

# Some Styles of Theory in Biophysics

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CAL POLY POMONA

Strike! Strike! Strike! —————

# Strike! Strike! Strike!

General models→Often fundamental limits

The most reductionist level→Molecular modeling

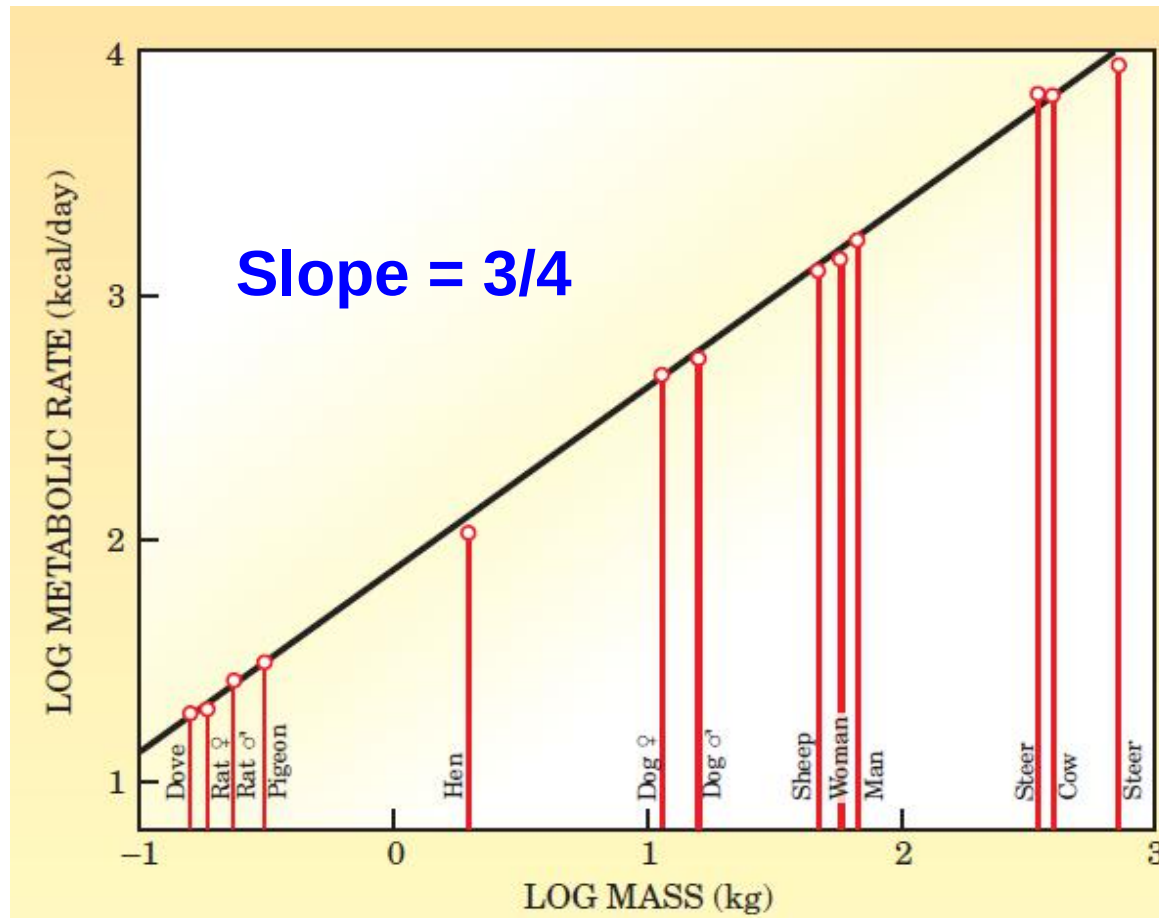
Modeling a single process

Models of experiments  **Phenomenology**  
**Information limits**

My own work on superresolution

# How I got into biophysics

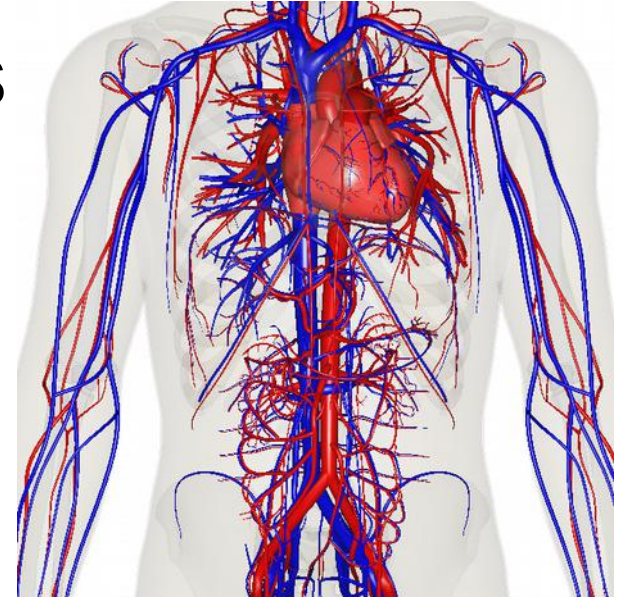
- 1997: West, Brown, Enquist (mostly) explain this:



- But heat loss  $\propto$  area  $\propto m^{2/3}$  ???

# Model Assumptions

1) Fractal network of blood vessels

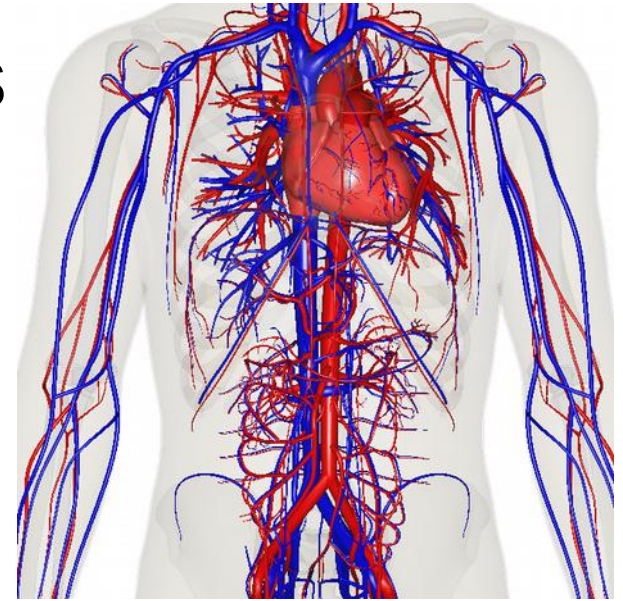


# Model Assumptions

1) Fractal network of blood vessels

2) Invariant terminal units

Capillaries have to be  $\sim 100\ \mu\text{m}$  apart and thick enough for a red blood cell

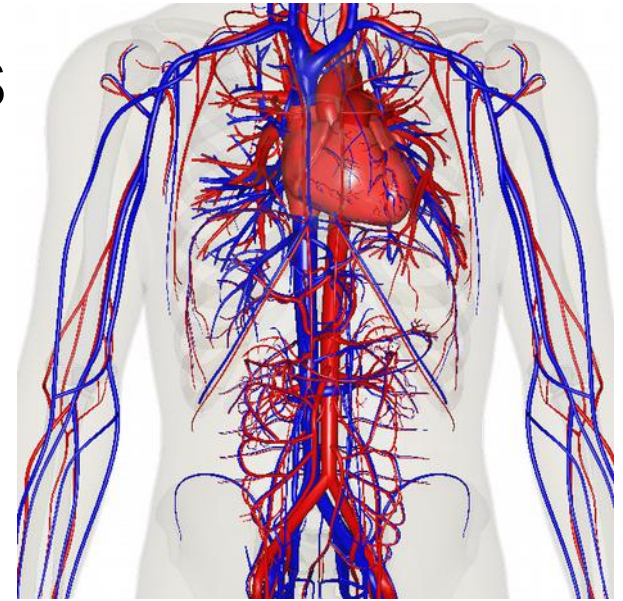


# Model Assumptions

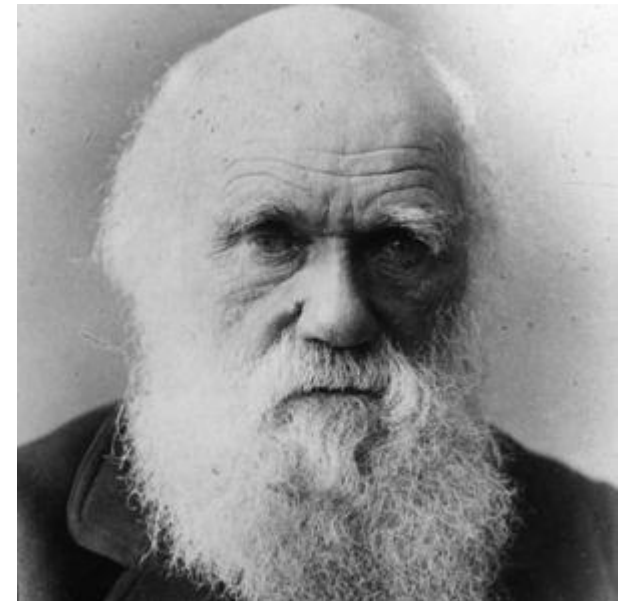
1) Fractal network of blood vessels

2) Invariant terminal units

Capillaries have to be  $\sim 100\ \mu\text{m}$  apart and thick enough for a red blood cell



3) Minimize energy dissipation



# (Caveats)

- The  $\frac{3}{4}$  scaling law is more accurate for large mammals than small ones.
- The fluid dynamics assumptions in WBE are only valid asymptotically.
- WBE errs in wrong directions for small mammals.
- More accurate models fix it.



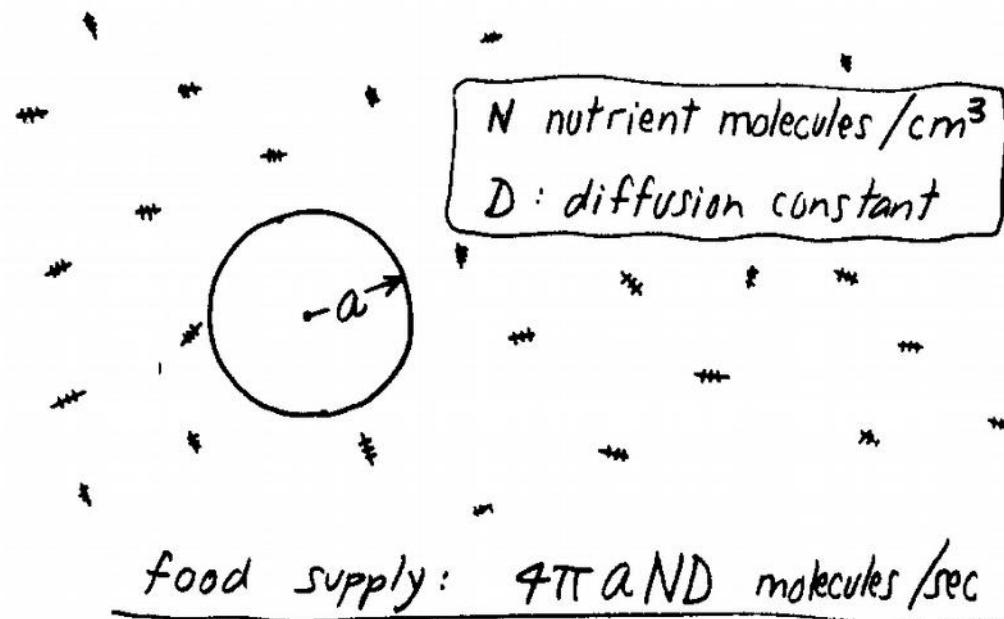
# Biophysics of fundamental questions

## Life at low Reynolds number

E. M. Purcell

*Lyman Laboratory, Harvard University, Cambridge, Massachusetts 02138*

*American Journal of Physics, Vol. 45, No. 1, January 1977*



# Biophysics of fundamental questions

Minimize conduction delays  
Maximize number of connections  
Minimize wire length

→ Brain should be 60% wire

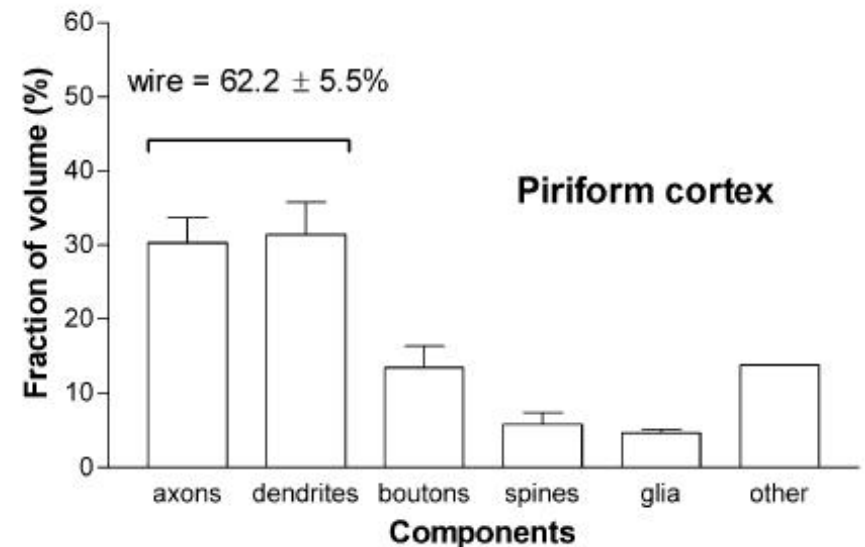
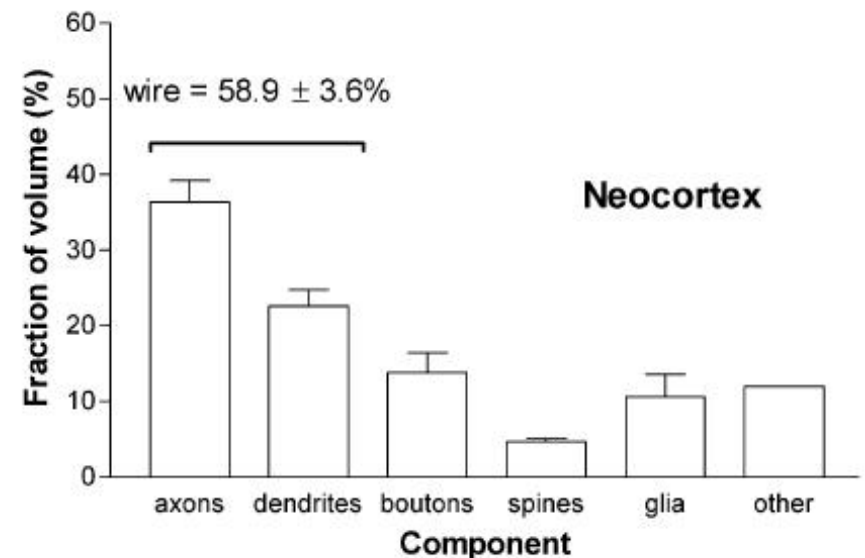
Neuron, Vol. 34, 341–347, April 25, 2002, Copyright ©2002

## Wiring Optimization in Cortical Circuits

Dmitri B. Chklovskii,<sup>1</sup> Thomas Schikorski,<sup>2</sup>  
and Charles F. Stevens<sup>2,3</sup>

<sup>1</sup>Cold Spring Harbor Laboratory  
Cold Spring Harbor, New York 11724

<sup>2</sup>Howard Hughes Medical Institute  
The Salk Institute



# Other fundamental results

- Optimal ratio of macromolecules to small molecules in bacteria (Vazquez, 2010)
- Morphogen gradients optimized against noise (Saunders, 2009)
- Limits to concentration sensing and chemotaxis (Endres, Levine, Wingreen, etc.)
- Leaf size bounded by diminishing returns to resource investment and flow impedance of tall trees (Jensen and Zwieniecki, PRL, 2013)

*And many, many more...*

Not all biophysics theory is done at  
50,000 feet

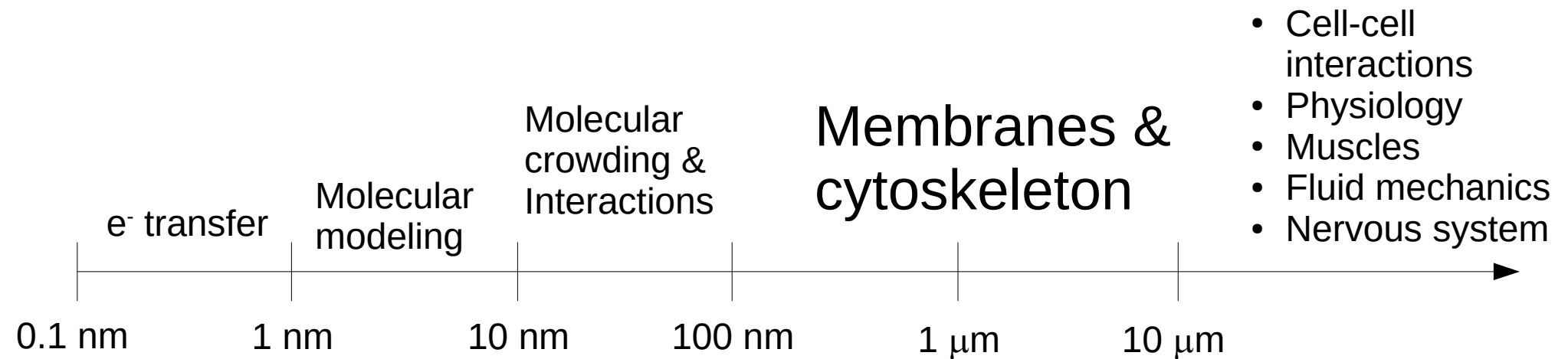
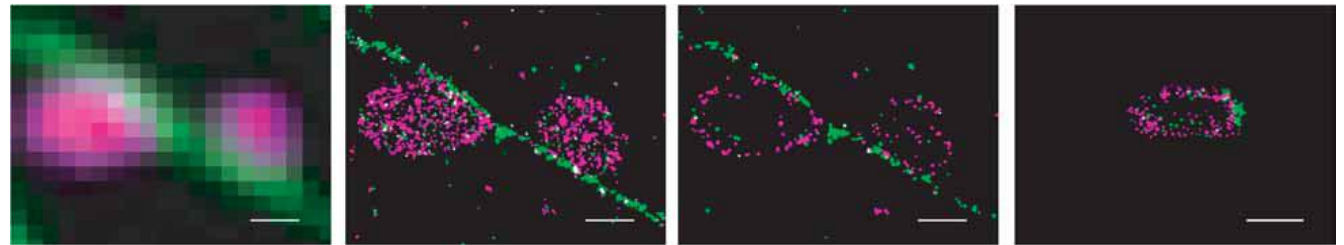


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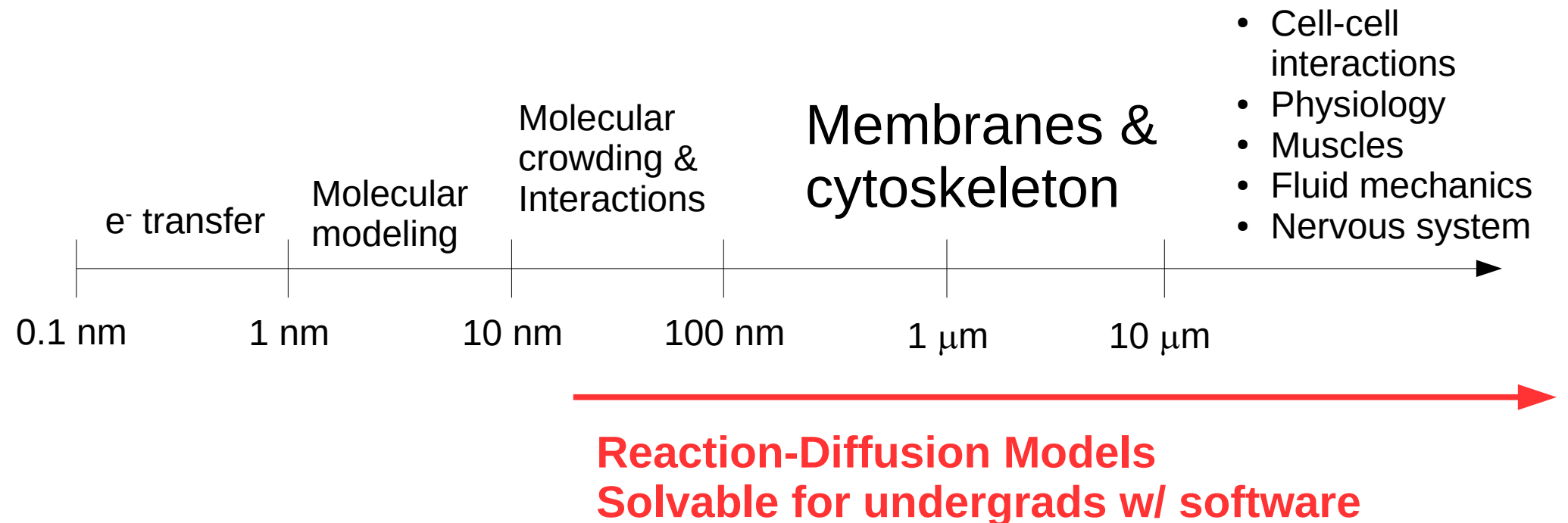
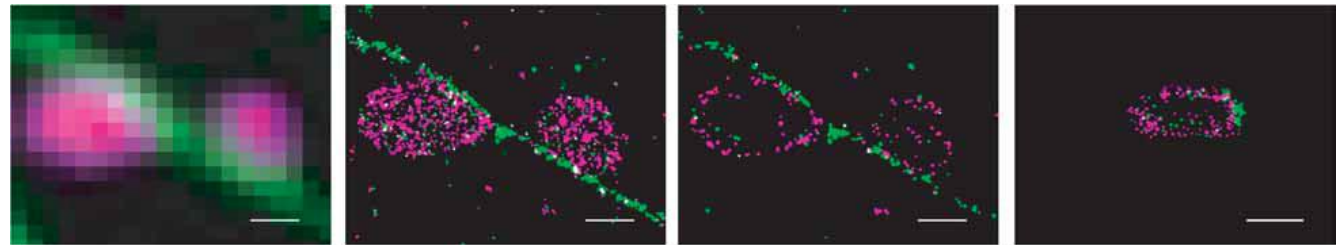
# Biophysics of particular systems

- Often borrow approaches from soft matter physics (especially cell membranes, cytoskeleton)



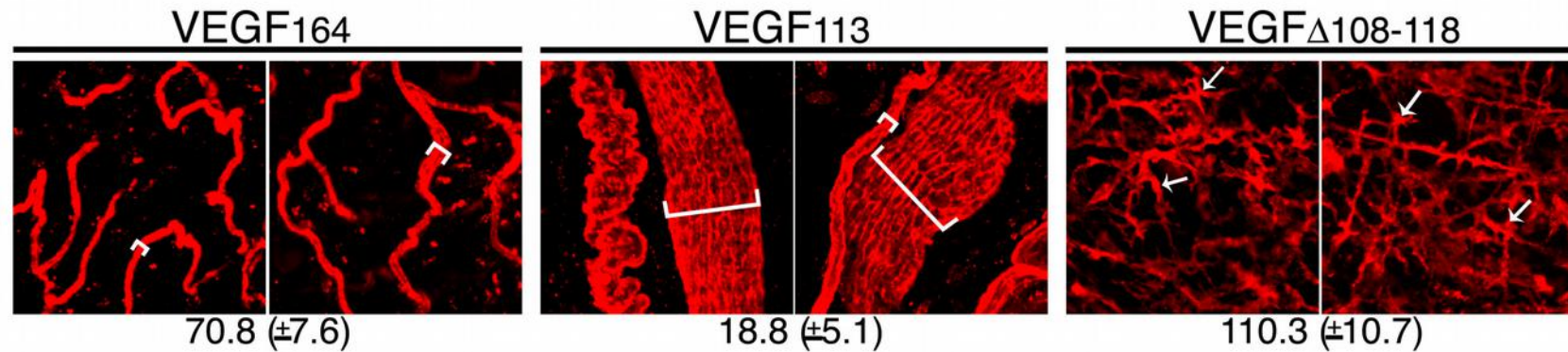
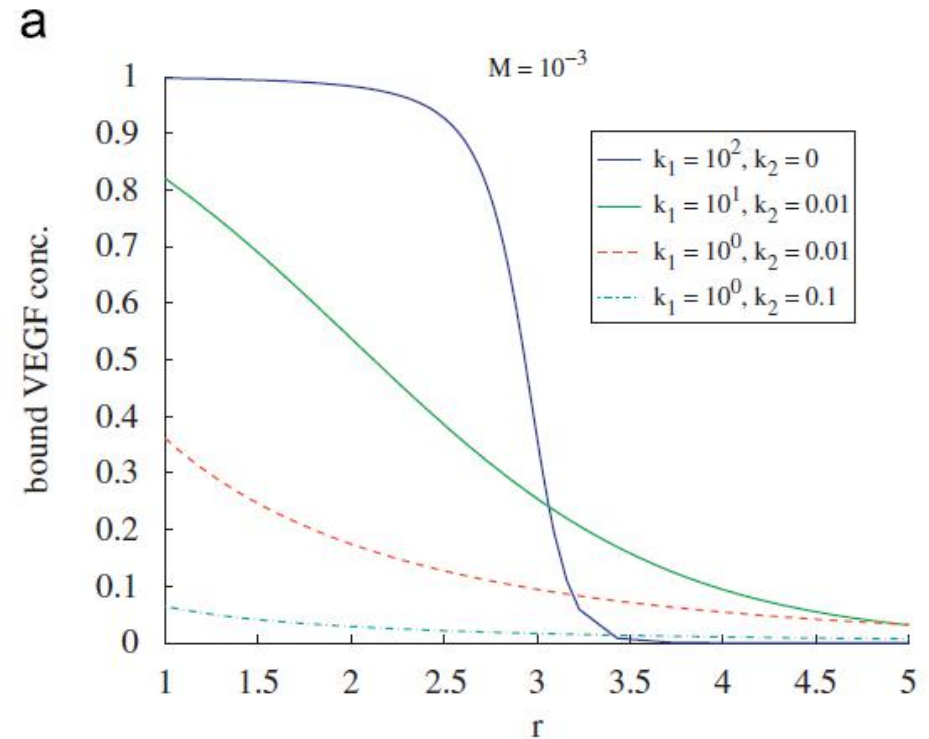
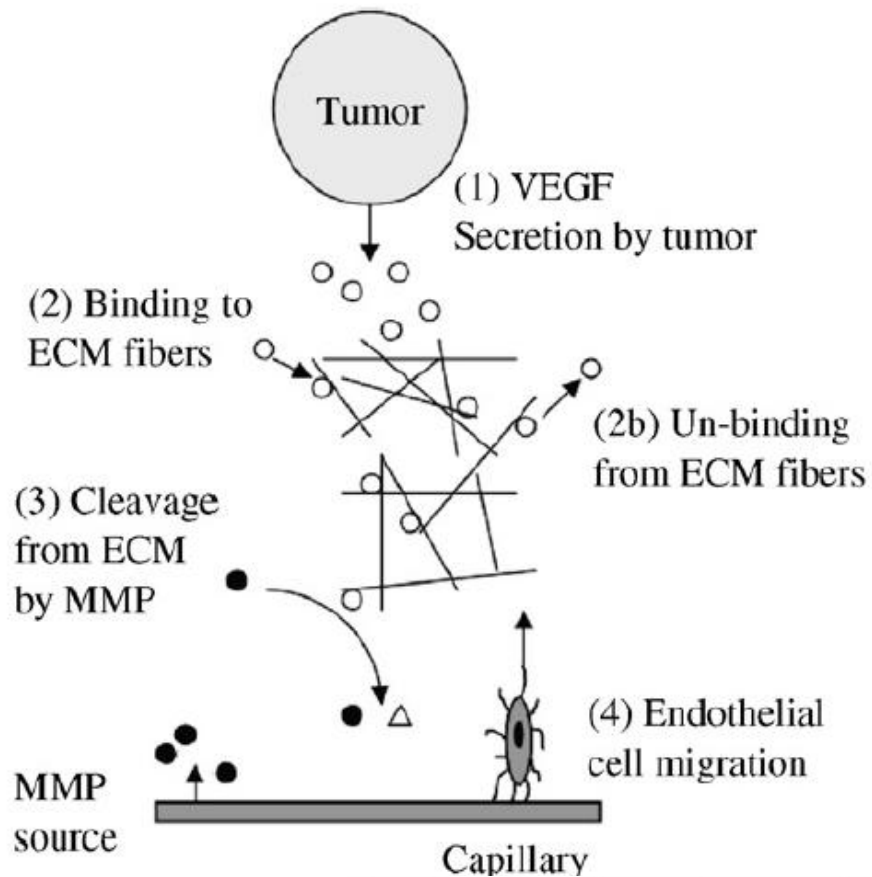
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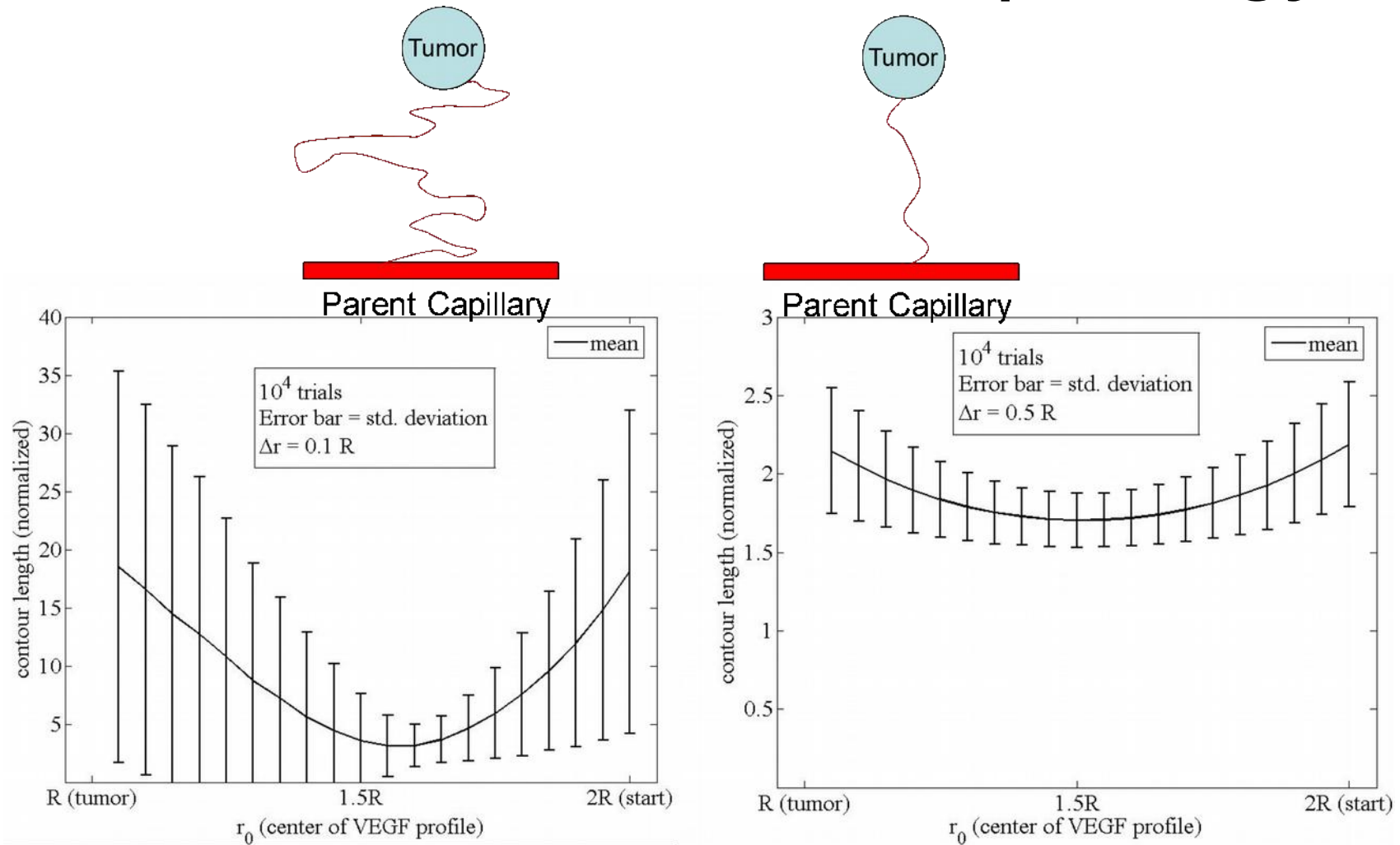


# Example: Tumor Angiogenesis





# Relation to vessel morphology



- Step-like gradients lead to longer migration times
- More tortuous morphology, fewer vessels reach tumor
- Consistent with differential potencies of VEGF<sub>189</sub> vs. VEGF<sub>165</sub>

# Who does what

- Fundamental limits: Primarily physicists
- Membranes, cytoskeletons, ordering, interactions: Primarily physicists
- Reaction-diffusion models: Physicists and mathematicians
- Cascades of reactions: Primarily mathematicians
- Developmental biology: Mostly mathematicians, and also Bill Bialek

# Molecular simulation

- Mostly done by chemists
- A few physicists→Note on history of the field
- Enough canned and standardized code that you can get undergrads involved on some level.
- Now for some slides from Paul Nerenberg (formerly of Claremont Colleges, about to start at CSU LA)

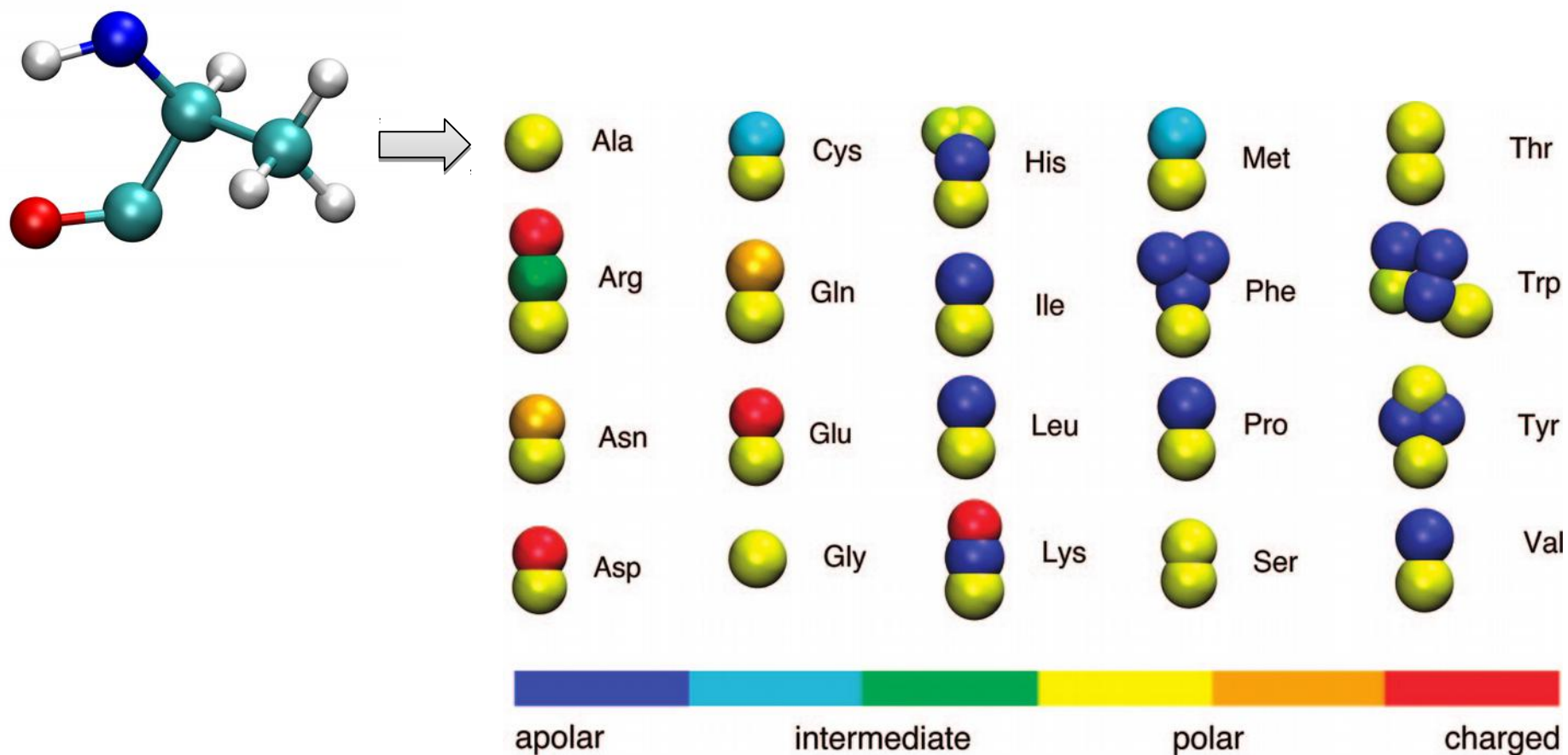
# What are molecular dynamics (MD) simulations?

- Basic idea: simulate molecules in time by calculating forces at each instant and applying Newton's second law
- Forces are derived from a potential energy function or *force field*
- A fundamental trade-off: accuracy of the potential energy function vs. sampling

# The various flavors of MD

Simulation type	Resolution	Number of atoms	Time scale
<i>ab initio</i> MD	Atomic+	$10^2$	200 ps
All-atom polarizable MD	Atomic	$10^2$ - $10^4$	50 ns
All-atom fixed-charge MD	Atomic	$10^3$ - $10^5$ (max: $10^7$ )	1 $\mu$ s (max: 1 ms)
Coarse-grained MD	$\sim$ Residue	$10^3$ - $10^5$	100 $\mu$ s

# The MARTINI force field



**Figure 1.** Coarse-grained representation of all amino acids. Different colors represent different particle types.

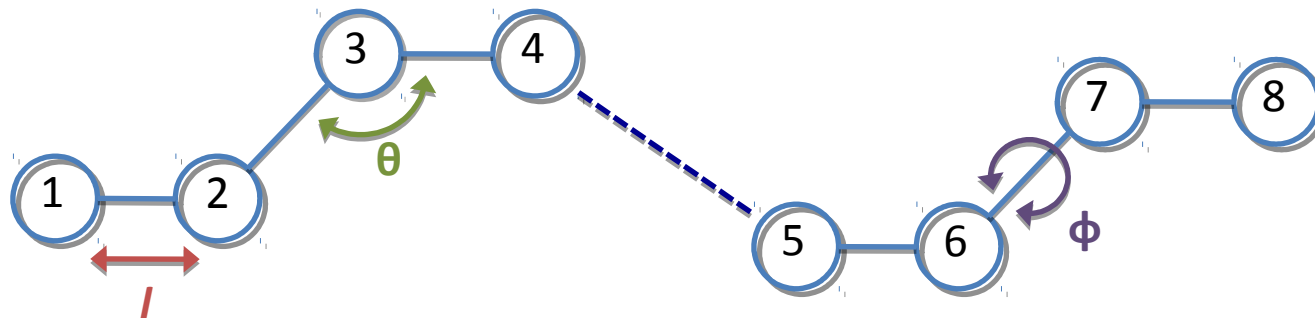
# CG MD in a nutshell

- Quasi-atoms are spheres of mass  $m_i$ , charge  $q_i$ , and van der Waals size  $\sigma_i$
- Potential energy function (force field):

$$U = U_{\text{bonded}} + U_{\text{non-bonded}}$$

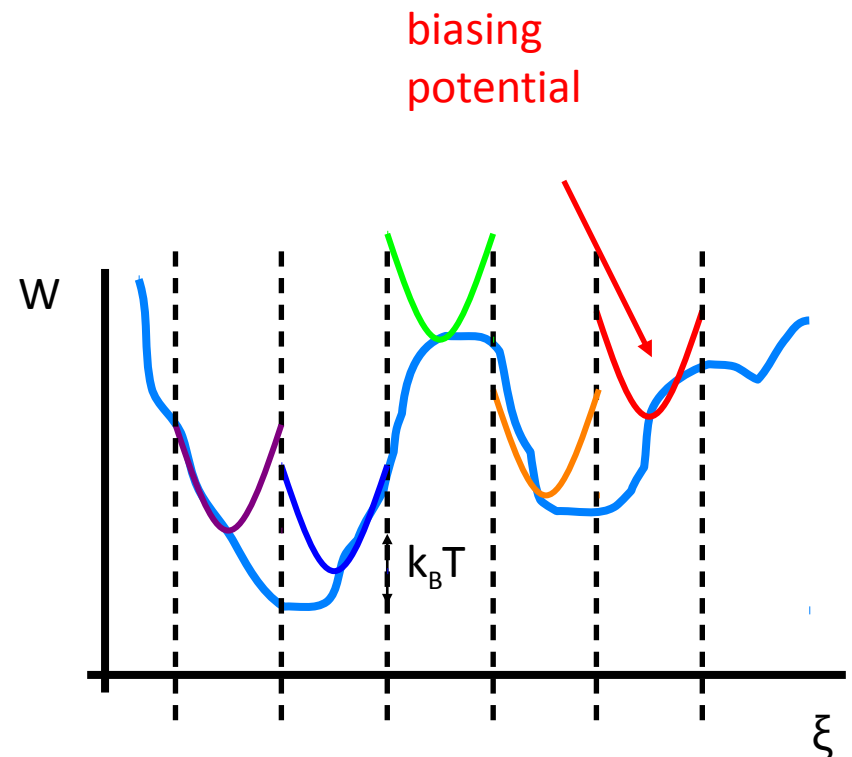
$$= \sum_{\text{bonds}} \frac{1}{2} k_b (l_{ij} - l_{ij}^{eq})^2 + \sum_{\text{angles}} \frac{1}{2} k_a (\cos \theta_{ijk} - \cos \theta_{ijk}^{eq})^2 + \sum_{\text{dihedrals}} k_d (1 + \cos(n\phi_{ijkl} - \phi_{ijkl}^{eq}))$$

$$+ \sum_{\text{nonbonded}} \left[ \frac{k_e q_i q_j}{\epsilon r_{ij}} + 4\epsilon_{ij} \left( \left( \frac{\sigma_{ij}}{r_{ij}} \right)^{12} - \left( \frac{\sigma_{ij}}{r_{ij}} \right)^6 \right) \right]$$



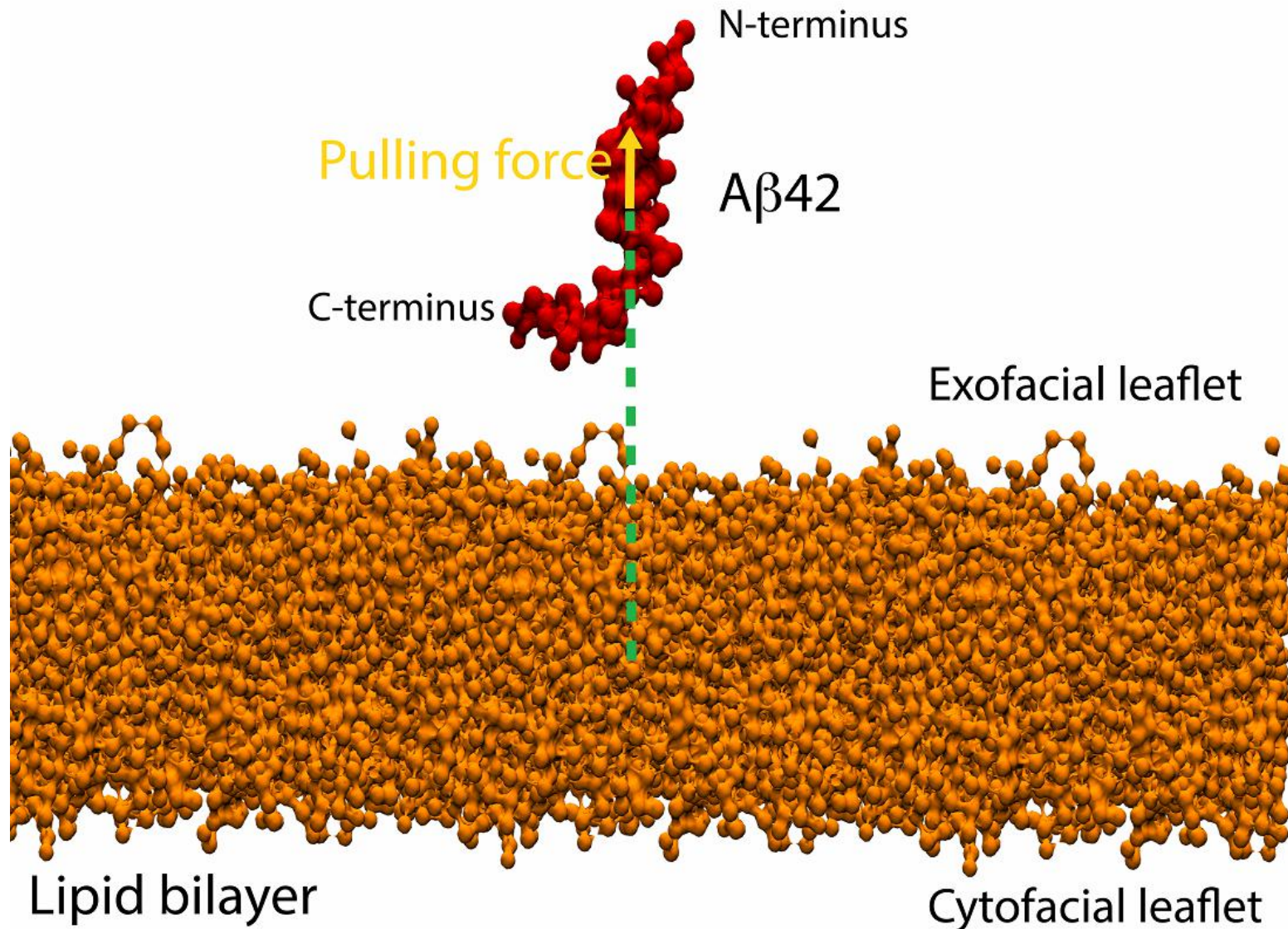
# Umbrella sampling

- Enhanced conformational sampling method along reaction coordinate: can traverse large energy barriers between states
- Create a biasing potential to sample small regions of the reaction coordinate
- Sample many windows to cover full range of reaction coordinate
- Unbias data at the end to recover underlying energy landscape (WHAM)





# Typical graphics



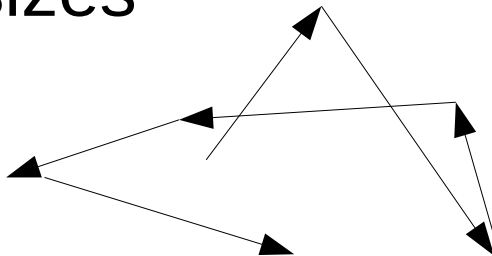
# Phenomenology and Information Limits

- Biological systems are messy
- Predicting the sorts of signals that we'll see from an underlying model/mechanism is hard
- Examples: Protein folding, random walks, imaging

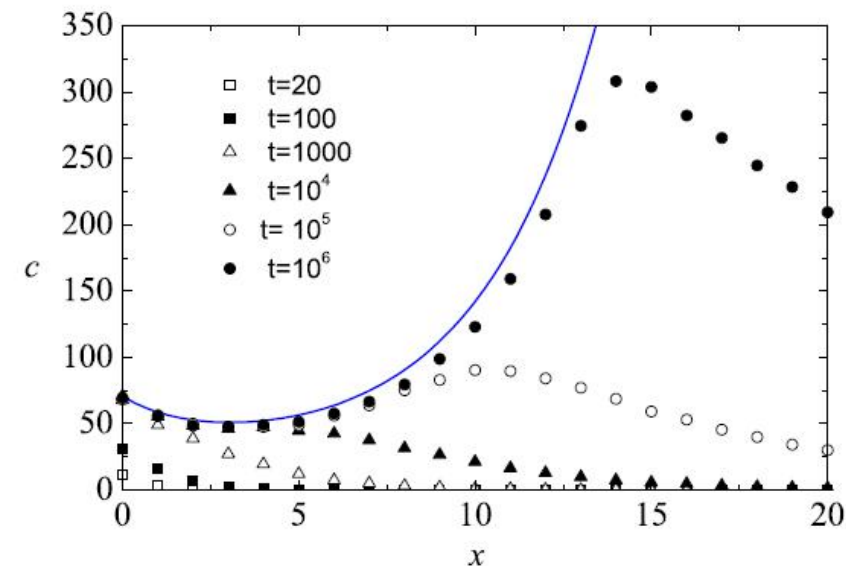
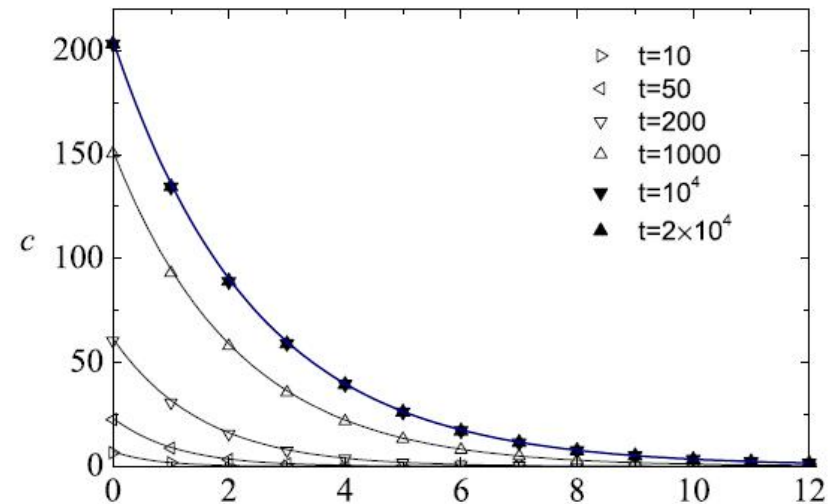
# Random Walks

- Simple diffusion is simple
- Anomalous diffusion: Often modeled with Continuous Time Random Walk

→ Interesting for power-law distribution of wait times and/or step sizes



→ Even more interesting for spatially varying reaction rates



# Single molecule experiments

PRL **110**, 158105 (2013)

PHYSICAL REVIEW LETTERS

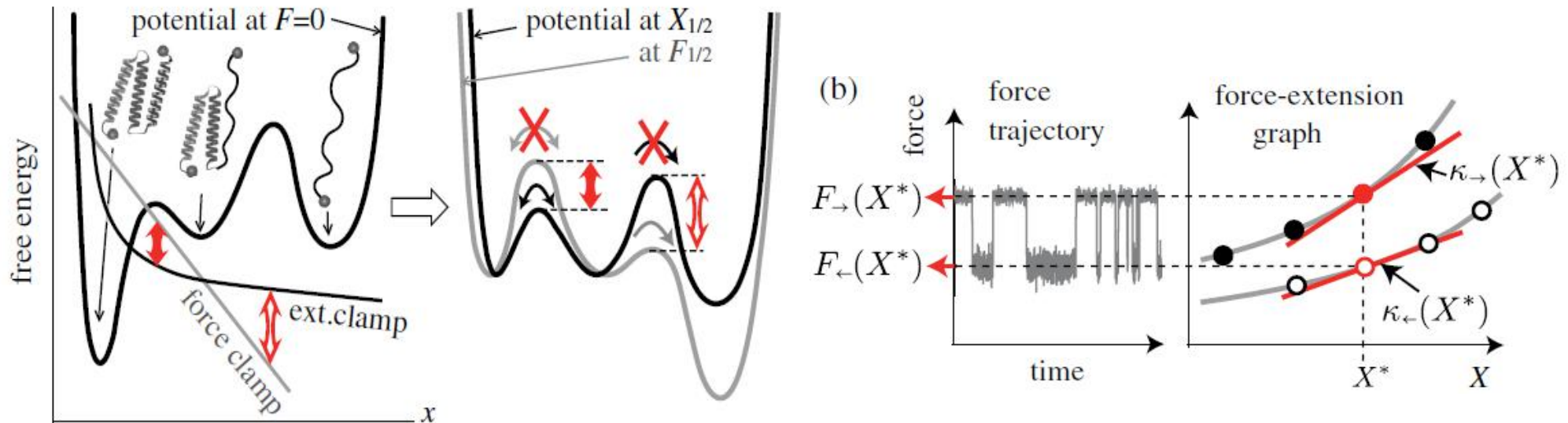
week ending  
12 APRIL 2013

## Single Molecules in an Extension Clamp: Extracting Rates and Activation Barriers

Yohichi Suzuki and Olga K. Dudko

*Department of Physics and Center for Theoretical Biological Physics, University of California at San Diego,  
La Jolla, California 92093, USA*

(Received 4 December 2012; published 12 April 2013)

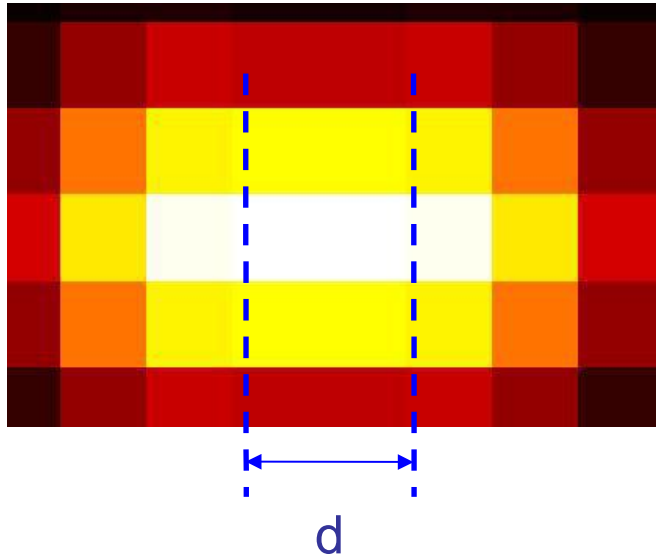


# Theory of superresolution

- Every new technique requires a theory to predict the maximum information obtainable
- That's what I've been doing

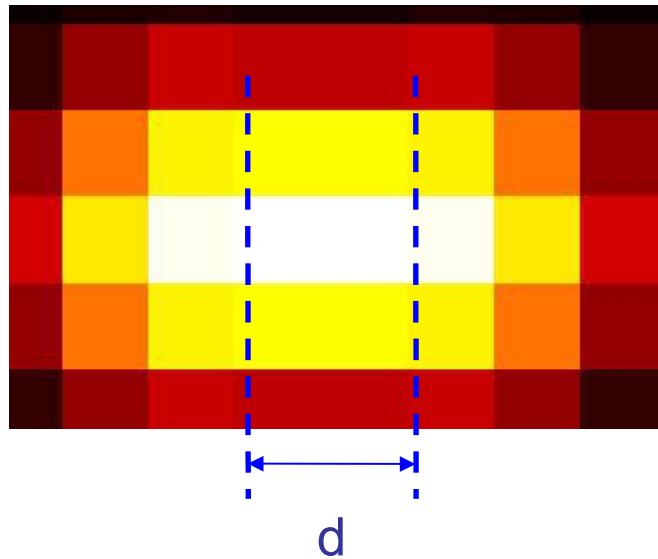
# Single-Molecule Localization

- 2 fluorescent molecules close together would look like this under microscope

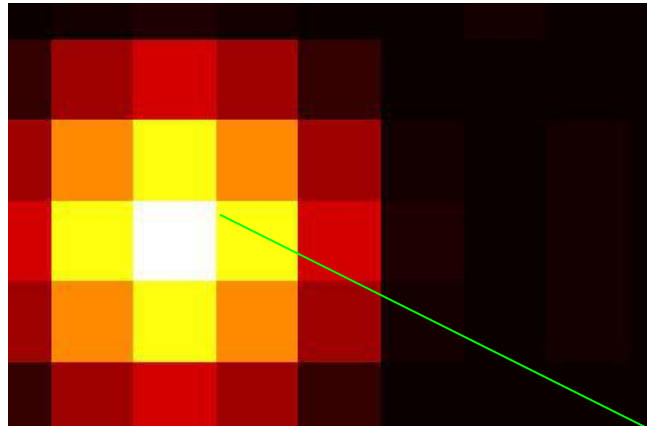


# Single-Molecule Localization

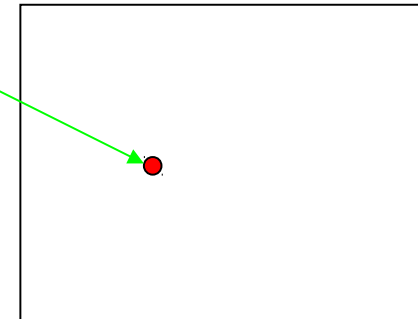
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What if only one at a time is shining?

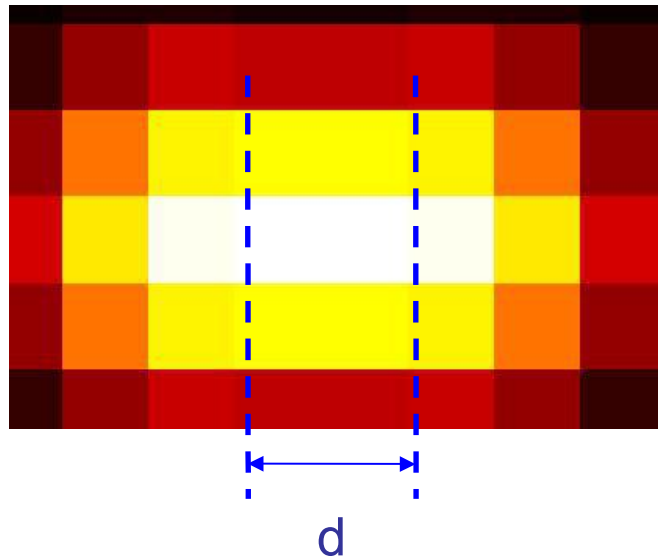


Find centers and infer molecule locations

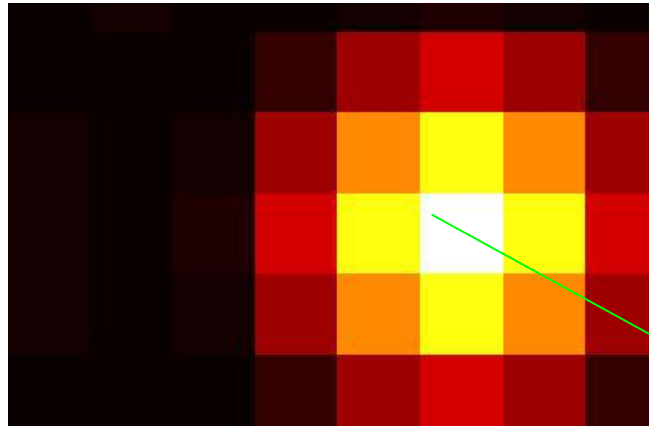


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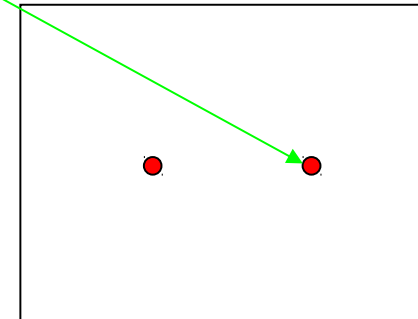
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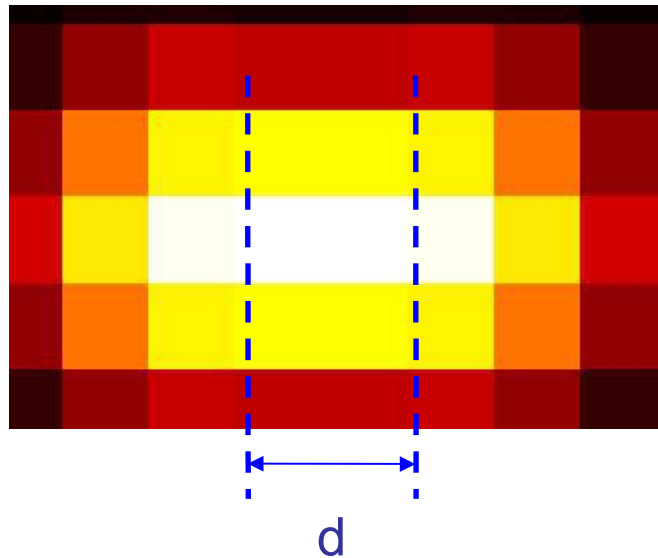
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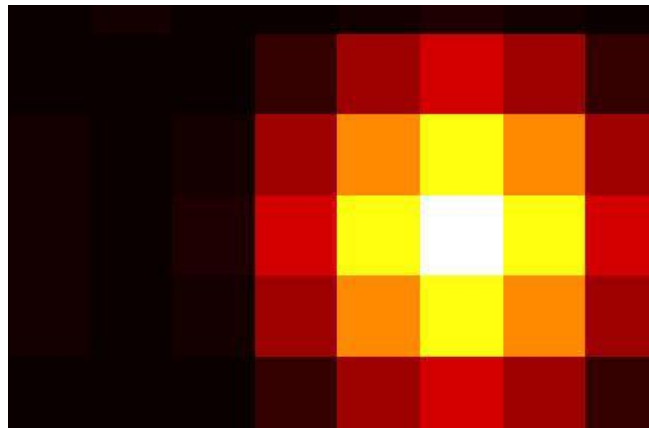


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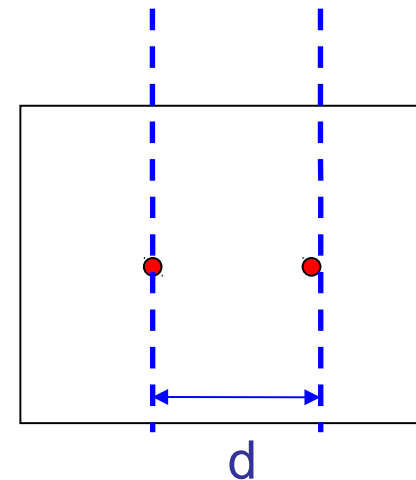
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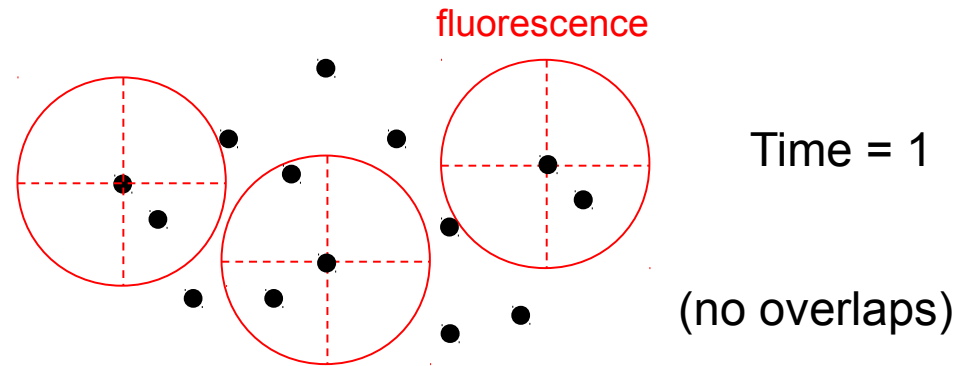


**Do this enough times, and eventually you know where every molecule is!**

# Sequential Localization

Not all the molecules in a crowded image are “on” at once.

Localize those that are on

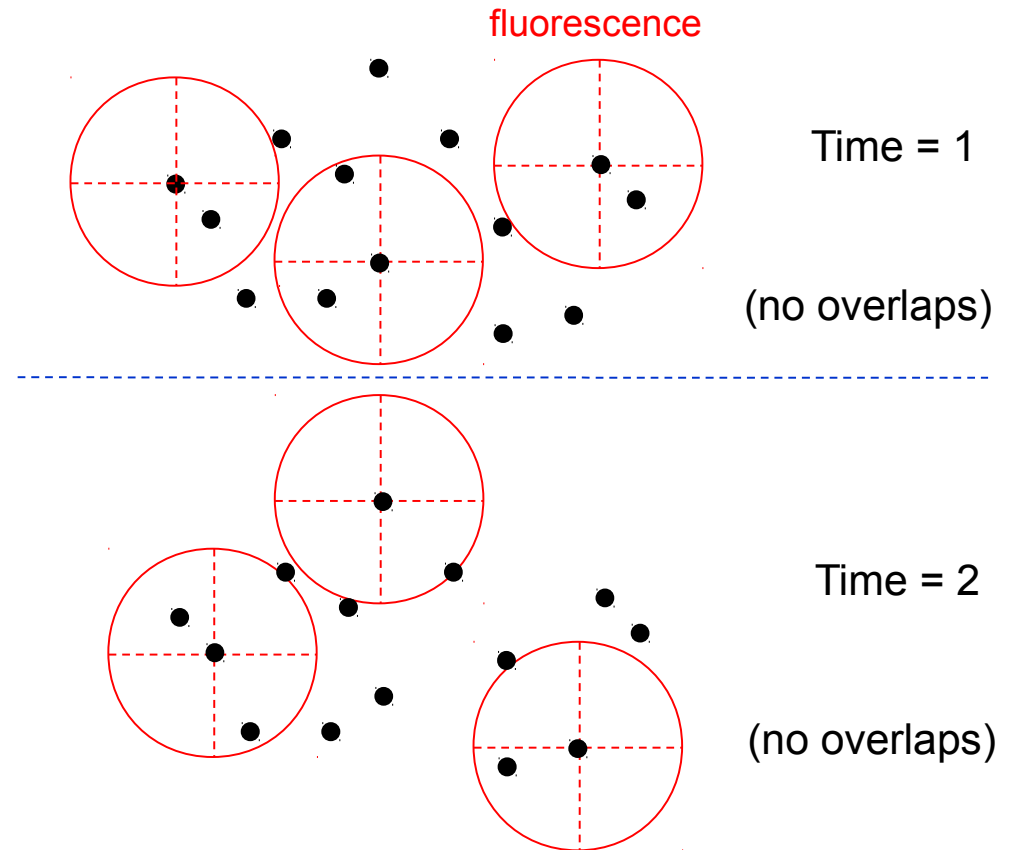


# Sequential Localization

Not all the molecules in a crowded image are “on” at once.

Localize those that are on

Repeat for different set of on molecules



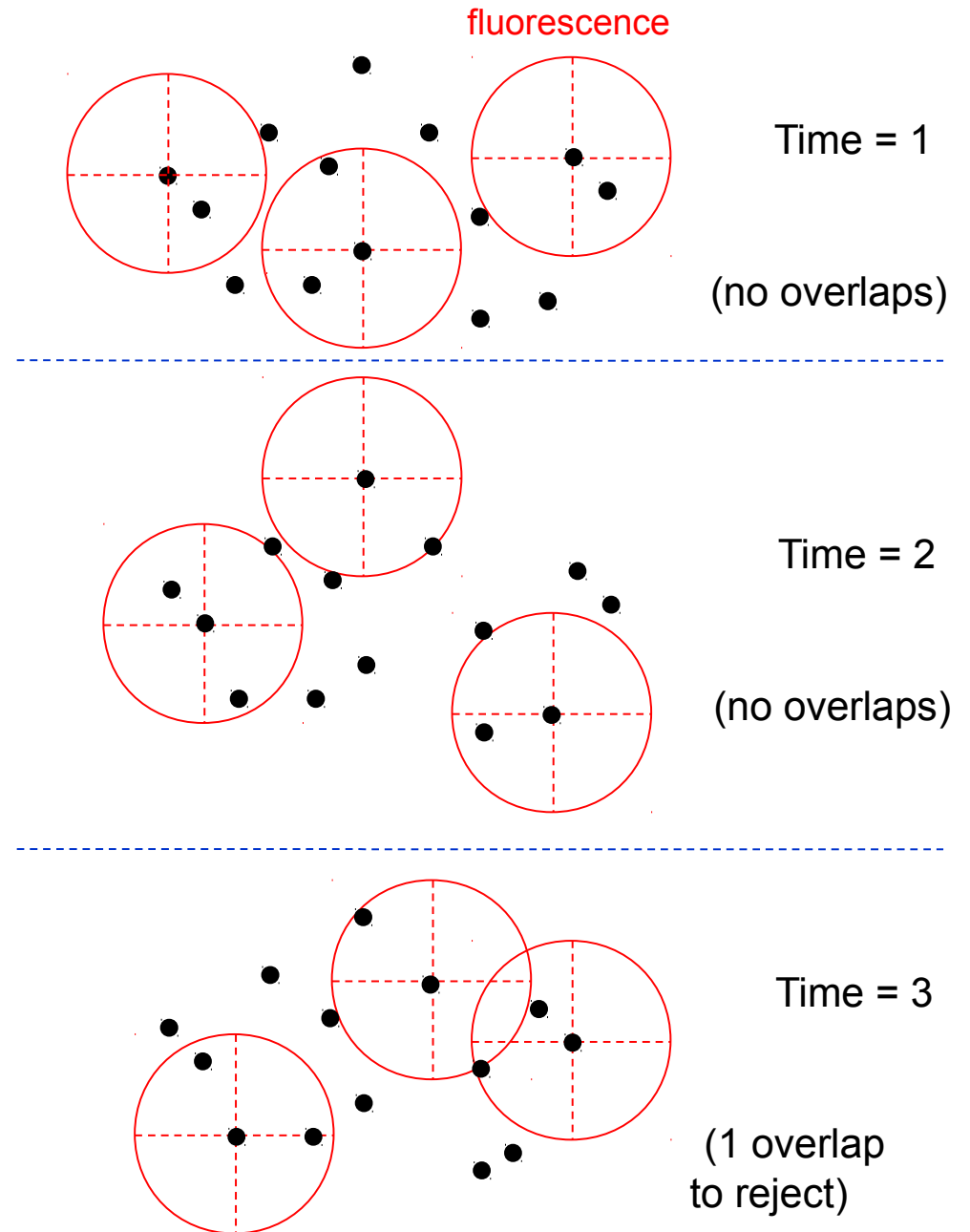
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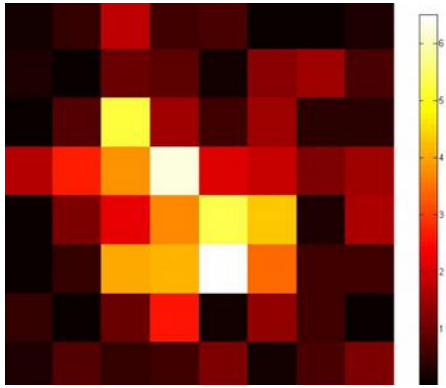
Repeat for different set of on molecules

2 close molecules are “on” at the same time: Need to discard that image.



# Noise and Single-Molecule Resolution

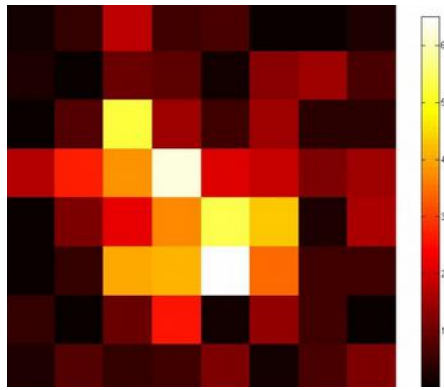
- No image is perfect
- We get photons one at a time, in an unpredictable sequence.



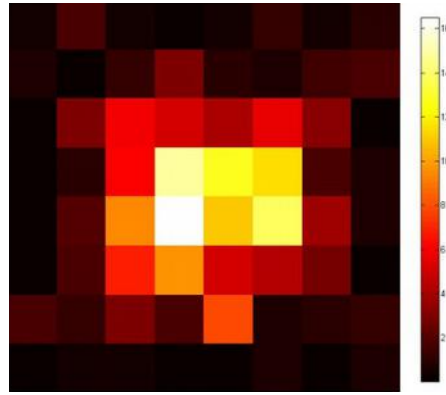
50 photons

# Noise and Single Molecule Resolution

- No image is perfect
- We get photons one at a time, in an unpredictable sequence.



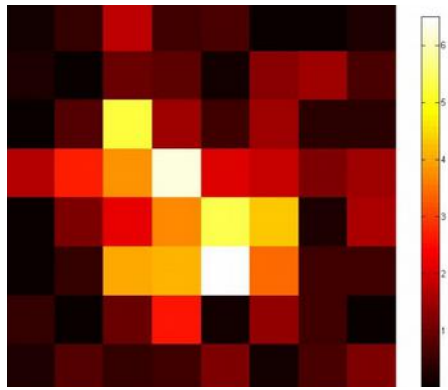
50 photons



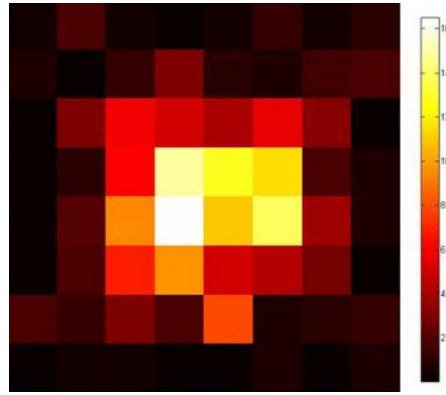
200 photons

# Noise and Single Molecule Resolution

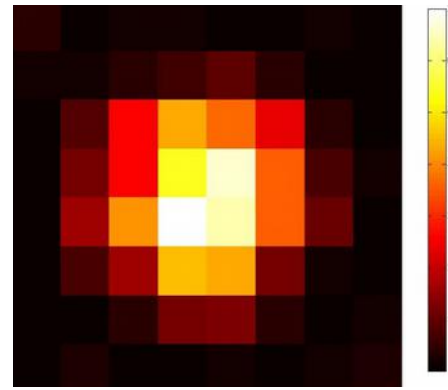
- No image is perfect
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50 photons



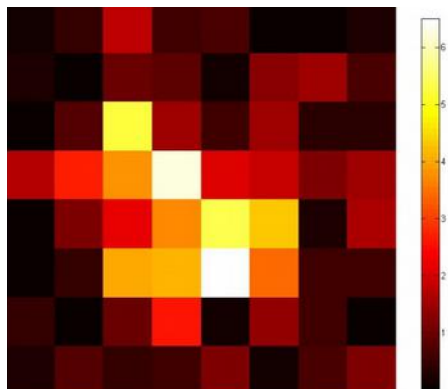
200 photons



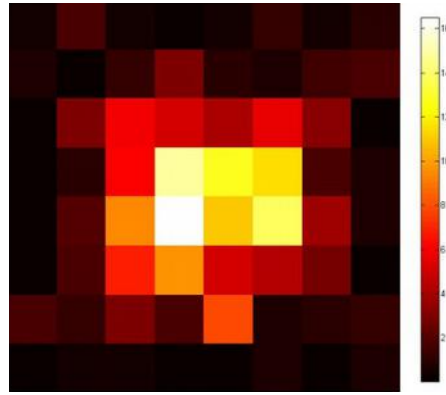
800 photons

# Noise and Single Molecule Resolution

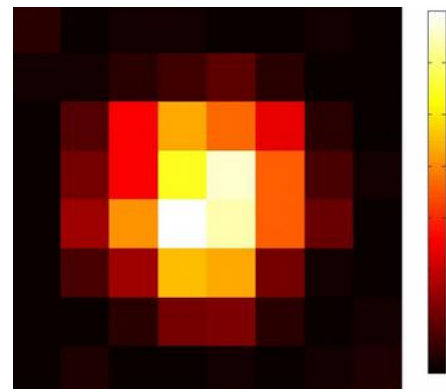
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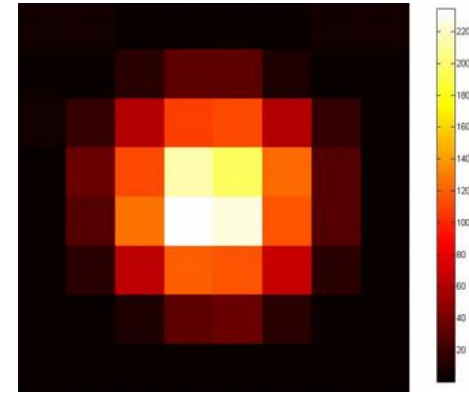
50 photons



200 photons



800 photons



2500 photons

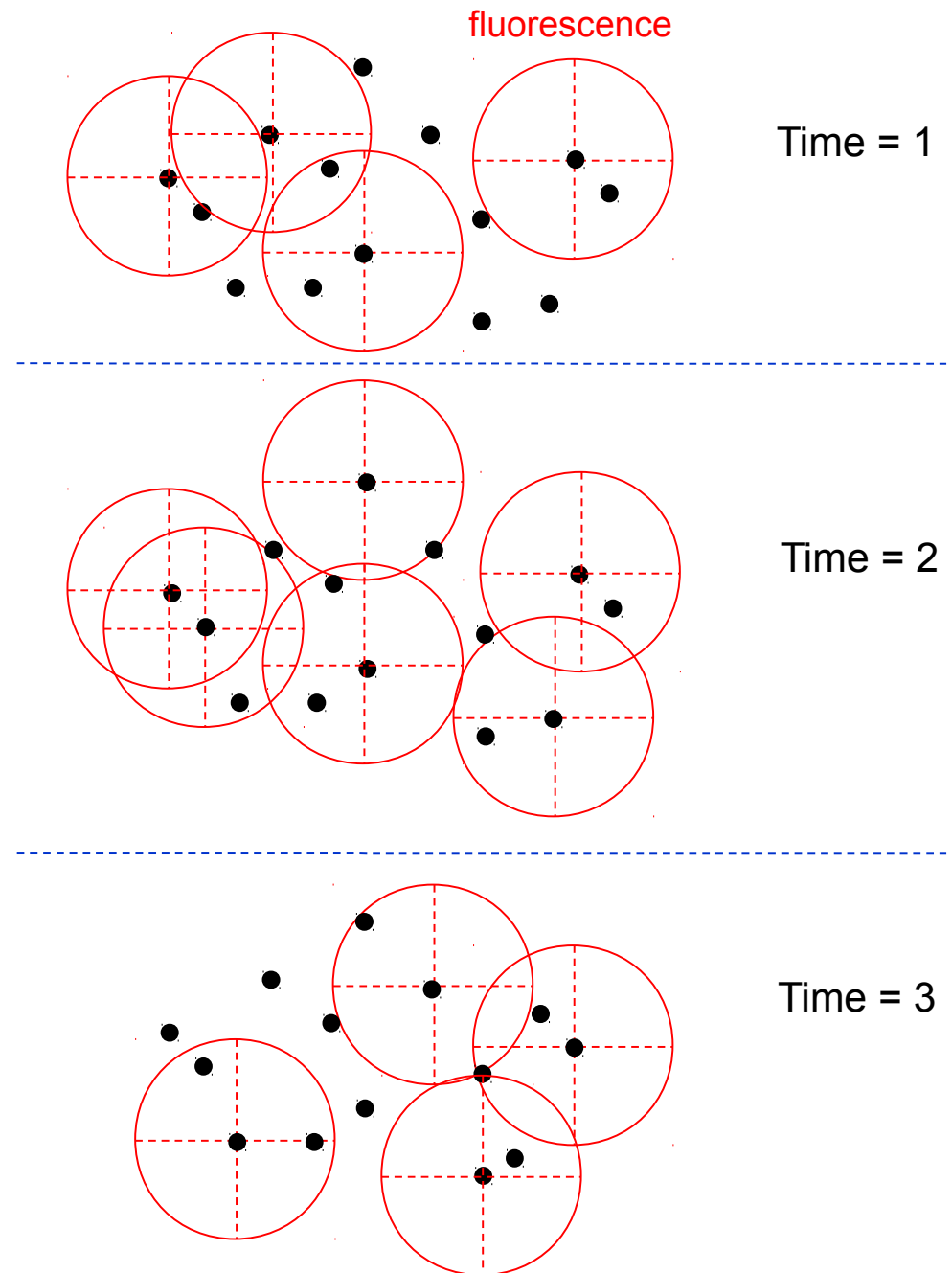
Best possible precision in  
estimate of molecular position:

$$\approx \frac{\lambda}{\sqrt{\# \text{ of photons collected}}}$$

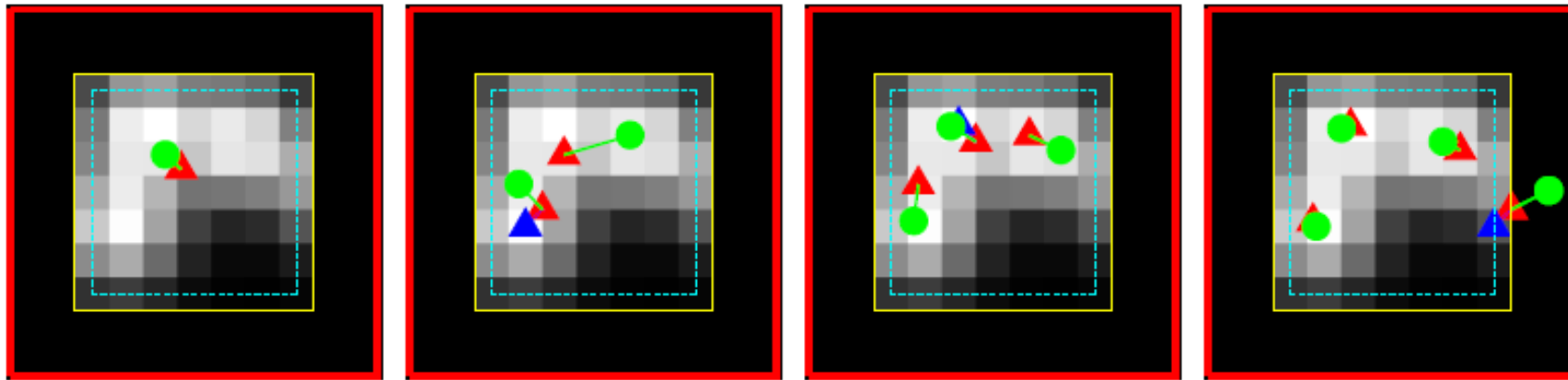


# Simultaneous Localization

Use software to estimate many positions at once, from overlapping images.



# Multifluorophore Localization



Huang, Lidke, et.  
al. 2011

(b) Single emitter fitting    (c) Two emitter fitting    (d) Three emitter fitting    (e) Four emitter fitting

- Select model that best matches data

**Caution:** You can always add more fluorophores with low intensity to “mop up” residuals

# Motive: Speed

Max Fluorophores per Frame	Normalized Min. # frames needed	Speed improvement
1	1	1
2	0.44	2.27
3	0.26	3.85
4	0.19	5.26
5	0.14	7.14
10	0.023	43.5

# What about resolution?

- MLE software can localize multi-fluorophore images w/ theoretical limit to precision.
- Most approaches consider images of N~5 fluorophores at once.
- Densities up to 8-10 fluorophores/ $\mu\text{m}^2$ .

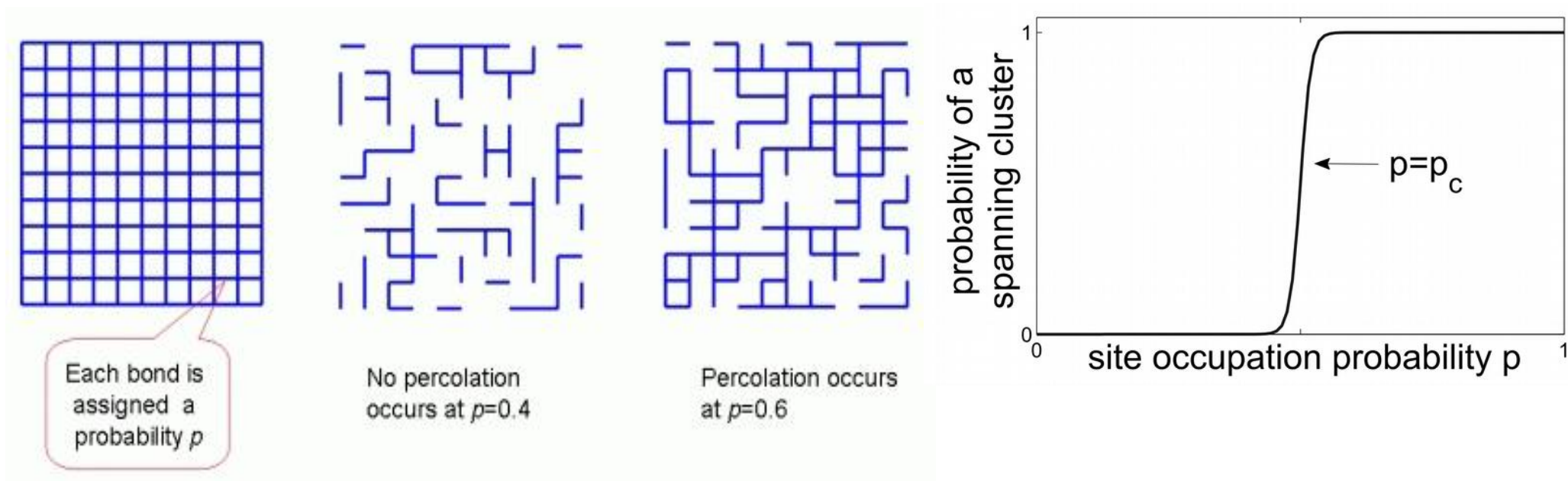
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*Is this the limit?*

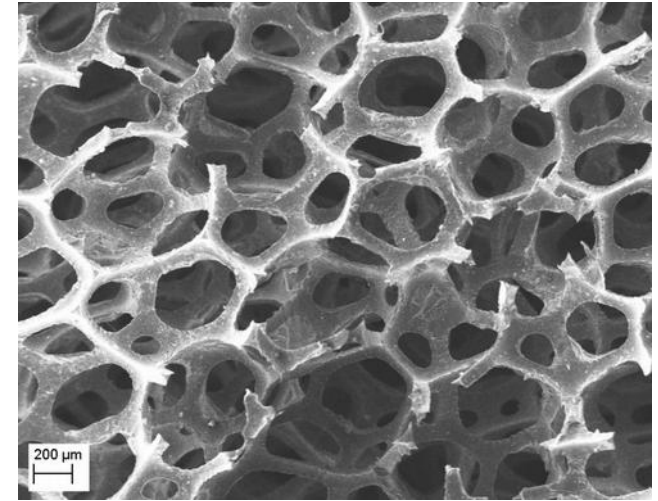
# Percolation Theory

- In a system with randomly-placed components, how many do you need to be assured a path from side to side?



# Why is there interest in percolation?

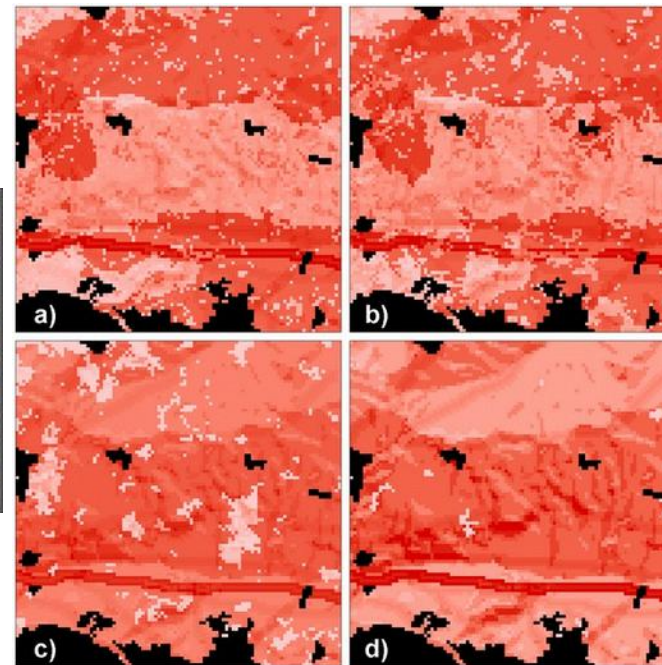
- Porous flow
- Electrical conduction
- Disease
- Forest fires



Bibikov and Prokof'ev, *Composite Materials for Some Radiophysics Applications*



Aerial photo of rural Spain

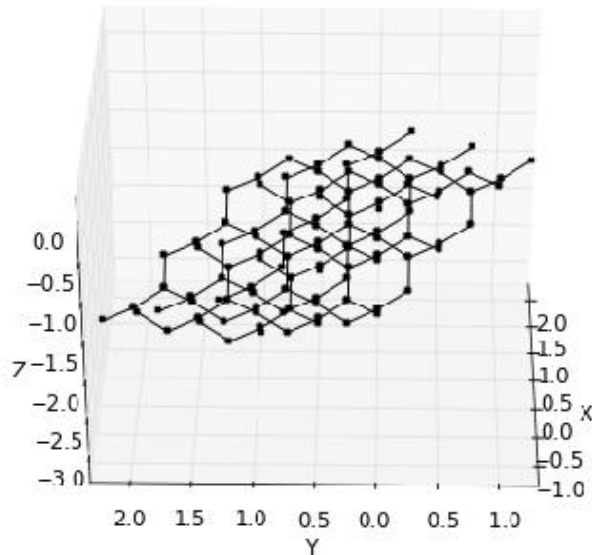


Computer simulations of wildfire risk in different land use scenarios.

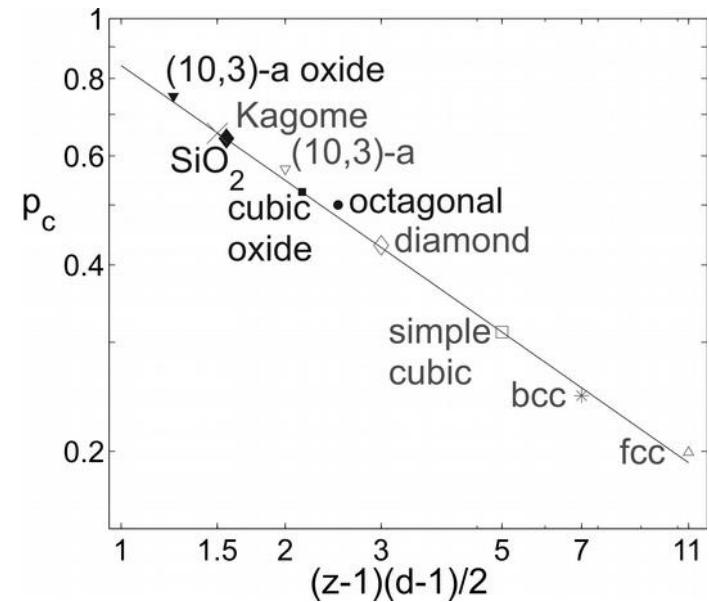
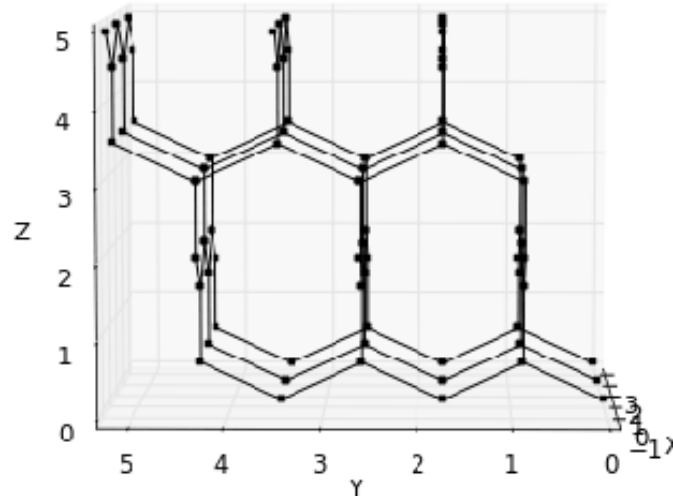


# Why did I get into percolation?

- (10,3)-a



- (10,3)-b



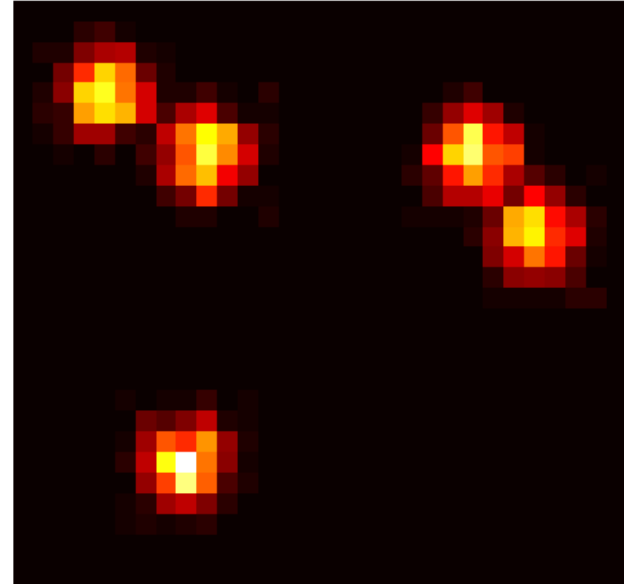
Low-coordinated lattices: Side project for my undergrads

Tran, JStat, 2013  
Yoo, Jstat, 2014

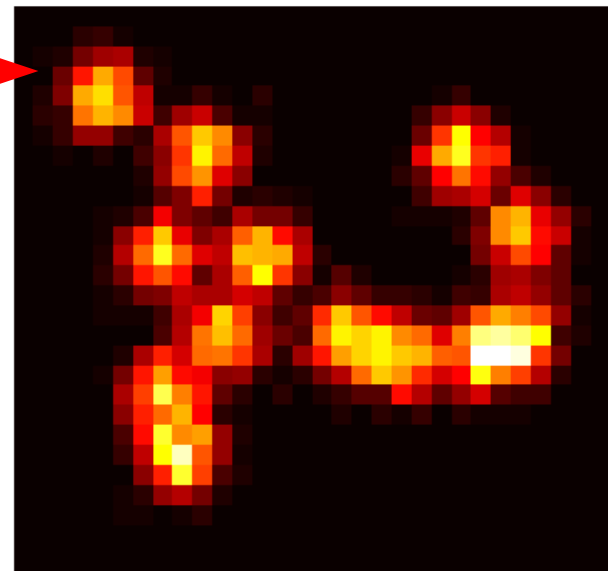


# Why should microscopists care about percolation?

- Low density
- Small clusters are independent

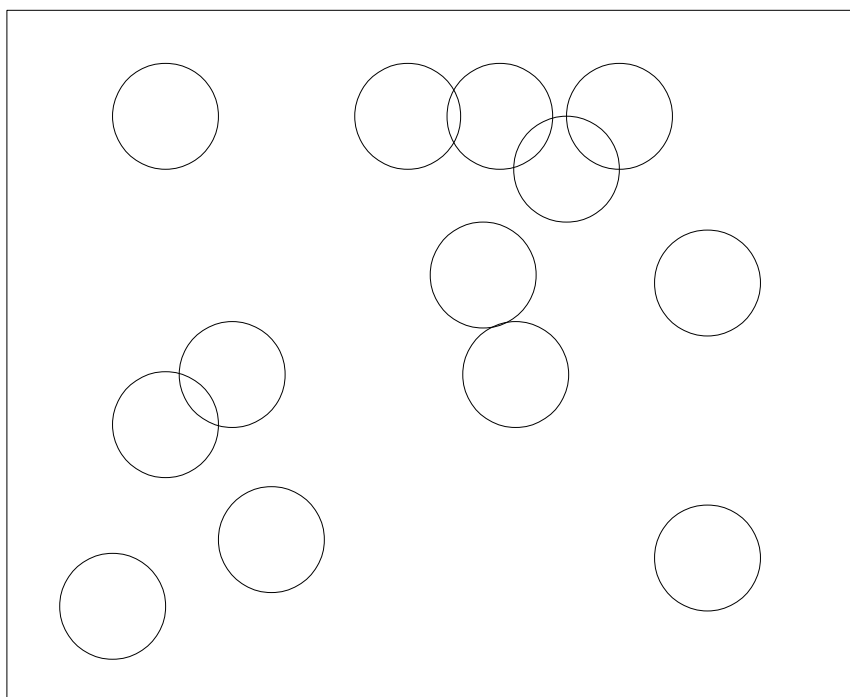


- **High density (percolation)** →
- Many molecules are coupled
- Position estimates have higher variance.

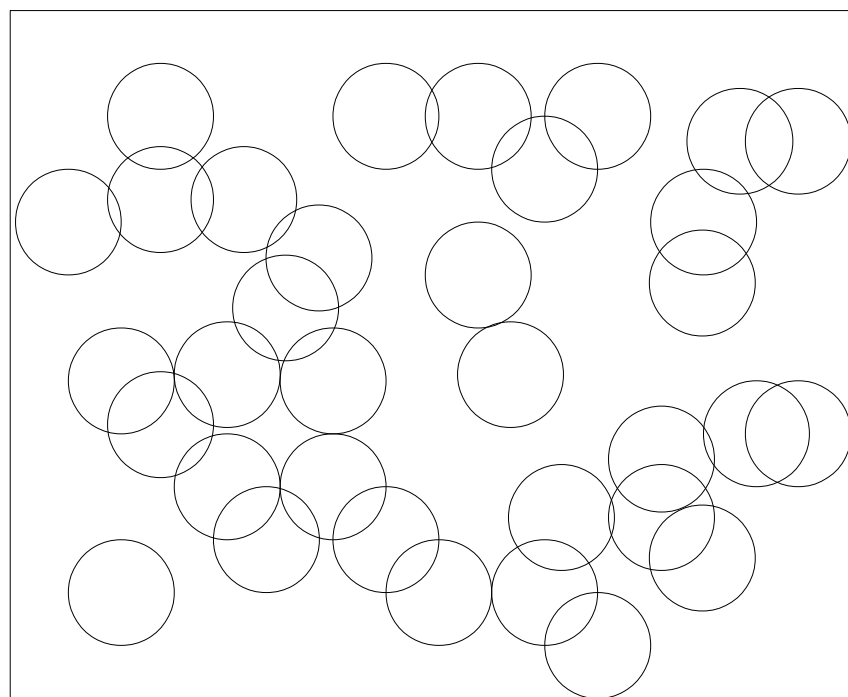


# Continuum Percolation

- We don't need to do this on graph paper.



Low coverage

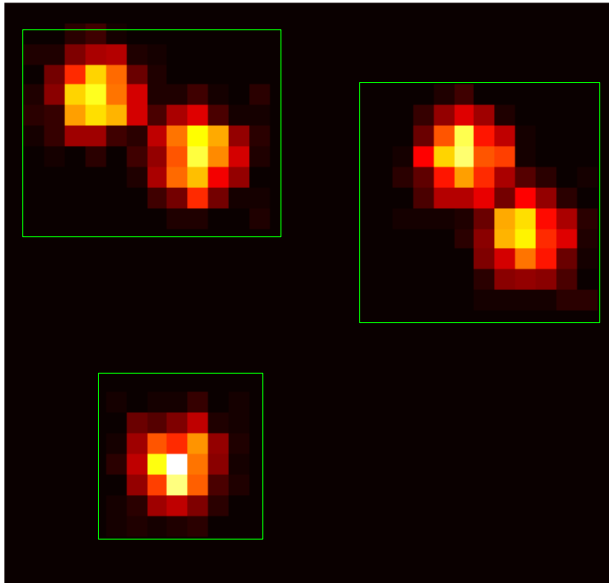


High coverage ( $\geq 67.6\%$ )

# Boundary Issues

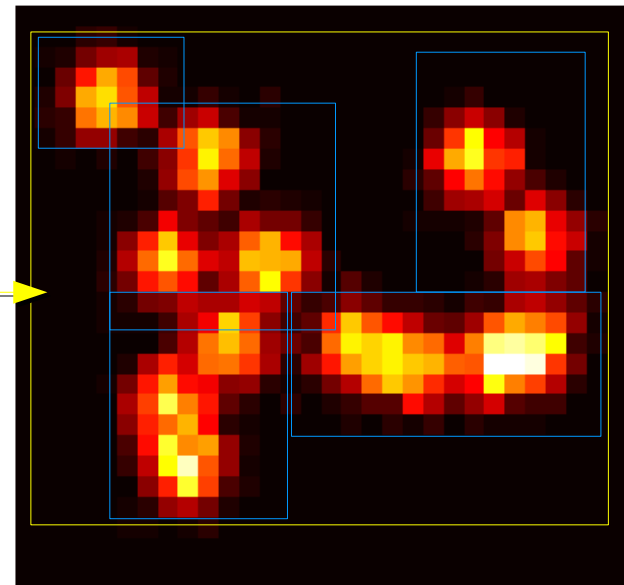
Percolation→Spanning Cluster→Images that can't be contained in small windows

Tractable estimation problems.



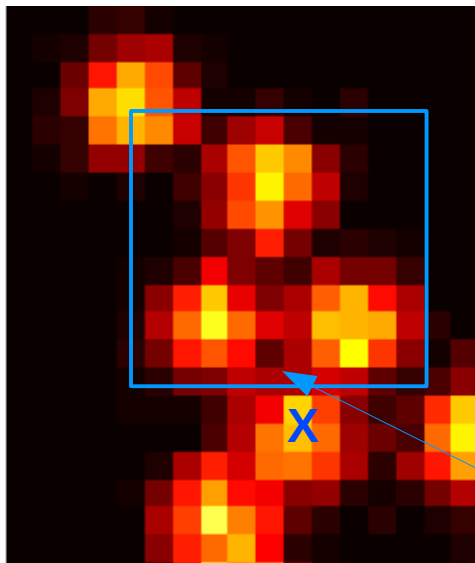
Intractable

Regions with “spill-over”



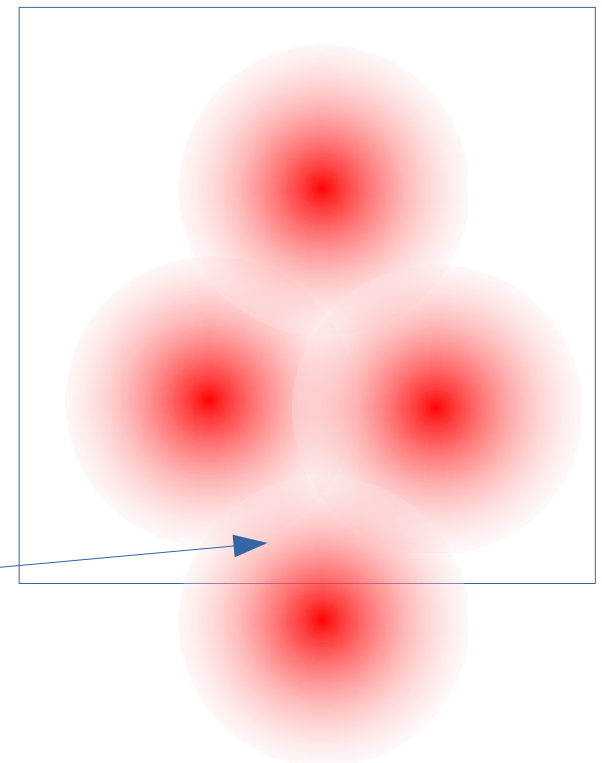
# Boundary Issues (2)

- Fluorophore outside the window cannot be estimated well
- But it still biases other estimates



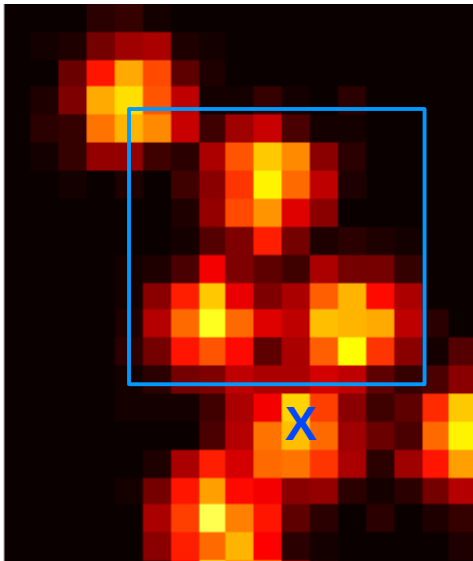
Spill-over  
photons from  
molecule X  
outside ROI  
(blue box)

Schematic



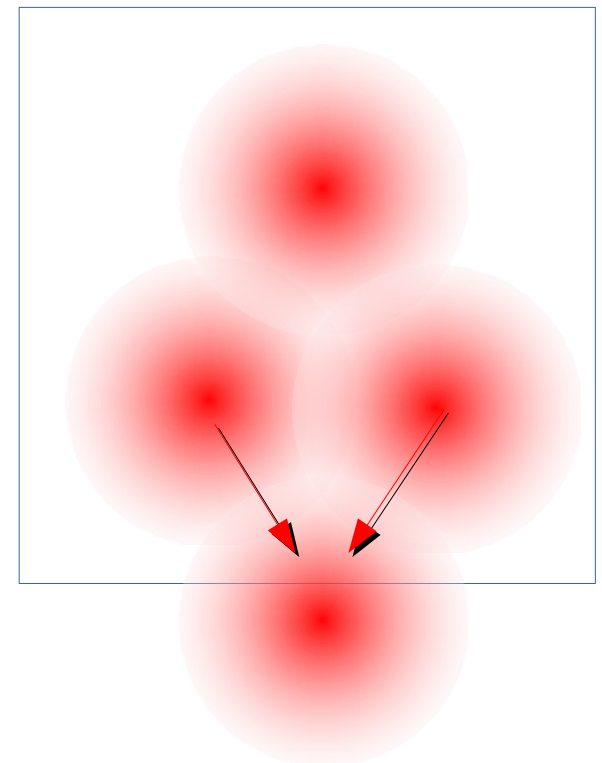
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Bias in position estimates due to spill-over photons.

Schematic



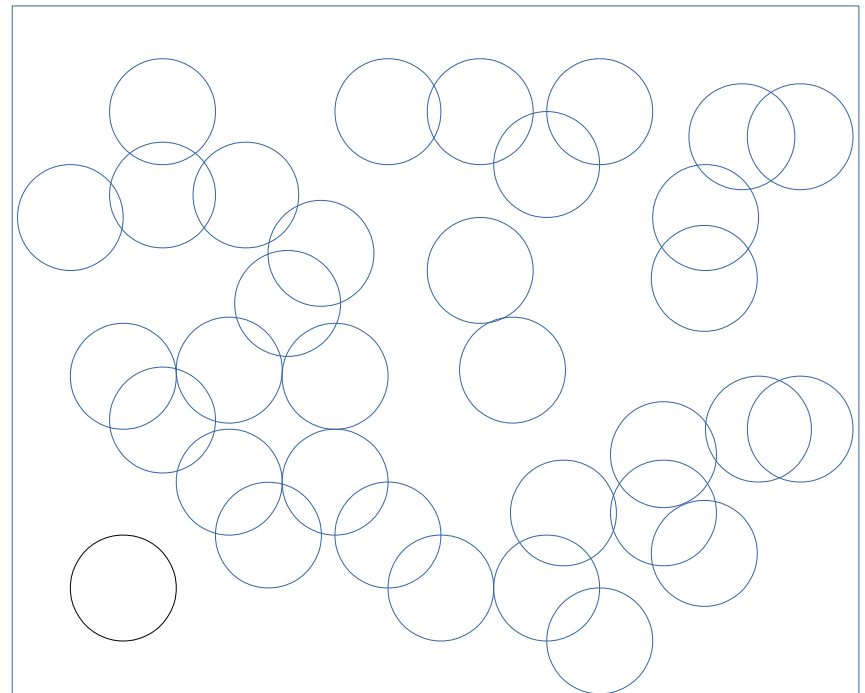
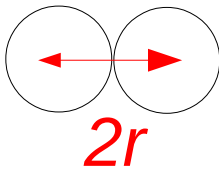
# When is overlap inevitable?

- Easy answer: Above continuum percolation threshold (67.6% coverage).

- $\sigma$ =density of circles

$$=1.13/\pi r^2$$

$$=4.52 \text{ neighbors}/\pi(2r)^2$$



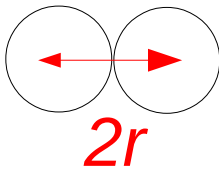
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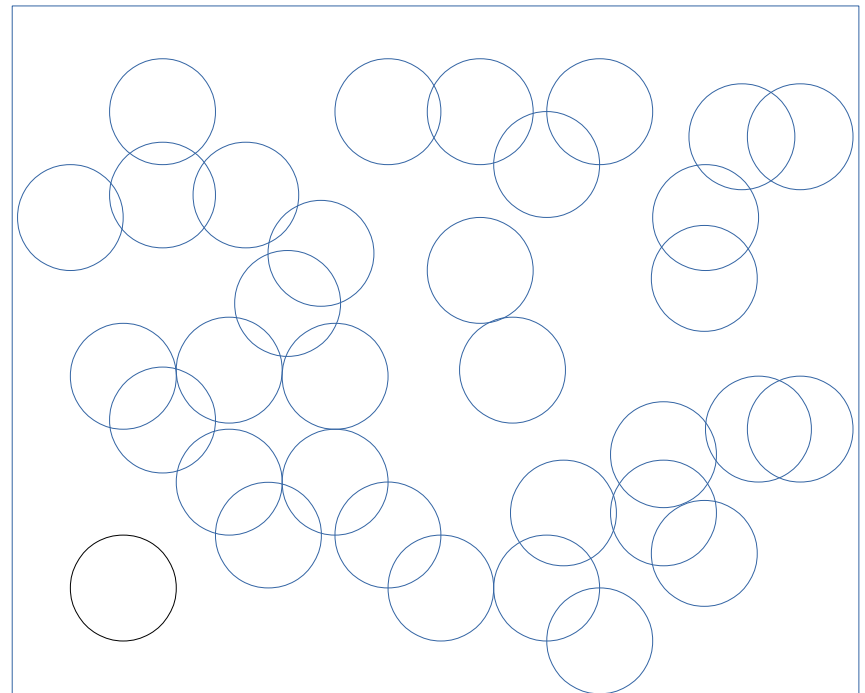
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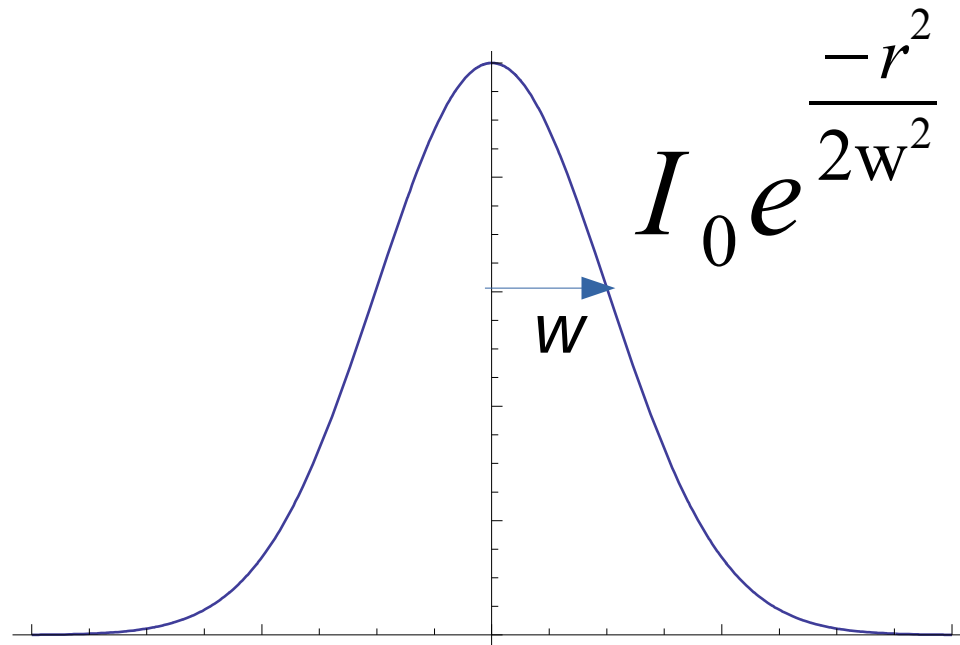


**What is  $r$ ?**



# Radius for overlaps

- $w$ ?  $2w$ ?



- Useful answer: “Close enough to matter”

Bias = std. dev. position estimate

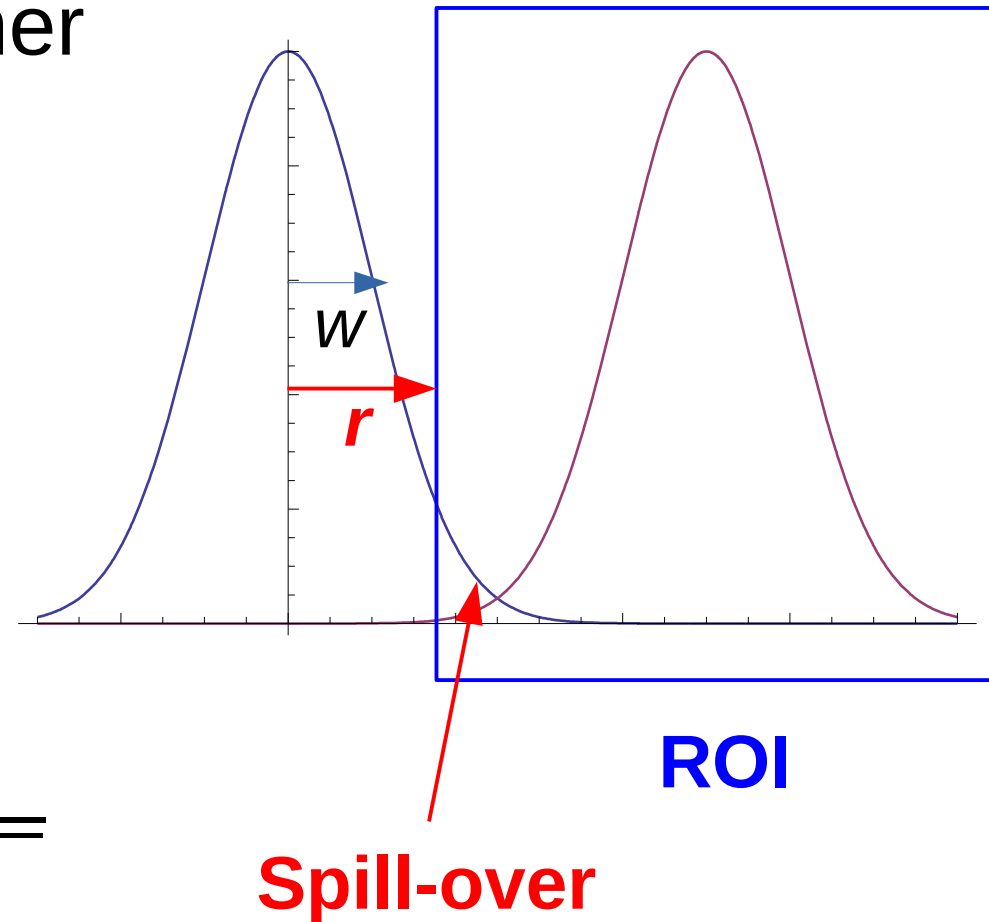


# Bias and Uncertainty

- Cut off a Gaussian PSF at  $r$  and overlap with another PSF

$$\text{bias} = \frac{w e^{-r^2/2w^2}}{\sqrt{2\pi}}$$

$$\text{std. dev.} = \frac{w}{\sqrt{N_{\text{photons}}}}$$



r depends on error tolerance

$$\text{bias} \leq \alpha^* \text{std. dev.}$$

$$r = w \sqrt{\log \frac{N_{\text{photons}}}{2\pi\alpha^2}}$$


# Prediction

$$\sigma = \text{Density} \leq \frac{\eta}{\pi r^2} = \frac{8\pi\beta\eta\text{NA}^2}{\lambda^2 \log \frac{N_{\text{photons}}}{2\pi\alpha^2}}$$

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Depends on  
percolation model,  
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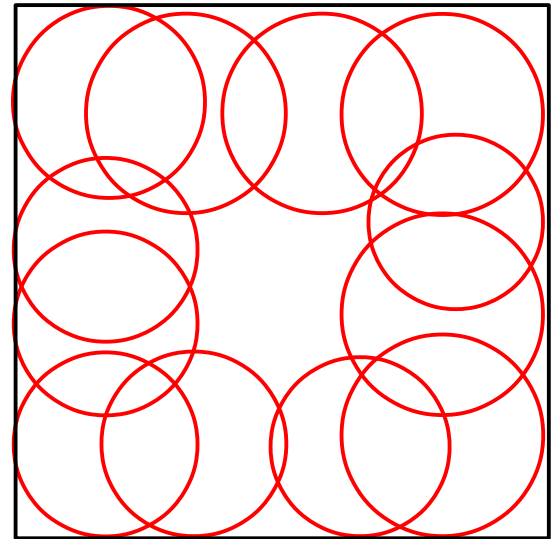
#'s: 647 nm light (**Alexa**), 1.45 NA objective, ~750 photons,  
 $\alpha=1$

$\sigma=8.35$  fluorophores/ $\mu\text{m}^2$  before edge effects matter

→*Near-exact match to Huang & Lidke 2011*

# Interpretation:

- Higher density: molecules not in middle of ROI are mis-localized.
- For  $<20$  fluorophores/ROI (typical)  
and  
density  $> p_c$   
majority of fluorophores  
contaminated with spill-over



# Consequence: Speed

Max Fluorophores per Frame	Normalized Min. # frames needed	Speed improvement
1	1	1
2	0.44	2.27
3	0.26	3.85
4	0.19	5.26
5	0.14	7.14
10	0.023	43.5

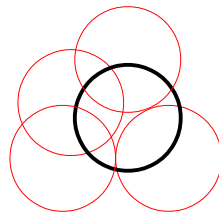
# Localization precision

- Fundamental Limit: Cramer-Rao Lower Bound (CRLB)
- CRLB calcs can be done for multi-flourophore fits (e.g. Yi Sun, JBO, 2013)



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- Percolation: ~4-5 neighbors per fluorophore
- Can a “mean-field” model get close to more precise calculations?

# Thompson, Larson, Webb 2002

- Formula not exact, but often useful.
- For spill-over background and pure shot noise, formula simplifies to:

$$\text{Var}(x) = \frac{w^2}{N} \left( 1 + \frac{4 \eta w a_{\text{pixel}}}{\sqrt{\pi} r^2} \right) + \frac{a_{\text{pixel}}^2}{N}$$

Typical #'s:  $w \sim 0.5r$ ,  $a \sim 0.4r$

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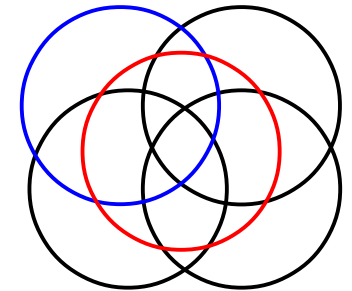
**→Off from Sun by ~2**

# Simplest cluster model

Explicitly construct Fisher information matrix

Central: 4 equidistant neighbors @  $2w$

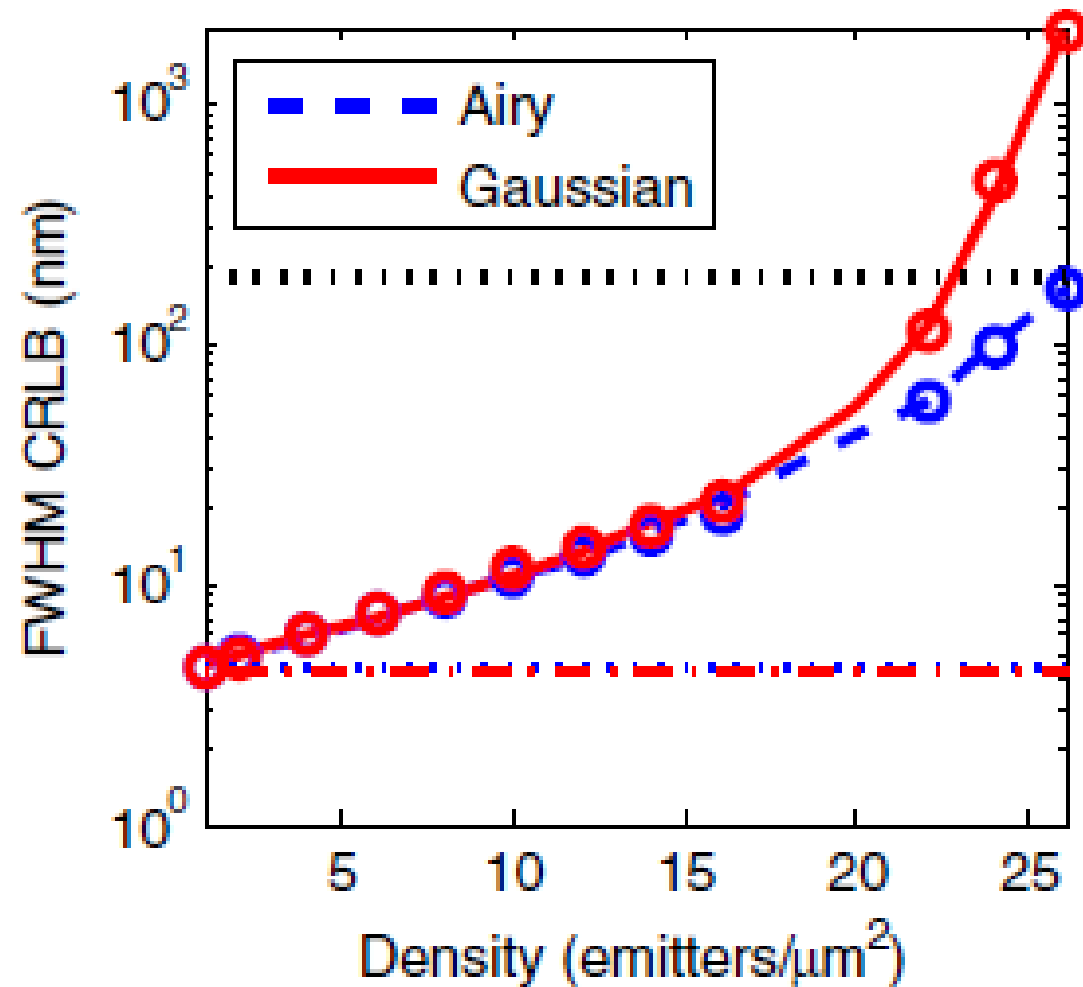
→ Off-diagonals dominated by  $r=w$ ,  
→ diagonal/ $e^{0.5}$



Corners: 1 neighbor @  $w$  (diagonal/ $e^{0.5}$ )  
2 neighbors @  $\sqrt{2}w$  (diagonal/ $e$ )  
1 neighbor @  $2w$  (diagonal/ $e^2$ )

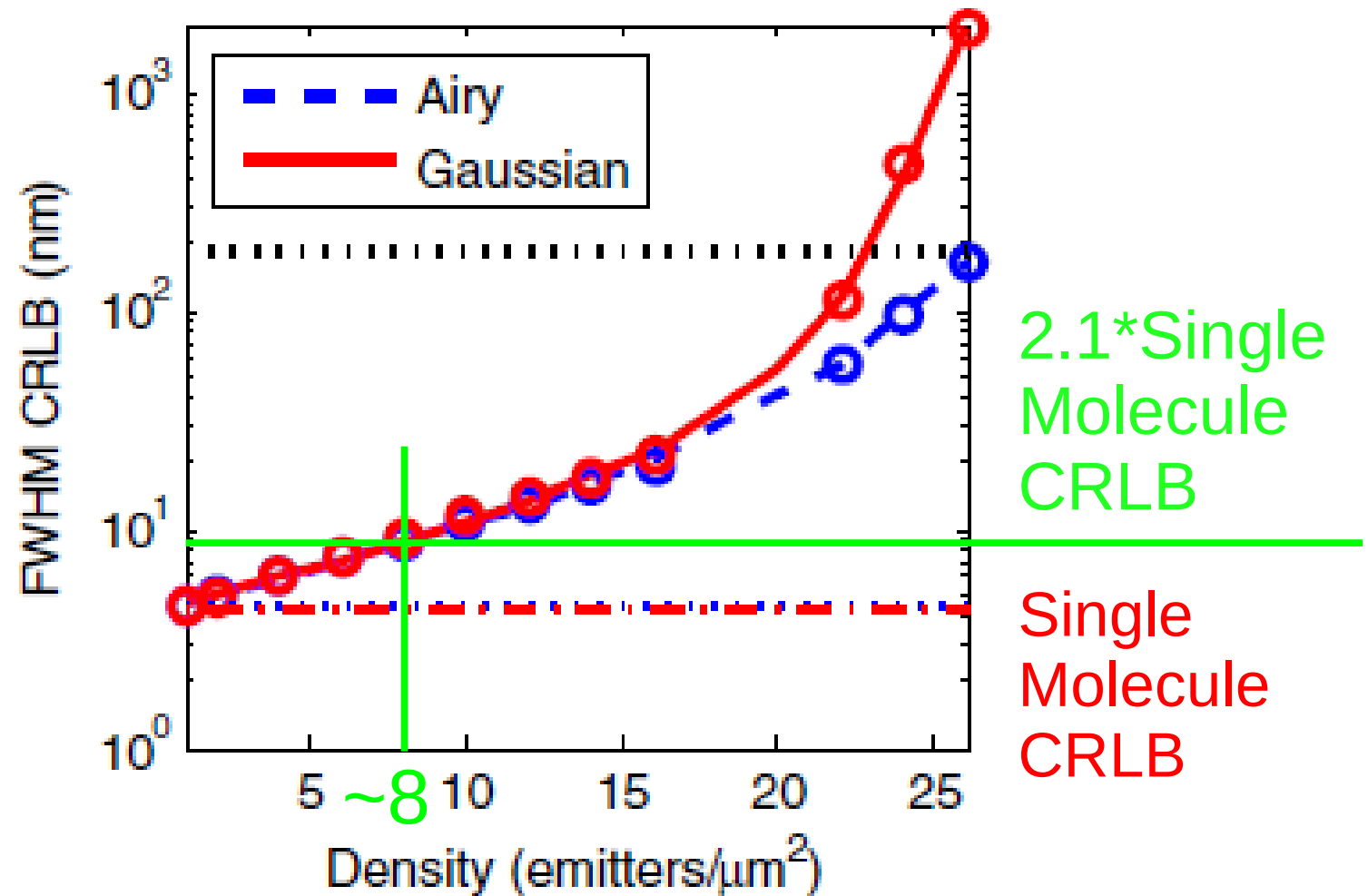
***Std. Dev. Of center = 2.1\*Single-Molecule Case!***

# From Sun, 2013



Single  
Molecule  
CRLB

# From Sun, 2013



# Conclusions

- Multi-fluorophore localization performance degrades due to percolation effects
- Localization precision: fluorophore has an average environment of 4 neighbors
- Speed improvements beyond  $\sim 7$  unlikely w/o accuracy trade-offs and/or large ROIs

# Acknowledgments

- Some of this work started during stay at LFD (Enrico Gratton, UCI)
- Discussions of percolation w/ Suketu Bhavsar (CPP)
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  - Jon Tran
  - Ted Yoo
  - Carina Kaainoa
  - Kevin Djepang
  - Brianna Thierjung