generated the base populations and provided technical
549 support for the experimental evolution experiment. M.M., R.B., R.V. and S.E.M.

designed the cell screen. S.G. and R.V. tested and defined cell screen assay conditions. S.G. and M.B. performed the CRISPR/Cas9 screen and prepared DNA for sequencing. R.V. analysed the screen data, and M.M., R.B., R.V., S.E.M and N.P. contributed to interpretation of screen results. J.D. and C.R. performed histological experiments, supervised by D.H. M.M. and R.B. wrote the paper with input from other authors. All authors approved the final version of the manuscript.

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Declaration of interests

The authors declare no competing interests.

Data availability

Source data is provided in the supplementary data files. All other data, new materials and code are available from the corresponding author upon reasonable request.

585 **Materials and Methods**
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587 ***Drosophila* strains and husbandry**
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589 *Drosophila* stocks were maintained on standard wheat flour–yeast–fruit–juice
590 medium under a 12 h light:12 h dark cycle at 25°C. Wild-type strains, as well as
591 mutant and transgenic lines used in this study are listed in Table S1. Experimental
592 evolution was performed using a subset of published *D. simulans* isogenic strains
593 (Signor, et al. 2018) (Table S2).
594
595 **Generation of mutant and transgenic flies**
596
Loss-of-function mutants: we generated *D. melanogaster* mutants for *kraken* and
Alkbh7 using in vivo CRISPR/Cas9 mutagenesis (Port, et al. 2020) by crossing flies
expressing Cas9 in the germline (*nanos-Cas9*) to those expressing sgRNAs
targeting the desired gene. For *kraken*, we used
$P\{hsFLP\}1,y^1,w^{1118};P\{HD\_CFD02326\}attP40$ as a source of sgRNAs driven by
Act5C-GAL4; for *Alkbh7*, we used the CG14130 sgRNA line described in (Lenče
2021). Mutants were validated by genomic PCR and sequencing (Fig. S4B).
D. sechellia mutant flies were generated by WellGenetics Inc. using modified
methods of (Kondo and Ueda 2013). For *kraken* (GM16772/LOC6617307) the
upstream sequence CTGTAATGCTGGACGCGGAT[TGG] (PAM in square brackets)
and the downstream sequence CCTGGTCACCCCGGATCGAG[TGG] were cloned
separately into a U6 promoter plasmid. The 3xP3-RFP cassette, which contains a
floxed 3xP3-RFP flanked by a 5' homology arm (999 bp: from -1043 nt to -45 nt
relative to the *kraken* ATG) and a 3' homology arm (1063 bp: +1388 nt to +2450 nt)
were cloned into *pUC57-Kan* as a donor template for repair. For *Alkbh7*
(GM25300/LOC6605344) the upstream sequences
ACTAATAAGAGTCGACCGGG[TGG], AACAGTCAAACAGCTGTTCC[TGG] and
downstream sequences GTCCCAGTTTCAGGGTACGC[TGG],
ATGGAAATGCGACGTGTCCT[GGG] were cloned separately into a U6 promoter
plasmid. The 3xP3-RFP cassette, which contains a floxed 3xP3-RFP, 5' homology
arm (1037 bp: -1121 nt to -85 nt relative to the *Alkbh7* ATG) and 3' homology arm
(964 bp: +940 nt to +1903 nt) were cloned into *pUC57-Kan*. DNA plasmids encoding
the sgRNAs and the corresponding homology-directed repair template were
microinjected into embryos of *Dsec.07 pBAC(nos:Cas9,3P3-YFP)* (Auer, et al.
2020). In both cases, a single F1 line expressing the 3xP3-RFP selection marker
was obtained, which was validated by genomic PCR and sequencing (Fig. S4B).
D. melanogaster and *D. sechellia* mutant flies were backcrossed for six
generations to Canton-S and *Dsec*07, respectively.
625
Transgenic flies: to make the *UAS-Dmel\kraken* and *UAS-Dsec\kraken* transgenes,
we amplified (using oligonucleotides listed in Table S6) the regions from the start to
the stop codons from w^{1118} and *Dsec07* genomic DNA and subcloned them into a
pUASTattB vector (Bischof, et al. 2007). Transgenic flies were obtained by phiC31
integrase-mediated transgenesis into attP2 performed by BestGene Inc. and verified
by sequencing.
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Toxin resistance assays

Tube assay: resistance to OA for the experimental evolution setup was scored adapting a previous protocol (Amlou, et al. 1997; Colson 2004). In brief, flies were lightly anesthetized with CO_2 and ~50 animals were placed in standard cultured tubes (25 mm diameter × 95 mm height, Milian SA) and provided with 1 ml of a 3% glucose solution on a tissue paper (Kimtech, 7552). 3 µl of pure OA (Sigma-Aldrich, CAS 124-07-2, C2875) were directly pipetted onto the tissue while the flies remained anesthetized. Direct contact with the acid was scrupulously avoided until all flies recovered. The fraction of alive flies was scored after 24 h. For the experimental evolution experiment, the surviving flies were collected (see below).

Plate assay: to measure OA resistance in the genetically-manipulated *D. melanogaster* and *D. sechellia* strains, we used a plate assay (Jones 1998). Compared to the experiments in tubes, this assay offers greater temporal resolution and higher dynamic range (in contrast to the experimental evolution experiment, we did not strive to maintain a certain fraction of surviving flies), as well as greater statistical power from tests of fewer animals. Ten female flies were lightly anesthetized with CO_2 and placed in a Petri dish (60 mm diameter × 15 mm height) in which 2 µl of pure OA (or 30 µl for *D. sechellia*) were spotted in the centre of the underside of the lid. The flies were allowed to recover (typically <5 min), and immobile/dead were scored every 5 min for 1 h. The survival curves were calculated and plotted using the R package survminer (10.32614/CRAN.package.survminer). The statistical analysis was conducted fitting a mixed effect Cox regression model as implemented in the R package coxme (Therneau 2024). Genotype was included as a fixed effect, and replicate dishes were modelled as a random effect.

Noni resistance assay: noni fruit was collected from *M. citrifolia* plants grown in the University of Lausanne greenhouses. Five ripe (pale yellow) fruits were homogenized together, and the pulp puree frozen at -20°C. The noni assay was performed with the same setup as for tube assay described above, using 2 g of thawed noni puree placed on the bottom of the tube. Noni contains 3.06 g OA/kg fruit (Pino, et al. 2009); assuming an equivalent quantity in our fruits, 2 g noni pulp should contain 6.73 µl OA, approximately double the amount used for the experimental evolution between generations 1 to 25, and two thirds of the concentration during generations 30-50.

Long-term assay: To assess the long-term effects (i.e., beyond 24 h) of OA or noni pulp, two-day-old mated female flies were individually housed in vials containing either food supplemented with OA or 2g of noni puree. OA-supplemented food was prepared by thoroughly mixing OA into Formula 4-24® Instant Drosophila Medium, Blue (Carolina Biological Supply) to a final concentration of 0.9%. Flies were flipped into fresh food every two days, and mortality was monitored daily. Statistical analysis was run as for the plate assay.

Fecundity and *developmental viability*: 10 mated adult female flies (4-7 days old) were allowed to lay eggs for 24 h on agar plates (2% agar, 2.5% sucrose in a 1:3 grape juice:water mix). The total number of eggs on each plate was counted. From each plate, 100 eggs were collected and transferred to standard fly food. Emerging adults were counted and removed daily until no further flies emerged.

685
686 *Lifespan*: Two-day-old mated female flies were individually housed in vials containing
687 standard fly food and monitored daily until death. Fresh food was provided every 2–3
688 days.
689
690 **Experimental evolution**
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Selection procedure: to generate the base populations for experimental evolution, we
used 93 isogenic lines (Table S2) of the *D. simulans* panel established by (Signor, et
al. 2018). We first crossed each of the lines together following a round-robin design
to ensure equal contribution of each line to the starting genetic pool; 5 males and 5
females from each cross were then combined in ten independent replicates (Fig.
S1D). These ten populations were maintained in bottles (57 mm diameter ×103 mm
height, Milian SA) of standard fly food at high population size (~1000 individuals per
population) to minimize the effect of genetic drift. Each population was subdivided
into smaller groups of ~250 individuals per bottle (i.e., four bottles per population) to
reduce competition. At each generation, individuals from the four bottles were mixed
to prevent bottle-specific genetic drift. The experimental evolution experiment was
initiated 10 generations after the creation of these populations.
From each population, 500 individuals (2-7 day old) were randomly collected
and further split into 10 groups of 50 mixed-sex flies (i.e., ~5000 flies total). Each
group of 50 flies was then placed under selection with 3 μl of OA for 24 h (as
described above); the survivors were collected and allowed to breed in food vials for
6-8 days, transferring them to fresh vials every 2-3 days. In the next generation, 500
flies per population were collected only from the last seeded vial, to maximize the
probability of these flies being parented by the OA-resistant males and females and
not offspring generated by females that mated before the selection procedure. These
flies were then subjected to the same selection procedure. This process was
repeated for 25 generations, and survivability was scored at each generation.
A second round of experimental evolution was performed by increasing the
amount of OA to 9 μl between generations 31 and 50; this amount was determined
empirically during generation 26-31 to re-establish the survivability close to the initial
level (Fig. 1G). All successive generations were maintained following the same
procedure.
As a control, the ten original populations were maintained in parallel in the
same conditions as previously described (i.e. ~1000 individuals per population
subdivided into smaller groups of ~250 individuals per bottle).
722
DNA sequencing and analysis: we pooled ~200 individual flies per population per
time point (generations 0, 25 and 50). Genomic DNA (and total RNA for qPCR, as
described below) extraction was performed using the Quick DNA/RNA Miniprep plus
kit (Zymo Research, D7003) following the manufacturer's instructions. Samples were
stored at -80°C.
We performed paired-end whole genome sequencing on an Illumina NovaSeq
6000 following Nextera DNA flex preparation, at an average coverage of >200×.
Read quality was assessed with Fastqc (Wood 2011) and adapters were trimmed
with Trimmomatic (Bolger, et al. 2014) providing the Nextera adapters sequences.
We aligned the trimmed reads to the *D. simulans* reference genome (NCBI
GCF_016746395.2) using bwa-mem (Li 2013) and the results were sorted and

compressed using samtools (Danecek, et al. 2021). Duplicate reads were then
removed with Picard MarkDuplicates (https://broadinstitute.github.io/picard/).
Unplaced scaffolds and alignments with mapping quality <20 were removed with
samtools. We then combined all the aligned sequences into a samtools pileup file
and split this into individual chromosomes. Regions with zero coverage were
removed with custom scripts. Sync files for each chromosome were generated with
the mpileup2sync.pl script implemented in Popoolation2 (Kofler, et al. 2011) and
used as input files for all the Evolve-and-Resequence analyses.
To explore the genetic variation of our samples we generated vcf files with
vcftools 0.1.17 (Danecek, et al. 2011) and binary files with plink1.9 (Purcell, et al.
2007; Purcell 2020). We calculated the identity-by-state (IBS) matrix as implemented
in plink1.9 (function -distance 'ibs').

Evolve-and-Resequence analysis: allele frequency changes between generation 0
and generations 25 or 50 were investigated via Popoolation2 and Bait-ER (Barata, et
al. 2023).
We performed a Cochran–Mantel–Haenszel (CMH) test via the cmh-test.pl
script implemented in Popoolation2 comparing base and evolved populations with
parameters --min-count 12 and --min-coverage 50. We estimated a plausible
significance threshold approximating the top 0.000005% of the simulated p-values
based on the effect of genetic drift alone by running genome-wide forward
simulations under neutrality in mimicree2 (Vlachos and Kofler 2018). We used the
same genomic information from the original 93 lines as input and the average real
survivability ratios observed throughout the selection experiment as –selection-
regime parameter and 500 as –population-size parameter (Fig. S1B).
For Bait-ER, the input sync files were filtered to match --min-count 12 and --
min-coverage 50 and scaled to a uniform coverage of 100 as in (Vlachos, et al.
2019) to avoid numerical problems due to high coverage. We ran the algorithm on
each chromosome separately. Bait-ER models the effect of genetic drift directly
using estimates of the effective population size (N_e). We calculated N_e estimates for
each chromosome in each replicate population with the R package poolSeq (Taus,
et al. 2017) and used the median value of each chromosome as "Population_size"
parameter (Table S3). Other parameters included "Sampling correction" = 0 (i.e.,
binomial) and "Prior parameters" =0.001 (shape), 0.001 (rate). Data points were
considered significant when the absolute natural logarithm of the Bayes Factor
exceeded the conservative threshold of log(0.99/0.01) as in (Barata, et al. 2023).
The linkage disequilibrium decay in the 10 evolved populations was analysed
across the whole genome using PopLDdecay (Zhang, et al. 2019). Linkage
disequilibrium maps of focal regions were generated with LDBlockShow (Dong, et al.
2021) using the Solid Spine method based on D' statistics originally introduced in
Haploview (Barrett, et al. 2005).
For the genome-wide screen, all the genes within selection blocks were
considered potential candidates. Based on the results on linkage disequilibrium
decay, a block was arbitrarily defined as a genomic interval where at least two
significant SNPs were found at a distance less than 50 kb. Consecutive SNPs
respecting this definition were grouped into larger blocks and visualized as circus
plots using the R package circlize (Gu, et al. 2014). The genes overlapping such
blocks were extracted using BEDtools intersect (Quinlan and Hall 2010).
Orthologous genes in *D. melanogaster* were retrieved using a combination of the
OMA browser (Altenhoff, et al. 2021), DAVID (Huang da, et al. 2009; Sherman, et al.

2022), BLAST (Camacho, et al. 2009), NCBI (Sayers, et al. 2025) and FlyBase
(Ozturk-Colak, et al. 2024), and subsequent manual curation.
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787 **Genome-wide CRISPR/Cas9 screen in *D. melanogaster* S2R+ cells**
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Library design, construction, and transfection: design, synthesis and cloning of the
genome-wide sgRNA library are previously described (Viswanatha, et al. 2025). To
determine sub-lethal doses of OA appropriate for screening, the effect of a gradient
of OA supplementation into media on attP+ Cas9+ cells was determined using
CellTiter-Glo (Promega) following the manufacturer's protocol; luminescence
measurements were made using a Spectramax Paradigm (Molecular Devices).
For the screen, attP+ Cas9+ *D. melanogaster* S2R+ cells
(RRID:CVCL_UD30) were transfected using Effectene and a 1:1 molar ratio of
sgRNA library and pIntAC, as described (Viswanatha, et al. 2025). Briefly,
transfected cells were distributed in 5 ml aliquots across ten 100 mm culture dishes
and incubated overnight, then 5 ml of fresh media was added to each plate. After
4 days, cells were subjected to puromycin selection for 12 days, subculturing every
4 days, to establish a stable sgRNA-expressing cell library. For the selection assay,
2.5×10^7 cells were plated onto four 15 cm diameter cell culture dishes to ensure
sufficient sgRNA coverage, with each sgRNA represented approximately 1000 times
(Viswanatha, et al. 2025).
805
OA selection assay: library-containing cells were left untreated or treated with either
0.075 µg/µl or 0.05 µg/µl OA (Sigma-Aldrich, CAS 124-07-2, C2875). Each replicate
consisted of four 15 cm diameter cell culture dishes seeded with cells, achieving
~1000× coverage. Two replicate sets were included for each of the two OA treatment
conditions and one replicate for the untreated control. During the selection period,
cells were split 1-2 times per week to avoid overgrowth. An aliquot of cells from each
replicate was collected and stored for genomic prep following each split and at the
end of the assay. The final samples were collected after 26 days of OA or control
treatment.
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Genomic DNA extraction and sequencing: genomic DNA was extracted from
samples collected during the screen and at the final endpoint using a Zymo Quick-
gDNA Miniprep kit (Zymo Research, D3025). DNA fragments containing the sgRNA
sequences were amplified by PCR using what we calculated to be at least 1000
genomes per sgRNA as templates for each sample. PCR fragments were in-line
barcoded using a previously described approach (Viswanatha, et al. 2018;
Viswanatha, et al. 2019; Viswanatha, et al. 2025). Next-generation sequencing
(Illumina NextSeq) was performed at the Biopolymers Facility at Harvard Medical
School (RRID:SCR_007175). FASTQ sequence files were demultiplexed using in-
line barcodes as sequence tags as described previously (Viswanatha, et al. 2018;
Viswanatha, et al. 2025). All further processing was done using MAGeCK (v.0.5.4
and 0.5.9.4) (Li, et al. 2014) to generate readcount files and perform robust rank
aggregation (RRA) analysis and Maximum likelihood estimation directly on the
generated readcount files using default parameters. Plasmid readcounts were used
to normalize RRA values.
After removing essential genes at a 2% FDR (2,517 genes; Data S4), we
considered as candidates the top 5% genes overlapping across the two conditions
that were over-represented or under-represented in the OA-exposed cell populations

compared to the control cells (i.e., representing "susceptibility" and "resistance"
genes, respectively). The gene-set enrichment analysis was performed with
PANGEA 1.1 Beta using the *Drosophila* GO subsets (GO slim) and a p-value cut-off
of 0.05 (Hu, et al. 2023).
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839 **qPCR analysis**
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Reverse transcription of total RNA (extracted as described above) was performed
with the PrimeScript first-strand cDNA synthesis kit (Takara, 6110A) following the
manufacturer's instructions. Samples were stored at -80°C.
Quantitative PCR experiments were conducted on a QuantStudio 6 Flex Real-
Time PCR System using the PowerUp™ SYBR™ Green Master Mix following the
manufacturer's instructions. Each sample consisted of at least three biological
replicates. We performed three technical replicates per reaction and removed
replicates where the standard deviation was equal or greater than 0.5. We
determined target specific amplification efficiencies for each gene (Table S4, Table
S6). Relative quantification and statistical analysis were performed with the software
qbase+ v3.4 (Hellemans, et al. 2007) using Calibrated Normalized Relative
Quantities (CNRQs). Normalization was performed on the geometric mean of the
relative quantities for the three most commonly-used reference targets: *Act5C*,
αTub84B and *RpL32* (Lu, et al. 2018). We assessed reference target stability by the
geNorm expression stability value (M) and the coefficient of variation of the
normalized reference gene relative quantities (CV) (Table S5) (Vandesompele, et al.
2002; Hellemans, et al. 2007).
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859 **Population genetics analysis**
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Publicly available WGS data for 41 *D. sechellia* and 20 *D. simulans* genomes were
downloaded from the Sequence Read Archive (PRJNA215932 and PRJNA395473)
(Rogers, et al. 2014; Schrider, et al. 2018) with fasterq-dump (Leinonen, et al. 2010).
FASTQ files were processed and aligned as described in the "Experimental
evolution" section. Aligned files were phased with samtools phase (Danecek, et al.
2021) and the genes of interest were extracted with pilon (Walker, et al. 2014) to find
variation among populations. CDS sequences were aligned based on amino acid
information using translatorX (Abascal, et al. 2010). Gene sequences were aligned
using Clustal Omega (Madeira, et al. 2024). McDonald-Kreitman test, Tajima's D and
Fu and Li's statistics were calculated in DNAsp 6 (Rozas, et al. 2017). FUBAR
analyses (Fast, Unconstrained Bayesian AppRoximation) were performed in HyPhy
2.5 (Kosakovsky Pond, et al. 2020). The consensus sequences of *D. simulans* and
D. sechellia were computed by SnapGene (www.snapgene.com) "export consensus"
with a >50% threshold. The sequence alignment and secondary structures were
depicted using ESPript 3 (Robert and Gouet 2014).
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877 **RNA fluorescence *in situ* hybridization**
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To detect *kraken* mRNA *in situ*, we used the RNAscope Multiplex Fluorescence
Detection Reagents v2 (Advanced Cell Diagnostics, 323110) and RNAscope H_2O_2
and Protease Reagents (Advanced Cell Diagnostics, 322381). Surfaces and
equipment were cleaned with RNAzap (Invitrogen, AM9780) prior to dissection.
Intestinal tissues were dissected from *D. melanogaster* (Canton S) and *D. sechellia*

(*D. sechellia* 07) females in RNAlater (Sigma-Aldrich, R0901) diluted 1:50 in DEPC-
treated water and briefly rinsed in 1× PBS (prepared in ultrapure water)
(ThermoFisher Scientific, 70011044). Dissected tissues were mounted on poly-L-
lysine-coated slides in 1× PBS and fixed immediately in cold 4% formaldehyde
(Sigma-Aldrich, F8775) for 30 min. The fixative was removed, and tissues were
incubated with few drops of hydrogen peroxide for 10 min at room temperature (RT).
Tissues were then washed 3× 5 min in wash buffer (Advanced Cell Diagnostics,
320058) preheated to 40°C. For tissue dehydration, slides were incubated with 200
µl of ethanol solutions in the following order: 50% ethanol for 5 min at RT, 70%
ethanol for 5 min at RT and 100% ethanol for 5 min at RT (repeated twice). Slides
were air-dried for 5 min, and samples incubated with few drops of Protease III for 30
min at RT.

To hybridise probes, tissues were washed in 1× PBS for 3× 5 min. 50 µl of C1
(D. melanogaster-kraken-RA (NM_057917), Advanced Cell Diagnostics, 1850601-
C1); and 1 µl of C2 *(D. melanogaster-Rpl32* (NM_079843.4), Advance Cell
Diagnostics, 533451-C2) RNAscope FISH probes were preheated to 40°C for 10
min, cooled at RT for 5 min, applied to tissues and the slides were placed in a
humidified chamber for hybridization at 40°C for 2 h. Slides were hybridized with
AMP reagents in sequential steps. A few drops of AMP1, warmed to RT, were
applied to slides, which were incubated at 40°C for 30 min, followed by washes in
wash buffer 3× 5 min. Next, a few drops of AMP2, warmed to RT, were applied to
slides and incubated at 40°C for 30 min, followed by washes in wash buffer 3× 5 min.
Finally, a few drops of AMP3, warmed to RT, were applied to slides, and incubated
at 40°C for 15 min, followed by washes in wash buffer 3× 5 min.

A few drops of HRP-conjugated probe was applied for C1 and slides were
incubated at 40°C for 15 min. After, rinsing in wash buffer 3× 5 min, Opal dyes 520
nm (Akoya Biosciences, OP-001001) and/or 570nm (Akoya Biosciences, OP-001003
(1:1000 in TSA multiplex buffer, Advanced Cell Diagnostics, 322809) were applied to
slides and incubated at 40°C for 30 min. Slides were then washed and treated with
HRP blocker (a few drops per slide) for 15 min at 40°C and washed in wash buffer
for 3× 5 min. Tissues were counterstained with DAPI for 5 min at RT and washed
once in the wash buffer, before mounting in Vectashield (Vector Laboratories,
H1000).

A Leica STELLARIS-DIVE confocal microscope was used to generate all
confocal images, using a water-immersive 40× objective or an oil-immersive 63×
objective. The images were acquired using both HyDs as well as PMTs tailored for
the fluorophores of each sample.

Statistics and reproducibility

Data were analysed and plotted in R 4.0.3 (R Core Team 2021).

Figure legends

Fig. 1. Experimental evolution of *D. simulans* with increased OA resistance.
A Phylogeny of the drosophilid trio studied in this work. Mya: million years ago.
B Schematic of octanoic acid (OA) resistance tube assay (see Materials and Methods) and selection procedure.
C Species differences in resistance of *D. simulans* 04 and *D. sechellia* 07 strains to 3 µl OA.
D Schematic illustrating the generation of the base populations for Evolve-and-Resequence (see Materials and Methods).
E Changes in survivability to 3 µl OA of the ten *D. simulans* populations over generations 0-25. Box plots represent the interquartile ranges of the survivability for all the replicates; blue lines shows the generalized additive model (GAM) regression; grey shading represents the 95% confidence interval for the mean responses at each time point.
F As in **E** for the 10 populations combined.
G OA resistance for the combined 10 populations with increasing OA volumes during generation 25-31.
H As in **E**, with 9 µl OA during generations 31-50.
I As in **H** for the 10 populations combined.
J Comparison of OA resistance of unevolved *D. simulans* (i.e., the original 10 outcrossed populations maintained on standard fly food; see Materials and Methods), evolved (generation 50 (G50)) *D. simulans*, and *D. sechellia*.
K-N Phenotypic comparison of unevolved and evolved (G50+) *D. simulans* populations 1-3 for **K** noni resistance (G54), **L** fecundity (G54), **M** developmental viability (G57), and **N** lifespan (G57). For **K-M**, significance was assessed using the unpaired t-test correcting for multiple testing. NS p>0.05, ***p<0.001. For **L-N**, tests were run on standard media. No significant differences between unevolved and evolved flies were detected using the Cox regression model. See Materials and Methods for details and Data S1 for raw data.

Fig. 2. Dynamic directional selection of allele frequencies during experimental evolution.
A Schematic of the analysis.
B Identity-by-state analysis. The heatmap depicts the proportion of shared alleles between populations, with values ranging from 0.8 (blue, lowest shared proportion) to 1 (red, identical alleles).
C,D Manhattan plot showing the probability of allele frequency changes between **C** generations 0 and 25, and **D** generations 0 and 50. Top panel shows the results of the CMH test; orange dots indicate significant variants at a threshold of $-\log_{10}(p) \geq 40$, a conservative approximation of the top 0.000005% simulated SNPs under neutrality (Fig. S1B). Bottom panel shows the absolute log-Bayes factors estimated by Bait-ER; orange dots indicate significant variants at the conservative threshold log(0.99/0.01). Raw data in Data S2.
E Circular plots illustrating overlap of predicted genomic regions under directional selection between the two statistical approaches at generation 25 (left) and generation 50 (middle), or between generation 25 and 50 (right). Raw data in Data S3.

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Fig. 3. A genome-wide pooled CRISPR screen for OA susceptibility and
resistance genes in *D. melanogaster* cultured cells.
A Left: images of S2R+ cells cultured in the absence or presence of OA. Scale bar,
50 µm. Right: dose-dependent effects of OA on S2R+ cell viability, through
measurement of ATP levels with a CellTiter-Glo assay (see Materials and Methods).
B Schematic of CRISPR screen design.
C Top 5% "susceptibility" genes (see Materials and Methods and Data S4). Box plots
represent the interquartile ranges for all the replicates of the Z-scores obtained
through maximum likelihood estimation to test for selection. The inset (bottom-right)
displays the gene-ontology analysis of these genes.
D As in **C**, for the top 5% "resistance" genes.
E Venn diagram showing the overlap between all the genes identified in the
experimental evolution at G25 and G50, and genes associated with susceptibility
and resistance in the CRISPR screen.
F Subportion of the top panel (CMH test) of the Manhattan plot from Fig. 2C for
chromosome arms 2L and 3L, with the red arrows highlighting the regions magnified
on the right showing the significant variants and linkage disequilibrium blocks
(triangles) for *kraken* and *Alkbh7* (CG14130).
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Fig. 4. Sequence and expression analyses of *kraken* and *Alkbh7*.
A AlphaFold protein models for *D. melanogaster* Kraken (AF-O18391-F1-model_v4)
and Alkbh7 (AF-Q9VTP1-F1-model_v4) (Jumper, et al. 2021; Varadi, et al. 2022),
highlighting a conserved putative catalytic serine in the active site in Kraken, and the
mitochondrial targeting sequence in Alkbh7 (predicted by MitoFates (Fukasawa, et
al. 2015)). See also Fig. S2A,B.
B Tissue-specific expression of *D. melanogaster kraken* and *Alkbh7* from bulk RNA-
sequencing data from the Fly Atlas 2.0 (Krause, et al. 2022).
C tSNE plots illustrating *D. melanogaster kraken* expression in single-cell
transcriptomes of the Malpighian tubules and gut from the Fly Cell Atlas (Stringent
10x datasets) (Li, et al. 2022).
D Left: RNAscope detection of *kraken* (green) transcripts in the gut and Malpighian
tubules, with nuclei counterstained with DAPI (blue). Right: higher-magnification
images showing *kraken* transcript expression in the indicated tissues. Scale bars,
200 µm. This expression pattern was observed in tissues from >20 individuals.
E Expression levels of the *kraken* and *Alkbh7* in the indicated laboratory strains (left
panel) and wild caught strains (right panel) of adult female *D. sechellia* and *D.*
simulans measured by qPCR (Data S6). Expression is represented as calibrated and
normalized relative quantities (CNRQ). Significance was assessed using the
unpaired t-test correcting for multiple testing and represented using letter codes.
F Expression levels of *kraken* and *Alkbh7* in the indicated strains and species at
larval or adult stages as indicated from published RNA-seq data. (References: 1,
{Watanabe, 2019 #6929}, 2 {Ma, 2018 #8721}, 3 {Kalra, 2024 #10088}). Significance
was assessed using the unpaired t-test correcting for multiple testing and
represented using letter codes when multiple replicates were present in the original
data.
G Expression levels of the candidate genes at generation 0, 25 and 50 of the
experimentally-evolved *D. simulans* populations measured by qPCR (Data S6).

Significance was assessed using the paired t-test correcting for multiple testing and
represented using letter codes.
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Fig. 5. Functional validation of the contribution of *kraken* and *Alkbh7* to OA
resistance.
A-C Survival curves of adult flies of the indicated genotypes tested in the plate assay
with 2 µl OA for *kraken* and *Alkbh7* RNAi (**A**), mutants (**B**) and overexpression (**C**).
Each replicate consisted of 10 2-7 day-old female flies. N replicates ≥6, n flies ≥60.
Shading indicates 95% confidence intervals. Genotypes: **A** *act-Gal4/+;UAS-*
kraken^{RNAi}/+, UAS-kraken^{RNAi}/+, act-Gal4/+, act-Gal4/+;UAS-Alkbh7^{RNAi}/+, UAS-
Alkbh7^{RNAi}/+, act-Gal4/+ **B** *kraken^{1}, Alkbh7^{1}, Canton-S* (wild-type genetic
background control for both mutants), **C** *act-Gal4/+;UAS-kraken$^{o/e}$/+, UAS-*
kraken$^{o/e}$/+, act-Gal4/+, act-Gal4/+;UAS-Alkbh7$^{o/e}$/+, UAS-Alkbh7$^{o/e}$/+, act-Gal4/+.
Significance is based on the resulting mixed effect Cox regression model. NS
p>0.05, *p<0.05, **p<0.01, ***p<0.001. The same data for *D. sechellia* are shown in
both plots in **C**.
D Survival curves of wild-type (*Dsec07*), and *kraken* (*Dsec\kraken^{RFP}*) and *Alkbh7*
(*Dsec\Alkbh7^{RFP}*) mutant *D. sechellia* in the plate assay with 30 µl OA. Statistics as in
in **A**.
E Long-term survivability of the same genotypes as in **D** (n = 60 per condition) in
control conditions (no OA) (left) and food supplemented with 0.9% OA (right).
Statistics as in **A**.
Raw data in Data S7.
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1052 **Supplementary Information**
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1054 **Supplementary Tables**
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1056 **Table S1.** *Drosophila* strains.
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Genotype	Reference/Source
D. sechellia 07 (*Dsec07*)	*Drosophila* Species Stock Center (DSSC):14021-0248.07
D. simulans 04 (*Dsim04*)	DSSC:14021-0251.004
D. melanogaster Canton-S *(DmelCS)*	
D. sechellia 28 (*Dsec28*)	DSSC: 14021-0248.28
D. simulans 196 (*Dsim196*)	DSSC: 14021-0251.196
D. sechellia Aride 3311	(Shahandeh, et al. 2026)
D. sechellia Mahe 1311	(Shahandeh, et al. 2026)
D. sechellia Praslin 7612	(Shahandeh, et al. 2026)
D. simulans H1	D. Matute
D. simulans LZV47	D. Matute
D. simulans MD221	D. Matute
y[1] w[*]; P{Act5C-GAL4-w}E1/CyO (act-Gal4)	RRID:BDSC_25374
y[1] w[1118] (used as background for act-Gal4)	RRID:BDSC_6598
y[1] w[67c23] (background of Alkbh7$^{o/e}$)	RRID:BDSC_6599
y[1] w[67c23]; P{y[+mDint2] w[+mC]=EPgy2}EY05855 (Alkbh7$^{o/e}$)	RRID:BDSC_16679
y,w[1118];P{attP,y[+],w[3']}} (background of UAS-kraken^{RNAi} and UAS-Alkbh7^{RNAi})	VDRC:v60100
P{KK100356}VIE-260B (UAS-kraken^{RNAi})	VDRC:v105604
P{KK102727}VIE-260B (UAS-Alkbh7^{RNAi})	VDRC:v109669
y[1] M{w[+mC]=nanos-Cas9.P}ZH-2A w[*] (nanos-cas9)	RRID:BDSC_54591
P{hsFLP}1, y1 w1118; P{HD_CFD02326}attP40/CyO-GFP (sgRNA line for Dmel\kraken CRISPR mutant)	VDRC:v342579
CG14130 cris (gRNA 1+2)/CyO *(sgRNA line for Dmel\Alkbh7 CRISPR mutant)*	(Lenče 2021)
y[1] w[67c23];;UAS-kraken (kraken$^{o/e}$)	This study
y[1] w[67c23];;UAS-Dsec\kraken (Dsec\kraken$^{o/e}$)	This study
y[1] w[67c23]; P{y[+t7.7]=CaryP}attP2 (background of kraken$^{o/e}$)	This study
Dmel\kraken^{1}	This study
Dmel\Alkbh7^{1}	This study
Dsec\kraken^{RFP}	This study
Dsec\Alkbh7^{RFP}	This study

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1060 **Table S2. *D. simulans* lines used to generate the base populations.**
1061 *- see separate Excel file.*
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1063 **Table S3. Effective population size estimates from the experimentally-evolved**
1064 **populations.**

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	2L	2R	3L	3R	4	X
pop1	80.65681	101.5796	107.7145	100.5748	369.3642	107.3616
pop2	135.7243	56.29977	76.40593	91.41878	333.638	53.66758
pop3	74.08474	119.4491	95.20849	128.4216	193.5982	99.40552
pop4	83.05221	89.62312	112.3929	116.7979	171.2875	112.1581
pop5	123.0386	106.8179	107.448	81.31624	364.4287	83.55315
pop6	98.8018	77.03804	60.71772	147.1788	124.9405	75.48775
pop7	119.4038	101.4515	96.22689	114.8849	431.7489	114.6792
pop8	136.6513	176.4308	70.8519	130.9822	639.7962	78.70108
pop9	95.58733	101.7351	122.8352	78.80166	109.3757	69.50865
pop10	141.862	101.6789	107.512	83.0109	376.8733	71.34583
mean	108.8863	103.2104	95.73134	107.3388	311.5051	86.58684
median	109.1028	101.6293	101.8374	107.7299	349.0333	81.12711

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1067 **Table S4. qPCR primer efficiency.**
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Gene	Efficiency	SE
kraken	1.998	0.041
Alkbh7	1.977	0.016
Act5C	1.976	0.022
αTub84B	2.027	0.023
RpL32	2.053	0.031

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1070 **Table S5. geNorm analysis of reference target stability.**
1071 *- see separate Excel file.*
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1073 **Table S6. Oligonucleotides.**
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Supplementary Figure legends

Fig. S1. Linkage disequilibrium decay, and comparison of simulated and observed p-value distributions in the experimentally evolved populations.
A Linkage disequilibrium decay at generations 25 (left) and 50 (right). Black curves represent the trajectory of r2 values (y-axis left) as genomic distances increase (x-axis). Red curves represent the same but for D' values (y-axis right).
B Violin plots showing the distribution of p-values from the CMH test, comparing allele frequency changes between generation 25 and 0 (left), and generation 50 and 0 (right). Blue violins represent p-values derived from forward evolution simulations, while orange violins correspond to p-values from experimentally evolved populations. Red dashed lines mark the top 0.000005% of simulated p-values, serving as the significance threshold for the experimental data.

Fig. S2. Sequence and molecular evolutionary analyses of *kraken* and *Alkbh7*.
A,B Protein sequence alignment and secondary structure of Kraken (**A**) and Alkbh7 (**B**) from the reference sequence of *D. melanogaster* (NP_477265.1 and NP_648511.2) and the consensus sequences of *D. simulans* and *D. sechellia*. Sequence data from PRJNA215932 and PRJNA395473 (see Materials and Methods). Black triangles indicate the residues where signs of selection were detected by the FUBAR analysis (see **D**).
C Results from the McDonald-Kreitman test. NI = neutrality index; alpha = proportion of substitutions driven by positive selection; p-value = p-value resulting from a two-tailed Fisher's exact test.
D FUBAR analysis to detect pervasive selection as implemented in HyPhy. position = codon position; alpha = mean posterior synonymous substitution rate; beta = mean posterior nonsynonymous substitution rate; posterior prob = posterior probability of positive selection; codons = alternative codons, Freq Dsec = frequency of the alternative codons in *D. sechellia* in the same order as codons; Freq Dsim = frequency of the alternative codons in *D. simulans* in the same order as codons.
E Survival curves of *D. melanogaster* overexpressing *Dmel\kraken* and *Dsec\kraken* exposed to 2 µl OA. Each replicate consisted of 10 2-7 day-old female flies. Genotypes: *act-Gal4/+;UAS-Dmel\kraken/+*, *act-Gal4/+;UAS-Dsec\kraken/+*. N replicates ≥6, n flies ≥60. Shading indicates 95% confidence intervals. Significance is based on the resulting mixed effect Cox regression model. NS p>0.05. Raw data in Data S5.

Fig. S3. Additional expression analyses of *kraken* and *Alkbh7*.
A Bar plots showing the expression levels of *kraken* and *Alkbh7* across various adult *Drosophila* tissues and species. Data from (Bontonou, et al. 2024).
B Bar plots showing the expression levels of *kraken* and *Alkbh7* across various larval *Drosophila* tissues and species. Data from (Watanabe, et al. 2019) ("medium diet" condition).
C Bar plots showing the effect of exposure to 9 µl OA (in the tube assay) on *kraken* and *Alkbh7* expression in *D. sechellia* and three evolved (G50) *D. simulans* populations.

D Left: RNAscope detection of *kraken* (green) and *Rpl32* (magenta) transcripts in the gut and Malpighian tubules. Right*:* higher-magnification showing *kraken* transcript expression in the indicated tissues*.* Scale bars, 200 μm. This expression pattern was observed in tissues from >10 individuals.

Fig. S4. Validation of RNAi and overexpression lines and generation of mutants for *kraken* and *Alkbh7*.
A Quantitative RT-PCR analysis of RNAi-mediated knockdown of *kraken* and *Alkbh7* in *D. melanogaster*. Genotypes as in Fig. 5A.
B Schematics of the CRISPR/Cas9 design to generate *D. melanogaster* and *D. sechellia* mutants for *kraken* and *Alkbh7*. Validation was performed by PCR and Sanger sequencing. The *Dse*c*kraken^{RFP}* mutant comprised a 1431 bp deletion (-44 to +1,387 nt relative to the ATG), replaced with the 3xP3-RFP cassette. The *Dsec**Alkbh7^{RFP}* mutant was a 1023 bp deletion (-84 nt to +939 nt relative to the ATG), replaced with the 3xP3-RFP cassette.
C Quantitative RT-PCR analysis of Gal4/UAS-mediated overexpression of *D. melanogaster kraken* and *Alkbh7*. Genotypes as in Fig. 5C.

Fig. S5. Control survival curves in the absence of OA.
A Survival curves of adult flies of the indicated genotypes tested in the plate assay without OA. Each replicate consisted of 10 2-7 day-old female flies. N replicates ≥5, n flies ≥50. All genotypes display full viability, rendering datapoints invisible beneath the topmost displayed genotype (*UAS-kraken^{RNAi}*). Genotypes as in Fig. 5A-B.
B Survival curves of wild-type (*Dsec07*), and *kraken* (*Dsec**kraken^{RFP}*) and *Alkbh7* (*Dsec**Alkbh7^{RFP}*) mutant *D. sechellia* on homogenized noni pulp. N = 50 per strain. Significance is based on the resulting mixed effect Cox regression model. NS p>0.05.
C Survival curves of *kraken* (*Dsec**kraken^{RFP}*, top) and *Alkbh7* (*Dsec**Alkbh7^{RFP}*, bottom) mutant *D. sechellia* compared to wild-type (*Dsec07*) flies at 30, 33, 36 degrees Celsius. N ≥ 50 per strain/condition. Statistics as in **B**.

Data S1. Source data for Fig. 1
Data S2. Experimental evolution data available at:
https://drive.google.com/file/d/1Gy3k9eYvQ8xQMjIsTCqWevRaO_b1__JY/view?usp=sharing
Data S3. Source data for Fig. 2
Data S4. Source data for Fig. 3
Data S5. Source data for Fig. S2
Data S6. Source data for Fig. 4
Data S7. Source data for Fig. 5 and Fig. S5.

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