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Scale-dependent sharpening of interfacial fluctuations in shape-based models of dense cellular sheets

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The properties of tissue interfaces – between separate populations of cells, or between a group of cells and its environment – has attracted intense theoretical, computational, and experimental study. Recent work on shape-based models inspired by dense epithelia have suggested a possible “topological sharpening” effect, by which four-fold vertices spatially coordinated along a cellular interface lead to a cusp-like restoring force acting on cells at the interface, which in turn greatly suppresses interfacial fluctuations. We revisit these interfacial fluctuations, focusing on the distinction between short length scale reduction of interfacial fluctuations and long length scale renormalized surface tension. To do this, we implement a spectrally resolved analysis of fluctuations over extremely long simulation times. This leads to more quantitative information on the topological sharpening effect, in which the degree of sharpening depends on the length scale over which it is measured. We compare our findings with a Brownian bridge model of the interface, and close by analyzing existing experimental data in support of the role of short-length-scale topological sharpening effects in real biological systems.

1 Introduction

Interfaces between populations of cells, or between cells and their environment, is pivotal for a wide variety of biological processes, ranging from embryonic development to wound healing to tumor metastasis to lineage sorting^{1–4}. The stability and structure of the interface unveils information of cell and tissue mechanics, and the corresponding pathological changes in certain diseases. In perhaps the most famous model for cell-cell interfaces, the differential adhesion hypothesis (DAH)^{5,6} distinct cell groups are treated as immiscible liquids with effective surface tension originating from different cell-cell adhesion. Other cellular mechanisms, such as cortical tension, can also contribute to the effective surface tension^{7–11}. While theoretically appealing, there have long been questions about whether the DAH and related theories are sufficient to understand the boundaries between cellular populations^{3,12–14}. At a qualitative level, it is often observed that very sharp, low-roughness boundaries exist between coexisting groups of cells, and it is not clear at a quantitative level whether the molecular mechanisms of generating adhesion and tissue surface tension are sufficient to account for this observed boundaries⁷.

Recently it has been observed that a popular simplified class of models of dense epithelial tissues – vertex and Voronoi models of confluent cells^{15–17} – might possess an unusual mechanism that

could lead to sharp interfaces at very little energetic cost. We describe these geometrical, shape-based models in more detail below, but the essential idea is that at interfaces they might possess a non-analytic, cusp-like potential mediated by the coordination of four-fold vertices (corresponding to either highly-coordinated vertices or to very short edges in real cellular systems)¹⁸. This cusp-like potential for cells at the interface can give rise to an apparent effective surface tension larger than the microscopic surface energy term – this can suppress interfacial fluctuations, leading to spatial registry of cells in which cell positions are highly correlated across the interface, yet still lead to relatively compliant behavior with respect to mechanical perturbations^{18–20}.

This proposed underlying mechanism involves a cusp-like non-analyticity on the scale of small numbers of cells affecting the strength of the effective surface tension on a much larger length scale. Much like any coarse-grained analysis of a specific system²¹, from this mechanism one would expect that on short length scales the interface should be dominated by the specific microscopic physics governing the model, and at very large length scales these specific features should eventually average out into an interface governed by a standard surface energy but with a renormalized value of the surface tension. However, it is difficult to analytically compute this renormalization, and existing numerical studies give little quantitative indication of how strong the sharpening effect is at different length scales.

Additionally, although previous studies showed that both Voronoi and vertex models have cusp-like potentials near four-fold vertices, the specific mechanisms are different and result in different interfacial behavior from the perspective of microscopic

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restoring forces²⁰. From the experimental perspective, although four-fold vertices are observed in biological systems, there is as yet no evidence directly supporting the hypothesis that the effect observed in these computational models plays a role in sharpening the boundary in real systems. This is in part due to the fact that when only the average width of the experimental interfaces is measured this allows one to estimate an *effective* surface tension; it is difficult to further estimate which part of the effective surface tension originates from the explicit differences of cell-cell adhesion and cortical tension, and which which part is due to the proposed topological sharpening effect.

In this paper, we revisit the interfacial fluctuations of these shape-based models, focusing on this distinction between a short length scale reduction of interfacial fluctuations and long length scale renormalized surface tension. We do this by studying both the average width of cell-cell interfaces, and also implementing a spectrally resolved analysis of interfacial fluctuations over extremely long simulation times. We show that the spectral analysis provides us more quantitative information of this topological sharpening effect, especially as it differently impacts fluctuations of different wavelengths. We find substantial differences in the interfacial behavior of vertex and Voronoi models. In Voronoi models on long length scales there is an essentially constant degree of interfacial sharpening, independent of the imposed amount of microscopic surface energy or the temperature of our simulations. On short length scales, in contrast, the sharpening effect is extremely strong and strongly depends on the temperature and microscopic surface energy. We compare these predictions with a Brownian bridge model, suggesting that the interfacial effects we observe are due to a combination of a harmonic and a cusp-like potential with a varying population of four-fold vertices that changes the relative strength of these potentials. Unlike our results on the Voronoi model, we do not observe an interfacial sharpening effect in sufficiently equilibrated vertex models. Finally, we analyze existing experimental data²², which we find supports the role of a short length scale topological sharpening effect in real biological systems.

2 Models and methods

2.1 Voronoi and vertex models

We simulate Voronoi and vertex models in a 2D square box with size L and periodic boundary conditions. In the Voronoi model the cell shapes are determined by an instantaneous Voronoi tessellation of the current cell positions. In the vertex model the vertices themselves are the degrees of freedom, and cell shapes are traced out by connecting vertices around a cell. We initialize the system with strips of two different types of cells with width $L/2$, as shown in the inset of Fig. 1. The forces on cells or vertices are calculated as the negative gradient of a cell-shape dependent energy^{18,23}:

$$E = \sum_{i=1}^N \left[k_A (a_i - a_0)^2 + (p_i - p_0)^2 \right] + \sum_{\langle ij \rangle} \delta_{[i],[j]} \gamma_0 l_{ij}. \quad (1)$$

In this energy, N cells have a quadratic energy penalty between instantaneous cell area a_i and perimeter p_i compared to the pref-

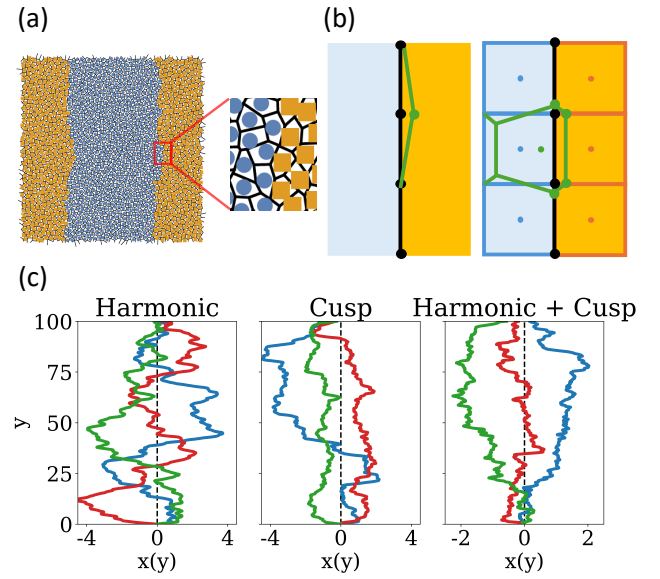


Fig. 1 (a) Sample configuration of coexisting cell populations in the Voronoi model. (b) Schematics of the energy-related changes when a point or cell at the interface is perturbed for particle-based systems with metric interactions (left) and systems with topological interactions (right). Green: the related changes after the perturbation. (c) Sample trajectories for Brownian bridges in harmonic, cusp, and combined potentials. Brownian bridge trajectories were generated by Eq. 9 using the free energy in Eq. 13 with $(\gamma, c) = (1.0, 0.0)$, $(\gamma, c) = (0.0, 1.0)$, and $(\gamma, c) = (1.0, 1.0)$ in the left, center and right plots respectively, with a fixed value of $k_B T = 0.5$.

ferred area a_0 and preferred perimeter p_0 . The second sum introduces an explicit interfacial tension over the edges l_{ij} between neighboring cells with different types ($[i]$ means the type of cell i). Here and below we use γ_0 to denote the microscopic surface energy, and γ to denote the effective value of the surface energy (e.g., as inferred from macroscopic observations of a simulation or experiment). We set the parameter k_A , which controls the relative area stiffness to perimeter stiffness, and the preferred area a_0 to unity. The preferred perimeter $p_0 = 4.0$ so that this confluent cellular systems are deep in the fluid phase²⁴. Unless otherwise specified we perform our simulations with overdamped Brownian dynamics at temperature T :

$$\frac{d\vec{r}_i}{dt} = \vec{F}_i + 2\sqrt{k_B T/dt} \vec{\eta}, \quad (2)$$

where $\vec{\eta}$ is an uncorrelated Gaussian noise of zero mean and unit variance. We use the cellGPU package (<https://github.com/sussmanLab/cellGPU>) for all simulations, and further details of the implementation of these equation of motion for Voronoi and vertex models are contained in Ref. ²³.

2.2 Interfacial width and interfacial spectrum

The width of the interface between cell populations, w , is estimated by analyzing the density profile of one type of cell across the system. Given arrangements of cell populations as in Fig. 1, we average the density of a given cell type along the vertical direction and fit the resulting density profile, $\rho(x)$, to an error

function²⁵ $\rho(x, w) = \frac{1}{2} \left(1 + \operatorname{erf} \left(\frac{x}{\sqrt{2}w} \right) \right)$. Here x is the position relative to the mean location of the interface. The fluctuation spectrum of the interface $|h_k|$ can be obtained by applying a Discrete Fourier transform (DFT) to evenly spaced points on the interface $h(y)$, which are generated through shape-preserving piecewise cubic interpolation. The interpolation and DFT are implemented in MATLAB R2024a with the functions *interp1* and *fft*. The relation between the width and the spectrum is:

$$w^2 = \sum_{k=-n}^n |h_k|^2 = |h_0|^2 + 2 \sum_{k=1}^n |h_k|^2, \quad (3)$$

where $h_0 = \int h(x) dx$, as the average position of the interface, can be set to zero. We choose the number of points on the interface, n , to be equal to L – given that our unit of length is the square root of the average cell size, this is approximately the number of cells on the interface.

Based on the capillary wave theory (CWT)²⁶, in the absence of external forces the spectrum of the thermal capillary waves is expected to be:

$$\langle |h_k|^2 \rangle = \frac{k_B T}{\gamma L} \frac{1}{q^2}, \quad (4)$$

where $q = 2\pi k/L$ and γ is the effective surface tension. This can be derived with the energy equipartition theorem applied to the surface energy:

$$\begin{aligned} E_s &= \gamma \int_0^L \left(\sqrt{1 + h'(y)^2} - 1 \right) dy \\ &\approx \frac{\gamma}{2} \int_0^L h'(y)^2 dy \\ &= \gamma L \sum_{k=1}^n q^2 |h_k|^2, \end{aligned} \quad (5)$$

which gives $\gamma L q^2 |h_k|^2 = k_B T$ for each component of the spectrum. As a result, the average width can be obtained based on Eq. 3 as:

$$\langle w^2 \rangle = \frac{k_B T L}{2\pi^2 \gamma} \sum_{k=1}^n \frac{1}{k^2} \approx \frac{k_B T L}{12\gamma}, \quad (6)$$

where the final approximate equality assumes $n \gg 1$.

An important insight of earlier work¹⁸ is that in a Voronoi model the perturbation of a cell center near the interface may have a surface energy quite different from that of Eq. 5. Near interfaces there is a tendency for these models to spatially coordinate four-fold vertices along the interface, with cell displacements from these highly-registered configurations results in a (topological) discontinuous change in the set of interacting neighbors of the displaced cell, as shown in Fig. 1(b). This suggests a surface energy that contains *both* a standard harmonic contribution (second term) as in CWT (Eq. 5) and a cusp-like term reflecting discontinuities in the topology of the cellular neighbor network (first term):

$$\varepsilon_s = g_0 |h'(y)| + g_1 h'(y)^2. \quad (7)$$

In the highly idealized geometry of square cells perfectly registered along the interface as in Fig. 1(b), the coefficients for the cusp and harmonic terms above are $g_0 \propto \gamma_0(\sqrt{2} - 1)$ and

$g_1 \propto \gamma_0(1 - 1/\sqrt{2})$ ¹⁸, as determined by calculating the energetic cost of cell configurations before and after a small perturbation. The precise prefactors are not particularly relevant – they will be sensitive to the density of four-fold vertices along the interface, and the geometry of cells near the interface – but importantly γ_0 sets the overall scale of both the cusp-like and harmonic terms.

The cusp-like potential gives a non-zero constant restoring forces for very small perturbations, stronger than the restoring forces proportional to the perturbations in the harmonic cases, which results in a sharpening effect^{18,20}. In order to understand the effect of the combined cusp-like and harmonic surface energy, we combine large-scale simulations along with a numerical analysis of a toy model of these interfaces, as described below.

2.3 Brownian Bridge Simulations

In our cell-based simulations (e.g. Fig. 1) the position of the interface between cells of different type, $h(y)$, is constrained by the periodic boundary conditions of our simulation to have $h(0) = h(L)$, and the position of the interface in between the periodic boundaries reflects some sort of effective transverse confining potential for this interface. In this sense we can interpret each instance of $h(y)$ as the result of a “Brownian bridge” process²⁷. Brownian bridges involve stochastic processes that are, e.g., constrained to both start and end at specific locations after a given amount of time. We thus interpret y as the “time” over which a stochastic process evolves, simultaneously subject to a transverse restoring force of different functional forms and the Brownian-bridge constraint that $h(0) = h(y) = 0$.

This is, clearly, a highly idealized framework for thinking about the interface between cellular populations. It assumes a complete separation of length scale between whatever is generating the local energetics of the interface and the size of the interface as a whole – as such, at best it would be able to capture the scaling of the longest-wavelength interfacial properties. Indeed, in the appendix we show analytically how this Brownian bridge framework can reproduce the same scaling of fluctuations predicted by the capillary wave theory described above. That is, we explicitly show that the width of interfaces generated by Brownian bridges subject to the surface energy of Eq. 5 scales as $\langle \sigma^2 \rangle \propto k_B T L / \gamma$. It also gives no insight into the properties of cells that might control the functional form or the strength of the transverse restoring forces. This is further complicated by the fact that at an actual interface different portions of the interface might be governed by *different* transverse restoring forces (e.g., corresponding to local arrangements of the cells that have three-fold vs four-fold coordination) – in this sense, the Brownian bridge framework gives only a mean field prediction.

This is particularly relevant since, as discussed above, in the Voronoi model one expects that the interface lives not in a harmonic potential, but rather in a potential which combines harmonic *and* a cusp-like terms which are each proportional to the microscopic surface energy γ_0 . Including these cusp-like potentials makes analytic solutions to the Brownian bridge process more difficult, and so we turn to straightforward numerical implementations of this toy model of the cellular interface. We proceed

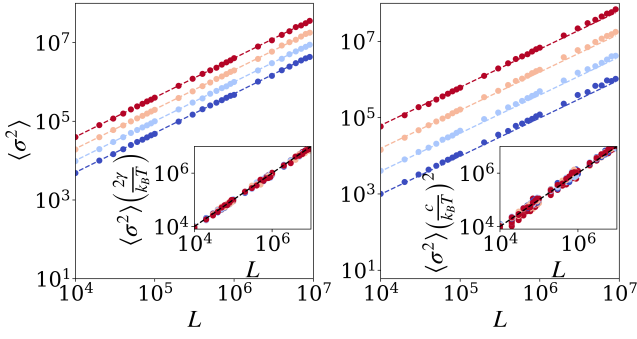


Fig. 2 Brownian bridge variance scaling. The scaling of the average variance of interfaces $x(y)$ generated by numerical simulations of Eq. 9 with a harmonic (left) and cusp (right) potential given by Eqs. 5 and 10 respectively. Colors correspond to temperatures $k_B T \in \{0.005, 0.01, 0.02, 0.04\}$ (blue to red), and dashed lines in the main plot are fits to $\propto (k_B T / \gamma) L$ and $\propto (k_B T / c)^2 L$. Insets in both figures demonstrate the expected collapse of the variance of these processes by the quantities indicated in the main text. Each contains data for $\gamma \in \{0.005, 0.01, 0.02, 0.04\}$ and $c \in \{0.004, 0.008, 0.016, 0.032\}$, and in each the dashed black line denoting the scaling $\propto L$.

as follows. First, we assume that local fluctuations of the interface away from $h' = 0$ obey Landau theory relaxation dynamics for a non-conserved order parameter field²⁸, we write

$$\partial_y h' = -\Gamma \frac{\delta E_s}{\delta h'} + \sigma \eta(y), \quad (8)$$

where $h' = \partial h / \partial y$, Γ is a constant of proportionality with units of inverse energy, and η is a white noise process modulated by a strength σ . In order to impose the Brownian bridge constraint we modify the above equation to²⁷

$$\frac{dh'(y)}{dy} = -\frac{h(y)}{L-y} - \frac{\delta E_s[h'(y)]}{\delta h'} + \sigma \eta(y). \quad (9)$$

The $(-h/(L-y))$ term serves to enforce the Brownian bridge constraint $h(0) = h(L)$, and we work in units of $\Gamma = 1$. In addition to the analytical derivation in the Appendix, we explicitly show in Fig. 2 that our numerical simulations of these equations using the CWT free energy in Eq. 5 reproduces all of the scalings of the CWT interfacial width result in Eq. 6.

With this framework in place, we consider instead a cusp-like surface energy,

$$E_s[h'(y)] = \int_0^L dy c |h'|, \quad (10)$$

where c is analogous to surface tension but has units of energy. As shown in Fig. 2, we find that in this case that the average variance of a bridge scales as

$$\langle \sigma^2 \rangle = \frac{k_B T}{\gamma_{\text{eff}}} L, \quad (11)$$

where

$$\gamma_{\text{eff}} = \frac{c^2}{k_B T} \quad (12)$$

is a temperature-dependent effective surface tension produced by the cusp potential.

Finally, given the “topological sharpening” argument for the Voronoi model presented in Ref.¹⁸ and outlined above, we ex-

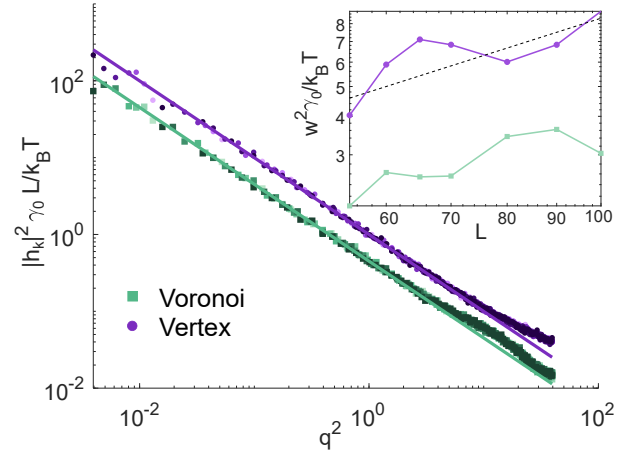


Fig. 3 Large-scale interfacial sharpening. The interfacial spectrum is obtained for systems with sizes $L = 55, 60, 65, 70, 80, 90, 100$ (colors from light to dark) and $\gamma_0 = 0.02$, each being average of 5 independent simulations. In each simulation, spectrum is calculated and averaged for data drawn at 50 evenly spaced time points after the system has reached equilibrium ($t > 3 \times 10^5 \tau$). After rescaling, the interfacial spectra for the vertex model and Voronoi models collapse onto separate curves. The solid lines are the fits of the low-wavevector data ($q^2 < 5$) to the expected form $\frac{|h_k|^2 \gamma_0 L}{k_B T} = \frac{A}{q^2}$ for $A \equiv \gamma_0 / \gamma_{\text{eff}}$. γ_0 is the explicit surface tension in the models and γ_{eff} is the effective surface tension. We find $A = 0.998$ for vertex simulations and $A = 0.450$ for Voronoi simulations. (inset) The rescaled interfacial width for systems of different sizes. The dashed line is the prediction of CWT.

tend this to the case of a surface energy with both cusp-like and harmonic terms in the surface energy which are each proportional to the microscopic surface energy. As such, below when comparing with the results of the Voronoi model we present additional Brownian bridge simulations governed by an energy cost

$$E_s[h'(y)] = \int_0^L dy \left(c |h'| + \frac{\gamma}{2} h'^2 \right), \quad (13)$$

where both c and γ are set proportional to the microscopic surface energy γ_0 . We find numerically that over a broad range of parameters the Brownian bridge “interfaces” generated by Eq. 13 have an effective surface tension which is approximately the sum of the harmonic and renormalized cusp surface tensions:

$$\gamma_{\text{eff}} \approx \frac{c^2}{k_B T} + 2\gamma. \quad (14)$$

3 Large-scale sharpening in Voronoi models

In order to compare the interfacial fluctuations in Voronoi and vertex models with CWT, we run simulations with system size L ranging from 55 to 100 and explicit surface tension $\gamma_0 = 0.02$ for a long enough time to make sure the system has reached equilibrium. For each system, we measure the average width w^2 , shown in the inset of Fig. 3. We find that vertex simulations’ results are in fact *consistent* with CWT: at large length scales there is no obvious sharpening effect. In contrast, our Voronoi model simulations do show interfacial sharpening. Suprisingly, this sharpening effect ($w^2/w_0^2 \approx 1/2$, with w_0 as the width predicted by CWT) is weak

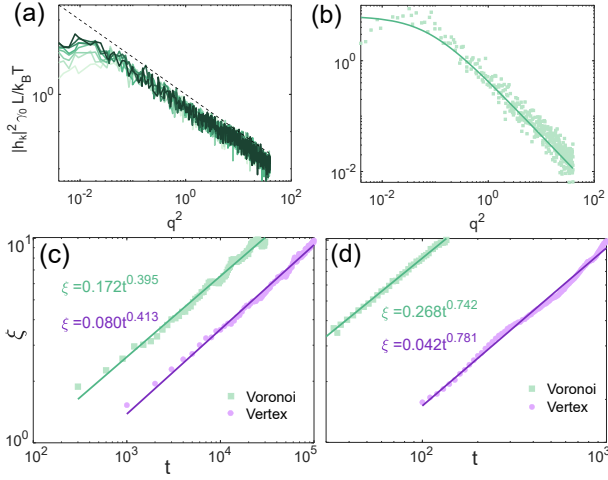


Fig. 4 Dynamics of interfacial Spectrum (a) The interfacial spectra between $t = 3000$ and $t = 24000$ (from light to dark) are plotted Voronoi Brownian Dynamics simulations in systems with sizes $L = 65, 70, 80, 90, 100$ and $\gamma = 0.02$, each with 5 independent simulations. We have confirmed that after rescaling the spectrum by L the data from simulations of different sizes collapse onto a single curve. As such, data for multiple L are combined here.. The dashed line is the theoretical prediction of classical CWT. (b) An example of fitting with equation 15 for the interfacial spectrum at $t = 2700$. (c-d) $\xi(t)$ for Voronoi and Vertex models with Brownian Dynamics (c) and Molecular Dynamics (d), fitted with equation (15). The cutoff of ξ in (c) and (d) corresponds to $q^2 = 10^{-2}$ in spectra of (a) and (b), when the plateau contains fewer than four data points and results in fits with large variance.

compared to the findings in previous literature¹⁸.

To begin resolving this finding, we calculate the full spectrum of interfacial fluctuations, as shown in Fig. 3. We see that, except for the high- q regime (approximately $q^2 > 5$, which corresponds to sizes smaller than roughly three cells), the spectrum is consistent with the relation $|h_k|^2 \propto 1/q^2$. After rescaling, the spectra for systems of different sizes collapse onto one. By fitting the collapsed data with equation $\frac{|h_k|^2 \gamma_0 L}{k_B T} = \frac{A}{q^2}$, where $A \equiv \gamma_0 / \gamma_{\text{eff}}$ is the sharpening factor, we get $A = 0.998$ for vertex simulations and $A = 0.450$ for Voronoi simulations. This is consistent with the conclusion we reached by directly measuring average width in the last paragraph, but fitting the spectrum allows a much more precise estimate of the sharpening compared to the much rougher estimate obtained from measuring interfacial widths. Given the much weaker effect we observe in the Voronoi model, it is perhaps understandable that the vertex model shows almost no effect at the relatively low surface tension used here, as the non-analytical behavior near four-fold vertices have been found not as strong as in Voronoi models²⁰ (a subtlety compounded by the fact that most standard implementations of vertex models do not allow arbitrarily short edges between vertices; instead there is often a minimum edge length below which a T1 transition is triggered, which may in turn mask the effect of stable higher-order vertices²⁹).

Analyzing the full spectrum of interfacial fluctuations indeed helps address why we find much weaker sharpening effects in our Voronoi simulations. In Fig. 4(a) we plot the spectrum of in-

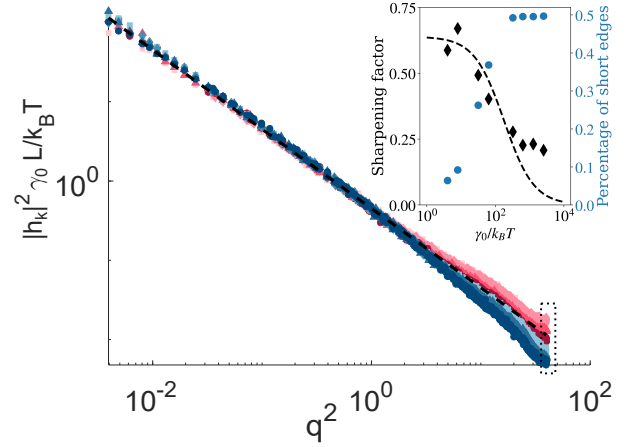


Fig. 5 Interfacial Spectrum with different γ_0 and $k_B T$. All the red symbols are for $k_B T = 0.005$ and the blue ones are for $k_B T = 0.001$. For each temperature group, from light to dark, the γ_0 increases. Specifically, for $k_B T = 0.005$, $\gamma_0 = 0.02, 0.04, 0.16, 0.32$ respectively and for $k_B T = 0.001$, $\gamma_0 = 0.32, 0.64, 1.28, 2.56$ respectively. The black dashed line is for $\frac{\gamma_0 L}{k_B T} \langle |h_k|^2 \rangle = A q^{-2}$ with $A = 0.45$ obtained from the fitting in Fig. 3. The original time step used in our simulation is $dt = 0.01$, but for larger surface tension, to maintain numerical accuracy, we use smaller dt . Specifically, for $\gamma_0 = 0.64, 1.28$, $dt = 0.001$ is used and for $\gamma_0 = 2.56$, $dt = 0.0001$ is used. (inset) The sharpening factor $A \equiv \gamma_0 / \gamma_{\text{eff}}$ versus $\gamma_0 / k_B T$ measured for high- q regime corresponding to about single-cell size (within the dotted box in the main figure) on the left y axis, and percentage of very short edges (smaller than 0.05) on the interface on the right y axis. Dashed black line denotes a fit to the the Brownian model prediction in Eq. 17 with $B_1 = 0.093$ and $B_2 = 1.565$.

terfacial fluctuations over time in our simulations. The magnitude of each fluctuation mode, $\langle |h_k|^2 \rangle$, grows over time until reaching the equilibrium state, and naturally the small- q modes take the longest time to equilibrate. Not only does this lead to spuriously low interfacial widths if the interfacial width is measured before the small- q modes equilibrate, but the correlation time for the small- q modes is also quite long. Thus, within a fixed-duration window of observation time, fewer time points can be used as independent data in the ensemble averaging. This is why in Fig. 3, the data points for small q modes are more noisy. In a linear regression as shown in Fig. 3, the noisy data points in the low- q regime do not influence the fitting result greatly. However, as the magnitude for the small q modes dominate in the width, the interfacial width measurement are less precise, as revealed in the inset of Fig. 3. We thus argue that spectral analyses are a preferred way to quantify the effective surface tension than directly measuring the interfacial width. We believe these issues affected previous measurements of interfacial sharpening in the Voronoi model¹⁸, and also measurements of surface fluctuations in a variety of other soft-matter systems. It is worth checking whether the plateau in the spectrum within the low- q regime or sharpening effect for large-scale systems observed in some previous studies reflects the long equilibration times hinted at above^{30,31}.

We additionally point out that both our and many other measurements have been done for systems evolving according to Brownian dynamics. It is perhaps underappreciated that the approach to equilibrium under Brownian dynamics can be surpris-

ingly slow, especially for the low- q modes at the interface. To quantify this, we fit the spectra before equilibrium with the equation

$$\frac{\gamma L}{k_B T} \langle |h_k(t)|^2 \rangle = \frac{A}{q^2 + \xi(t)^{-2}}, \quad (15)$$

where ξ is defined as a characteristic length. When $\xi \gg q^{-1}$, Eq. 15 reduces to the equilibrium spectrum (Eq. 4). One example of the fitting is shown in Fig. 4(b) and the fitted $\xi(t)$ for Voronoi and vertex simulations is shown in Fig. 4(c). As a comparison, we implement the same process for simulations with Molecular Dynamics, which is suitable for regular liquids with momentum conservation, as shown in Fig. 4(d). The $\xi(t)$ within a certain range, can be well fitted with power functions $\xi(t) = Bt^{1/z}$, with $z \approx 2.5$ for systems with Brownian Dynamics and $z \approx 1.3$ for systems with Molecular Dynamics. This means that for the interface of length L to equilibrate, the characteristic time $\tau \propto L^{2.5}$ for Brownian Dynamics and $\tau \propto L^{1.3}$ for Molecular Dynamics. Because of this, although the simulation time in previous literature¹⁸ is long enough for the interface of a regular liquid system to reach equilibrium, the Brownian Dynamics simulations have not equilibrated so that the measured width (dominated by the low- q modes) is much smaller. This large-scale slow dynamics in systems with microscopic Brownian Dynamics, as also shown in a recent paper on coalescence³², again underscores the need for extra caution when studying the equilibrated state of highly dissipative biological systems. For example, in the advancing front of an epithelial monolayer, determining quasi-equilibrium is difficult without comparing the time scales of boundary equilibration and boundary movement. The time scale for boundary equilibration can be significantly increased by cell-substrate friction, particularly in large systems. Only when the front stops, such as due to contact inhibition when confronting another piece of epithelial monolayer, can measurements be confidently considered at equilibrium.

We note that our estimates of the scaling exponent z is roughly consistent with field theoretical calculations of interfacial roughening in the presence or absence of momentum conservation^{33–36}. It is well-known that even for small fluctuations in equilibrium surfaces one does not expect Edwards-Wilkinson *dynamics* – which would predict $z = 2$ – since there are bulk fluxes whose relaxation is fast on the time scale of long fluctuations of the interfacial height^{33,34}. In the absence of momentum conservation (as in our Brownian dynamics), a non-linear one-loop renormalization predicts $z = 3 - 1/3$ for an interface of spatial dimension $d_I \leq 2$ ³⁶. The prediction for fully wet systems is $z = 1$ ³⁷. Considering the difficulty of extracting these exponents, our results of $z \approx 2.5$ and $z \approx 1.3$ may be consistent with these predictions. We also note other potential discrepancies, such as the non-monotonic behavior for the early stage spectrum. It cannot be well-fit to a $\frac{A}{q^2 + \xi(t)^{-2}}$ form, as seen in Fig. 4(b). Both of these points will be investigated in future studies targeted at more precise scaling analyses of interface roughening. It would also be interesting to extend these studies to models with self-propulsion or other non-equilibrium dynamics, to more explicitly test theories of active phase separation^{34,36,37}.

The above results are for systems with relatively low surface

tension. Then, we run more simulations by varying γ_0 and $k_B T$ to check how they influence the sharpening factor A , as shown in Fig. 5. For Voronoi models, on large scales ($q > 1$), after rescaling, all the spectra collapse onto one and the sharpening factor A is always around 0.5. However, on small scales ($q < 1$), the sharpening factor varies, which we will dig into in the next section.

4 Small-scale sharpening in Voronoi models

In contrast to the relatively weak but robust sharpening effects at the longest length scales, we continue to find very strong sharpening effects at smaller length scales. From Fig. 5, in the regime $q > 1$, corresponding to length scale of about a few cells, the magnitude of the modes decreases as we increase $\gamma_0/k_B T$ (and the effect grows strong at higher q). In the inset of Fig. 5 (left y axis), we show that the sharpening factor $A \equiv \gamma_0/\gamma_{\text{eff}}$ on the length scale of about one cell decreases from around 0.7 to about 0.2 when we increase $\gamma_0/k_B T$ from 4 to 2560. This change is consistent with the trend of increasing percentage of short edges on the interface, as shown on the right y axis in the inset of Fig. 5, supporting our hypothesis that this small-scale sharpening is due to the four-fold vertices in the Voronoi model. To better understand how the cusp-like potential near the four-fold vertices influences the sharpening, we build a Brownian bridge model to compare with the Voronoi model simulations.

As discussed in Section 2.3, our Brownian bridge model predicts that an interface in the presence of a cusp potential should have a temperature-dependent *effective surface tension* (γ_{eff}) given by Eq. 14. The corresponding predicted sharpening factor is then given by

$$\frac{\gamma_0}{\gamma_{\text{eff}}} = \frac{\gamma_0}{\frac{c^2}{k_B T} + 2\gamma}. \quad (16)$$

As noted above, both the cusp and harmonic terms have strength set by the microscopic surface energy. In the real cell systems we do not a priori know the relative density of four-fold vertices generating the cusps nor the precise geometry which sets the balance between the harmonic and cusp terms. Although they are in principle themselves dependent on the ratio $\gamma_0/(k_B T)$, for simplicity we simply adopt parameters B_1 and B_2 in the expression

$$\frac{\gamma_0}{\gamma_{\text{eff}}} = \frac{1}{B_1^2(\gamma_0/k_B T) + B_2}. \quad (17)$$

as free but constant parameters quantifying these effects.

We see that even with this simplifying assumption, Eq. 17 provides a reasonable fit to the observed sharpening in the Voronoi model simulations over several orders of magnitude of $\gamma_0/k_B T$, as shown by the comparison between the black diamonds and the dashed curve in the inset of Fig. 5. In the limit $\gamma_0/k_B T \rightarrow \infty$ our Brownian bridge model predicts infinite sharpening as $\gamma_{\text{eff}} \rightarrow \infty$, whereas in Voronoi model simulations the sharpening appears to plateau at a non-zero value. The fact that the plateau in sharpening coincides with a plateau in the number of four-fold defects populating the interfaces suggests that the discrepancy with the Brownian bridge model is likely connected to non-negligible geometry-based variations of the parameters B_1 and B_2 in the $\gamma_{\text{eff}} \rightarrow \infty$ regime. Regardless, these results demonstrate

that the small wavelength deviations from CWT scaling observed in Voronoi models can be understood as a consequence of the presence of four-fold vertices along the interface inducing a dependence of the effective surface tension on $\gamma_0/k_B T$.

5 Discussion and outlook

In this paper we investigated in detail the strength of interfacial fluctuations in models of cellular monolayers. Our focus has been on the quantification of how an unusual proposed “topological sharpening” effect at the microscopic scale renormalizes the effective surface tension at different length scales. Combining long-time simulations of Voronoi models with a spectrally resolved analysis of interfacial fluctuations, we quantitatively measured this sharpening effect on different lengths scales. We find that on large scales (more than ~ 10 cells), this non-analytical local potential is renormalized to an equivalent harmonic potential with a modestly larger effective surface tension. In contrast, on small scales the interfacial fluctuations are strongly suppressed by the presence of cusp-like contributions to the restoring force at the interface. We verified that the broad features of this sharpening effect on small scales can be captured by a Brownian bridge model of the interface that combines cusp and harmonic contributions; the coefficients of this model would in general depend on both the fraction of four-fold vertices at the interface and the precise geometry cells adopt near it. It is worth mentioning that although our simulations are based on Voronoi and vertex models, the underlying mechanism is the cusp-like potential energy. Thus, we expect that these results will also be of relevance for other tissue models with topological interactions. This could potentially include deformable polygon models³⁸ in density regimes where perturbation of interfacial cells results in an energy cost with a cusp-like term.

The fact that there are short length scale deviations from the predictions of CWT is, of course, not in itself surprising: long lengthscale descriptions of physical systems always coarse grain over microscopic details, but at the scale of interacting particles these coarse grained descriptions discard relevant physical interactions. Importantly, we note that the *trend* of this effect when studying interfacial fluctuations in particulate systems is typically in the opposite direction of what we find in these models of cellular interfaces: rather than a *sharpening* effect, the finite size of the interacting particles lead to a roughening of the interface relative to CWT scaling^{31,39,40}. Similar roughening effects are also observed in vertex models (Fig. 3) with small γ_0 when the topological effect of shape-based models is weak.

Interestingly, this behavior may be a point of difference between Voronoi and vertex models. The properties of both of these similar shape-based models are often quite similar, and although there are some subtle differences between them (for instance, it has been found that they have different athermal jamming transitions⁴¹), they are often regarded as largely interchangeable when it comes to modeling. However, when we study their interfacial fluctuations we find qualitatively different behavior which is strongly sensitive to the imposed microscopic surface energy γ_0 . For small values of γ_0 , Fig. 3 shows that for sufficiently equilibrated vertex models clear sharpening effects do *not* seem to

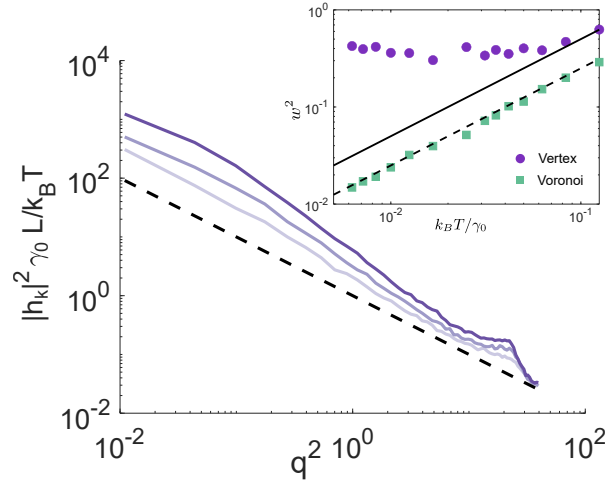


Fig. 6 Interfacial width for Voronoi and vertex models with different $k_B T$ and γ_0 .

be present. Furthermore, the large wavevector behavior shows a roughing effect that is similar to the typical behavior of particulate systems. This may be understandable, since in vertex models the perturbation of a single vertex does not necessarily lead to topological changes around the whole cell of the sort shown for Voronoi tessellations in Fig. 1. Previous studies have also indicated at most weak topological sharpening effects in vertex models by measuring the cusp-like restoring forces with small perturbations²⁰.

The interfacial behavior of vertex models seems more complicated for large values of γ_0 . The inset of Fig. 6) shows a plateau in the interfacial width over a broad range of non-dimensionalized surface energies, where the magnitude of the plateau indicates an interface substantially rougher than that predicted by CWT. This is further reflected in the complicated changes in the spectrum seen in Fig. 6. We speculate that this roughening is, again, connected to the very different microscopic interfacial behaviors near four-fold vertices in Voronoi and vertex models observed in previous studies²⁰. The authors of that work found that the non-zero plateau of restoring forces due to the cusp-like potential have a very different dependence on explicit surface tension γ_0 . A natural speculation is that this may also be connected to the introduction of short length scales that trigger topological transitions in common implementations of vertex models, but we have not found any direct evidence that varying this T1 length scale affects our findings. Further work may be needed to disentangle the subtle difference in interfacial behavior between Voronoi and vertex models.

Given the above discussion, an important question is whether topological sharpening effect can be seen in real biological systems. Previous studies have certainly observed the presence of four-fold vertices and of cell registration in epithelial tissues^{20,42}, but this does not on its own imply that, e.g., four-fold vertices in cells play the same role that they do in highly simplified models of dense tissue. Measurements of interfacial width give an estimation of the *effective* surface tension, but without knowing the microscopic details (such as precise measurements of cellular ad-

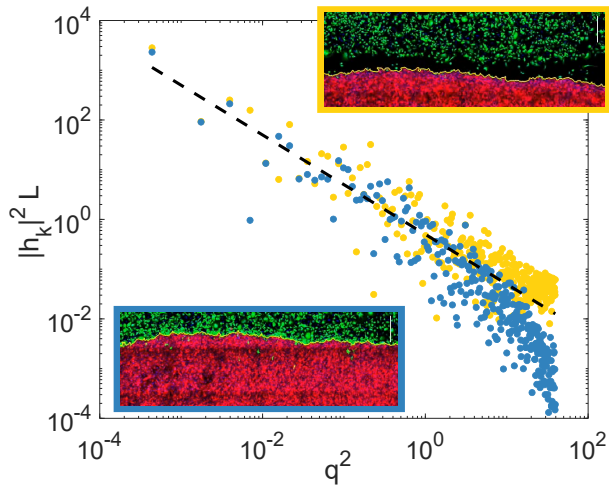


Fig. 7 Interfacial spectrum calculated from experimental images (extracted from movie "Confrontation of MCF10A E-cad KD/MDA-MB-231 monolayers yields a straight interface" of Ref.²²). Yellow data points are estimated from a single snapshot at an early time (before the two types of epithelial cells contact). Blue data points are from a snapshot after a stable interface between the cell sheets has formed. The black dashed line is guide to the eye with the expected CWT scaling relation $|h_k|^2 L = 0.5/q^2$.

hesion energies, cortical tensions, etc.), one cannot tell whether the effective surface tension does or does not match the microscopic one. We suggest that spectral analyses are a robust way to probe this question.

As an example, we analyzed video data of epithelial cell sheets presented as supplemental material in Ref.²², with estimates of two interfacial spectra shown in Fig. 7. A relevant experiment in that paper involved the confrontation of two epithelial monolayers, composed of MCF10A-E-cad KD and MDA-MB-231 cells. At early times in the experiment – before the two monolayers are in contact and, hence, before the possibility of forming any four-fold vertices – we observe a spectrum of fluctuations consistent with CWT even at the highest wavevectors. In contrast, when the two cell sheets come into contact there is an obvious regime of high- q sharpening, starting near $q \approx 1$. Fitting the low- q regime of this data to the predictions of CWT, we find $k_B T / \gamma_{\text{eff}}^y \approx 0.538(0.3277, 0.7482)$ for the cells before the sheets contact and $k_B T / \gamma_{\text{eff}}^b \approx 0.3981(0.2654, 0.5307)$ after the two cell sheets have formed a stable interface (where the numbers in parentheses represent a 95% confidence bound). Thus, within the precision of our analysis of this existing experimental data, by looking at only the low- q modes it is hard to determine whether there is an overall two-fold sharpening that our Voronoi model simulations would predict. On the other hand, the transition between the low- q and high- q regimes shows an effect quite like that in Fig. 5, and supports the Voronoi model prediction of strong interfacial sharpening at small lengthscales. This reiterates the value of spectral analysis in the study of biological interfaces.

We note that there are still some quantitative differences between our analysis of these experiments and our model simulations. Based on Fig. 5, the high- q regime sharpening becomes most obvious when the effective surface tension as estimated

from the low- q regime is of order $\gamma_{\text{eff}} \approx 2\gamma_0 \approx 200k_B T$. In Fig. 7, the effective surface tension for blue points, which already show obvious high- q sharpening, is apparent even at our estimate of $\gamma_{\text{eff}} \approx 2.5k_B T$. As the obvious high- q regime sharpening is enhanced when there are a large percentage of short edges (inset of Fig. 5), the above inconsistency may suggest that in the experiments, the interfaces between two epithelial monolayers can have high-level of four-fold vertices, i.e. cell registration, even when the large-scale effective surface tension is not very strong. In the Voronoi model registration between cells across an interface is only enforced by the surface tension term, but of course in biological systems there may be other molecular mechanisms that support this. For example, previous studies found that actin-based protrusions, like filopodia and lamellae, help with the correct matching of opposing cells along the fusion seam during dorsal closure^{42,43}.

To further verify the above speculations, we propose the need for more experiments that combine large enough systems to provide a large range of the spectrum (as in Refs.^{22,44}), which also have clear images of cell membranes that can clearly show relative fractions of highly-coordinated vertices along the interface (as in Refs.^{45,46}). We believe that integrating modeling studies like those presented here with biological experiments involving genetic mutations of adhesion or actin cortex-related proteins is a promising route to enhancing our understanding of the specific roles these proteins play in the interfacial behaviors of epithelial tissues.

Conflicts of interest

There are no conflicts to declare.

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6 Appendix: Variance of Brownian bridge processes

Here we derive the scaling relationship for the interfacial widths expected in our Brownian bridge framework. Brownian bridges $h(y)$ satisfy

$$P_{\text{bridge}}(h, y) = \frac{P(h = \delta, y = L | h_0 = h, y_0 = y) P(h, y | h_0 = \delta, y_0 = 0)}{P(h = \delta, y = L | h_0 = \delta, y_0 = 0)} \quad (18)$$

As in the main text, we begin by assuming that local fluctuations of the interface away from $h' = 0$ obey Landau theory relaxation dynamics for a non-conserved order parameter field:

$$\partial_y h' = -\gamma \frac{\delta E_s}{\delta h'} + \sigma \eta(y). \quad (19)$$

As before, η is a white noise process modulated by a strength σ and γ is a constant of proportionality with units of inverse energy (which we here take to be proportional to the harmonic surface energy term in Eq. 5). In the smoothly varying (over-damped)

limit, $|\partial_y h'| \ll 1$, Eqs. 5 and 8 simplify to

$$\gamma \frac{dh}{dy} = \sigma \eta(y), \quad (20)$$

where in thermal equilibrium

$$\sigma \equiv \sqrt{2\gamma k_B T}. \quad (21)$$

The free-particle Smoluchowski equation corresponding to Eq. 20 is given by

$$\partial_y P(h, y | h_0, y_0) = D \partial_h^2 P(h, y | h_0, y_0), \quad (22)$$

where $D \equiv \frac{1}{2} (\sigma/(\gamma))^2 = (k_B T)/(\gamma)$, and admits the solution

$$P(h, y | h_0, y_0) = \frac{e^{-\frac{(h-h_0)^2}{4D(y-y_0)}}}{\sqrt{4\pi D(y-y_0)}}. \quad (23)$$

Inserting Eq. 23 into Eq. 18, then yields

$$P_{\text{bridge}}^{\text{CWT}}(h, y) = \frac{1}{2} \sqrt{\frac{L}{\pi D y (L-y)}} e^{-\frac{L h^2}{4 D y (L-y)}}. \quad (24)$$

The variance of this distribution is computed straight-forwardly to be

$$\text{Var}(y) = 2D \frac{y(L-y)}{L}, \quad (25)$$

which, averaged across the entire bridge, reduces to

$$\langle \text{Var}(y) \rangle = \frac{1}{3} \frac{k_B T}{\gamma} L. \quad (26)$$

Although there is a difference in prefactor, we note that Eq. 26 yields precisely the temperature, length, and surface-tension scaling relation expected for a two-dimensional interface by standard CWT.

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