

# The emergence of geometric order in proliferating metazoan epithelia.

by Gibson *et al* – Computational Biology and Bioinformatics Seminar

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Philip Hartout

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ETH Zurich

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# Introduction

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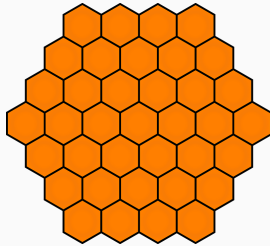
*'The organisation of cells into epithelial sheets is an essential feature of animal Fdesign.'*

Gibson et al [2]

Predictable geometries arise when *minimising surface energy* or *maximising space filling* in biological and non-biological structures:

Examples

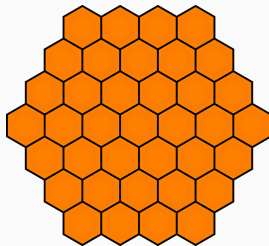
- *Drosophila Melanogaster*'s retinal cells
- Honeycombs
- Coins on tabletops



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How is the geometry of *growing* epithelial cells organized and how can it be explained?

# Geometry of *growing* epithelial cells

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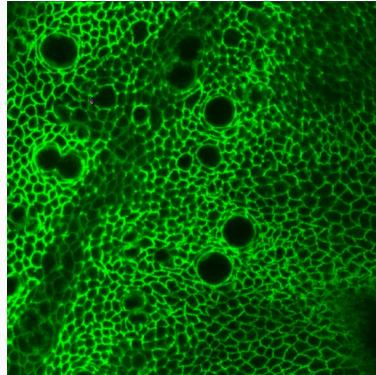
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# Geometry of *growing* epithelial cells

Time-lapse movies collected revealed the following findings:

- Initially polygonal prophase cells rounded up and divided
- Cell-neighbour relationships were stably maintained throughout the cell cycle
- The topology of the divided cells is invariant over time.



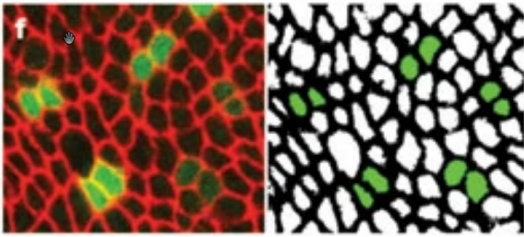
**Figure 1:** General hexagon-dominated topology of growing epithelial in *Drosophila Melanogaster*.

## Model Assumptions

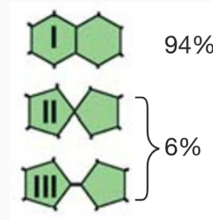
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# Post-mitotic relationship between two daughter cells

Stochastically marked cell lineages with GFP revealed stable post-mitotic patterns.



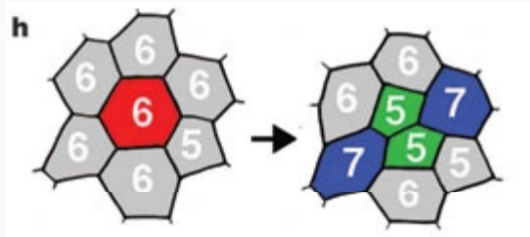
**Figure 2:** Daughter-cell relationship.



**Figure 3:** Observed post-mitotic relationship between daughter cells.

# Typical topology changes of a dividing cell

**Figure 4:** Topology changes during cell division



# **A Discrete Markov Model for Proliferating Epithelia**

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# A discrete Markov model to model stochastic cell proliferation

$s$	state of the cell as its number of sides, where $s > 3$ .
$p_s$	the relative frequency of $s$ -sided cells in the population
$\mathbf{p}^{(t)}$	the infinite row vector $\mathbf{p}^{(t)} = [p_4, p_5, p_6, \dots]$ state of the population at generation $t$ .
$\mathbf{p}^{(t+1)} = \mathbf{p}^{(t)}PS$	the state dynamics, where $P$ and $S$ are the probabilistic transition matrices.
$P_{ij}$	the probability that an $i$ sided cell divides to produce a $j$ -sided daughter cell.
$S_{ij}$	the probability of an $i$ -sided cell will gain sides from dividing neighbour cells divisions to become $j$ -sided.

## Derivation of Markov state dynamics - transition matrix $P$

- $P_{ij}$ : the probability that an  $i$  sided cell divides to produce a  $j$ -sided daughter cell.
- $K_t$ : the number of junctions distributed to one daughter cell on division at generation  $t$ .  $s_{t-1} - K_t$  are left for the other.



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- Because no triangular cells are observed, each daughter receives at least two junctions, leaving  $s_t - 4$  junctions to be distributed among the daughters with equal probability.
- So, the number of junctions received by the first daughter is  $K_t - 2 \sim \mathcal{B}(n = s_{t-1}, p = \frac{1}{2})$ .

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- Therefore, the unnormalized entries of  $P$  are the coefficients of Pascal's triangle:

$$P = \begin{matrix} & 4 & 5 & 6 & 7 & 8 & 9 & \dots \\ \begin{matrix} 4 \\ 5 \\ 6 \\ 7 \\ 8 \\ 9 \\ \vdots \end{matrix} & \left( \begin{array}{cccccc} 1 & & & & & \\ 1 & 1 & & & & \\ 1 & 2 & 1 & & & \\ 1 & 3 & 3 & 1 & & \\ 1 & 4 & 6 & 4 & 1 & \\ 1 & 5 & 10 & 10 & 5 & 1 \\ & & & & & \ddots \end{array} \right) \end{matrix}$$

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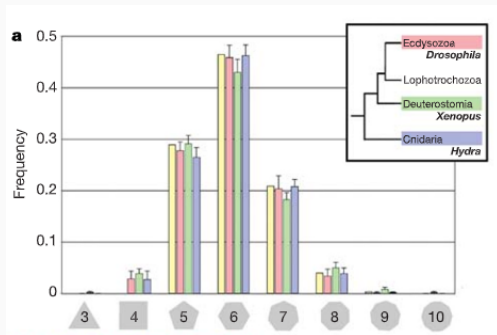
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- On average the number of sides gained per cell is  $\frac{2N}{2N} = 1$ . That is, a cell will gain 1 side every cycle on average.
- Thus,  $S_{ij} = 1$  if  $j = i+1$  and 0 otherwise.

$$S = \begin{matrix} & \begin{matrix} 4 & 5 & 6 & 7 & 8 & 9 & \dots \end{matrix} \\ \begin{matrix} 4 \\ 5 \\ 6 \\ 7 \\ 8 \\ 9 \\ \vdots \end{matrix} & \begin{pmatrix} 0 & 1 & & & & & \\ & 0 & 1 & & & & \\ & & 0 & 1 & & & \\ & & & 0 & 1 & & \\ & & & & 0 & 1 & \\ & & & & & 0 & 1 \\ & & & & & & 0 & 1 \\ & & & & & & & \ddots \end{pmatrix} \end{matrix}$$



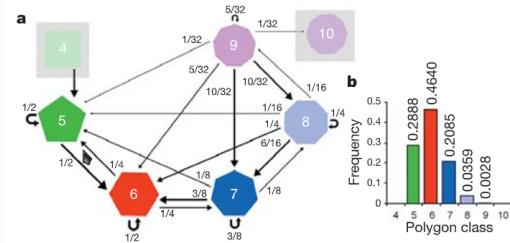
## Predicting the equilibrium $E$ from $T = PS$

The Perron-Frobenius theorem guarantees that a Markov chain will converge to a unique equilibrium  $\mathbf{p}^*$  where  $\mathbf{p}^*$  is the principal eigenvector of the transition matrix  $T$ .

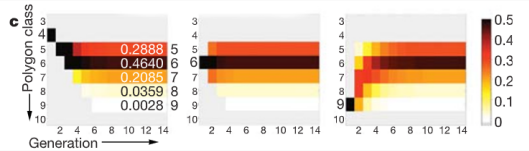


**Figure 5:** Distribution of the predicted and observed polygons

# Robust equilibrium topology



**Figure 6:** Discrete state dynamics

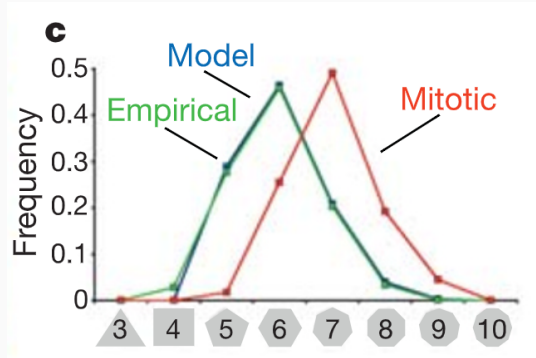


**Figure 7:** Figure showing the same attained equilibrium, no matter the initial conditions

## Additional insights on mitotic cells

As seen in the construction of  $S$ , the average mitotic cell gains approximately one side, and daughter cells have one less side. Experimental evidence confirms this:

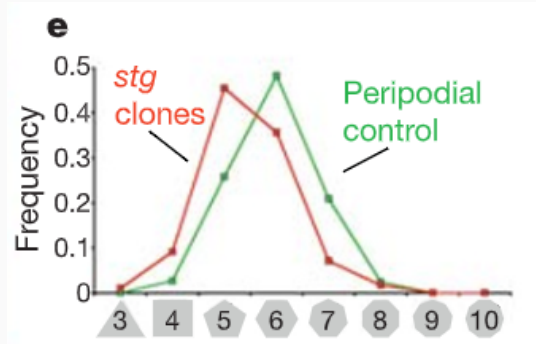
<b>Non-mitotic cells</b>	$(5.94 \pm 0.06)$
<b>Mitotic cells</b>	$(6.99 \pm 0.07)$



**Figure 8:** Polygon distributions of mitotic cells vs. non-mitotic cells

## Polygon distribution of mitotic cells in non-proliferating tissues

The polygon distribution of proliferating cells in a non-proliferating tissue is not the same as that in proliferating tissue. This is demonstrated by forcing mitosis (using a protein named *string*, abbreviated *stg*) in some cells in quiescent tissue.



**Figure 9:** Distribution of polygons in non-proliferating tissue

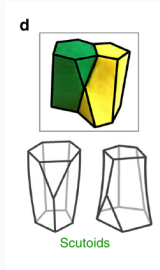
→ Differential proliferation influences cell shape and morphogenesis (not modelled here).

## Subsequent works

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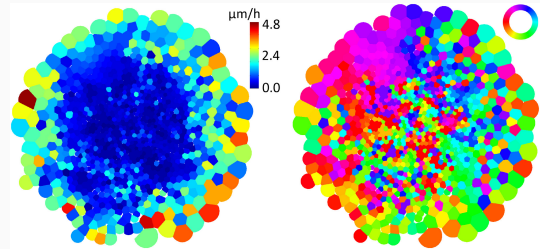
# Subsequent work

Work on 3D structures:



**Figure 10:** The scutoid has been discovered as a solution to 3D cell packing.[3]

Work on dynamics



**Figure 11:** Modelling and prediction of speed and direction of proliferating epithelia. [1]

## Take home messages

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The proposed **discrete Markov chain**:

1. shows how epithelial topology can be irregular, but **not random**.
2. **explains the hexagonal topology** of growing epithelia
3. predicts the **overall distribution** of polygonal cell types
4. shows an **emergent mechanism** by which epithelia accommodate **rapid proliferation** while maintaining **structural integrity**.
5. Provides a **framework** for investigating other models for **cell division** such as **cleavage plane** choices or **aberrant cell division**.

**Questions?**

## Backup slides - Assumptions of the recurrence system

From the aforementioned observations we make the following assumptions about the mathematical modelling of cell epithelia:

- Cells are polygons with a minimum of four sides
- Cells do not resort
- Mitotic siblings retain a common common junctional interface
- Cells have asynchronous but uniform cell cycle times
- Cleavage places cut a side rather than a vertex of the mother polygon
- Mitotic cleavage orientation randomly distributes existing tricellular junctions to both daughter cells.

# Recurrence system

**Table 1:** Cell features, topological equivalence and evolution at division  $t$




Cell feature	Graph equivalent	Evolution at division $t$
Tricellular junction	Vertex, $v_t$	$v_t = v_{t-1} + 2f_t$
Cell side	Edge, $e_t$	$e_t = e_{t-1} + 3f_{t-1}$
Apical cell surface	Face, $f_t$	$e_t = e_{t-1} + 3f_{t-1}$

So we can construct the following system:

$$s_t = \frac{2(e_t + 3f_{t-1})}{2f_{t-1}} = \frac{s_{t-1}}{2} + 3$$

which exponentially converges to 6 provided the initial condition:

$$s_t = 6 + 2^{-t}(s_0 - 6)$$

-  S. Aland, H. Hatzikirou, J. Lowengrub, and A. Voigt.  
**A mechanistic collective cell model for epithelial colony growth and contact inhibition.**  
*Biophysical journal*, 109(7):1347–1357, 2015.
-  M. C. Gibson, A. B. Patel, R. Nagpal, and N. Perrimon.  
**The emergence of geometric order in proliferating metazoan epithelia.**  
*Nature*, 442(7106):1038–1041, 2006.
-  P. Gómez-Gálvez, P. Vicente-Munuera, A. Tagua, C. Forja, A. M. Castro, M. Letrán, A. Valencia-Expósito, C. Grima, M. Bermúdez-Gallardo, Ó. Serrano-Pérez-Higueras, et al.  
**Scutoids are a geometrical solution to three-dimensional packing of epithelia.**  
*Nature communications*, 9(1):1–14, 2018.