Computational study of the impacts of mechanical and physical cell properties on mitotic cell rounding in developing epithelia

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Abstract

Mitotic rounding (MR) during cell division is critical for the robust segregation of chromosomes into daughter cells and is frequently perturbed in cancerous cells. MR has been studied extensively in individual cultured cells, but the physical mechanisms regulating MR in intact tissues are still poorly understood. A cell undergoes mitotic rounding by simultaneously reducing adhesion with its neighbors, increasing actomyosin contraction around the cortex, and increasing the osmotic pressure of the cytoplasm. Whether these changes are purely additive, synergistic or impact separate aspects of MR is not clear. Specific modulation of these processes in dividing cells within a tissue is experimentally challenging, because of off-target effects and the difficulty of targeting only dividing cells. In this study, we analyze MR in epithelial cells by using a newly developed multi-scale, cell-based computational model that is calibrated using experimental observations from a model system of epithelial tissue growth, the *Drosophila* wing imaginal disc. The model simulations predict that increase in apical surface area of mitotic cells is solely driven by increasing cytoplasmic pressure. MR however is not achieved within biological constraints unless all three properties (cell-cell adhesion, cortical stiffness and pressure) are simultaneously regulated by the cell. The new multi-scale model is computationally implemented using a parallelization algorithm on a cluster of graphic processing units (GPUs) to make simulations of tissues with a large number of cells feasible. The model is extensible to investigate a wide range of cellular phenomena at the tissue scale.

Author Summary

Mitotic rounding (MR) during cell division is critical for the robust segregation of chromosomes into daughter cells and is frequently perturbed in cancerous cells. MR has been studied extensively in individual cultured cells, but the physical mechanisms regulating MR in intact tissues are still poorly understood. The newly developed computational model Epi-scale enables one to produce new hypotheses about the underlying biophysical mechanisms governing MR of epithelial cells within the developing tissue micro-environment. In particular, our simulations results predict that robust mitotic rounding requires co-current changes in cell-cell adhesion, cortical stiffness and cytoplasmic pressure, and explains how regulation of each property impacts the shapes of dividing cells in tissues.

Introduction

Epithelia are tissues composed of tightly adherent cells that provide barriers between internal cells of organs and the environment and are one of the four basic tissue types in the human body [1–3] (Fig. 1). Epithelial expansion driven by cell proliferation is a key feature throughout development, and it also occurs in hyperplasia, a precursor to cancer. Cell divisions during development must occur robustly, as mis-segregation of chromosomes leads to severe genetic abnormalities (aneuploidy) in daughter cells. Over 90% of all human tumors are epithelially-derived [4], and the accumulation of genetic errors during cell division can lead to all of the hallmarks of cancer [5]. In tissues, mitotic cells must become sufficiently round to avoid the mis-segregation of chromosomes all the while still remaining connected with their neighbors [6]. A deeper understanding of the biophysical mechanisms governing the behavior of mitotic cells in epithelia will result in a better understanding of many diseases including cancer.

Epithelial cells entering mitosis rapidly undergo structural changes that result in the apical area of the cell becoming larger and rounder, in a process known as mitotic rounding (MR) [7,8]. MR occurs in detached cells, cells adherent to a substrate as well as in epithelial cells within tissues [9–11]. The beginning of MR in epithelia coincides with an increased polymerization of actomyosin at the cell cortex, which results in an increase in cortical tension and is necessary for MR [10,12]. Simultaneously, intracellular pressure increases [10], and cells partially reduce adhesion to their neighbors and the substrate [12]. Experiments that can specifically target only dividing

cells and measure physical properties of individual cells within tissues are very

challenging.

Computational modeling coupled with experimentation has become a powerful tool for

identifying the biophysical principles governing organogenesis [13]. Biophysically-

derived computational models can complement current experimental methods by

predicting the response of tissue to mechanical perturbations of individual cells. MR is

investigated in this paper by using a novel multi-scale sub-cellular element model (SEM)

called Epi-scale that simulates epithelial cells in growing tissues. Novel biologically-

relevant features of the model include: i) separate representations of the sub-cellular

elements, as well as cell-cell interactions; ii) a detailed description of cell properties

during mitotic rounding; and iii) a systematic calibration of model parameters to provide

accurate biological simulations of tissue growth. We performed parameter sweeps on

the extent that a mitotic cell regulates cell-cell adhesion, membrane stiffness, or internal

pressure and analyzed the impacts of such parameter variations on cross-sectional

areas of mitotic cells at the apical surface as well as the roundness of mitotic cells.

Model simulations demonstrate that cell-cell adhesion and stiffness significantly impact

roundness but do not increase cell area during MR. Solely increasing cell pressure

during MR increases both cell area and roundness. However, the internal pressure

increase needed to achieve experimentally observed levels of roundness leads to

nonphysical levels of cell area expansion (Fig. S5). The model predicts that a cell must

regulate all three cellular properties simultaneously to achieve physiological levels of

area expansion and roundness without adversely affecting tissue integrity.

The paper is organized as follows. The Methods section describes the modeling

background and model development. The Results section provides details of model

calibration of single cell parameters using quantitative biophysical data. The calibrated

model predicts emergent properties of epithelial topology. The model is then used to

investigate new questions into the relative contributions of cell-cell adhesion, membrane

stiffness and intracellular pressure impact the extent of mitotic rounding. The paper

ends with a detailed discussion of the models' predictions in the more general biological

context. It also describes future extensions of the computational model environment, for

simulating epithelial tissue mechanics in greater biological detail.

Methods

Modeling background

Multiple computational approaches have been utilized to model various aspects of

epithelial tissue dynamics, each with its particular focus and applications [14]. For

example, the cellular Potts modeling (CPM) approach has been used successfully to

take into account cell adhesivity to study cell aggregation as well as cell morphogenesis

[15,16]. Finite element models (FEMs) have also been implemented to investigate

epithelial cell behavior [17,18]. Vertex based models (VBM) provided an efficient

approach to study the regulation of cell topology, tissue-size regulation, tissue

morphogenesis, and the role of cell contractility in determining tissue curvature [19–24].

Cells are defined in VBMs by the several vertices representing the meeting points of

shared cell-cell contacts (as reviewed in [19]). An implementation and comparison of the

five popular cell-based modelling approaches for simulating the self-organization of

multicellular tissues within a consistent computational framework, Chaste [25]

(http://www.cs.ox.ac.uk/chaste), was recently described in [26].

The Subcellular Elements Model (SEM), developed initially by Newman's group [27] for

simulating multi-cellular systems to encompass multiple length scales, is currently

actively used as a general computational modeling approach. SEMs have been

extended to predict how mechanical forces generated by cells are redistributed in a

tissue and for studying tissue rheology, tissue fusion, thrombus formation, and cell-cell

signaling. SEM was also used to study aspects of epithelial cell mechanics without

making assumptions about cell shapes [28]. Each cell in a SEM consists of a set of

nodes representing a coarse-grained representation of subcellular components of

biological cells. Node-node interactions are represented by energy potentials. Another

SEM developed by Jamali et al. [29] represents the membrane and nucleus of the cell

by nodes connected by overdamped springs. Gardiner et al. [30] described a SEM with

locally-defined mechanical properties. Christely et al. [31] have developed an efficient

computational implementation of the SEM simulating role of Notch signaling in cell

growth and division, on GPU cluster to decrease computational time. A particular

advantage of the SEM approach is that it can provide local representations of

mechanical properties of individual cells which can be directly related to the

experimental data [32].

The Epi-scale computational model

This paper represents detailed simulations obtained using novel multi-scale SEM called Epi-scale, of developing epithelia with a focus on the two-dimensional (2D) planar cell shapes near the apical surfaces of cells. This is a simplifying approximation that was used in many previous models of wing disc growth [22,23,33–35]. In particular, it is reasonable to use a 2D model for studying many epithelial processes in the *Drosophila* wing disc pouch because it consists of a single layer of cells and the essential structural components of those cells, including E-cadherins and actomyosin, are concentrated on the apical surface of the epithelia (Fig. 1c-d). E-cadherin is responsible for adhesion between two neighboring cells, and actomyosin, which is concentrated at the apical MR drives cell contractility. The nucleus and most of the cytoplasm are pushed up to the apical surface during cell division. Using a 2D approximation also allows us to model a large number of cells with high resolution and special attention to mechanical cell properties. However, future development of the Epi-scale simulation platform implemented on GPU clusters, will also enable 3D simulations with reasonable computational costs.

In what follows, we first describe different types of the subcellular elements that are used to simulate each cell, and the interactions between them. Then, the equations of motion of each subcellular element are provided. Finally, approaches for modeling cell's growth, transition to mitotic phase, and division are described.

Subcellular elements

Epi-scale represents individual cells as collections of two types of interacting subcellular

elements: internal nodes and membrane nodes (Fig. 2). The internal nodes account for

the cytoplasm of the cell, and the membrane nodes represent both plasma membrane

and associated contractile actomyosin cortex. The internal and membrane nodes are

placed on a 2D plane, representing the apical surface of epithelia.

Interactions between internal and membrane nodes are modeled using potential energy

functions as shown in Fig. 2a [31,36]. The internal–internal nodes interactions represent

the cytoplasmic pressure of a cell, and the interactions between membrane nodes of the

same cell are used to model the cortical stiffness. Cell-cell adhesion is modeled by

membrane-membrane nodes interactions between two neighboring cells. List of all

potential functions used in the Epi-scale to model mechanical properties of cells and

epithelial tissue and description of their biological relevance are provided in Table 1.

Epi-scale utilizes spring and Morse potential functions. Linear and torsional springs are

used for modeling interactions E_{ik}^{MMS} and E_i^{adh}, while Morse potential functions are used

for modeling interactions E_{ij}^{MI}, E_{ik}^{II}, and $E_{\alpha_i}^{MMD}$ (Fig. 2). Morse potential consists of two

terms, generating short-range repulsive and long-range attractive forces [27]. The

following expression is a Morse potential function for the E_{ij}^{MI}:

$$E_{ij}^{MI} = \left[U^{MI} \exp\left(-\frac{|x_i - x_j|}{\xi^{MI}} \right) - W^{MI} \exp\left(-\frac{|x_i - x_j|}{\gamma^{MI}} \right) \right] \tag{1}$$

190   where U^{MI}, W^{MI}, ξ^{MI}, and γ^{MI} are Morse parameters for the E_{ij}^{MI} which are carefully

191   calibrated using specific experimental data. The same form of the potential with different

192   sets of parameters is also used for E_{ij}^{MI} and E_{il}^{MMD} (Table 2). These potential functions

193   govern the motion of internal and membrane nodes inside the cells resulting in the

194   deformation and rearrangement of cells within the tissue.

Equations of motion

Equations of motion differ for membrane nodes and internal nodes. For each internal or

membrane node and at each time-point, displacement of a subcellular element is

calculated based on the potential energy functions. The model assumes that nodes are

in an overdamped regime [22,36,37] so that inertia forces acting on the nodes can be

neglected. This leads to the following equations of motion describing movements of

membrane and internal nodes, respectively:

$$\eta \dot{x}_i^I = -\left(\sum_j \nabla E_{ij}^{MI} + \sum_m \nabla E_{im}^{II} \right) \qquad i = 1,2,\dots\dots N^I \qquad (2)$$

$$\eta \dot{x}_j^M = -\left(\sum_i \nabla E_{ij}^{MI} + \sum_k \nabla E_{kj}^{MMS} + \sum_l \nabla E_{lj}^{MMD} + \nabla E_j^{adh} \right) \qquad j = 1,2,\dots N^M \qquad (3)$$

204   where η is the damping coefficient, x_i^M and x_j^M are positions of internal and membrane

205   nodes, respectively. The dot represents a time derivative.

207 Eqns. (2) and (3) are discretized in time using forward Euler method and positions of

208 nodes x_{I_i} and x_{M_i} are incremented at discrete times. The forward Euler discretization of

209 the equation of motion of internal nodes (Eqn. (2)) has the form:

$$x_i^I\,(t+\Delta t) = \ x_i^I\,(t) - \left(\sum_j \nabla\big(E_{ij}^{II}\big)(t) + \sum_k \nabla\big(E_{\alpha ik}^{MI}\big)(t)\right)\frac{\Delta t}{\eta} \tag{4}$$

where Δt is the time step size. The same discretization technique is used for the

equation of motion of the membrane nodes.

The Epi-scale was computationally implemented on the cluster of Graphical Processing

Units (GPUs). This enabled us to run large number of simulations with subcellular

resolution at micro-scale with low computational cost and to study the impact of

changes in individual cell physical properties on the tissue development at the macro-

scale. (See Supplementary Information (SI-S1) for details.)

Cell cycle

The *Drosophila* wing disc, which was used for calibrating the model, has a spatially

uniform growth rate which decreases over time [38]. The growth rate for cell *ii* is

modeled by an exponentially decaying function fit to the specific experimental data [38],

with a random term representing stochastic variation among cells:

$$g_{ii}(t) = (g_{0_{Avg}} + Rnd[-g_0, g_0])e^{-k_g t} \tag{5}$$

where $g_{0_{Avg}}$ is the average growth rate of cells in the beginning of a simulation and

$Rnd[-g_0, g_0]$ is a random number chosen from a uniform distribution in the range

$[-g_0, g_0]$. k_g is the decay constant of the growth rate.

Cells evolve through interphase and mitosis phases. Variable Cell Progress ($CP \in [0,1]$)

describes in the model cell's progress between the beginning of the interphase ($CP = 0$)

to the end of the cell division ($CP = 1$). CP is updated based on cell's growth rate as

follows:

$$CP_{ii}(t + \Delta t) = CP_{ii}(t) + g_{ii}(t) \cdot \Delta t \qquad (6)$$

The number of internal nodes inside the cell increases as the cell grows (Fig. 2). It has

been shown experimentally that epithelial cells undergoing mitosis increase their

intracellular pressure by adjusting their osmolarity relative to their surroundings [39].

Additionally, the actomyosin cortex is enriched, and cellular adhesion to the substrate

and to neighboring cells are downregulated [10,40–44]. Since these changes in mitotic

cells occur concurrently, the relative impact on mitotic cells cannot easily be

decomposed into separable effects in experiments.

To simulate MR, parameters regulating cell-cell adhesion, actomyosin cortex, and

internal pressure of cells in the mitotic phase (M phase) are modified (Table 2),

representing changes of cell physical properties during mitosis [9,10,43]. Mitosis is

modeled by a linear transition from the interphase parameter range to the mitotic

parameter range determined based on the experimental observations. For example

U^{MI}, a Morse parameter used for representing cytoplasmic pressure on the membrane

of the cell (see SI-S2.3), is varied from the interphase value (U^{MI}_{Inter}) to the mitotic value

(U^{MI}_{Mit}), by using the following function of *CP*:

$$U^{MI} = U^{MI}_{Inter} \frac{1 - CP}{1 - CP_{Mit}} + U^{MI}_{Mit} \frac{CP - CP_{Mit}}{1 - CP_{Mit}} \tag{7}$$

where U^{II}_{Mit} is the parameter value in the mitosis range. Similar linear variations of

parameter values are used for representing enrichment of actomyosin cortex and

reduction in cell-cell adhesion with neighboring cells in mitotic phase (see Table 1).

Cells in the mitotic (M) phase – which lasts approximately 30 minutes – divide into two

daughter cells (Fig. 2). As *CP* becomes equal to 1, cytokinesis occurs that is modeled

by separating internal and membrane elements of the mother cell into two sets

representing daughter cells. The axis of division is implemented perpendicular to the

cell's longest axis, following Hertwig's rule [45]. New membrane elements are created

along the cleavage plane for each daughter cell, and cell parameters for nodes of each

daughter cell are set to values from the interphase range and *CP* is reset to zero for

both daughter cells.

Membrane nodes in the beginning of a simulation are arranged in a circle for each cell,

and internal nodes randomly placed within each cell (Fig. 3a). After initialization, internal

nodes rapidly rearrange in a simulation, and cells self-organize into a polygonal

network, similar to the experimentally observed cell packing geometry of epithelia (Fig.

3b). Cells constantly grow, divide and interact with each other resulting in a detailed

simulation of the developing epithelial tissue (Fig. 3c-d).

Results

Model Calibration

Before running predictive model simulations, the model parameters, described in the

Epi-scale computational model section, were calibrated using experimental data for the

third instar *Drosophila* wing disc, which is a powerful model for studying organ formation

[46,47] (Fig. 4, SI Movie S4.1). Experimental values for epithelial cell lines were used to

calibrate the model parameters when experimental data for *Drosophila* wing disc were

not available.

Mechanical stiffness of the actomyosin cortex is primarily responsible for the modulus of

elasticity (E) and the Poisson's ratio of the cells [48]. E is experimentally obtained by

applying known force to either side to stretch a cell, and measuring the cell's

deformation based on the applied force [49,50]. This experiment can be reproduced *in*

silico by applying forces to membrane nodes on either side of a simulated cell, and

measuring the deformation (Fig. 4a-a'', 4c). Parameters corresponding to cortical

stiffness (k_{Inter}^{St} and L_{Inter}^{St}) were calibrated to have $E = 19\,kPa$, which is within the

biological range of $10 - 55\,kPa$ [49,50] measured for epithelial cells.

The cell-cell adhesive force (F_{adh}), or the force needed to detach two adhered cells

from each other, is dependent on the strength of cell-cell adhesions. F_{adh} is

experimentally obtained by measuring the force needed to detach two adhered cells

from each other. This experiment can be reproduced *in silico* by applying forces to

membrane nodes on either side of two adhered cells, and measuring the force needed

to separate them (Fig. 4b-b"). Parameters corresponding to cell-cell adhesion (k_{Inter}^{adh}

and L^{adh}) were calibrated such that $F_{adh} = 20\ nN/\mu m$ (Fig. 4d), which is in the range of

experimental results for epithelial MDCK cells [51] and E-Cadherin-transfected S180

cells [52].

Cells in the wing disc have spatially-uniform growth-rates that slow down as the tissue

approaches its final size [38]. The growth rate in the Epi-Scale model described by Eqn.

(5) was calibrated (Table 3) such that the number of cells in time as the tissue grows

follows the experimental data [38] (Fig. 4e). Cells in mitosis deviate from cells in

interphase in their area and roundness. The area and roundness of interphase and

mitotic cells are calibrated based on the data from the wing disc (Fig. 5e-f). SI-S2.2

provides additional details about the methods used in this study to measure the size

and roundness of cells in the imaginal wing disc pouch.

Tissue topology emerges from cell self-organization driven by cellular mechanics

After model calibration, simulations were run to determine whether this cellular-scale

calibration was sufficient to recapitulate topological properties of the tissue (Fig. 6)

[53,54]. One metric for tissue topology is the distribution of cell neighbor numbers, or

polygon class distribution. Based on the simulation results for studying the tissue growth

by using the Epi-Scale model, the polygon class distribution approaches steady state

after 35 hours (Fig. 5a, 6b). This distribution matches with the ones reported

experimentally for the wing disc and other epithelial systems [32] (Fig. 6d) as well as

obtained using other computational models for simulating growing tissues such as

vertex based model [22].

Another way to quantify tissue topology is through evaluation of the three laws

describing topological relationships: Euler's law, Lewis law, and Aboav-Weaire Law.

Euler's law states that, on average, cells forming a packed sheet should be hexagonal

[54]. The Lewis law states that cells with more neighbors should have a larger

normalized area [55]. The Aboav-Weaire law states that the average polygon class of

each cell's neighbors decreases as the cell's polygon class increases [56]. Simulation

results obtained using the calibrated model show the average side of cells to be equal

to 5.96 which is in a very good agreement with the Euler's law. The model simulations

also satisfy two other laws as shown in Fig. 6c.

Impacts of adhesion, stiffness, and cytoplasmic pressure on MR

The Epi-scale model is suitable for generating and testing new hypothesis regarding

mechanical mechanisms of the MR because it is capable of representing non-polygonal

shapes of cells, and parameters representing mechanical cell properties in the model

can be directly related to the properties of cells measured in experiments. Simulations

were conducted to predict relative contributions of different cell properties to the relative

size (A_{Mit}/A_{Inter}) and roundness (R) of mitotic cells. Cell-cell adhesion, cortical

stiffness and internal pressure were individually varied in two sets of simulations (Table

4) to decouple the effects of these properties on the final size and shape of mitotic cells.

In the first set (Fig. 6, blue lines), only one of the three properties of mitotic cells is

varied in each simulation, while the other two properties are kept constant and equal to

interphase values. In the second set of simulations (Fig. 6, black lines), two of the three

parameters are set to calibrated values for mitosis. The third parameter under

investigation is varied with respect to its calibrated mitotic value. The extent of variation

of each parameter ranges from 100% below to 100% higher the calibrated value. Each

of these parameter sweeps can be interpreted conceptually as changing the relative

degree to which a mitotic cell is regulating each property during the mitotic phase of the

cell cycle.

Fig. 7a-c shows that variation of each of these three cell properties considerably affects

the roundness of mitotic cells. As seen in Fig. 7a, decreasing only cell-cell adhesion

during the M phase increases the roundness of mitotic cells. Increasing the cortical

stiffness (Fig. 7b) and the internal pressure (Fig. 7c) will increase the roundness of

mitotic cells. The levels of internal pressure needed to achieve "wild type" values of

roundness result in unphysical levels of cell areas (SI Movie S4.2) with noticeable

incidence of cell-cell rearrangements. Reducing adhesion leads to rounder cells, but

only regulating adhesion levels during mitosis would require a complete loss of

adhesion (100% below calibrated values) to reach wild-type levels of roundness.

Fig. 7d-f show normalized cross-sectional areas of mitotic cells (A_{Mit}/A_{Inter}) as cell-cell

adhesion, cortical stiffness, and the internal pressure are varied. Varying the adhesivity

of mitotic cells results in negligible impact on the area of mitotic cells as is evidenced by

a flat line (Fig. 7d, blue line). Increasing only the cortical stiffness (Fig. 7e, blue line)

slightly decreases the size of mitotic cells, while increasing cytoplasmic pressure

significantly increases it (Fig. 7f, blue line). Table 4 provides detailed information on the

impact of variation of different cell properties on mitotic cell's roundness and ratio of

maximum cross-sectional area during mitosis to area at interphase. The schematic

diagram in Fig. 8 recaps the results shown in Fig. 7 and represents the impact of

variation of each individual mechanical property on the final size and roundness of

mitotic cells. For a cell to be both large and round requires not only modulating internal

pressure but also reducing both adhesion and increasing stiffness co-currently.

In summary, the obtained results suggest that increase in cross-sectional area of mitotic

cells is solely driven by increasing cytoplasmic pressure. Mitotic roundness however is

not achieved within biological constraints unless all three properties (cell-cell adhesion,

cortical stiffness and pressure) are simultaneously regulated by the cell. Without

concurrent regulation reducing cell-cell adhesion and increasing cortical stiffness,

unrealistic high levels of pressure increase would be required to enforce mitotic

roundness, resulting in unphysical levels of cell areas.

Discussion

General models for investigating epithelial mechanotransduction, including MR, require

coupling of biologically calibrated mechanical components capable of representing non-

polygonal cell shapes, and simulating the membranes as well as cytoplasm of individual

cells as separate entities. To accomplish this, a novel multi-scale sub-cellular model,

called Epi-scale, was developed in this paper for simulating mechanical and adhesive

properties of cells in the developing columnar epithelium of the wing disc which consists

of a single layer of cells. The model approximates the tissue as a 2D surface since the

majority of the contractile and adhesive forces are localized at the apical surface of the

epithelium. Parameter ranges for the computational model were obtained by calibrating

the model using single cell stretching experiments, double cell stretching experiments,

as well as experimentally observed size distributions of cells during mitosis and

interphase, and tissue growth rate of *the Drosophila* wing disc. Cell-cell adhesion and

cell elasticity were calibrated using experimental data on epithelial tissues. The

calibrated model was tested by successfully reproducing emergent properties of

developing tissue such as the polygon class distributions.

Epi-scale enables one to produce new hypotheses about the underlying biophysical

mechanisms governing mitotic rounding of epithelial cells within the developing tissue

micro-environment. In particular, results of the model simulations predict that robust

mitotic rounding requires co-current changes in cell-cell adhesion, cortical stiffness and

cytoplasmic pressure. Regulating only one of the cellular properties does not result in

the experimentally observed levels of apical areal expansion and degree of mitotic

roundness. The individual contributions of changes in these three mechanical properties

to the mitotic cell roundness and area were characterized through detailed parameter

sensitivity analysis (Fig. 7).

Changes in all three cell properties were shown to contribute to the degree of

roundness *in vivo*, but internal pressure was shown to be the primary driver of mitotic

cell area increase. Cell properties can be modulated experimentally in tissue as a

whole. However, it is currently challenging to target only dividing cells in a tissue. Cell-

cell adhesion is dictated by the adhesive interactions of AJs, which can be modulated

pharmacokinetically, or through genetic modification of E-cadherin molecules to alter

their binding affinities. Cell stiffness can be adjusted by reducing the contractility of the

cortex through pharmacological perturbations. Internal pressure of cells is primarily

dictated by osmotic channels regulating the flow of water and ions through the cell

membrane, and can be adjusted by modulating those channels, or by changing the

osmolarity of the media.

The computational model simulations provide insight into the individual contributions of

cell properties to MR and can predict the consequences of dysregulation of mitotic cell

rounding on the development and homeostasis of epithelial tissues. One experimental

approach that could in future be used for testing the model predictions would be to

regulate the expression of E-Cadherin, Myosin-II, and osmotic channel antagonists

under a Cyclin B promotor, active during mitosis, resulting in modulation only in dividing

cells [57,58]. Alternatively, opto-genetic methods could be employed to selectively

regulate individual cell properties [59].

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Fig. 1. Epithelial mechanics. (a) Subapical surface of epithelial cells within the *Drosophila* wing imaginal disc that are marked by E-cadherin tagged with fluorescent GFP (DE-cadherin::GFP). Multiple cells within the displayed region are undergoing mitotic rounding. (b) Experimental and cartoon abstraction of epithelial cells, which are polarized with apical and basal sides. Actomyosin and mechanical forces during mitotic rounding are primarily localized near the apical surface. (c) At the molecular scale, the boundary between cells consists of a lipid bilayer membrane for each cell, E-cadherin molecules that bind to each other through homophilic interactions, and adaptor proteins that connect the adhesion complexes to an underlying actomyosin cortex that provides tensile forces along the rim of apical areas of cells. Arrows indicate mitotic cells. Scale bars are 10 micrometers.

Fig. 2.Diagram of the underlying physical basis of model simulations. (a) Intracellular and intercellular interactions between different elements of the model. (b) Implementation of the simulation of cell cycle in the model.

Fig. 3. Initial conditions and sample simulation output. (a) Initial condition of a simulation with seven initially non-adherent circular cells. Each cell starts with 100 membrane elements and 20 internal elements. (b) Initial formation of an epithelial sheet after cells adhere to each other. An equilibrium distribution of internal nodes is reached for each cell. (c) Epithelial sheet after 55 simulated hours of proliferation. (d) Enlarged view of

625   the selected region showing different cell shapes and sizes due to interactions between

626   cells. The large cell is undergoing mitotic rounding (MR). Arrows indicate mitotic cells.

Fig. 4: Calibration of model parameters through simulations. (a-a″) Calibration test to determine parameters for cell elasticity, analogous to experimental single cell stretching tests [48], (a) Initial condition t=0, (a′) 6 minutes after simulation with no force applied, (a″) after 72 minutes cell is completely on tension (b-b″) Cell adhesivity test, analogous to experimental tests [52] for calibrating the level of cell-cell adhesion between adjacent cells. (b) Initial condition t=0, (b′) 6 minutes after simulation begins with no force applied, (b″) after 72 minutes, 15 nN force is applied. (c) Stress versus strain for single cell calibration (red line) and stress versus strain for calibrating the level of adhesivity between the two cells (blue line)[51,52]. Initial negative strain in adhesivity test is due to strong adhesion between two cells. (d) Force and strain as a function of time for adhesivity test. (e) Tissue growth rate calibration by comparing with the experimental data by Wartlick et al. [38].

Fig. 5: (a-a''') Time-lapse confocal images of cell undergoing mitosis in the wing disc with E-Cadherin:GFP-labeled cell boundaries. Scale bar is 5 µm. Arrows indicate daughter cells. (b-d''') Time series from Epi-Scale simulation of a cell undergoing mitosis and division with illustration of: (b) adhesive spring stiffness, (c) cortical spring stiffness, and (d) internal pressure, normalized to their interphase values. (e-f) Comparison of size and roundness of mitotic cells with experimental data for the *Drosophila* wing disc. Arrow represents mitotic cell.

Fig. 6. Emergence of tissue-level statics from model simulations. (a) Sample simulation output showing cells with different numbers of neighbors as different colors (corresponding to legend in b). (b) Average relative area ($A/\overline{A}$) and average polygon class of neighbors for cells of different polygon classes verifying that simulation results follow Lewis law and Aboav-Weaire law. (c) Simulations initiated from seven cells reaches steady-state polygon-class distribution after approximately 35 h of simulated time. (d) Comparison of polygon class distributions obtained by Epi-Scale model with various biological systems (data extracted from [60]) and a vertex based model by Farhadifar et al. [22].

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Fig. 7. Predicted sensitivity of MR to perturbations in cell mechanical properties. (a-c) Dependence of roundness (R) on mitotic parameters. Effects of (a) adhesivity between cells, (b) cortical stiffness, and (c) internal pressure on the roundness of the mitotic cells. (d-f) Predicted increase in area during mitosis (A_{Mit}/A_{Inter}). Effects of (d) adhesivity, (e) cortical stiffness, and (f) internal pressure on the roundness of the mitotic cells. The grey shaded region shows the calibrated wild type values for roundness and area ratio when all three parameters are varied (black circles) vs. when only the single parameter is varied (blue squares) during mitosis. P-values of t-tests between the means of cell roundness and area ratio values for single parameter variation and multi-parameter variation are: (a): p=0.068, (b): p =0.0046, (c): p = 0.14, (d): p = 8.7x10^{-6}, p = 3.18x10^{-15}, p= 0.876.

Roundness ↑
Area ratio ↓

Increasing cortical stiffness

Increasing cytoplasmic pressure

Roundness ↑
Area ratio ↑

Reducing adhesion

Actomyosin

E-cadherin

Cytoplasmic pressure

Roundness ↑
Area ratio —

Fig. 8. Schematic detailing the contribution of individual cellular mechanical properties on roundness and size of mitotic cells within constrained tissues as revealed by Epi-Scale simulations. This model extending the graphical model for mitotic rounding presented by Stewart et al.[10] to include the relative effects of each mechanical change in mitotic cells undergoing MR within the constraints of packed tissues.

678   **Table 1:** Potential energy functions in Epi-scale model

Potential function	Type of potential function	Biological concept
Internal-internal nodes (E^{II}),	Morse	Internal pressure
Membrane-internal nodes (E^{MI})	Morse	Keeps the cytoplasm inside the cell and applies pressure from the cell's cytoplasm to the cell's membrane
Membrane-membrane nodes of neighboring cells (E^{MMD}).	Morse	Prevent membranes of neighboring cells invading into the cell
Membrane-membrane nodes of neighboring cells (E^{adh}).	Linear spring	Adhesion between neighboring cells
Membrane-membrane nodes of the same cell (E^{MMS}).	Linear and torsional spring	Membrane and cortex stiffness of the cell

Table 2: Calibrated parameters in Epi-sale

Parameter	Interphase	Mitotic phase	Values during interphase & mitosis	Source
E^{II}	U_{Inter}^{II}	U_{Mit}^{II}	$0.488 \,\&\, 19 \; nN.\mu m$	Matched to experimental data (SI-Appendix S3)
	W_{Inter}^{II}	W_{Mit}^{II}	$0.146 \,\&\, 5.86 \; nN.\mu m$	
	ξ_{Inter}^{II}	ξ_{Mit}^{II}	$0.3125 \,\&\, 0.54 \; \mu m$	
	γ_{Inter}^{II}	γ_{Mit}^{II}	$1.25 \; \mu m$	
E^{MI}	U_{Inter}^{MI}	U_{Mit}^{MI}	$0.78 \,\&\, 3.81 \; nN.\mu m$	Boundary condition
	ξ_{Inter}^{MI}	ξ_{Mit}^{MI}	$0.125 \,\&\, 0.25 \; \mu m$	
E^{MMD}	U_{Inter}^{MMD}	U_{Mit}^{MMD}	$3.91 \; nN.\mu m$	Boundary condition
	W_{Inter}^{MMD}	W_{Mit}^{MMD}	$3.91 \; nN.\mu m$	
	ξ_{Inter}^{MMD}	ξ_{Mit}^{MMD}	$0.125 \; \mu m$	
	γ_{Inter}^{MMD}	γ_{Mit}^{MMD}	$1.5625 \; \mu m$	
E^{adh}	k_{Inter}^{adh}	k_{Mit}^{adh}	$20 \,\&\, 8 \; nN/\mu m$	[43,51,52]
	L_{adh}	L_{adh}	$0.2 \; \mu m$	
E^{MMS}	k_{Inter}^{St}	k_{Mit}^{St}	$200 \,\&\, 250 \; nN/\mu m$	[49,50]
	L_{Inter}^{St}	L_{Mit}^{St}	$0.06 \,\&\, 0.125 \; \mu m$	
	L_{Inter}^{RSt}	L_{Mit}^{RSt}	$6 \,\&\, 7 \; nN.\mu m/rad$	

682   ** Other Morse parameters are equal to zero.

684   **Table 3:** Tissue scale parameters in Epi-scale

Parameter	Value	Reference
η	$36\ nN.s/\mu m$	[38]
$g_{0_{min}}$ & $g_{0_{max}}$	2×10^{-3} & $4\times10^{-3}P/s$	[38]
k_g	$4\times10^{-4}\ 1/s$	[38]

686 **Table 4: Impact of adhesion, stiffness, and cytoplasmic pressure on MR**

Fig. panels	Adhesion in Mitotic phase	Cortical stiffness in M phase	Pressure in M phase	Total variation of roundness*	Total variation of A_{Mit}/A_{Inter}*
Fig. 7a, 7d (Blue line)	varied	same as interphase	same as interphase	-0.42	0.04
Fig. 7b, 7e (Blue line)	same as interphase	varied	same as interphase	0.41	-0.26
Fig. 7c, 7f (Blue line)	same as interphase	same as interphase	varied	0.58	4.69
Fig. 7a, 7d (Black line)	varied	same as wild type	same as wild type	-0.32	-0.95
Fig. 7b, 7e (Black line)	same as wild type	varied	same as wild type	0.59	-0.60
Fig. 7c, 7f (Black line)	same as wild type	same as wild type	varied	0.18	4.69

689 *These values are reported based on varying the desired parameter from 100% below the

690 calibrated value to 100% above the calibrated value