

Edge statistics in natural versus laboratory images

Implications for understanding lateral connectivity in primary visual cortex with respect to animal environments

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Society for Neuroscience 2011

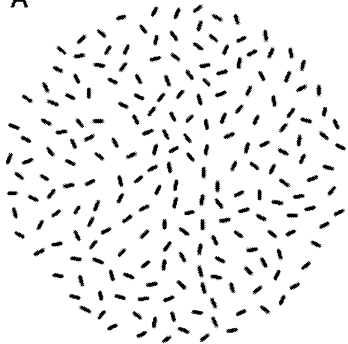
Nanosymposium on "Development of Motor and Sensory Systems"

Abstract Control Number: 17671, Presentation Number: 530.04.

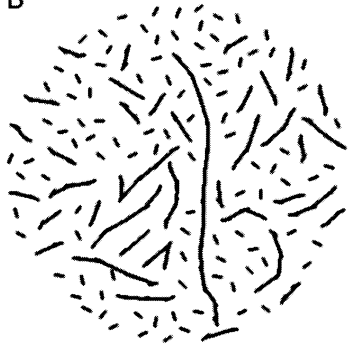
This work is supported by European Union project Number FP7-269921, "BrainScaleS".



A

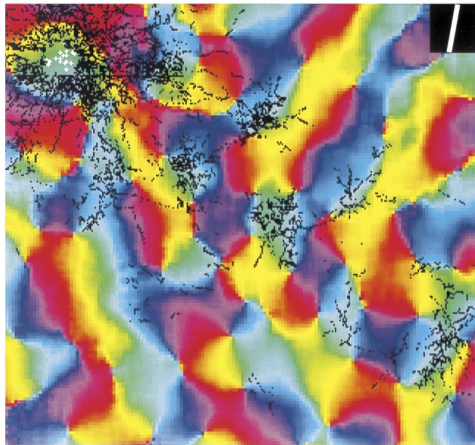


B

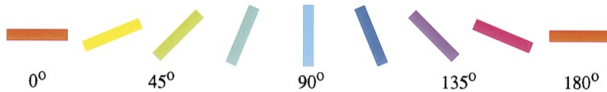
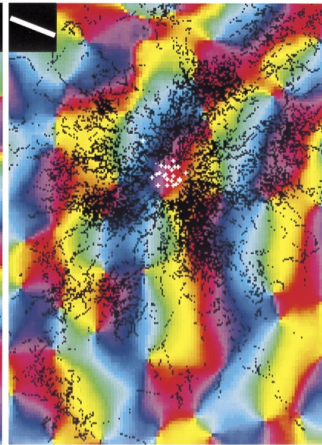


(Geisler et al., 2001, Vision Research)

A

500 μm

B



(Bosking et al, 1997, Journal of Neuroscience)





Outline: Edge statistics in natural versus laboratory images

Introduction: linking neural structure to natural scenes

Geisler et al, 2001

Bosking et al, 1997

Problem statement

Method: detection of edges

Geisler et al, 2001

Log Gabor representation / Sparse coding

Results: natural vs. laboratory images

Some examples of edge extraction

Second-order statistics

Quantitative difference using classification

Take-home message

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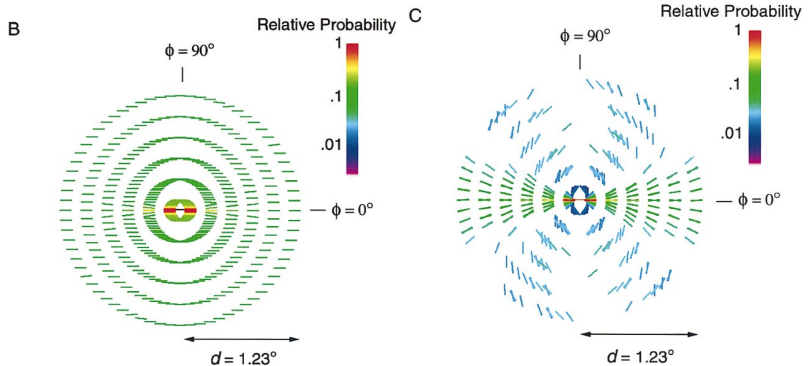
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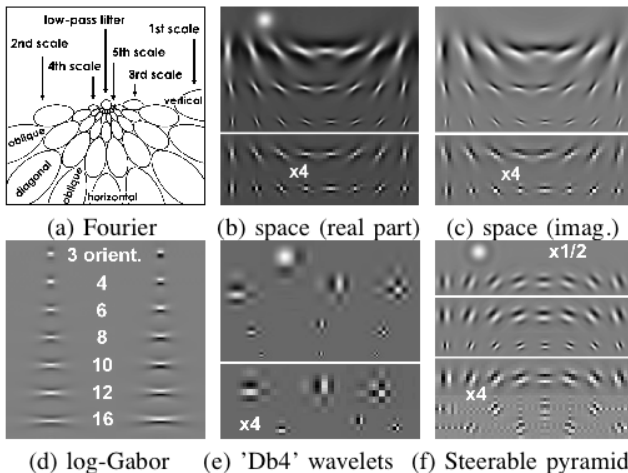


(Geisler et al., 2001, Vision Research)



(Geisler et al., 2001, Vision Research)

Log Gabor representation / Sparse coding



(Fischer et al, 2007, International Journal of Computer Vision)

(Perrinet, 2010, Neural Computation)

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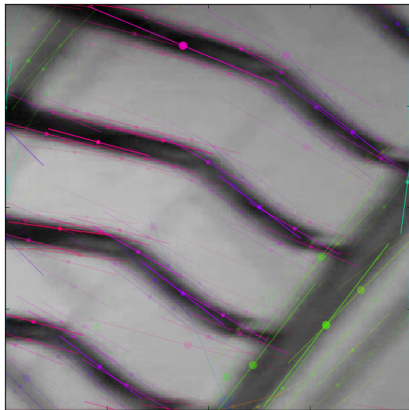
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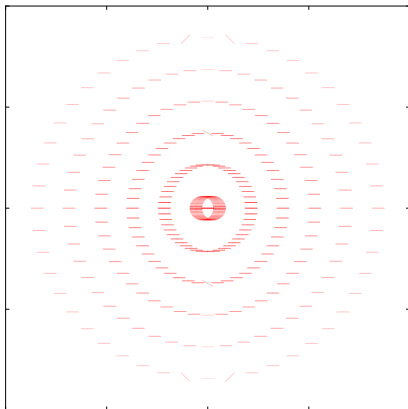
Natural



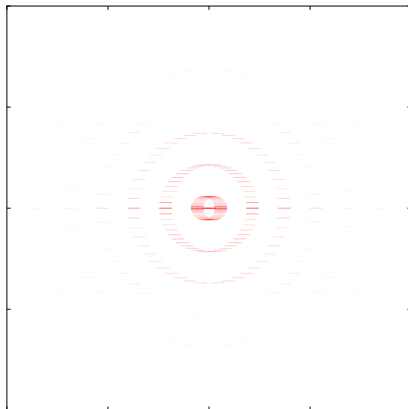
Laboratory

Second-order statistics

$$\arg \max_{\theta} p(d, \phi, \theta, \sigma | \pi_0)$$



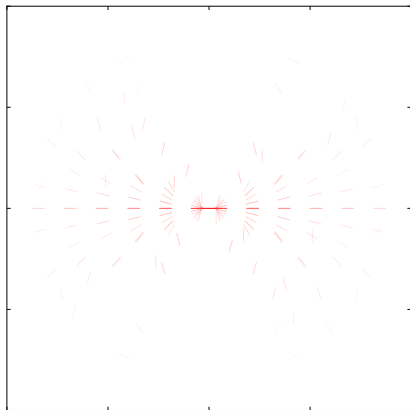
Natural



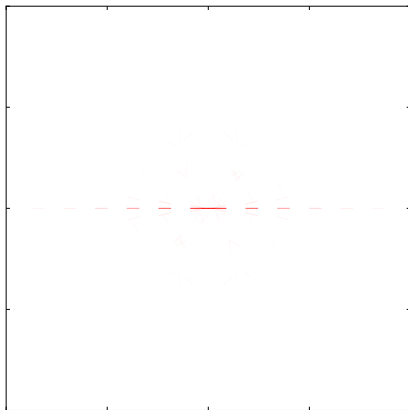
Laboratory

Second-order statistics

$$\arg \max_{\phi} p(d, \phi, \theta, \sigma | \pi_0)$$



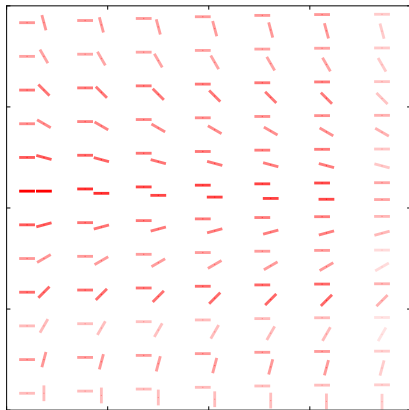
Natural



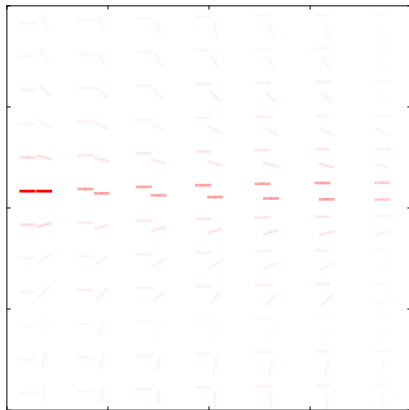
Laboratory

Second-order statistics

$$p(d, \phi, \theta, \sigma | \pi_0) \approx p(d, \sigma | \pi_0) p(\theta, \phi | \pi_0)$$

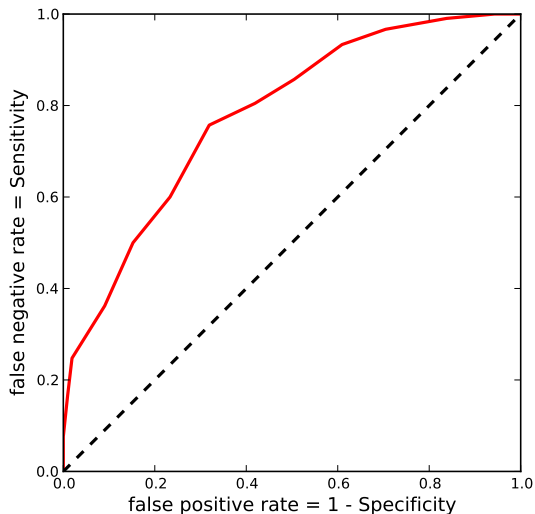


Natural



Laboratory

Quantitative difference using classification



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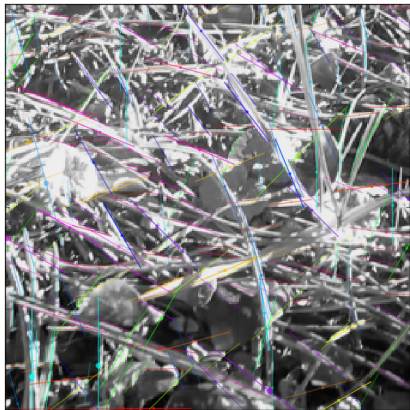
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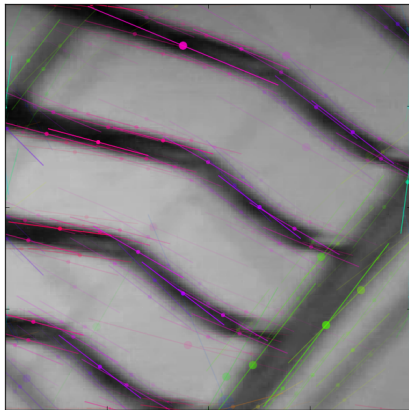
Quantitative difference using classification

Take-home message

Summary



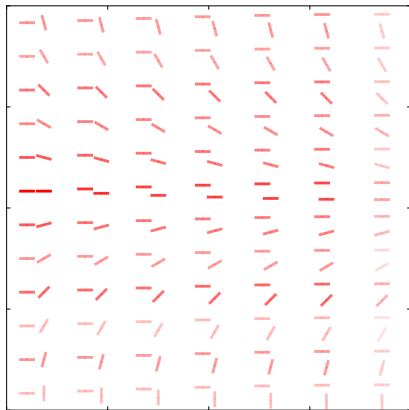
Natural



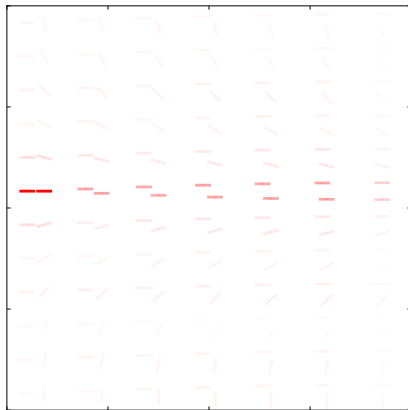
Laboratory

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$$p(d, \phi, \theta, \sigma | \pi_0) \approx p(d, \sigma | \pi_0) p(\theta, \phi | \pi_0)$$



Natural



Laboratory

References



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