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**Outcrossing Complicates Mutation Purging by Trapping Single Nucleotide Polymorphisms in Structural Variant Mutations**

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**Abstract**

Outcrossing influences a population's ability to purge deleterious mutations, but its impact on larger structural genomic changes like insertions, deletions, and inversions remains unclear. To test this, we developed a theoretical model of structural variant (SV) mutation dynamics, implemented it in population simulations, and tested our predictions by exposing three *C. elegans* strains to ethyl methanesulfonate and formaldehyde to induce diverse mutations. Contrary to expectations for single nucleotide polymorphisms (SNPs), our model and whole-genome sequencing revealed that outcrossing could retain SVs in populations even after recovery from mutagenesis. The strain with highest levels of outcrossing accumulated the largest number of SVs, including megabase-scale rearrangements, and showed the highest transposable element activity. Moreover, SNPs persisted when embedded within retained SVs. These findings contradict classical population genetic expectations, showing that outcrossing can hinder SV removal and complicate mutation purging by trapping SNPs within SVs.

**Introduction**

The accumulation of deleterious mutations can have severe consequences for populations, leading to long-term fitness decline or, in extreme cases, extinction[1]. The ability to effectively eliminate or purge harmful mutations is critical for reducing genetic load and ensuring long-term evolutionary fitness[2]. Based on their size and scale, mutations are broadly categorized into two types: single nucleotide polymorphisms (SNPs) and structural variants (SVs). SNPs are small-scale changes involving a single base pair, often resulting in localized effects on amino acids or regulatory elements[3]. In contrast, SVs are large-scale genomic changes including deletions, duplications, inversions, and translocations. These can impact larger portions of the genome, frequently altering multiple genes or regulatory networks[4].

While mutation purging operates on both SNPs and SVs, the process may be more straightforward for SNPs due to their smaller scale, localized effects, and simpler genomic dynamics[5]. In contrast, purging

SVs may be more complex due to their larger genomic footprint, reduced recombination, and pleiotropic

effects[6]. Evolutionary theory has traditionally viewed the spectrum of genomic changes under the broad

category of "mutation". Much of the research on mutation purging has disproportionately focused on

SNPs as they are more easily and accurately detected[7]. However, the advent of large-scale genome

sequencing has revealed the ubiquity of SVs. Despite the foundational role of mutation purging in

evolutionary theory[8–10], our understanding of how it operates on SVs remains unexplored. Addressing this

is essential to refining theoretical models, validating predictions with empirical evidence and

understanding how mutation size and severity interact with established evolutionary predictions.

Mating systems influence mutation purging by affecting genetic diversity, recombination, and the

expression patterns of deleterious mutations[11,12]. Current consensus in the field is that outcrossing

populations, through recombination, create novel genetic combinations which facilitates the purging of

deleterious variation[13]. However, outcrossing can also potentially dilute the effects of deleterious

mutations and slow their purging[14]. Reduced recombination and genetic diversity in populations that

reproduce through self-fertility may lead to the accumulation of mildly deleterious mutations through

genetic drift[15]. Selfing however, can also increase homozygosity, exposing harmful recessive mutations to

selection and enabling more effective purging[15]. These processes may differ for mutations of different

physical sizes. SNPs, with largely localized effects, may be more directly exposed to selection. Purging

deleterious SVs may be complicated by their larger size, impacts, and pleiotropic effects, each of which

can vary significantly with mating system[16,17]. In this study we combine theoretical modelling with

experimental evolution to address two key questions: First, how do quantitative levels of outcrossing

influence the genome's ability to purge deleterious mutations? Second, how does this ability vary with

mutation size and genomic location?

To investigate these questions, we developed a mathematical model describing the evolution of

structural variants in the genome and conducted population simulations to predict how SV mutations

behaved under varying levels of outcrossing. We experimentally tested our predictions using three strains

of *Caenorhabditis elegans* which differ in natural frequency of males in the population: N2, AB1 and

CB4856. As an androdiecious species *C. elegans* populations consist of both males and hermaphrodites,

allowing for reproduction through either self-fertilization or outcrossing with males. While self-

fertilization predominates in *C. elegans*, the frequency of males and outcrossing varies drastically across

different strains[18] making *C. elegans* a prime system to test these ideas.

We subjected *C. elegans* strains to two contrasting mutagens, EMS to induce single nucleotide

mutations and formaldehyde to introduce complex lesions in the genome[19]. We mutagenized worms for

five consecutive generations and allowed them to recover from induced mutations for three generations.

We then tracked the population's relative fitness, changes in male frequency, and outcrossing rates to

investigate their ability to cope with mutational input and purge deleterious mutations. We employed

both long-read and short-read sequencing to quantify the size, location, and types of mutations that

persist after recovery.

<u>Results</u>

*Theoretical Expectations*

Although there is a body of work modelling the influence of outcrossing on mutation in populations these

models typically assume that mutations are occurring at the level of an individual SNP or broader, abstract

'allele' level[20,21]. Importantly, the assumptions of these models do not fit our current understanding of SVs

in natural populations. Therefore, we developed a set of mathematical equations with assumptions better

suited to what we currently know about SVs, describing evolution of SVs in the genome. The full model is

described in the Methods; here, we outline critical assumptions.

Our model described a genomic system where mutations occur as SVs, either insertions,

deletions or inversions. Each individual in a population of size K carried n structural mutations that

occurred at a rate µ each generation and were removed at a rate $v$, for example by back mutation via

gene conversion[22]. We did not explicitly model the physical size of mutations but assumed that

mutations had discrete phenotypic effect sizes that cumulatively influenced fitness. The distribution of

fitness effects was drawn from a gamma distribution, an empirically supported assumption[23–25]. With

this function the majority of mutations had small effects on the phenotype and fitness and a few large

effect mutations were possible (SFigure 1).

Individual SVs interacted to decrease fitness faster-than-linearly and the fitness landscape had a

negative curvature or synergistic epistasis (SFigure 2). The fitness of an individual carrying n SVs was

$w(n) = \exp^{-(an+(bn^2)/2)}$ where a and b were parameters determining the curvature of the fitness

landscape (Charlesworth 1990). We implemented this model in stochastic simulations using the SLiM 4.3

software[26] and measured the impact of outcrossing rates on SVs in the population.

Contrary to the predictions of population genetic models studying SNPs and alleles[20,21], our model

predicted that outcrossing retains SVs in populations (Figure 1). Our fully self-fertile populations rapidly

purged SVs while populations with >30% of individuals reproducing through outcrossing retained SVs

throughout their evolution. Between these two extremes our model predicted that outcrossing could

have a quantitative impact on the retention of SVs. Increasing levels of outcrossing increased both the

number of populations retaining SVs and the number of mutations retained.

Figure 1. A) Simulated populations evolving with less than 10% of individuals reproducing through
outcrossing do not retain SVs. B) The mean number of structural variant mutations is lower in populations
with lower quantitative levels of reproduction through outcrossing. Here K=1000, a=0.001, b =0.0006,
μ=0.01, ν=0.001 and recombination rate r=0.0001.

*Experimental Test of the Outcrossing Prediction*

To test the predictions of our theoretical model we subjected three laboratory strains of *C. elegans* (N2

(Bristol), AB1, and CB4856 (Hawaiian), obtained from the *Caenorhabditis* Genetics Center, University of

Minnesota) to five generations of mutagenesis followed by three generations of recovery at standard

laboratory plates with large and high-density populations. During both the mutagenesis and recovery

phases, worms were maintained on standard laboratory plates. However, during the mutagenesis phase,

worms were suspended in mutagen solutions for 4 hours every fourth day before being transferred back

to standard plates. In contrast, during the recovery phase, worms were not exposed to the mutagen and

were allowed to grow normally on standard laboratory plates. We measured both male frequency and

outcrossing rates at two key time points: before and after mutagenesis. Additionally, we assessed relative

fitness for each strain at three time points: before mutagenesis, directly after mutagenesis, and after

recovery. Each strain was maintained in two distinct lineages: one mutagenized with EMS and the other

with formaldehyde, with four replicate populations per lineage to ensure statistical rigor.

*Male frequency varied between C. elegans strains*

To assess the impact of five generations of mutagen exposure, we measured male frequency in both non-

mutagenized and 5 generation mutagenized *C. elegans*. Age-synchronized L4 hermaphrodites were

assayed individually for lifetime progeny sex ratio across replicate plates. We used a zero-inflated beta

regression model[27] to analyze male frequency, followed by pairwise post hoc comparisons using Tukey's

HSD comparison[28]. We found that non-mutagenized parents of CB4856 had the highest male frequency,

which was significantly greater than non-mutagenized parents of AB1 ($p < 0.0001$) and non-mutagenized

parents of N2 ($p < 0.01$).

When mutagenized with EMS for 5 consecutive generations, CB4856 showed a significant drop in

male frequency compared to non-mutagenized CB4856 parent ($p = 0.03$). When mutagenized with

formaldehyde for 5 generations, compared to parental populations, the drop in male frequency was not

significant ($p = 0.11$). For N2 non-mutagenized parental populations, male frequency was not significantly

different from N2 EMS ($p = 1.00$) or Formaldehyde treatment ($p = 1.00$). Similarly, AB1 Formaldehyde and

AB1 EMS populations did not have male frequencies different from the AB1 non-mutagenized parental

populations ($p = 1.00$ for both) or from any treatments of N2 (all $p$-values were $> 0.05$; Supplementary

Material Table 1; Fig. 2A).

*Outcrossing frequencies varied between C. elegans strains*

Outcrossing rates were assessed across strains by monitoring sex ratios in progeny from single L4 worm

pairs over four days. Plates exceeding strain-specific male frequency thresholds were classified as

outcrossed. CB4856 populations consistently exhibited the highest outcrossing frequencies under all

conditions, with peaks following mutagenesis (Fig. 2A). N2 populations, in contrast, displayed the lowest

outcrossing frequencies. AB1 populations were intermediate to these with notably higher values in EMS

mutagenized populations compared with both parent strains and the AB1 formaldehyde mutagenized

population. These findings suggest that mutagens differ in their ability to stimulate outcrossing behavior

(Fig. 2A).  Notably this difference in the outcrossing rate likely led to a difference in effective population

sizes of these strains. The highest outcrossing rate of CB4856 likely led to the highest effective population

size both before and after mutagenesis among the three strains.

*Fitness decline varied between C. elegans strains and populations*

Relative fitness was assessed by pairing focal worms with a GFP-marked competitor and quantifying

progeny composition after 8 days. Fitness was compared across non-mutagenized, post-mutagenized, and

recovered lineages using automated image analysis (Supplementary Material Table 2). To validate the

accuracy of the image analysis pipeline, we randomly selected a subset of samples and manually counted

individuals. This confirmed that the automated method produced reliable counts, with no observed false

positives or false negatives. We found that three generations of recovery from mutagenesis were

sufficient for each strain to recover pre-mutagenesis fitness levels for both the mutagens (Fig. 2B). For

CB4856, relative fitness before mutagenesis was not significantly different from relative fitness after three

generations of recovery for formaldehyde ($p$ = 1.00) and EMS treatments ($p$ = 0.18). Similarly, for AB1,

post-recovery relative fitness was not significantly different from parental fitness for formaldehyde ($p$ =

0.94) or EMS ($p$ = 0.76). N2 populations also exhibited no difference between parental and post-recovery

fitness for formaldehyde ($p$ = 1.00) or EMS (p = 1.00).

CB4856 fitness decreased slightly in EMS mutagenized populations but the effect was not

significant ($p$ = 0.18). For formaldehyde mutagenized populations fitness levels were nearly identical

before and after recovery ($p$ = 1.00). In contrast, this was not observed for N2 and AB1. For N2, pre-

mutagenesis and post-recovery relative fitness was not different for formaldehyde ($p$ = 1.00) or EMS ($p$ =

1.00). Similarly, for AB1, pre-mutagenesis and post-recovery fitness under formaldehyde ($p$ = 0.94) and

EMS ($p$ = 0.76) did not show changes in fitness.

N2 exhibited a distinct response compared to the other strains. Regardless of the mutagen used,

N2 experienced a temporary increase in relative fitness immediately following mutagen exposure

compared to its non-mutagenized parental levels (formaldehyde: $p$ = 0.03; EMS: $p$ = 0.46). However, this

fitness gain was transient. During recovery, N2 population fitness returned to pre-mutagenesis levels

(formaldehyde: $p$ = 1.00; EMS: $p$ = 1.00). Overall, three generations of recovery from mutagenesis were

sufficient to restore fitness to pre-mutagenesis levels across all strains. However, CB4856 exhibited a

persistent, albeit non-significant, reduction in fitness following EMS exposure, suggesting it may be less

resilient to mutagenic stress compared to N2 and AB1.

Figure 2. Male frequency, outcrossing behaviour, and relative fitness reveal strain-specific responses to
mutagenic stress in C. elegans. (A) Mean male frequency across strains, represented by solid circles,
before mutagenesis (non-mutagenized) and following EMS or formaldehyde exposure. CB4856 exhibited
the highest male frequency across all treatments, followed by AB1 and N2. Outcrossing frequency, shown
as solid triangles, was consistently higher in CB4856 across treatments, suggesting strain-specific
differences in mating behaviour. (B) Relative fitness across strains and generations, comparing non-
mutagenized (lightest shade), mutagenized (medium shade), and recovered (darkest shade) populations.
AB1 (green) demonstrated the highest post-mutagenesis resilience, while CB4856 (red) showed the
lowest recovery in fitness.

*Genomic analyses showed numerous SVs and SNPs were retained after recovery*

To comprehensively identify SV and SNP mutations we sequenced whole genome DNA libraries with both

Pacific Biosystems long-read and Illumina short-read DNA sequencing. Short-read data was used for calling

SNPs, whereas structural variants (SVs) were called from both short-reads and long-read data generating

a high-confidence SV dataset capturing mutagen-induced genomic alterations.

We performed t-tests to compare de novo SNP counts between strains and found that CB4856

exhibited significantly higher susceptibility to mutagenesis compared to AB1 and comparable de novo

SNPs to N2 when the mutagen was formaldehyde. Specifically, CB4856 showed a 1.12-fold increase in de

novo SNPs compared to AB1 (t=5.55, *p*=0.0014) and comparable SNPs to N2 (t < 0.01, *p*= 0.996) under

formaldehyde treatment. In contrast, with EMS as the mutagen, CB4856 showed a significantly higher

(1.16-fold) de novo SNP count when compared with AB1 (t = 6.79, p < 0.01) and a marginally higher (1.21-

fold) de novo SNP count when compared with N2 (t = 2.45, p = 0.05) as identified through short-read

sequencing (Fig. 3A).

Due to marked heteroscedasticity observed in SV counts we performed Mann–Whitney U tests
(Wilcoxon rank-sum tests in R) to compare de novo SV counts among strains. Under EMS treatment,
CB4856 populations had significantly higher de novo SV counts compared with AB1 populations (W=16, *p*
= 0.03) and compared to N2 (W=16, *p*=0.03). Similarly, with formaldehyde treatment, CB4856 showed
significantly higher de novo SV counts compared to AB1 (W=16, *p*=0.03) and compared to N2 (W=16,
*p*=0.03) (Fig. 3A).

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Figure 3: Comparison of de novo SNPs and SVs Mutations and Their Localization in Exons Across Strains
(A) Mean counts of de novo SNPs (Circles) and SVs (Triangles) induced by EMS and Formaldehyde across
AB1 (Green), CB4856 (Red), and N2 (Blue). De novo SNPs are significantly more abundant than de novo
SVs under both mutagen treatments. CB4856 has the maximum number of de novo SNPs and de novo
SVs. (B) Mean number of de novo SNPs and de novo SVs localized in exons for the same strains and
mutagen treatments, with de novo SNPs again being more prevalent than de novo SVs and CB4856 having
more SVs in exons than the other two strains and more SNPs in exons than AB1. Compared to N2, CB4856
had more SNPs in exon than N2 when treated with EMS and comparable SNPs in exon when treated with
Formaldehyde.
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*CB4856, the strain with highest male frequency and outcrossing rates, harbored high numbers of both*
*SNPs and SVs in exons*
The higher number of de novo SNPs in CB4856 was also seen in the increased number of SNPs that were
present in exons. When formaldehyde was used as the mutagen, CB4856 had 2.20 times more SNPs in
exons than AB1 (W=0, *p*= 0.03) and 1.12 times more than N2 (W=12, *p*=0.34). With EMS as the mutagen,
CB4856 had 2.26 times more SNPs in exons compared to AB1 (W=0, *p*=0.03) and 1.38 times more SNPs in
exons compared to N2 (W=16, *p*=0.03) (Fig. 3B).
Similar trends were observed with exons affected by SVs. AB1 showed the most resilience with
the least number of exons affected by the SVs, whereas CB4856 had the maximum number of SVs affecting

exons. When formaldehyde was used as mutagen, AB1 had 16.25-fold fewer SVs in exons than CB4856

(W = 0, p = 0.02) and 6.92-fold fewer SVs than N2 (W = 16, p = 0.03). With EMS as the mutagen, CB4856

had 22.31-fold more SVs affecting the exons compared to AB1 (W = 0, p = 0.02), and N2 had 7.81-fold

more exons affected by SVs than AB1 (W = 16, p = 0.03) (Fig. 3.B).

*Active transposable elements were responsible for a small portion of SV mutations*

SV mutations can be associated with transposable element (TE) activity. To test this, we used

TransposonUltimate[29] to annotate repeats and transposable elements in our whole genome sequences.

We compared repeats and TEs present in our populations with SVs before and after mutagenesis to focus

on de novo TE-associated SV mutations and potentially active TEs[29]. The total percentage of the genome

annotated as repetitive was 25.71% in AB1, higher than that identified in N2 (23.19%) and CB4856

(21.68%). Despite this, the total number of transposition events was highest in CB4856 (Fig. 4A). CB4856

exhibited 2.85-fold more TE events than N2 and AB1 when treated with formaldehyde, and 3.82-fold more

TE events than N2 and 4.20-fold more TE events than AB1 when treated with EMS (Fig. 4B). While some

TEs were strain-specific, some like the DNA transposon Zator (Fig. 5B) were consistently mutation-

associated in all populations and across both mutagens, indicating that Zator may be commonly activated

by mutagenic stress in *C. elegans* (Fig. 5B). Interestingly, some TEs, such as Copia, Gypsy, and Sola, were

uniquely active in CB4856 and nearly absent from mutation characterization in N2 and AB1 populations.

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243 Figure 4: Transposable Element (TE) Activity and Genomic Repeat Content Across Strains and Treatments.
244 **(A)** Percentage of the genome annotated as repetitive elements across the three C. elegans strains (AB1,
245 CB4856, and N2). **(B)** Total TE transposition events observed post-mutagenesis, showing higher activity in
246 CB4856 compared to AB1 and N2 under both EMS and formaldehyde treatments.
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 *SNP Enrichment Within SVs Varies by Strain and Mutagen*

CB4856 had the highest number of SNPs trapped within SVs (77.13% for formaldehyde and 82.46% for

EMS treatment; Fig. 5A). This strain, characterized by high male frequency and outcrossing (Fig. 2A, B),

also had a high intersection of these mutations with exonic regions (Fig. 3B). In contrast, while the total

number of SNPs in AB1 and N2 strains appears similar, a closer examination reveals that AB1 harbors a

higher proportion of its SNPs within structural variants (SVs) compared to N2 when the mutagen was EMS

and almost comparable numbers when the mutagen was formaldehyde. (6.36% vs. 6.80% for

formaldehyde and 12.01% vs. 7.12% for EMS treatment). While most SVs were small (<$10^4$ bp), CB4856

harbored a notable proportion of larger SVs compared to N2 and AB1 under both EMS and formaldehyde

treatments, suggesting elevated genome instability in this strain (Fig. 6).

Figure 5: Mutational and TE landscape across strains and mutagens **(A)**: Number of SNPs, SVs and SNPs
in SVs for AB1, CB4856 and N2; **(B)**: Classification and count of active TE families contributing to
transposition events across strain-treatment combinations. DNA transposons such as Zator and Tc1-
Mariner were among the most active elements, with CB4856 showing the broadest TE activity profile.
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**SV Density Distribution by Strain and Mutagen**

Figure 6: Size distribution of structural variants (SVs) across C. elegans strains and mutagens. Density distribution of SVs sizes for all the three strains for both the mutagens. x- axis Size of SVs, Y-axis: density of SVs.

*Genome-wide structural variants and transposon activity reveal distinct patterns*

The distribution of structural variations (SVs) was almost uniform across all chromosomes, regardless of strain or mutagen used (Fig. 7). In the N2 formaldehyde-treated population, insertion mutations created 471,238 nucleotides of new genetic material, approximately 0.48% of the 100.8 Mb C. elegans genome (Supplementary Material Table 3). In comparison, 74,044 nucleotides (0.06%) were deleted, 44,038 nucleotides (0.04%) were duplicated, and 1,289,368 nucleotides (1.29%) were affected by inversions. Mutations in the EMS-mutagenized N2 populations were of similar scale.

AB1 populations showed the smallest number of nucleotides affected by SV mutations. In formaldehyde-treated AB1, 35,468 nucleotides were inserted and 46,363 deleted. The notable exception was inversion mutations, which spanned 2,539,375 nucleotides in formaldehyde-treated and 4,181,603

nucleotides in EMS-treated populations. CB4856 populations harbored substantially more DNA affected

by structural variation. Roughly 1 Mb of new DNA was inserted under both treatments. Compared to AB1,

CB4856 had 14.11-fold more insertions under formaldehyde and 15.18-fold more under EMS. Relative to

N2, it had 1.50-fold and 1.66-fold more insertions under formaldehyde and EMS, respectively. CB4856

also had large amounts of duplicated DNA: 1.2 Mb under formaldehyde and 2.1 Mb under EMS. Strikingly,

inversion mutations accounted for 49.7 Mb—nearly half the genome—in formaldehyde-treated CB4856,

including large 13 Mb and 8 Mb inversions, with additional megabase-scale and smaller rearrangements.

In EMS-treated CB4856, inversions spanned 32.8 Mb—almost a third of the genome (Fig. 7.A).

Given the potential of transposable elements to contribute to structural variation, we further

evaluated TE-associated mutations across strains and treatments. TE-associated duplications and

inversions were less frequent than deletions across all populations (Fig. 7.B). Many TE-associated

mutations occurred on chromosome IV, particularly in the CB4856 and N2 populations (Fig. 7.B).

Additionally, we found shared transposition events between the two mutagens. The highest overlap

between formaldehyde and EMS TE-associated mutations was observed in N2 where 33.3% of the

mutations were shared. In CB4856 17.9% of the TE-associated mutations were shared between EMS and

formaldehyde treated populations while in AB1 it was just 4.5%. These findings indicate that while

transposable elements contribute to structural variation, they do not account for the full spectrum of

observed SVs, and their activity patterns are distinct in both scale and genomic distribution.

Figure 7: Chromosomal distribution of SVs and TE-associated SVs across strains and mutagens. (A) Stacked
bar plots showing the total count and types of SVs including Deletions, Insertions, Duplications, Inversions,

Translocations, and Unclassified (NA) across chromosomes 1–5 and X, for each strain (AB1, CB4856, N2)
and mutagen (EMS and formaldehyde). (B) TE associated SVs were primarily Deletions.

<u>Discussion</u>

In this study we investigated how quantitative levels of outcrossing influence the genome's ability to purge
deleterious mutations and how this ability varies with mutation size and genomic location. By developing
a theoretical model to explore the evolution of structural variants (SVs) in populations with varying
outcrossing rates and leveraging the unique mating system of *C. elegans*, we were able to create testable
predictions and directly answer our questions with quantitative analyses. CB4856, characterized by
highest male frequency, outcrossing rate and elevated TE activity, showed a trend toward reduced fitness
after mutagenesis and recovery. While this decline was not statistically significant, it contrasted with the
complete recovery seen in AB1 and N2, potentially reflecting strain-specific differences in genomic
resilience. Mutation analysis agreed with our theoretical predictions and showed that CB4856 populations
retained the highest number of SVs. Importantly, we found many SNPs were trapped within SVs and
overlapped with exonic regions, making it difficult to purge these variants through recombination and
outcrossing. Overall, these findings support the predictions of our theoretical model and show that higher
outcrossing rates may in fact hinder the purging of larger genomic changes such as TEs and SVs.

*C. elegans* is one of the most tractable model systems and has emerged as a powerful model
organism for unraveling the intricate dynamics of mutation accumulation. Traditionally, researchers have
employed two primary approaches to investigate this phenomenon. The first approach involves
manipulating DNA repair pathways, creating "mutation accumulation" lines with compromised repair
mechanisms [30]. This allows for controlled observation of how increased mutation rates impact population
fitness and adaptation over generations. The second approach takes a more observational stance, tracking
the accumulation and consequences of naturally occurring mutations over extended periods[31]. Both
approaches have yielded valuable insights into the remarkable ability of *C. elegans* populations to tolerate
and potentially recover from mutational stress. However, despite these advancements, a crucial aspect
remains underexplored: how mating system variation, reproductive strategies, and underlying genetic
architecture interact with mutation size and its impact on genome-wide fitness.

We explored these factors in our theoretical model and predicted that increased levels of
outcrossing would retain SVs. The number of deleterious mutations segregating in our simulated
outcrossing populations was small, ranging between 10 and 20 mutations per individual. Importantly, the
phenotypic effect size of the segregating mutations was also small. In a fully outcrossing simulated
population, there was an average of 15 mutations per individual after 100,000 generations, each with an

average phenotypic effect size of 0.0035. Larger mutation effect sizes were rapidly purged from

outcrossing populations. However, deleterious mutations with small effects on the phenotype and fitness

were stably maintained in the simulated outcrossing populations for long periods.

We found that similar to our theoretical predictions, the *C. elegans* strain with the highest levels

of outcrossing, CB4856, retained the highest number of SVs. Our *C. elegans* strains displayed small

depressions in relative fitness, indicating that these genomic mutations also had small deleterious effects.

The mechanisms by which genomes cope with mutational threats are multifaceted and depend heavily

on the type of lesion incurred. Point mutations like SNPs predominantly affect gene expression and

protein structure and are typically repaired through efficient pathways like base excision repair (BER) or

nucleotide excision repair (NER). In contrast, complex lesions like SVs (structural variants) pose a more

significant challenge to genome stability, often leading to chromosomal aberrations[32]. Repairing these

intricate injuries often relies on more intricate mechanisms like homologous recombination (HR) and non-

homologous end joining (NHEJ)[33]. In our study we used two types of mutagens that are known to affect

the genome in different ways. EMS is known to cause more point mutations and formaldehyde is known

to cause complex lesions. We expected that EMS should be creating more SNPs and formaldehyde should

be causing more structural variations in the genome. However, we found that long term mutagenic abuse

(5 consecutive generations in our case) with either of the mutagen had similar mutational effects on the

*C. elegans* genome and resulted in SNPs and SVs ranging from 10s of nucleotides to multiple megabases

in length.

The potential impact of male frequency and outcrossing on mutational burden remains highly

debated. Studies in dioecious species suggest that intense male competition for mates drives stronger

selection against deleterious mutations, potentially benefiting populations facing genomic stress[34].

However, bottlenecks arising from such strong selection can ironically counteract this advantage by

allowing the accumulation of detrimental mutations. Additionally, the relevance of these findings to

androdiecious species like *C. elegans* is unclear. Due to their self-fertilizing nature, *C. elegans* males might

not experience the same level of selective pressure as those in dioecious species[35]. Consequently,

increased male frequency in *C. elegans* might not directly translate to mutational resilience. Nevertheless,

higher male proportions could offer another advantage: increased outcrossing, which can introduce new

genetic variation. This influx of novel alleles could enhance the population's adaptability and resilience in

the face of unpredictable challenges like ongoing mutagenic stress. Moreover, mutation purging

processes differ significantly between self-fertilizing and outcrossing mating systems, particularly for SNPs

and SVs. In self-fertilizing populations, increased homozygosity facilitates rapid purging of highly

deleterious mutations from the genome. However, reduced genetic variation weakens selection against
slightly deleterious mutations, potentially leading to the accumulation of slightly deleterious SNPs and
SVs. In contrast, outcrossing populations can maintain both highly deleterious and slightly deleterious
mutations if they are recessive.

Our results suggest that there are strain- and mutagen-specific effects, with CB4856 showing the
highest male frequency among the three strains. However, it exhibited the worst relative fitness after
three generations of recovery from mutagenic treatment and carried the highest mutational load for both
mutagens. AB1, in contrast, had second largest male frequency but was least affected by mutagenic abuse
and had the minimum mutagenic load in the genome when compared with the other two strains. N2 had
the lowest male frequency but relative fitness during mutagenesis, which dropped in recovery. These
results suggest there might be some costs associated with the mechanisms that resulted in N2s high
relative fitness in an unstable environment.

Strain-specific differences in coping with mutagenic stress may reflect distinct genomic
architectures. For example, in AB1 few SVs had high proportions of SNPs trapped within them and
overlapped infrequently with exonic regions. This genomic configuration, where SNPs are more frequently
associated with SVs but largely avoid exonic regions, could enhance a population's genomic resilience by
minimizing disruptions to exonic regions. This "buffering effect" could facilitate resilience in response to
mutagenic stress. SV-associated regions could serve as hotspots for repair or regulation, further
contributing to fitness and survival under stress.

CB4856 had the highest number of de novo SNPs and SVs. A high proportion of SNPs were trapped
within SVs and a significant number of these mutations intersected with exonic regions, potentially
disrupting essential genes. This genomic configuration likely explains CB4856's reduced resilience under
mutagenic stress, as the interplay between structural and single nucleotide variants can amplify the
deleterious effects of mutations, complicating the purging process. The broad and evenly distributed
mutational landscape especially for SV ranging from 1,000 base pairs to $10^7$ base pairs (Fig. 6) may favor
greater genomic plasticity, improving a species generation of genetic variation and potential adaptation
in uncertain environments. However, CB4856 experienced the most pronounced decline in fitness
following mutagenesis. The delicate balance between evolutionary plasticity and mutational resilience
may tip toward instability under the stress of chemical mutagenesis.

N2 populations showed relatively stable relative fitness responses without significant increases in
male frequencies or outcrossing frequencies and relatively low genomic responses to mutagenesis. This
suggests that N2 relied heavily on DNA repair mechanisms, rather than increased genetic variation

through outcrossing, to mitigate the negative effects of induced mutations. The combination of self-
fertilization and the retention of small SVs may minimize genomic disruptions, ensuring genomic integrity
over generations. This may balance minimizing risk through self-fertilization and small SV accumulation,
while still allowing for some level of genomic flexibility. This type of response may allow an organism to
persist in environments where maintaining genomic integrity is essential for survival. Interestingly, the
relative fitness of N2 with both the mutagens increased which came back to the pre-mutagenizes levels
after recovery phase. This suggest that the initial increase in fitness might come with a long-term cost,
potentially limiting its adaptive potential over time.

Transposable elements (TEs) can play a critical role in shaping genomic resilience by introducing
genetic variation and driving structural changes in response to stress[36]. Increased outcrossing has been
associated with elevated TE activity, likely due to the enhanced recombination rates and reduced
efficiency of silencing mechanisms in outcrossing populations[37,38]. Our results highlight significant strain-
specific differences in TE activity post-mutagenesis. CB4856, with the highest outcrossing frequency,
exhibited the highest transposition rates, reflecting its propensity for increased genomic instability under
mutagenic stress. Interestingly, although AB1 had the highest number of annotated transposons, its
transposition rates were lower, suggesting fewer active TEs. Elevated TE activity in CB4856, coupled with
unique transposons such as Copia, Gypsy, and Sola, may indicate a potential for adaptability through
increased genetic variation and a capacity to respond to environmental challenges. Interestingly, N2
exhibited the highest overlap in TE events between mutagens (33.3%), suggesting common TE activations
under stress. Across all strains, chromosome IV emerged as a TE hotspot, likely reflecting shared genomic
characteristics that tolerate or facilitate transposition in this region.

Increased outcrossing enhances recombination rates, creating genomic instability that can
activate TE mobilization[39]. Additionally, the introduction of genetic variation through outcrossing may
temporarily disrupt TE silencing mechanisms, further contributing to higher TE activity in CB4856[40]. These
results indicate that while higher outcrossing may promote the generation of genetic diversity, it can also
trigger processes that increase susceptibility to mutagenic stress. This susceptibility may stem from strain-
specific genomic architectures, the accumulation of deleterious mutations, or the difficulty of purging
large-scale genomic changes under conditions of frequent outcrossing.

In conclusion, we found strain specific fitness and mutation responses which may reflect larger
patterns of evolutionary response. Although the N2 "Bristol" strain originally collected by Sydney Brenner
(Brenner 1974) has been developed as a model system its fitness, behavior and genomics differ
substantially from other strains of *C. elegans*. N2 is a highly lab adapted strain with little genetic variation,

few males and low outcrossing frequencies[41]. In contrast, CB4856 and AB1 are more recently collected

"wild worms" with substantial genomic differences[42] and high natural variation in male frequency[43] and

outcrossing [44]. The spectrum of natural variation in *Caenorhabditis* will continue to be an important

resource in scientific understanding of mutation.

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**Data Availability**

Scripts, bioinformatic workflows and analyses are available at github.com/RohitKapila. Python code used

for image analysis is available at github.com/FierstLab/WormImage Fitness data are available as

Supplemental Materials associated with the manuscript. Sequence reads are available at the NCBI SRA

under BioProject PRJNA1201181.

**Author contributions**

J.L.F. secured funding. R.K. and J.L.F. designed the experiments. R.K., with assistance from S.S., performed

the fitness assays, and, with assistance from G.B., conducted the male frequency and outcrossing rate

assays. V.K.E. carried out the transposable element analysis. R.K. performed the genomic analysis with

guidance from J.L.F. R.K.V. developed the Python code for worm counting. J.L.F. developed the theoretical

model and conducted the population simulations. R.K. and J.F.K. co-wrote the first draft. R.K. and J.F.K.

prepared the final manuscript with contributions from V.K.E.

**Methods**

*Theoretical model*

In order to model the evolution of structural variant mutations in a population we adapted a model of

transposable element mutations developed by Dolgin and Charlesworth (2006, 2008). This framework

built on previous work[45,46] studying the population dynamics of transposable elements. The fitness of an

individual was determined by the number of mutations they carried and calculated according to a

synergistic fitness function where fitness declined faster-than-linearly as mutations accumulated in the

genome. This synergistic fitness function was a fundamental assumption in theoretical work addressing

the evolution of sex and mutation-selection balance [46,47]. Importantly, free recombination is assumed

between each mutation.

We adapted this model in several ways. We envisioned a genomic system where structural variant

mutations could take the form of insertions, deletions or inversions and, as described in the Results above,

each mutation carried a discrete phenotypic effect size drawn from a gamma distribution. The phenotypic

effects of individual mutations cumulatively determined fitness. Following Dolgin and Charlesworth

(2006) the change in the population mean mutation number was $\Delta \bar{n} \approx V\_n \ (\partial \ln w\_\bar{n})/(\partial \bar{n}) + \bar{n} \ (\mu -$

$v)$. Here, $V\_n$ was the variance in copy number in the population and $w\_\bar{n}$ was the fitness of an individual

with $\bar{n}$ SV mutations in their genome.

Mutational effect sizes were $\Gamma(k, \theta)$ random variables giving $\Delta n \approx n \ (\theta \ (\partial \ln w\_n)/(\partial n) + \mu - v)$.

At equilibrium (defined as $\Delta n = 0$) the mean number of mutations carried in an individual's genome was

$n = (\mu - v - \theta \, a)/(\theta \, b)$ where a and b were the parameters determining the curvature of the fitness

landscape (Charlesworth 1990). According to this, we expected that the number of SV mutations

segregating in a population at equilibrium would be inversely proportional to the variance of mutational

effect sizes. With $\theta = 2, \mu = 0.01, v = 0.001, a = 0.001$ and $b = 0.0006$ we expected a mean of 5.83 SV

mutations in a fully outcrossing population.

*Population simulations*

To quantitatively examine the influence of outcrossing on SV purging and retention we implemented our

theoretical model in population simulations using the software SLiM 4.3[48]. We constructed dioecious

populations containing equal numbers of males and females and self-fertile populations with a proportion

of individuals reproducing through outcrossing and a proportion reproducing through selfing. Each

population started at a specified size K and evolved through overlapping generations with reproduction

and fitness differences creating variable population sizes. Each individual had a diploid chromosome

10,000 nucleotides in length. Individuals were randomly chosen for reproduction according to a uniform

function. Recombination occurred along the chromosomes with uniform probability at a specified rate r.

After reproduction SV mutations occurred at a rate $\mu$ and were removed at a rate $v$. The probability of a

SV mutation occurring at any location was uniform across the chromosome and each SV mutation

conferred a small deleterious phenotypic effect drawn from $\Gamma(k, \theta)$.

Our simulated populations evolved for 100,000 generations with the parameters specified in our

deterministic equations, $k = 1, \theta = 2, \mu = 0.01, \nu = 0.001, a = 0.001$ and $b = 0.0006$ We evolved

populations under a range of evolutionary scenarios including recombination rates $r$=0.5-1x10$^{-5}$, $N$=100

to 5,000 and levels of self-fertility ranging from 0 to 100%.

*Strains and mutagens*

Three C. elegans strains with varying male frequencies were chosen: N2 (Bristol), CB4856 (Hawaiian), and

AB1. ST2 served as a common competitor in relative fitness assays. All strains were obtained from the

Caenorhabditis Genetics Center and maintained frozen until the experiment. Frozen strains were thawed

on 75 mm × 13 mm NGM plates seeded with E. coli OP50. After one generation (~3.5 days at 20 °C) for

recovery, each strain was divided into nine replicate populations. Prior to the experiment, replicates were

synchronized by bleaching. Over five generations, four replicates per strain were mutagenized with

formaldehyde, four with EMS, and one remained untreated as a non-mutagenized control. All replicates

for a given strain were handled in parallel on the same day using the same batch of plates to minimize

intra-strain variability due to plate effects. Following the mutagenesis phase, all populations were frozen

and stored at −80 °C for downstream analyses.

*Mutagenesis*

Four replicates per strain were treated with 1mM Ethyl methane sulfonate (EMS) in M9 solution, while

the other four replicates were treated with 0.1% (w/v) Formaldehyde solution in M9. We followed the

worm mutagenesis protocol described by Brenner et al. (1974) with minor modifications. Every generation

(~4 days), worms were washed from the plates with M9, collected in tubes, and resuspended in their

respective mutagen solutions for 4 hours at 20 °C. After treatment, worms were transferred to fresh plates

with NGM and *E. coli* OP50, ensuring maintenance of replicate identity and avoidance of mixing.

It is important to note that the population size of the worms was not altered at any point during

the experiment. During both transfers—from the plates to the tubes and from the tubes back to the

plates—all worms were transferred using 300 microliters of liquid. Specifically, M9 buffer was used for

the transfer from plates to tubes, and the mutagen was used for the transfer from tubes back to plates.

After the mutagen treatment, the liquid was carefully transferred onto the plate, causing parts of the *E.*

*coli* colony to be dislodged. As a result, the worms would initially swim on the plate. We then left the plate

open, allowing the liquid to be absorbed within 30-40 minutes, after which the worms resumed crawling

on the plate.

*Male frequency assay*

To evaluate the impact of five generations of mutational exposure on male frequency, a male frequency assay was conducted using both non-mutagenized (parent) worms and worms immediately after five generations of mutagen treatment. Worms from each replicate population were bleached for age synchronization. The lifetime progeny production of worms was monitored by placing a single L4 worm on a 75mm x 13mm Petri plate, transferring it to a fresh plate every 24 hours. After 96 hours, the worm was discarded, and the progeny produced over the four days were sexed and counted. Any worms that died during the transfer process were excluded from the analysis. To ensure consistent sampling across treatments, 30 replicate plates were prepared for each non-mutagenized parental group and 15 replicate plates for each mutagenized population.

*Outcrossing frequency assay*

As with the male frequency assay, outcrossing frequency was assessed using the same set of replicate plates — 30 for each non-mutagenized parental group and 15 for each mutagenized population. The lifetime progeny production of each worm was monitored by placing a single L4 worm on a 75mm x 13mm Petri plate, transferring both male and hermaphrodite to a fresh plate every 24 hours. After 96 hours, both the male and hermaphrodite were discarded, and the progeny produced over the four days were allowed to grow to adulthood.

Outcrossing in *C. elegans* is known to produce approximately 50% male progeny[49]. Based on male frequency assays, we determined that spontaneous male production in the N2 and AB1 strains, both before and after mutagenesis, was relatively low. Thus, plates for these strains were visually scanned for males, and plates with male frequencies exceeding the expected spontaneous range (0.01–0.05) and approaching 50% were categorized as outcrossing plates. For the CB4856 strain, which naturally exhibits higher baseline male frequencies (15–17%), male frequency was calculated for each plate. A threshold of 25% males was set, and plates with male frequencies meeting or exceeding this value were marked as outcrossing plates.

Outcrossing produces male and hermaphrodite offspring in a 1:1 ratio. Given the significantly low male frequency in both the non-mutagenized and mutagenized N2 and AB1 strains, plates were scanned for increased male frequency. If an unusually high number of males were observed, the mating pair was considered to have outcrossed. In contrast, for the CB4856 strain, where the male frequency in non-

mutagenized parents was already around 20%, each progeny was sexed, and if the total male frequency

exceeded 40%, the mating pair was counted as an outcrossed pair.

*Relative fitness assay*

After five generations of mutagenesis, worms were frozen at -80°C and later recovered for relative fitness

assay at sixth generation- representing the fitness immediately after five generations of mutagenesis and

again after three generations of recovery from mutations using the standard *C. elegans* maintenance

protocol at 20 °C. During recovery, worms were maintained using the standard worm maintenance

protocol. For the relative fitness assay, one early-stage L1 worm from each focal strain (N2, CB4856, AB1)

was paired with one L1 worm from the non-mutagenized common competitor strain ST2 expressing green

fluorescent protein (GFP). Both strains were co-cultured on NGM plates with *E. coli* OP50 for eight days

(~2 generations) at 20 °C. Washed worms from replicate plates (11 per generation/mutant/strain

combination) were collected in M9 on a slide, imaged with and without a GFP filter on a fluorescent

microscope, and counted using a Python-based code specific for worm morphology (available on the

Github). Plates with zero worms of either type were excluded from analysis. Images captured without the

GFP filter counted all worms present on the plate- both the focal strain and the common competitor strain.

Conversely, images captured with the GFP filter only revealed the green, fluorescent competitor worms,

allowing for estimation of their abundance relative to the total worm population. With every relative

fitness assay of mutagenized and non-mutagenized worms an assay of relative fitness of non-mutagenized

control worm was also done.

*DNA extraction and Sequencing*

To assess the genomic impact of mutagenesis, high molecular weight DNA was extracted from worms

(after 3 generations of recovery) using the Cytiva kit (detailed procedure available in the Supplementary

Materials). Before starting out the DNA extraction we washed a minimum of ~10,0000 worms with M9

buffer from the plate into a tube and kept the worms rocking overnight to remove associated microbiome.

Pooled PacBio long-read sequencing was performed for parental worms of all three strains (N2, CB4856,

AB1) and one replicate per strain per mutagen. Additionally, Illumina short-read sequencing was

conducted for each replicate of each treatment (control, EMS, formaldehyde) for all strains. This strategy

enabled comprehensive analysis of genomic variants induced by mutagenesis and their phenotypic

consequences in subsequent generations.

For the comprehensive genomic analysis, we sequenced a total of 36 samples. Using PacBio long-

read sequencing, with an average depth of coverage of 94.52x, we sequenced nine samples: the three

parental strains (N2, CB4856, AB1) as controls, and six mutagenized replicates i.e., one replicate per strain

treated with each mutagen. Additionally, we performed Illumina short-read sequencing on 27 treatment

replicates, achieving an average depth of coverage of 72.11x. These 27 samples included the three

parental strains, 12 EMS-treated replicates, and 12 formaldehyde-treated replicates, with four replicates

per treatment for each strain.

The male frequency and outcrossing rate assays were conducted on worms derived from the same

batch that underwent five generations of mutagenesis. In contrast, the worms used for the relative fitness

assay and DNA extraction came from a different set, although they were mutagenized in the same manner

as the first batch.

**Data analysis**

<u>Male frequency assay</u>

Male frequency was calculated as the ratio of male progeny to the total progeny laid by each worm, and

this proportion was used as the unit of analysis. We assessed the normality of the data using the Shapiro-

Wilk test and by visualizing its distribution with a histogram. As expected, the distribution was right-

skewed, with a high proportion of zero values, indicating zero-inflation. Given the bounded nature (0 to

1) of the data and the excess zero values, we used a zero-inflated Beta regression for analysis. Since this

model requires proportion values strictly between 0 and 1, we applied an epsilon adjustment by adding a

small constant to all zero values to make the data more appropriate for the model. We performed the

analysis using the glmmTMB package in R (R version 4.1.0, released 2021-05-18)[50], implementing the

following model:

Male Frequency~Strain× Mutagen

In this model Strain (N2, AB1, CB4856) and Mutagen treatment (EMS, Formaldehyde and non-

mutagenized parents were included as fixed factors and an interaction term was added to check if the

effect of strain on male frequency varied by mutagen identity. This model accounts for both (1) the

continuous variation in male frequency and (2) the excess zero values separately, making it a better fit

than standard regression models. Further we did Tukey's HSD comparison [51] using emmeans package in R

[52].

*Relative fitness assay*

To quantify the relative fitness of the *C. elegans* strains N2, CB4856, and AB1, each plate was washed with

M9 buffer solution to prepare the worms for imaging. Two aliquots from the washed plate were then

placed onto a microscope slide and covered with a coverslip. We performed imaging to count the GFP-

expressing competitor strain and the non-GFP-expressing focal strain worms within these aliquots**.** The

allelic frequency of the non-fluorescent allele was calculated as the square root of proportion of non-GFP-

expressing worms to the total worm population. For statistical analysis, two-way ANOVAs were conducted

separately for each mutagen to assess the effects of strain and treatment duration on relative fitness.

Data normality was examined using the Shapiro-Wilk test and the residual distribution showed slight

variation from normality (Shapiro-Wilk test: W = 0.9868, *p* < 0.01), indicating variation from normality.

However, visual inspection of the residuals with histogram suggested an approximately normal

distribution. Since ANOVA is robust to minor deviations, we proceeded with ANOVA followed by pairwise

comparisons using Tukey's HSD test. The following linear model was applied:

anova(lm(RF ~ Strain*Generation, data = a)).

Following ANOVA, we performed pairwise comparisons using Tukey's Honest Significant

Difference (HSD) test to assess differences in relative fitness across strains and generations. We conducted

separate Tukey HSD tests within each strain to compare relative fitness before mutagenesis, after five

generations of mutagenesis, and after three generations of recovery.

*Bioinformatic analysis*

The bioinformatic analysis for identifying de novo mutations following PacBio sequencing involved a

comprehensive workflow to process and analyse long-read data, distinguishing mutations that arose in

the worms after mutagenesis from those existing in parental strains. Initially, raw PacBio reads were

aligned and indexed to reference genomes using minimap2 [53] and bcftools [54]; for the N2 and CB4856

strains, NCBI reference genomes were used, while the AB1 strain employed the recently annotated

genome of its descendant strain, SX3368, to account for genomic divergences within the lineage.

Subsequent variant calling was performed using Sniffles [55] for structural variants (SVs), such as insertions,

deletions, and inversions, capitalizing on the accuracy offered by long-read data.

For the specific identification of de novo mutations induced by mutagenesis, a comparative

analysis using bcftools contrasted the variant profiles of generation 8 worms (post three generations of

recovery) with their control parents, filtering out pre-existing genetic variations. Further quality control

and data filtering ensured only high-quality SNPs and SVs (quality score ≥ 20, depth ≥ 10) were retained.

We then used bedtools to provide insights into their potential impacts on exonic regions and regulatory

elements.

Additionally, similar bioinformatic steps were conducted with Illumina sequencing data.

Alignment and indexing were again performed on reference genomes, followed by SNP and SV calling

using Freebayes [56]and Manta[57], respectively. We initially called variants in pooled-population mode

(FreeBayes 'pooled') to account for minor alleles in the combined worm samples. However, when

intersecting these SNP calls with structural variants, we observed inflated counts due to multi-allelic sites

being represented multiple times. To avoid this over-counting artifact, we generated a separate VCF using

non-pooled calling parameters and used these calls for all subsequent intersection and annotation

analyses. Enhanced SV detection was achieved with VGtoolkit [58] through graph-based variant calling,

which was then integrated with SVs identified by Sniffles 2.0 using Survivor 1.0.3[59], creating a

comprehensive view of the genomic alterations induced by the mutagens.

Long-read sequencing was performed on the three non-mutagenized parent strains and one

replicate per strain for each mutagen treatment. Short-read sequencing covered all replicates across all

strains. For structural variation analysis, we merged variants identified by both long-read and short-read

callers using Survivor 1.0.3. For SNP analysis, we relied solely on variants identified by short-read callers.

<u>Annotation of Transposable Elements and Intersection with Structural Variants</u>

Transposon annotation is facilitated in complete and contiguous sequences, thus annotation of TEs was

done in the N2, AB, and CB reference genomes rather than directly from the raw sequence data. We used

the TransposonUltimate reasonaTE[60] pipeline for TE annotation in each of the reference genomes. Briefly,

TEUltimate reasonaTE wraps 13 previously published TE annotation tools, which include: RepeatMasker

v4.1.1 **,** RepeatModeler v2.0.1, SINE-Scan , SINE-Finder,  LTRPred , LTRHarvest, TIRvish , MUSTv2 , MITE-

Tracker, MITEFinderII**,** HelitronScanner, TransposonPSI  for transposon discovery. While some functions

are optional, all subprograms were run with default parameters in this study. The resulting annotations

are filtered, merged, and clustered using CD-HIT v4.8.1 [61] and BLASTN v2.10.1 [62]. The final annotations

from FinalAnnotations_Transposons.gff3 in the final Results folder were used for downstream analysis for

each of the three focal strains.

Evidence of transposon movement was assumed from structural variants which overlapped with

transposon annotations. The TEUltimatedeTEct pipeline [60] was used for intersection of transposon and

structural variant annotations. De novo VCF structural variant files, as generated above, were intersected

with transposon annotations and reported as an event if overlapping by at least 10% and lengths were

similar by at least 50%. Finally, any annotation smaller than 50 base pairs, or larger than 1% of the genome

was discarded. Plots were generated from the transpositionEvents.gff3 file in RStudio with R v4.3.2 (R

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**A** Mean and Standard Error of Mutations by Strain and Mutagen

**B** Mean and Standard Error of Mutations in Exon by Strain and Mutagen

SV Density Distribution by Strain and Mutagen