# Capabilities and Limitations of Tissue Size Control Through Passive Mechanical Forces

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#### Abstract

Embryogenesis is an extraordinarily robust process, exhibiting the ability to control tissue size and repair patterning defects in the face of environmental and genetic perturbations. The size and shape of a developing tissue is a function of the number and size of its constituent cells as well as their geometric packing. How these cellular properties are coordinated at the tissue level to ensure developmental robustness remains a mystery; understanding this process requires studying multiple concurrent processes that make up morphogenesis, including the spatial patterning of cell fates and apoptosis, as well as cell intercalations. In this work, we develop a computational model that aims to understand aspects of the robust pattern repair mechanisms of the Drosophila embryonic epidermal tissues. Size control in this system has previously been shown to rely on the regulation of apoptosis rather than proliferation; however, to date little work has been done to understand the role of cellular mechanics in this process. We employ a vertex model of an embryonic segment to test hypotheses about the emergence of this size control. Comparing the model to previously published data across wild type and genetic perturbations, we show that passive mechanical forces suffice to explain the observed size control in the posterior (P) compartment of a segment. However, observed asymmetries in cell death frequencies across the segment are demonstrated to require patterning of cell mechanical properties in the model. Finally, we show that distinct forms of mechanical regulation in the model give rise to distinct cell shapes in the P compartment, as quantified through experimentally accessible summary statistics.

## Introduction

The mechanisms underlying tissue size control during embryonic development are extremely robust. There are many cases where the rates of proliferation, growth, or death are perturbed significantly yet patterns are maintained or repaired during later stages of development. For example, even after 80% of the material in a mouse embryo is removed, accelerated growth can give rise to correctly proportioned, albeit not viable, offspring [1]. In fruit fly embryos, overexpressing the maternal effect gene *bicoid* leads to stark overgrowth in the head region, but the excess tissue is removed during later stages of development through apoptosis (programmed cell death), leading to viable adults [2]. Tetraploid salamanders of the species *Amblystoma mexicanum* have half the number of cells as their diploid counterparts, yet are the same size [3].

The robustness of tissue size control relies on tight coordination of cellular processes including growth, proliferation, apoptosis and movement at a tissue level. However, the fundamental mechanisms underlying such coordination remain largely unknown. In particular, the role of mechanical regulation in tissue size control is not well understood. The regulation of cellular mechanical properties is known to play a key role during morphogenetic events, such as tissue folding, elongation and cell sorting [4,5]. For example, upregulation of myosin II generates tension that helps to straighten compartment boundaries in the *Drosophila* wing imaginal disc [6], while controlled cell death provides the tension required for invagination during *Drosophila* leg development [7]. It has been illustrated theoretically how mechanical feedback might facilitate uniform growth in epithelia in the face of morphogen gradients [8]. Could mechanical forces also play a significant role in robustly maintaining tissue size?

To explore questions of pattern repair, we develop a computational model of a patterned epithelium, with application to the segments of the *Drosophila* embryonic epidermis (Fig. 1C). These tissues define the body plan along the head-tail axis. They are first defined during stage 6 of embryonic development and are visible as stripes in the epidermis of the larva [9]. The segments are subdivided into anterior (A) and posterior (P) compartments, which are marked by distinct gene expression patterns. In particular, cells in the P compartment express the gene engrailed [10] (Fig. 1D). While the initial specification and establishment of segments is relatively well studied [11], maintenance of segment identities have received much less attention. However, it is known that compartment dimensions can be robustly restored in the presence of genetic manipulations made during earlier developmental patterning events [2,12–14]. Both the conserved epidermal growth factor receptor (EGFR) and Wnt/Wingless (WG) pathways have been implicated in regulating apoptosis to achieve pattern repair for perturbations made in each of the compartments and are known to antagonize each other [2,14].

A major strength of *Drosophila* as a model organism is the availability of genetic tools

that enable the ectopic expression of gene products or RNA interference (RNAi) constructs to manipulate cell growth, proliferation and signaling in a spatio-temporally controlled manner [15–17]. For example, the bipartite GAL4-UAS system can be used to drive expression of ectopic genes in embryos through a cross of one line containing a tissue-specific enhancer driving expression of the heterologous yeast transcription factor GAL4 with a second line that activates expression of a transgene upon binding of GAL4 to the UAS promoter region. Using this approach, Parker [14] investigated P compartment size using the GAL4 driver line as the control genotype engrailed-GAL4, UAS-GFP, in the following referred to as wt (wild type). This was compared to various perturbations (Fig. 1B). In particular, these included crosses between the driver line and UAS-CyclinE (which we shall term en > CycE) and UAS-dacapo lines (further specified as en > dap), which perturbed the amount of final proliferation events towards the end of the normal range of proliferation in the epidermis (Fig. 1A).

Parker [14] observed an increase in final cell number of more than 30% (Fig. 1E, right bar) in the P compartments of en>CycE embryos, which exhibited an additional round of cell division. However, the size of the P compartment was much less affected by this perturbation (Fig. 1F, right bar), as measured in first instar larvae [14]. Conversely, in en>dap embryos that exhibited a loss of one round of cell division, Parker [14] observed a reduction in cell number of 25% while, again, the compartment size was relatively unchanged (Fig. 1E and 1F, middle bars).

Parker's findings also suggest that epidermal growth factor receptor (EGFR) signalling, through the activating ligand Spitz, patterns apoptosis inside the P compartment. Spitz is released by a column of cells inside the anterior (A) compartment that is directly adjacent to the P compartment. Identifying cell death events through TUNEL staining [18], Parker [14] observed apoptosis much more frequently in the 'front' (more anterior) half of the P compartment, away from the Spitz source (Fig. 1G), than the 'back' half. These numbers differed by a factor of nearly 40 in wt [14]. Counter-intuitively, inhibiting apoptosis by expressing the caspase inhibitor p35 inside the P compartment of en>CycE embryonic segments resulted in compartment shrinkage by nearly 10%.

The above findings shed light on tissue size control in the Drosophila embryonic epidermal tissues, demonstrating its reliance on the regulation of apoptosis rather than proliferation. However, potential roles of cellular mechanics in repairing growth defects in patterned tissues remains poorly understood. To address this, we develop a vertex model of an embryonic segment to test hypotheses about the emergence of size control. Comparing the model to previously published data across wt and genetic perturbations, we investigate the extent to which passive mechanical forces might suffice to explain the observed size control and asymmetries in cell death frequencies across the P compartment. Our results suggest that the basis of size control relies to a significant degree on the passive mechanical responses of cells. However, the observed spatial

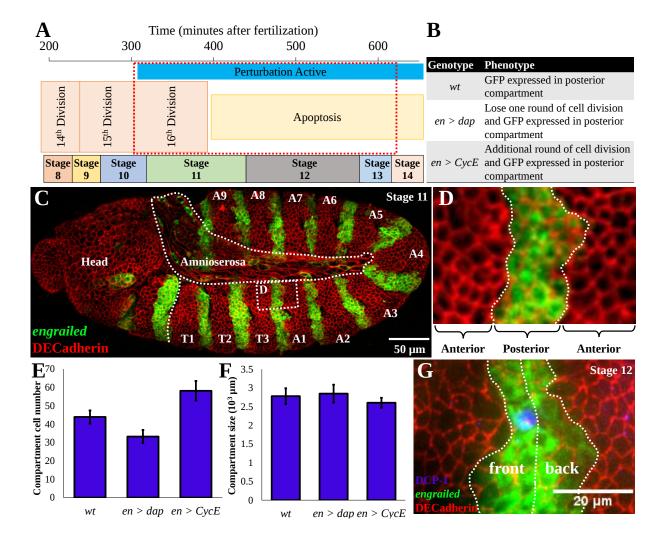


Figure 1. The *Drosophila* embryo as a model system for size homeostasis. (A) Specification of embryonic stages over time; the red boxed region represents the time period of simulations [19]. (B) Summary of genetic perturbations simulated in this study. The wt genotype is *engrailed>GAL4*, *UAS>GFP*. The perturbations are crosses between the *wt* and *UAS>CyclinE* and *UAS>dacapo* lines, respectively. (C) Stage 11 embryo expressing GFP in the posterior compartment, stained for DE-cadherin to show cell boundaries. (D) High magnification image of simulation domain. (E, F) Data extracted from [14] demonstrating that compartment dimensions are robust to manipulations that change the number of cells. (G) Cell death, indicated by cleaved *Drosophila* caspase-1 antibody staining [20], is statistically more likely to occur in the front half of the posterior compartment in *en>CycE* embryos [14].

asymmetry in cell death frequencies requires patterning of mechanical properties by intercellular communication. These results also provide a basis for identifying experimentally how extracellular signaling pathways like EGFR and WG might impact cellular decision making processes through predictions of observable cellular morphologies.

### Materials and Methods

We use a vertex model to simulate cell movement, intercalation, shape changes and apoptosis during the sixteenth round of divisions in a segment of the *Drosophila* embryonic epidermis. Vertex models were first introduced to study the structure of foams [21], and have since been applied to study a variety of epithelial tissues [6, 22–25]. For more information on vertex models and their application to epithelial morphogenesis, we refer the reader to two recent reviews [26, 27].

Equations of motion. Vertex models approximate cells in epithelial sheets as polygons. The polygons represent the cells' apical surfaces, where most inter-cellular forces originate [4]. The terms in the model account for the mechanical effect of the force-generating molecules that accumulate in the apical surface of the cells, such as actin, myosin, and E-cadherin. Vertices correspond to adherens junctions, and their positions are propagated over time using an over-damped force equation, reflecting that adherens junctions are not associated with a momentum. The force equation takes the form

$$\mu \frac{\mathrm{d}\mathbf{x}_i}{\mathrm{d}t} = -\nabla_i E. \tag{1}$$

Here,  $\mu$  is the friction strength (which we assume to take the same constant value for all vertices), t is time,  $\mathbf{x}_i$  is the position vector of vertex i, and E denotes the energy of the whole system. The total number of vertices in the system may change over time due to cell division and apoptosis. The symbol  $\nabla_i$  denotes the gradient operator with respect to the coordinates of vertex i. The forces act to minimise a phenomenological energy function, based on the contributions thought to dominate epithelial mechanics [22]:

$$E = \sum_{\alpha} \frac{K}{2} (A_{\alpha} - A_{0,\alpha})^2 + \sum_{\langle i,j \rangle} \Lambda l_{i,j} + \sum_{\alpha} \frac{\Gamma}{2} p_{\alpha}^2.$$
 (2)

Here, the first sum runs over every cell in the sheet,  $A_{\alpha}$  denotes the apical surface area of cell  $\alpha$  and  $A_{0,\alpha}$  is its preferred area, or target area. This energy term penalises deviations from a target area for individual cells, thus imposing cellular bulk elasticity. The second sum runs over all edges  $\langle i,j\rangle$  in the sheet and penalizes long edges (we choose  $\Lambda > 0$ ), thus representing the combined effect of E-cadherin, myosin, and actin at the binding interface between two cells. The third sum

also runs over all cells, and  $p_{\alpha}$  denotes the perimeter of cell  $\alpha$ . This term models the effect of a contractile acto-myosin cable along the perimeter of each cell [22]. The parameters K,  $\Lambda$ , and  $\Gamma$  together govern the strength of the individual energy contributions. Although this description of cell mechanics is phenomenological, a variety of previous studies have demonstrated its ability to match observed junctional movements and cell shapes in epithelial sheets through validation against experimental measurements [6, 22, 25].

In contrast to many previous vertex model applications, we allow the mechanical parameters  $\Lambda$ ,  $\Gamma$ , and  $A_0$  to vary between cells. In particular, we consider  $A_0$  to be a function of cell generation and introduce the parameter

$$R_A = A_{0,\text{daughter}}/A_{0,\text{mother}} \tag{3}$$

as the ratio of target areas of daughter cells to mother cells. To ensure that the target areas of all cells add up to the total size of the spatial domain, which is assumed to be fixed, we choose  $R_A = 0.5$  unless stated otherwise. In each simulation, the initial area of each cell,  $A^s$ , equals its initial target area,  $A_0^s$ , with  $A^s = A_0^s = 121 \text{ µm}^2$  (see discussion below for the choice of length scales in the model). In contrast to several previous applications [22, 25] of the vertex model the spatial domain in this study is constrained due to the fact that there is no net organism growth during embryogenesis.

Cell intercalation and apoptosis. In addition to evolving vertex positions in accordance with Eq. (1), we must account for cell intercalation and cell removal through topological rearrangements. One such topological rearrangement is a T1 swap, which simulates cell neighbour exchanges. In a T1 swap an edge shared by two cells is removed and the cells are disconnected, while a new perpendicular edge is created that then connects the cells that were previously separated by the edge (see Fig. 2B). In our implementation T1 swaps are executed whenever the length of a given edge decreases below a threshold  $l_{\min} = 0.11 \, \mu \text{m}$ , which is 100 times smaller than the approximate length of a cell at the beginning of the simulation. The length of the new edge,  $l_{\text{new}} = 1.05 l_{\text{min}}$ , is chosen to be slightly longer than this threshold in order to avoid an immediate reversion of the swap.

A second topological rearrangement in vertex models is a T2 transition, during which a small triangular cell is removed from the tissue and replaced by a new vertex (see Fig. 2B). In our implementation any triangular cell is removed if its area drops below a threshold  $A_{\min} = 0.121$   $\mu$ m<sup>2</sup>, which is 100 times smaller than the initial area of each cell. The energy function (2) in conjunction with T2 transitions can be understood as a model for cell removal: cells are extruded from the sheet by a T2 transition if they become mechanically unstable. Note that we do not discriminate between cell removal by cell death or by delamination, since this distinction is

immaterial for our purposes. However, delamination has been shown to provide an alternative way of cell removal from epithelial sheets that is distinct from apoptosis [28]. Rates of cell removal predicted by previous vertex model applications have coincided with experimental measurements in the *Drosophila* wing imaginal disc [22] and notum [28].

Cell proliferation. All simulations start with  $N_P^s = 24$  cells in the P compartment and  $N_A^s = 40$  cells in the anterior compartment, to approximately match observed cell numbers [14] and to ensure that there are similar amounts of anterior tissue on either side of the P compartment.

In the case of a wt embryonic segment each cell divides once, with cell cycle times drawn independently from the uniform distribution on 0 to  $\tilde{t}_{wt} = 600$  time units. This corresponds to the duration of the sixteenth division cycle in the epidermis, which occurs during late stage 10 and early stage 11 and takes roughly 50 minutes [19]. After the round of divisions is complete, the system is allowed to relax for 200 more time units, corresponding to a total simulation time of  $t_{wt} = 800$  time units.

For an en>CycE embryonic segment, the first round of divisions is implemented as for wt, but each cell in the P compartment then has a probability  $P_{CycE}=0.54$  of dividing a second time once the first round of divisions is complete, with cell cycle times drawn independently from the uniform distribution from  $\tilde{t}_{CycE}=600$  to  $\tilde{t}_{CycE}=1200$  time units. This probability is inferred from published data on the en>CycE+p53 perturbation [14]; in this case apoptosis is blocked, allowing us to infer the average number of cell division events. The second period of 600 time units corresponds to the duration of the ectopic divisions in the en>CycE embryos, which occur during late stage 11 and early stage 12 [14]. After the second round of divisions is complete, the system is allowed to relax for 200 more time units, corresponding to a total simulation time of  $t_{CycE}=1400$  time units.

For an en>dap embryonic segment, each cell in the P compartment has a fixed probability  $P_{dap} = 0.6$  of not participating in the single round of divisions. This probability is inferred from published data on the en>dap perturbation [14]. As with wt, divisions occur during the first  $\tilde{t}_{wt} = 600$  time units, after which the system is allowed to relax for 200 more time units, corresponding to a total simulation time of  $t_{wt} = 800$  time units.

These simulation times are chosen such that the system is at quasi-steady state at each time point. This quasi-steady state assumption is commonly used in vertex models [6, 22, 28, 29] and reflects the fact that the times associated with mechanical rearrangements (seconds to minutes) are an order of magnitude shorter than typical cell cycle times (hours) [22].

At each cell division event, a new edge is created that separates the newly created daughter cells. The new edge is drawn along the short axis of the polygon that represents the mother cell [30]. The short axis has been shown to approximate the division direction (cleavage plane) of

cells in a variety of tissues [31], including the *Drosophila* wing imaginal disc [32]. The short axis of a polygon crosses the centre of mass of the polygon, and it is defined as the axis around which the moment of inertia of the polygon is maximised. Each daughter cell receives half the target area of the mother cell upon division unless stated otherwise.

Geometry and boundary conditions. In order to simulate the subsections of the P compartment we consider a spatial domain comprising two adjacent cell populations, the cells in the P compartment and parts of the adjacent tissue in the anterior compartment on either side of it. Sample simulation images are shown in Fig. 2A and Fig. 2C. Motivated by the repeated pattern of A and P compartments along the anterior-posterior axis of the embryo, as well as by the fact that single P compartments stretch farther dorso-ventrally than the simulated region, doubly periodic boundary conditions are applied (Fig. 2A). These boundary conditions keep the simulated region of interest at a fixed size. Compartment size changes are analysed as changes in the relative proportions of the anterior and posterior compartment within the simulated region.

An analysis of the sensitivity of P compartment sizes and cell numbers to this choice of boundary condition is provided in Text S1 and Choice of simulation boundary condition does not significantly influence simulation results. The precise choice of boundary condition imposed in the model simulations does not significantly affect predicted compartment sizes and cell numbers.

To enable comparison of cell death rates in the front and back halves of the P compartment (see Fig. 1G), a cell is defined to be in the front or back half if its centroid is located to the anterior ('left') or posterior ('right') side of the centre of the tissue, respectively. The tissue centre is defined to be the horizontal midpoint of the sheet at time t=0 and is held fixed at all times.

When computing measures of cell shape in our analysis of simulation results, we define the area and perimeter of a cell to be those of the associated polygon in the vertex model, while 'cell elongation' is defined as the square root of the ratio of the largest to the smallest eigenvalues of the moment of inertia of that polygon. This latter measure provides a robust way to measure elongations of arbitrary shapes [30] and is comparable to the ratio of the lengths of the long and short axis of the best fit ellipse to a cell.

Compartment boundary line tension. Unless stated otherwise, the line tension along the compartment boundaries is set to  $\Lambda_b = 2\Lambda$ , twice the value of the line tension in the remainder of the tissue. High tension along compartment boundaries is known to promote cell sorting and boundary straightness [6, 29], and the presence of myosin cables that can generate this tension between A and P compartments in the *Drosophila* embryonic epidermis has been reported [33]. Choice of simulation boundary condition does not significantly influence simulation results shows that while the increase in line tension along compartment boundaries does affect the straightness

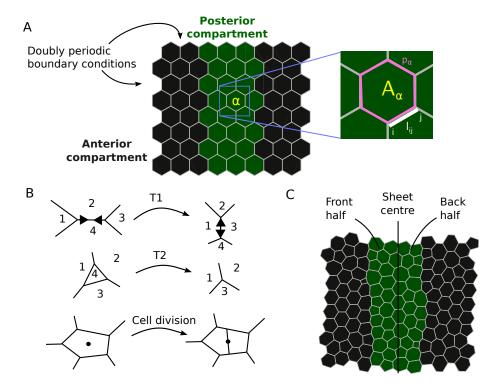


Figure 2. Vertex model of posterior compartment dynamics during the last division cycle in the *Drosophila* embryonic epidermis. (A) Snapshot of the initial tissue configuration for each simulation, with mechanical parameters in Eq. (2) annotated. (B) Schematic diagram of a junctional rearrangement (T1 swap), a cell removal (T2 transition), and cell division in the vertex model. Numbers indicate cell indices. (C) Snapshot of a *wt* simulation at the final time point, once all cell divisions have occurred, with annotation for the front and back halves of the P compartment. Parameter values are listed in Table 1.

of the boundary between A and P compartments in the model simulations, it does not significantly affect compartment sizes or cell numbers.

**Incorporating mechanical asymmetry.** To investigate the consequences of asymmetries in cell mechanical properties on P compartment size control and patterning of apoptosis, we consider three distinct cases.

In the first case, we allow for asymmetry in cell target areas in the P compartment. This is implemented by modifying the target area of each cell in the P compartment to take the form

$$A_0' = (R_A)^g (1 \pm \lambda_A),\tag{4}$$

where  $R_A = 0.5$  as listed in Table 1,  $g \in \{0, 1, 2\}$  denotes the generation of the cell, and the – and + signs apply to cells located in the front and back halves of the compartment, respectively. We refer to the parameter  $\lambda_A$  as the area asymmetry.

In the second case, we allow for asymmetry in line tensions in the P compartment. This is implemented by modifying the line tension of each cell-cell interface (edge) inside the P compartment to take the form

$$\Lambda = \Lambda_r (1 \pm \lambda_l), \tag{5}$$

where  $\Lambda_r$  is the value of the line tension when no asymmetry is imposed. The + sign applies to all edges between P compartment edges whose midpoint is the front half of the compartment, while the - sign applies to all edges whose midpoint is in the back half of the compartment. We refer to the parameter  $\lambda_l$  as the *line tension asymmetry*.

In the third case, we allow for asymmetry in perimeter contractility in the P compartment. This is implemented by modifying the perimeter contractility of each cell in the P compartment to take the form

$$\Gamma = \Gamma_r (1 \pm \lambda_p),\tag{6}$$

where  $\Gamma_r$  is the value of the perimeter contractility when no asymmetry is imposed, and the + and – signs apply to cells in the front and the back halves of the P compartment, respectively. We refer to the parameter  $\lambda_p$  as the *perimeter asymmetry*.

The asymmetry parameters are all fixed at 0 in Figs. 3, S1 and S2, and are varied in Figs. 4, 5 and 6.

Numerical implementation. Prior to solving the model numerically, we non-dimensionalise it. Non-dimensionalising reduces the number of free parameters in the system and facilitates comparison of parameter values to previous implementations of the vertex model [22]. Rescaling space by the characteristic length scale L and time by the characteristic time scale T, equations (1) and (2) become

$$\frac{\mu L}{T} \frac{\mathrm{d}\mathbf{x}'_i}{\mathrm{d}t'} = -\frac{1}{L} \nabla_i' E,\tag{7}$$

$$E = \sum_{\alpha} \frac{KL^4}{2} (A'_{\alpha} - A'_{0,\alpha})^2 + \sum_{\langle i,j \rangle} \Lambda L l'_{i,j} + \sum_{\alpha} \frac{\Gamma L^2}{2} p'^2_{\alpha},$$
 (8)

where  $\mathbf{x}'_i$ ,  $A'_{\alpha}$ ,  $A'_{0,\alpha}$ ,  $l'_{i,j}$  and  $p'_{\alpha}$  denote the rescaled  $i^{\text{th}}$  vertex positions, the rescaled cell area and cell target area, and the rescaled cell perimeter, respectively. The symbol  $\nabla'_i$  denotes the gradient with respect to the rescaled  $i^{\text{th}}$  vertex position. Multiplying the first equation by  $T/\mu L$ , we obtain

$$\frac{\mathrm{d}\mathbf{x}'_i}{\mathrm{d}t'} = -\nabla_i' \frac{T}{\mu L^2} E,\tag{9}$$

$$E' = \frac{T}{\mu L^2} E = \sum_{\alpha} \frac{TKL^2}{\mu} \frac{1}{2} (A'_{\alpha} - A'_{0,\alpha})^2 + \sum_{\langle i,j \rangle} \frac{\Lambda T}{L\mu} l'_{i,j} + \sum_{\alpha} \frac{\Gamma T}{\mu} \frac{1}{2} p'^2_{\alpha}.$$
 (10)

Finally, by introducing the time scale  $T = \mu/KL^2$ , and the rescaled mechanical parameters  $\overline{\Lambda} = \Lambda T/(L\mu) = \Lambda/KL^3$ ,  $\overline{\Gamma} = \Gamma T/\mu = \Gamma/(KL^2)$  the non-dimensionalised equations read

$$\frac{\mathrm{d}\mathbf{x}'_i}{\mathrm{d}t'} = -\nabla_i' E',\tag{11}$$

$$E' = \sum_{\alpha} \frac{1}{2} (A'_{\alpha} - A'_{0,\alpha})^2 + \sum_{\langle i,j \rangle} \overline{\Lambda} l'_{i,j} + \sum_{\alpha} \frac{\overline{\Gamma}}{2} p'^2_{\alpha}.$$
 (12)

We choose the characteristic length scale  $L=11~\mu m$  such that  $L^2$  is the mean cell area in the P compartment at the start of the simulation period, i.e.  $121~\mu m^2$ ; the P compartment occupies a total area of  $2.76\times10^3~\mu m^2$  [14] and is initialized with 24 cells. The precise value of the characteristic time scale T depends on tissue properties ( $\mu$  and K) and could be inferred from the duration of vertex recoil after laser ablation experiments, for example. In the non-dimensionalised model, cell shapes are governed by the rescaled target area of each cell and the rescaled mechanical parameters,  $\overline{\Lambda}$  and  $\overline{\Gamma}$ . For these parameters we use previously proposed values [22], unless stated otherwise. A complete list of parameters used in this study is available in Table 1.

Table 1. Description of parameter values used in our simulations.

| Parameter  | Description   | Value            | Reference |
|--|---|------------------|-----------|
| $\overline{\Lambda}$   | Adhesion parameter between cells in same compartment          | 0.12             | [22]      |
| $rac{\overline{\Lambda}_r}{\overline{\Lambda}_b} \ \overline{\Gamma}$ | Reference adhesion parameter for asymmetry simulations        | 0.12             | [22]      |
| $\overline{\Lambda}_b$   | Adhesion parameter between cells in different compartments    | $2 \times 0.12$  | [6]       |
| $\overline{\Gamma}$  | Cell perimeter contractility                                  | 0.04             | [22]      |
| $\overline{\Gamma}_r$  | Reference perimeter contractility for asymmetry simulations   | 0.04             | [22]      |
| $\Delta t'$  | Time step   | 0.01             | [30]      |
| $A'_{\min}$  | T2 transition threshold                                       | 0.001            | [30]      |
| $l'_{ m min}$  | T1 swap threshold   | 0.01             | [30]      |
| $l'_{ m new}$  | Distance between new edge nodes after swaps                   | $1.05 l'_{\min}$ | [30]      |
| $t_{wt}$   | End time of the simulation for $wt$ and $en>dap$              | 800              | _         |
| $t_{CycE}$   | End time of the simulation for $en>CycE$                      | 1400             | _         |
| $	ilde{t}_{wt}$  | Time when first round of divisions finishes                   | 600              | _         |
| $t^s_{CycE}$   | Time when second round of divisions starts in $en>CycE$       | 600              | _         |
| $	ilde{t}_{CycE}$  | Time when second round of divisions finishes in $en>CycE$     | 1200             | _         |
| $A'^s$   | Initial cell area   | 1.0              | [22]      |
| $A_0^{\prime s}$   | Initial cell target area                                      | 1.0              | [22]      |
| $N_P^s$  | Initial cell number inside P compartment                      | 24               | [14]      |
| $N_A^s$  | Initial cell number inside A compartment                      | 64               | [14]      |
| $P_{dap}$  | Probability for P cells to not divide in $en>dap$             | 0.6              | [14]      |
| $P_{CycE}$   | Probability for P cells to divide twice in $en>CycE$          | 0.54             | [14]      |
| $R_A$  | Ratio between target areas of mother cells and daughter cells | 0.5              | _         |
| $\lambda_A$  | Area asymmetry  | 0.0              | _         |
| $\lambda_l$  | Line tension asymmetry  | 0.0              | _         |
| $\lambda_p$  | Perimeter asymmetry   | 0.0              | _         |

For parameter values for which no reference is given, please see main text for details on how these values were estimated.

To solve Eqns. (11) and (12) numerically we use an explicit forward Euler scheme:

$$\mathbf{x}'_{i}(t' + \Delta t') = \mathbf{x}'_{i}(t') - \nabla'_{i}E'(t')\Delta t'. \tag{13}$$

The time step used in the forward Euler scheme is 0.01 rescaled time units and is manually chosen to ensure that the numerical scheme converges and that a further reduction in the time step does not change the results.

We implement the model within Chaste, an open source C++ library that provides a systematic framework for the simulation of vertex models [30,34]. All code used to implement model simulations and to generate results presented in this work is provided (see Software S1).

## Results

We first analyse the extent to which passive mechanical forces can lead to stable tissue size control as observed in [14]. We then investigate the effect of spatial regulation of cellular mechanical properties on P compartment sizes, cell numbers, and cell death locations.

Compartment size control can emerge from passive mechanical forces. As an initial study, we analyse simulations where compartment size is governed solely by passive mechanical properties of individual cells, and no further regulatory mechanism for size control is assumed. In particular, all cells in the tissue are specified to have the same mechanical properties, with the exception of interfaces shared by cells at the compartment boundary. As we shall show, such passive mechanical interactions are sufficient to explain the robustness of compartment size to hyperplastic manipulations.

Fig. 3A shows snapshots of individual simulations of wt, en>dap and en>CycE embryonic segments. We observe cells that are larger but fewer in number in en>dap than in wt, while the en>CycE compartment contains more smaller cells. Generating statistical distributions by running 100 simulations in each case, we obtain the summary statistics visualized in Fig. 3B-C. To allow for comparison with observed values we superimpose on each panel in Fig. 3B-C either the upper and lower bounds in observed P compartment areas [14] across the three perturbations (shaded gray) or the upper and lower limits in cell numbers for each perturbation separately (blue, green, red for wt, en>CycE, and en>dap, respectively). We do not plot the distinct shaded regions in the case of P compartment areas since the regions for individual perturbations overlap. Fig. 3B shows that, for wt and en>dap, the average P compartment sizes and cell numbers at the end of the final round of divisions predicted by the model closely match observed values. The difference in cell number between simulation and experiment for en>CycE is statistically significant (17%), indicating that the model underestimates the number of cell deaths in this perturbation.

These simulation results were achieved using literature values of the parameters  $\overline{\Lambda}$  and  $\overline{\Gamma}$  [22], and by assigning daughter cells to have half the target area of their mother cells ( $R_A = 0.5$ ). Although the model is a drastic simplification of epithelial compartment size homeostasis, the *in silico* results provide a close match to experimental values without any parameter tuning. The model thus provides a simple explanation for the emergence of P compartment size control [14]: size control can be achieved through passive mechanical forces without any further regulation of cellular properties through signalling gradients.

To explore how robust the observed size control is to the model parameters, a single parameter sensitivity analysis was performed while fixing the remaining parameters at their values listed in Table 1 (Fig. 3C). For most parameter values considered, the simulation results fall within the

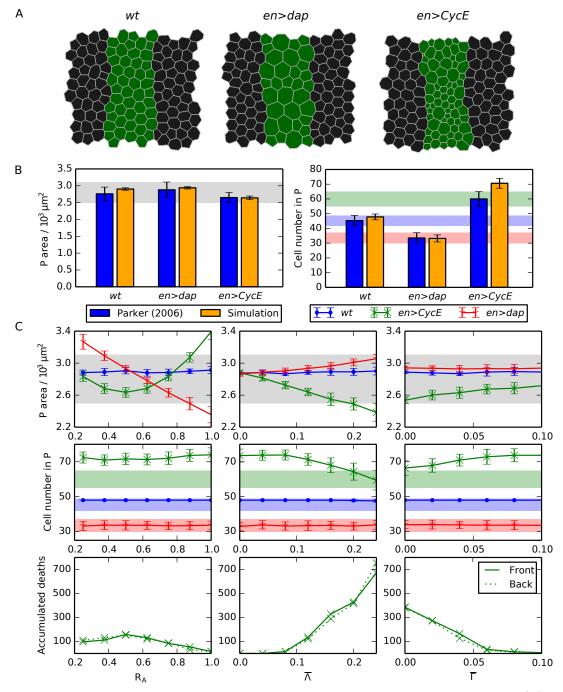


Figure 3. Compartment size control can emerge from passive mechanical forces. (A) Snapshots of wt, en>dap and en>CycE simulations, each following the final round of division. Parameter values are listed in Table 1. (B) Comparison of simulated P compartment areas and cell numbers with observed values [14]. Mean values from 100 simulations are shown and error bars are standard deviations. (C) Variation of P compartment area (upper row) and cell number (middle row), and of the number of accumulated cell deaths in the en>CycE perturbation over 100 simulations in the front and back halves of the P compartment (lower row), as each mechanical parameter is varied individually, holding all other parameters at their values listed in Table 1. Shaded areas in (B) and (C) mark the ranges of experimentally observed values and are added for reference (see main text for details).

bounds of experimentally observed values, except for values of the target area ratio  $R_A$  smaller than 0.4 and larger than 0.9, and for values in  $\Lambda$  larger than 0.2.

Focusing on the results of en>CycE simulations, the model exhibits some counter-intuitive dependence on parameter values. In particular, uniformly low perimeter contractility,  $\overline{\Lambda}$ , or high line tension,  $\overline{\Gamma}$ , leads to mechanically induced P compartment shrinkage. In addition, an increase or decrease of  $R_A$  away from 0.5 will increase compartment sizes for the en>CycE perturbation. We may interpret these results as follows.

In the model, the apical area of a cell is determined by the balance of forces. The perimeter contractility and line tension terms act to minimise edge lengths and perimeters of cells. These force contributions can be counteracted by the area term, which acts to keep the cell close to its target area, or by stretching forces exerted by neighbouring cells. Upon division, a new edge is created, which adds an inward contractile force that any expansive forces must counteract. Therefore, daughter cells occupy a smaller area than their mother cell once they reached mechanical equilibrium. The observation that an increased rate of cell division leads to tissue shrinkage is counterintuitive, yet not unrealistic; data from [14] for en>CycE and en>CycE+p53 embryonic compartments show a similar trend, in which the more cells are present, the smaller the compartment area. Inhibition of cell death in the en>CycE+p53 leads to more cells, but smaller compartments.

A similar mechanism explains the dependence of the size of the en>CycE compartment on the target area ratio,  $R_A$ . Mitosis induced shrinkage is a result of the perimeter contractility and line tension terms in the mechanical model. If we choose a value for  $R_A$  that is not equal to 0.5, then the target areas of all cells will no longer add up to the total area of the tissue, and more cells will have areas that are far away from their actual target areas. This increases the absolute value of the area elasticity term in the energy equation, and hence reduces the relative strength of the perimeter contractility and line tension terms. As the relative strength of these two terms decreases, the extent of mitosis-induced shrinkage is also reduced.

All the observed changes in P compartment sizes and cell numbers remain within experimentally measured values (Fig. 3, shaded regions), the exception being the P compartment cell numbers for the en>CycE perturbation. The discrepancy between observed values and  $in\ silico$  results for the P compartment cell numbers in en>CycE is insensitive to parameter variation.

The robustness of the simulation results in Fig. 3B to parameter values provides further confirmation that size control is a natural outcome of passive mechanical cellular interactions in our model. However, this model fails to capture the observed asymmetry in cell death locations, as measured by the ratio of accumulated cell death occurrence between the front and the back half of the P compartment across multiple embryos. The third row of Fig. 3C shows that the total number of cell deaths across 100 simulations is the same between the front half and the

back half of the P compartment. Here we only plot the cell death occurrences of the en>CycE simulations, since no cell deaths were observed in any wt or en>dap simulations. This is in close agreement with experimental results [14], where only 0.7 (wt) or 0.2 (en>dap) cell deaths where identified by TUNEL staining per embryo.

We draw two main conclusions from the simulations presented in Fig. 3: (i) mechanical interactions between identical cells can explain robust size control of all considered genetic perturbations (wt, en>CycE, en>dap), even if the parameters are varied significantly; (ii) passive mechanical interactions of cells with uniform mechanical properties cannot explain the observed asymmetry in cell death occurrence, nor completely recapitulate the changes in cell numbers for the en>CycE perturbation.

Spatial patterning of cell death emerges from differential growth or differential mechanical regulation. We next use the model to analyse how asymmetries in cellular mechanical properties across the P compartment may lead to the observed spatial patterning of apoptosis. We consider three cases (Fig. 4A): (i) 'area regulation', which refers to patterning of the cell target area ratio  $R_A$  through the parameter  $\lambda_A$ ; (ii) 'line tension regulation', which refers to patterning of the line tension  $\overline{\Lambda}$  through the parameter  $\lambda_l$ ; and (iii) 'perimeter regulation', which refers to the patterning of the perimeter contractility  $\overline{\Gamma}$  through the parameter  $\lambda_p$ . These parameters are defined in the *Materials and Methods* section.

Fig. 4B-D shows the impact of small amounts of asymmetry on P compartment dynamics. In each of the cases (i)-(iii), we set the relevant asymmetry parameter to 0.2, while keeping the other two asymmetry parameters fixed at 0. Fig. 4B shows snapshots of simulation outcomes for each case. A visual inspection suggests that these three cases give rise to a P compartments with similar cell sizes and shapes as in Fig. 3A.

Fig. 4C shows that P compartment sizes and cell numbers are not affected by these small amounts of asymmetry in the tissue. In each case, the *in silico* compartment area and cell number is as close to the observed values [14] as the passive mechanical model. Although cell mechanical properties are now patterned, compartment size control still emerges within the model. Fig. 4D shows the total number of cell deaths in the front and the back halves of the P compartment across 100 simulations in each asymmetry case. We find that each case can explain the observed spatial asymmetry in cell death locations.

Robustness of compartment size and compartment cell number to cellular asymmetry. To assess to which extent P compartment sizes and cell numbers are robust to spatial asymmetry in cell mechanical properties, we next vary each of the three asymmetry parameters in turn while keeping the others fixed at 0. Fig. 5 shows that increases in asymmetry lead to

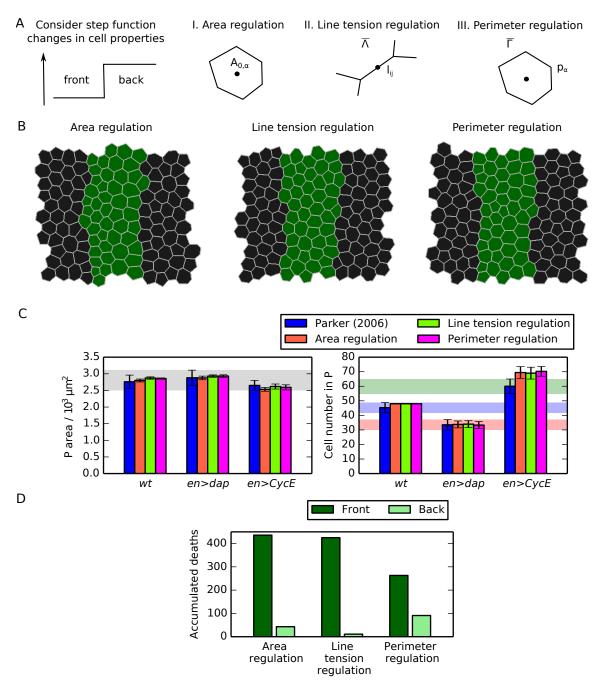


Figure 4. Spatial regulation of mechanical cell properties can induce asymmetry of cell death occurrence inside posterior compartments. (A) Schematic of the distinct forms of mechanical asymmetries considered in this work. (B) Snapshot of final configuration of simulations for each considered perturbation. (C) Comparison of P compartment areas and cell numbers for each of the considered perturbations with experimental values. Mean values from 100 simulations are shown and error bars are standard deviations. Parameter values in Table 1 and in the main text. Shaded areas mark the ranges of experimentally observed values and are added for reference and comparison with Figs. 3, S1 and S2. (D) Comparison of accumulated number cell death over 100 simulations in the front and back halves of the P compartment for each of the considered perturbations.

decreases in P compartment sizes and cell numbers (top and middle row) and the degree of asymmetry in cell death across the front and back halves of the compartment (bottom row).

In the model, P compartment sizes and cell numbers are most sensitive to asymmetry in cell target areas; for example, a value of  $\lambda_A > 0.9$  can result in loss of the entire P compartment. In contrast, P compartment sizes and cell numbers remain within experimentally measured regimes for values of  $\lambda_p$  or  $\lambda_l$  from 0 up to 0.4.

Differential growth and mechanical regulation generate distinct distributions of cell shapes. To identify experimentally observable signatures to differentiate between modes of regulating compartment homeostasis, we examined the distributions of four measures of cellular morphology for the scenarios described in Fig. 4. We extract the distributions of cell areas, cell perimeters, lengths of edges between cells, and cell elongations within the P compartment at the end of each simulation. We observe distributions of these four measurements in the posterior compartment as a whole, and in the front and the back half of the compartment separately. The results of this investigation are summarized in Fig. 6.

The top two rows of Fig. 6 show that the distributions of cell areas and cell perimeters (row 1 and 2) for the area regulation scenario are distinct from the corresponding distributions for the line tension and the perimeter regulation scenario. In particular, the distributions of all areas and all perimeters are bimodal for the area regulation scenario, and unimodal for the line tension and perimeter regulation scenarios. This is marked by nearly non-overlapping distributions of cell areas and cell perimeters in the front and the back halves of the compartment for the area regulation scenario, whereas these distributions are overlapping in the line tension and perimeter regulation scenarios. Upon decomposing cell area distributions into contributions from the front and back halves of the P compartment, we see that the mean cell area is different between these two halves in the area and perimeter regulation scenarios. Cell elongations and edge lengths have similar shapes and mean values for all three asymmetry scenarios (rows 3 and 4 of Fig. 6).

The results in Fig. 6 suggest that it is possible to distinguish the scenario 'area regulation', which corresponds differential growth across the compartment, from the 'line tension regulation' and 'perimeter regulation' scenarios, which correspond to regulation of apical mechanical properties, by measuring the distributions of cell areas or cell perimeters in the front and the back halves of the posterior compartment separately. Measuring edge lengths or cell elongations will not reveal differences between the scenarios.

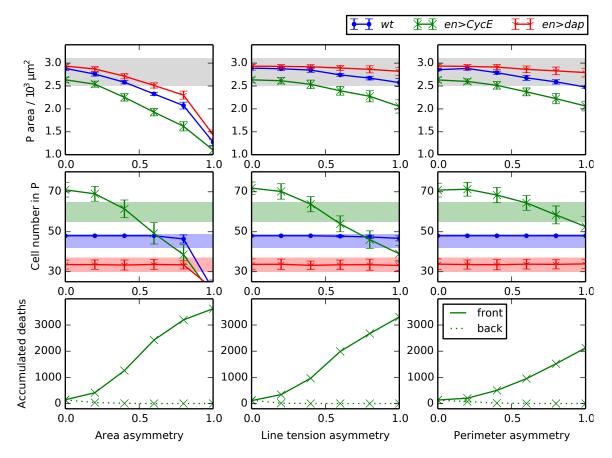


Figure 5. Sensitivity of P compartment size and cell number to asymmetry. Variation of P compartment area (upper row) and cell number (middle row), and of the number of accumulated cell deaths over 100 simulations in the front and back halves of the P compartment (lower row), as the asymmetry parameters  $\lambda_A$ ,  $\lambda_l$ , and  $\lambda_p$  individually while holding all other parameters at their values listed in Table 1. Shaded areas are added for comparison with Figs. 3 and 5.

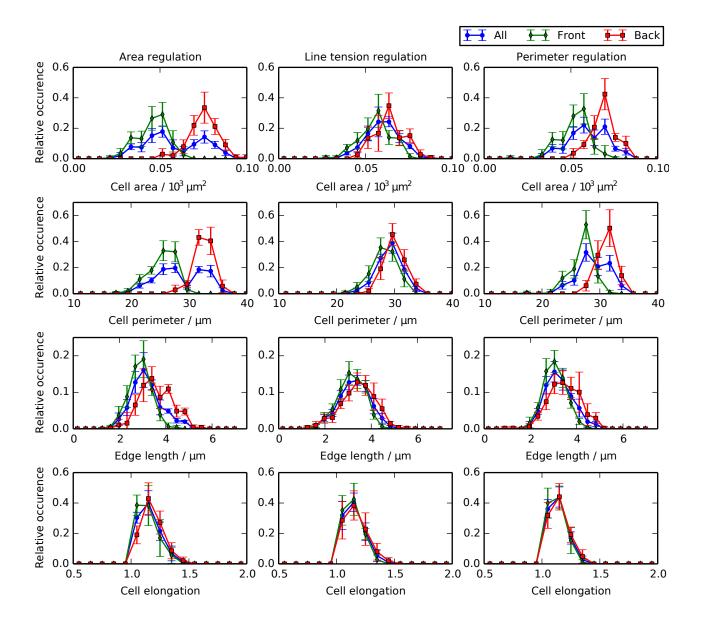


Figure 6. Differential growth and mechanical regulation generate distinct distributions of cell shapes. Distributions of cell areas (row 1), cell perimeters (row 2), cell edge lengths (row 3), and cell elongations (row 4) for each scenario of cellular asymmetry. We distinguish distributions for all cells in the posterior compartment ('All'), for cells the cells in the front half ('Front'), and cells the back half ('Back').

# Discussion

We have employed a vertex model of a *Drosophila* embryonic segment to test hypotheses about the emergence of size control. A comparison of the *in silico* segment with extant literature values indicated that passive mechanical forces suffice to explain the observed size control. However, the observed spatial heterogeneity in cell death frequencies requires some form of patterning of mechanical properties across the tissue. Several conceptually distinct modifications of the model can explain size control while also recapitulating the spatially varying rates of cell death: first, individual cells could regulate their sizes through differential growth; and second, cells could regulate their apical mechanical properties through differential expression of tension regulating protein activities. It is possible to distinguish these two scenarios within the model by the spatial distribution of P compartment cell areas and perimeters. These results hint at two possible mechanistic functions of trophic signalling pathways, such as EGFR or Wg [14, 35, 36]: they could either cause growth of individual cells, or else modulate cell shape through regulation of contractile cytoskeletal activity, either of which would explain the experimentally observed shrinkage or growth when the pathways are perturbed [14].

Connecting robustness of proportional size control to cell mechanics. Understanding the mechanism of tissue size control is particularly challenging due to the interconnected and complex nature of cell signaling and the high degree of feedback between cell- and tissue-level processes. Computational models therefore offer an important tool for investigating and testing hypothesised mechanisms and to abstract the principles underlying developmental robustness [37–39].

The development of multicellular organisms requires control of total cell numbers and relative proportions of cell types with tissues. Size control can be divided into two steps: initial specification and maintenance [12]. While much work has focused on the regulation of the position of cellular fates during early embryonic development, the mechanisms that ensure the maintenance of tissue size and of boundaries between tissues is less well understood. In particular, the physics of size homeostasis for patterned epithelia are not well understood, yet they are a recurring theme in development [40,41] and it is increasingly recognised that mechanical feedback plays a role in controlled tissue behaviour [8,42].

Advancing our knowledge of how embryos achieve robustness to defects or damage to the initial patterning of tissue domains is important for understanding the underlying causes of birth defects, as well as diseases with an underlying basis of misregulated growth, such as cancers.

Providing predictions to guide future experimental inquiries into pattern repair. Although several studies have investigated the robustness of sizes of patterned epidermal segments

of *Drosophila*, quantification has been somewhat sporadic and diffuse. This will in general require a thorough systems-level characterization of later stages of *Drosophila* morphogenesis for multiple experimental perturbations. The present study provides a basis for guiding future experiments that seek to identify possible modes of size control in late stages of epithelial development in *Drosophila*.

How could model predictions be validated against such experiments? Several previous vertex models of developing epithelia have been validated against key summary statistics. Such studies have focused primarily on the *Drosophila* wing imaginal disc, which undergoes up to 9 rounds of divisions to arrive at a distinct distribution of cell polygon numbers [22,24]. In these studies, it is safe to assume that the initial distribution of cell polygon numbers will not affect simulation outcomes, due to the high levels of proliferation. Here, we considered one or two rounds of divisions; over such a short developmental timespan we expect the initial sheet topology to influence final polygon distributions. Hence, for a quantitative comparison of this summary statistic between model and data, experimentally informed cell shapes in late stage 10 segments may be required. Such summary statistics remain lacking for the *Drosophila* embryonic epidermis during its development, and poses an experimental challenge due to the small system size (20-60 cells). Large sample sizes will be required to obtain accurate distributions of cell polygon numbers. Future iterations of the model may be further constrained through inference of mechanical parameters from laser ablation [6] or less invasive experimental protocols [43].

Current limitations. Embryogenesis is an extremely complex process. To make headway into the factors that influence robustness of tissue size maintenance, there needs to be conscious decoupling and abstraction through studies of simpler systems. This is also part of the rationale for studies in genetic model organisms from the worm and fly to mouse [39].

Due to the lack of kinematic data on cell shapes and compartment sizes during the latter stages of embryogenesis, we have not included an analysis for this initial study and have focused on more local mechanisms. In particular, we assumed that the overall tissue dimensions are constant during the considered time frame, since the epidermis forms at the outside of the embryo during stage five of *Drosophila* development and as a whole does not change dimensions for the remainder of the development. However larger scale tissue morphogenetic movements, which are undoubtedly important for aspects of morphogenesis [44], may affect the exact size of a given subsection of the tissue. For example, dorsal closure occurs during the considered time frame, which leads to an extension of the tissue that we study [45]. The assumption that this extension should not affect the relative proportions in A and P compartment size requires future experimental validation. In addition, our finding that elevated tension along compartment boundaries does not affect compartment sizes may be contrasted with theoretical and experimental studies showing how

differential line tension, either at compartment boundaries or across tissues, may drive convergent extension [46, 47]. A key conceptual difference between the present work and these studies is the assumption of a fixed, or free, boundary to the tissue.

Our model relies on the quasi-steady state assumption that the tissue is at mechanical equilibrium at each time point. We justified this assumption on the basis that individual cell cycle times of the 16th division cycle in Drosophila development are around an hour [9]. However, if cell divisions occur highly synchronously, then, this assumption might not hold. In en>CycE embryonic segments, the numbers of cell divisions events in the model were inferred from data where apoptosis was blocked by expressing the protein p35 in the P compartments [14]. It has previously been reported that epithelial sheets can extrude cells that are not undergoing apoptosis [28]; if this occurs to a great extent in the Drosophila embryonic epidermis, then our inferred numbers of mitotic events would require adjustment. In this case, an  $in\ vivo$  cell tracking study would be necessary to measure the levels of cell division and extrusion events. Such data would also help to shed light, for example, on the possible impact of mitotic cell rounding on local cell shapes and possible short-range correlations between mitosis and apoptosis events.

Due to a current lack of data in the literature, our model does not include a description of upstream patterning of cell types. Instead, we infer the necessity of patterning of cell mechanics through simulations. This study is timely as it provides some guidance into important parameters and considerations that should be taken into account in future quantitative analyses of late stages of epidermal development including germband retraction and head involution. From the results presented here, further questions arise. If a passive mechanical model is sufficient to explain compartmental size control, then what is the functional role of Spitz-mediated EGFR regulation? It is known that EGFR signalling is required for dorsal closure during *Drosophila* development [48]. Hence, it is possible that the influence of EGFR signalling on larval compartment sizes reflects the role of EGFR signalling in convergent extension during dorsal closure. If the asymmetry in our model reflects patterning of mechanical properties through trophic signalling, then a more detailed experimental analysis of the spatio-temporal dynamics of cellular signalling will allow more detailed modelling of how these properties may be patterned.

Summary and larger implications. Our study serves as an example of using computational models as an abstraction of the maintenance of tissue sizes with implications for a broad range of studies. Significant advances in stem cell engineering have resulted from understanding how to unlock the potential for multicellular aggregates to self-organize. Recent examples include the morphogenesis of optic eye cups in organ culture conditions [49] and the engineering of beating mini-hearts [50]. We posit that great success in developing tissue repair strategies will come through the reverse engineering of pattern repair mechanisms in situations where pattern repair is

perturbed. Such reverse engineering will require guiding experimental efforts through modelling studies that identify the information needed to distinguish between mechanisms.

# **Supporting Information**

#### Software S1

Zipped folder containing implementation of the computational model and analysis described in this study. A thoroughly documented example of how to run the code is provided in the file README.txt.

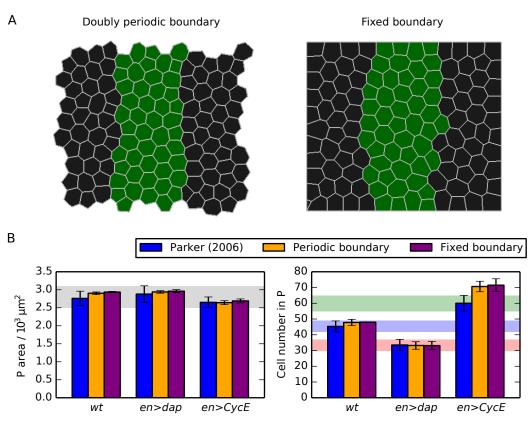
#### Text S1

Immunohistochemistry. A *Drosophila* line expressing GAL4 under the engrailed (en) promotor and CD8::GFP under the UAS promotor was used. Immunohistochemistry on embryos was performed as described previously [51] with rabbit anti-dpERK (1:100, Cell Signaling), rat anti-DCAD2 (1:100, DSHB) and DAPI (5 μg/ml, Invitrogen DU1306), with goat anti-rat IgG 561 (1:500, Invitrogen), and goat anti-rabbit IgG 647 (1:500, Invitrogen).

Confocal microscopy. Confocal z-stacks were collected using an Andor spinning disc confocal microscope with a piezo stage at 1.0 µm intervals. For each embryo, six by six grids with thirty-three percent overlap were collected for each of the four channels. Image collection was performed with MetaMorph® version 7.0.11.

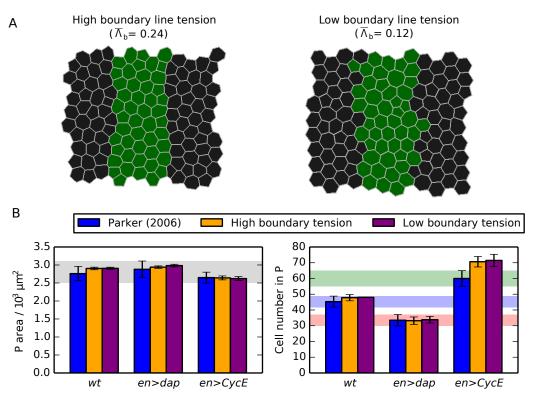
Choice of simulation boundary condition does not significantly influence simulation results. Fig. S1 compares the results of simulations using the doubly periodic boundary conditions described in the main text to those where fixed boundary conditions are imposed on the tissue as follows. Vertices on the boundary of the tissue experience the same forces as non-boundary vertices. At each time step, after applying the forces to all vertices and updating their positions, we map the boundary vertices perpendicularly back onto the boundary. If a cell on the boundary undergoes a T2 transition, then the newly created vertex is also mapped onto the boundary. The dimensions of this fixed boundary are chosen such that the total area of the tissue equals the sum of the target areas of all cells of the tissue. The tissue occupies a total of 64 rescaled area units, or  $7.744 \times 10^3 \, \mu \text{m}^2$ . The length of each fixed boundary is 8 L, i.e. 88  $\mu$ m. Fig. S1 shows that imposing this alternative boundary condition gives similar predicted P compartment areas and cell numbers to those obtained using doubly periodic boundary conditions across wt, en > dap and en > CycE simulations.

Figure S1



Choice of boundary condition does not affect P compartment sizes and cell numbers. (A) Snapshots of a wt simulation at the final time point, once all cell divisions have occurred, where doubly periodic (left) or fixed (right) boundary conditions are imposed. Parameter values are listed in Table 1. (B) Comparison of P compartment areas and cell numbers for wt, en>dap and en>CycE simulations where doubly periodic or fixed boundary conditions are imposed. Mean values from 100 simulations are shown and error bars are standard deviations. Shaded areas mark the ranges of experimentally observed values and are added for reference and comparison with Figs. 3, 4 and S2.

Figure S2



Compartment boundary line tension does not affect P compartment sizes and cell numbers. (A) Snapshots of a wt simulation at the final time point, once all cell divisions have occurred, where either a high (left) or low (right) line tension,  $\overline{\Lambda}$ , is imposed at the boundary between A and P compartments. Parameter values are listed in Table 1. Compartment boundary line tension promotes cell sorting and straightness of the boundary, but does not affect compartment sizes. (B) Comparison of P compartment areas and cell numbers for wt, en>dap and en>CycE simulations where a high (left) or low (right) compartment boundary line tension is imposed. Values for  $\overline{\Lambda}$  at the compartment boundary are those given in (A). Mean values from 100 simulations are shown and error bars are standard deviations. Shaded areas mark the ranges of experimentally observed values and are added for reference and comparison with Figs. 3, 4 and S1.

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