1	The effects of mutations on gene expression and alternative splicing
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### **Abstract**

Understanding the relationship between mutations and their genomic and phenotypic consequences has been a longstanding goal of evolutionary biology. However, few studies have investigated the impact of mutations on gene expression and alternative splicing on the genome-wide scale. In this study, we aim to bridge this knowledge gap by utilizing whole-genome sequencing data and RNA sequencing data from 16 obligately parthenogenetic *Daphnia* mutant lines to investigate the effects of EMS-induced mutations on gene expression and alternative splicing. Using rigorous analyses of mutations, expression changes, and alternative splicing, we show that trans-effects are the major contributor to the variance in gene expression and alternative splicing between the wildtype and mutant lines, whereas cis mutations only affected a limited number of genes and do not always alter gene expression. Moreover, we show that there is a significant association between differentially expressed genes and exonic mutations, indicating that exonic mutations are an important driver of altered gene expression.

#### Introduction

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One of the fundamental goals of evolutionary biology is to understand the genomic and phenotypic consequences of mutations. As mutations are rare, long-term mutation accumulation (MA) experiments in a variety of model organisms, e.g. *Drosophila* (Keightley et al. 2009, Krasovec 2021), Caenorhabditis (Denver et al. 2009), Arabidopsis (Keightley and Lynch 2003, Ossowski et al. 2010), Saccharomyces cerevisiae (Lynch et al. 2008, Zhu et al. 2014), and Daphnia (Bull et al. 2019, Flynn et al. 2017, Xu et al. 2012) have been instrumental in our understanding of mutational consequences. These MA experiments minimized the power of natural selection relative to genetic drift, thus allowing for the persistence of non-lethal mutations in a nearly unbiased manner and offering insight into the genome-wide rates, spectrum, and fitness effects of spontaneous mutations. The findings of these studies have yielded important implications for our understanding of genotype-phenotype relationships, the evolution of mutation rate, and the impacts of mutations and selection on molecular evolution. To date, many MA studies have moved beyond simply measuring the mutation rate, spectrum, and fitness effects on the organismal level. For example, MA studies have been utilized to study transposable element mutation rates (Díaz-González and Domínguez 2020, Ho et al. 2021), enzymatic activity (Aquadro et al. 1990, Harada 1995), methylation frequency (Denkena et al. 2021, Jiang et al. 2014), phenotypic plasticity (Latta et al. 2015), and patterns of genotype-environment interactions (Chu and Zhang 2021, Scheffer et al. 2022, Xu 2004). MA studies have also been supplemented with transcriptome data to study the evolution of copy number changes (Konrad et al. 2018), the rate at which mutations produce new variation in gene expression (Rifkin et al. 2005), effects of mutations on gene expression (Landry et al. 2007) and allele-specific expression (White et al. 2022). Moreover, how targeted mutagenesis (Iofrida et al. 2012, West et al. 2016) and somatic mutations in cancer genomes affect transcription has been investigated (Gerstung et al. 2015, Jia and Zhao 2017). Despite these previous studies, the impact of chemically induced germline mutations on gene expression on the genome-wide scale remains understudied (but see Gruber et al. 2012 and Vande Zande et al. 2022). Addressing this understudied issue would yield insight into how different types of environmentally-induced mutations (e.g., cis- and trans-mutations) can contribute to the evolution of gene expression in both short- and long-term Furthermore, the impact of chemically induced mutations on alternative splicing at the whole-genome level remains elusive. Gaining a better understanding of the impact of mutations affecting the splicing machinery and the role of cis- and trans-mutations in regulating alternative splicing will have important implications for understanding phenotypic changes such as disease (reviewed in Ward and Cooper 2010).

Gene expression is a major link between an organism's genotype and phenotype.

Numerous studies have demonstrated the critical role of altered gene expression in phenotypic variation such as the beak morphology in Darwinian finches (Abzhanov *et al.* 2006), cold adaptation in lizards (Josephs 2021), melanin production in *Drosophila* (Rebeiz *et al.* 2009), body mass due to latitudinal adaptation and color pattern in mice (Mack *et al.* 2018, Manceau *et al.* 2011), and adaptation to predation in Trinidadian guppies (Ghalambor *et al.* 2015).

Variation in gene expression can arise through cis and trans-mutations. Cis-mutations are often located within a gene, or up- and downstream of a gene, affecting the interaction between the gene and its regulators (e.g., promotors and enhancers). On the other hand, trans-mutations can be located anywhere within the genome (other than within or nearby the gene of interest) and mediate the expression of genes through the use of e.g., transcription factors and long noncoding RNAs (Albert *et al.* 2018, Vande Zande *et al.* 2022). Trans-mutations are thought to be

primarily responsible for the variation in gene expression within species since trans-factors represent a much larger mutational target and are thus more likely to arise than cis-mutations (Chen *et al.* 2015, Coolon *et al.* 2014, Metzger *et al.* 2016, Rhoné *et al.* 2017, Wittkopp *et al.* 2004). However, presumably due to the pleiotropic effects of trans-mutations on expression ( see Vande Zande *et al.* 2022), trans-mutations are often selected against as populations adapt and diverge. With cis-mutations mostly affecting the expression of specific genes, they survive selection more often, and therefore are mostly responsible for gene variation between populations and between species (Coolon *et al.* 2015, Schaefke *et al.* 2013).

These theories and empirical observations in natural populations lead us to expect that *de novo* mutations (spontaneous or environmentally induced) upon arising in the genome should exhibit stronger trans-effects than cis-effects. However, direct empirical support for this hypothesis remains limited. This is mainly because many studies inferred the impact of mutations on expression using F1 hybrids of two diverging species, because mutation accumulation studies are restricted to a small number of model organisms, and because many previous studies mainly utilized targeted mutagenesis to examine the effects of cis-mutations in a variety of different organisms (Gruber *et al.* 2012, Hornung *et al.* 2012, Kwasnieski *et al.* 2012, Maricque *et al.* 2017, Metzger *et al.* 2015).

Furthermore, a very much ignored aspect of mutations' impact resides in alternative splicing. Alternative splicing is critically involved in the origin of phenotypes, e.g., the flowering time of *Arabidopsis* (Macknight *et al.* 2002), seed development in sunflowers (Smith *et al.* 2018), pigmentation of cichlid fishes (Terai *et al.* 2003), and thermogenesis in mice (Vernia *et al.* 2016). With the involvement of trans-acting splicing factors and cis-acting regulatory motifs, alternative splicing is often highly regulated and vulnerable to mutations. One can imagine that

point mutations can occur in both introns and exons, leading to the alteration of existing splice sites or the generation of new ones, and affecting splicing enhancers and silencers (Anna and Monika, 2018). However, it remains unclear to what extent alternative splicing is affected by trans-mutations, which could directly or indirectly affect the various aspects of the transcription and translation of splicing factors.

To examine the effects of mutations on gene expression and alternative splicing on a genome-wide scale, we employed the chemical mutagen ethyl methanesulfonate (EMS) to induce heritable mutations in the microcrustacean *Daphnia*. EMS induces DNA damage through the alkylation of guanine, causing O<sup>6</sup> ethylguanine to mispair with thymine instead of cytosine in subsequent replications, resulting in EMS-induced mutations mainly consisting of G:C to A:T transitions (Greene *et al.* 2003). EMS-induced mutations are randomly distributed throughout the genome, enabling the generation of loss or gain of function mutants and weak nonlethal alleles (Greene *et al.* 2003, Lee *et al.* 2003).

Previously we established an EMS mutagenesis protocol for *Daphnia* (Snyman *et al.* 2021), where we showed that the base substitution rate for EMS exposed *Daphnia* reached  $1.17 \times 10^{-6}$  and  $1.75 \times 10^{-6}$  per base per generation for 10mM and 25mM EMS concentrations, respectively (Snyman *et al.* 2021). These EMS-induced mutation rates are greatly elevated compared to the spontaneous base substitution rate estimates in *Daphnia* of  $2.30 \times 10^{-9}$  and  $7.17 \times 10^{-9}$  per base per generation (Flynn *et al.* 2017, Keith *et al.* 2016). The EMS-induced mutations were also randomly distributed in different types of genomic regions (e.g., exon, intron, intergenic regions).

In this study, we selected an obligately parthenogenetic (OP) *Daphnia* isolate to examine the impact of EMS-induced heritable mutations on gene expression and alternative splicing. A

major motivation for using an OP isolate is that we can avoid the maternal effects imposed by EMS exposure in the ancestral mutant as much as possible by examining the asexual offspring of later generations that are genetically identical (albeit extremely rare spontaneous mutations) to the ancestors. *Daphnia* typically reproduces by cyclical parthenogenesis where they switch between asexual and sexual reproduction depending on environmental cues. However, these obligate parthenogenetic isolates (**Supplementary Figure S1**) reproduce solely via asexual reproduction (Xu *et al.* 2022).

In this study, we examined the gene expression and alternative splicing mechanisms functional in 16 *Daphnia* mutant lines derived from the same ancestor in comparison to the wildtype using whole-genome DNA and RNA sequencing data. Our main goals were three-fold. We first investigated the alterations of gene expression (DE) and alternative spicing (AS) in the mutant lines due to the EMS-induced mutations, second the contribution of cis- and transmutations to DE and AS, and lastly if whether any genomic regions carrying cis-mutations were significantly associated with DE and AS.

### **Materials and Methods**

Sampling and maintenance of isolates

An OP *Daphnia* isolate collected from a pond in the US was used in this study. This isolate has been kept as a clonally reproducing line in artificial lake (Kilham *et al.* 1998) water under a 16:8 (light: dark) cycle at 18°C and fed with the green algae, *Scenedesmus obliquus* twice a week.

Generating Daphnia mutants using EMS

To generate *Daphnia* mutants with heritable mutations from the wildtype, we used EMS mutagenesis following the established methodology by Snyman *et al.* 2021. Specifically, eight

sexually mature Daphnia females were exposed to 25 mM ethyl methanesulfonate for 4 hours to introduce mutations into their oocytes. After exposure, these females were individually isolated (**Figure 1**). We collected two progenies ( $G_1$ s) in the first asexual brood from each of the eight exposed females because each  $G_1$  was derived from an oocyte that was individually mutagenized, thus all the  $G_1$ s are genetically distinct. In total, we established 16 mutant lines.

To avoid maternal effect (i.e., from exposure to EMS) on gene expression in the offspring, each of the mutant lines was propagated asexually for at least 3 generations to reach a high enough animal density to induce the parthenogenetic production of resting eggs. 5-10 ephippia (each containing 2 resting eggs) were collected from each mutant line, and the decapsulated resting eggs were hatched using the protocol established by Luu *et al.* 2020. Briefly, the collected resting embryos were kept in the dark at 18°C for two weeks before exposure to UV light to stimulate embryo development. If no sign of embryo development was observed after five days of UV exposure, the process was repeated until at least one resting embryo hatched per mutant line. The hatched mutant offspring was then asexually propagated until a high enough animal density was reached for DNA and RNA extraction.

DNA extraction and whole-genome DNA sequencing

A total of 30-40 clonal offspring were collected for each mutant line for DNA extraction using a CTAB (Cetyl Trimethyl Ammonium Bromide) method (Doyle and Doyle 1987). DNA quality and concentration were assessed by electrophoresis on a 2% agarose gel and a Qubit 4.0 Fluorometer (Thermo Scientific, Waltham, MA, USA). DNA sequencing libraries were prepared following standard MGI sequencing library protocol by the Beijing Genomics Institute (BGI, Cambridge, MA, USA). All 16 mutant lines were sequenced on an MGI DNBseq platform with

150-bp paired-end reads, with a targeted sequencing coverage of 30X per line. The raw DNA sequence data were deposited at NCBI SRA under PRJNA892919.

RNA extraction and transcriptomic sequencing

Experimental animals were maintained in the same environmental conditions i.e., 18°C with a 16:8 light/dark cycle, and three replicates of 2 or 3-day old offspring were collected from the wild-type (i.e., control), and each mutant line. RNA was extracted using the Promega SV Total RNA Isolation kit (Madison, WI, USA) following the manufacturer's instructions. RNA quality was examined by electrophoresis on a 2% agarose gel, and RNA concentration was measured using a Qubit 4.0 Fluorometer (Thermo Scientific, Waltham, MA, USA). RNA sequencing libraries were prepared using the NEBNext Ultra II RNA Library Prep Kit for Illumina (Ipswich, MA, USA). Transcriptomic sequencing was done by Novogene Corporation Inc. (Sacramento, CA, USA) following standard Illumina sequencing protocol. Each library was sequenced on an Illumina NovaSeq6000 platform with at least 20 million 150-bp paired-end reads. The raw RNA sequence data were deposited at NCBI SRA underPRJNA892982.

Quality control and mapping

Quality of the raw reads was examined using FastQC (Andrews 2010). No adapter contamination was observed for the whole-genome sequencing data, thus further analysis was completed using the raw reads. Our RNAseq dataset showed adapter contamination, therefore Trimmomatic v.0.39 (Bolger *et al.* 2014) was used to perform adapter trimming and quality filtering. Lastly, reads were reassessed using FastQC to confirm the removal of low-quality reads and adapter sequences.

*Identification of EMS-induced heritable mutations* 

We identified heterozygous EMS-induced heritable mutations following the EMS mutagenesis analysis protocol established by Snyman et al. (2021). Since these mutant lines are obligate parthenogens (i.e., no segregation and no sexual reproduction) all EMS-induced mutations should be in the heterozygous state, ignoring rare ameiotic gene conversions (Omilian et al 2006). We used the Burrows-Wheeler Alignment Tool BWA-MEM version 0.7.17 (Li and Durbin 2009) with default parameters to align the whole-genome DNA sequencing raw reads of each mutant line to the Daphnia pulicaria reference genome (Jackson et al. 2021). SAMtools (Li et al. 2009) was used to remove reads mapped to multiple locations, and the MarkDuplicates function of Picard tools (http://broadinstitute.github.io/picard/) was used to locate and tag PCR duplicates. We used BCFtools (Li 2011) mpileup and call functions with default parameters to generate genotype likelihoods and genotype calls in a VCF file. The following additional FORMAT and INFO tags were added to the VCF file: AD (allelic depth), DP (number of highquality bases), ADF (allelic depth on forward strand), and ADR (allelic depth on reverse strand). Tentative EMS-induced mutations for all treatment lines were further filtered with BCFtools' filter function to retain only single nucleotide polymorphisms (SNPs) with a sequencing depth  $(DP) \ge 10$  and  $\le 60$ , quality score  $(QUAL) \ge 20$ , and a distance of at least 50-bp from an indel.

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Furthermore, a custom python script (https://github.com/Marelize007/Functional\_impact \_of\_EMS-induced\_mutations) was used to filter the tentative mutations using a consensus method (Snyman *et al.* 2021). Briefly, all genotype data from the 16 EMS mutant lines were added to one VCF file, and a consensus genotype was established by majority rule. Out of the 16 mutant lines, at least 12 mutant lines had to agree to generate a consensus genotype call per site. This allowed for the inclusion of mutation sites where up to 3 out of the 16 mutant lines had no

genotype call due to inadequate depth. A tentative mutation was identified if the mutant line showed a different genotype than the consensus. All final mutations had to be supported by at least two forward and two reverse reads to limit false positives due to allele drop and inadequate sequence coverage.

Differential expression analysis

The trimmed RNAseq reads were mapped to the *D. pulicaria* (Jackson *et al.* 2021) reference genome utilizing the STAR aligner (Dobin *et al.* 2013) with default parameters. Reads mapped to multiple locations in the genome were removed using SAMtools (Li *et al.* 2009), and raw transcript counts were obtained for each sample using featureCounts (Liao *et al.* 2014). Differential expression analysis was performed using DESeq2 v.1.34.0 (Love *et al.* 2014). First, the regularized log (rlog) transformation function was used to normalize the mapped read counts, and DE genes were determined using the Wald negative binomial test with the design formula  $\sim$  genotype, where genotype represents either the mutant or wildtype genotype. The Benjamini-Hochberg method (FDR < 0.05) was used to adjust p-values for multiple testing, and genes with a p-value <0.05 were considered significantly differentially expressed. DE genes were additionally filtered according to fold-change. Genes with a fold-change > 1.5 were considered upregulated, and < -1.5 were considered downregulated in the mutant lines compared to the wildtype (i.e., control).

*Mutation rate calculation* 

The formula,  $\mu = m/n$  \*l was used to calculate the per site per generation mutation rate for all 16 mutant lines, where m is the total number of mutations identified in each line, n is the total number of genomic sites with a sequencing depth >=10, and <=60, QUAL >= 20, and where

each site is at least 50-bp from the nearest indel in each mutant line. Furthermore, l represents the number of generations. To calculate the per gene per generation mutation rate, we used the formula  $\mu_g = m_g/n_g *1$ , where  $m_g$  represents the total number of mutations detected in genic regions (including UTRs, exons, and introns),  $n_g$  represents the total number of genes analyzed in each mutant line, and l represents one generation. The non-synonymous mutation rate was calculated utilizing the same formula, except  $m_g$  represented the number of non-synonymous mutations per mutant line.

Annotation of EMS-induced mutations

Functional annotation based on genomic locations and effect prediction of EMS-induced heritable mutations were done using the cancer mode (-cancer) of SnpEff version 4.0 with default parameters (Cingolani *et al.* 2012). This mode was utilized since it allowed direct comparison between the EMS mutant genotypes and the wild type.

Alternative splicing analysis

We used the tool rMATS v4.1.1 (Shen *et al.* 2014) to detect alternatively spliced (AS) events using reads mapped to both exons and splice junctions. The following alternatively spliced events were detected: skipped exon (SE), alternative 5' splice site (A5SS), alternative 3' splice site (A3SS), retained intron (RI), and mutually exclusive exons (MXE). SE events take place when an exon along with its flanking introns are spliced out, and A3SS and A5SS result when different parts of exons are either included or excluded from the resulting transcript. During RI events, introns are retained; during MXE events, only one of two exons are retained in the resulting mRNA (Pohl *et al.* 2013, Wang *et al.* 2015). Alternatively spliced events had to be supported by at least four uniquely mapped reads and have a minimum anchor length of 10nt.

Additionally, the Benjamini-Hochberg adjusted (FDR < 0.05) p-value had to be less than 0.05, and the difference in exon inclusion level ( $\Delta |\psi|$ ) greater than 5% (Shen *et al.* 2014, Suresh *et al.* 2020).

## Results

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Whole-genome DNA and RNA sequencing data We obtained ~6GB of raw DNA sequence data per mutant line, containing an average of ~34 million reads (SD = 3.7 million) per sample (Supplementary Table S1). After removing PCR duplicates and reads mapped to multiple locations, each mutant line had on average ~29 million (SD = 2.5 million) mapped reads with an average coverage of 25 reads (SD = 2) per site. Furthermore, ~20 million (SD = 4.9 million) RNA reads were obtained per line, totaling 51 samples (3 replicates per mutant line and one wildtype (control)). After adapter trimming and quality control, reads were mapped to the *D. pulicaria* reference genome (Jackson et al. 2021) with an average mapping rate of 95% (SD = 2.37%, Supplementary Table S2). EMS-induced heritable mutation rate and spectrum To identify EMS-induced mutations we used a rigorous bioinformatic EMS mutation identification pipeline previously tested with PCR mutation verification (false positive rate < 0.05, Snyman et al. 2021). The number of EMS-induced heritable heterozygous mutations in each mutant line ranged from 125 to 576 (median =152), translating to a base substitution rate

identification pipeline previously tested with PCR mutation verification (false positive rate < 0.05, Snyman *et al.* 2021). The number of EMS-induced heritable heterozygous mutations in each mutant line ranged from 125 to 576 (median =152), translating to a base substitution rate ranging from 1.51 to  $6.98 \times 10^{-6}$  (median=  $1.84 \times 10^{-6}$ ) per site per generation. The number of mutations found within genic regions (including UTRs, exons, and introns) ranged from 49 to 320 (median = 74) in each line, resulting in a mutation rate of  $2.0 \times 10^{-3}$  to  $1.3 \times 10^{-2}$  (median =  $3.1 \times 10^{-3}$ ) per gene per generation (**Figures 2A and 2B, Supplementary Table S3**).

Across all 16 EMS-induced mutant lines the number of non-synonymous mutations ranged from 21 to 140 (median = 29), resulting in a median non-synonymous mutation rate of  $1.14 \times 10^{-3}$  per gene per generation (**Figure 2B**). The number of synonymous mutations ranged from 6 to 72 (median = 17, **Supplementary Table S3**) across mutant lines. Assuming a random distribution of EMS-induced mutations, we expect to see a non-synonymous vs synonymous mutation ratio of 3:1. However, in 6 of the mutant lines we observed a significant deviation (one proportion Z test p-value < 0.05) from the expected 3:1 ratio (**Figure 2 B**, Graur and Li 2000).

Lastly, in accordance with previous work on *Daphnia* (Snyman *et al.* 2021) and other model organisms such as *C. elegans* (Flibotte *et al.* 2010) and *D. melanogaster* (Pastink *et al.* 1991), the majority of EMS-induced mutations were G:C to A:T transitions (mean = 90%, SD = 4%), yielding a transition-transversion ratio greater than 5.06 for all mutant lines (**Figure 2C**, **Supplementary Table S4**).

Differentially expressed genes in mutant lines

To assess how EMS-induced mutations altered the transcript abundance across the genome, we examined the transcriptomic dataset derived from 16 mutant lines in comparison to the wildtype. Our principal component analysis based on the normalized read counts (rlog transformation) showed the tight clustering of replicates derived from the same mutant, suggesting strong overall similarity among replicates of the same mutant background (**Figure 3A**). The top two principal components (PC) were responsible for >50% of the variance, with PC1 accounting for 32% and PC2 for 21%. It was unclear what biological factors these two principal components potentially represent. Most likely, they jointly captured the variability between mutants, between sibling mutant lines, and between replicates.

Our differential expression analysis contrasting each mutant line against the wildtype revealed that the total number of differentially expressed (DE) genes ranged from 1176 to 6606 across all lines (p < 0.05 with FDR < 0.05), with a median of 3545. In nearly all mutant lines the ratio of down- vs. upregulated genes was greater than 1 (1.05-3.63, Figure 3B, Supplementary Table S5), demonstrating that downregulated genes were more abundant.

Next, we assessed the number of DE genes carrying an EMS-induced mutation and DE genes free of a mutation. We defined that a gene carries an EMS-induced mutation if a mutation was located 5kb upstream or downstream of the gene, in the 3' UTR, 5' UTR, introns, or exons. The number of mutation-carrying DE genes in the mutant lines ranged from 5 (0.43%) to 423 (6.40%), with a median of 50.5 (Supplementary Table S5). Furthermore, the ratio of down- vs. upregulated mutation-carrying DE genes ranged from 0.5 to 3, with mutant lines equally split between showing a majority of downregulated vs a majority of upregulated genes (Figure 3B, Supplementary Table S5).

Except for two lines where ~6% of DE genes carried a mutation, the remaining 14 lines all showed less than 2% of DE genes carrying a mutation (Supplementary Table S5). On the other hand, mutation-free DE genes in each mutant line ranged from 1171 (94.6%) to 6383 (99.6%) with a median of 3502 (). Therefore, with most DE genes being mutation-free, it strongly suggests that trans-effects drive the transcriptional changes for most DE genes and tend to result in more downregulation than upregulation, whereas cis-effects only influence a small proportion of genes, although trans-effects cannot be excluded from imposing their impacts. Lastly, cis-mutations may not alter gene expression because we identified non-DE genes carrying a mutation. Out of the total number of genes carrying a mutation, the number of genes

that were non-DE ranged from 215 (51.6%) to 1021 (77.3%) across the lines, with a median of 303.

## *Impact of EMS-induced mutations*

Across all 16 mutant lines, the EMS-induced mutations were randomly distributed throughout the genome (chi-squared test, all p values > 0.05), i.e., the proportion of mutations in a specific type of genomic region does not exceed expectation based on random distribution of mutations. The distribution of all mutations, mutation in DE genes, and mutations in non-DE genes across the genome can be viewed in **Table 1** and **Supplementary Tables S6, S7 and S8**. The distribution of mutations among these three categories was highly similar.

The occurrence of mutation-carrying DE and non-DE genes presents an opportunity to see whether differential expression is associated with EMS-induced mutations in a specific genomic region. Using a multiple logistic regression model, we find that there was a significant association (p-value = 0.028) between differential expression and mutations in exons. Our results showed that for every mutation occurring in the exonic region of a gene, there is an increase of 0.73 in the log odds of that gene being differentially expressed, suggesting that exonic mutations are an important driver of altered gene expression.

Moreover, we categorized the mutations in DE and non-DE genes as high impact, moderate impact, low impact, or modifier variants. The mutations identified in DE genes consisted of 27 (1%) high impact variants (i.e., loss of a start codon, gain of a stop codon, splice acceptor variant), 184 (10%) moderate impact variants (i.e., missense variants), 113 (6%) low impact variants (i.e., splice region variant, synonymous variant, and variants producing a premature start codon in the 5' UTR region), and 1553 (83%) modifier variants (i.e., variants 5kb

upstream or downstream of a gene, variants within the 3' UTR and 5' UTR region, and variants within introns (Supplementary Table S9).

Interestingly, for the high-impact variants, two genes carrying splice acceptor variants were both downregulated, while two other genes carrying start-lost variants were both upregulated. Out of 23 genes carrying stop-gained variants, only four were downregulated whereas the remaining 19 were upregulated (Supplementary Figure S2). From the genes carrying moderate impact (i.e., missense) variants, low impact variants and modifier variants; 129, 83 and 765 were upregulated, and 55, 36 and 793 were downregulated, respectively, suggesting moderate and low impact variants tend to be more associated with upregulated expression than downregulated expression (Supplementary Table S10). On the other hand, mutations identified in non-DE genes have a similar distribution of the impact categories (Supplementary Table S11 and S12). It is interesting to note that even some high-impact variants such as loss of a start codon and gain of a stop codon did not result in altered expression for the mutation carrying genes.

## Alternative splicing

Alternatively spliced (AS) genes were identified by comparing all 16 EMS-induced mutant lines to the wildtype. The number of AS genes ranged from 212 to 627 across all mutant lines, with a median of 393 (Supplementary Table S13). The number of mutation-carrying AS genes ranged from 2 (0.5%) to 57 (9.6%) with a median of 12, while that of mutation-free AS genes ranged from 202 (~90%) to 611 (~99%) across mutant lines, with a median of 373 (Supplementary Table S14 and 15). The dominance of mutation-free AS genes strongly suggests that the transeffects caused by mutations are a major driver of alternative splicing.

A summary of the AS events detected for all AS genes, mutation-carrying AS genes and mutation-free AS genes can be viewed in **Table 2**, with the full results in **Supplementary Table S13**, **S14** and **S15**. No significant difference (chi-square test, p-value = 0.25) was observed in the distribution of AS events between mutation-carrying and mutation-free AS genes. Lastly, across all 16 EMS-induced mutant lines, a total of 44 genes were both AS and DE (median=2, **Supplementary Table S14**).

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Impact and effect of genic EMS-induced mutations on alternative splicing

Mutations located in the alternatively spliced genes were further categorized according to their impact, with 6 (~3%) being high impact, 14 (~7%) low impact, 20 (~10%) moderate impact, and 159 (~80%) modifier impact variants across all 16 EMS-induced mutant lines (**Supplementary Table S16**). The high impact variants consisted of 1 (~0.5%) splice acceptor variant, 2 (~1%) splice donor variant, and 3 (~2%) stop-gained variants. Moderate impact variants consisted of 20 (~10%) missense variants, while low impact variants consisted of 1 (~0.5%) splice region variant and 13 (~7%) synonymous variants. Modifier impact variants consisted of 1 (~0.5%) 3' UTR variant, 40 (~20%) downstream variants, 22 (~11%) intron variants, and 97 (~49%) upstream variants across all mutant lines (**Supplementary Table S17**). Lastly, results from a logistics regression showed that there was no significant association between AS event and variant type.

# Discussion

How mutations alter genome-wide gene expression and alternative splicing has been understudied, with only a few attempting to address this issue using mutation-accumulation experiments in a small number of species including *C. elegans*, *D. melanogaster*, *S. cerevisiae*, and zebrafish (Denver *et al.* 2005, Huang *et al.* 2016, Konrad *et al.* 2018, Landry *et al.* 2007,

Rifkin *et al.* 2005, White *et al.* 2022, Zalts and Yanai 2017). In this study, we examined the alterations in genome-wide gene expression and alternative splicing patterns in EMS *Daphnia* mutants to bridge this knowledge gap. Furthermore, we dissected the contribution of cis- and trans-mutations to differential gene expression and alternative splicing.

Global gene expression changes due to EMS-induced mutations

In our mutant lines the EMS-induced base substitution rate, per gene mutations rate, and non-synonymous mutation rate were estimated to be (medians) 1.84×10<sup>-6</sup> per site per generation, 3.1×10<sup>-3</sup> per gene per generation and 1.14×10<sup>-3</sup> per gene per generation respectively (**Figure 2A**, **B**). These estimates are derived from a rigorous mutation filtering pipeline that has been calibrated with PCR verification of mutations (false discovery rate <0.05) in a previous study (Snyman *et al.* 2021). These rate estimates are also consistent with our previous work (Snyman *et al.* 2021), showing mutation rates significantly higher than the spontaneous mutation rates in *Daphnia* (Bull *et al.* 2019, Flynn *et al.* 2017, Keith *et al.* 2016). Additionally, the identified EMS-induced mutations showed the expected enrichment of G:C to A:T transitions (**Figure 2C**) and random distribution across the genome (**Table 1**, Greene *et al.* 2003).

Considering that EMS may cause acute gene expression changes in the exposed females and the lingering maternal effects may alter gene expression in immediate offspring, we measured gene expression in the clonal mutant lines several generations after exposure. This method ensures that the observed gene expression changes are most likely caused by the EMS-induced mutations, rather than by maternal effects.

Our differential expression analyses reveal that downregulated genes are more abundant than upregulated genes in nearly all mutant lines, with the ratio of down- vs. upregulated genes

ranging between 1.05 and 3.63 (**Figure 3B**, **Supplementary Table S5**). It is not clear whether mutations induced by other mutagens would cause a similar pattern as gene expression changes in mutagen-induced mutant lines remains understudied. However, as we discuss below the dominance of trans-effects in gene expression changes, the higher abundance of downregulation is likely caused by trans-acting mutations mainly acting as silencers to reduce transcription (Johnson *et al.* 2015).

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Examining whether DE genes carry EMS-induced mutations across our mutant lines unveiled a significant contribution of trans-mutations to expression changes. DE genes carrying EMS-induced mutations only accounted for a small portion of the total (ranging from 0.43% to 6.40% across our mutant lines), while mutation-free DE genes were more common (ranging from 94.6% to 99.6% across mutant lines). This pattern strongly suggests that the pleiotropic effects of trans-mutations are a major contributor to the variance in gene expression between mutant and wildtype lines (Supplementary Table S5), while cis-mutations only play a relatively small role. The pleiotropic effects of trans-mutations were previously reported in D. melanogaster and C. elegans mutation accumulation lines, where a relatively small number of trans-acting mutations with multiple downstream effects induced the majority of the transcriptional differences (Denver et al. 2005, Huang et al. 2016). The dominant pleiotropic effects of trans-mutations are probably due to trans-factors representing a larger mutational target and are more likely to arise than cis-mutations (Chen et al. 2015, Coolon et al. 2014, Gruber et al. 2012, Metzger et al. 2016, Rhoné et al. 2017, Wittkopp et al. 2004). As it is often assumed trans-mutations have negative pleiotropic effects on fitness, unfortunately we did not collect fitness data on these lines. It would be interesting for future studies to address this issue.

The presence of cis-mutations in DE genes indicates that cis-mutations may directly cause gene expression changes. However, it cannot be ruled out that trans-effects also contribute to the variance in gene expression of these genes. It is noteworthy that the presence of cismutations does not necessarily alter expression because 51.6% - 77.3% of the observed genes with cis-mutations were non-DE. However, our logistic regression analysis provides a clear signal that exonic mutations are strongly associated with altered gene expression. Interestingly, it was observed in EMS mutagenized *S. cerevisiae* mutants that trans regulatory mutations were also mostly located within the coding sequences (Duveau *et al.* 2021).

It is possible that some cis-mutations do not directly impact gene expression. However, the unchanged expression of genes with cis-mutations may also be a result of complex interactions between cis- and trans-effects, for which we do not have the capacity to further test in this study. Furthermore, because the mutated sites are heterozygous for the mutation and wildtype SNP allele, the invariance of total gene expression may have concealed the signature of allele-specific expression changes caused by the EMS-induced mutations. It would be of great interest for future studies to address whether and how mutations cause the changes in allele-specific transcript abundance.

Impact of different classes of mutations on gene expression

Our results also showed that moderate-impact (missense) and low-impact variants tend to be associated with an upregulation of gene expression (129 upregulated genes vs 55 downregulated genes for moderate impact, 83 upregulated vs 36 downregulated genes for low impact). Missense variants are known to alter the amino acid sequence, and have been shown to impact the interacting DNA-transcription factors, resulting in changed expression of other genes (Ding *et al.* 2015). Low-impact variants including synonymous variants have also been shown to

influence gene expression by disrupting transcription and splicing, as well as mRNA stability (Pagani *et al.* 2005, Presnyak *et al.* 2015, Stergachis *et al.* 2013, Wang *et al.* 2021).

High impact variants consisted of two downregulated splice acceptor variants, two upregulated start-lost variants, and the upregulation of most stop-gained variants (Supplementary Figure S2). The downregulation of some stop-gained (nonsense) variants could indicate nonsense-mediated RNA decay (NMD) is at play. NMD is a surveillance pathway that helps maintain RNA quality and cellular homeostasis by detecting and eliminating transcripts containing premature stop codons (Nickless *et al.* 2017). The upregulation of most stop-gained variants could indicate that these genes escape NMD, that a stop-codon read-through occurs, or that transcriptional adaptation is triggered causing an upregulation of the affected as well as related genes (El-Brolosy *et al.* 2019).

On the other hand, mutation-carrying non-DE genes showed a similar distribution to DE genes (Supplementary Table S7 and S8). Interestingly, some high-impact mutations in these genes did not result in an altered expression. It is plausible that these mutations simply do not impact gene expression or some of these mutations are false positives that slipped through our mutation identification pipeline, although the rate of false positives should be < 5%. Another possibility is that our RNA-seq experimental design did not provide enough experimental power to detect the expression changes of these genes. It should also be noted that only one allele of the affected genes was mutated, which means that a functional wildtype allele still exists for the affected genes. Thus, it is possible that while the mutated allele may have had reduced transcription, the wildtype allele likely experiences enhanced transcription, resulting in unchanged expression of the gene in question.

Alternative splicing

Changes in pre-mRNA splicing are another important source of phenotypic variation. If the splicing process is not correctly regulated, premature stop codons and transcripts with an altered amino acid sequence can arise. Alternative splicing (AS) can result from both cis- and trans-regulatory mutations. Cis-mutations can interact with exonic or intronic splicing enhancers or silencers, whereas trans-mutations affect splicing factors potentially impacting genes throughout the genome (Juan *et al.* 2014, Kornblihtt *et al.* 2013). To date, the impact of cis-versus trans-mutations on alternative splicing remains understudied. Most studies have suggested that cis-regulatory (Ast 2004, Keren *et al.* 2010, Thatcher *et al.* 2014) mutations may be the main contributor to alternative splicing. However, a recent study raised the opposite view, suggesting that the pleiotropic effects of trans-mutations may be the main culprit (Smith *et al.* 2018).

Our results are consistent with the latter view. Our analyses show that the number of mutation-carrying AS genes ranged from 0.5% to 9.6% across mutant lines (Supplementary Table S14). Most AS genes are free of EMS-induced mutations (ranging from 90% to 99% of total AS genes), strongly suggesting that trans-effects are a major driver of AS (Supplementary Table S15). Furthermore, our results showed little overlap between DE and AS genes, a pattern further seen in *Drosophila* (Jakšić and Schlötterer 2016), aphids (Grantham and Brisson 2018), and salmonids (Jacobs and Elmer 2021), presumably because of non-overlapping regulation between DE and AS genes

*Implications for understanding the biological impacts of anthropogenic events* 

As we are preparing this manuscript, on February 3<sup>rd</sup> 2023, a train derailment in East Palestine, Ohio, USA, resulted in the spill of > 100,000 gallons of vinyl chloride, a carcinogenic mutagen. Accidents like this (along with Chernobyl disaster in 1986 and Fukushima nuclear disaster in 2011) and many unreported, smaller-scale anthropogenic events occur at a sharply increased rate

since the Industrial Revolution. The epigenetic effects of some of these environmental chemicals have been investigated (reviewed in Baccarelli and Bollati 2009, Goyal *et al.* 2022, Hou *et al.* 2012). However, much work still needs to be done to determine the impact of chemically-induced mutations on gene expression and alternative splicing on the genome-wide scale. Based on our results, mutations exerting trans-effects can have a genome-wide impact by altering the expression and splicing of thousands of genes (up to 99% of the total DE and AS genes respectively in *Daphnia*), while high-impact and moderate-impact mutations could lead to amino acid changes, with potential functional implications. Understanding the consequences of these anthropogenic stressors are thus vital to protecting and preserving the health of our ecosystems and ourselves.

# **Conclusions**

Our results show that trans-effects are the major contributor to the variance in gene expression and alternative splicing between the wildtype and EMS-induced mutant lines in *Daphnia*, while cis-mutations do not always alter gene expression and only affected a small portion of genes. Furthermore, our results showed a significant association between DE genes and exonic mutations, indicating that exonic mutations are an important driver of altered gene expression.

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537	Data Accessibility
538	DNA sequence data: NCBI SRA-PRJNA892919.
539	RNA sequence data: NCBI SRA-PRJNA892982.
540	Author Contributions
541	SX designed the study, and MS and SX wrote the manuscript. MS performed tissue collection,
542	molecular work, and data analysis.
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Figure 1. Illustration of exposure method to generate EMS mutant lines.

Figure 2. (A) The base substitution rates. (B) Non-synonymous and synonymous mutation rate, and non-synonymous vs synonymous mutation ratio in all mutant lines. (C) The proportion of base substitutions caused by EMS-induced mutations.

Figure 3. (A) PCA plot based on transcriptomic data of mutant lines. (B) Log2 fold-change of DE genes (red dots) and DE genes with mutations (blue dots) with the ratio of down- vs. upregulated DE genes for each line.

Table 1. The number of mutations found in different genomic regions.

Category	5kb downstream region	5kb upstream region	Exons	Intergenic regions	Introns	Splice Sites	3'UTR	5'UTR
All mutations	3047	3142	1068	1762 (~18%)	538	58 (~0.6%)	144	147
	(~31%)	(~32%)	(~11%)		(~5%)		(~1.5%)	(~1.5%)
Mutation in DE	650 (~34%)	698 (~37%)	311	N/A	140	16 (~1%)	43 (~2%)	33 (~2%)
genes			(~17%)		(~7%)			
Mutation-in non-	2397	2444	757	N/A	398	42 (~1%)	101	114
DE genes	(~38%)	(~39%)	(~12%)		(~6%)		(~2%)	(~2%)

**Table 2.** Summary of alternative splicing events of genes.

Category	A3SS	A5SS	MXE	RI	SE
All AS genes	665 (~11%)	665 (~11%)	1051 (~17%)	2027 (~32%)	1880 (~30%)
Mutation-carrying AS genes	20 (~8%)	22 (~9%)	48 (~19%)	55 (~22%)	104 (~42%)
Mutation-free AS genes	645 (~11%)	643 (~11%)	1003 (~17%)	1972 (~33%)	1776 (~29%)