

# Intra-Specific Regulatory Variation in *Drosophila* pseudoobscura

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#### **Abstract**

It is generally accepted that gene regulation serves an important role in determining the phenotype. To shed light on the evolutionary forces operating on gene regulation, previous studies mainly focused on the expression differences between species and their inter-specific hybrids. Here, we use RNA-Seq to study the intra-specific distribution of *cis-* and *trans-*regulatory variation in *Drosophila pseudoobscura*. Consistent with previous results, we find almost twice as many genes (26%) with significant *trans-*effects than genes with significant *cis-*effects (18%). While this result supports the previous suggestion of a larger mutational target of *trans-*effects, we also show that *trans-*effects may be subjected to purifying selection. Our results underline the importance of intra-specific analyses for the understanding of the evolution of gene expression.

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# Introduction

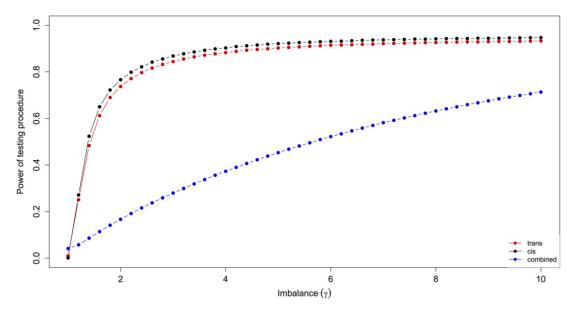
It is well understood that variation in gene expression is an important source of phenotypic differences. Hence, it has been a long-term goal in biology to understand the regulation of gene expression. The genetic basis of variation in gene expression can be divided into two classes: 1) Variation in regulatory domains (cisregulatory elements) that modulates gene expression or polymorphisms that influence the stability of mRNA. 2) Variation in transacting proteins, e.g. transcription factors that regulate the expression of a set of target genes. The combined effects of cisand trans-regulation determine the expression of every gene. Different approaches ranging from diallel crosses [1] to eQTLs [2] have been pursued to understand the regulation of gene expression, but the highest level of resolution has been obtained from allele-specific gene expression measurements in parents and their offspring. Since offspring share the trans-acting factors from both parents the contrast of gene expression in parents and offspring provides an estimate for the magnitude of cis- and transeffects [3].

The contribution of *cis*- and *trans*-effects to the evolution of gene expression has been studied in several species (e.g. in *Drosophila*: [3,4,5,6], yeast: [7,8], *Arabidopsis*: [9]). Independent of the species studied, frequently more *cis*- than *trans*-effects were found when two closely related species and their hybrids were analyzed. This observation is particularly interesting since the mutational target of *trans*-effects (i.e.: all *trans*-regulating factors) is larger than the one of *cis*-effects [10,11]. The interpretation of the greater amounts of *cis*-effects found is that *trans*-effects are more pleiotropic than *cis*-

effects. Thus, *trans*-mutations are more difficult to establish and become fixed even if they have a beneficial effect on one target gene as other target genes can be negatively affected by the mutation. In contrast, *cis*-mutations affect a single gene only. Therefore, they tend to be less deleterious than *trans*-effects and are more likely to become fixed [4].

The comparison of mutation accumulation lines, with very low selection efficacy due to a small effective population size, to natural isolates also showed that *trans*-effects are subject to strong purifying selection [11]. Furthermore, *D. sechellia*, a species with a very small effective population size (and thus a lower selection efficacy) shows more *trans*-effects than the cosmopolitan *D. simulans* [6,12,13].

For intra-specific studies, the relative importance of cis- and transeffects is still controversial. Early attempts to understand the contribution of cis- and trans-regulatory components within Drosophila species showed more trans- than cis-effects in D. simulans males [14]. A study in D. melanogaster identified 90% cis-effects and only very few trans-effects, but it was not clear to what extent this observation results from low power to detect *trans*-effects [5]. A comparison of the two D. melanogaster strains with strong assortative mating, Z and M, suggested that the expression differences between them could be largely attributed to cis-regulatory variation [15]. In yeast, transeffects were found to contribute more to intra-specific expression variation [7]. Among 863 of differentially expressed genes 558 (~64%) had been influenced by trans-regulation, whereas 421  $(\sim 49\%)$  genes were found to have significant cis- regulation [7]. In Arabidopsis thaliana, more pure trans-effects than pure cis-effects were found [16], which contrasts with an earlier study that found an excess of *cis*-effects in A. thaliana [1].



**Figure 1. Power analysis of** *cis*- **and** *trans*-**tests.** We used computer simulations to determine the power to detect *cis*-effects (black), pure *trans*-effects (red), and *trans*-effects in genes with combined regulatory modes (blue). The power was determined for 10000 randomly selected genes over a range of imbalance indices ( $\gamma$ ) and the averages are shown. The imbalance indices for genes with combined effects were calculated as  $|\gamma_P - \gamma_H| + 1$ . doi:10.1371/journal.pone.0083547.q001

Since it is not clear to what extent the contrasting observations reflect differences in the experimental design and applied technologies, we studied the relative contribution of cis- and trans-effects to the intra-specific expression variation in D. pseudoobscura. Using the latest next generation sequencing technology we show that trans-effects are more abundant than cis-effects, probably reflecting their larger mutational target size. Furthermore, we show that the distribution of effect sizes differs between cis- and trans-effects, most likely caused by different selective forces operating on them.

# **Materials and Methods**

#### Sample preparation and sequencing

The *D. pseudoobscura* strains ps94 (stock number 14011-0121.94) and ps88 (stock number 14011-0121.88) were obtained from the UC San Diego Drosophila Species Stock Center. Libraries of the parental strains were those used by Palmieri et al. [17]. Reciprocal hybrids were reared in parallel and under the same conditions as the parental strains by either crossing a virgin ps94 female with a ps88 male or a virgin ps88 female with a ps94 male. Flies were reared on standard cornmeal-molasse-yeast-agar medium and maintained at 19°C under constant dark conditions. Virgin female offspring of 15-20 replicate crosses from either direction were collected, allowed to age for three to seven days and shock-frozen in liquid nitrogen. For extraction of total RNA, we used equal numbers of females from the two directions of crosses, and performed two replicate RNA extractions that were pooled for library construction. This strategy aimed to reduce possible imprinting effects and the biological variation among crosses. Paired-end Illumina mRNA libraries were generated from 10 ug total RNA using the mRNA Sample Prep Kit (Illumina, San Diego, CA) as previously described [17]. All libraries were sequenced on the same flow cell as the female libraries of the two parental strains used in [17]. We did not sequence replicate libraries. The RNA-Seq read libraries were deposited in the ArrayExpress database https://www.ebi.ac.uk/arrayexpress/under the accession number E-MTAB-1424.

# Read mapping

Reads were trimmed using the Mott algorithm implemented in PoPoolation [18] (minimum read length = 40, quality threshold = 20). We mapped three RNA-Seq paired-end  $(2 \times 100 \text{ bp})$ libraries derived from ps88 females, ps94 females and hybrid females, against the D. pseudoobscura reference genome (FlyBase, assembly r2.23) using GSNAP [19]. Only non-ambiguously mapping read pairs were retained. To minimize the mapping bias, we first mapped the reads from the parental lines to the reference and identified SNPs fixed between the two strains. Then, we remapped the reads including the SNP information with GSNAP. Using an improved D. pseudoobscura annotation [17] we identified SNPs that were fixed (allele frequency = 1) for different alleles in the two parental lines. Based on these SNPs we generated two distinct reference genomes, one for each parental strain. Allele specific gene expression was measured by mapping the RNA-Seq reads to both genomes simultaneously and counting the number of reads mapping unambiguously to one of the two genomes (i.e. read pairs which covered at least one of the SNPs distinguishing the two genomes). The generation of two parental genomes not only identifies reads from either parental allele but also aims to avoid a well described mapping bias for different allelelic variants [20,21]. Moreover, it has been shown that after the adjustment for the different parental alleles via two reference genomes a mapping bias can remain for some genes [21]. To quantify and account for this remaining mapping bias we determined an empirical correction factor through simulations. These simulations included the following steps: 1) For each gene the same number of RNA-Seq reads (paired ends) were generated for the identical positions in the two genomes matching the read length and insert size of the empirical data. (i.e.: differing only by the identified SNPs). 2) The simulated RNA-Seq read libraries were mapped against both genomes simultaneously retaining only non-ambiguously mapped reads. 3) For genes with a different number of reads mapping to each of the genomes, we determined a correction factor, which is the inverse of the ratio of the mapped reads. This correction factor

was then included in the G-test to correct the Null-expectation for allele specific expression in hybrids.

Finally, we normalized the expression data using the TMM method [22]. The pipeline described above to identify allelespecific gene expression analysis is implemented in the Allim software package (http://code.google.com/p/allim/, [21]).

# Data analysis

Cis- and trans-effects were estimated by contrasting gene expression in the parental and F1 libraries. In order to test for significance of the regulatory scenarios we used the G-test of independence followed by correction using the Benjamini-Hochberg method at a false discovery rate cutoff (FDR) of 0.05. Testing was performed in a gene-wise manner. Total differential expression (TDE) between parents was estimated by comparing the ratio of the counts in the two parents to the ratio of total counts (i.e.: ratio of library sizes). For allele specific differential expression in F1 individuals (cis), we applied the same logic to the F1 library and compared the ratio of allelic counts at a given gene to the ratio of the allele counts summed over all genes. For the trans-test, we compared the ratio of parental gene expression counts to the ratio of allele specific counts in the F1 library. We note that this testing strategy results in a lower power to detect differences in trans compared to cis-effects since the total number of counts is higher in our tests for cis-effects than in our tests for trans-effects (see below). Following previous studies [6,23], we distinguish between different types of regulation namely cis, trans, cis-by-trans, cis+trans, compensatory, conserved and ambiguous (Table S1). Modes of inheritance were classified as additive, dominant, over-dominant, underdominant and conserved expression as described by Landry et al. [23] and McManus et al. [6] (Table S2).

# Statistical power estimation for *cis*- and *trans*-tests through simulations

We hypothesized that statistical power of the G-test for allelespecific differential expression (cis) is higher than for trans-test due to the fact that only one "noisy" allele-specific expression measurements were included into the cis-tests whereas two such measurements appeared in a trans-test. To support this hypothesis we performed computer simulations, in which we compared the power for different effect sizes. Since the power critically depends on the expression level of a given gene, we based our simulations on the observed expression level n (averaged over the three libraries) of randomly selected genes. The allelic imbalance  $\gamma$  was modulated on a grid of values ranging from 1 to 10 with the step size of 0.2. We simulated allele expression counts taken from a Poisson ( $\lambda$ ) distribution for the parental lines (P1 and P2) and from a Poisson  $(\lambda/2)$  distribution for the hybrid lines (H1 and H2). For each gene two different values  $\lambda_1$  and  $\lambda_2$  were computed according to  $\lambda_1 = 2n/(\gamma+1)$ ,  $\lambda_2 = 2n \gamma/(\gamma+1)$ . For pure *cis*-effects we simulated expression counts from H1 according to Poisson  $(\lambda_1/2)$  and for H2 according to Poisson ( $\lambda_2/2$ ), and Poisson ( $\lambda_1$ ) for P1 and Poisson  $(\lambda_2)$  for P2. For pure trans-effects, we took Poisson  $(\lambda_1)$  for P1 and Poisson ( $\lambda_2$ ) for P2, and Poisson (n/2) for H1 and H2. Combined cis- and trans-effects can be simulated by taking  $\lambda_{P1} = 2n/(\gamma_P + 1)$ ,  $\lambda_{P2} = 2n \gamma_P/(\gamma_P+1)$ ,  $\lambda_{H1} = 2n/(\gamma_H+1)$  and  $\lambda_{H2} = 2n \gamma_H/(\gamma_H+1)$  with allelic imbalances  $\gamma_{\rm P}$  and  $\gamma_{\rm H}$  for parents and hybrids. Using the proportion of rejections, the statistical power (1-Type II error) of the tests was estimated by simulating expression counts for 10000 genes under each imbalance parameter value ( $\gamma$ , and ( $\gamma_P$ ,  $\gamma_H$ )). All statistical analyses were performed using R.

### GO analysis

GO analysis were performed with FuncAssociate 2.0 [24] using the gene IDs of the *D. melanogaster* orthologs of *D. pseudoobscura*.

#### Results

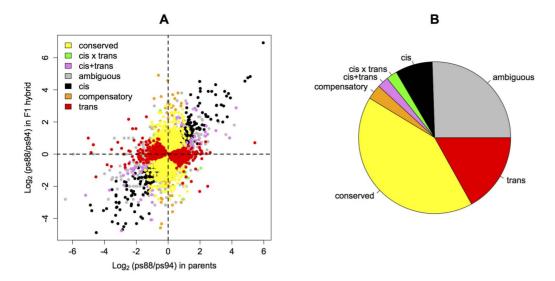
We performed RNA-Seq for two highly inbred parental lines (ps88 and ps94) and F1 individuals obtained from bi-directional crosses of the two inbred lines. On average, we obtained about  $41 \times 10^6$  reads for each library (Table S3), which resulted on average in about  $4 \times 10^6$  unambiguously mapping read pairs for each library after all filtering steps. Out of 16743 annotated genes in *D. pseudoobscura* [17], we identified 8116 (48.5%) genes with at least one fixed difference between the two lines. On average, we detected 4.4 fixed SNPs per gene between the two lines. 7631 genes with more than 20 reads in both parental lines (ps88+ps94 $\geq$ 20) were included in the subsequent analysis.

# Classification of cis- and trans-regulation

Gene expression is determined by *cis*- and *trans*-regulatory variation. Since variation in *cis*-regulatory elements results in allele specific gene expression, it is possible to measure *cis*-effects in F1 individuals. We detected significant (FDR≤0.05) differences in gene expression due to *cis*-effects between the two parental alleles for 1359 genes (including 154 genes from the ambiguous category, Table S1), which correspond to 18% of the expressed genes. *Trans*-effects were estimated by a significant (FDR≤0.05) difference in gene expression for the same allele in the parental and F1 background [3,4,5,6]. We identified 1982 (26%) genes (including 89 genes from the ambiguous category, Table S1) with significant *trans*-regulation.

Following the classification introduced by [23] we grouped the expressed genes in further sub-categories (Table S4). We identified 604 genes for which only cis-regulation was significant. More than twice as many genes (1292) had only significant trans-effects. We used computer simulations to test if the large excess of genes with trans-effects could be an artifact of our method. Interestingly, our simulations showed a slightly lower power to identify trans-effects over a wide range of imbalances (Fig. 1). For genes with combined cis- and trans-effects the power of the testing procedure was substantially reduced on average, as shown by the total area under the power curve that is smaller for combined effects than for pure cis and trans effects (Fig. 1). This lower power to detect trans- than cis-effects makes the observed excess of genes with trans-effects particularly relevant since it is most likely an underestimate. For 601 genes we identified combined cis- and trans-effects: cis-by-trans (176), cis+trans (183) and compensatory (242). For 3187 (42%) genes the gene expression pattern was conserved among the three samples. The remaining 1947 genes had an ambiguous expression pattern with no clear biological interpretation, i.e. the genes that cannot be placed in any category of regulation based on their patterns of parental and hybrid expression (Figure 2A,B). Using more stringent significance FDR thresholds of 0.01 and 0.005 did not change the overall pattern (Figure S1).

Our observation of more genes with *trans*-effects than *cis*-effects is consistent with the larger mutational target of *trans*-effects. To further scrutinize this hypothesis we determined the frequency distribution of *cis*- and *trans*-regulatory effect sizes (*cis*-magnitude = |log2(ps88 hybrid/ps94 hybrid)|, *trans*-magnitude = |log2(ps88/ps94) - log2(ps88 hybrid/ps94 hybrid)| and noticed a striking difference. In the small effect size class we detected a significant excess of genes with *trans*-effect. In all classes with a larger effect size, we found an excess of genes with *cis*-effects (Figure 3). Qualitatively the same result was obtained when genes with

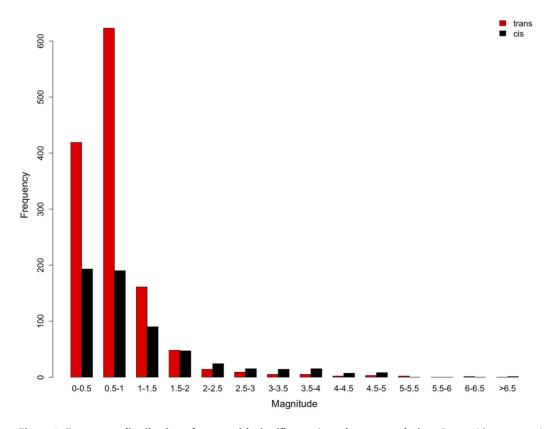


**Figure 2.** *Cis-* **and** *trans-***regulatory differences.** A) Distribution of regulatory effects depending on parental expression and allele specific expression in the hybrid. B) Relative distribution of the regulatory types (*cis*: 604 (7.9%); *trans*: 1292 (16.9%); *cis*-by-*trans*: 176 (2.3%); *cis*+*trans*: 183 (2.4%); compensatory: 242 (3.2%); conserved: 3187 (41.8%); ambiguous: 1947 (25.5%)). doi:10.1371/journal.pone.0083547.g002

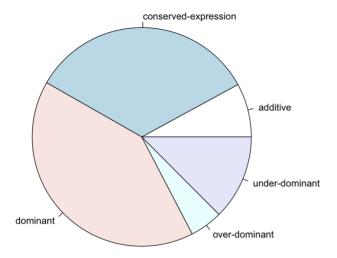
combined effects were also included (data not shown). This marked difference between effect size classes is consistent with stronger purifying selection operating on genes with *trans*-effects, possibly due to their pleiotropic nature.

# Dissecting modes of inheritance

The comparison of expression intensities in F1 individuals relative to the one in their parents provides information about the mode of inheritance (Table S2). Out of 7631 genes analyzed, 607 genes were classified as additive and 3120 genes as dominant. 1326 genes were misexpressed in the F1 individuals, with 369 genes



**Figure 3. Frequency distribution of genes with significant** *cis*- **and** *trans*- **regulation.** Genes with pure *trans* (red) and *cis* (black) effects were grouped according to the magnitude of their effect. doi:10.1371/journal.pone.0083547.g003



**Figure 4. Modes of inheritance.** Relative proportions of the inheritance modes (dominant: 3120 (40.9%); additive: 607 (8%); overdominant: 369 (4.8%); under-dominant: 957 (12.5%); conserved-expression: 258 (33.8%)).

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being over-dominant (higher expression in the F1) and 957 genes being under-dominant (lower expression in the F1 than in both parents). 2578 genes had no significant difference between parents and the F1 (conserved, Figure 4). We note, that the high fraction of genes with conserved inheritance reflects statistical power of our experiment rather than a true biological signal since the genes with conserved mode of inheritance have lower expression intensities (ps88 parent+ps94 parent+F1) than those genes with additive, dominant or misexpressed modes of inheritance (Wilcoxon's one-tailed rank-sum test, P<2.2e-16). As expected, we found no indication of one parent being over-represented among the genes with a dominant mode of inheritance: among 3 120 genes 1518 (~49%) and 1602 (~51%) genes have ps88-like and ps94-like expression.

Intra-specific comparisons suggested that *cis*-regulation tends towards additivity whereas *trans*-regulation shows higher degrees of dominance [25]. Consistent with these previous results, we found that *cis*-magnitude ( $|\log 2(\text{ps88 hybrid/ps94 hybrid})|$ ) within genes with additive mode of inheritance was significantly greater than the *cis*-magnitude of genes with other significant inheritance types (Wilcoxon's one-tailed rank-sum test, P = 0.007738) (Figure S2).

# GO analysis

We tested for an enrichment of functional categories among the different modes of gene regulation using a gene ontology (GO) analysis. While some marginal significance was detected for some modes of gene regulation, none of them survived a correction for multiple testing. Similarly, no significant enrichment of a GO category was observed for the different modes of inheritance. Hence, we conclude that our data do not support a functional differentiation among the categories studied.

# Discussion

This report is the first study of intra-specific *cis*-and *trans*-effects using RNA-Seq in *Drosophila* and the first report for *D. pseudoobscura*. We observe about twice as many genes with pure *trans*-effects (1292) as genes with pure *cis*-effects (604). This result is consistent with earlier studies in *Drosophila* [4], *Arabidopsis* [16] and yeast [7]. Hence, we suggest that this pattern is most likely general

and contrasting results [5] probably reflect more methodological differences than an alternative biological phenomenon.

Of particular interest is the contrast to studies of cis- and transeffects relying on the analysis of expression differences between closely related species and their corresponding hybrids. These analyses typically observed a clear excess of genes with cis-effects [4,8,9,26]. The prevailing explanation for this apparent discrepancy is that cis-effects affect only a single gene, while trans-effects are typically pleiotropic [3,4,27]. Hence, cis-effects become more easily fixed between species [28,29]. Within species, however, the larger mutational target of trans-effects results in the observed excess of trans-effects [4,7]. In our report, we have further scrutinized the hypothesis of differential selection operating on cis- and trans-effects by contrasting the frequency distribution of different effect sizes. Reasoning that trans-effect mutations with a large effect size are more effectively purged than cis-effect ones, we predicted an underrepresentation of genes with trans-effects in the large effect size classes. In fact, our data show a trend of a more pronounced underrepresentation of genes with *trans*-effects with an increasing effect size (Figure 3). Furthermore, the larger number of genes with trans-effects is consistent with them being a larger mutational target. Alternatively, the observed distribution of cisand trans-effects (Figure 3) could be also obtained if new transmutations have, in general, smaller regulatory effect sizes than mutations in regulatory regions.

Overall, we conclude that the patterns of intra-specific and inter-specific cis- and trans-effects discussed above are compatible with the possibility that most regulatory variation is deleterious and their distribution reflects the balance between the occurrence of new mutations and their different fixation probabilities. This idea is in line with other results showing that purifying selection is the major force affecting the evolution of gene expression [30]. Nevertheless, if purifying selection shapes the pattern of trans-effects while positive selection contributes largely to cis-effects [31], the distribution of intra-specific cis- and trans-effect sizes may also differ.

We caution that the evolution of gene expression is probably more complex. One indication for this could be obtained from the comparison of the effect sizes in intra- and inter-specific comparisons. While we found an underrepresentation of genes with large *trans*-effects, inter-specific comparisons suggest large effect sizes of *trans*-effects [6]. We hypothesize that this discrepancy could be due to species specific *trans*-regulatory mutations that may, for example, be involved in compensatory changes.

To obtain a complete picture of the evolutionary dynamics of *cis*and *trans*-effects it is important to analyze their context specific
effects. Depending on the developmental stage, tissue, and
environment the regulatory landscape may differ and consequently
also the effects of *cis*- and *trans*-regulatory variation. More
experiments are needed to understand to what extent the patterns
described for a single sex, developmental stage and environment can
be generalized. Furthermore, future studies may benefit from
replication, which will improve the accuracy of the estimated effects
beyond the already high repeatability of RNA-Seq studies [32].

# **Supporting Information**

Figure S1 Effect of different FDR threshold on the identification of regulatory effects. A) FDR < 0.05. B) FDR < 0.01. C) FDR < 0.005. (PDF)

Figure S2 Cis-magnitude comparison of genes with additive and non-additive modes of inheritance. Genes with additive mode (A) of inheritance exhibit higher cis-magnitude

vs. all other genes with non-additive modes (NA) excluding genes with conserved expression mode.

(PDF)

**Table S1** Determination of regulatory categories based on total differential expression between parents (TDE), allele-specific differential expression in F1 hybrid (cis) and trans-test. All the statistical tests were performed using G-test, followed by FDR correction (q $\leq$ 0.05). (DOC)

**Table S2** Identification of inheritance modes based on differential expression and directionality of expression between parental and total expression in F1 hybrids. (DOC)

**Table S3** Library size information. (DOC)

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**Table S4** Gene summary table. This table provides information about raw and normalized expression counts for parents and hybrids, statistical tests, regulatory types, inheritance modes and *cis* and *trans* magnitudes for each gene revealed by RNA-seq. (XLS)

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#### **Author Contributions**

Conceived and designed the experiments: CS. Performed the experiments: VN. Analyzed the data: AS. Contributed reagents/materials/analysis tools: RVP SF AF. Wrote the paper: CS VN AS.

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