Non-parametric change detection methods in fluorescence life cell imaging for sub-cellular trafficking and exocytosis analysis

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Outline

- Introduction to biological context and motivations
- Non-parametric method for sparse event detection in TIRF microscopy
- Experimental results and validation

Transmembrane proteins under study

Two constitutively recycling receptors

Transferrin receptor

Well studied canonical recycling transmembrane protein

Langerin

- Major component of Birbeck granules
- C-Type Lectin with carbohydrate recognition domain (CD 207)



Birbeck granule in EM microscopy [Lipsker 2003]



Langerin-YFP in M10 stable cell lines (spinning disk microscopy)

Experimental set-up

Is endocytosis/recycling altered by shape constraints? Are events located in specific cell regions?



2D+time TIRF microscopy (100ms / frame)

Micro-patterning : shape constraint for variability reduction [Thery 2005]

Comparison of dynamic recycling events

Two independent image sequences and two different micro-patterned cells



Langerin-YFP and TfR-YFP in TIRF microscopy using micro-patterning [Thery 2005]

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Two independent image sequences and two different micro-patterned cells



Langerin-YFP and TfR-YFP in TIRF microscopy

Langerin's recycling is Rab11 dependent ...

Dual view of a single micro-patterned cell / two synchronized TIRF image sequences





Rab11 (L) and Langerin (R) in TIRF microscopy

Langerin / Rab11 colocalization

- Space-time organization and coordination of membrane trafficking
- Biogenesis of the recycling compartment and related ultra-structures
- Molecular events and recycling scenarii
- Specificity of addressing mechanisms

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Framework for detection in independent sequences: *"Automated Event Detection"*

Non-parametric detector: Probability of finding no similar patch within a neighborhood in the subsequent images is very low



PATCH-BASED ALGORITHMS

✓ <u>DENOISING</u>

C. Kervrann & J. Boulanger, IEEE IP, 2006, Boulanger et al., IEEE PAMI, 2007; C. Kervrann & J. Boulanger, IICV, 2008; J. Boulanger et al., IEEE TMI, 2010; Carlton et al., PNAS, 2010: Saliba et al., PNAS, 2010; Dokudovskaya et al., Mol Cell Prot, 2011...

✓ EVENT DETECTION

J.Boulanger et al., PLoS One, 2010; A. Chessel et al., Proc. of IEEE ISBI'10; C. Kervrann et al., SIAM MMS, 2011



How to detect ONE "meaningful" event on average in an image pair?



Lateral movement of vesicles (trafficking) and global movement of the cell prevent the application of usual "motion detection" algorithms.

- Things to do :
 - Track all the vesicles as much as possible
 - Extract the beginning and ending of the trajectories...
- Problems :
 - Isn't it an overkill ?
 - Robustness of the tracking method ~ over-detection ?
 - How to control the number of false alarms?

Image redundancy and notations

- Definition and notations : Let u = (u(x))_{x∈Ω} and v = (v(x))_{x∈Ω} be an image pair defined over a bounded domain Ω ⊂ ℝ².
- Non-parametric detection :
 - A change occurs at pixel $x \in \Omega$ if we find no match between a *n*-dimensional patch $\underline{u}(x)$ from *u* and patches $\underline{v}(x)$ from *v*.
 - We consider a fixed size search window $B(x) \subset \Omega$ (semi-local neighborhood) where N = |B(x)| is the number of tested patches.



Local scores and collective decision fusion

Step 1 : For each pixel y ∈ B(x), we compute a score (dissimilarity measure) between patches :

 $\Phi(\underline{u}(x),\underline{v}(y))$

and we compare the score to a threshold $\tau(x)$ (e.g. $\Phi(\cdot) = \|\cdot\|^2$).

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• Step 2 : At each pixel x, we count the total number of positive decisions and we define $D(x) \in \{0, 1\}$ as :

$$D(x) = \begin{cases} 1 & \text{if } \sum_{y \in B(x)} \mathbb{1}[\Phi(\underline{u}(x), \underline{v}(y)) \ge \tau(x)] = N \\ 0 & \text{otherwise} \end{cases}$$

... a change occurs at location x if all the scores are higher than $\tau(x)$ (maximum vote).

Advantages and difficulties...

The setting of thresholds $\tau(x)$, the search window size *N* and the patch size *n* must be addressed to make the procedure successful!

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- Motion (traffic) in the background is not well defined
- Collaborative neighborhood-wise decisions (space and scale) is attractive for implicit spatial regularization (patch overlapping) [Boulanger 2010]
- Performance analysis of detectors in terms of false alarm rates
- No energy minimization procedure
- No optical flow estimation

Analysis of background motion amplitude

• Local detection threshold : Given B(x), $\tau(x)$ is defined as the highest score computed from the reference image u :

$$\tau(x) \stackrel{\triangle}{=} \sup_{y \in B(x)} \Phi(\underline{u}(x), \underline{u}(y))$$

• Size of the search window B(x): The choice of N depends on the motion amplitude we do not want to detect... The value $\frac{\sqrt{N}}{2}$ is related to the expected amplitude of background dynamics¹.

1. Let $s(x) \in \mathbb{R}^2$ be a 2D brownian motion. If $u \in \mathbb{R}^{\Omega}$ is K-Lipschitz $(|u(x) - u(y)| \le K ||x - y||)$, then

$$\mathbf{P}\left(\|s(x)\| \geq \frac{\sqrt{N}}{2}\right) \leq \frac{nK^2\mathbb{E}[\|s(x)\|^2]}{\tau(x)}$$

From local to global decisions

 Motivation : We need to make a pointwise decision from large spatial and multiscale contexts... detecting unusual events (one spot on average) when comparing two images.

Multiscale modeling :

- Let $\{D_1(x), \dots, D_L(x)\}$ be the set of binary decisions at pixel x, where $D_{\ell}(x)$ is the decision made for a given patch size $n_{\ell} = (2\ell + 1)^2, 1 \le \ell \le L$ and L is the number of sizes considered at each location.
- ² The binary Bernouilli variables $D_{\ell}(x)$ at pixel *x* are correlated because the patches with different sizes are nested.

Patch-space framework

- **Property** : Let *X* be a random variable defined as the sum of *L* correlated Bernouilli variables. *X* is known to tend to a Poisson law in distribution as $L \rightarrow \infty$ (Chen-Stein method [Arratia 1989]).
- **Multiscale decision** : The probability of false alarm $P_{FA}(x)$ is given by the Poisson tail with parameter λ . A change occurs at pixel x if $(S_L(x) = \sum_{\ell=1}^{L} D_\ell(x))$:

$$egin{array}{rcl} \mathsf{P}_{\mathit{FA}}(x) & \stackrel{ riangle}{=} & \mathsf{P}(X \geq \mathcal{S}_L(x) | \mathcal{H}_0) \ pprox \ 1 - \sum_{k=0}^{\mathcal{S}_L(x)} (\lambda)^k \, rac{e^{-\lambda}}{k!} \ & \leq & rac{arepsilon}{|\Omega|} \end{array}$$

... on average 1 pixel is falsely detected if $\varepsilon = 1$ [Desolneux, Moisan and Morel, 2000].

A practical algorithm ...

Kervrann et al., SIAM J. Multiscale Modeling & Simulation, 2011

Let us consider 3×3 search windows B(x) and L patch sizes.

1 For $\ell = 1 \cdots L$

- For each pixel $x \in \Omega$ compute $\tau_{\ell}(x) = \sup_{y \in B(x)} \Phi(\underline{u}_{\ell}(x), \underline{u}_{\ell}(y))$
- For each pixel $x \in \Omega$ compute $D_{\ell}(x) \in \{0, 1\}$:

$$\mathcal{D}_\ell(x) = \mathbb{1}\left[\sum_{y\in \mathcal{B}(x)} \mathbb{1}[\Phi(\underline{u}_\ell(x), \underline{v}_\ell(y)) \geq au_\ell(x)] = \mathcal{N}
ight]$$

² Compute $\lambda = \frac{e^{-N}}{|\Omega|} \sum_{\ell=1}^{L} \sum_{y \in \Omega} e^{\sum_{z \in B(y)} \mathbf{1}[\Phi(\underline{u}_{\ell}(x), \underline{v}_{\ell}(y)) \ge \tau_{\ell}(x)]}$

a "meaningful" change occurs a location $x \in \Omega$ if $\mathsf{P}_{F\!A}(x) \leq \frac{\varepsilon}{|\Omega|}$.



Rab11 (L) and Langerin (R) in TIRF video-microscopy (frame # 192)



Rab11 (L) and Langerin (R) in TIRF video-microscopy (frame # 193)



Difference image (Rab11 (L) and Langerin (R))



Probabilities of false alarms (regions of interest are in red)



Analysis on significant fluorescence signals



Probabilities of false alarms from significant fluorescence signals



Localization of "meaningful" appearances



Cumulative change detection map

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Patch-Based Event Detection (PBED) algorithm



Detection of of appearances (green) and disappearances (rent) for YFP-Langerin in M10 stable cell lines acquired in TIRFm (100ms / frame)

Basic statistical analysis



Rab11a escorts the vesicles until the fusion with the plasmic membrane



Workflow from photons to signal clusters



Sparse detection of "meaningful" events

- Signal descriptors : convex hulls (computational geometry and alpha-shapes / "Hullkground" algorithm²)
- Event classification : Hausdorff's distance and spectral clustering

^{2.} A.Chessel et al., Proc. of SSVM'09; Proc of IEEE ISBI'10

Classification of two cargo proteins



Quantification of the heterogeneity of Langerin recycling behaviors
Measurement of "docking-fusion" times of vesicles

Specific regulation of the Langerin cargo recycling by the Rab11A/Rab11FIP2/MyoVb platform³



- MyoVb plays a crucial role
- Rab11A dissociates concomitantly to the docking/fusion process
- Rab11FIP2 is a sensor for regulation of fusion
- Cargo specificity

3. Boulanger et al, PLoS ONE 2010; Gidon et al., Traffic 2012; Chessel, Cinquin et al., in revision

PBED : General tool for endo-exocytosis event analysis

- Automatic detection of docking/membrane fusion events
- Quantification of space-time interactions

Change detection algorithm :

- Patch-based image representation
- Intuitive algorithm parameters and implicit regularization
- Collaborative neighborhood-wise decisions (space and scale)
- Performance analysis (false alarm probabilities)
- So energy minimization procedure and no optical flow estimation

 $\sim 85\%$ of success rate wrt manual labeling (several hundred image pairs)!

Space-time coordination of membrane trafficking

Langerin's recycling is Rab11 dependent ...



Platforms and molecular complexes

Traffic overview

- Biogenesis of the recycling compartment and related ultra- structures
- Molecular events and recycling scenario
- Specificity of addressing mechanisms