FRET Live Cell Imaging and Quantitation Workshop UCSD-2015

Live-Cell Image Processing, Segmentation, Tracking, and Interpretation

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Image Analysis Bridges Experiments and Computational Models



Live-Cell Imaging Provides Tremendous Amount of Data



Live-Cell Imaging Provides Tremendous Amount of Data 40,000 images in 1 hour



Strength of Automated Analysis

- Large Scale, Quantitative, and Objective
- Reveals Non-obvious and Hidden Information
- Systematic Understanding of Dynamic Images
- Multi-channel and Multi-dimensional

Three levels of Image Analysis



MATLAB for Image Processing



MATLAB Basics – Imagetool

>> im = imread('corn.tif'); >> imtool(im);





MATLAB Basics – Figure

>> im = imread('ngc6543a.jpg');
>> figure; imshow(im);



MATLAB Basics – Figure



In the matrix im: y --- row number ; x --- column number.

Color Image – Red, Green, and Blue (RGB)

HSV/HSI Color Image Hue, Saturation, Value/Intensity



An HSV color system is close to human color interpretation.

- Hue : a pure color with values in the range of [0 360]
- Saturation: the degree to which a pure color is diluted by white light [0 1]
- Value/Intensity: brightness [0 1]

Example: ECFP/FRET/mCherry/DIC Image (Fluocell) http://code.google.com/p/fluocell/

ECFP



FRET



mCherry







ECFP/FRET Ratio in Pseudo Color



Color Image - Intensity Modified Display

HSV Color Image : hue = ECFP/FRET Ratio, s = 0, value = ECFP intensity

ECFP/FRET Ratio in Pseudo Color



Live-cell Video Images of ECFP/FRET Ratio



Quantitative Analysis of Live-cell Images



Image Segmentation - Thresholding

Suppose that the histogram corresponds to the image (matrix) imA(i, j).



The thresholded (binary mask) image maskA can be defined as

$$maskA(i, j) = \begin{cases} 1, & \text{if } imA(i, j) > T \\ 0, & \text{if } imA(i, j) \le T \end{cases}$$

Use Image Smoothing to Improve Global Threshold



abc def

FIGURE 11.15 (a) Noisy image, and (b) its histogram. (c) Result obtained using Otsu's method. (d) Noisy image smoothed using a 5×5 averaging mask, and (e) its histogram. (f) Result of thresholding using Otsu's method.

Image Segmentation – The water algorithm

The Water Algorithm for Separating Bright Objects Zamir E et al. 1999, J Cell Sci 112,1655–1669 Balaban NQ et al. 2001, Nat Cell Biol 3, 466–472



Image Segmentation – Focal Adhesion Detection



Two problems:

- 1. The diffusive background is not uniform.
- 2. Bright spots of focal adhesions are connected.

Tracking – Single Particle Tracking

- Jaqaman et al 2008 Nature Methods 5, 8, 695
- Jaqaman et al 2013 Cell, 145, 593-606
- Chenouard N et al 2014 Nature Methods 11, 3, 281

Particle Tracking/Linking

- An objective comparison of particle tracking methods based on an open competition in 2012.
- Research groups world wide were invited to participate.
- Registered teams were given 1 month to prepare their methods.
- After release of the actual competition data set, without ground truth, the teams submitted tracking results to an independent evaluator.

Quantitative Analysis of Live-cell Images



Correlative FRET Imaging Microscopy (CFIM)



Lu S. et al, Scientific Reports, 2014

Growth Factor Induced Src Activation and Focal Adhesion Disassembly



FA Detection and Quantification





Putting Single Cell Data Together



Putting Different Signals Together



Magnitude Coupling



- <u>Linear Correlation Coefficient R</u>: measures the degree of Src-paxillin magnitude coupling.
- <u>Slope S</u>: measures the capacity of Src activity in causing paxillin disassembly, or the amount of paxillin disassembly per unit of Src activation.

Dynamic Coupling



- <u>Cross-correlation Peak– K</u>: measures the similarity between the time courses of Src activation and paxillin disassembly.
- <u>Time Lag T</u>: The time lag between Src activation and paxillin disassembly.

Effect of Matrix Protein Concentration



Molecular Model of coupling between Src activity and FA disassembly



Lu S. et al, Scientific Reports, 2014

Photobleaching and Photoactivation Following Intracellular Protein Dynamics

- Lippincott-Schwartz J et al. 2003 Nat Cell Biol.
- Klonis N et al. 2002 European Biophys J.
- Lu S et al. 2008 PLoS Comput Biol.
- Capoulade J et al.
 2011 Nat Biotech.



Photobleach Analysis

	Classical Analysis Based on Recovery	Finite Element Model (FEM) Based Analysis	Fluorescence Correlation Spectroscopy (FCS)
Experimental Duration	Long	Short	Short
Signal to Noise Ratio	High	High	Low
Flexible PB Pattern	No	Yes	N/A
Flexible Cell Geometry	Νο	Yes	Yes

Fluorescence Recovery After Photobleaching (FRAP)



Lu S. et al, Scientific Reports, 2014

Diffusion Model (2D) and FE-discretization

$$\frac{\partial u(x, y, t)}{\partial t} = D \bullet \Delta u(x, y, t)$$

Finite Element Discretization (Crank-Nicholson Scheme in Time)

$$\frac{M(u^{n+1} - u^n)}{dt} = D \cdot \frac{-K(u^{n+1} + u^n)}{2} + r$$

Comparison of Experiment and Simulation

Photobleach





Comparison of Experiment and Simulation



Subtraction of Diffusion in FRET Imaging



Fluocell Image Analysis Software Package



Download: http://code.google.com/p/fluocell

Email: kalu@ucsd.edu

Acknowledgments

P. I. and Lab Members:

Dr. Yingxiao Wang Dr. Mingxing Ouyang Dr. Ji-Hye Seong Dr. Yue Zhuo Yi Wang Eddie Chang Lexie Qin Qin Shirley Yi-Qian Wu

Collaborators:

- Dr. Shu Chien, UCSD
- Dr. Jin Zhang, JHU
- Dr. Randy Bank, UCSD
- Dr. Philip Gill, UCSD
- Dr. Michael Holst, UCSD

NIH, NSF, NSF/NIH Math/Bio Initiative, Beckman Foundation IEM, CCOM, and Bioengineering at UCSD