Quantitative Microscopy:

Bridge between "wet" biology and computer science

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System Biology: Experiment Driven Model and Model Driven Experiment





Quantitative Microscopy in wild : Learning endocytosis kinetic model from pulse-chase experiment

Endocytosis: Cellular Organelles Transport Cargo













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Endocytosis: Cellular Organelles Transport Cargo



Simple Cargo Flow Model



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Pulse-chase experiment



≻HeLa cells

≻30'' pulse of fluorescent cargo

Chasing at set of time points

≻Fixing and staining by antibodies endosomal markers

➢Imaging

>Endosome identification

>Colocalization cargo with endosomal markers

First step from "wet" experiment to computer: imaging



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Endosome identification: What is challenge?



- •Wide dynamic range of vesicle intensity
- Variety of shapes
- Non-homogenous background
- Multiplicative (Poisson) noise

Consider intensity distribution as a surface in 3D-space



Intensity distribution of individual small endosome can be modeled by hat-like function

$$I(x, y) = \frac{A_0}{1 + \left\{ \left[\frac{((x - x_0)\cos(\alpha) - (y - y_0)\sin(\alpha)}{w} \right]^2 + \left[\frac{(x - x_0)\sin(\alpha) + (y - y_0)\cos(\alpha)}{h} \right]^2 \right\}^p}$$

where A_0 - intensity at the center

 x_0, y_0 - center coordinates

w, h - width by perpendicular dimensions

$$\alpha$$
 - angle between main axis and axis Y

p – power factor



Base function is fitted to the image by minimization χ^2



Endosomes with complex shape can be modeled by sum of base functions



Fitting requires image noise analysis

$$\chi^{2} = \sum_{i,j} \frac{\left(I_{i,j} - F_{i,j}\right)^{2}}{\sigma_{i,j}^{2}}$$

Photon flux calibration



Bonus: Number of photons per object

a) Counting number fluorescent molecules

b) Single molecule experiments



Result of image fitting



Benefits of approach

- ✓ High dynamic range
- ✓ Sub-pixel resolution

✓ Accurate estimate of mean intensity, integral intensity and endosome size

✓ Possibility to count number of fluorescent molecules per endosome

Are we ready to model cargo flow?

We did not check two major microscope problems:

≻Uneven illumination of view filed

Chromatic shift/aberration

Quantitative Microscopy: uneven illumination



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Quantitative Microscopy: uneven illumination







$$\delta x_1 = 7.31, \, \delta y_1 = 3.37, \, \alpha_1 = 0.00285$$

 $\delta x_2 = 0.26, \, \delta y_2 = 10.83, \, \alpha_2 = 0.0041$







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Colocalization: World is 3D



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- 3. Noise in neighbour pixels is correlated

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Does it matter?

Correct model choice and noise amplitude: How many components model has?



Deconvolution of artificial two-exponential curve (τ_1 =0.2 and τ_2 =0.4, signal-tonoise ratio = 200) by one exponent (τ =0.286) (panels A, C) and two exponents (τ_1 =0.192 and τ_2 =0.385) (panels B, D).

Model search is χ^2 minimization problem



Model fit and noise correlation

Simulation model:

 $y_i = \exp(-x_i) + \varepsilon$, where ε has distribution N(0.0, 0.05)

Regression model:



Model fit and noise correlation





Take home message:

- Analyze image (if it is possible) before any correction and apply correct to the result of analysis latter.
- Avoid automatic (hidden) correction of images by microscope software with unknown characteristics.
- Check actual image noise on consistence with noise model, which is assumed by analyzing software.

Next step: We need 'just' quantitative colocalization to compare model with experiment



Pearson & Overlap colocalization criteria

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Colocalization of Fluorophores in Confocal Microscopy

Two or more fluorescence emission signals can often overlap in digital images recorded by confocal microscopy due to their close proximity within the specimen. This effect is known as colocalization and usually occurs when fluorescently labeled molecules bind to targets that lie in very close or identical spatial positions. This interactive tutorial explores the quantitative analysis of colocalization in a wide spectrum of specimens that were specifically designed either to demonstrate the phenomenon, or to alternatively provide examples of fluorophore targets that lack any significant degree of colocalization.

Specimen Image Colocalization Scatterplot Channels: 0 Intensity (Green Channel) 🔽 Red 🔽 Green 🔽 Blue G 2D G 3D Choose A Specimen **Colocalization Channels:** pithelial Cell Organelles Red-Green Red-Blue Green-Blue Colocalization Coefficients: Colocalization Overlay: Pearson's R(r): 0.19 🔽 Full Color 🔲 Binary

Overlap (R): 0.85

Global:

Red: 0.99 Green: 1.0

Freehand ® Rectangle © Ellipse



Large flat rabbit kidney epithelial cells (RK-13 line) were transfected with a DsRed2 (red) fluorescent protein vector localized to the endoplasmic reticulum, fixed, and stained for filamentous actin with Alexa Fluor 488 (green). The cell nuclei were counterstained with Hoechst 33342 (blue). Colocalization is seen in the red and blue channels between the nuclei and endoplasmic reticulum.

Region of Interest:

Pearson's colocalization coefficient:



Overlap coefficient:



Estimation of random (apparent) colocalization





Pearson colocalization time

course



Pearson colocalization time

course



Are domain structures colocalized?



Definition of asymmetric colocalization

The green vesicle is colocalized with red if more then \mathbf{x} % of its area is covered by red vesicle(s)



Lets threshold = 50%, then green is colocalized with red, but red is not colocalized with green. *This definition of colocalization is asymmetric, but... better fit our intuition*

We don't have deal with pixels anymore. Now we are working with objects



Original image



Image reconstructed from the fitted functions

Object-based estimation of random (apparent) colocalization



Random Colocalization



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Unknown microscopy: Is chromatic shift constant over experiment? (How long did you worm microscope before imaging?)

The 10min uptake of Tf-Alexa488 and Tf-Alexa647 mixture by HeLa cells. Automatic microscope imaging

(12 images per well)



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Problems of 'wet' experiment: Analysis of pulse-chase experiments (chromatic shift)





Left and right images have indexes 'a' and 'e' in same imaging session.

HeLa: 30 sec pulse EGF & Tf, chase 10 min. Red: EEA1; Green: EGF; Blue:APPL1; Cyan: Tfn

Problems of 'wet' experiment: Analysis of pulse-chase experiments (chromatic shift)





Shift correction on 'per-frame' basis.

HeLa: 30 sec pulse EGF & Tfn, chase 10 min. Red: EEA1; Green: EGF; Blue:APPL1; Cyan: Tfn

Problems of 'wet' experiment: Analysis of pulse-chase experiments (chromatic shift)



Cargo for degradation (EGF)

Cargo for recycling (Tfn)



Simple Cargo Flow Model



Simple Cargo Flow Model

$$\frac{dC_{pm}}{dt} = \Phi_{in}(t) - k_1 C_{pm}$$

$$\frac{dC_{ccv}}{dt} = k_1 C_{pm} - k_2 C_{ccv}$$

$$\frac{dC_{APPL}}{dt} = k_2 C_{ccv} - k_3 C_{APPL}$$

$$\frac{dC_{APPL-EEA 1}}{dt} = k_3 C_{APPL} - k_4 C_{APPL-EEA 1}$$

$$\frac{dC_{EEA 1}}{dt} = k_4 C_{APPL-EEA 1} - (k_5 + k_6) C_{EEA 1}$$

$$\frac{dC_{RE}}{dt} = k_5 C_{EEA 1} - k_7 C_{RE}$$

$$\frac{dC_{LE}}{dt} = k_6 C_{EEA 1}$$

Model meet Experiment: Do we get global optimum?



Local minimum traps in the model fitting procedure

The application of theoretical models of complex shape to the fitting of experimental spectra having closely overlapping bands

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The real life is more complicate than our simple model (to be continued...)



Conclusion

- Quantitative comparison model and experiment requires quantitative analysis of all steps, which connects "wet" experiment and "digital" data
- "Small" imperfections of experimental equipment could lead to wrong model selection

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