Collective sensing by communicating cells

Andrew Mugler Purdue Physics

Cheung et al, *Cell*, 2013

50 µm

From communication to collective behavior



Physics of collective cell behavior



Physics of collective cell behavior



Gradient sensing





Bacteria

п



Should the minimum run time be... ...long? ...or short?





PHYSICS OF CHEMORECEPTION

BIOPHYSICAL JOURNAL VOLUME 20 1977





Howard Berg

Edward Purcell

It's all about counting molecules.

A good measurement requires

$$\bar{n}_2 - \bar{n}_1 > \sigma_{n_2 - n_1}$$



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Poisson statistics:

$$\bar{n} \sim a^3 c \qquad \sigma^2 = \bar{n}$$

Diffusive refreshing:

$$\sigma^2 \to \frac{\bar{n}}{T/(a^2/D)}$$

"Berg-Purcell limit":

$$\frac{\sigma}{\bar{n}} \sim \frac{1}{\sqrt{TDac}}$$

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The Berg-Purcell limit persists



Using the fluctuation-dissipation theorem:

Bialek & Sateyeshgar, PNAS, 2005

$$\frac{\delta c}{c} = \sqrt{\frac{1}{\pi D\sigma cT} + \frac{2}{k_a c(1-\overline{n})T}}.$$



Using reaction-diffusion theory:

Kaizu et al, Biophys J, 2014

$$\frac{\delta c}{c} = \sqrt{\frac{1}{2\pi\sigma Dc(1-\overline{n})T}} + \frac{2}{k_a c(1-\overline{n})T}.$$

Berg-Purcell term sets the "noise floor"





Gradient sensing in larger cells



Difference in molecule numbers:

$$\Delta \bar{n} \sim a^3 \Delta c = a^3 g a$$

$$\sigma \sim \frac{a^3 c}{\sqrt{T D a c}}$$

Error in gradient sensing:

$$\frac{\delta g}{g} = \frac{\sigma}{\Delta \bar{n}} \sim \sqrt{\frac{c}{g^2 a^3 T D}}$$

Or
$$\left| \frac{\delta g}{c/a} \sim \frac{1}{\sqrt{TDac}} \right|$$

Berg-Purcell limit

Once again, the Berg-Purcell limit persists



Using an analogy to electrostatics:

Endres & Wingreen, PNAS, 2008

$$\frac{\langle (\delta c_{\vec{r}})^2 \rangle}{(c_0/a)^2} = \frac{1}{4\pi Dac_0 T},$$



Using the fluctuation-dissipation theorem: Endres & Wingreen, *Prog Biophys Mol Biol*, 2009

$$\frac{\left< \left[\delta(c_1 - c_2) \right]_{\tau}^2 \right> / r^2}{(c_0 / r)^2} = \frac{1}{\pi D a' c_0 \tau}$$



Modeling as an Ising spin chain: Hu et al, *PRL*, 2010

$$\sigma_{p,\mathcal{T}}^2 \gtrsim \frac{8}{\pi \mathcal{T} DLC_0}$$

Amoebae appear to approach the limit



Physics of collective cell behavior

What are the fundamental limits to cellular sensing?

Concentration sensing (Berg & Purcell, 1977)



Gradient sensing



Can cells surpass these limits by communicating?





Cell-cell communication

Juxtacrine signaling (short-range)



Autocrine signaling (long-range)

Bacteria

Youk & Lim, *Science*, 2014



Amoebae

Mehta & Gregor, *Curr Opin Gen Dev*, 2010



Outline of this talk



Concentration sensing by a single cell



 $\begin{array}{lll} \boldsymbol{C} + \boldsymbol{U} \stackrel{\tilde{\alpha}}{\underset{\mu}{\longrightarrow}} \boldsymbol{R} & & \boldsymbol{R} \stackrel{\beta}{\to} \boldsymbol{R} + \boldsymbol{M} \\ & & \boldsymbol{M} \stackrel{\nu}{\to} \boldsymbol{\emptyset} \end{array}$

$$\dot{c} = D\nabla^2 c - \delta(\vec{x})\dot{r} + \eta_c$$

$$\dot{r} = \alpha c(\vec{0}, t) - \mu r + \eta_r \qquad (\alpha \equiv \tilde{\alpha}r_{\text{total}})$$

$$\dot{m} = \beta r - \nu m + \eta_m$$

$$\frac{\sigma_m^2}{\bar{m}^2} = \frac{1}{2} \frac{1}{\pi a \bar{c} D T} + \frac{2}{\mu T \bar{r}} + \frac{2}{\nu T \bar{m}}$$
extrinsic noise intrinsic noise

See also:

- Gardiner, Handbook of stochastic methods, 1985
- Bialek & Setayeshgar, PNAS, 2005

Outline of this talk



Conc. sensing with short-range communication



$$\dot{c} = D\nabla^2 c - \sum_{i=1}^N \delta(\vec{x} - \vec{x}_i)\dot{r}_i + \eta_c$$
$$\dot{r}_i = \alpha c(\vec{x}_i, t) - \mu r_i + \eta_{ri}$$
$$\dot{m}_i = \beta r_i - \nu m_i + \gamma \sum_{j \in \mathcal{N}_i} (m_j - m_i) + \eta_{mi}$$

$$N = 2: \quad \frac{\sigma_m^2}{\bar{m}^2} = \frac{3}{8} \frac{1}{\pi a \bar{c} D T}, \quad \gamma \gg \nu$$

Note: $\frac{3}{8} > \frac{1}{2} \times \frac{1}{2}$ single two cell cells

Outline of this talk



Conc. sensing with long-range communication



Conc. sensing with long-range communication



- Long-range communication outperforms short-range, even for small populations
- Optimal separation can be many cell radii



PNAS

Glioblastoma cellular architectures are predicted through the characterization of two-cell interactions

Nataly Kravchenko-Balasha^{a,1}, Jun Wang^{a,1}, Francoise Remacle^{b,c}, R. D. Levine^{c,d,2}, and James R. Heath^{a,d,2}





Collective ATP sensing in fibroblasts



Outline of this talk





Collective sensing

EGF (0.5 nM/mm) Single cells Organoid + Endothelin-I нон нон -0-1 100 µm unbiased unbiased right unbiased right left left left right

Gradient sensing: Berg-Purcell estimate



Difference in molecule numbers:

$$\Delta \bar{n} \sim a^3 \Delta c = a^3 g(Na)$$

Diffusive fluctuations:

$$\sigma \sim \frac{a^3 c}{\sqrt{T D a c}}$$

Error in gradient sensing:

$$\frac{\sigma_g}{g} = \frac{\sigma}{\Delta \bar{n}} \sim \frac{1}{gNa} \sqrt{\frac{c}{TDa}}$$

Gradient sensing: Berg-Purcell estimate



Compartments need to *communicate* to integrate information.

Gradient sensing w/ short-range communication



"Local excitation—global inhibition (LEGI)":



Gradient sensing w/ short-range communication





$$\dot{c} = D\nabla^2 c - \sum_{i=1}^N \delta(\vec{x} - \vec{x}_i)\dot{r}_i + \eta_c$$
$$\dot{r}_i = \alpha c(\vec{x}_i, t) - \mu r_i + \eta_{ri}$$
$$\dot{\ell}_i = \beta r_i - \nu \ell_i + \eta_{\ell i}$$
$$\dot{m}_i = \beta r_i - \nu m_i + \gamma (m_{i-1} + m_{i+1} - 2m_i) + \eta_{mi}$$

$$\Delta_N = \ell_N - m_N$$



For 2D, 3D systems, see: Smith*, Fancher*, Levchenko, Nemenman, Mugler, *Phys Biol*, 2016









Comparing theory with experiment



Communication length scale: $2.9 < n_0 < 4.2$ cells

What is the communication molecule?

Gap-junction blocker: 50 nM Endothelin-I



Same for:

- 50 µM Carbenoxolone
- 50 µM Flufenamic acid
- 0.5 mM Octanol

Calcium depletion: 100 nM Thapsigargin









Collective cell migration



Cellular Potts model:

$$U = \sum_{\langle x, x' \rangle} J_{\sigma(x), \sigma(x')} + \lambda \sum_{i} \delta A_{i}^{2}$$

Current work w/ Bumsoo Han, Purdue:





Outline of this talk







- Simple models provide powerful bounds on biological information processing
- Communication allows collective systems to outperform single cells
- Long-range communication can reduce measurement correlations, leading to optimal cell separations
- Communication is ultimately imperfect, which fundamentally limits sensory precision

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Fancher, Mugler, arXiv:1603.04108

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