

Bio - ML

TASKS RELATED TO BIOLOGICAL VISION

Ask Jesús!

(kick off meeting MINECO sept. 2021)

NEW THEORETICAL AND EXPERIMENTAL METHODS

1.3 Improving biological models through experiments driven by statistics

* Psychophysics - Noise

- MAXimum Differentiation (MAD)

* fMRI - information transfer and connectivity between voxel regions

EXTENSION OF CONVENTIONAL TECHNIQUES

2.1 Bio-inspired architectures. Divisive Normalization (DN) versus dynamical models (Wilson-Cowan/Amari) and Intrinsically Nonlinear Receptive Fields (INRF)

2.3 Noise in natural and artificial networks. Estimated noise and Fisher information.

RELATIONS BETWEEN BRAIN AND STATISTICS

3.2 Similarities and differences between artificial and natural NN

* Contrast Sensitivity Functions

* Visual Illusions

* Distortion metrics

* Nets with natural noise induce human behavior?

3.3 Information theory in the visual system

* Information flow

* Brain connectivity

* Information-based distortion metrics

1.3 New experiments to improve biological vision models

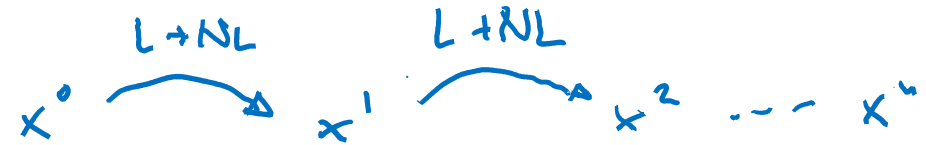
- Bio vision models

- Psychophysics

- fMRI

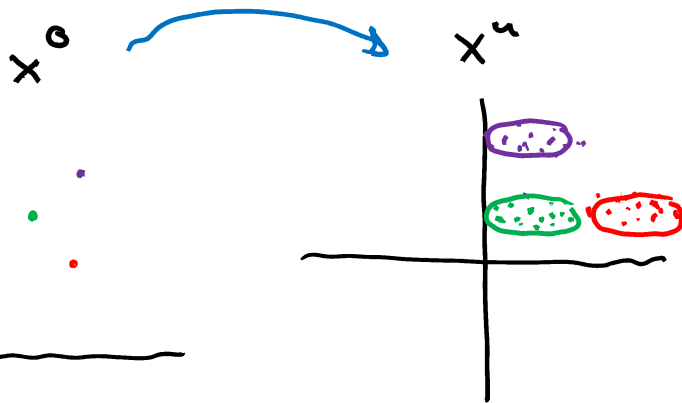
1.3 New experiments to improve biological vision models

- Bio vision models



$L \equiv$ Chromatic opponency / center-surround / wavelets } convolutional
 $NL \equiv$ Saturation (activation) } invariance via pooling
 Adaptive (similar to batch normalization) } invariance to contrast texture & illumination

Long tradition in fitting the parameters of $L+NL$



IEEE TIP OCa
IEEE TIP OCb

JOSA A 10

JOSA A 17

PLOS 18

Frontiers 19

Scient. Rep. 20

J. Neurophysiol. 20

J. Math. Neurosci. 20

<https://ieeexplore.ieee.org/document/1556625>

<https://ieeexplore.ieee.org/document/1556637>

<https://www.osapublishing.org/josaa/abstract.cfm?uri=josaa-27-4-852>

<https://www.osapublishing.org/josaa/abstract.cfm?uri=josaa-34-9-1511>

<https://journals.plos.org/plosone/article?id=10.1371/journal.pone.0201326>

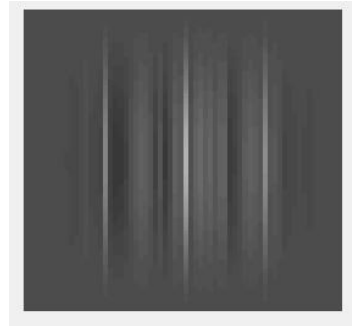
<https://www.frontiersin.org/articles/10.3389/fnins.2019.00008/full>

<https://www.nature.com/articles/s41598-020-73113-0>

<https://journals.physiology.org/doi/full/10.1152/jn.00487.2019>

<https://mathematical-neuroscience.springeropen.com/articles/10.1186/s13408-020-00095-8>

1.3 New experiments to improve biological vision models



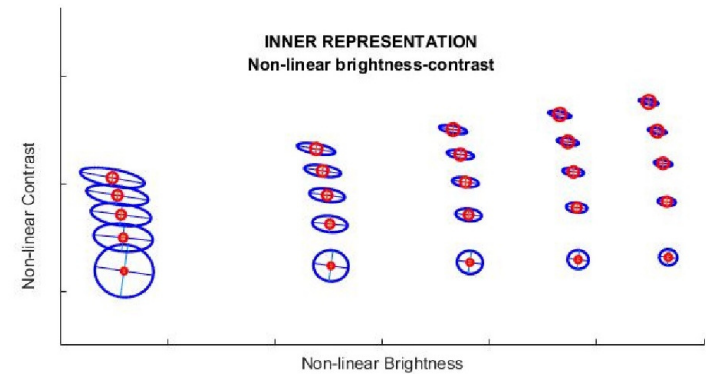
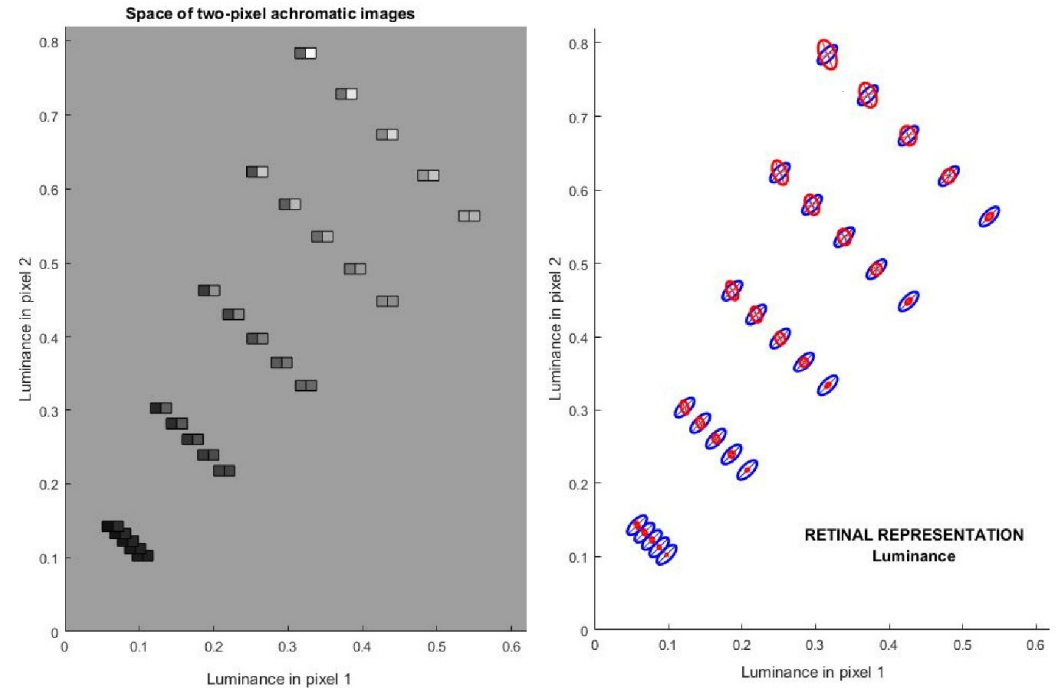
Noise

- Psychophysics

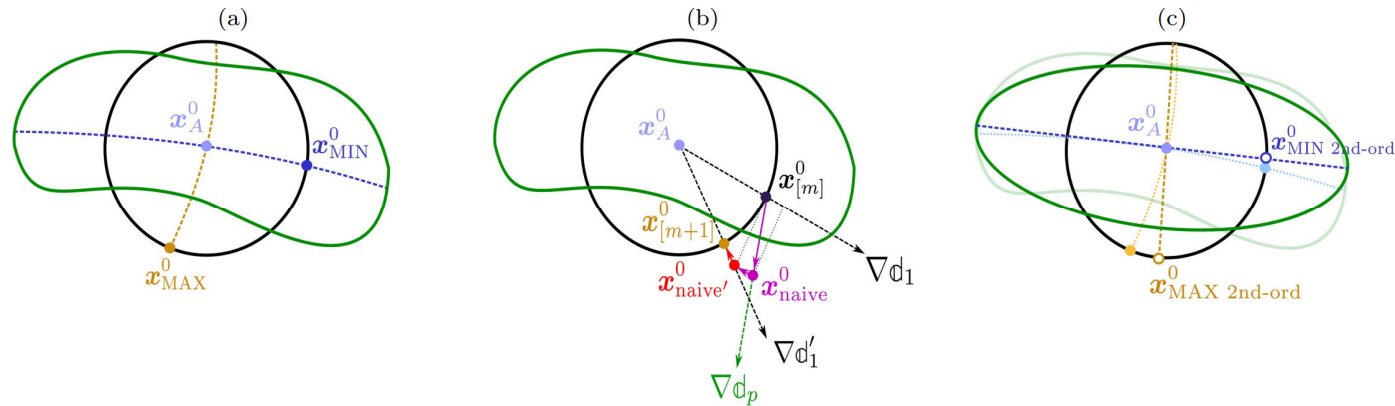
<https://arxiv.org/abs/2012.06608>

We are starting to
measure the noise!

(not only the parameters
of the transform $L + NC$)

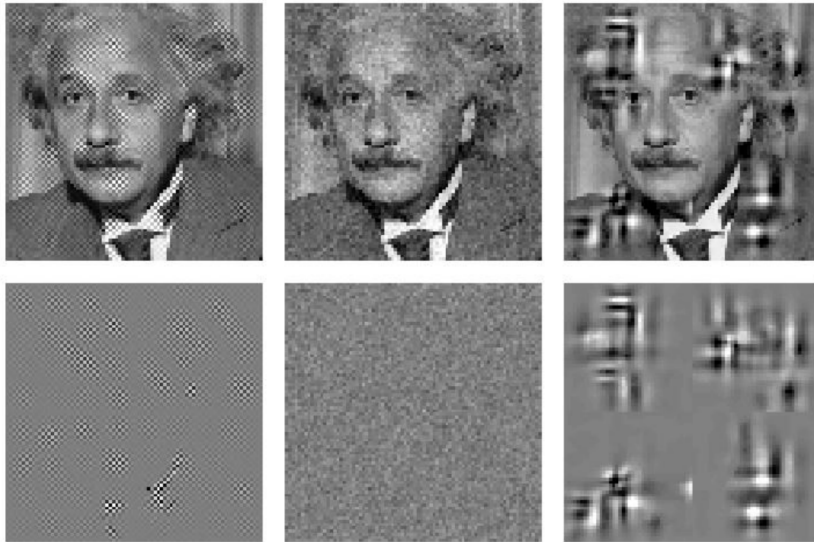


1.3 New experiments to improve biological vision models



- Psychophysics

Maximum Differentiation (MAD)



<https://sci-hub.st/https://doi.org/10.1117/12.2085653>

<https://journals.plos.org/plosone/article?id=10.1371/journal.pone.0201326>

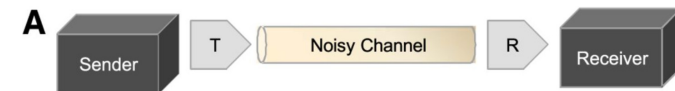
MAD can be used to determine the parameters better!

1.3 New experiments to improve biological vision models

Attention Improves Transfer of Motion Information between V1 and MT

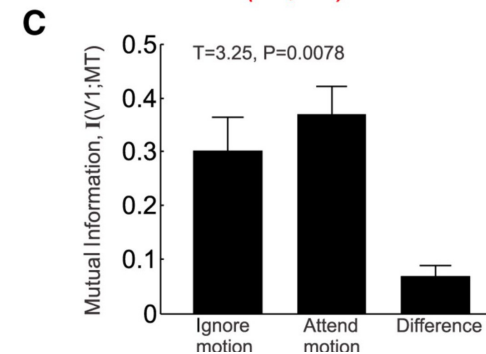
Sameer Sapru and John T. Serences

Journal of Neuroscience 5 March 2014, 34 (10) 3586-3596; DOI: <https://doi.org/10.1523/JNEUROSCI.3484-13.2014>



T: Transmitted message
R: Received message

B Probability of decoding error: $P \propto e^{-I(T,R)}$
where, $I(.)$ is the mutual information



- fMRI

Inform flow
&
Connectivity
from REAL DATA!

2.1 Bio inspired architectures: models of the non-linearity NL



- Divisive normalization

$$x = \frac{e}{b + H \cdot e}$$

$$x_i = \frac{e_i}{b_i + H_{ij} e_j}$$

- Dynamical models
(recurrence)

$$\dot{x}_i = \alpha_i e_i - W_{ij} f(x_j)$$

- Intrinsically Nonlinear Receptive fields

$$x_i = M_{ij} e_j - W_{ij} f(e_j - G_{jk} e_k)$$

J. Neurophysiol. 20

ArXiv 19

Sci. Rep. 20

<https://journals.physiology.org/doi/full/10.1152/jn.00487.2019>

<https://arxiv.org/abs/1906.08246>

<https://www.nature.com/articles/s41598-020-73113-0>

Ready to
apply!

2.3 Noise in artificial and natural networks

- NeurIPS 17 <https://arxiv.org/abs/1710.02266>
 - Covariance back in the input space
 - Fisher information
 - MAD in artificial networks
- J. Math. Neurosci. 20 <https://mathematical-neuroscience.springeropen.com/articles/10.1186/s13408-020-00095-8>
- Entropy Conference 21 <https://arxiv.org/abs/1912.12093>
- VSS - J. Vision 21 <https://jov.arvojournals.org/article.aspx?articleid=2777344>

ALL USED
APPROXIMATED
ESTIMATES OF
NOISE

Now we can
do it BETTER
with real
noise estimator

3.2 Similarities and differences between artificial / biological networks

HOT TOPIC and many ongoing works from us

* Contrast Sensitivity Functions ArXiv 21 <https://arxiv.org/abs/2103.00481>

- Tasks (restoration / color constancy) reconstruction
- statistics
- Bottleneck

* Visual illusions Vision Research 20 <https://doi.org/10.1016/j.visres.2020.07.010>
ArXiv 21 <https://arxiv.org/abs/1911.09599>

* Distortion metric

IEEE ICIP 20

ArXiv 21

Metric \rightarrow Masking?

Stats \rightarrow Metric

<https://ieeexplore.ieee.org/document/9190691>

<https://arxiv.org/abs/2106.04427>

* Networks with natural noise (follow up at NeurIPS 17)

3.3 Information theory in human vision

- Information flow

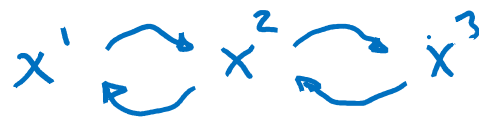
$$x \xrightarrow{\quad} y = S(x) + n$$

<https://mathematical-neuroscience.springeropen.com/articles/10.1186/s13408-020-00095-8>

$$I(x, y) = \sum_i h(y_i) - T(y) - h(n)$$

$$I(x, y) = h(x) + E_x \left[\log |\nabla_x S| \right] - h(n) + E_x \left[D_{KL}(S+n|S) \right]$$

- Connectivity



$I(x^i, x^j)$ is widely used

$T(x^1, \dots, x^M)$ is better (ANALYT + DFMU!)

And we have tools to estimate T (RBIG)

- Extension of Visual Information Fidelity VIF: $\frac{I(x, y_1)}{I(x, y_0)}$ We are doing it Better!

<https://jov.arvojournals.org/article.aspx?articleid=2777344>