Title

Effective mechanical potential of cell–cell interaction explains basic structures of

three-dimensional morphogenesis

Short title

Effective potential of cell–cell interaction during morphogenesis

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Author contribution

H.K. designed the work, H.K., A.M.I., and H.O. contributed to the conception. H.K. and

T.F. designed experiments, T.O. provided experimental materials, and H.K. performed

experiments and image processing. H.K. and H.O. designed models, H.K. developed

computational algorithms, and H.K. and K.N. statistically analyzed the data. H.K., T.O.,

H.O., A.M.I., K.N., K.K., and T.F. wrote the manuscript, and all authors contributed to

the interpretation of the results.

Conflicts of interest

49 The authors declare no competing financial interests.

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Abstract

Mechanical forces of cell–cell interactions have been suggested to be critical for the

emergence of diverse three-dimensional morphologies of multicellular organisms. The

direct evaluation of the forces in living systems has been difficult due to technical

limitations. Here, we developed a framework for inferring and modeling mechanical

forces of cell–cell interactions. First, by analogy to coarse-grained models in molecular

and colloidal sciences, cells were assumed to be spherical particles, where the mean

forces (i.e. effective forces) of pairwise cell–cell interactions were considered. Then, the

forces were statistically inferred from live imaging data, and subsequently, we

successfully detected effective mechanical potentials of cell–cell interactions as a

function of the cell–cell distances in Madin-Darby canine kidney (MDCK) cells,

C.elegans early embryos, and mouse pre-implantation embryos. The qualitative and

quantitative differences in the inferred potentials can be a control parameter for

morphological transition during the mouse compaction process, and can also reproduce

various three-dimensional morphologies including aggregates, cavities, tubes, cups, and

two-dimensional sheets, which constitute basic structures observed during

morphogenesis. We propose that effective potentials of cell–cell interactions are

measurable, and their qualitative and quantitative features are critical for the emergence

of diverse three-dimensional morphogenesis.

Significance statement

Emergence of diverse morphologies of multicellular organisms is one of the most

intriguing phenomena in nature. Mechanical forces generated by cells play central roles

in morphogenesis, however, their measurement is technically limited. Furthermore, due

to the complex situations in living systems, a model for describing the emergent

properties of multicellular systems has not been established. Here, we developed a

method for inferring mechanical potential energy of cell–cell interactions, and showed

that the quantitative differences in the potential is alone sufficient to describe basic

three-dimensional morphologies observed during embryogenesis and organogenesis.

This framework sheds light on the emergent properties of multicellular systems.

Introduction

In multi-cellular living systems, various three-dimensional morphologies are

observed. The emergence of these diverse morphologies is thought to be primarily

dependent on the mechanical properties of the constituent cells. In particular,

mechanical forces of cell–cell interactions are involved in morphogenetic events such as

epithelial cell movement and cell sorting (1–5). However, the mechanical basis of

morphogenesis, which gives rise to a variety of structures, remains to be elucidated.

The mechanical forces of cell–cell interactions are regulated by various factors.

In epithelial cells, whose shapes are typically columnar polygons, cell–cell adhesion

energy is controlled by cadherin family proteins (1, 2, 6). Moreover, cellular polarities

such as apico-basal polarity are linked to biased localization of cadherin proteins,

leading to directional differences in the forces of cell–cell interactions. In addition to

these direct interactions, there are indirect interactions through external factors. In

mesenchymal cells, cell–cell interactions are regulated through extracellular matrix (i.e.

substrate) originally generated by the cells (7–9). Therefore, cell–cell interactions entail

many microscopic parameters determined by various proteins and cellular processes,

resulting in very complex physics with many degrees of freedom. To elucidate the

principles underlying the emergence of morphological diversity, it is essential to

investigate how these cellular features can be described by models with simple meso or

macroscopic parameters.

The mechanical potential of cell–cell interaction is conceptually well

recognized, and has been used in theoretical studies where various potentials such as a

hard- or soft-core potential were considered (Fig. 1A-B) (7, 10–18). Although adhesive

forces between two isolated cells have been measured (i.e. forces required to dissociate

one cell from the other) (4, 19), they have neither been measured nor inferred in more

complex multi-cellular systems, which differ from isolated cells as the consequences of

the cell polarities, the indirect interactions, etc. If effective potentials of cell–cell

interactions could be defined as a model integrating these various factors (Fig. 1C), the

effective potentials would be useful to explain the emergence of morphological diversity

of multi-cellular systems. However, it is challenging to identify all these factors and

measure the values of their related-parameters, which is a bottleneck to obtain effective

potentials through bottom-up approaches. Consequently, it remains poorly understood

whether qualitative and quantitative differences in the effective potentials are detectable

in living systems and whether the differences can explain the emergence of various

three-dimensional morphologies.

In this study, by analogy to non-living materials (Fig. 1D-G), we tried a

top-down approach: we developed a method for statistically inferring the effective

potential of cell–cell interactions using nuclear tracking data from live imaging as

reference data, which are relatively easy to obtain. Thus, shapes and polarities of cells

were not explicitly considered at this moment, whose effects we tried to decode after

acquiring inference results (Fig. 1I). Our method was validated by using various

simulation data with or without external factors including egg shells and cavities.

Influence of boundary conditions such as egg shells on the inference results was

revealed to be negligible. Then, we tested whether effective potentials of cell–cell

interactions can be detected in living systems. We applied this method to relatively

small multi-cellular systems (4–350 cells) including mouse pre-implantation embryos

and early embryos of the nematode *Caenorhabditis elegans*. We clearly detected

effective potentials as a function of cell–cell distances in these systems. Importantly, we

identified a control parameter from the effective potentials, which can describe

morphological transition during the compaction of the mouse embryos. Further

theoretical analyses demonstrated that effective potential can reproduce various

structures including cell aggregates, cavities, tubes, cups (round and hollow shapes),

and two-dimensional sheets (Fig. 1H). The inferred potentials would be useful to predict

cellular behaviors and mechanical properties of the systems.

Theory and principle for inferring effective potential

Overview of strategy for inferring effective potentials/forces of cell–cell interaction

To measure or infer the effective potentials or forces of cell–cell interactions *in*

vivo, we developed a particle-based cell model in which attractive or repulsive forces

are exerted between each pair of cells (Fig. 1A, B and D). This coarse-grained model

does not explicitly incorporate microscopic features of cells such as cell shape,

polarities, size, heterogeneity, cell type or origin (e.g., epithelial, mesenchymal, or

blastomere cells), etc. (Fig. 1C). Simultaneously, we obtained time series of

three-dimensional cell positions from confocal microscopic images (Fig. 1E) and

statistically inferred the forces of cell–cell interactions by systematically searching for

force values that could reproduce the cell movements observed *in vivo* (Fig. 1F). To

validate the method, we utilized synthetic data generated by simulations under given

potentials of cell–cell interactions, and confirmed that the inferred forces were

consistent with the given-potentials. The details of each part of the analysis are

explained in the following sections.

Particle-based cell models

In our model, particles interact with each other and attractive or repulsive

forces (F_i) are assumed (Fig. 1D), where i is an identifier for particle–particle

interactions. In three-dimensional cellular systems with no attachment to substrates, we

did not assume persistent random walks which are originated from cellular traction

forces on substrates (20, 21).

The equation of particle motions is defined below. In cellular-level phenomena,

viscous drag force or frictional force provided by the surrounding medium or tissue is

dominant, as expected from the low Reynolds number; consequently, the inertial force is

negligible in general (3, 7, 9, 22, 23). The forces in such a system can be assumed to be

correlated with the velocities of the objects (3, 9, 15, 24, 25). Thus, the velocity (V_p) of

a particle is calculated by the net force (F_p) exerted on the particle as follows:

$$V_p = F_p / \gamma \qquad\qquad\qquad \text{(Eq. 1)},$$

where p is an identifier for particles and γ is the coefficient of viscous drag and

frictional forces. F_p is the summation of F_i: $F_p = \sum_{i=1}^{I} \delta(p) F_i$, where I is the total number

of interactions, and in the case that the ith interaction is formed with the pth particle,

$\delta(p)$ is 1, otherwise $\delta(p)$ is 0 (Fig. 1D). We assumed the simplest situation, i.e., γ is

constant (=1.0). Thus, by giving the values of F_i from, for instance, a distance–potential

curve, a simulation can run.

Data acquisition of time series of cell positions

To define cell positions, we focused on nuclei, because they are most easily

imaged by microscopy (Fig. 1E) in a wide range of organisms from *C. elegans* to

mammals, and these data are accumulating (26–30). We utilized publicly available

nuclear tracking data of a developing embryo of *C. elegans* (26); nuclear tracking data

of developing mouse embryos and MDCK cultured cells were obtained in this study (SI,

Section 2, 8, and 11 with text data).

Development of method for inferring effective potentials/forces of cell-cell interaction

In the case of ions, molecules, etc., radial distribution functions are often

measured and used to infer the effective potentials of their interactions (31, 32).

Although this method is simple and applicable to thermodynamically equilibrium

systems, it is not suitable for non-equilibrium systems, including morphogenetic events.

To infer the effective potentials of non-equilibrium systems, we fitted the particle-based

cell model to the time series of nuclear positions in the spirit of data assimilation, where

the fitting parameters are forces between all pairs of cell–cell interactions (the total

number is *I*, as previously defined) for each time frame. Data assimilation is a technique

to solve an inverse problem for a simulation-based model (33, 34). One of well-known

model fittings is based on least squares in combination with a linear model

(function-based model), resulting in inference of parameter values of the linear function.

In data assimilation, models to be fitted are simulation-based models instead of

function-based models, whereas least squares are also often used to infer parameter

values which are usually numerically solved. Through repeated cycles of simulations of

the model, we systematically searched for the values of the effective forces between all

pairs of cell–cell interactions that minimized the differences (i.e. corresponding to least

squares) between the particle positions in the simulations and the *in vivo* nuclear

positions (Fig. 1). The differences (G_{xyz}) to be minimized are defined as follows (SI,

Section 4-1 and 4-2).

$$G_{xyz} = \sum_{t=1}^{T} \sum_{p=1}^{P(t)} \frac{\left\{ x_p(t) - x_p^{\text{ref}}(t) \right\}^2 + \left\{ y_p(t) - y_p^{\text{ref}}(t) \right\}^2 + \left\{ z_p(t) - z_p^{\text{ref}}(t) \right\}^2}{\Delta t}$$

219   (Eq. 2)

Here, p is an identifier for particles, t is an identifier for time frames, and Δt is

the time interval between the time frames. The x, y, and z coordinates of the pth particle

obtained from microscopic images are x_p^{ref}, y_p^{ref}, and z_p^{ref}; ref means reference (Fig.

S2C). The coordinates during the repeated cycles of simulations are x_p, y_p, and z_p. The

effective forces were inferred for each cell–cell pair for each time frame (Fig. S2B and

C). Note that, because γ was assumed to be constant in Equation 1, we can only infer

relative but not absolute values of effective forces.

To determine whether our method can correctly infer effective forces, we first

applied it to synthetic data generated by simulations under given potentials of particle–

particle interactions, and examined whether the inferred effective forces were consistent

with the given potentials (Fig. S3A). We tried the Lenard–Jones (LJ) potential as a test

case of the given potentials. We applied the inference method to the time series of the

simulation data, yielding effective forces for each pair of particles for each time frame

(Fig. S3B). The inferred effective forces were plotted against particle–particle distance

(Fig. S2D, S3C, and SI, Section 5). By averaging the plots for each shallow bin of

distances, we obtained distance–force (DF) curves (Fig. S3C). We found that the DF

curves were not consistent with the curve from the LJ potentials, suggesting that

Equation 2 was not optimal. Therefore, we tried to incorporate various additional

constraints (i.e. cost functions during the minimization procedures) into Equation 2. We

found that, when a cost function was set so that force values approach zero at long

distant regions of cell–cell interactions (Equation 3 and SI, Section 4-3), inferred DF

curves became consistent with that from the LJ potentials (Fig. S3). The cost function is

as follows:

$$G = G_{xyz} + G_{F_0} \quad (G_{F_0} = \omega_{F_0} \sum_{t=1}^{T} \sum_{i=1}^{I(t)} \psi(D_i) F_i^2 \Delta t) \qquad \text{(Eq. 3)},$$

where the cell–cell interaction ID (i) defined in each time frame is from 1 to $I(t)$, D_i is

the cell–cell distance of the ith interaction, and ω_{F_0} is a coefficient for determining the

relative contribution of the second term to G. $\Psi(D_i)$ is a distance-dependent weight that

increases gradually toward the cut-off distance and eventually becomes extremely large,

defined as Equation S4 and Figure S2A-iii. Detailed principles, rational, and procedures

are described in SI (Section 4). Additionally, if this cost function was not incorporated,

this minimization problem exhibited indefiniteness (i.e., a unique solution could not be

determined) and over-fitting also occurred. This cost function is considered as the prior

in the Bayesian inference, and, in general, a unique solution could be ensured by

introducing such a prior (33).

Using Equation S6 containing the above cost function (SI, Section 4-3), we

also tried other potentials generated by freehand drawing, called "FH" potentials in this

study, as an alternative to the LJ potentials (Fig. S3A and Movie S1B). The FH

potentials were chosen because the simulation outcomes under the FH potentials differ

from those under the LJ potentials as shown later (Fig. S12). The inferred DF curves

were almost consistent with the given potentials (Fig. S3C, E, and SI, Section 4-3-4).

Conversely, we also used time series of random–walk simulations as negative controls

(Fig. S3A and Movie S1C), resulting in DF curves with zigzag and disorganized profiles

(Fig. S3C and E). Similar disorganized profiles were also detected in two-dimensional

cellular systems where each cell almost randomly migrates due to its self-propelled

activity under negligible cell–cell interactions (Fig. S5). These results indicate that our

method can correctly infer DF curves.

Results

Systematic validation of inference method using simulation data

We first systematically validated our inference method using simulation data

which include various external factors. Specifically, we assessed whether spatial

constraints such as egg shells or cavities can affect inference results of effective forces

of cell–cell interactions. Similar to the previous section, we generated simulation data

under LJ potential, to which we applied our inference method.

In the absence of spatial constraints, inferred distance–force (DF) curves were

well consistent with the DF curves derived from the LJ potential under various settings

of simulations: the settings are related to time interval of sampling, cell proliferation,

and fluctuation of forces between cell–cell interactions (Fig. 2A-B, S4, SI, Section 4-4,

7-4, and Movie S8A-B). Next, we introduced spatial constraints corresponding to egg

shells (SI, section 7-4-2 and Movie S8C). Even under narrow cylindrical constraints,

inferred DF curves were well consistent with the DF curves from the LJ potential (Fig.

2C; length of cylinder = 25μm, S4, and SI, Section 4-4). By contrast, under constraints

with smaller volume where each particle was highly compressed, the profiles of inferred

DF curves were shifted leftward in the distance–force graph (Fig. 2C; length of cylinder

= 15μm). This shift can be reasonable, if radial distribution functions under compressed

conditions are considered. However, in cell biology, we think that the above

compressive conditions are not physiologically relevant, because cells are essentially

composed of non-compressive liquid and their volumes are not changed by external

pressure. These results suggest that inferred DF curves are not significantly affected by

boundary conditions provided from spatial constraints.

Next, we introduced cavities. Cells can generate cavities by secreting liquid,

and then indirectly push other cells away (35–37), possibly resulting in indirect

interactions. Although the generations of cavities are derived from the intrinsic features

of cells under *in vivo* situations, we implemented cavities as spatial constraints under

simulations. We assumed that particles located on the surface of a cavity cannot

penetrate into the cavity (Fig. 2D, SI, S4, Section 7-4-2, and Movie S8D). Inferred DF

curves were almost consistent with the DF curves from the LJ potential (Fig. 2D, S4,

and SI, Section 4-4). However, we detected additional repulsive forces at distant regions,

which we call distant energy barrier (DEB). Additionally, profiles similar to DEB are

well-known in molecular and colloidal sciences: by considering hydration of particles

by solvents as an external factor, effective potentials are modified so as to contain

energy barriers around distant regions (31, 32). We conclude that the cavities can

qualitatively change the profiles of effective potential of cell–cell interactions.

Related   comprehensive   analyses   are   provided   in   bioRxiv,   doi:

https://doi.org/10.1101/812198, supporting that, despite the wide distributions of the

data points as shown by the heat maps, the inferred DF curves were reliable.

Inference of effective forces in C. elegans early embryo

We next investigated whether effective potential could be detected as a function

of cell–cell distance in three-dimensional *in vivo* systems. The nematode *C. elegans* has

well-defined embryogenesis: cell movements in early embryos are almost identical

among different individuals, and the time series of nuclear positions have been reported

previously (26). During early *C. elegans* embryogenesis, a fertilized egg (i.e. one cell)

repeatedly undergoes cell division and cell differentiation, ultimately forming an ovoid

embryo containing ~350 cells (Fig. S6A and Movie S3A). The cells become smaller

through cell division (38), and the total volume of the embryo remains constant because

the embryo is packed into a stiff egg shell. Thus, the volume of each cell is reduced by a

factor of 350 relative to the original egg, meaning that cell diameter is reduced by a

factor of 7. Because particle diameter is approximately reflected on a DF curve as the

distance with force = 0 (Fig. 1B) corresponding to the minimum of the potential, we

expected that inferred DF curves should be gradually shifted to the left side of the

distance–force graph throughout embryogenesis.

Figure 3A is a snapshot that shows nuclear positions and inferred effective

forces in the embryo; the temporal change is also visualized (Fig. S6B and Movie S3B),

where dynamic changes in the effective forces were observed. It was confirmed that our

inference problem has essentially a unique solution (Fig. 3B, S6D, and SI, Section 4-4).

We divided the whole embryogenesis into segments containing different time frames,

and obtained DF curves (Fig. 3C and S6C, time frame = 16–55, 36–75, 76–115, 116–

155, and 156–195). The DF curves exhibited organized patterns. All of the DF curves

had very similar profiles: repulsive and attractive forces were detected around the core

of the cells and longer interaction distances, respectively (Fig. 3D). We also calculated

the distance–potential (DP) curves by integrating the force values (Fig. 1B, S2D, and SI,

Section 5). Importantly, the distances at the minimum of the effective potentials were

gradually shifted toward the left side (Fig. 3D), as we had expected (Fig. 3E), indicating

that the inferred effective potentials were quantitatively consistent with the reduction in

cell volume during embryogenesis (Fig. 3E). Taken together, the patterns of the DP

curves seem to be physically reasonable.

Effect of inferred distance–force curves on C.elegans embryonic morphologies

We next investigated whether the simple DF curves can explain embryonic

morphologies. Because the dynamics of all cells during embryogenesis are very

complicated, largely due to the repeated cell division and cell differentiation, we

focused on a few typical features of the morphologies: aggregated cells and the ovoid

embryo shape. The embryonic cells can keep a cell aggregate even when the eggshell is

removed, and subsequent culture leads to a cell aggregate with an ovoid or distorted

shape, except for very early stage of development ($< \sim$20 cells) (39, 40).

We performed simulations based on the particle-based cell model by

considering the DF curves, and investigated whether aggregated states with ovoid

shapes were stable. In multi-cellular systems, morphologies of systems are determined

by both DF curves and initial configurations, because energetic local minimum states

(i.e. metastable state) as well as global minimum states are meaningful. To find stable

states under the DF curves, we performed simulations starting from various initial

configurations of the particles, and the systems were relaxed. When we used the DF

curves obtained from the *C. elegans* embryos, we observed a tendency for the particles

to aggregate with an ovoid or distorted shape (Fig. 6A and S12). These results are

striking contrast to the simulation outcomes under the LJ or FH potential (Fig. S12).

These results suggest that the inferred effective DF curves are capable of recapitulating

the basic morphology of the *C. elegans* embryos.

Inference of effective force in mouse pre-implantation embryos

To further investigate whether effective forces could be obtained in

three-dimensional systems, we focused on mouse pre-implantation embryos, including

morulae at the 8-cell, compaction, and blastocyst stages (Fig. 4A, illustration). In 8-cell

stage embryos before compaction, cell–cell adhesion is weak, and individual cells can

be easily discerned (Fig. 4A, bright field). In the compaction-stage embryos composed

of ~16-32 cells, cell–cell adhesion becomes stronger due to elevated expression of

cadherin proteins on cell membrane, and the cells are strongly assembled (41). The

surface of the embryo becomes smooth, and the embryonic shape becomes more

spherical (Fig. 4A, bright field). In blastocyst-stage embryos composed of >64 cells, an

inner cavity is formed through liquid secretion from the cells, and the embryos expand

while maintaining their spherical shape. Trophectoderm (TE) cells form a single–

cell-layered structure at the outermost surface of the embryo, whereas the cells of the

inner cell mass (ICM) were tightly assembled, forming a cluster (Fig. 4A, illustration).

374   Figure 4B shows the snapshots of the inferred effective forces of cell–cell

interactions during the three embryonic stages that we examined (Movie S4D–G). The

uniqueness of the solutions was confirmed (Fig. S7B). We calculated the DF and DP

curves in a manner similar to *C. elegans* study described above, and obtained curves for

all three stages (Fig. 4C and D). The DF curves derived from the 8-cell and compaction

stages had typical profiles, with repulsive and attractive forces detected at short and

long interaction distances, respectively (Fig. 4D and S7A). On the other hand, the DF

curves derived from the blastocyst stage contained DEB (Fig. 4D, arrows), similar to

the case of simulation data considering a cavity (Fig. 2D). Moreover, when we inhibited

the cavitation by an inhibitor for Na^+/H^+ exchanger (S3226) (42), DEB was diminished

in the inferred potentials (Fig. S20), indicating that the cavity yielded the DEBs. These

results also indicate that the DF curves experience changes in their profiles during the

development.

Effect of inferred distance–force curves on mouse embryonic morphologies

We next asked whether the differences in the DF curves derived from the mouse embryos were sufficient to explain the key morphological features of the embryos. The DF curve derived from the compaction stage embryo yielded a spherical shape where particles were aggregated (Fig. 6B and S12). In the case of the 8-cell stage embryo, simulation results were similar to those obtained for the compaction stage, except that the spherical shape seemed to be slightly distorted (Fig. S12). In the case of the blastocyst stage embryo, when aggregated particles were given as initial configurations, the particles gradually scattered and formed a spherical structure harboring a cavity at the center (Fig. 6B). To examine whether DEB contribute to the formation of a cavity, we artificially eliminated DEB from the inferred DF curves, and performed a simulation. We found that no cavity was generated (Fig. 6B). These results suggest that all of the DF curves from the various embryonic stages are sufficient to explain their different morphologies. Nevertheless, because the DEB is not derived from a "real" cell–cell interactions but from the indirect effect of the cavity (i.e. external environment), we should notice that the predictive capabilities of the DP curves bearing such DEBs would be limited.

Inference of effective distance–force curves and their effect on morphology in MDCK

cyst

To further test the applicability of our method to structures harboring cavities,

we focused on MDCK cells, derived from dog kidney epithelium, which can form cysts

containing a cavity when cultured in suspension (Fig. 5A, illustration) or in gels

composed of extracellular matrix (ECM) (43). To exclude adhesion between the cysts

and the ECM, we chose suspension conditions in which a cyst can be assumed to be a

mechanically isolated system. Figure 5D and S10 shows the DF and DP curves from a

MDCK cyst. A DEB was detected in the DF curves, implying that a DEB is generally

yielded in cavity-bearing structures.

Simulations based on inferred DF curves revealed that, a cavity was stably

maintained in MDCK cells (Fig. 6C and S12). Moreover, the DF curves from which the

DEB was artificially eliminated could not maintain the cavity-bearing structure (Fig.

6C). Taken together, the DF curves with the DEBs in the blastocyst and the MDCK cyst

seem to reproduce the formation and maintenance of a cavity, respectively, though their

predictive capabilities would be limited due to the indirect effect of the cavities.

Perturbation experiments for mouse compaction stage embryo

To evaluate the predictive capability of the DF curves inferred by our method,

we performed perturbation experiments. In the mouse compaction stage, the cells are

tightly assembled to form a spherical and symmetric embryonic shape. In other words,

we think that there are two morphological features, 1) increase in symmetry of the

embryos, and 2) smoothing of the embryonic surfaces. The embryos are surrounded by

external structures called zona pellucida which are almost no effect on embryogenesis

(44 and references therein). Because both E-cadherin and actomyosin are essential for

the compaction process (41, 45, 46), we inhibited these proteins by chemicals: EDTA

for E-cadherin, cytochalasin D for F-actin, and blebbistatin for non-muscle myosin II

(Fig. 7A and S13).

The effective DP curves were inferred under these drugs (Fig. 7B and S13).

Then, we quantitatively evaluated the differences among the DP curves and found that

the distances at the potential minima were increased under the three inhibitors (Fig. 7C,

left panel). This result predicted that the cell–cell distances were increased, that was

supported through histological observations: each cell exhibited a nearly spherical shape

and seemed that the cells were not so densely assembled especially under the condition

of EDTA compared with the no-drugs condition (Fig. 7D and S16). The embryonic and

cellular shapes are schematically depicted in Figure 7E. In addition, we found another

quantitative difference among the DP curves. We measured distances at 10% energy of

potential minima as relative values to the distances at the potential minima, and found

that the relative distances were decreased under EDTA and blebbistatin (Fig. 7C, right

panel, and S15A-B for other % energies). This result means that attractive forces of

long-range interactions were reduced under these drugs. Several other parameters were

also tested (Fig. S15).

Predictive capability of effective potential for mouse compaction stage embryo

Simulations under the DP curves from the drugs-treated embryos were

performed. The cell particles were spherically or almost spherically assembled under the

potentials of the normal embryos or the cytochalasin D-treated embryos, respectively

(Fig. 8A). By contrast, the cell particles were not spherically assembled under the

potentials of the EDTA or blebbistatin-treated embryos. All simulation data are provided

in Figure S17.

To relate the profiles of the DP curves to morphological changes in the

compaction process, we first introduced two indices for the morphologies as follows.

Sphericity ($=(36\pi V^2)^{1/3}/S$) is essentially defined as the ratio of volume (V) to surface

area (S) so that the value becomes 1.0 for spheres (SI, Section 7-1). Shapes other than

spheres show values less than 1.0. Thus, both the increases in symmetry and in

smoothness of surfaces contribute to this index. Aspect ratio is defined as the ratio of

the length of the longest axis to that of the shortest axis in an ellipsoid which is fitted to

an object of interest (SI, Section 7-1). The increase in symmetry leads to that the aspect

ratio becomes 1.0, and otherwise the aspect ratio shows more than 1.0. The sphericities

of all simulation results obtained from the drug and no-drug conditions in Figure 8A and

S17 were measured, and we found that the sphericities were decreased under EDTA and

blebbistatin (Fig. 8B, left panel). Similarly, the aspect ratios were increased under

EDTA and blebbistatin (Fig. 8C, left panel).

We searched for control parameters which can describe the sphericities or

aspect ratios. Among several parameters tested (Fig S19), the distances at potential

minima measured in Figure 7C showed a tight relationship to the sphericities (Fig. 8B,

right panel). The sphericities showed an abrupt change along the x-axis. These results

indicate that the sphericity is the order parameter for the morphological transition of the

compaction process and that the distance at the potential minimum is the control

parameter. On the other hand, the relative distances at 10% potential minima showed a

clear relationship to the aspect ratios (Fig. 8C, right panel). Thus, the morphological

transitions of the compaction process are induced by the quantitative changes in the DP

curves.

Modeling of distance–force curves

481   To comprehensively understand the effect of the differences of the DF curves

on morphogenesis, we modeled the DF curves as a simple mathematical equation. We

selected the following equation as described in SI (Section 6):

$$F(D) = \varepsilon(D - D_0)^{-N} \cos\left\{\frac{2\pi(D - D_0)}{\delta}\right\}, \text{ for } D > D_0 \qquad \text{(Eq. 4)}$$

Here, F is the force, D is the distance, and N is the exponent of D, affecting the decay of

the DF curves along D (X-axis). D_0 and δ affect the profile of the DF curves as shown in

Figure 9A. Specifically, D_0 can transform the DF curves along D, and δ affects the

wavelength of the cosine function by which DEB is also introduced. All parameters and

variables related to lengths/distances were normalized by $(\delta/4)$, leading to generation of

dimensionless parameters and variables such as $D_0^* = D_0/(\delta/4)$. This enabled us to

present simulation outcomes in two-dimensional space $(D_0^*$ and $N)$ in a manner similar

to a phase diagram (Fig. 9).

Systematic analyses of morphologies generated by distance–force curves

495   Using the DF curves defined by Equation 4, we systematically investigated

what kind of morphologies could be generated or stably maintained, and what kind of

morphological transformation could be achieved. Here, we tried to identify possible

stable states, including metastable states. We set various initial configurations of

particles, performed simulations under various DF curves, and then searched for local

minimum states. A cavity-bearing structure was set as the initial configuration (Fig. 9B

and S21-I). The simulation outcomes included cavity-bearing structures ("cavity" and

"cavity with extra particles"), cups, tubes, and lattices / aggregates as possible stable

states. Multiple clusters of particles were formed or disorganized morphologies were

generated which include distorted rings, etc. Importantly, when the DEBs were

eliminated from the DF curves, one of the cavity-bearing structures ("cavity"), cups, or

tubes were not generated (Fig. 9C), indicating that the DEBs are essential for the

formation or maintenance of these structures.

Next, we tried other initial configurations of particles. A two-dimensional sheet

was applied as the initial configuration which corresponds to a monolayered cell sheet

(Fig. 10). The simulation outcomes included spherically aggregated structures,

cavity-bearing structures, and two-dimensional sheets (Fig. 10A). In contrast to Figure

9B, the cavity-bearing structures were newly generated even in the absence of the DEBs

(Fig. 10B). DF curves were found, which can maintain the two-dimensional sheet, and

the maintenance was largely dependent on the DEBs (Fig. 10A vs. 10B). We confirmed

that the generated two-dimensional sheets were not broken by slight positional

fluctuations (Movie S7), indicating that the structures were stable.

Other analyses considering different initial configurations and/or different

particle numbers are shown in Figure S21. We also found that modulation of the profiles

of the DF curves can induce transition among some 2D and 3D structures such as 2D

sheet, bending sheet, aggregate etc. (Fig. S22, Movie S6, and SI, section 7-3).

Discussion

In this study, we developed a top-down method for statistically inferring

effective forces of cell–cell interactions using nuclear tracking data obtained from

microscopic images, which is applicable for non-equilibrium systems. We then

demonstrated that effective potentials of cell–cell interactions can be extracted from the

inferred forces in various living systems. Our findings provide for the first time the

experimental quantification of effective potentials, which have been recognized

conceptually (7, 10). Furthermore, we identified distances at potential minima as a

control parameter for describing the morphological transition during the compaction

process of mouse embryos, where sphericity was defined as an order parameter. By

considering the qualitative and quantitative differences in the effective potentials, we

successfully reproduced various three-dimensional structures, including cell aggregates,

cavity-bearing structures, cups, tubes, two-dimensional sheets, etc. These structures are

often observed during embryogenesis, morphogenesis, and organoids prior to the

formation of complex tissue structures. For instance, cell aggregates are observed in

early embryos in various species and spheroids (47–50); cavity-bearing structures in

early embryos of various species, organoids, and cystic diseases (43, 47, 51–53); cell

sheets in epithelia and germ layers (49, 54–56); tubes in tubular organs (57, 58); and

cups in retina (i.e. optic cup) and murine epiblast (48, 59). Thus, we consider these

structure as basic structural units for morphogenesis. These results suggest that the

qualitative and quantitative differences in the effective potentials are powerful to

explain the emergent properties of multicellular systems.

Effect of external factors on effective potentials

We detected DEBs in the effective potentials obtained from blastocysts and

MDCK cysts, which are derived from indirect effect of cellular processes generating

cavities on cell–cell interactions. In the case of the blastocysts, the trophectoderm cells

transport liquid from the outside of the embryos to the cavities, resulting in increased

hydrostatic pressure, and subsequent expansion of the cavities (35–37). Therefore, the

TE cells push each other through the liquid in the cavities. This indirect mechanical

repulsion might be reflected in the DEBs (Fig. S20 and S23A). In the case of the

MDCK cysts, tubes, and sheets, the DEBs may reflect properties of cell sheets as

discussed in Figure S23B.

How should we evaluate these indirect effects? Because cavities are generated

by cellular activities through liquid secretion, DEB is derived from cell intrinsic features.

However, the appearance of the DEBs depends on the existence of such external factors,

meaning that the DEBs are not derived from "real" interactions and thus their effects are

context-dependent. Our simulations using the potentials with the DEBs successfully

generated cavities, implying that some external factors can be approximately

incorporated into effective potentials. Nevertheless, the predictive capabilities of such

potentials should be carefully assessed. Our goal is to develop a framework for

explaining diverse morphogenesis in a minimal model. If indirect cell–cell interactions

provided from external factors could be approximately incorporated into effective

potential, the model based on effective potential might effectively reduce degrees of

freedom.

Comparison of other inference methods

To understand morphogenetic events, mechanical states and simulations based

on those states are essential. Models such as vertex and Cellular Potts models are often

utilized, especially for epithelial cells (3, 51, 60). These models successfully

recapitulated various morphogenetic events, including the formation of the mouse

blastocyst (60) and very early embryogenesis in *C.elegans* (61). However, it is still

challenging to measure cellular and non-cellular parameters in three-dimensional

systems with high spatiotemporal resolution, although high-resolution inferences of

parameters such as cell–cell junction tensions and stresses have been performed in

two-dimensional situations (33, 62). Our method enables to identify a control parameter

from the effective potentials, and provides a framework for quantitatively connecting

model parameters to *in vivo* parameters under three-dimensional situations.

Conclusion

Our framework, composed of the inference method and the particle-based

model, has various potential uses for investigating the properties of cell–cell interaction,

586 material properties (Fig. 7C and S15), 3D morphogenesis (Fig. 6), 3D mechanical maps

587 (e.g. Fig. 3A), single-cell analysis of mechanics (e.g. Fig. 3A), relationships between

588 mechanical parameters and gene products (Fig. 7 and S5), and quantitative comparisons

589 of mechanical properties and behaviors among different living species and non-living

590 systems (Table S1, Fig. 6, and S12). Although some of these issues will be further

591 examined, we propose that our framework provides a novel physical scenario for

592 understanding the biophysics and physical biology of diverse morphogenetic events.

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Figures

Figure 1: Overview of strategy for inferring the effective potential of cell–cell

interactions

A. Cells were modeled as particles. A blue sphere corresponds to a cell. B. Distance–

force (DF) and distance–potential (DP) curves are considered. Typical profiles of the

curves, their physical meaning, and their origins are shown. C-I. Strategy for inferring

effective forces of cell–cell interactions is described. C. Mechanics of cell–cell

interactions is coarse-grained, leading to effective forces of cell–cell interactions as

shown in D. Detailed conceptual explanations are described in Figure S1. D. attractive

or repulsive force between the particles (blue spheres #1–4), was considered; force

vectors (F_i) are illustrated by black arrows in the case of particle #3. The net force (F_p)

is shown by a red vector, and the relationship between F_p and velocity (V_p) is described.

E. From a time series of microscopic images of nuclei, nuclear tracking data were

obtained. F. Effective forces of cell–cell interactions were inferred by solving an inverse

problem. The inference method is explained in Figure S2. G. The inferred force values

(F) were plotted against the distance between cell–cell interactions, and whether DF or

DP curves are emerged from the plots were examined. H. Whether the DF or DP curves

can reconstitute tissue shape or cell dynamics was examined. Components related to

cell–cell interactions described in C can be decoded from the DF or DP curves was

tested.

Related figure: Figure S1 and S2.

Figure 2: Validation of inference method using simulation data

Simulations were performed on the basis of a distance–force (DF) curve obtained from

the Lenard-Jones potential (LJ). The simulation conditions contain "Steady state" (A),

"Cell proliferation" (B), "Egg shell-like spatial constraints" (C), and "Cavity-like spatial

constraints" (D). The left panels in A-D are snapshots of the simulations. The right

panels in A-D are DF curves obtained from the inference. Effective forces of

interactions for each pair of cells for each time frame were plotted against the distance

of the cell–cell interactions, and shown as a colored heat map representing the

frequencies of the data points (Frequency index (FI)). Averaged values of the forces are

shown in yellow (binned average). Detailed procedures are described in SI (Section 5).

In some graphs, the DF curves from the LJ potential are overlaid by orange broken lines

(LJ). In A and B, simulation data with different sampling intervals were applied. In C,

under the cylindrical constraint with a shorter length where particles are highly

compressed, the DF curve was shifted leftward along the distance (white arrow). In D,

on the expanding cavities, repulsive forces at distant regions were detected, which we

called distant energy barriers (DEB). Related comprehensive analyses are shown in

Figure S4 which also demonstrated that the DEBs are derived from the cavities.

Detailed assumptions and conditions of the simulations are described in SI, Section 7-4.

Related figures: Figure S3, S4, and S5. Related movies: Movie S1A-C, S2A-B, and

S8A-D.

Figure 3: Inference of effective force of cell–cell interaction in *C.elegans* embryos

A. Snapshot of the nuclear positions in the *C.elegans* embryo (left panel) and snapshots

with inferred effective forces (right three panels) were three-dimensionally visualized at

the time frames as described (t16, t76, and t195); the interval between time frames is 1

min. The spheres correspond to cells whose colors represent cell lineages, including AB,

C, D, E, MS, and P & Z. The lines represent cell–cell interactions; the colors indicate

the values of the effective forces (red, attractive; blue, repulsive). Forces are depicted in

arbitrary units (A.U.); 1 A.U. of the force can move a particle at 1μm/min as described

in SI (Section 4-4). The nuclear tracking data were obtained from a previous report (26).

Related figures and movies are provided (Fig. S6A, B, Movies S3A and B). B.

Uniqueness of solution of effective force inference was examined. The minimizations of

Equation S6 were performed from different initial force values as described in the x-

and y- axes, and the inferred values of each cell-cell interaction were plotted by crosses.

C. The inferred effective forces of cell–cell interactions were plotted against the

distance of cell–cell interactions in a manner similar to Figure 2. The graphs for time

frame 76–115 and 156–195 are shown. Data at other time frames are provided in Figure

S6C. D. Distance–force (DF) and distance–potential (DP) curves were estimated.

Details are described in SI (Section 5-3). The curves for each time frame are shown. E.

Mean diameters of cells at each time frame were estimated from cell numbers and the

DP curves. Given that the volume of the embryos is constant (= *Vol*) during

embryogenesis, the mean diameters were estimated from the cell numbers (N_c) at each

time frame as follows: mean diameter = $\{Vol\ /\ (4/3\ \pi\ N_c)\}^{1/3}$. The diameters relative to

that at time frame = 16 are shown with cell numbers (parentheses). The sizes of the

circles reflect the diameters, whose colors roughly correspond to the colors in the graph

in D. The diameters were also estimated from the DP curves in D; the distances at

minima of the DP curves roughly reflect the diameters (Fig. 1B and S2D).

Related figure: Figure S6. Related movies: Movie S3A-B.

Figure 4: Inference of the effective force of cell–cell interaction in mouse

pre-implantation embryos

A. Eight-cell, compaction, and blastocyst stages of mouse embryo are illustrated, and

their confocal microscopic images are shown: bright field, maximum intensity

projection (MIP) and cross-section of fluorescence of H2B-EGFP. Snapshots of nuclear

tracking are also shown; blue spheres indicate the detected nuclei. In the nuclear

tracking image of the blastocyst, ICM cells are located around the bottom region of the

image, as indicated by *. Scale bars = 15μm. Related movies are provided (Movie S4A–

C). B. Snapshots of nuclear positions with inferred effective forces were visualized

three-dimensionally for the three embryonic stages in a manner similar to Figure 3A.

Related movies are provided (Movie S4D-G). The uniqueness of the solutions was

confirmed in Figure S7B. C. The inferred effective forces of cell–cell interactions were

plotted against the distance of cell–cell interactions in a manner similar to Figure 2. D.

Distance–force (DF) and distance–potential (DP) curves were estimated from (C) in a

manner similar to Figure 3D. A distant energy barrier (DEB) was detected, as indicated

by arrows. Data from other embryos are provided in Figure S7A. The differences in DF

and DP curves between ICM and TE cells and between the outer and inner cells in the

compaction state were described in Figure S8. DP curves between adjacent cells are

provided in Figure S9. DP curves from the embryos whose cavitation was inhibited by

chemicals are shown in Figure S20.

Related figures: Figure S7, S8, S9, and S20. Related movies: Movie S4A-G.

Figure 5: Inference of effective force of cell–cell interaction in cysts formed by MDCK

cells

A. MDCK cysts formed under suspension conditions are illustrated, and confocal

microscopic images are shown: bright field, maximum intensity projection (MIP) and

cross-section of fluorescence of H2B-EGFP. A snapshot of nuclear tracking is also

shown; blue spheres represent nuclei. Scale bars = 10μm. A related movie is provided

(Movie S5A). B. A snapshot of the nuclear positions with inferred effective forces was

three-dimensionally visualized in a manner similar to Figure 3A. The blue spheres

correspond to cells. The lines represent cell–cell interactions and the colors indicate the

values of the effective forces [red, attractive (attr.); blue, repulsive (repl.)]. A related

movie is provided (Movie S5B). The uniqueness of the solutions was confirmed in

Figure S10B. C. The inferred effective forces of cell–cell interaction were plotted

against the distance of cell–cell interaction in a manner similar to Figure 3C. D.

Distance–force and distance–potential curves were estimated from (C) in a manner

similar to Figure 3D. DEB was detected as indicated by arrows. Data from other cysts

are provided in Figure S10A. DP curves between adjacent cells are provided in Figure

S11.

Related figures: Figure S10 and S11. Related movies: Movie S5A-B.

Figure 6: Outcomes of simulations based on inferred distance–force curves

Simulations results under the DF curves derived from *C. elegans* (A) and mouse

embryos (B), and the MDCK cyst (C) are shown. Blue spheres correspond to cells. In

the *C. elegans* embryos, the DF curves were derived from time frames 36–75 and 156–

195 in Figure 3D. In the mouse embryos, the DF curves were derived from the

compaction and blastocyst stages in Figure 4D. Cross-sections are also visualized. In

the case of the mouse blastocyst, two parallel cross-sections (shown in blue and green)

are merged. In the MDCK cyst, the DF curves are derived from Figure 5D. In the

blastocyst and the MDCK cyst, the DF curves without DEB were also applied (w/o

DEB). The diameters of the blue spheres are set to be equivalent to the distance at the

minimum potential (Fig. S2D). The initial configurations of the simulations are

described in Figure S12, along with other simulation results.

Related figure: Figure S12.

Figure 7: Inference of effective potentials of cell–cell interaction in compaction-inhibited mouse embryos

A. Microscopic images of embryos are shown, where the compaction states were inhibited by chemicals. The experimental design is described in Figure S13. CytoD, cytochalasin D; Bleb, blebbistatin. B. DP curves of the drug-treated embryos are shown. N=6~10 for each condition. Enlarged views are provided in Figure S14. C. Quantitative

differences in the DP curves were calculated. Distances at the potential minima in the

DP curves are plotted (left panel). Distances providing 10% energy of the potential

minima are plotted as the relative value to the distances at the potential minima (right

panel). Mann–Whitney–Wilcoxon tests were performed (SI, Section 5-4-2) and the

resultant p-values for "No Drugs" vs. "EDTA", vs. "CytoD", and vs. "Bleb" are 0.011,

0.23, and 0.00040, respectively, in the left panel. The p-values in the right panel are

shown in Figure S15B. Other quantification data are provided in Figure S15, S24, and

Table S1. D. Confocal microscopic images of cell shapes in the drug-treated embryos

are shown. Cell shapes were visualized by staining F-actin or E-cadherin. Other images

of staining are provided in Figure S16. E. The embryonic and cellular shapes in the

drug-treated embryos are illustrated based on D.

Related figures: Figure S13, S14, S15, S16, S24, and Table S1.

Figure 8: Simulations under inferred distance–potential curves in drug-treated embryos,

and identification of parameters explaining morphological transition

A. Simulation results for the drug-treated embryos are shown. All simulation data are

provided in Figure S17, and the representative results which nearly showed the mean

values of the sphericities for each condition are chosen here. Simulation results from a

different initial configuration was provided in Figure S17. B. Sphericities in the

simulations of the drug-treated embryos were calculated (left panel), and the sphericities

were plotted against the distances at the potential minima calculated in Figure 7C.

Mann–Whitney–Wilcoxon tests were performed (SI, Section 5-4-2) and the resultant

p-values for "No Drugs" vs. "EDTA", vs. "CytoD", and vs. "Bleb" are 0.011, 1.0, and

0.00080, respectively, in the left panel. Relationships between the sphericities and other

parameters are shown in Figure S19. C. Aspect ratios in simulations of the drug-treated

embryos were calculated (left panel), and the aspect ratios were plotted against the

relative distances providing 10% energy of the potential minima which were calculated

in Figure 7C. Mann–Whitney–Wilcoxon tests were performed (SI, Section 5-4-2) and

the resultant p-values for "No Drugs" vs. "EDTA", vs. "CytoD", and vs. "Bleb" are 0.21,

0.97, and 0.33, respectively, in the left panel. Relationships between the aspect ratios

and other parameters are shown in Figure S19. Sphericities and aspect ratios from the

different initial configuration of the simulation are provided in Figure S18.

Related figures: Figure S17, S18, and S19.

Figure 9: Modeling of distance–force curves and diagrams of simulation outcomes

A. DF curves were mathematically modeled using four parameters: N, D_0, δ, and ε. In

the case of the DF curves without the cosine term, the curves ($F = \varepsilon (D - D_0)^{-N}$) are

always >0 (repulsive) (green broken line). Upon introduction of the cosine (purple

broken line), the curves acquire regions with <0 (attractive) and DEB (red line). Upon

introduction of D_0, the curves are translated along the distance (blue line). N affects the

gradient of the curves and the height of DEB (right panel). DF curves from $D = 0$ to D

at the end of DEB (w/ DEB) or the curves from $D = 0$ to D before starting DEB (w/o

DEB) were used in the following simulations. B. Outcomes of simulations under

various values of N and D^* are shown in the diagrams. A cavity-bearing structure was

given as an initial configuration; the blue spheres correspond to cells, and a

cross-section is visualized. According to the conditions, a cavity was maintained,

whereas cups, tubes, etc. were generated. Some conditions on the diagram were

designated by symbols (circle, square, and triangle), and their simulation results are

visualized (blue spheres). Example profiles of the DF curves are shown. The diameters

of the blue spheres are set to be 0.5× the distance at the minimum potential, which

corresponds to the distance with $F = 0$ (arrows in the profiles of the DF curves, Fig.

S2D). In the case of the cup, the mouth of the cup is marked by orange rings, and the

particles located on the ring are labeled by white dots. Note that the particles labeled

particles by crosses are located neither on the ring nor at the center of the ring. In the

case of the tube, the open ends of the tube are shown by orange arrows. In the case of

the cavity with extra particles, the extra particles are shown by arrows. The definition of

each phase (i.e. the cavity, cup, etc.) is explained in SI (section 7-2). Data points for

depicting the lines separating each phase are shown by gray crosses. A phase diagram

under different numbers of particles is provided in Figure S21-I-A, where the overall

relative positions among the phases were conserved though the parameter values for

generating each phase became different. C. Simulation outcomes are shown on

diagrams similar to B, except that the simulations were performed without DEBs.

Related figure: Figure S21-I.

Figure 10: Simulation outcomes and their diagrams, based on distance–force curves;

starting from 2D sheet

A. Simulations were performed from a two-dimensional (2D) sheet with slight

positional fluctuations under various distance–force curves defined by Equation 4 (Fig.

9A), and a diagram of the outcomes is shown. The number of particles was set to 25.

Cross-sections are also presented (blue circles). According to the conditions, the 2D

sheet was maintained, whereas aggregates, cavity-bearing structures etc. were generated.

The diameters of the blue spheres are set to be 0.2× the distance at the potential

minimum.

B. Similar to A, except that the simulations were performed without distant energy

barriers (DEBs). 2D sheets were maintained, but the area in the diagram became

significantly smaller than that in A. Other diagrams are provided in Figure S21. Possible

shape transformations are exemplified in Figure S22.

Related figures: Figure S21-II, S22, and S23. Related movie: Movie S6.

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