

1 **Modulation of protein-DNA binding reveals mechanisms of spatiotemporal gene control in**  
2 **early *Drosophila* embryos**

3 Sahla Syed<sup>1</sup>, Yifei Duan<sup>1,2</sup>, and Bomyi Lim<sup>1\*</sup>

4 <sup>1</sup> Department of Chemical and Biomolecular Engineering, University of Pennsylvania,  
5 Philadelphia, PA 19104

6 <sup>2</sup> Master of Biotechnology Program, University of Pennsylvania, Philadelphia, PA 19104

7 \* To whom correspondence may be addressed: [bomyilim@seas.upenn.edu](mailto:bomyilim@seas.upenn.edu)

8 **Abstract**

9 It is well known that enhancers regulate the spatiotemporal expression of their target genes by  
10 recruiting transcription factors (TFs) to the cognate binding sites in the region. However, the role  
11 of multiple binding sites for the same TFs and their specific spatial arrangement in determining  
12 the overall competency of the enhancer has yet to be fully understood. In this study, we utilized  
13 the MS2-MCP live imaging technique to quantitatively analyze the regulatory logic of the *snail*  
14 distal enhancer in early *Drosophila* embryos. Through systematic modulation of Dorsal and Twist  
15 binding motifs in this enhancer, we found that a mutation in any one of these binding sites causes  
16 a drastic reduction in transcriptional amplitude, resulting in a reduction in total mRNA production  
17 of the target gene. We provide evidence of synergy, such that multiple binding sites with moderate  
18 affinities cooperatively recruit more TFs to drive stronger transcriptional activity than a single site.  
19 Moreover, a Hidden Markov-based stochastic model of transcription reveals that embryos with  
20 mutated binding sites have a higher probability of returning to the inactive promoter state. We  
21 propose that TF-DNA binding regulates spatial and temporal gene expression and drives robust  
22 pattern formation by modulating transcriptional kinetics and tuning bursting rates.

23 **Key words**

24 Transcription kinetics; live imaging; MS2-MCP; transcription factors; bursting; Hidden Markov  
25 model

26

27 **Introduction**

28 Development of a *Drosophila* embryo is a highly precise and coordinated process, occurring with  
29 little variability despite intrinsic and extrinsic noise and perturbations (Arias and Hayward 2006;  
30 Houchmandzadeh, Wieschaus, and Leibler 2002). Proper levels of essential genes and correct  
31 positioning of expression patterns are regulated by short non-coding DNA sequences known as  
32 enhancers (Banerji, Rusconi, and Schaffner 1981). Enhancers tightly control their target gene  
33 expression both in space and time via transcription factor (TF) recruitment. Complex patterning  
34 and cell fates are established through TFs recognizing and binding to specific short DNA  
35 sequences within enhancers with varying degrees of affinity at different developmental stages  
36 (Long, Prescott, and Wysocka 2016; Reiter, Wienerroither, and Stark 2017; Ramos and Barolo  
37 2013b). Concerted action of TFs with other transcriptional machinery has been found to reposition  
38 nucleosomes, initiate chromatin remodeling, recruit additional activating co-factors, and generate  
39 distinct transcriptional outputs (Spitz and Furlong 2012). However, it remains to be understood  
40 how these brief, yet frequent, interactions between TFs and regulatory DNAs facilitate efficient  
41 and specific transcription on the timescale of minutes. Although we know that TFs influence  
42 various facets of transcription such as timing or probability of activation, we have yet to determine  
43 their role in orchestrating an enhancer's transcriptional competency at a mechanistic level. For  
44 example, does the spatial arrangement of the binding sites influence transcriptional capability?  
45 How does each TF binding site shape transcriptional dynamics of individual nuclei and contribute  
46 to overall pattern formation?

47 Recently, it was shown that the genomic context of an enhancer provides an optimal environment  
48 for driving normal expression patterns and preventing misregulation upon induced perturbations.  
49 Mutating a single Giant repressor binding site in a minimal *even-skipped* stripe 2 enhancer region  
50 caused misexpression of the target gene, whereas those effects were buffered in an extended  
51 enhancer containing more TF binding sites and expression levels were comparable to the wildtype  
52 (López-Rivera et al. 2020). However, the role of multiple TF binding sites with varying affinities  
53 within the enhancer in regulating transcription has yet to be established. Recent studies have  
54 explored the role of low-affinity binding sites in producing specific expression patterns and found  
55 that enhancers containing optimal TF motifs may lead to overexpression and result in  
56 developmental defects (Farley et al. 2015; Ramos and Barolo 2013a; Tsai et al. 2017). Previous  
57 work has shown that modulating the strength of a single TF binding site was sufficient to disrupt  
58 transcriptional activity, such that a mutation of an activator Dorsal (DI) site in the *t48* enhancer  
59 delayed activation and almost completely abolished transcriptional activity, while optimization of  
60 the site to a consensus motif induced ectopic transcriptional activity with a broader gene  
61 expression domain (Keller et al. 2020). However, systematic removal of binding sites of varying  
62 affinities for another activator Bicoid seems to affect its target gene, *hunchback*, expression to a  
63 similar degree, indicating that each site has a nearly equal contribution to the overall expression  
64 pattern (Eck et al. 2020). Yet, since many studies have relied on fixed tissue experiments to derive  
65 the role of TFs in transcriptional regulation, the changes in real-time transcription kinetics that  
66 drive the observed misexpression are often overlooked. Dynamic interplay among TFs, cofactors,  
67 and DNA occurs on the order of seconds, a time resolution that cannot be resolved solely through  
68 RNAi and single molecule *in situ* hybridization experiments (Mir et al. 2017). Since TF binding  
69 events affect the expression of regulatory genes both spatially and temporally, incorporation of  
70 both live imaging techniques and predictive modeling is crucial to correlate transient TF-DNA  
71 binding to downstream transcriptional activity in single-cell resolution.

72 In this study, we investigated the effects of perturbing TF-DNA binding strength on the  
73 transcriptional dynamics of *snail* (*sna*) in early *Drosophila* embryos. *sna* is a well-characterized,  
74 key patterning gene that encodes a zinc finger protein and is responsible for the differentiation of  
75 the mesoderm (Rembold et al. 2014; Ip et al. 1992; Leptin 1991). *Sna* represses the expression  
76 of genes responsible for neuroectoderm formation and establishes the mesoderm-neuroectoderm  
77 boundary (Kosman et al. 1991). Embryos lacking *sna* fail to undergo gastrulation, resulting in  
78 embryonic lethality (Hemavathy et al. 2004). Previous studies have demonstrated that *sna*  
79 expression is controlled by a proximal enhancer and a distal (shadow) enhancer located directly  
80 upstream and ~7 kb upstream of the promoter, respectively (Perry et al. 2010). The distal  
81 enhancer is necessary for proper *sna* expression and the viability of the developing embryo,  
82 especially under genetic and environmental stresses (Perry et al. 2010; Dunipace, Ozdemir, and  
83 Stathopoulos 2011). *sna* is a target gene of the DI morphogen, and the nuclear gradient of  
84 maternally deposited DI protein controls the sharp boundaries of *sna* expression, such that only  
85 nuclei with high concentrations of nuclear DI express *sna* (Figure 1B) (Hong et al. 2008). Through  
86 binding assays like EMSA and ChIP-seq, it was determined that the distal *sna* enhancer contains  
87 multiple, low affinity binding sites for DI, Twist (Twi), and the pioneer factor Zelda (Zld) (Figure  
88 1A) (Zeitlinger et al. 2007; Ferraro et al. 2016). Indeed, *sna* expression is completely abolished in  
89 embryos lacking DI or Twi, and Zld null embryos show a delay in *sna* activation (Dufourt et al.  
90 2018; Liang et al. 2008).

91 Here, we utilize a combination of quantitative live imaging and mathematical modeling to probe  
92 the underlying regulatory mechanisms that TFs employ to initiate transcription, regulate gene  
93 expression levels, and establish spatial boundaries. Using MS2-MS2 coat protein (MCP) based  
94 live imaging, we visualized transcription dynamics driven by the wildtype minimal *sna* distal  
95 enhancer in the cases of with and without various TF binding site mutations. We find that mutating  
96 a single TF (DI or Twi) binding site in the enhancer significantly reduces mRNA production of the

97 target gene, mainly by lowering transcriptional amplitude, without significantly delaying the timing  
98 of initiation or affecting the probability of activation. Surprisingly, we found that modulating the  
99 same TF binding site in the context of the full *sna* distal enhancer results in a similar reduction in  
100 expression levels, underscoring that additional TF binding sites in the full enhancer are not  
101 sufficient to rescue reduced transcriptional activity. Using a thermodynamic equilibrium binding  
102 model, we show that the TF binding sites must cooperate with each other to establish the correct  
103 spatial pattern of *sna*. Moreover, a two-state stochastic model of transcription indicates that TF  
104 binding site mutations affect transcriptional bursting, specifically by increasing the probability of  
105 the promoter switching out of the ON state,  $k_{off}$ , and reducing burst durations. Together, our data  
106 highlight the distinct mechanisms by which TF binding sites regulate transcriptional kinetics and  
107 spatial patterning during embryonic development.

## 108 **Results**

### 109 **Single TF binding site mutation greatly diminishes the transcriptional capability of the** 110 **distal *sna* enhancer**

111 The 519 bp *sna* minimal enhancer is located within the full *sna* distal enhancer (1.8 kb) and has  
112 been shown to recapitulate proper *sna* expression (Figure 1A) (Ferraro et al. 2016). We focus on  
113 the analysis of the minimal enhancer since it contains fewer TF binding sites than the full distal  
114 enhancer, which allows systematic perturbation in a sensitized background to gain a functional  
115 understanding of the role of TF binding site arrangements in gene regulation. The minimal  
116 enhancer contains binding sites for many TFs, including Df, Twi, and Zld, and our study focuses  
117 on the sites with the strongest binding affinities, three Df sites and one Twi site (Figure 1A). We  
118 note that all Df and Twi sites still have relatively low binding affinities compared to the consensus  
119 motif, because *sna* is activated only in the domain with high nuclear Df concentration (Hong et al.  
120 2008). We systematically deleted one site at a time by introducing point mutations in each binding

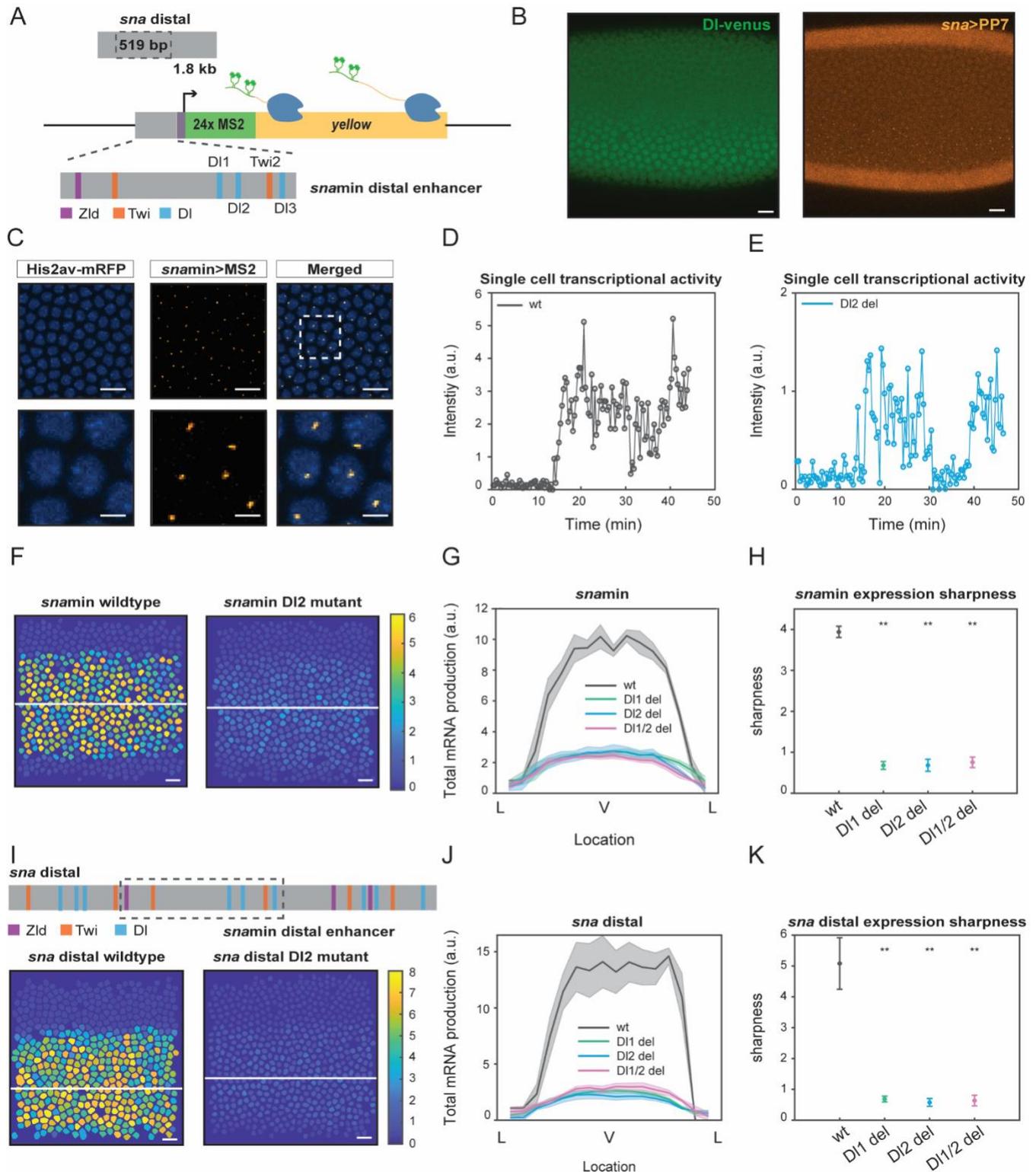
121 motif. The mutations were created by inducing double-nucleotide substitutions that cause the  
122 sequence to no longer be recognized as a TF motif match (Table 1).

123

124 We utilized the MS2-MCP live imaging technique to quantitatively analyze the effects of  
125 modulating TF binding sites within the minimal *sna* distal (*sna<sub>min</sub>*) enhancer. Specifically, the 10x  
126 repeat sequences of MS2 are integrated into the 5' UTR of the *yellow* reporter gene (Figure 1A).  
127 Upon transcription, each MS2 sequence forms a stem loop which is detected by maternally  
128 deposited MCP tagged with fluorescent proteins. To visualize nascent transcription dynamics,  
129 females carrying His2Av-mRFP and MCP-GFP were crossed with males carrying the desired  
130 MS2 construct, resulting in progeny expressing distinct fluorescent puncta in active transcription  
131 loci (Figure 1C, Video 1). *sna* distal enhancer drives gene expression in the endogenous *sna*  
132 domain, where the nuclear DI level is the highest (Figure 1B). The fluorescence intensity trajectory  
133 for each nucleus is extracted for subsequent analysis and correlated to its instantaneous  
134 transcriptional activity (Figure 1D,E).

135

136 We found that a mutation of any single DI or Twi binding site within the minimal enhancer resulted  
137 in a dramatic reduction in mRNA production (Figure 1F, Video 1,2, Figure 1 supplement 1A).  
138 Nuclei in the center of the *sna* expression domain of a mutant embryo produce 65% less mRNA  
139 than those in the wildtype counterpart. The reduction occurs uniformly across the *sna* expression  
140 domain without significantly affecting the width (Figure 1G, Figure 1 supplement 1C). It is  
141 important to note that mutations of different TFs (DI and Twi) and of binding sites with different  
142 affinities (DI1 and DI2) all result in a similar degree of decreased mRNA production. Surprisingly,  
143 even the embryos with two mutated binding sites exhibit a similar mRNA reduction (Figure 1G –  
144 pink, Figure 1 supplement 1A - orange). These results suggest that each site in the minimal  
145 enhancer is necessary to drive normal *sna* expression, perhaps due to the sensitized background  
146 in which the mutations were induced.



**Figure 1. Single TF binding site mutation greatly diminishes mRNA production from the distal *sna* enhancer**

(A) Schematic of the reporter construct containing the minimal *sna* distal enhancer, *sna* core promoter, MS2 stem loops, and the *yellow* reporter gene. The minimal enhancer contains binding sites for TFs Zld, Twi, and DI.

(B) Embryo expressing maternally deposited DI-Venus protein (green) and *sna* distal>*PP7-yellow* reporter gene (orange). *sna* is expressed in the region with high nuclear DI.

(C) Snapshots of an embryo expressing minimal *sna* distal>*MS2-yellow*. The nuclei are marked with Histone-RFP (blue) and the *MS2-yellow* reporter gene is observed with MCP-GFP (orange). Each nucleus has one distinct fluorescent punctum, indicating nascent transcription. Bottom row are magnifications of the embryo within the rectangle.

(D,E) Single nucleus transcriptional trajectories for a wildtype and DI2 mutant embryo, respectively.

(F) Heatmap showing that mRNA production is higher in a wildtype embryo compared to a DI2 mutant embryo. The white line indicates the ventral midline.

(G) Average mRNA production of all nuclei in wildtype embryos and DI1, DI2, and DI1/2 mutant embryos across the *sna* expression domain (Lateral-Ventral-Lateral).

(H) Plot indicating the wildtype steepness of *snamin* expression is significantly higher than the mutants.

(I) Top: Schematic showing the additional TF binding sites present in the full *sna* distal enhancer. Bottom: Heatmap showing higher mRNA production of *MS2-yellow* in a wildtype embryo containing the full *sna* distal enhancer compared to the embryo containing the full enhancer with DI2 site mutations.

(J) Average mRNA production of all nuclei in wildtype embryos and DI1, DI2, and DI1/2 mutant embryos containing the full *sna* distal enhancer across the expression domain.

(K) Plot indicating the wildtype steepness of *sna* distal expression is significantly higher than the mutants.

Shaded error bars in (E), (G), and (J) indicate standard error of the mean (SEM).

Error bars in (H) and (K) indicate SEM.

1524 nuclei from 3 replicate wildtype embryos, 1672 nuclei from 4 replicate DI1 mutant embryos, 2091 nuclei from four replicate DI2 mutant embryos, and 1788 nuclei from three replicate DI1/2 mutant embryos were analyzed.

\*\* denote  $p < 0.001$  from the student's t-test.

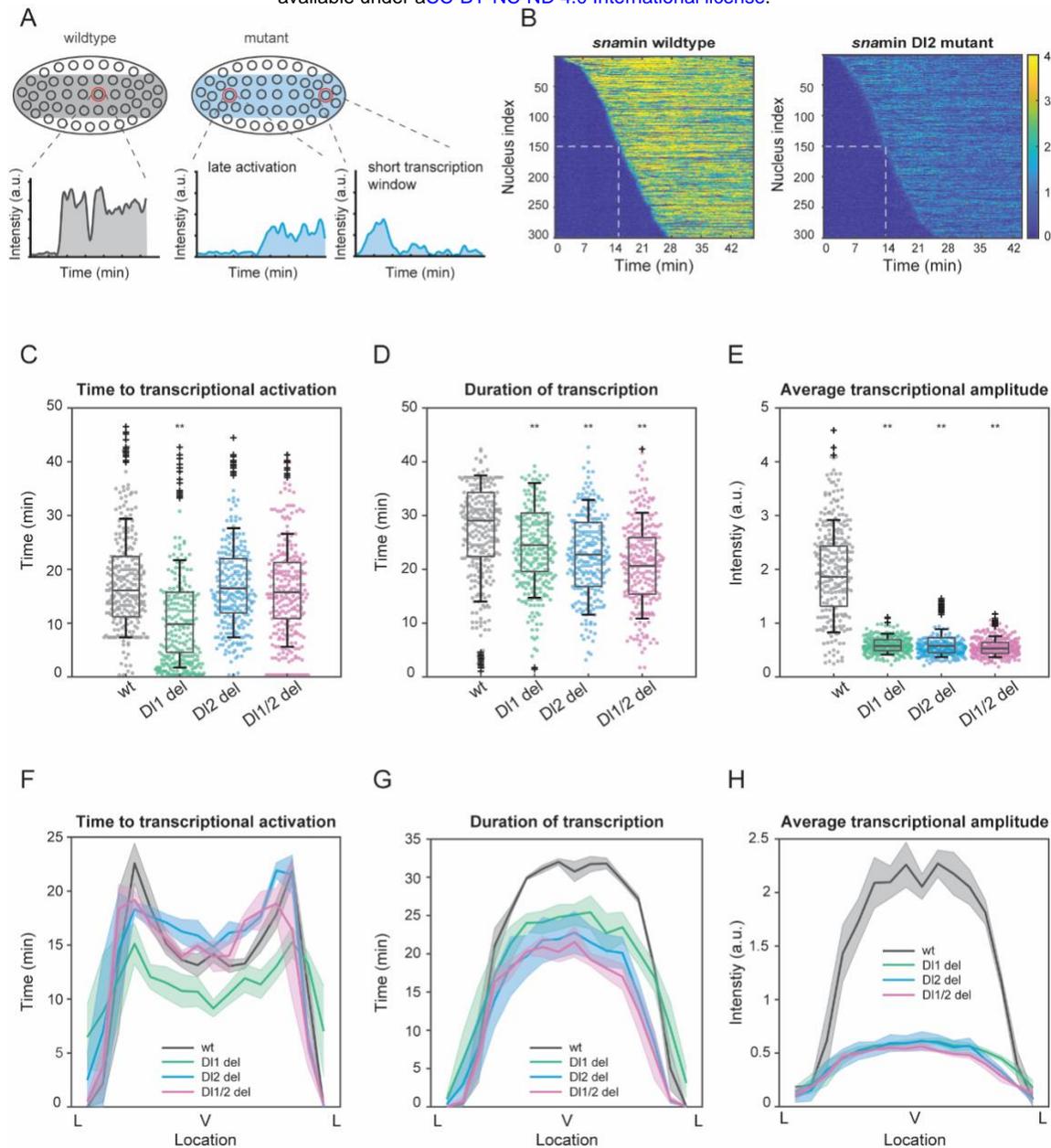
147

148 To further examine the significance of these core TF binding sites in the minimal enhancer, we  
149 investigated if the additional TF binding sites in the full distal enhancer would buffer against the  
150 drastic changes in mRNA production caused by the mutations. Despite several additional DI and  
151 Twi sites flanking the minimal enhancer, we found that the same mutations on the same TF motifs  
152 caused a similar decrease in mRNA production (Figure 1I,J). Furthermore, we found that in both  
153 the minimal and full distal enhancers, the mutant embryos have shallower gradients and less  
154 sharp boundaries, highlighting the importance of proper TF-DNA interactions in regulating the  
155 sharpness of the expression domain (Figure 1H,K, Figure 1 supplement 1B). Shallower  
156 expression of *sna* may lead to higher uncertainty in germ layer formation between mesoderm and  
157 neuroectoderm. It is interesting to note that the introduction of the mutations in the larger genomic  
158 context (i.e., full distal enhancer) did not lessen their effect on transcriptional activity and pattern  
159 formation. Here, it is evident that additional TF sites are not able to rescue normal *sna*  
160 transcriptional activity and that each site within the minimal enhancer region plays a critical role  
161 in ensuring robust expression.

162

### 163 **Mutations cause lower mRNA production, mainly due to reduced transcriptional amplitude**

164 After establishing the dramatic reduction in transcriptional activity, we delved into the underlying  
165 causes of the low mRNA production in both a single nucleus and across the entire *sna* expression  
166 domain. We hypothesized that the reduced mRNA production may occur through multiple different  
167 modes. The mutations may affect the transcriptional capability of the enhancer by altering the  
168 time of transcriptional activation or by reducing the transcriptional window, thereby effectively  
169 lowering the mRNA production compared to a wildtype embryo. Or the mutations may reduce the  
170 enhancer's ability to transcribe and effectively load Pol II, resulting in decreased instantaneous  
171 transcriptional amplitude (Figure 2A). We found that the differences in the time for half the nuclei



**Figure 2. Mutations cause lower mRNA production, mainly due to lower transcriptional amplitude**

(A) Schematic comparing wildtype and mutants. The mutant embryo may have fewer nuclei transcribing and at a lower intensity. Each nucleus in mutant embryos may have late activation or a shorter transcription window, all of which may contribute to the observed low mRNA production.

(B) Heatmap of transcription activation times for representative wildtype and DI2 mutant embryos. The time at which half of the nuclei are activated is indicated by the dotted white line and there is no significant difference.

(C-E) Boxplots showing (C) the time to transcriptional activation (D) the duration of active transcription, and (E) the transcriptional amplitude for all actively transcribing nuclei. Decreased transcriptional amplitude contributes the most to the low mRNA production in mutants.

(F-G) (F) Average time to transcriptional activation, (G) average duration of transcription for all actively transcribing nuclei, and (H) average transcriptional amplitude for all nuclei across the *sna* expression domain (Lateral – Ventral – Lateral). Nuclei in the middle of the expression domain are affected more, but there is no significant change in the expression width.

Shaded error bars in (F-H) indicate SEM.

250 individual data points are overlaid on the respective boxplots.

1124 nuclei from 3 replicate wildtype embryos, 1011 nuclei from 4 replicate DI1 mutant embryos, 1123 nuclei from four replicate DI2 mutant embryos, and 943 nuclei from three replicate DI1/2 mutant embryos were analyzed.

Transcription of the *MS2-yellow* reporter gene is driven by the minimal *sna* distal enhancer.

\*\* denote  $p < 0.001$  from the student's t-test.

172 to begin transcription as well as the transcription initiation time per nucleus were not sufficient to  
173 explain the low mRNA output (Figure 2B,C, Figure 2 supplement 1A). Although the duration of  
174 active transcription is slightly lowered in embryos containing the induced mutations (Figure 2D,  
175 Figure 2 supplement 1B), the main cause of the low mRNA output was the average transcriptional  
176 amplitude (Figure 2E, Figure 2 supplement 1C). Here, we observed a significant decrease in  
177 transcriptional intensity, leading us to conclude that the mutations mainly modulate transcription  
178 by lowering Pol II loading rate. At single-cell resolution, we find that the mutants with two deleted  
179 binding sites have a slightly bigger impact on transcriptional activity than those with one site  
180 removed, but the differences are minimal (Figure 2C-E, Figure 2 supplement 1A-C).

181

182 Since *sna* is a patterning gene and is responsible for the formation of the ventral furrow and  
183 presumptive mesoderm, we wanted to determine how the mutations spatially affect the  
184 aforementioned transcription parameters and the spatial boundaries of the expression pattern.  
185 The nuclei in the center of the *sna* expression domain are more substantially affected, exhibiting  
186 lower duration of transcription and average transcriptional amplitude, confirming the trend we  
187 observed for mRNA production (Figure 2F,G, Figure 2 supplement 1D-F). Our results agree with  
188 previous studies, in which the center nuclei within the *eve* stripe 2 domain had significantly longer  
189 transcription windows and overall, higher rates of mRNA production (Lammers et al. 2020).  
190 Interestingly, we observed that the DI1 mutation induced earlier transcriptional activation than  
191 wildtype (Figure 2C,F). Since the DI1 and DI2 binding sites are less than 10 bp apart, mutating  
192 the DI1 site may create a more favorable steric conformation, allowing DI to bind to the single site  
193 more efficiently. However, the earlier activation time is not sufficient to buffer against the severe  
194 reduction in transcriptional amplitude and causes the DI1 mutant to exhibit similarly reduced  
195 mRNA production (Figure 1G).

196

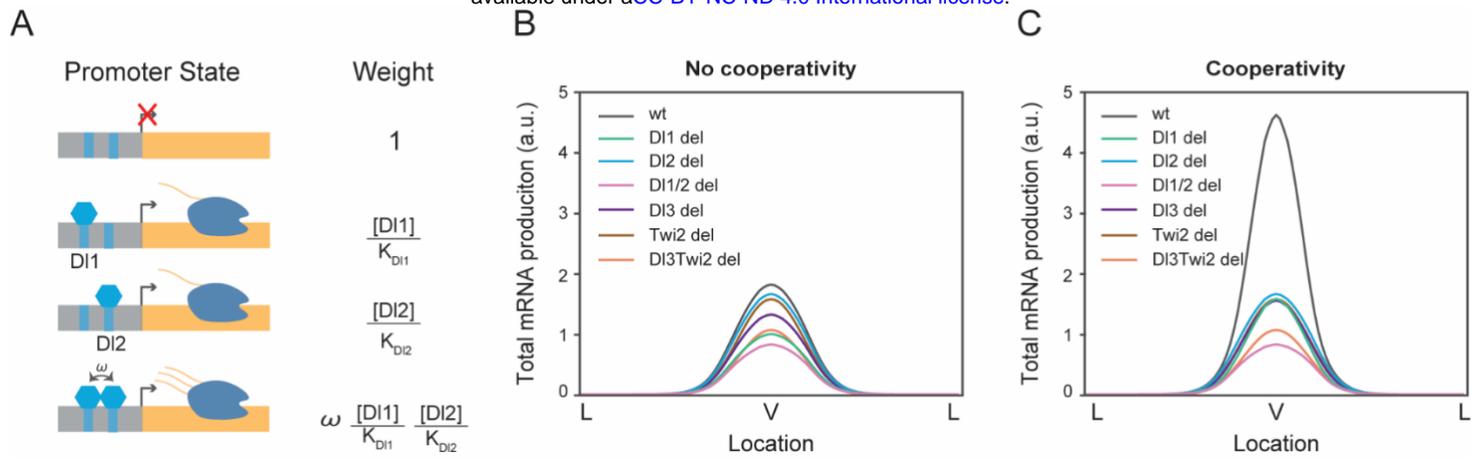
197 Interestingly, in addition to modulating the average transcriptional intensity, the mutations in the  
198 full distal enhancer affect the time to activation and the transcription window as well.  
199 Transcriptional activation is more significantly delayed and the duration of transcription is  
200 substantially lower in the mutants than in the wildtype (Figure 2 supplement 2A-C). As with the  
201 case with the minimal enhancer, here we do not observe any significant spatial modulation of  
202 these parameters and the expression width remains unaffected (although the boundary is less  
203 sharp) (Figure 2 supplement 2D-F).

204

## 205 **Thermodynamic equilibrium binding model reveals synergistic interactions among TF** 206 **binding sites**

207 The mechanistic role that multiple TF binding sites with different affinities play in regulating  
208 enhancer activity and capability is still unclear. Is a single binding site sufficient to establish the  
209 correct pattern and expression levels? In that case, why does an enhancer contain multiple TF  
210 binding sites for the same TF? Our findings seem to indicate a nonadditive behavior between  
211 binding sites, such that both a single and double mutation affect transcriptional dynamics to a  
212 similar degree. We utilized a thermodynamic Monod-Wyman-Changeux (MWC) model to examine  
213 the contributions of each DI and Twi site to the overall competency of the enhancer (Monod,  
214 Wyman, and Changeux 1965). Here, we assume that the microstates (unbound, bound with  
215 activator(s)) are in equilibrium and that the probability of each state can be correlated with its  
216 Boltzmann weight (Eck et al. 2020). As described in (Eck et al. 2020; Kanodia et al. 2012), the  
217 Boltzmann weights can be calculated in terms of activator concentration  $C$ , dissociation constant,  
218  $K$ , and a cooperativity term  $\omega$  (Figure 3A). The probability of transcription initiation is represented  
219 by the probability of an activator (i.e., DI) binding to its cognate site, which can be written as:

$$220 \quad p_{bound} = \frac{\frac{C_{DI1}}{K_{DI1}} + \frac{C_{DI2}}{K_{DI2}} + \omega \frac{C_{DI1}C_{DI2}}{K_{DI1}K_{DI2}}}{1 + \frac{C_{DI1}}{K_{DI1}} + \frac{C_{DI2}}{K_{DI2}} + \omega \frac{C_{DI1}C_{DI2}}{K_{DI1}K_{DI2}}}$$



**Figure 3. Thermodynamic equilibrium binding model reveals synergy among TF binding sites**

(A) Promoter states and statistical weights for each microstate. A bound activator will yield transcription. Cooperativity term  $\omega$  is included when more than one TF is bound which will result in higher mRNA production. Dissociation constants are given by  $K_i$ , which is correlated with the binding affinity of each site.

(B-C) mRNA production curves generated by assuming (B) no cooperativity and (C) cooperativity among TF binding sites. Modeling results support experimental data with the cooperativity term included.

221

222 This equation can be expanded to include all the combinations of DI and Twi interactions (See  
223 Methods, Equation 1). The equilibrium binding constants were correlated to the p-values  
224 calculated from the PWM scanning software, FIMO (Bailey et al. 2015) (see Methods). We first  
225 tested our model in the case of completely independent binding, or no cooperativity. In this  
226 condition, the model failed to predict the experimental data and we found that the wildtype has  
227 only slightly higher expression levels compared to the mutants (Figure 3B). When we modified  
228 the model to include the cooperativity terms among all the TFs, we were able to qualitatively  
229 recapitulate the experimental mRNA production curve as well as determine which TFs work  
230 synergistically to help recruit more TFs to the enhancer region (Figure 3C). We found that  
231 DI1/DI2/Twi, DI2/DI3/Twi, and DI1/DI2/DI3/Twi cooperativities were the key parameters to reach  
232 a stable solution, confirming the role of cooperativity among weak binding sites as a mechanism  
233 for precise gene control. We note that the deletion of one or two sites would render no  
234 cooperativity, and hence would result in a similar effect on transcription. Our data demonstrate  
235 that TF binding sites must coordinate with one another to some degree to recreate the correct  
236 pattern and levels of *sna*. Synergistic interactions among TFs have been observed in other genes  
237 as well, such that the transcriptional output of multiple TFs is significantly higher than the sum of  
238 their individual activities (Keller et al. 2020; Park et al. 2019). We propose that the cooperativity  
239 allows TF binding sites with moderate or weak affinities to recruit more TFs to the enhancer,  
240 generate sharp transcriptional responses, and drive strong and robust expression in the narrow  
241 *sna* expression domain (see Discussion).

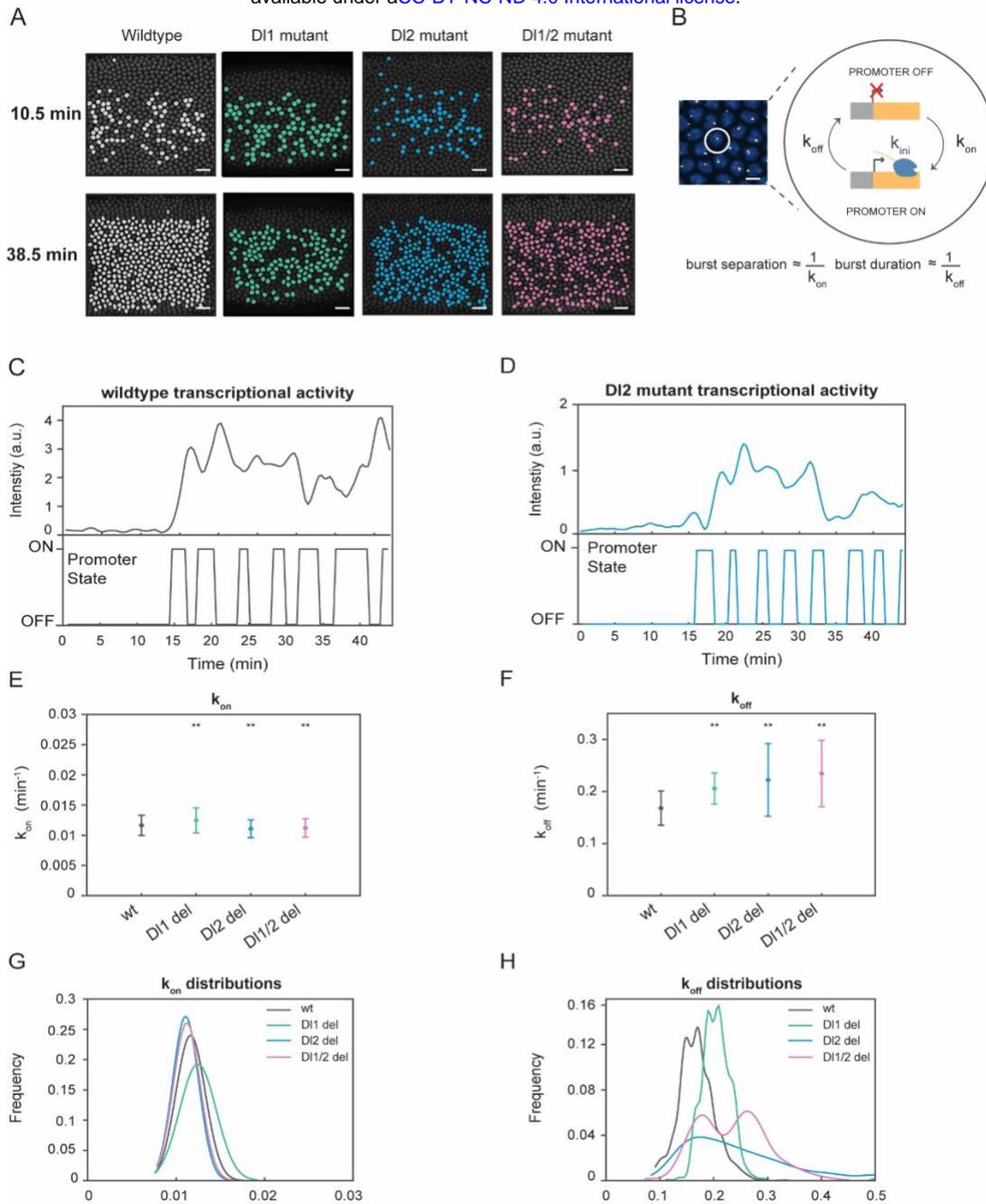
242

### 243 **A two-state model of transcription reveals differences in $k_{\text{off}}$ rates and burst duration**

244 As shown previously, transcription occurs discontinuously in distinct, stochastic bursts of activity  
245 punctuated by quiescence (Corrigan et al. 2016; Senecal et al. 2014; Donovan et al. 2019).  
246 Bursting has been proposed as an evolutionary mechanism for driving heterogeneity in gene

247 expression, giving rise to cell-to-cell variability and overall diversity (Rodriguez and Larson 2020).  
248 Bursting parameters, such as burst frequency, burst duration, and promoter switching rates  
249 provide a glimpse into the underlying mechanisms of dynamic transcription regulation, such as  
250 kinetic rates and promoter states. We find that the wildtype and mutant embryos show comparable  
251 numbers of actively transcribing nuclei at the beginning of the nuclear cycle 14. However, at later  
252 times, most wildtype nuclei in the *sna* expression domain are active in a given frame while nuclei  
253 from mutant embryos exhibit stochastic activity (Figure 4A, Video 3,4). Indeed, by quantifying the  
254 number of nuclei transcribing at every time point, we find that the embryos with mutations have  
255 far fewer active nuclei in each frame compared to the wildtype, despite their cumulative number  
256 of active nuclei being comparable (Figure 4 supplement 1A,B). We use stochastic modeling of  
257 transcription to investigate if TF binding site mutations cause changes in transcriptional bursting  
258 characteristics of each nucleus. The two-state model, in which a promoter can switch between an  
259 active (ON) and inactive (OFF) state, has been widely implemented to gain a functional  
260 understanding of bursting control on transcription (Figure 4B) (Bothma et al. 2014; Corrigan et al.  
261 2016).

262  
263 Hidden Markov models (HMMs) are utilized to reveal hidden states not directly observable based  
264 on a sequence of observed events (Bronson et al. 2009). They are extensively used to recover  
265 rates of promoter switching and other bursting parameters from transcriptional trajectories  
266 (Bothma et al. 2014; Lammers et al. 2020; Hoppe et al. 2020). In this study, we utilize an HMM to  
267 infer the promoter state based on the observed fluorescence intensity curves. The inferred  
268 promoter states of the mutant show slightly shorter burst durations compared to its wildtype  
269 counterpart (Figure 4C,D, Figure 4 supplement 1C,D). Based on previous studies, burst  
270 separation and burst duration can be correlated to bursting parameters,  $k_{on}$  and  $k_{off}$  (Hoppe et al.  
271 2020; Zoller, Little, and Gregor 2018). Using our modeling approach, we extracted the kinetic  
272 rates of the promoter returning to an active state from an inactive state ( $k_{on}$ ) and vice versa ( $k_{off}$ ).



**Figure 4. A two-state model reveals differences in  $k_{off}$  rates and burst duration**

(A) Actively transcribing nuclei are false-colored for early and late NC14. Mutant embryos show more sporadic transcriptional activity in a given frame.

(B) Schematic depicting a single nucleus that can be in an OFF or ON state and switches between the two with rates  $k_{on}$  and  $k_{off}$ . The rates of  $k_{on}$  and  $k_{off}$  can be correlated to burst separation and burst duration, respectively.

(C-D) Representative transcriptional trajectory of (C) wildtype and (D) D12 mutant with the inferred promoter states derived from the Hidden Markov model (HMM).

(E-F) Plots showing the rates of (E)  $k_{on}$  and (F)  $k_{off}$ .  $k_{on}$  is not significantly affected, whereas  $k_{off}$  rates are higher in the mutant embryos. Error bars indicate standard deviation (SD).

(G-H) Probability distributions of (G)  $k_{on}$  rates and (H)  $k_{off}$  rates for wildtype, D11, D12, and D11/2 mutants. The distribution of  $k_{on}$  rates follow a tight normal distribution while the  $k_{off}$  distributions vary widely.

Error bars in (E-F) indicate standard deviation (SD).

1124 nuclei from 3 replicate wildtype embryos, 1011 nuclei from 4 replicate D11 mutant embryos, 1123 nuclei from four replicate D12 mutant embryos, and 943 nuclei from three replicate D11/2 mutant embryos were analyzed.

Transcription of the *MS2-yellow* reporter gene is driven by the minimal *sna* distal enhancer.

\*\* denote  $p < 0.001$  from the student's t-test.

273 Our results reveal that  $k_{on}$  is only slightly affected by the induced mutations, while  $k_{off}$  is  
274 significantly increased in mutants (Figure 4E,F, Figure 4 supplement 1E,F). Moreover, all of the  
275 perturbations do not change the normal distribution trend of  $k_{on}$  rates (Figure 4G, Figure 4  
276 supplement 1H). However, we observe heterogeneity and high variability in the distribution of  $k_{off}$   
277 rates as well as shorter burst durations in the mutated embryos, explaining the bursty and noisy  
278 transcriptional activity we observed (Figure 4G,H, Figure 4 supplement 1G,I). This leads us to  
279 conclude that the mutations affect the ability of the promoter to remain in the active state, causing  
280 it to become more unstable and more likely to revert to the inactive state, supporting the  
281 observation of lower mRNA production. In the full distal enhancer, we observe similar stochastic  
282 transcriptional activation, where the number of actively transcribing nuclei is about 50% less in  
283 the mutants than in the wildtype at any given time (Figure 4 supplement 2A). We also see a similar  
284 increase in the rate of  $k_{off}$  in the full distal enhancer, with minimal effect on  $k_{on}$  (Figure 4 supplement  
285 2D-F). The trend in the distributions of the promoter switching rates remains the same as well  
286 (Figure 4 supplement 2B-C). Taken together, our data demonstrate that TF-DNA binding  
287 modulates mRNA production by increasing the rate of promoter inactivation ( $k_{off}$ ).

288

## 289 Discussion

290 Organisms have evolved to contain enhancers with multiple binding sites for the same TF, not  
291 only for robustness under varying biological conditions, but as a molecular regulatory mechanism.  
292 Here, we studied transcriptional dynamics driven by different degrees of TF-DNA interactions  
293 through inducing mutations in the *sna* regulatory module. By utilizing quantitative live imaging, we  
294 dissected the effects of modulation of TF binding sites and determined that mRNA production of  
295 the target gene is severely reduced when a single binding site is mutated. This reduction is mainly  
296 due to the decreased transcriptional amplitude (i.e., Pol II loading rate) but is also slightly affected  
297 by shorter duration of transcription, delayed transcriptional activation, and lower probability of  
298 activation. Although a previous study demonstrated that an extended enhancer with additional

299 binding sites could buffer against the effects of mutating a single TF binding site (López-Rivera et  
300 al. 2020), we find that the same effect in the minimal enhancer is also observed in the full distal  
301 enhancer. It is interesting to note that despite the presence of additional TF binding sites,  
302 modulating a single site can catastrophically reduce gene expression levels. We confirm that the  
303 minimal enhancer is the core region that is needed to recapitulate normal *sna* levels and that TFs  
304 in that region employ distinct mechanisms to regulate transcription in response to genetic  
305 perturbations.

306

307 Moreover, we determined that the TF binding sites in the *sna* enhancer work synergistically to  
308 drive the proper pattern and levels of the target gene expression. Cooperativity is a necessary  
309 mechanism by which a hub of weaker sites can coordinate to synergistically generate the correct  
310 expression pattern and levels in a developing embryo. In the context of *sna*, the *sna* enhancer  
311 only contains weak DI binding sites, since high affinity DI sites would cause *sna* to be expressed  
312 in nuclei with intermediate DI levels, repressing *short gastrulation (sog)* expression in the region  
313 and preventing those cells from developing into neuroectoderm (Hong et al. 2008). Hence, it is  
314 crucial that the *sna* enhancer contains only weak DI binding sites to drive expression exclusively  
315 in the region with high DI concentration (i.e., mesoderm). Indeed, previous works have shown that  
316 other enhancers also utilize weak, sub-optimal binding sites to drive specific target gene  
317 expression. It was determined that despite the *shavenbaby* enhancers containing low-affinity  
318 binding sites, a microenvironment of high concentrations of Ubx and other cofactors can mediate  
319 efficient and specific transcription of the target locus (Tsai et al. 2017). Moreover, a single  
320 optimized DI binding site in the *t48* enhancer resulted in earlier activity and ectopic expression  
321 patterns (Keller et al. 2020). Suboptimization of enhancers was shown to be an important  
322 characteristic of gene regulation to drive restricted expression, whereas optimizing TF motifs  
323 resulted in the loss of specificity and an increase in aberrant transcriptional activity (Farley et al.

2015). We propose that the *sna* enhancer utilizes synergy among multiple weak binding sites to drive stable and robust expression of *sna* in its narrow expression domain.

Interestingly, we found that the TF binding site mutations affect the rate of the promoter switching OFF. The heterogeneity in  $k_{\text{off}}$  rates confirms that the mutations abolish the ability of the embryo to develop robustly by causing promoter instability and shorter transcription windows, leading to overall lower mRNA production. As a result of higher  $k_{\text{off}}$  rates, the mutant embryos have fewer actively transcribing nuclei within the domain at any given time point (Figure 4A, Figure 4 supplement 1A, Figure 4 supplement 2A, Video 3,4). Furthermore, the shallower gradient of the expression pattern highlights the role of TFs in controlling the sharpness of the expression gradient, potentially affecting the robustness in germ layer formation (Figure 1H,K). In sum, we believe that multiple TF binding sites are imperative not only for pattern formation but also for eliminating extrinsic and intrinsic variabilities that may occur during development.

We illustrate that the minimal enhancer is an adequate model system to study transcription kinetics. We note that although it may not always be possible to generalize the effects of mutations from a reporter construct to the endogenous setting, we can still gain valuable insights and broaden our understanding of transcription regulation. Since *sna* is responsible for mesoderm formation as well as for the repression of other patterning genes like *sog* that specify neuroectoderm, it will be interesting to characterize the phenotypic effects of these mutations. Specifically, if the mutation causes 65% reduction in *sna* activity endogenously, would the developing embryo undergo proper gastrulation and remain viable? However, since *sna* has at least two well-characterized enhancers (proximal and distal) that compensate for each other, investigating the phenotypic effects of endogenously modulating TF binding sites remains challenging. In order to correlate transcriptional dynamics with downstream development it will be critical to carefully design experiments that will disentangle the contributions of the individual enhancers and the role of specific TF binding sites.

350

351 Moreover, the *sna* enhancer contains weaker binding sites for DI and hence is activated by high  
352 concentrations of DI. For this reason, it will be of interest to note how the induced mutations affect  
353 *sna* activity in the presence of reduced maternally deposited DI levels. While DI heterozygotes  
354 flies are viable and fertile, analyzing development under reduced DI level conditions may reveal  
355 effects on cell-to-cell variability in transcriptional activity. Lastly, the spatial arrangement of the TF  
356 binding may also play a role in ensuring normal development. Mutation of the DI1 site (Figure 1A)  
357 seems to cause earlier transcriptional activation, which may indicate a potential competing  
358 conformation due to the close proximity between the two DI sites (DI1 and DI2).

359

360 In this work, we demonstrate that TFs can regulate transcriptional dynamics by tuning bursting  
361 parameters and modulating transcriptional activity in response to genetic perturbations. Using our  
362 quantitative live imaging platform, we find that low-affinity TF binding sites can create an  
363 environment of increased transcriptional activity to drive localized, specific, and sharp expression  
364 patterns. The evidence of a dual modality of regulation and synergy highlights the importance of  
365 moving beyond fixed tissue studies and focusing on experiments that can tease apart subtle  
366 kinetic changes that occur during development. Collectively, our findings provide novel insights  
367 into enhancer-mediated transcriptional dynamics and expand our understanding of enhancer-TF  
368 binding through a combination of experimental and modeling approaches.

369

## 370 **Materials and methods**

### 371 **Motif scanning**

372 TF binding sites were found through the FIMO (Find Individual Motif Occurrences) (Bailey et al.  
373 2015) tool using motifs from (Keller et al. 2020) and JASPAR (Castro-Mondragon et al. 2022).  
374 The cut-off p-value for motif match was set to  $p < 1e-3$ . Mutations of the TF binding sites were

375 confirmed by scanning the mutated sequence through FIMO and ensuring that it was no longer  
376 recognizable as a motif. The wildtype and mutated sequences are shown in Table 1.

377

### 378 **Plasmid and transgenic fly generation**

379 The minimal distal *sna* enhancer was characterized in (Ferraro et al. 2016). TF binding sites were  
380 mutated using PCR-mediated site-directed mutagenesis and confirmed via Sanger sequencing  
381 (Table 1). The mutated enhancers were cloned into a plasmid containing the core 100 bp *sna*  
382 promoter, 10 copies of MS2 stem loops, and the *yellow* reporter gene. Transgenic reporter lines  
383 were created using PhiC31-mediated integration and the transgene was inserted to the VK33  
384 locus (Venken et al. 2006). Injection was performed by the BestGene, Inc.

385

### 386 **Live imaging**

387 Wild-type embryos were produced by crossing *yw;His2Av-mRFP,nanos>MCP-GFP* (Fukaya,  
388 Lim, and Levine 2016) virgin females to the desired *y,w;MS2* males. The embryos from the cross  
389 were laid at 25°C, dechorionated, and staged with Halocarbon oil. All images were taken using a  
390 Zeiss LSM800 confocal laser scanning microscope. Images were acquired with a Plan-  
391 Apochromat 40x1.3 NA oil objective using a 488 nm and 561 nm laser to visualize MCP:GFP and  
392 His2Av-mRFP, respectively, with a time resolution of 21s/frame. Images were created using  
393 maximum projection of 14 z-stacks with 0.75 µm steps. The same exposure and laser settings  
394 were used for all minimal *sna* replicates and a different set of settings were used for all the full  
395 distal *sna* replicates. All images were acquired in 16-bits. Images were taken as the embryo  
396 entered the 14<sup>th</sup> nuclear cycle until the embryo began gastrulation.

397

### 398 **Quantification and Statistical Analysis**

399 All the image processing methods and analyses were implemented in MATLAB

400 (R2018b, MathWorks). Histograms of all the snapshots and movies shown in all figures were  
401 adjusted for visualization purposes only. Analysis of all data was performed using raw images. To  
402 determine statistical significance, two sample t-tests were performed. \*\* indicates  $p < 0.001$ .

403

#### 404 **Image analysis**

405 Segmentation, nuclei tracking, and MS2 signal extraction were performed as described in (Syed  
406 et al. 2021). Actively transcribing nuclei were labeled if they exceeded a predetermined  
407 fluorescence intensity threshold. mRNA production was calculated by integrating under the  
408 fluorescence intensity trajectories of actively transcribing nuclei. Activation time was defined as  
409 the time at which the MS2 signal increased beyond the threshold. Transcription duration was  
410 calculated to be the time a nucleus was above the given threshold. The mean transcriptional  
411 amplitude was determined by the averaging of the MS2 signal for all transcriptionally active nuclei.  
412 The average transcriptional trajectory was obtained by averaging the intensity of all active nuclei  
413 at each timepoint. The sharpness of the expression gradient was determined by finding the  
414 maximum derivative of the mRNA production curves. Spatial analysis was performed by dividing  
415 the embryo into 16 bins along the dorsoventral axis. All the nuclei data within each bin was  
416 averaged to obtain the plots.

417

#### 418 **Equilibrium binding model**

419 The thermodynamic model utilized in this study is built on those described in (Eck et al. 2020;  
420 Kanodia et al. 2012). Concentrations of nuclear DI were assumed to follow a normal distribution.  
421  $T_{wi}$  concentration was calculated from (Lim et al. 2015) and normalized. The dissociation  
422 constants were chosen to reflect the relative p-values obtained from the FIMO analysis, correlated  
423 to the affinity of each binding site. Assuming that the microstates are in equilibrium, the probability  
424 of transcription occurring is given by Equation 1:

425  $p_{bound} =$

426

427 
$$\frac{\frac{C_A}{K_A} + \frac{C_B}{K_B} + \frac{C_C}{K_C} + \frac{C_D}{K_D} + \omega_1 \frac{C_A C_B}{K_A K_B} + \omega_2 \frac{C_B C_C}{K_B K_C} + \omega_3 \frac{C_A C_C}{K_A K_C} + \omega_4 \frac{C_A C_D}{K_A K_D} + \omega_5 \frac{C_B C_D}{K_B K_D} + \omega_6 \frac{C_C C_D}{K_C K_D} + \omega_7 \frac{C_A C_B C_C}{K_A K_B K_C} + \omega_8 \frac{C_A C_B C_D}{K_A K_B K_D} + \omega_9 \frac{C_B C_C C_D}{K_B K_C K_D} + \omega_{10} \frac{C_A C_C C_D}{K_A K_C K_D} + \omega_{11} \frac{C_A C_B C_C C_D}{K_A K_B K_C K_D}}{1 + \frac{C_A}{K_A} + \frac{C_B}{K_B} + \frac{C_C}{K_C} + \frac{C_D}{K_D} + \omega_1 \frac{C_A C_B}{K_A K_B} + \omega_2 \frac{C_B C_C}{K_B K_C} + \omega_3 \frac{C_A C_C}{K_A K_C} + \omega_4 \frac{C_A C_D}{K_A K_D} + \omega_5 \frac{C_B C_D}{K_B K_D} + \omega_6 \frac{C_C C_D}{K_C K_D} + \omega_7 \frac{C_A C_B C_C}{K_A K_B K_C} + \omega_8 \frac{C_A C_B C_D}{K_A K_B K_D} + \omega_9 \frac{C_B C_C C_D}{K_B K_C K_D} + \omega_{10} \frac{C_A C_C C_D}{K_A K_C K_D} + \omega_{11} \frac{C_A C_B C_C C_D}{K_A K_B K_C K_D}}$$

428

429 where  $C_i$  is the concentration of a TF,  $K_i$  is the dissociation constant, and  $\omega_i$  is the cooperativity  
 430 factor. To find the minimum of the nonlinear multivariable functions in the previous equation, we  
 431 utilized a nonlinear programming solver, *fmincon*. The solver returns a vector of cooperativity  
 432 values that minimize the objective function. The objective function uses the root mean square  
 433 error between the wildtype and mutant conditions to determine the cooperativity terms that would  
 434 satisfy the constraints (i.e., mutant condition must have 65% reduction in expression level  
 435 compared to the wildtype). A stable solution was defined once the solution converged and the  
 436 solver returned cooperativities that satisfied the objective function and constraints within a step  
 437 tolerance of  $1e-10$ . Using these evaluated cooperativities, curves were generated to predict  
 438 mRNA production. In the case of no cooperativity (Figure 3B), all cooperativity values were set to  
 439 1 and the results were plotted.

440

441 **Two-state model**

442 Two-state model fitting is similar to that described in (Keller et al. 2020). Transcriptional  
 443 trajectories were smoothed using local regression (LOESS) method. Each trajectory from a  
 444 given nucleus was converted into a binary plot indicating promoter ON (1) and promoter OFF (0)  
 445 states as described below. The slope between two consecutive time-points of active transcription  
 446 was obtained to define promoter ON and OFF states. Time points with positive slope were  
 447 considered as ON promoter states (1), while those with a negative slope were assumed as OFF  
 448 promoter states (0). This binary data was used as the input for the Baum-Welch based HMM.  
 449 Initial transition probabilities were assumed to be equal (i.e., 0.5). These probabilities were

450 adjusted in each iteration to individual burst traces until they converged. The Viterbi algorithm was  
451 used to determine the most likely sequence of (ON/OFF) promoter states (shown in Figure 4C,D).  
452 Burst separation and burst duration were correlated to  $k_{on}$  and  $k_{off}$ , respectively.

453

#### 454 **Acknowledgements**

455 We thank members of the Lim lab for helpful discussions and comments on the manuscript. We  
456 especially thank Noel Buitrago for his help in imaging. D1-venus fly was kindly provided by the  
457 Shvartsman Lab at Princeton University. We also thank the FlyBase for providing useful  
458 information (Gramates et al. 2022). This study was funded by NIH R35GM133425 awarded to  
459 BL.

460

#### 461 **Competing interests**

462 The authors declare that there are no competing interests.

463

#### 464 **References**

- 465 Arias, Alfonso Martinez, and Penelope Hayward. 2006. “Filtering Transcriptional Noise during  
466 Development: Concepts and Mechanisms.” *Nature Reviews Genetics* 2006 7:1 7 (1): 34–44.  
467 <https://doi.org/10.1038/nrg1750>.
- 468 Bailey, Timothy L., James Johnson, Charles E. Grant, and William S. Noble. 2015. “The MEME  
469 Suite.” *Nucleic Acids Research* 43 (W1): W39–49. <https://doi.org/10.1093/NAR/GKV416>.
- 470 Banerji, Julian, Sandro Rusconi, and Walter Schaffner. 1981. “Expression of a Beta-Globin Gene  
471 Is Enhanced by Remote SV40 DNA Sequences.” *Cell* 27 (2 Pt 1): 299–308.  
472 [https://doi.org/10.1016/0092-8674\(81\)90413-X](https://doi.org/10.1016/0092-8674(81)90413-X).
- 473 Bothma, Jacques P., Hernan G. Garcia, Emilia Esposito, Gavin Schlissel, Thomas Gregor, and  
474 Michael Levine. 2014. “Dynamic Regulation of Eve Stripe 2 Expression Reveals

475 Transcriptional Bursts in Living Drosophila Embryos.” *Proceedings of the National*  
476 *Academy of Sciences of the United States of America* 111 (29): 10598–603.  
477 [https://doi.org/10.1073/PNAS.1410022111/SUPPL\\_FILE/PNAS.1410022111.SM05.MOV](https://doi.org/10.1073/PNAS.1410022111/SUPPL_FILE/PNAS.1410022111.SM05.MOV).

478 Bronson, Jonathan E., Jingyi Fei, Jake M. Hofman, Ruben L. Gonzalez, and Chris H. Wiggins.  
479 2009. “Learning Rates and States from Biophysical Time Series: A Bayesian Approach to  
480 Model Selection and Single-Molecule FRET Data.” *Biophysical Journal* 97 (12): 3196.  
481 <https://doi.org/10.1016/J.BPJ.2009.09.031>.

482 Castro-Mondragon, Jaime A., Rafael Riudavets-Puig, Ieva Rauluseviciute, Roza Berhanu  
483 Lemma, Laura Turchi, Romain Blanc-Mathieu, Jeremy Lucas, et al. 2022. “JASPAR 2022:  
484 The 9th Release of the Open-Access Database of Transcription Factor Binding Profiles.”  
485 *Nucleic Acids Research* 50 (D1): D165–73. <https://doi.org/10.1093/NAR/GKAB1113>.

486 Corrigan, Adam M., Edward Tunnacliffe, Danielle Cannon, and Jonathan R. Chubb. 2016. “A  
487 Continuum Model of Transcriptional Bursting.” *ELife* 5 (FEBRUARY2016).  
488 <https://doi.org/10.7554/ELIFE.13051>.

489 Donovan, Benjamin T, Anh Huynh, David A Ball, Heta P Patel, Michael G Poirier, Daniel R  
490 Larson, Matthew L Ferguson, and Tineke L Lenstra. 2019. “Live-Cell Imaging Reveals the  
491 Interplay between Transcription Factors, Nucleosomes, and Bursting.” *The EMBO Journal*  
492 38 (12): e100809. <https://doi.org/10.15252/EMBJ.2018100809>.

493 Dufourt, Jeremy, Antonio Trullo, Jennifer Hunter, Carola Fernandez, Jorge Lazaro, Matthieu  
494 Dejean, Lucas Morales, et al. 2018. “Temporal Control of Gene Expression by the Pioneer  
495 Factor Zelda through Transient Interactions in Hubs.” *Nature Communications* 2018 9:1 9  
496 (1): 1–13. <https://doi.org/10.1038/s41467-018-07613-z>.

497 Dunipace, Leslie, Anil Ozdemir, and Angelike Stathopoulos. 2011. “Complex Interactions

498 between Cis-Regulatory Modules in Native Conformation Are Critical for *Drosophila* Snail  
499 Expression.” *Development* 138 (18): 4075–84. <https://doi.org/10.1242/DEV.069146>.

500 Eck, Elizabeth, Jonathan Liu, Maryam Kazemzadeh-Atoufi, Sydney Ghoreishi, Shelby Blythe,  
501 and Hernan G. Garcia. 2020. “Quantitative Dissection of Transcription in Development  
502 Yields Evidence for Transcription Factor-Driven Chromatin Accessibility.” *ELife* 9  
503 (October): 1–99. <https://doi.org/10.7554/ELIFE.56429>.

504 Farley, Emma K., Katrina M. Olson, Wei Zhang, Alexander J. Brandt, Daniel S. Rokhsar, and  
505 Michael S. Levine. 2015. “Suboptimization of Developmental Enhancers.” *Science* 350  
506 (6258): 325–28.  
507 [https://doi.org/10.1126/SCIENCE.AAC6948/SUPPL\\_FILE/FARLEY.SM.PDF](https://doi.org/10.1126/SCIENCE.AAC6948/SUPPL_FILE/FARLEY.SM.PDF).

508 Ferraro, Teresa, Emilia Esposito, Laure Mancini, Sam Ng, Tanguy Lucas, Mathieu Coppey,  
509 Nathalie Dostatni, Aleksandra M. Walczak, Michael Levine, and Mounia Lagha. 2016.  
510 “Transcriptional Memory in the *Drosophila* Embryo.” *Current Biology* 26 (2): 212–18.  
511 <https://doi.org/10.1016/J.CUB.2015.11.058>.

512 Fukaya, Takashi, Bomyi Lim, and Michael Levine. 2016. “Enhancer Control of Transcriptional  
513 Bursting.” *Cell* 166 (2): 358. <https://doi.org/10.1016/J.CELL.2016.05.025>.

514 Gramates, L Sian, Julie Agapite, Helen Attrill, Brian R Calvi, Madeline A Crosby, Gilberto dos  
515 Santos, Joshua L Goodman, et al. 2022. “FlyBase: A Guided Tour of Highlighted Features.”  
516 *Genetics* 220 (4). <https://doi.org/10.1093/GENETICS/IYAC035>.

517 Hemavathy, Kirugaval, Xiaodi Hu, Shovon I. Ashraf, Stephen J. Small, and Y. Tony Ip. 2004.  
518 “The Repressor Function of Snail Is Required for *Drosophila* Gastrulation and Is Not  
519 Replaceable by Escargot or Worniu.” *Developmental Biology* 269 (2): 411–20.  
520 <https://doi.org/10.1016/j.ydbio.2004.01.029>.

- 521 Hong, Joung Woo, David A. Hendrix, Dmitri Papatsenko, and Michael S. Levine. 2008. “How  
522 the Dorsal Gradient Works: Insights from Postgenome Technologies.” *Proceedings of the*  
523 *National Academy of Sciences of the United States of America* 105 (51): 20072–76.  
524 [https://doi.org/10.1073/PNAS.0806476105/ASSET/23BB268E-511A-45FD-BD8E-](https://doi.org/10.1073/PNAS.0806476105/ASSET/23BB268E-511A-45FD-BD8E-6CC0CCED1BBE/ASSETS/GRAPHIC/ZPQ9990852040003.JPEG)  
525 [6CC0CCED1BBE/ASSETS/GRAPHIC/ZPQ9990852040003.JPEG](https://doi.org/10.1073/PNAS.0806476105/ASSET/23BB268E-511A-45FD-BD8E-6CC0CCED1BBE/ASSETS/GRAPHIC/ZPQ9990852040003.JPEG).
- 526 Hoppe, Caroline, Jonathan R. Bowles, Thomas G. Minchington, Catherine Sutcliffe, Priyanka  
527 Upadhyai, Magnus Rattray, and Hilary L. Ashe. 2020. “Modulation of the Promoter  
528 Activation Rate Dictates the Transcriptional Response to Graded BMP Signaling Levels in  
529 the *Drosophila* Embryo.” *Developmental Cell* 54 (6): 727-741.e7.  
530 <https://doi.org/10.1016/J.DEVCEL.2020.07.007>.
- 531 Houchmandzadeh, Bahram, Eric Wieschaus, and Stanislas Leibler. 2002. “Establishment of  
532 Developmental Precision and Proportions in the Early *Drosophila* Embryo.” *Nature* 2002  
533 *415:6873* 415 (6873): 798–802. <https://doi.org/10.1038/415798a>.
- 534 Ip, Y T, R E Park, D Kosman, K Yazdanbakhsh, and M Levine. 1992. “Dorsal-Twist Interactions  
535 Establish Snail Expression in the Presumptive Mesoderm of the *Drosophila* Embryo.”  
536 *Genes & Development* 6 (8). <https://doi.org/10.1101/gad.6.8.1518>.
- 537 Kanodia, Jitendra S., Hsiao Lan Liang, Yoosik Kim, Bomyi Lim, Mei Zhan, Hang Lu, Christine  
538 A. Rushlow, and Stanislav Y. Shvartsman. 2012. “Pattern Formation by Graded and  
539 Uniform Signals in the Early *Drosophila* Embryo.” *Biophysical Journal* 102 (3): 427–33.  
540 <https://doi.org/10.1016/J.BPJ.2011.12.042>.
- 541 Keller, Samuel H., Siddhartha G. Jena, Yuji Yamazaki, and Bomyi Lim. 2020. “Regulation of  
542 Spatiotemporal Limits of Developmental Gene Expression via Enhancer Grammar.”  
543 *Proceedings of the National Academy of Sciences of the United States of America* 117 (26):

544 15096–103.  
545 [https://doi.org/10.1073/PNAS.1917040117/SUPPL\\_FILE/PNAS.1917040117.SM04.MOV](https://doi.org/10.1073/PNAS.1917040117/SUPPL_FILE/PNAS.1917040117.SM04.MOV).  
546 Kosman, David, Y. Tony Ip, Michael Levine, and Kavita Arora. 1991. “Establishment of the  
547 Mesoderm-Neuroectoderm Boundary in the *Drosophila* Embryo.” *Science* 254 (5028): 118–  
548 22. <https://doi.org/10.1126/science.1925551>.  
549 Lammers, Nicholas C., Vahe Galstyan, Armando Reimer, Sean A. Medin, Chris H. Wiggins, and  
550 Hernan G. Garcia. 2020. “Multimodal Transcriptional Control of Pattern Formation in  
551 Embryonic Development.” *Proceedings of the National Academy of Sciences of the United*  
552 *States of America* 117 (2): 836–47.  
553 [https://doi.org/10.1073/PNAS.1912500117/SUPPL\\_FILE/PNAS.1912500117.SM07.AVI](https://doi.org/10.1073/PNAS.1912500117/SUPPL_FILE/PNAS.1912500117.SM07.AVI).  
554 Leptin, M. 1991. “Twist and Snail as Positive and Negative Regulators during *Drosophila*  
555 Mesoderm Development.” *Genes & Development* 5 (9): 1568–76.  
556 <https://doi.org/10.1101/GAD.5.9.1568>.  
557 Liang, Hsiao Lan, Chung Yi Nien, Hsiao Yun Liu, Mark M. Metzstein, Nikolai Kirov, and  
558 Christine Rushlow. 2008. “The Zinc-Finger Protein Zelda Is a Key Activator of the Early  
559 Zygotic Genome in *Drosophila*.” *Nature* 456 (7220): 400–403.  
560 <https://doi.org/10.1038/NATURE07388>.  
561 Lim, Bomyi, Carmeline J Dsilva, Thomas J Levario, Hang Lu, Trudi Schüpbach, Ioannis G  
562 Kevrekidis, and Stanislav Y Shvartsman. 2015. “Dynamics of Inductive ERK Signaling in  
563 the *Drosophila* Embryo.” <https://doi.org/10.1016/j.cub.2015.05.039>.  
564 Long, Hannah K., Sara L. Prescott, and Joanna Wysocka. 2016. “Ever-Changing Landscapes:  
565 Transcriptional Enhancers in Development and Evolution.” *Cell* 167 (5): 1170–87.  
566 <https://doi.org/10.1016/J.CELL.2016.09.018>.

- 567 López-Rivera, Francheska, Olivia K. Foster Rhoades, Ben J. Vincent, Edward C.G. Pym,  
568 Meghan D.J. Bragdon, Javier Estrada, Angela H. DePace, and Zeba Wunderlich. 2020. “A  
569 Mutation in the *Drosophila Melanogaster* Eve Stripe 2 Minimal Enhancer Is Buffered by  
570 Flanking Sequences.” *G3: Genes/Genomes/Genetics* 10 (12): 4473.  
571 <https://doi.org/10.1534/G3.120.401777>.
- 572 Mir, Mustafa, Armando Reimer, Jenna E. Haines, Xiao Yong Li, Michael Stadler, Hernan  
573 Garcia, Michael B. Eisen, and Xavier Darzacq. 2017. “Dense Bicoid Hubs Accentuate  
574 Binding along the Morphogen Gradient.” *Genes & Development* 31 (17): 1784–94.  
575 <https://doi.org/10.1101/GAD.305078.117>.
- 576 Monod, Jacques, Jeffries Wyman, and Jean Pierre Changeux. 1965. “On the Nature of Allosteric  
577 Transitions: A Plausible Model.” *Journal of Molecular Biology* 12 (1): 88–118.  
578 [https://doi.org/10.1016/S0022-2836\(65\)80285-6](https://doi.org/10.1016/S0022-2836(65)80285-6).
- 579 Park, Jeehae, Javier Estrada, Gemma Johnson, Ben J Vincent, Chiara Ricci-Tam, Meghan DJ  
580 Bragdon, Yekaterina Shulgina, et al. 2019. “Dissecting the Sharp Response of a Canonical  
581 Developmental Enhancer Reveals Multiple Sources of Cooperativity.” *ELife* 8 (June).  
582 <https://doi.org/10.7554/ELIFE.41266>.
- 583 Perry, Michael W., Alistair N. Boettiger, Jacques P. Bothma, and Michael Levine. 2010.  
584 “Shadow Enhancers Foster Robustness of *Drosophila* Gastrulation.” *Current Biology : CB*  
585 20 (17): 1562. <https://doi.org/10.1016/J.CUB.2010.07.043>.
- 586 Ramos, Andrea I., and Scott Barolo. 2013a. “Low-Affinity Transcription Factor Binding Sites  
587 Shape Morphogen Responses and Enhancer Evolution.” *Philosophical Transactions of the*  
588 *Royal Society B: Biological Sciences* 368 (1632). <https://doi.org/10.1098/RSTB.2013.0018>.  
589 ———. 2013b. “Low-Affinity Transcription Factor Binding Sites Shape Morphogen Responses

590 and Enhancer Evolution.” *Philosophical Transactions of the Royal Society B: Biological*  
591 *Sciences* 368 (1632). <https://doi.org/10.1098/rstb.2013.0018>.

592 Reiter, Franziska, Sebastian Wienerroither, and Alexander Stark. 2017. “Combinatorial Function  
593 of Transcription Factors and Cofactors.” *Current Opinion in Genetics & Development* 43  
594 (April): 73–81. <https://doi.org/10.1016/J.GDE.2016.12.007>.

595 Rembold, Martina, Lucia Ciglar, J. Omar Yáñez-Cuna, Robert P. Zinzen, Charles Girardot,  
596 Ankit Jain, Michael A. Welte, Alexander Stark, Maria Leptin, and Eileen E.M. Furlong.  
597 2014. “A Conserved Role for Snail as a Potentiator of Active Transcription.” *Genes &*  
598 *Development* 28 (2): 167. <https://doi.org/10.1101/GAD.230953.113>.

599 Rodriguez, Joseph, and Daniel R. Larson. 2020. “Transcription in Living Cells: Molecular  
600 Mechanisms of Bursting.” *Https://Doi.Org/10.1146/Annurev-Biochem-011520-105250* 89  
601 (June): 189–212. <https://doi.org/10.1146/ANNUREV-BIOCHEM-011520-105250>.

602 Senecal, Adrien, Brian Munsky, Florence Proux, Nathalie Ly, Floriane E. Braye, Christophe  
603 Zimmer, Florian Mueller, and Xavier Darzacq. 2014. “Transcription Factors Modulate C-  
604 Fos Transcriptional Bursts.” *Cell Reports* 8 (1): 75–83.  
605 <https://doi.org/10.1016/J.CELREP.2014.05.053>.

606 Spitz, François, and Eileen E.M. Furlong. 2012. “Transcription Factors: From Enhancer Binding  
607 to Developmental Control.” *Nature Reviews Genetics* 2012 13:9 13 (9): 613–26.  
608 <https://doi.org/10.1038/nrg3207>.

609 Syed, Sahla, Henry Wilky, João Raimundo, Bomyi Lim, and Amanda A. Amodeo. 2021. “The  
610 Nuclear to Cytoplasmic Ratio Directly Regulates Zygotic Transcription in *Drosophila*  
611 through Multiple Modalities.” *Proceedings of the National Academy of Sciences* 118 (14).  
612 <https://doi.org/10.1073/pnas.2010210118>.

613 Tsai, Albert, Anand K. Muthusamy, Mariana R.P. Alves, Luke D. Lavis, Robert H. Singer,  
614 David L. Stern, and Justin Crocker. 2017. “Nuclear Microenvironments Modulate  
615 Transcription from Low-Affinity Enhancers.” *ELife* 6 (November).  
616 <https://doi.org/10.7554/ELIFE.28975>.

617 Venken, Koen J.T., Yuchun He, Roger A. Hoskins, and Hugo J. Bellen. 2006. “P[Acman]: A  
618 BAC Transgenic Platform for Targeted Insertion of Large DNA Fragments in *D.*  
619 *Melanogaster*.” *Science (New York, N.Y.)* 314 (5806): 1747–51.  
620 <https://doi.org/10.1126/SCIENCE.1134426>.

621 Zeitlinger, Julia, Robert P. Zinzen, Alexander Stark, Manolis Kellis, Hailan Zhang, Richard A.  
622 Young, and Michael Levine. 2007. “Whole-Genome ChIP-Chip Analysis of Dorsal, Twist,  
623 and Snail Suggests Integration of Diverse Patterning Processes in the *Drosophila* Embryo.”  
624 *Genes & Development* 21 (4): 385–90. <https://doi.org/10.1101/GAD.1509607>.

625 Zoller, Benjamin, Shawn C. Little, and Thomas Gregor. 2018. “Diverse Spatial Expression  
626 Patterns Emerge from Unified Kinetics of Transcriptional Bursting.” *Cell* 175 (3): 835-  
627 847.e25. <https://doi.org/10.1016/J.CELL.2018.09.056>.  
628