Structural Redundancy in Supracellular Actomyosin Networks Enables Robust Tissue Folding

Graphical Abstract

Highlights

- A supracellular myosin network links ventral cells in the folding Drosophila embryo
- Many more connections are formed than are minimally needed to fold
- Network redundancy protects tissue-scale connectivity under local damage
- Anterior-posterior network stiffness directs a robustly oriented fold

Authors

Hannah G. Yevick, Pearson W. Miller, Jörn Dunkel, Adam C. Martin

Correspondence

acmartin@mit.edu

In Brief

Development is robust, but how cells mechanically ensure correct shape change has been unknown. Yevick et al. show that a supracellular network of myosin connections encodes intrinsic robustness into the folding Drosophila embryo. Redundancy and oriented stiffness in network connections guarantee correct folding under a wide range of perturbations.

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Structural Redundancy in Supracellular Actomyosin Networks Enables Robust Tissue Folding

Hannah G. Yevick, 1 Pearson W. Miller, 2,3 Jörn Dunkel, 3 and Adam C. Martin 1,4,*

1Department of Biology, Massachusetts Institute of Technology, Cambridge, MA 02142, USA
2Department of Physics, Massachusetts Institute of Technology, Cambridge, MA 02142, USA
3Department of Mathematics, Massachusetts Institute of Technology, Cambridge, MA 02142, USA
4Lead Contact
*Correspondence: acmartin@mit.edu
https://doi.org/10.1016/j.devcel.2019.06.015

SUMMARY

Tissue morphogenesis is strikingly robust. Yet, how tissues are sculpted under challenging conditions is unknown. Here, we combined network analysis, experimental perturbations, and computational modeling to determine how network connectivity between hundreds of contractile cells on the ventral side of the Drosophila embryo ensures robust tissue folding. We identified two network properties that mechanically promote robustness. First, redundant supracellular cytoskeletal network paths ensure global connectivity, even with network degradation. By forming many more connections than are required, morphogenesis is not disrupted by local network damage, analogous to the way redundancy guarantees the large-scale function of vasculature and transport networks. Second, directional stiffening of edges oriented orthogonal to the folding axis promotes furrow formation at lower contractility levels. Structural redundancy and directional network stiffening ensure robust tissue folding with proper orientation.

INTRODUCTION

A hallmark of embryonic development is that it is robust. To achieve a stereotypic outcome, morphogenesis must be resistant to perturbations caused by small genetic and environmental variations. Progress has been made in understanding how the embryo can use genetic or molecular strategies to overcome perturbations (Félix and Barkoulas, 2015; Whitacre, 2012). For example, redundancy in gene regulatory elements, such as shadow enhancers, promotes more robust gene regulation during development (Frankel et al., 2010; Hong et al., 2008; Perry et al., 2010). Redundancy in protein function or chaperone activity ensures correct physiological outcomes (Burga et al., 2011; Siegal and Rushlow, 2012; Zheng et al., 2013). Although these studies provided important insights into how robustness is regulated at the cell level, how robustness is achieved for populations of physically interacting cells is unknown.

Tissue contraction is a key mode of morphogenesis where force is propagated across distances that are far longer than the cell length-scale (Fernandez-Gonzalez et al., 2009; Galea et al., 2017; Hutson et al., 2003; Martin et al., 2010; Varner and Taber, 2012). Tissue contraction can bring opposing cell sheets together, a mechanism used in both development and wound healing (Davidson et al., 2002; Kiehart et al., 2000). In addition, the selective contraction or expansion of one surface on an epithelial sheet (i.e., apical or basal) can result in tissue folding and cell invagination (Gutzman et al., 2008; Heer et al., 2017; Krueger et al., 2018; Polyakov et al., 2014; Sui et al., 2018). In several cases, where there is tissue contraction, a supracellular actomyosin network is present across the field of cells (Hannezo et al., 2015; Martin et al., 2010; Nishimura et al., 2012; Röper, 2013; Skoglund et al., 2008). The function of supracellular actomyosin networks in morphogenesis is poorly understood. For example, do cells connect equally with all neighbors or can some cells take on more important roles in connecting parts of the tissue? How does the pattern of mechanical interactions in a tissue enable robust morphogenesis?

Ascertaining the network of cell interactions that enables a tissue to change shape is critical to understand morphogenesis. A limitation of past studies was the lack of methods to analyze both the cellular and supracellular structure of actomyosin networks. Here, we adapted an algorithm, originally developed to identify filamentous structures in the universe (Sousbie et al., 2011), to trace the filamentous structure of the supracellular actomyosin network during the folding of the Drosophila ventral furrow. Drosophila ventral furrow formation is an essential developmental step, which occurs when the embryo consists of a single layer of cells surrounding a central yolk. At this stage of development, ventral cells undergo apical constriction, which causes monolayer folding and internalizes the ventral cells (Leptin and Grunewald, 1990; Sweeton et al., 1991). The furrow or fold is reproducibly oriented along the anterior-posterior (a-p) axis, which is the direction of highest tension (Chanet et al., 2017; Martin et al., 2010). Analysis of wild-type and experimentally degraded networks coupled with computational modeling identified two mechanisms that make tissue folding robust. First, redundancy in cytoskeletal connections creates multiple mechanical paths spanning the ventral tissue, which ensures at least one path across the tissue when there is local damage. Second, stiffening of network connections along the a-p axis enables anteroposterior-oriented furrow formation, even at lowered contractility levels.

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RESULTS

Ventral Furrow Formation Is Robust to Tissue Damage in the Absence of Wound Healing

To investigate whether tissue folding is robust to cell and tissue damage, we ablated specific numbers of cells and determined whether the tissue could fold. We systematically injured embryos (4–32 cells) using high-intensity, two-photon laser excitation in cells prior to myosin II (myosin) accumulation and apical constriction. These targeted ablations created patches of damaged cells that did not apically constrict (Figures 1A and 1C). In contrast to other stages in Drosophila development (Fernandez-Gonzalez and Zallen, 2013; Rodriguez-Diaz et al., 2008), wounding prior to apical constriction (i.e., late cellularization stage) did not result in an actomyosin purse-string or rapid wound closure (Figures 1C and 1E; Video S1). Instead, the wound increased in area as surrounding cells began to recruit myosin and constrict. In addition, during the folding process, the wound anisotropy grew, suggesting that surrounding contractile cells stretched the wound (Figure 1F). Tracking tissue flow toward the midline at the time of folding (t = 0) slows with increased tissue damage, but all embryos still fold. N = 1 embryo for each of the 4 wound sizes.

Supracellular “Mechanical Paths” Connect Hundreds of Cells in the Ventral Furrow

The striking robustness of the folding process led us to investigate mechanisms that could produce this mechanical resilience. We hypothesized that patterns of linkages between individual force-generating cells establishes robust folding. In the Drosophila ventral furrow, mechanical linkages are associated with supracellular, apical actomyosin fibers that connect between cells through...
Supracellular actomyosin networks have not been rigorously analyzed because of the difficulty of defining subcellular structure in noisy tissue-scale images (Figures 2B and 2C). To determine this global network structure, we adapted a topological structure identification algorithm to map the filamentous supracellular myosin network in embryos depleted of an essential adherens junction component, $\alpha$-catenin. As expected, $\alpha$-catenin-depleted embryos exhibited a significant reduction of the network length which is defined as the sum of the length of all the edges in the network. Scale bars, 20 $\mu$m.

**Figure 2. Developing a Method to Trace the Supracellular Actomyosin Network**

(A) Myosin is predominantly activated apically. It interacts with actin to form actomyosin structures across the apical surface of cells. These fibers connect between cells in an end-on manner at spot adherens junctions.

(B and C) Z-projection of myosin (green, sqh::GFP) (B), and subapical slice of membrane (Gap43::mCherry) channel at the time of folding initiation (C). Myosin forms a network across the entire apical surface of the ventral tissue just prior to folding.

(D) Apical myosin intensity (blue-red heatmap) was traced using the DisPerSE algorithm (Sousbie, 2011). The trace follows the topology of the intensity signal and forms a network (magenta) where nodes correspond to local maxima and edges correspond to the topological ridges between these peaks. The myosin network (magenta) does not correspond to the cell membranes (grayscale).

(E) The trace of the supracellular myosin network (magenta) across hundreds of cells in the ventral tissue.

(F) Myosin network trace (magenta) in an $\alpha$-catenin-RNAi embryo which has depleted cell-cell junctions. Almost all the network connections are severed. Myosin (Sqh::GFP) is labeled in green.

(G) $\alpha$-catenin-RNAi embryos exhibited a significant reduction of the network length which is defined as the sum of the length of all the edges in the network. Scale bars, 20 $\mu$m.
with the gradient in node density along the dorsal-ventral axis growing over time (Figure 3B). A dorsal-ventral gradient in the probability of having a myosin node is consistent with there being a multicellular gradient in average apical myosin activity (Heer et al., 2017). While the network increased in size, individual myosin nodes existed transiently, exhibiting a mean lifetime of 100–150 s (Figure 3C).

To determine if these node dynamics reflected myosin turnover, we examined node lifetime in a myosin light chain mutant, the sqh-AE mutant. This mutant served as further control for our DisPerSE method because the difference in myosin dynamics in this mutant is well characterized (Vasquez et al., 2014). The sqh-AE mutant disrupts myosin turnover, likely because it uncouples myosin regulation from dynamic upstream signals and because it compromises motor activity (Vasquez et al., 2016). A supracellular myosin network is still able to form in this tissue, and the tissue folds albeit more slowly because of reduced motor activity (Vasquez et al., 2016). Myosin node lifetime was significantly increased in sqh-AE mutant embryos, suggesting that the myosin node dynamics we observed in the network are due to cytoskeletal and signaling dynamics (Figures 3C, S2, and S2B). In summary, the supracellular myosin network maintains global connectivity while individual network nodes and, therefore, their corresponding connections are dynamic.

**Supracellular “Mechanical Paths” Are Redundant in the Wild-Type Actomyosin Network**

To determine how the global network connectivity couples distinct regions of the tissue, we sought to measure the relative importance of nodes in connecting the network. To identify nodes that heavily support the overall network connectivity, studies of network resilience in subway systems (Derrible, 2012), airports (Guimerà et al., 2005), and roads (Kemanshah and Derrible, 2017; Lämmer et al., 2006) have used the betweenness centrality, $C_B$, which is calculated for each node. Node $b$ (red) has the highest normalized betweenness as it is the most important node in connecting other sets of nodes.

![Image](https://example.com/image.png)

**Figure 3. Nodes in the Supracellular Myosin Network Become Increasingly Redundant during Ventral Furrow Formation**

(A) The number of network nodes grows leading up to folding ($t = 0$). Plotted is the number of nodes detected in an embryo as a function of developmental time; the different colors represent different embryos. Trend line (black) represents mean ± SD. Each color represents a different embryo. $N = 6$ embryos.

(B) Node density is present in a ventral-dorsal gradient. The highest node density is at the ventral midline ($0$ on x axis) with node density decreasing away from the ventral midline. The gradient increases in magnitude leading up to folding.

(C) Node lifetime depends on myosin regulation. Plot shows the mean ± SD of node lifetimes for wild-type or sqh-AE mutant tracked nodes. Pulsed cycles myosin accumulation and disassembly normally observed are abolished in the sqh-AE mutant.

(D) An example network and the measure of normalized betweenness centrality, $C_b'$, which is calculated for each node. $C_b'$ measures the fraction of times a node is on the shortest path between any other two nodes (Freeman, 1978). To ensure our calculated “betweenness” did not simply reflect network size, we normalized the measurement to the total number of paths, $C_B$ (Figure 3D;STAR Methods). Therefore, nodes with higher “betweenness” are more critical in connecting regions of the tissue.

We measured the normalized “betweenness” of all nodes in the network of 5 wild-type embryos (Figures 3E and S3) at each developmental time. For all embryos, the mean node “betweenness” decreased to a consistently low value at which point the tissue folded (Figure 3F). Therefore, individual nodes in the wild-type network become increasingly redundant (i.e., less important in connecting parts of the tissue) leading up to folding.

**Network Degradation Decreases Redundancy but Does Not Prevent Folding**

To test the function of redundancy in folding, we degraded the network using three different perturbations and assessed network structure. First, we examined the network structure of laser-ablated embryos (Figure 1C). Second, we overexpressed a constitutively active form of the myosin binding subunit (MBS$^{CA}$, which we refer to as MBS$^{CA}$) of myosin phosphatase...
to downregulate myosin activity (Figure 4A). In contrast to laser ablation, MBS<sup>CA</sup> overexpression resulted in a more uniform degradation of the supracellular myosin network; clusters of cells failed to accumulate apical myosin and constrict (Figure 4B). Finally, we examined cta mutants that disrupt signaling upstream of RhoA and cause uncoordinated apical constriction (Costa et al., 1994; Parks and Wieschaus, 1991; Xie et al., 2016). Myosin degradation reduced, but did not abolish, the gradient in node density along the ventral-dorsal axis (Figure S5 A). Similar to our results with laser ablation (Figures 1 C and 1D), folding was observed even when MBS<sup>CA</sup> expression caused a 65% reduction in the number of nodes across the tissue (Video S3). Thus, our results with laser ablation and MBS<sup>CA</sup> expression demonstrated that tissue folding is robust to network degradation.

Degradation of the supracellular actomyosin network increased the relative importance of individual network nodes in connecting the tissue, but did not completely disconnect components of the network (Figure 4C). The cumulative distribution function of “betweenness” values in wild-type, laser ablation, and MBS<sup>CA</sup> overexpression conditions at the time of folding demonstrated that nodes in the network have higher normalized “betweenness” values upon network degradation (Figures 4C and 4D). These results suggested that many more paths are present in the wild-type supracellular myosin network than are needed to fold the tissue.
Network redundancy in the supracellular actomyosin network ensures a minimal path across the tissue, even when there is damage (Figure 4E).

To further test the robustness of tissue folding, we analyzed concertina (cta) mutant embryos, which result in uncoordinated apical constriction and cell size heterogeneity (Parks and Wieschaus, 1991; Xie et al., 2016). The cta mutant embryos exhibited a reduced myosin network to 52% of the WT length, but the embryos were also still able to fold (Figures S4A and S4B).

We hypothesized that rigidity in the network could promote collective deformation of the tissue. Maxwell first described the isostatic point (Maxwell, 1864), which is the minimal average node degree in a network needed to achieve mechanical rigidity (i.e., infinite stiffness) (Figure 5A). Below the isostatic point, “floppy modes” exist, which are collective degrees of freedom that are not subject to a restoring force (Figure 5B). Adding sufficient connections between nodes, as is the case for a truss in a bridge, rigidifies the structure.

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To determine whether wild-type supracellular myosin networks achieve isostacity, we measured the degree of each node across the network, which represents the number of edges which connect to it. N = 6 embryos, 1,361 nodes.

The degree distribution of the nodes for wild-type networks at the time of folding. The degree of a node degree is defined as the number of edges which connect to it. N = 6 embryos, 1,361 nodes.

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Wild-Type Supracellular Networks Exhibit Connectivity below the Isostatic Limit

Because we showed that the ventral furrow folds even with decreased network connectivity, we investigated what properties of the remaining connections enabled robust folding. We hypothesized that rigidity in the network could promote collective deformation of the tissue. Maxwell first described the isostatic point (Maxwell, 1864), which is the minimal average node degree in a network needed to achieve mechanical rigidity (i.e., infinite stiffness) (Figure 5A). Below the isostatic point, “floppy modes” exist, which are collective degrees of freedom that are not subject to a restoring force (Figure 5B). Adding sufficient connections between nodes, as is the case for a truss in a bridge, rigidifies the structure.

To determine whether wild-type supracellular myosin networks achieve isostacity, we measured the degree of each node across the network, which represents the number of edges emanating from it. For passive networks formed from spring-like forces between nodes, a rigid structure is achieved in two dimensions in the large network limit when the average node degree is $\zeta \geq 4$ (Maxwell, 1864). The average node degree in the wild-type supracellular actomyosin networks was low at the time of folding ($\zeta = 2.38 \pm 0.2$; Figures 5C and 5D), and it did not increase during the folding process. MBS<sup>CA</sup> networks were less connected than the WT with an even lower average node degree (Figure 5D). Thus, the density of edges in the
Figure 6. Directional Network Stiffness Promotes Robust Furrow Formation

(A) To model ventral furrow formation, a stiff myosin network (magenta) was superimposed onto the side of an elastic shell (gray), which also contained a myosin contractility gradient (green). The insets show the effect of different fiber orientations (d-v, blue; a-p, yellow) on preferred curvature in the model.

(B) The a-p oriented stiff fibers accelerate folding in silico. Starting from an unfolded configuration, $M_0$ was slowly increased from 0 to 15, triggering folding. The depth of the furrow at the midpoint of the a-p axis was compared for three types of simulations: without a network (orange), with a network containing edges preferentially directed parallel to the a-p axis (yellow), and containing edges preferentially parallel to the d-v axis (blue).

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supracellular actomyosin network does not reach the isostatic point in either of these conditions, suggesting that connectivity controlled rigidity does not explain how the tissue folds.

**Folding Is Associated with the Stiffening of a-p-Oriented Connections**

Studies of filament networks have shown that sub-isostatic networks can stiffen under strain (Sharma et al., 2016). A signature of a network under strain is edge straightening, which we quantified using a measure called tortuosity. The tortuosity of an edge is the ratio of the arc length of the edge to the distance between interconnected nodes (Figure 5E). A value of 1 corresponds to an edge that is perfectly straight. We found that folding was associated with a preferential straightening of anterior-posterior (a-p) oriented edges. While a-p and dorsal-ventral (d-v) oriented edges began with the same tortuosity, over time edges oriented along the a-p axis (angle $0^\circ$–$30^\circ$) decreased their tortuosity more significantly than those oriented in the d-v direction (Figure 5F).

This bias occurred in both wild-type and MBS$^{CA}$ embryos (Figures 5G and S5B–S5E). The straightening of a-p edges is consistent with the higher level of tissue tension along this axis (Chen et al., 2017). A preferential straightening along the a-p axis was also observed in the cta mutants (Figure S4C). Our finding shows that this tissue-level property is ensured by connections between cells becoming preferentially taut along the a-p axis.

To determine the relationship between network strain and folding, we measured mean tortuosity as a function of time for all network edges. In the wild-type embryo, mean tortuosity decreased prior to folding initiation (Figure 5H). Interestingly, despite lower myosin levels, folding for the MBS$^{CA}$ overexpressing embryos occurred at the same mean tortuosity for the myosin network as the edges in wild-type networks (Figure 5H; WT and MBS$^{CA}$ compared by two-sample KS test, $p = 0.2$), suggesting that the increase in a-p stiffness is correlated with tissue folding. Therefore, folding is not associated with increasing network connectivity, but is instead closely associated with the onset of directional network stiffening.

**Stiff a-p Edges Promote Robust a-p Furrow Formation at Low Myosin Levels**

The directional bias in taut fibers along the a-p direction does not result in preferential area constriction in the a-p direction. Instead, constriction occurs predominantly in the d-v direction (Heer et al., 2017; Martin et al., 2010). Therefore, to test how directionally biased (a-p) stiffness in a supracellular actomyosin network promotes d-v bending of the embryonic tissue, we employed a 3D continuum elastic shell model of the *Drosophila* embryo. Folding of the model embryo was driven by a gradient of myosin contractility applied to the shell surface, which is present in the embryo (Figure 6A) (Heer et al., 2017). Myosin contractility induces folding by changing the local preferred curvature of the shell (Heer et al., 2017) (STAR Methods).

In addition to the myosin gradient, we modeled the supracellular actomyosin network by imposing a network of stiff elements on the surface. Because the edges of the network represent tense fibers connecting discrete points, we represented the mechanical effect of an edge by imposing an additional cost for the tissue to bend along the edge’s length (Figure 6A). To make a general theoretical prediction as to how the network influences tissue folding, we made the network 10$x$ stiffer than the background tissue. The trends in our model were confirmed by also implementing it with stiffnesses that were 4$x$ and 8$x$ the background.

First, we explored how the presence of a network influences folding. Starting with a level of myosin contractility that was just enough to generate a furrow, we simulated folding events with lower contractility levels and determined the effect of having a rigid network. For low myosin contractility levels, rather than forming an oriented a-p furrow, depressions were formed at the end of the contractile domain (Figure 6C; $M_0 = 3$). However, imposing a rigid network resulted in furrow formation at contractility levels that normally would not result in an a-p furrow (Figure 6C; $M_0 = 4$). Therefore, the presence of a stiff network decreases the amount of contractility needed to form a furrow, making it more robust.

To explore how the directionality of network connections influences folding, we compared folding without a stiff network, to networks with stiff edges exhibiting preferred orientations along either the a-p or d-v axis. For a given network size, we found that folding speeds were the highest when network edges were preferentially oriented along the long axis of the shell when compared to the case where there was only contractility and no structural network (Figure 6B, yellow versus blue line). In contrast, networks where edges were preferentially oriented along the d-v direction folded slower than both the wild-type network and the a-p biased network (Figure 6B, orange line). This result suggested that the structural properties of the supracellular actomyosin network promotes robust tissue folding and that the a-p connections in the network are most important for this effect.

Because the model predicted an important role of a-p oriented edges, we investigated whether there was a bias to network edges in embryos. In wild-type embryos, the myosin edges were slightly biased toward alignment along the a-p axis (Figures 6D and 6F). We then examined this bias in embryos overexpressing MBS$^{CA}$. MBS$^{CA}$ decreases myosin levels and the number of myosin nodes, but the remaining nodes were still distributed in a gradient from the ventral midline (Figure S5A). Interestingly, MBS$^{CA}$ exhibited a greater bias in edge orientation (p value $1.5 \times 10^{-12}$ two-sample KS test), preferentially retaining a-p edges (Figures 6E and 6G). The cta mutant embryos and...
embryos damaged via laser ablation also exhibited a bias in edge orientation similar to WT embryos (Figures 6H and 6I). It appears that MBS<sup>CA</sup> embryos preferentially preserve the a-p oriented connections, which we argue is an important mechanism for robustness. It is unclear why d-v edges are preferentially degraded, however, one possibility is that the greater tension and straightness of these edges makes them resistant to degradation (Fernandez-Gonzalez et al., 2009). Our model predicted that stiff edges oriented along the furrow axis increases folding efficiency and enables folding at lower motor activity. Our analysis of the network in MBS<sup>CA</sup> overexpressing embryos highlights the importance of a-p network connections. MBS<sup>CA</sup> embryos exhibited folding while predominantly having a-p connections, rather than d-v connections. Because contraction must happen in d-v for cells to obtain wedge shapes needed for the furrow shape, our results support a structural role for the supracellular actomyosin network, such as by providing directional tissue stiffness.

To test the origin and possible importance of the a-p bias in the network in folding we analyzed embryos depleted of Spn27A, a negative regulator of ventral cell fate whose depletion expands the contractile domain around the d-v axis and flattens the multi-cellular myosin gradient (Figure 7A) (Chanet et al., 2017; Heer et al., 2017). Spn27A depletion disrupted the a-p bias to the network (Figure 7D) at the time of folding. Strikingly, Spn27A depletion often caused folding to occur along the DV axis (11/19 embryos), which is orthogonal to the normal furrow orientation (Figures 7B and 7C). To investigate whether a-p oriented stiffness could promote correct folding orientation, we used our model to simulate Spn27A-depleted embryos. We adjusted our model to expand the contractile domain 2-fold with respect to the original model. We discovered that expanding the contractile domain around the embryo’s circumference and imposing a weak a-p myosin gradient, similar to that observed (Streichen et al., 2018), resulted in a perpendicular furrow (Figure 7E). This misoriented furrowing is prevented if an oriented myosin network is present (Figure 7F). Therefore, we conclude that an a-p oriented network of stiff mechanical elements also promotes robustness by preventing misoriented folding (Figure 7G).

**DISCUSSION**

In this study, we developed a strategy to trace supracellular actomyosin networks involved in tissue morphogenesis. This strategy will be generally applicable to further investigations of both collective supracellular behavior in tissues and network structure in individual cells. Using this new approach, we discovered: (1) Many more supracellular connections are present in wild-type supracellular actomyosin networks than are minimally needed to fold the tissue. This network redundancy makes tissue folding robust. (2) The supracellular network plays a passive, structural role in folding with stiff a-p edges enhancing folding at lower contractility levels. The network creates a “frame” that determines final tissue shape and orientation; being stiff along one axis and flexible along the orthogonal axis.

**Redundancy Is a General Strategy for Network Resilience**

Our analysis of the supracellular network revealed a striking feature of the supracellular actomyosin network—individual nodes in the network are not important to connect different parts of the tissue. We also demonstrated this redundancy experimentally, using both laser ablations, MBS<sup>CA</sup> overexpression, and cta mutants. In all three cases, we could significantly degrade the network, and tissue folding still occurred, arguing for redundancy in network connections. The presence of redundant paths across the tissue secured alternative spanning routes even upon the removal of some regions of the network. Therefore, the tissue can recover from damage and fold without wound healing.

Redundancy of tissue-scale connections is a novel mechanism to mechanically ensure robustness during embryonic development. Embryos employ a similar strategy to that used in many other types of networks that exhibit large-scale function upon loss of connections due to attack or random failure. For example, seminal work in the field of network science on network resilience carried out by the US Airforce described the use of redundancy as a means of building communication systems that can withstand heavy nuclear attacks (Baran, 1964). Well-constructed transportation networks incorporate redundancy to optimize transport through the network under local fluctuations in transport capabilities (Corson, 2010). The cerebral cortex relies on a highly interconnected network of communicating arterioles which can reroute flow to mitigate the effects of a single obstruction (Schaffer et al., 2006). Interestingly, while the myosin network in the ventral furrow is redundant, it is far from the optimal connectivity needed to maximize robustness under damage. Unlike the ventral furrow, scale-free networks, which have a power-law distribution of node degree (for example, the internet), are more resistant to being broken into two disconnected parts upon node removal. We speculate that the relatively low average node degree of the supracellular myosin network has optimized the need for folding robustness relative to the energetic cost of forming a large number of connections per node.

One challenge in our study was to measure the exact network degradation threshold past which folding does not occur. Percolation theory predicts that removing a small number of nodes has only a limited impact on the network’s integrity with a sudden network failure at a critical threshold (Barabási, 2016). We hypothesize that the existence of such a critical threshold, made it difficult to experimentally identify the minimal connectivity needed for folding using MBS<sup>CA</sup> overexpression. We either observed a network with reduced, yet less redundant a-p paths, or severe cellularization phenotypes and no network or folding. However, fully removing all a-p paths by severing the entire width of the constricting ventral domain via laser ablation did abolish folding (Chanet et al., 2017).

**The Supracellular Actomyosin Network Acts as a Structural Element that Guides Folding**

This work defines a new role for supracellular actomyosin networks by demonstrating that they can play a structural role (like a frame) that promotes robust tissue morphogenesis. The presence of stiff network edges on the embryo surface, especially a-p oriented edges, mechanically promotes furrow formation by changing the local bending energy of the tissue. We showed that an a-p orientation bias is present in the network for WT, under myosin degradation, laser ablation, and in cta mutants. Network edges preferentially straighten when oriented along the direction of the future furrow, and there are more network connections oriented along this axis. We modeled these...
Figure 7. An a-p Network Bias Suppresses Misoriented Folding

(A) Depleting the ventral fate inhibitor Spn27A expands the contractile domain circumferentially around the d-v axis. Images of a Spn27A-depleted embryo at the time of folding initiation with sub-apical membrane (Gap43::mCherry, grayscale), Z-projected myosin (Sqh::GFP, green), and the network trace (magenta).

(B) A furrow forms in the embryo along the dorsal-ventral axis at the time of folding initiation. Arrows indicate location of furrow.

(C) Folding occurs along the dorsal-ventral axis. Image was acquired 6.75 min after folding initiation.

(D) Network edges in Spn27A RNAi embryos do not exhibit the a-p directional bias seen in the WT. N = 3 embryos, 1,814 edges.

(E) The elastic shell model was adapted to model Spn27A-depleted embryos. The contractile domain was expanded 2-fold circumferentially. Increasing the strength in contractility resulted in a perpendicular furrow formation. Inset shows mid-sagittal planes through the shell.

(F) The misoriented d-v fold is prevented if an oriented myosin network is present on the surface of the elastic shell. Inset shows mid-sagittal planes through the shell.

(G) The presence of a stiff, oriented network (biased along the a-p direction, as seen in the WT) guides folding along the embryo’s long axis (a-p). Misoriented furrowing can improperly occur without this network orientation bias (i.e., Spn27A-depleted embryos), suggesting that it suppresses misaligned folding. Scale bars, 50 μm.
straight paths as stiff passive elements on an ellipsoid shell. Our physical model identified that these oriented connections promote furrowing along the correct axis at lower motor activity than is normally observed without the network. Additionally, our model demonstrates how oriented network edges can prevent misoriented furrow formation along the d-v axis, as seen in Spn27A-depleted embryos lacking an a-p network bias. We also showed that a-p oriented edges are preferentially retained in folding tissues under myosin degradation. The directionality of these edges is orthogonal to the axis of preferential constriction, suggesting they do not simply mediate contraction. Therefore, tissue-scale actomyosin assemblies can have a structural function in sculpting tissues, in addition to actively generating force needed to constrict cells. This ability of actomyosin to serve a structural role is reminiscent of supracellular actomyosin assembled at cell boundaries in some developing tissues (Monier et al., 2010; Umetsu et al., 2014). In these instances, tension directed along the boundary prevents the mixing of opposing groups of cells but does not promote cell or junction constriction. However, the tissue-scale actomyosin assemblies at boundaries do not trigger the 3D shape change.

**Encoded Robustness in the Absence of Wound Healing**

We report that robustness is encoded into morphogenetic movements even in the absence of a wound healing response, which is likely to be important for embryonic stages where such a response is not present. Wound healing is a well-described mechanism for tissue resiliency (Redd et al., 2004). During dorsal closure, for example, a purse string at the margin of two sheets of epidermis pulls and seals together the tissue (Kiehart et al., 2017). When the actomyosin purse string is damaged, cells assemble a secondary cable at the injury site via wound healing (Rodriguez-Diaz et al., 2008). While wound healing can keep an embryo on the correct developmental program, mechanical robustness encoded into the tissue can allow the tissue to adjust immediately to damage. Further study is needed to understand how inherent robustness mechanisms interact with wound healing to promote reproducibility in embryonic development. Exploring inherent mechanisms of robustness in other tissues undergoing morphogenesis has the potential to elucidate new ways to control and reprogram tissues both when treating disease and when striving to engineer reproducible tissue shape in vitro.

**STAR★METHODS**

Detailed methods are provided in the online version of this paper and include the following:

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**SUPPLEMENTAL INFORMATION**

Supplemental Information can be found online at https://doi.org/10.1016/j.devcel.2019.06.015.

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**AUTHOR CONTRIBUTIONS**


**DECLARATION OF INTERESTS**

The authors declare no competing interests.

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**REFERENCES**


STAR METHODS

KEY RESOURCES TABLE

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LEAD CONTACT AND MATERIALS AVAILABILITY

Further information and requests for resources and reagents should be directed to and will be fulfilled by the Lead Contact, Adam Martin (acmartin@mit.edu).

EXPERIMENTAL MODEL AND SUBJECT DETAILS

Drosophila Fly Stocks

All fly stocks were maintained at room temperature (~ 21°C). Fly crosses were maintained at 25°C in a humidified incubator. Embryos were collected on apple juice agar plates.

Wild-type embryos with marked membranes and myosin were obtained by crossing Gap43::mCherry/CyO; sqh::GFP flies to female virgin sqh^{AX3};sqh::GFP. Female progeny not expressing CyO were crossed to OregonR males. The resulting embryos were collected for imaging.

Sqh^{AE} phosphomutant sqh^{AE}::GFP was formed via substitutions of threonine-20 to alanine and serine-21 to glutamate as described in (Vasquez et al., 2014). Sqh^{AE} mutants were derived from sqh^{1} germline clones carrying sqh^{AE}::GFP transgene. Mutant/ovoD larve were placed at 37°C for 2 hr for 3–4 days to create a heat shock using the FLP-DFS method.

α-catenin-RNAi virgins of the shRNA line, y[1] sc[1] v[1]; P{y^{+t7.7} v^{+t1.8}=TRiP.HMS00317}attP2 were crossed with a maternal Gal4 driver line, mat67, sqh::GFP; mat15, gap43::mCherry. The mat67/+ sqh::GFP; mat15/+ gap43::mCherry-7/ P{y^{+t7.7} v^{+t1.8}=TRiP.HMS00317}attP2 females were mated with siblings and their progeny were examined.

The cta mutant embryos were formed by crossing cta^{RC10}; myosin::GFP, Gap43::mCherry/Tm3 virgin females with Df(2 L)PR31/CyO, myosin::GFP deficiency males.
Spn27A depleted embryos were formed by crossing virgin females from the Spn27A shRNA line y[1] sc[1] v[1]; P[y[+t1.8]=TRiP.HMC03159]attP2 with y,w; Sqh::GFP; mat15, Gap43::mCherry/(TM3, Sb[1]) males.

**METHOD DETAILS**

**Image Acquisition**

Embryos were dechorionated in 50% bleach, rinsed with water, and mounted with embryo glue (double-sided Scotch Tape glue dissolved in heptane) ventral side up on a slide. An imaging chamber was constructed with a #1.5 coverslip spacer and was filled with Halocarbon 27 oil (Sigma).

All images excluding the laser ablation experiments were acquired with a 40×/1.2 Apochromat water objective (Zeiss) on a Zeiss LSM 710 confocal microscope. Samples were illuminated with a 488-nm argon laser and a 561-nm diode laser.

Ablation images were acquired with a Zeiss LSM 710 NLO Laser Scanning Confocal and ablation was carried out with a Coherent Chameleon Ultra II femtosecond pulsed-IR laser and a 40×/1.1 objective.

**mRNA Injection**

Capped MBS mRNA was synthesized using the ~900 base pair N-terminal sequence of Myosin Binding Subunit fused to a glycine-serine linker and an HA tag. A mMESSAGE mMACHINE SP6 Transcription kit (Ambion) was used to generate mRNA. Embryos were dechorionated using same procedure as for live imaging and then desiccated for 5 min. The embryos were covered in a mixture of 75% halocarbon 700% and 25 percentage points halocarbon 27 (Sigma). The mRNA was injected into the lateral side of the embryo approximately 2.5 h before ventral furrow formation. The excessive injection oil was wicked off and spacer coverslips (No. 1.5) and a top coverside were added. The imaging chamber was filled with halocarbon 27 oil. Control embryos were injected with 0.1×PBS.

**Image Processing and Analysis**

Images were viewed with ImageJ to select embryos that were the best oriented (http://rsb.info.nih.gov/ij/). A multi-step process was used to project confocal z stacks of the myosin in MATLAB (MathWorks). The raw data stack was smoothed in 3D using box smoothing. The approximate location of the embryo’s surface was obtained using the height of the maximum intensity pixels at each position.

The resultant image was filtered using bandpass-filtering in the Fourier Domain to create a continuous surface. Myosin signal was averaged for 3 μm above the surface.

Velocity measurements were performed using PIVlab (Thielicke and Stamhuis, 2014) and subsequent analysis was done with custom MATLAB code.

The topological tracing of the myosin network was done on the projected 2D image with the DisPerSE package (Sousbie, 2011). A persistence cutoff of 5 was used to filter peaks in the signal of topological relevance. DisPerSE topologically smoothed the image and outputted the location of all the resultant critical points (local maxima, minima, saddle points) as well as the position of the edges connecting between these critical points. All of the data outputted was at subpixel accuracy. Spurious edges were removed by first removing the outputted DisPerSE trace with a mask of “active myosin” made using a thresholding technique described in (Heer et al., 2017). Edges that contained a minimum of a 75% overlap with the mask were retained (Figures S1B–S1E). Subsequent filtering the outputted DisPerSE trace with a mask of topological relevance.

Formally, the *betweenness* centrality for node u is defined as

\[
\text{CB}(u) = \frac{\sum_{s \neq t \neq u} n_{st}(u)}{\sum_{j} C_{B}(j)},
\]

which measures how often node u appears on a shortest path between nodes s and t. \(n_{st}(u)\) is the number of shortest paths from s to t and \(n_{st}(u)\) is the number of these paths that pass through u. We then normalize \(C_{B}(u)\) by the total *betweenness* of all nodes in the network \(C_{B}(u) = \frac{C_{B}(j)}{\sum_{j} C_{B}(j)}\). Betweenness of each node was calculated using customized codes written in MATLAB employing the built-in function “centrality.” Tortuosity of an edge was calculated by dividing the length along an edge by the distance between the edge’s two endpoints. The percent degradation of a network upon myosin phosphatase overexpression was defined by summing the total length of all the edges in the network divided by the average total length of all the edges in the WT data set.

**Computational Model**

To model the effect of supracellular myosin networks on tissue folding, we employed a continuum elastic shell model, approximating the tissue as an ellipsoid of revolution with major axis a and minor axis b = \(\frac{a}{2}\). Furrow formation is triggered by modification of the shell’s reference curvature as a function of a prescribed myosin-induced contractility gradient \(M = \frac{M_0}{(1 + e^{(x_0-y_0)/k})^2((1 + e^{k(x-y)})^2 + 1)}\) defined on the shell surface, with x and y the direction parallel and perpendicular to the a-p axis, respectively. In Figure 6, \(x_0 = y_0 = 0.2\) in Figure 6B and 0.34b in Figure 6C. In Figure 7, we expand the gradient to \(y_0 = 0.375b\) and a weak gradient is induced by setting the value of k in the x direction to 0.15. This model was adapted from (Heer et al., 2017), and full details of the
implementation can be found therein. For our present study, we made one substantial modification to this model by adding an extra term to the energy functional representing the myosin network:

\[
E_{\text{network}} = \frac{Y h^3}{24(1 - \nu^2)} \sum_i \int (1 + M) k_{\text{edge}} k_{ij}^2 d\sigma
\]

Here the sum is taken over all edges, where \( k_{ij} \) is the surface curvature along the tangent vector parallel to edge \( i \) at point \( \sigma \) (calculated by taking the projection of the mixed second fundamental tensor along the tangent vector’s direction), \( Y \) is Young’s modulus, \( \nu = 0.5 \) is the Poisson ratio, and \( h = 0.05b \) is shell thickness. The parameter \( k_{\text{edge}} \) is a dimensionless multiplier of the stiffness of each edge. In order to best represent the edge topology in a continuum model, we treat this stiffness as a scalar field across the surface given by \( k_{\text{edge}} = \sum_i k_{\text{max}} e^{-r_i^2/\sigma^2} \), where \( r_i \) is the distance of a given point on the surface from edge \( i \), and \( \sigma = 0.0125b \) defines the effective width of each edge. Each edge itself was represented as a straight line with start and end points defined directly from the experimental network trace. Directionally biased networks (Figure 6B) were constructed by starting with the wild-type network and removing the 50% of the edge most perpendicular to the chosen axis.

**QUANTIFICATION AND STATISTICAL ANALYSIS**

Statistical analysis was performed with the MATLAB statistics toolbox. \( p \) values were calculated against the null hypothesis using two-sample Kolmogorov–Smirnov test which is a nonparametric hypothesis test that the samples are unequal and from different continuous distributions. Statistical details of experiments can be found in figure legends. All error bars are +/− SD.

**DATA AND CODE AVAILABILITY**

Datasets generated during this study are available at Mendeley data [https://doi.org/10.17632/5269fdh6g8.1](https://doi.org/10.17632/5269fdh6g8.1).