# Interpreting Single Particle Tracking Data with a Correlated Random Walk and a first-passage time algorithm

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#### Outline

#### Introduction

Correlated Random Walk Model

Variance First-Passage Time

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# Single Particle Tracking

- Single Particle Tracking is an experimental technique used to address the following general question: How do individual biomolecules move within the cell?
- Individual particles are labeled with an optical bead or a fluorescent tag, and observed with video microscopy.

Data from SPT experiments is obtained as N + 1 position coordinates:



$$p_0 = (x_0, y_0), p_1 = (x_1, y_1), \vdots \vdots \\ p_N = (x_N, y_N).$$

# Mean Square Displacement and the Diffusion Coefficient

- ► A common measure used to characterize the movement of particles is the diffusion coefficient, *D*.
- D is estimated from Mean Square Displacement (MSD), a quantity that describes the average of the squared displacements of a particle's trajectory.



- MSD at time *t*, is defined by  $\rho(t) = \langle (r(t) r(0))^2 \rangle$ .
- In two dimensions,  $\rho(t) \approx 4Dt$ .

# Mean Square Displacement and the Diffusion Coefficient



# LFA-1 (Leukocyte function-associated antigen 1)

- Integrin T-cell adhesion receptor allows T-cells to adhere to endothelial cells by binding to ICAM-1 (intercellular adhesion molecule) on antigen-presenting cells (endothelial cells).
- LFA-1 facilitates transmigration of leukocytes across vascular endothelia in processes such as extravasation and the inflammatory response.



#### Macroheterogeneity Observed in LFA-1



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#### Macroheterogeneity – Microheterogeneity



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# Some Questions

- What is the mechanism of movement of LFA-1 proteins?
- Is there a common mechanism that governs the whole population or multiple movement mechanisms?
- Is there movement heterogeneity within the population (macroheterogeneity) and/or within individual trajectories (microheterogeneity)?

#### Our Approach

Movement tracks are also analyzed in ecology, to understand the movement of animals.

- Can ecological models be used to provide information about SPT data?
- Move lengths and turning angles are often used in Ecology. What can length and turning angle distributions tell us about SPT movement tracks?
- How can we use ecological tools to tell us more about both macro and micro heterogeneity?



Kareiva et al., 1983.

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#### Turning Angles and Step Lengths



#### Angle and Length Distributions



# Are these distributions alone enough to describe the observed LFA-1 trajectories?

#### Correlated Random Walk

- A random walk in which the directions of subsequent moves are correlated.
- Data is measured in terms of move lengths and turning angle probabilities.
- Conclusions are derived from analyzing collective distributions of data.

Can the movement mechanism of the LFA-1 protein be described by the length and turning angle distributions?

#### Correlated Random Walk: The Patlak Model

The model is given by the partial differential equation:

$$\frac{\partial u}{\partial t} = \frac{1}{n} \nabla \cdot \left[ \frac{1 + \psi \left( 2\frac{m_1^2}{m_2} - 1 \right)}{1 - \psi} \nabla \left( \frac{m_2}{2\tau} u \right) - \frac{\psi m_1^3}{\tau m_2 \left( 1 - \psi \right)} \nabla \left( \frac{m_2}{m_1} \right) u \right]$$

- $m_1$  = average move length
- $m_2$  = average squared move length
  - au = average move duration
- $\psi_{-}$  = average of the cosines of turning angles

#### Correlated Random Walk: The Patlak Model

This model gives rise to a theoretical formulation for MSD, called Net Squared Displacement:

$$\overline{R_n^2} = nm_2 + 2m_1^2 \left[ \frac{(\psi - \psi^2 - s^2)n - \psi}{(1 - \psi)^2 + s^2} + \frac{2s^2 + (\psi + s^2)^{\frac{n+1}{2}}}{((1 - \psi)^2 + s^2)^2} \gamma \right]$$

$$\gamma = ((1 - \psi)^2 - s^2) \cos((n+1)\alpha) - 2s(1 - \psi) \sin((n+1)\alpha)$$

where  $m_1, m_2, \psi$  and s are calculated from data as

$$m_1 = \frac{1}{k} \sum_{i=1}^k \ell_i$$
  

$$m_2 = \frac{1}{k} \sum_{i=1}^k \ell_i^2$$
  

$$\psi = \frac{1}{k} \sum_{i=1}^k \cos(\theta_i)$$
  

$$s = \frac{1}{k} \sum_{i=1}^k \sin(\theta_i)$$

#### Testing the CRW with a Bootstrapping Procedure



- The MSDs of the pseudotrajectories (blue) surround the theoretical Net Squared Displacement (red) of the experimental data.
- They do not encompass the MSD of the experimental data (black).

#### Testing the CRW for Individual Trajectories



Out of 75 experimental trajectories, only 10 passed the bootstrapping procedure.

What about the rest?

#### Anomalous Diffusion



Lateral transport modes on the cell surface. (A) Transient confinement by obstacle clusters (B) or by the cytoskelaton, (C) directed motion, and (D) free random diffusion.

www-cellbio.med.unc.edu

- (A) Transient Confinement Zones due to obstacles
- (B) Transient Confinement Zones due to cytoskeleton binding
- (C) Directed Motion
- (D) Random Brownian Motion

#### Modeling Transient Confinement

- ► At each step of a CRW, decide whether the particle should become confined with probability P - in
- If confined, decide whether the particle should leave the confinement zone with probability P - out



#### Transient Confinement and the Patlak Model



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#### First-Passage Time and its Distribution



FPT (First Passage Time): number of steps that a particle or individual takes in a circle of radius *r* centered on each step of the trajectory.

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Distribution given by the FPT for r

#### Variance First-Passage Time



S(r)(Variance First-Passage Time): Variance of the distribution given by the FPT for r. It is a measure of the amount of heterogeneity at the spatial scale r.

# How does variance FPT look for different transient confinement parameters?



(A)  $p_i = 1$  and  $p_o = 0$ ; 50nm, 100nm, 200nm, 500nm, 1000nm

(B)  $r_c = 50nm;$  $p_i = 0, 0.2, 0.4, 0.8, 1$  $p_o = 0.1$ 

(C)  $r_c = 50nm;$  $p_o = 0.05, 0.2, 0.4, 0.8, 1$  $p_i = 0.9$ 

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#### Application to LFA-1



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#### Looking at Macroheterogeneity



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# **Classifying Population Structure**

- Changes in population structure after treatment of PMA
- For a particular label (TS1/18), PMA activates the cell: increases mobility of LFA-1 proteins



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#### Distribution of Peaks in Variance First-Passage Time

 The areas of concentrated movement change after treatment of PMA



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#### **Detecting Cluster Sizes**

Combining MEM148 Antigen with PMA shows concentrated diffusion within ≤ 50nm ⇒ clustering?



#### LFA-1 Receptor Clustering



 Experimental evidence for LFA-1 clusters on activated cells is limited

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# Conclusions

- Ecological approaches can provide insights into the movement mechanism of individual proteins.
- Both the CRW model and variance FPT provide some ability to monitor and filter macroheterogeneity.
- Variance FPT is a useful method to detect microheterogeneity and suggests receptor clustering.

#### THANKS YOU!

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