# Canalization of Gene Expression in the Drosophila Blastoderm by Gap Gene Cross Regulation Supplementary Information 

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## S1 Gap gene circuits

A gene circuit [1-5] determines the time evolution of protein concentrations in the syncytial blastoderm of Drosophila melanogaster. The circuits used in this paper comprise the gap genes $h b$, $K r, g t$, and $k n i$ of the anteroposterior segmentation system. Their protein products are transcription factors that localize in the nuclei of the blastoderm [6-9]. Therefore, the state variables are the concentrations of the protein products of these genes inside nuclei. Anteroposterior (A-P) and dorsoventral ( $\mathrm{D}-\mathrm{V}$ ) patterning systems are largely independent of each other in the presumptive germ band of the blastoderm embryo. This allows us to consider a one-dimensional row of nuclei along the anteroposterior axis of the embryo, from $35 \%$ EL to $92 \%$ EL. The modeled region extends over $58 \%$ of the A-P axis, from the peak of the third $g t$ stripe to the posterior border of the posterior $h b$ domain (Fig. S4B).

The circuit functions according to three rules: interphase, mitosis and division. The first two rules describe the continuous dynamics of proteins during interphase and mitosis. During interphase the evolution of protein concentrations is determined by three processes: regulated protein synthesis, protein transport, and protein decay. During mitosis, transcription shuts down and nascent transcripts are destroyed [10]. Therefore, only protein transport and protein decay govern the dynamics in the mitosis rule.

The third rule, division, accounts for nuclear cleavages in the blastoderm. It models mitotic division as a discontinuous change in the state of the system. At the end of a mitosis, each nucleus is replaced with its daughter nuclei. The inter-nuclear distance is halved and the daughter nuclei inherit the state of the mother nucleus. The divisions are carried out according to a division schedule based on experimental data [11] (Fig. S2).

The gap gene circuits used in this study model events occurring during a period of time which begins at the onset of cleavage cycle 13 and ends at the onset of gastrulation at the end of cycle 14A [11]. $K r, g t$, and $k n i$ are exclusively zygotic, and their expression at the protein level is first detectable at early cycle 13 [12-17]. $h b$, which is expressed both maternally and zygotically, shows a large increase in expression in cycle 13 [18], indicating commencement of its zygotic expression. Time $t$ is measured in minutes from the start of cleavage cycle 13. The interphase of cycle 13 lasts for 16.0 min , and its mitosis from 16.0 to 21.1 min . At $t=21.1 \mathrm{~min}$, the thirteenth division is carried out by applying the division rule. The interphase of cycle 14 A starts immediately after division, and lasts until gastrulation at $t=71.1 \mathrm{~min}$.

## S1.1 Equations

The two continuous rules, interphase and mitosis, use a system of ordinary differential equations (ODEs) to describe the dynamics of protein concentrations. Let there be $M$ nuclei in the modeled region during a particular cleavage cycle. Let $i$ denote a particular nucleus, counting from anterior to posterior. We denote a particular segmentation gene by $a \in 1, \ldots, N$, where $N$ genes are represented in the circuit. $v_{i}^{a}(t)$ is the protein concentration of gene $a$ in nucleus $i$. The time evolution of the state variables $v_{i}^{a}(t)$ is given by the solution of the system of $M \times N$ ODEs,

$$
\begin{align*}
\frac{d v_{i}^{a}}{d t}=R^{a} g & \left(\sum_{b=1}^{N} T^{a b} v_{i}^{b}+m^{a} v_{i}^{\mathrm{Bcd}}+\sum_{\beta=1}^{N_{e}} E^{a \beta} v_{i}^{\beta}(t)+h^{a}\right)  \tag{S1}\\
+ & D^{a}(n)\left[\left(v_{i-1}^{a}-v_{i}^{a}\right)+\left(v_{i+1}^{a}-v_{i}^{a}\right)\right]-\lambda^{a} v_{i}^{a}
\end{align*}
$$

The first term on the right hand side of Eq. (S1) represents protein synthesis, the second one protein transport through Fickian diffusion and the last term represents first-order protein degradation.

The protein synthesis rate for gene $a$ is determined by the maximum synthesis rate $R^{a}$ and the regulatory input to $a, u^{a} \equiv \sum_{b=1}^{N} T^{a b} v_{i}^{b}+m^{a} v_{i}^{\mathrm{Bcd}}+\sum_{\beta=1}^{N_{e}} E^{a \beta} v_{i}^{\beta}(t)+h^{a}$. The rate of protein synthesis is the product of $R^{a}$ and the regulation-expression function $g\left(u^{a}\right)=\frac{1}{2}\left[\left(u^{a} / \sqrt{\left(u^{a}\right)^{2}+1}\right)+1\right]$ (see Fig. S1).


Figure S1: The regulation-expression function $g(u)$. The dashed vertical lines are at values of total input $u$ at which synthesis rate is $10 \%$ or $90 \%$ of maximum.

The regulatory input $u^{a}$ in turn accounts for the transcriptional regulation of gene $a$ by transcription factors. Each term of $u^{a}$ corresponds to a distinct type of factor. The first term $\sum_{b=1}^{N} T^{a b} v_{i}^{b}$ represents the regulation of gene $a$ by the genes $b \in 1, \ldots, N$ of the circuit. The elements of the ge-
netic interconnection matrix $T, T^{a b}$, characterize the regulatory effect of protein $b$ on the synthesis of gene $a$. A positive value of $T^{a b}$ represents activation and a negative value represents repression.

Spatially and temporally homogeneous maternal factors are represented in $u^{a}$ via its fourth term, $h^{a}$. The second term of $u^{a}$ represents the action of Bcd , which has a concentration that is spatially inhomogeneous, but constant in time [16]. $\mathrm{m}^{a}$ is the strength of the regulation of gene $a$ by Bcd, while $v_{i}^{\mathrm{Bcd}}$ is the concentration of Bcd in nucleus $i$.

Finally, we represent the effects of transcription factors which vary over space and time and regulate gap gene expression but are not themselves regulated by the gap genes through the third term of $u^{a}, \sum_{\beta=1}^{N_{e}} E^{a \beta} v_{i}^{\beta}(t) . N_{e}$ is the number of such factors in the circuit, $E^{a \beta}$ is the regulatory effect of the time varying external input $\beta$ on gene $a$, and $v_{i}^{\beta}(t)$ is the concentration of external input $\beta$ in nucleus $i$ at time $t$. The gene circuit used in this study has two such factors, $\operatorname{Cad}\left(v_{i}^{\mathrm{Cad}}(t)\right)$ and Tll $\left(v_{i}^{\mathrm{Tll}}(t)\right)$. cad is expressed from both the maternal and zygotic genomes. However, its expression pattern is not affected in mutants for $h b, K r, g t$ and $k n i$, except for a slight expansion of the posterior stripe in $h b^{-}$during late cycle 14 [19]. Similarly, tll expression is not affected in these mutants [20]. The concentrations of time varying external inputs Cad and Tll were determined by interpolation from data [16] (see Section S1.2 for details).

The second term of Eq. (S1) represents protein transport between nuclei as spatially discretized Fickian diffusion. The diffusion parameter $D^{a}$ is assumed to vary inversely with the squared distance between neighboring nuclei. The third term is protein degradation, in which $\lambda^{a}$ is the decay rate of the product of gene $a$. It is related to the protein half life of the product of gene $a$ by $t_{1 / 2}^{a}=\ln 2 / \lambda^{a}$.

Since Kr , Gt, and Kni proteins first appear only in cycle 13, they have initial conditions of zero, that is $v_{i}^{K r}(0)=v_{i}^{g t}(0)=v_{i}^{k n i}(0)=0$. For $h b$, the expression data from cycle 12 is used as the initial condition. These data are the maternal component of $h b$ since its expression intensifies only in cycle 13.

## S1.2 Numerical implementation of time varying external inputs

In order to specify the right hand side of Eq. (S1) fully, the concentrations of the time varying external inputs, $\mathrm{Cad}\left(v_{i}^{\mathrm{Cad}}(t)\right)$ and $\mathrm{Tll}\left(v_{i}^{\mathrm{Tll}}(t)\right)$, must be supplied for any time in the duration of the model. Average concentrations for these genes are known at ten time points $t_{k}, k=-1,0, \ldots, 8$. $t_{-1}=0 \mathrm{~min}, t_{0}$ is the midpoint of cycle 13 , and $t_{1}, \ldots, t_{8}$ correspond to the eight time classes $\mathrm{T} 1-\mathrm{T} 8$ in cycle 14 (Table 1). For nucleus $i$ and external input $\beta$, let $v_{i}^{\beta}\left(t_{k}\right)=v_{i, k}^{\beta}$, where $k=$ $-1,0, \ldots, 8$. The concentration of external input $\beta$ in nucleus $i$ is then determined at an arbitrary time $t$ by piecewise linear interpolation,


Figure S2: Division schedule and time classes. $t=0$ is at the start of cleavage cycle 13. The model runs until gastrulation at $t=71.1 \mathrm{~min}$. The duration of the cleavage cycles and mitosis was determined from experimental data [11]. The dashed lines demarcate time intervals in which different rules of the model apply. The intervals are labeled with the rules on the right. The nine time points at which model is compared to data are indicated. There is one time point for cycle 13 (C13), and eight points for cycle 14A (time classes T1-T8). Figure reproduced here with the permission of J. Jaeger [4].


Figure S3: Interpolation of time varying external inputs from data. (A) Cad data and linearly interpolated time profile at $50 \%$ EL. (B) Tll data and linearly interpolated time profile at $92 \%$ EL. Square boxes indicate data points. Cad concentration at $t=0 \mathrm{~min}$ is determined from cleavage cycle 12 data. Tll concentration is zero at 0 min since Tll is first detected in cleavage cycle 13 [16]. The other time points are midpoints of time classes (Table 1). Tll data were only available for time classes T3-T8. Lines are interpolated concentrations.

$$
v_{i}^{\beta}(t)=\frac{\left(t_{k+1}-t\right) v_{i, k}^{\beta}+\left(t-t_{k}\right) v_{i, k+1}^{\beta}}{t_{k+1}-t_{k}}, \quad t_{k} \leq t \leq t_{k+1} .
$$

Fig. S3 shows such interpolation at $50 \%$ EL for Cad, and at $92 \%$ EL for Tll. Higher order methods like cubic splines were not used because they gave rise to artifacts from experimental noise.

## S1.3 Optimization and selection procedure for gap gene circuits

The Parallel Lam Simulated Annealing algorithm [21, 22] produces many candidate circuits, and we selected a circuit using a three-step method [5]. First, only circuits with an RMS score less than 12.0 were considered. These circuits were screened further for patterning defects, and any circuit with major defects was discarded. Finally, experimental [23] and theoretical [24] investigations have shown that $K r$ represses $h b$, therefore, we only select circuits for further consideration if they show this property.

With this screening process, we obtained 23 circuits out of a total of 65 optimizations. This set of circuits have the same network topology (Table S3) as the circuits studied in earlier work [5, 24]. The signs of the regulatory parameters are the same in all circuits with three exceptions, $T^{g t \leftarrow k n i}$, $T^{K r \leftarrow g t}$, and $m^{k n i}$. Of these three, the last two change sign in only one circuit, k1_014 (Table S3). This level of consistency suggests that network topology has been inferred at a qualitative level, in agreement with the conclusions of a recent study of parameter determinability of gap gene circuits [25]. Bcd and Cad are activators of $h b, K r, g t$, and $k n i$. tll is an activator of $h b$, and a repressor of the other gap genes. The interaction between $h b, K r, g t$, and $k n i$ is one of mutual repression, with two exceptions: (1) $g t$ is an activator of $h b$ in all the circuits obtained. (2) kni is an activator of $g t$ in about half the circuits, and a repressor in the other half. This network topology is discussed in depth elsewhere [5, 24]. These results also hold true for circuits obtained using other Bcd profiles from the middle of the parameter scatter in Fig. 2A (black circles).

We chose one circuit ( $\mathrm{k} 1 \_007$ ), which has RMS score 10.76, for further analysis. Its parameters are shown in Tables S1 and S2. The chosen circuit's gap gene patterns (Fig. S4A,B) are consistent with data except for two minor defects. The first one is a bulge on the anterior border of the posterior $h b$ domain. The second is that the posterior border of the posterior $h b$ domain does not form fully. Some circuits reported in previous work [5] also suffer from these defects. The first defect is due to very low levels of spurious $t l l$ expression in the middle part of the embryo stemming from imperfect background removal. The second is due to the omission of the gene $h k b[20,26,27]$ from the model. $h k b$ is a terminal gap gene expressed near the posterior pole.


Figure S4: Gap gene expression patterns in the gene circuit chosen for analysis. Cleavage cycle 13 (A) and time class T8 (B) model output. The arrow shows the main patterning defect. (C,D) Averaged gene expression data for cleavage cycle C 13 (C) and time class T8 (D).

|  | Regulator gene $b$ |  |  |  |  |  |  |  |
| :--- | ---: | ---: | ---: | ---: | ---: | ---: | ---: | :---: |
| Target gene $a$ | $b c d$ | $c a d$ | $t l l$ | $h b$ | $K r$ | $g t$ | $k n i$ |  |
| $h b$ | 0.025 | 0.004 | 0.003 | 0.021 | -0.001 | 0.022 | -0.112 |  |
| $K r$ | 0.118 | 0.021 | -0.203 | -0.026 | 0.035 | -0.042 | -0.062 |  |
| $g t$ | 0.256 | 0.023 | -0.011 | -0.028 | -0.202 | 0.007 | 0.003 |  |
| $k n i$ | 0.012 | 0.020 | -0.187 | -0.082 | 0.000 | -0.017 | 0.013 |  |

Table S1: Regulatory parameters of the gap gene circuit chosen for analysis. The second, third, and fourth columns show $m^{a}, E^{a \leftarrow c a d}$, and $E^{a \leftarrow t l l}$ respectively. The columns of the $4 \times 4$ matrix $T^{a b}$ are shown in the last four columns of the table. The parameters are explained in Section S1.

|  | Gene $a$ |  |  |  |
| :--- | ---: | ---: | ---: | ---: |
| Parameter | $h b$ | $K r$ | $g t$ | $k n i$ |
| $R^{a}$ | 15.000 | 10.354 | 15.000 | 15.000 |
| $D^{a}$ | 0.166 | 0.200 | 0.103 | 0.200 |
| $t_{1 / 2}^{a}$ | 9.529 | 15.908 | 9.438 | 13.062 |

Table S2: Kinetic parameters of the gap gene circuit chosen for analysis. $h^{a}$ was kept fixed at -2.5 during optimization. The parameters are explained in Section S1.


Table S3: Equation parameters for the 23 gap gene circuits that have consistent network topology. $h^{h b}$, $h^{K r}, h^{g t}$, and $h^{k n i}$ were fixed to -2.5 . The circuit analyzed in this study, $\mathrm{k} 1 \_007$, is shown in the first column. See text for parameter definitions.

## S2 The simulation of Bcd variation

| Boundary | $\mathrm{k} 1 \_007$ |  | $\mathrm{k} 12 \_004$ |  | $\mathrm{k} 13 \_015$ |  | $\mathrm{k} 14 \_015$ |  | $\mathrm{k} 15 \_004$ |  |
| :--- | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
|  | raw | fit | raw | fit | raw | fit | raw | fit | raw |  |
| $h b$ posterior | 1.45 | 1.27 | 1.42 | 0.75 | 0.91 | 0.94 | 1.07 | 0.98 | 1.16 |  |
| $K r$ posterior | 0.78 | 0.58 | 0.74 | 0.51 | 0.58 | 0.41 | 0.48 | 0.59 | 0.61 |  |
| $k n i$ anterior | 0.96 | 0.62 | 0.89 | 1.00 | 0.99 | 0.38 | 0.68 | 1.13 | 0.99 |  |
| $k n i$ posterior | 1.05 | 1.17 | 0.86 | 0.89 | 0.86 | 0.90 | 1.03 | 1.28 | 1.15 |  |
| $g t$ anterior | 1.11 | 0.95 | 0.99 | 1.04 | 1.39 | 1.30 | 1.49 | 1.37 | 1.44 |  |
| $g t$ posterior | 1.56 | 1.30 | 1.51 | 0.57 | 1.59 | 1.44 | 1.93 | 1.08 | 1.83 |  |

Table S4: Positional variation of gap gene borders in circuits produced with Bcd gradients other than the median. Numbers are standard deviations in \% EL. The top row shows the circuits; $\mathrm{k} 1 \_007$ is the circuit produced with the median Bcd profile (Fig. 2B) and analyzed in this study. The other circuits were fit using the Bcd profiles highlighted in Fig. 2A with black circles. The first column shows the six borders that have low positional variation in the simulations of Bcd variation. Bcd variation was simulated in each circuit with two families of Bcd profiles. The first family consisted of raw, background-removed Bcd profiles (even-numbered columns). The second family consisted of the exponential fits of the profiles of the first family made using Eq. (2) (odd-numbered columns). The positional variation of the borders in simulations of circuit k1_007 using the second family of Bcd profiles is given in Table 2.

## S3 Regulatory analysis of gap gene borders

This section describes the methodology of the regulatory analysis presented in Section 2.3. At a border of the expression domain of gene $a$, the regulation-expression function $g\left(u^{a}\right)$ (see Eq. S1) changes from a value close to zero to a value close to one. In the regulatory analysis we determine the set of regulators of the gene $a$ that are responsible for this change in the value of $g\left(u^{a}\right)$. Here, $u^{a}=\sum_{b=1}^{N} T^{a b} v_{i}^{b}+m^{a} v_{i}^{\mathrm{Bcd}}+\sum_{\beta=1}^{N_{e}} E^{a \beta} v_{i}^{\beta}(t)+h^{a}$ is the total regulatory input to the gene $a$. Let $i_{B}$ be the indices of the nuclei over which the border of gene $a$ forms. Since the sigmoid $g\left(u^{a}\right)$ is approximately linear between $10 \%$ and $90 \%$ expression levels (Fig. S1), we can write $g\left(u_{i_{B}}^{a}\right) \approx u_{i_{B}}^{a}=\sum_{b=1}^{N} T^{a b} v_{i_{B}}^{b}+m^{a} v_{i_{B}}^{\mathrm{Bcd}}+\sum_{\beta=1}^{N_{e}} E^{a \beta} v_{i_{B}}^{\beta}+h^{a}$ for the purposes of this analysis.

Consider the posterior border of the expression domain of a gap gene $a$. Let the expression level of protein $a$ be at $90 \%$ of maximum at $i=i_{1}$, and at $10 \%$ maximum at $i=i_{M}$. The analysis is similar for anterior borders, but with $i=i_{1}$ at $10 \%$ maximum, and $i=i_{M}$ at $90 \%$ maximum. The total change in $u^{a}$ is $u_{i_{1}}^{a}-u_{i_{M}}^{a}$. Since this change is just a sum of regulatory contributions, we can divide it into individual regulatory terms

$$
\begin{equation*}
u_{i_{1}}^{a}-u_{i_{M}}^{a}=\sum_{b=1}^{N} T^{a b}\left(v_{i_{1}}^{b}-v_{i_{M}}^{b}\right)+m^{a}\left(v_{i_{1}}^{\mathrm{Bcd}}-v_{i_{M}}^{\mathrm{Bcd}}\right)+\sum_{\beta=1}^{N_{e}} E^{a \beta}\left(v_{i_{1}}^{\beta}-v_{i_{M}}^{\beta}\right) . \tag{S2}
\end{equation*}
$$

By comparing the magnitude of the change in different regulatory terms $\left(T^{a b}\left(v_{i_{1}}^{b}-v_{i_{M}}^{b}\right), m^{a}\left(v_{i_{1}}^{\mathrm{Bcd}}-\right.\right.$ $\left.v_{i_{M}}^{\mathrm{Bcd}}\right), E^{a \beta}\left(v_{i_{1}}^{\beta}-v_{i_{M}}^{\beta}\right)$, we can determine which regulators are driving the formation of the boundary. In Fig. 4A-C and 5A-C, the spatial derivatives of $u^{a}$ and individual regulatory terms are plotted so that the total change in $u^{a}$ is the area below or above the curve, and the total change in each regulatory term is the area between curves.

It is possible to simplify this analysis by eliminating many regulators that cannot set the border. Autoregulation, for instance, cannot form the border [5]. It can only make a border sharper. Since this is a posterior border, the concentration of $a$ reduces as one goes from anterior $\left(i_{1}\right)$ to posterior ( $i_{M}$ ). An activator whose concentration is increasing from $i_{1}$ to $i_{M}$ will tend to counteract the reduction in $a$ 's concentration. Thus an activator whose gradient falls in the opposite direction of the border cannot aid its formation. Similarly a repressor whose concentration is decreasing from $i_{1}$ to $i_{M}$, that is, a repressor gradient in the same direction as the border, cannot cause the border to form.

Such simplification is also possible for anterior borders (increasing from $i_{1}$ to $i_{M}$ ). Activator gradients in the opposite direction to the border, and repressor gradients in the same direction as the anterior border cannot cause it to form.

Here we list the regulatory inputs that were eliminated in the regulatory analysis shown in Fig. 4A-C and Fig. 5A-C following the procedure described above. For the posterior border of the anterior $h b$ domain (Fig. 4A), Hb autoactivation and Cad activation were eliminated. Of these two, Cad activation has a negligible contribution and hence only Hb activation is shown in red. In the analysis of the regulation of the posterior border of the central Kr domain (Fig. 4B), Kr autoactivation, Cad activation, Tll repression, and Hb repression were eliminated and their combined contribution is shown in red. For the posterior border of the posterior kni domain (Fig. 4C), Kni autoactivation, Kr repression, and Cad activation cannot set the border and their contribution is shown in red.

In the analysis of the regulation of the anterior border of the posterior kni domain (Fig. 5A), Kni autoactivation, Bcd activation, and Tll repression were eliminated and are shown in red. For the anterior border of the posterior $g t$ domain (Fig. 5B), Gt autoactivation, Bcd activation, Kni activation, and Tll repression were eliminated and are shown in red. For the posterior border of the posterior $g t$ domain (Fig. 5C), Gt autoactivation, Cad activation, and Kr repression cannot set the border and are shown in red.

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