# Spatial statistics and attentional dynamics in scene viewing

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#### Gaze control in natural scenes



(from: Henderson, 2003, Trends Cogn Sci)

# Computational modeling of visual attention

Input image



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### Examples of saliency computations

#### Stimulus display



Saliency map

(from: Einhäuser & König, 2010, Curr Opin Neurobiol)

Human fixations

# Examples of saliency computations (cont'd)



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# Application of the theory of spatial point processes



- red: fixation locations recorded from human observers
- shading: local intensity λ(x), bandwidth σ = 3°
   red lines: single-trial scanpath (fixations 2 to 24)

### Spatial point processes: First-order statistics

 A point process N = {x<sub>1</sub>, x<sub>2</sub>, ...} is stationary (or statistically homogeneous) if N and the translated point process

$$N_x = \{x_1 + x, x_2 + x, ...\}$$

have the same distribution for all x

For an inhomogeneous or non-stationary point process N the local intensity is a function of space, i.e., λ = λ(x)

#### Second-order statistics for homogeneous processes



(from: Illian, Penttinen, Stoyan & Stoyan, 2008)

#### Second-order spatial statistics: Pair density

- Definition: \(\rho(x\_1, x\_2)dx\_1dx\_2 = \probability\) of points located in each of two discs \(b\_1\) and \(b\_2\) with linear dimensions \(dx\_1\) and \(dx\_2\) at positions \(x\_1\) and \(x\_2\), respectively
- $\rho(x_1, x_2)$  is typically a function of the distance  $||x_1 x_2||$



#### Inhomogeneous pair correlation function

Inhomogeneous pair correlation function (PCF)

$$\hat{g}_{\textit{inhom}}(r) = \sum_{x_1,x_2 \in W}^{
eq} rac{1}{\hat{\lambda}(x_1)\hat{\lambda}(x_2)} rac{k(\|x_1-x_2\|-r)}{2\pi r A_{\|x_1-x_2\|}} \; ,$$

where k(.) is the *Epanechnikov* kernel, i.e.,

$$k(x) = \left\{ egin{array}{cc} rac{3}{4h}(1-rac{x^2}{h^2}), & ext{for } -h \leq x \leq h \ 0, & ext{otherwises} \end{array} 
ight.$$

and  $A_{\xi}$  denotes an edge correction at distance  $\xi = x - y$ 

Interpretation:

- For a random pattern without clustering:  $\hat{g}_{inhom}(r) \approx 1$
- If ĝ<sub>inhom</sub>(r) > 1, then pairs of fixations are more abundant than on average at a distance r
- If ĝ<sub>inhom</sub>(r) < 1, then pairs of fixations are less abundant than on average at a distance r
- Thus, the pair correlation function ĝ<sub>inhom</sub>(r) measures how selection of a particular fixation location is influenced by other fixations at distance r

#### Estimation of the inhomogeneous PCF

Numerical estimation involves two steps:

- First, we estimate the overall intensity 
   Â(x) for all fixation positions obtained for a given scene. In this procedure we borrow strength from the full set of observations to obtain reliable estimates of the inhomogeneity.
- Second, we compute the pair correlation function from single-trial data to measure correlations within the scanpath.
- Important: The behavior of the estimator should be analyzed over a range of bandwidths h.

#### Dependence of PCF deviation on bandwidth

• The deviation from g(r) = 1 is measured by

$$\Delta_g = \int_0^\infty (\hat{g}(r) - 1)^2 dr , \qquad (1)$$

Plot of the PCF deviation as a function of h for an inhomogeneous point process:



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#### Pair correlations for gaze patterns





#### Inhomogeneous null model



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#### Replication across datasets

#### Image set 1



Image set 2



#### Le Meur et al.





# Summary & Conclusions

- We can investigate second-order statistics for gaze patterns
- Pair correlation function can be estimated in two steps using the overall (inhomogeneous) intensity and the pair density for single trials
- Null model: inhomogeneous point process
- Experimental data for gaze pattern indicate clustering for length scales < 4° for our data</p>
- Scene content: no effects found
- Image size: strong effects

# Thank you for your attention!

#### **DFG** Deutsche Forschungsgemeinschaft

- Preprint: arxiv.org/abs/1405.3270
- R-Code: www.rpubs.com/Ralf
- More on spatial point processes & vision: Barthelmé, S. et al. (2013). Modeling fixation locations using spatial point processes. *Journal* of Vision (doi: 10.1167/13.12.1)