

Visual search as active inference

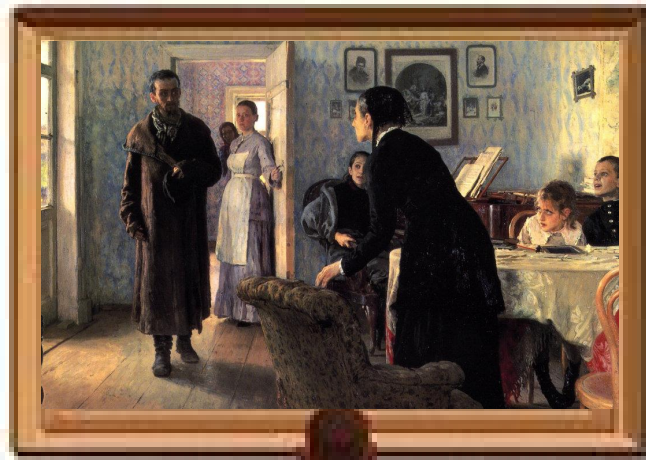
Emmanuel Dauce & Laurent Perrinet



1st International Workshop on Active Inference (IWAi*2020), 14/9/2020

Visual search as active inference

SCENE



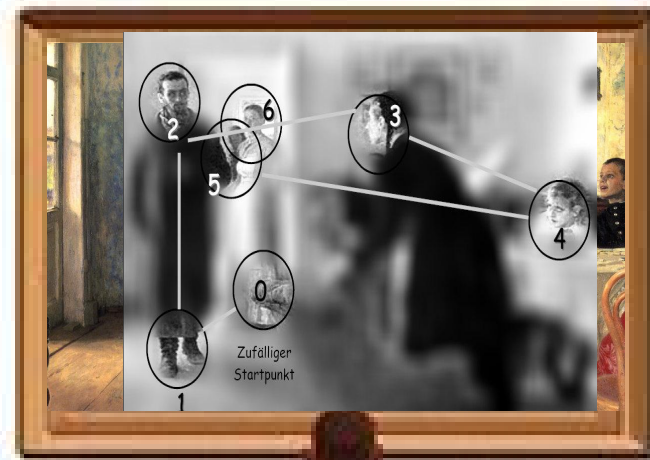
OBSERVER



partial.com

Visual search as active inference

SCENE



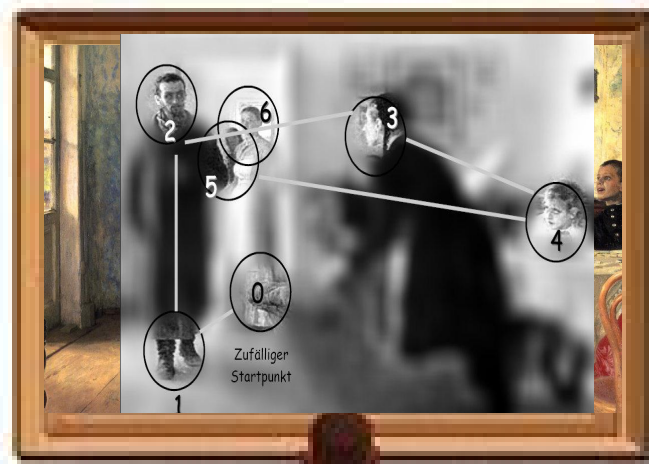
OBSERVER



partial forms

Visual search as active inference

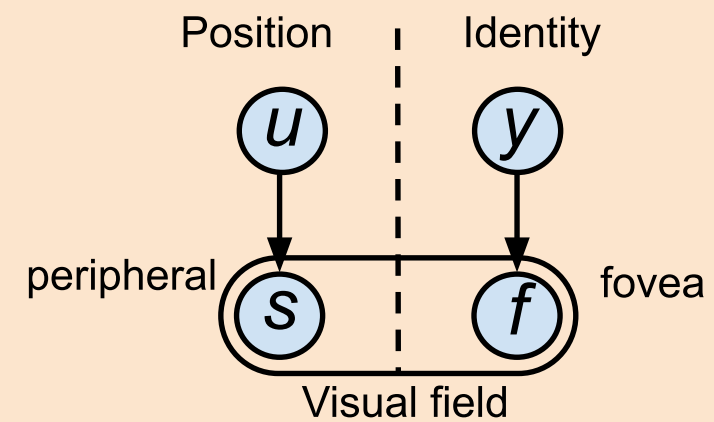
SCENE



OBSERVER

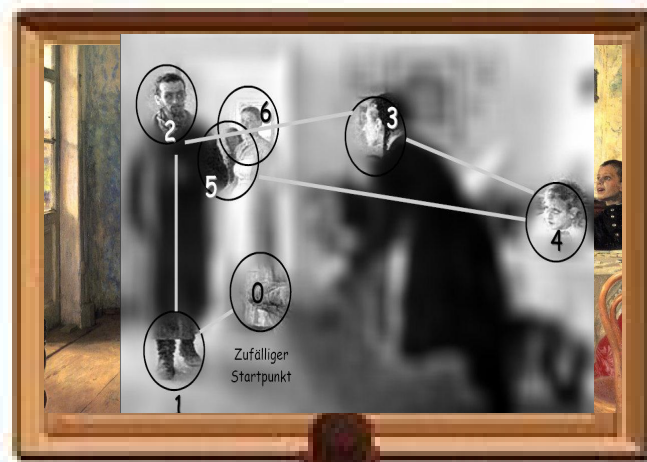


Generative model



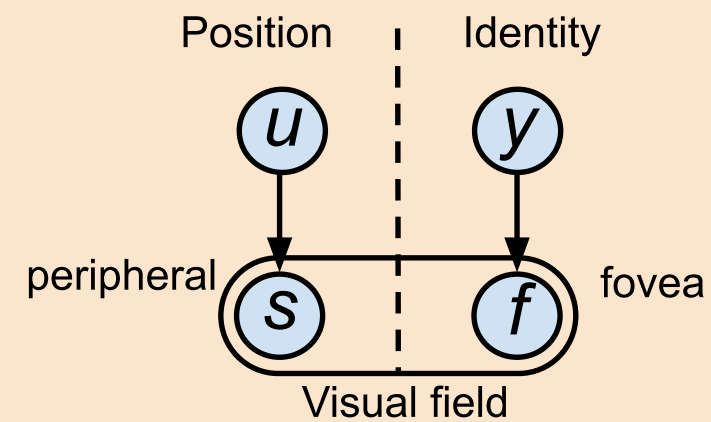
Visual search as active inference

SCENE



OBSERVER

Generative model



Information gain

$$\max_a E_y \log P(y|f(a)) - \log P(y|f_0)$$

Peripheral
processing

PREDICTED

Central
processing

CURRENT

Outline

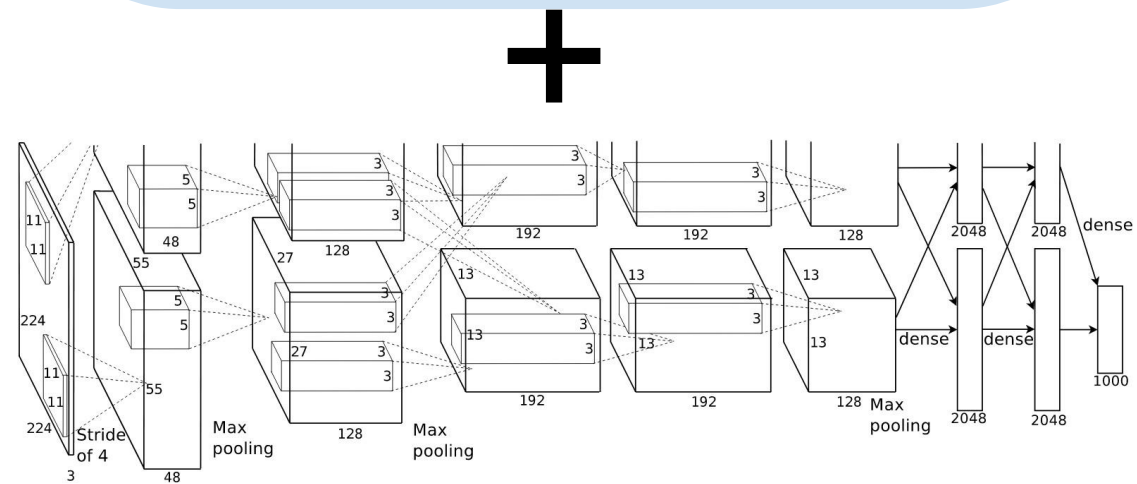
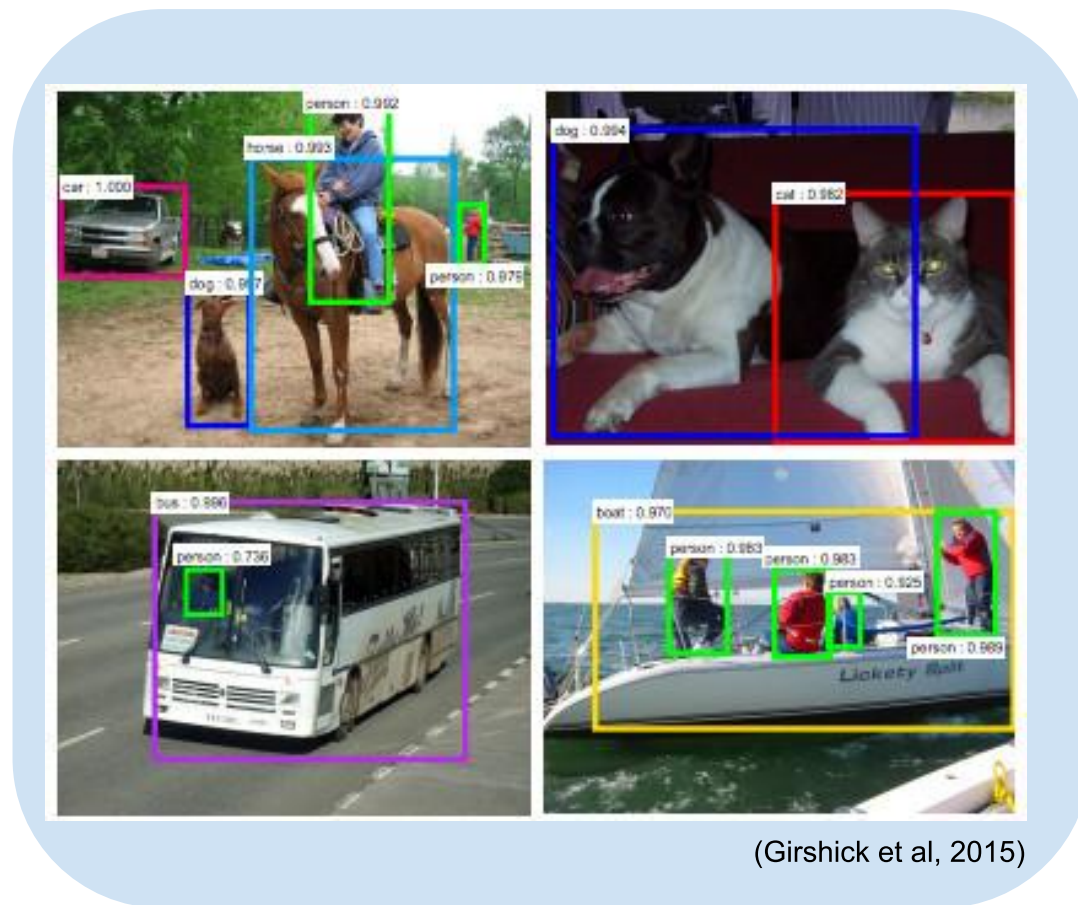
1. Motivation
2. Methods
3. Results
4. Conclusion

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Computer vision

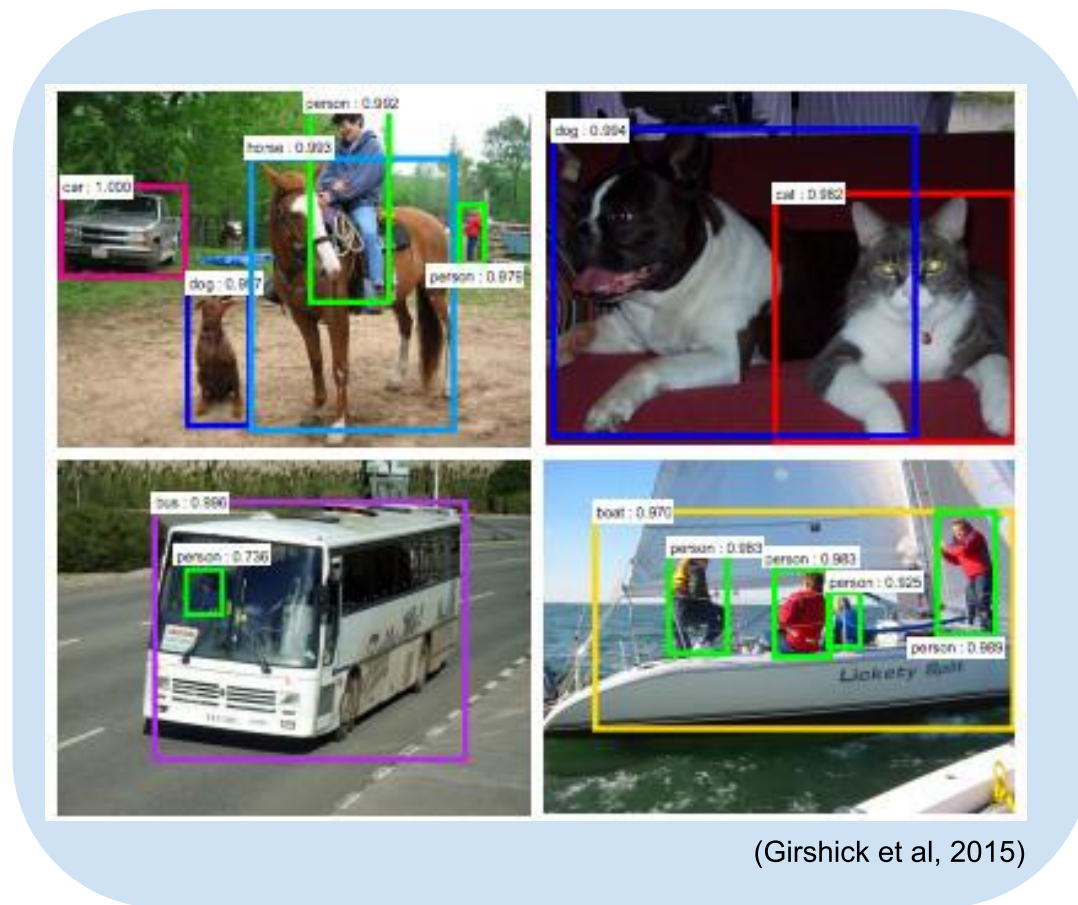
COMPUTER VISION



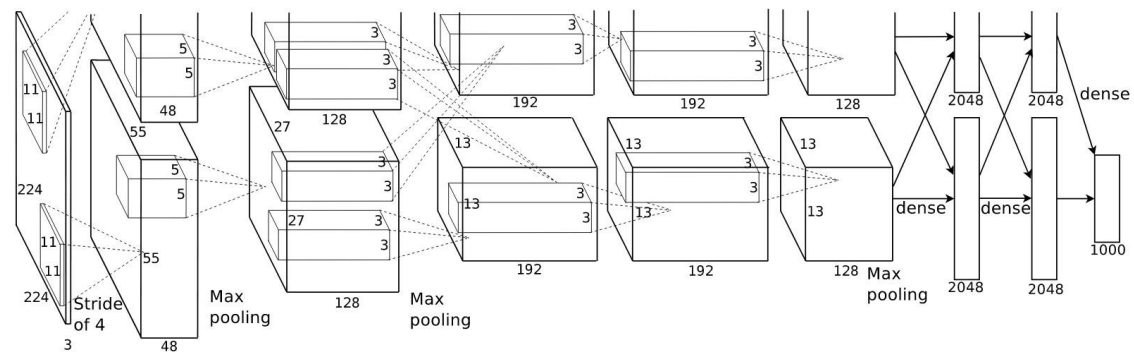
(Hinton et al, 2012)

Human vision

COMPUTER VISION

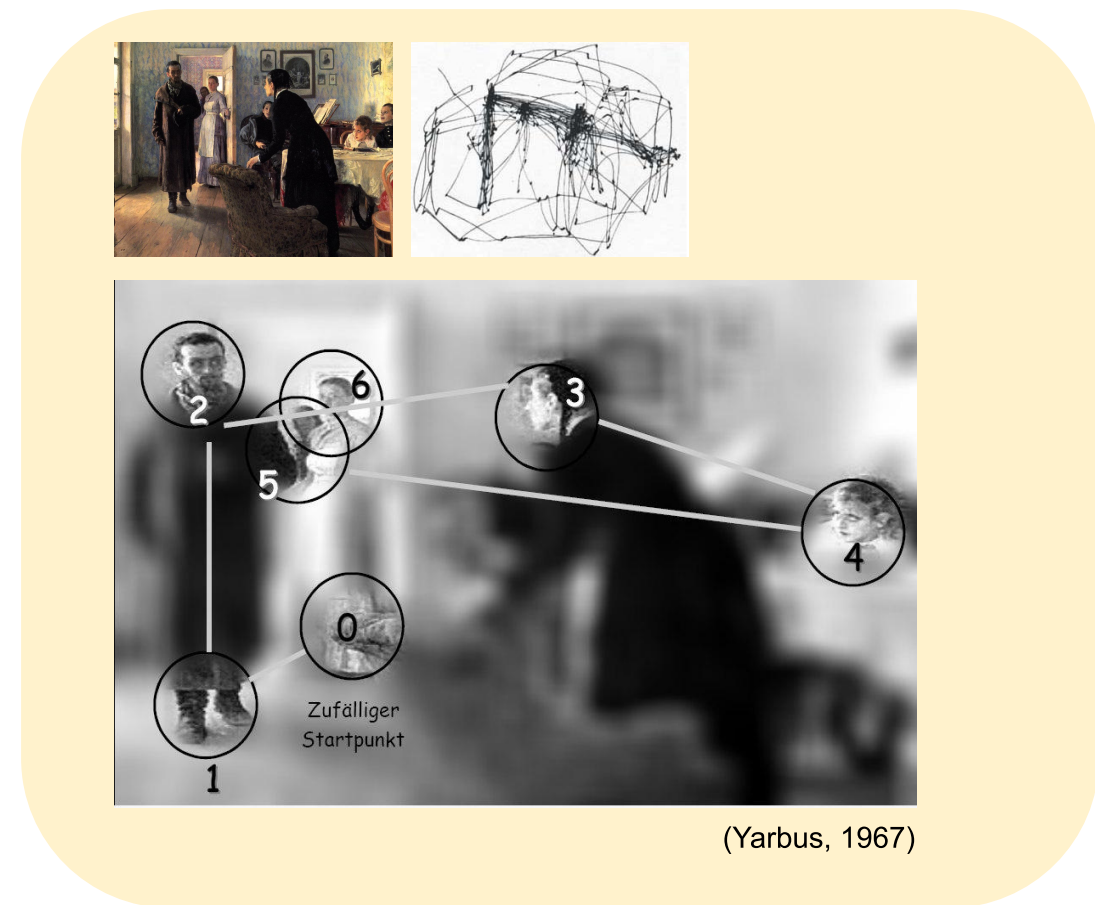


+

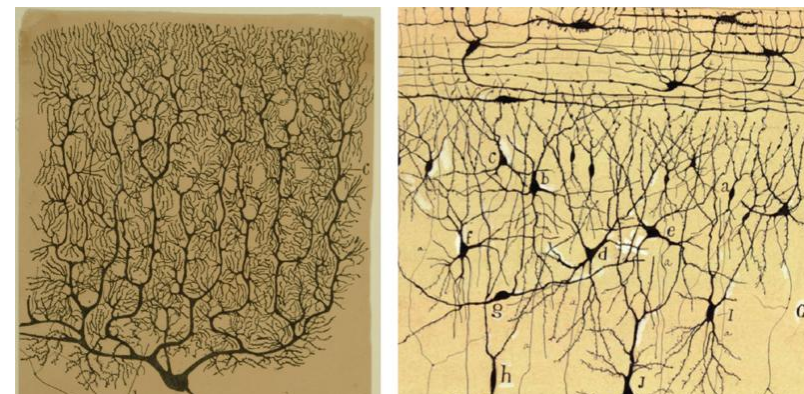


(Hinton et al, 2012)

HUMAN VISION



+



(Cajal, 1888)

Attention vs. Scene Understanding

Bayesian surprise
(Information Gain)

(Itti & Baldi, 2009)

$$E_y [\log P(Y|x, u) - \log P(Y)]$$

BOTTOM - UP

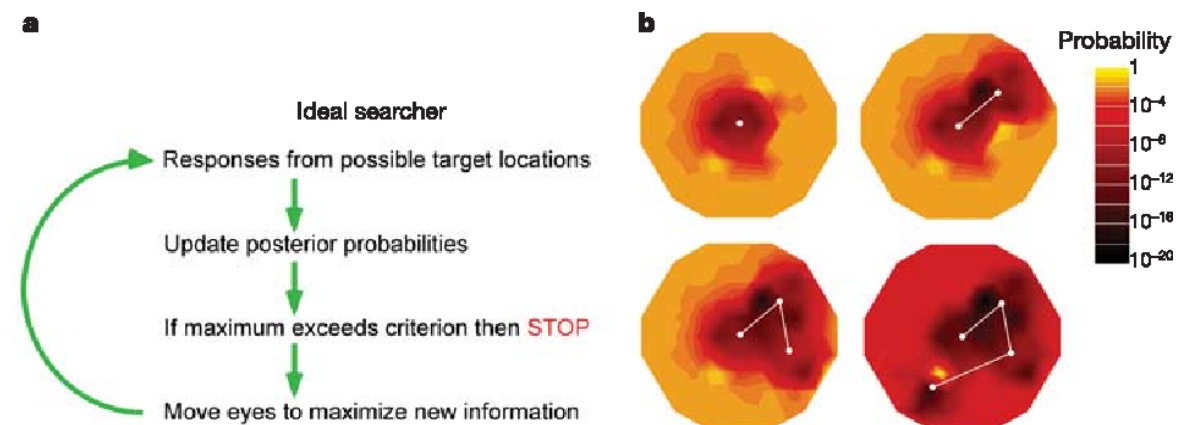
Visual attention
Saliency Maps



Itti and Koch (2000)
Kümmerer et al (2015)

TOP - DOWN

Active Inference
Recurrent Attention



Najemnik, J., & Geisler, W. S. (2005)
Butko & Movellan (2010)
Fu et al (2017)

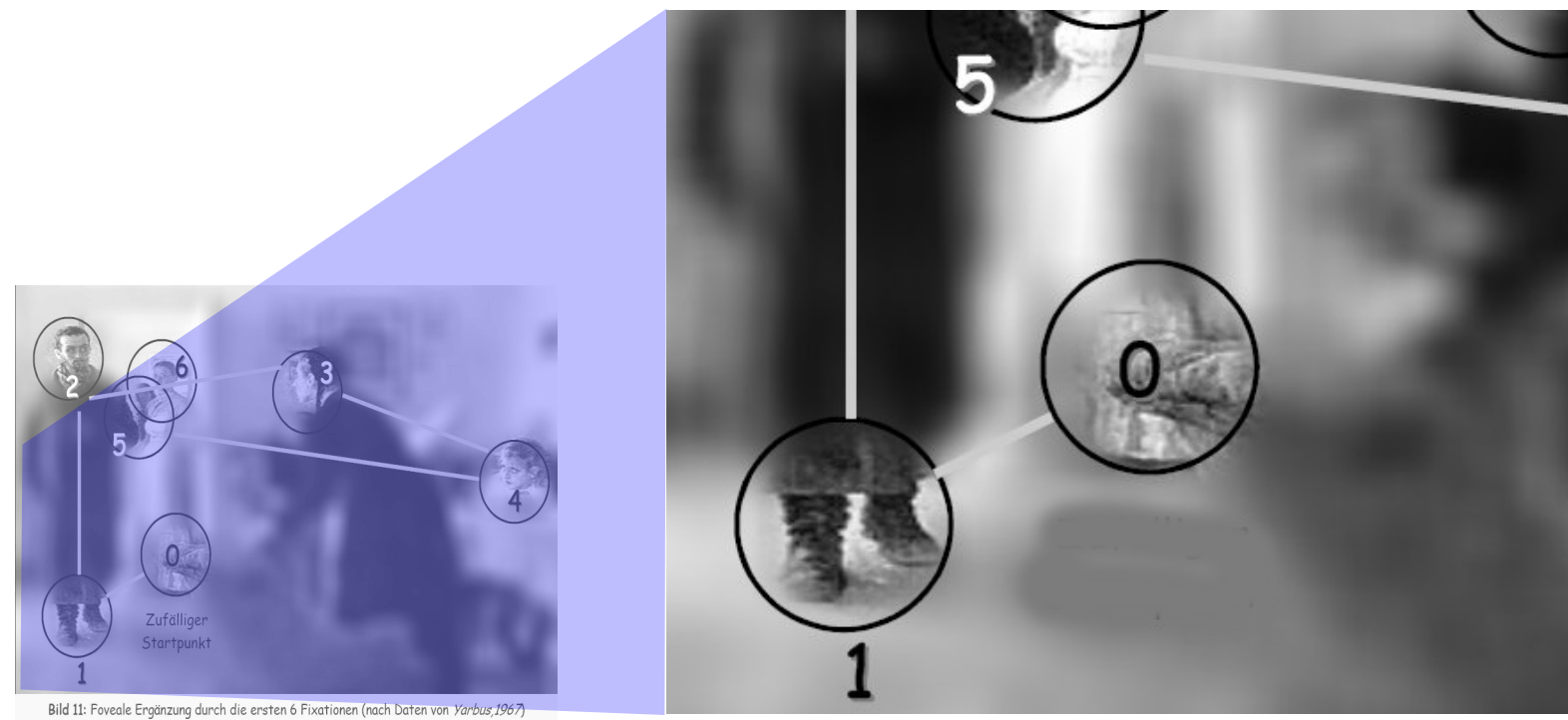
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Principles for central and peripheric vision

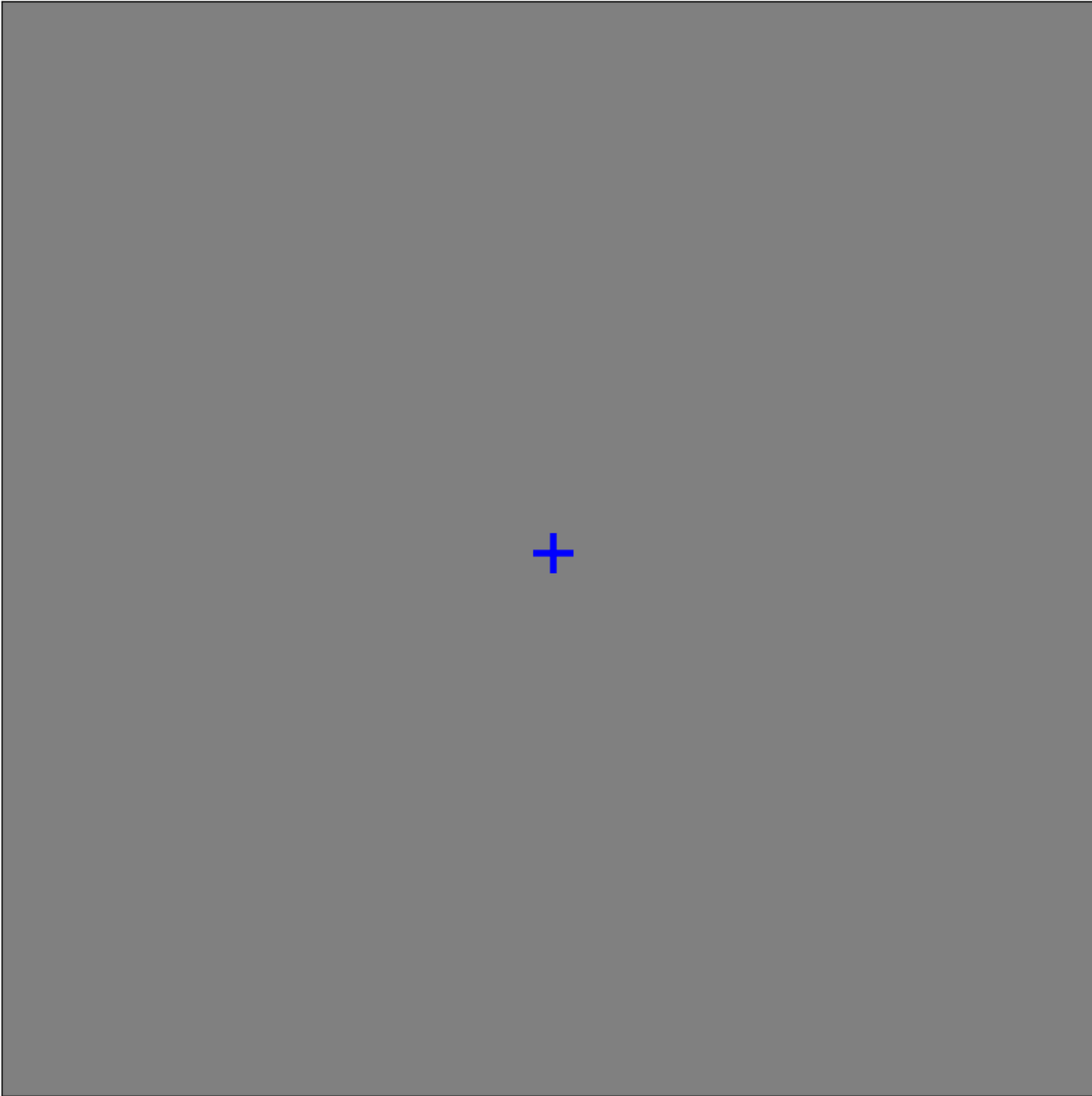


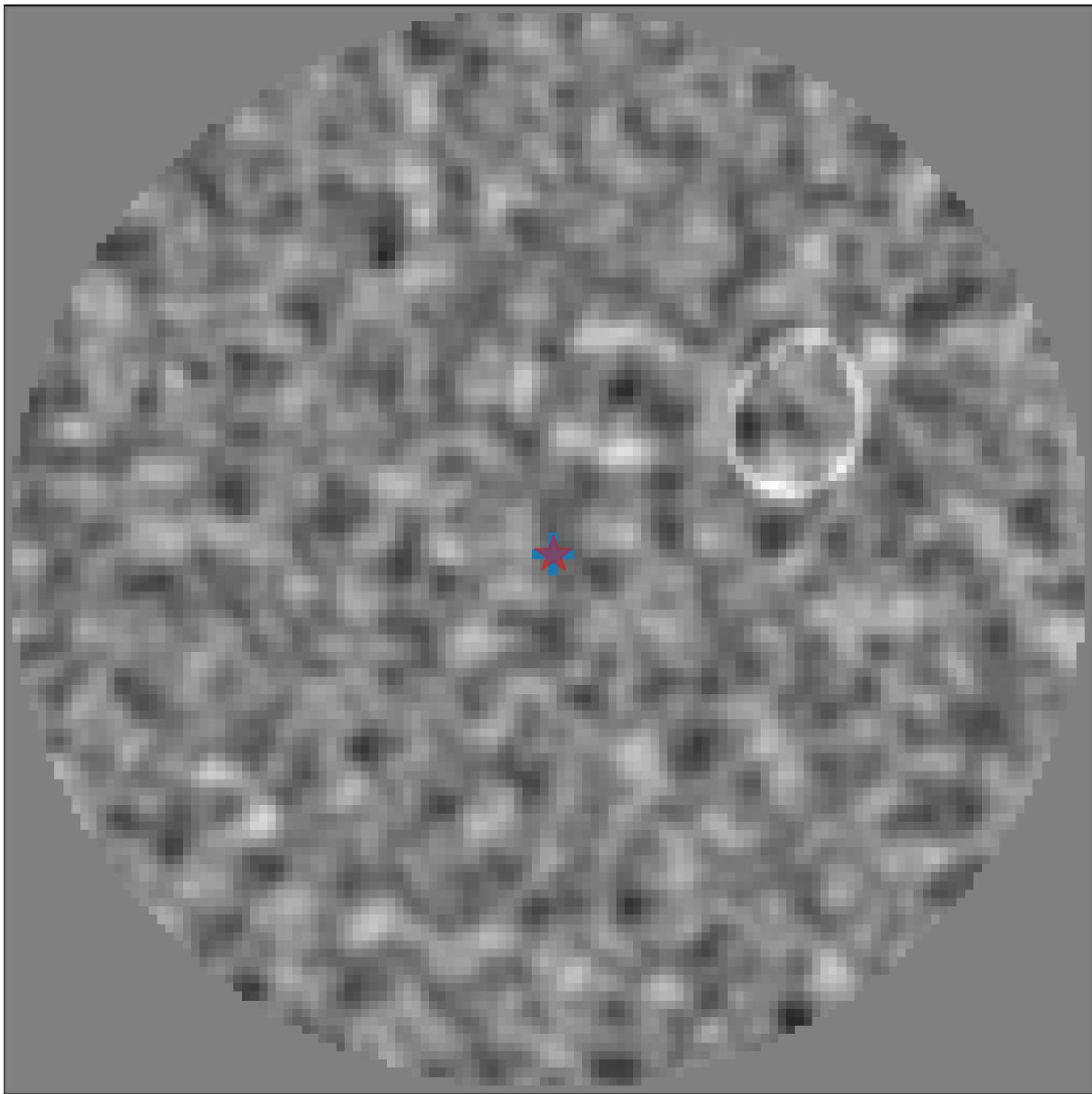
Information Gain :

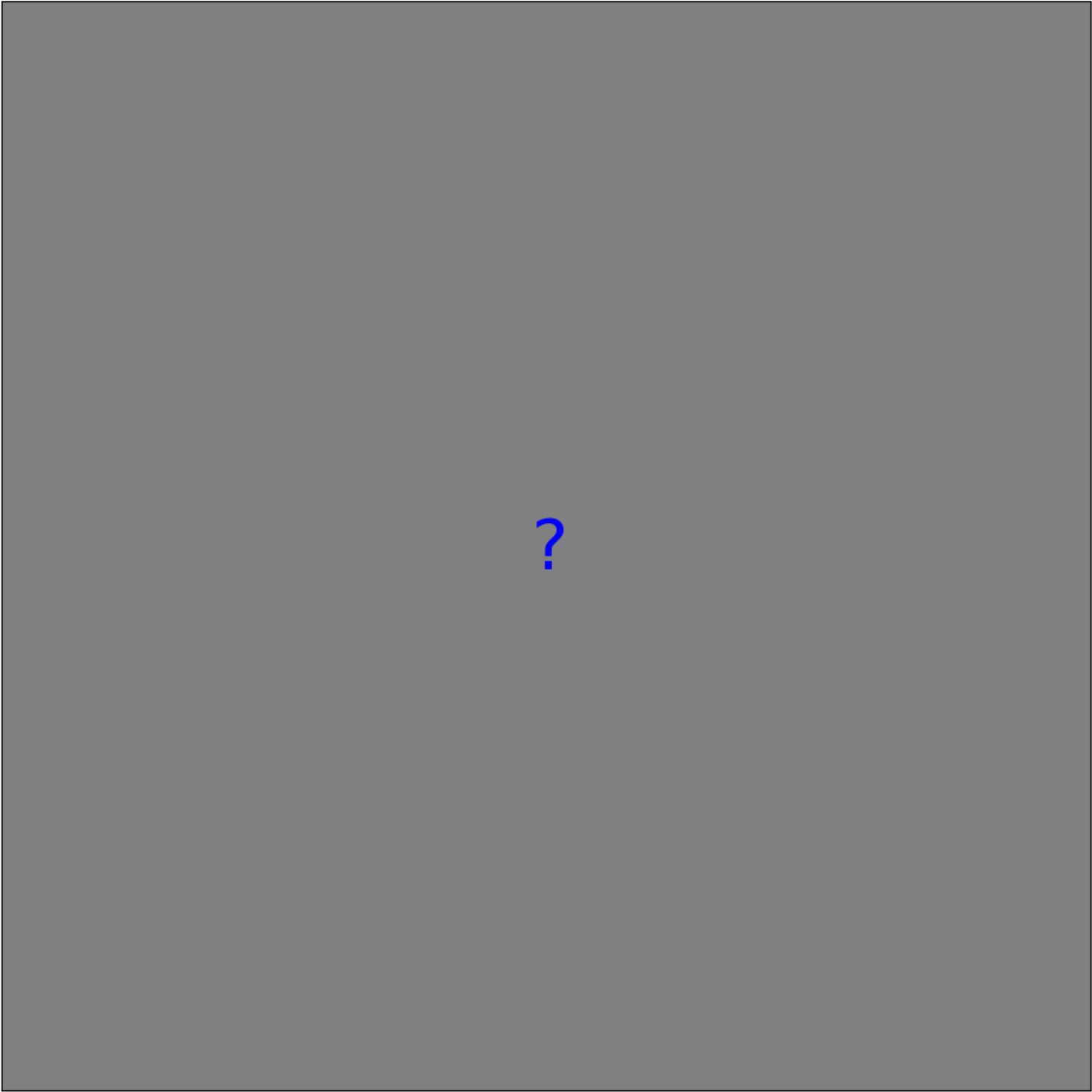
$$E_y [\log P(Y|x, u) - \log P(Y)]$$

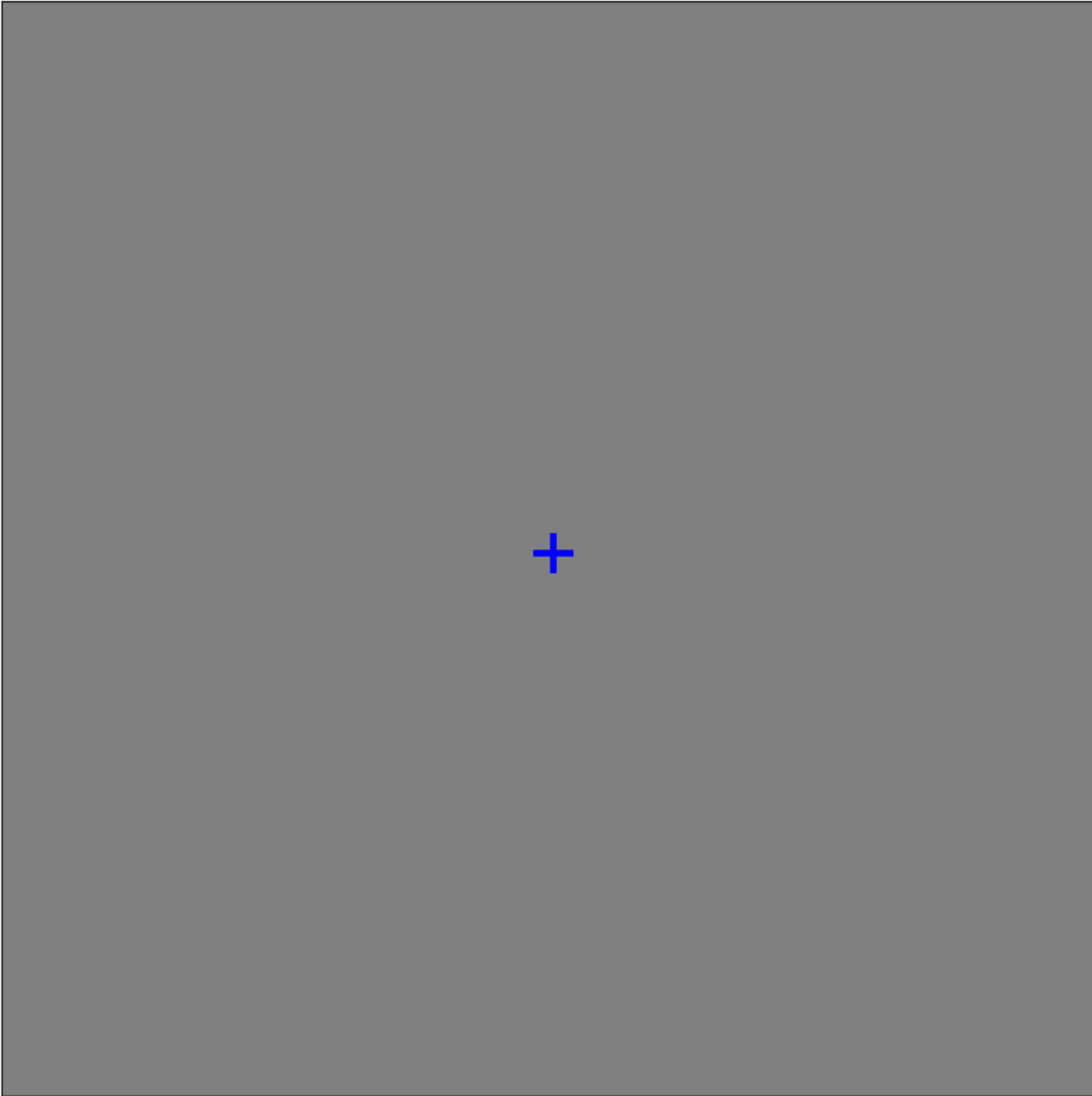
Peripheral
processing

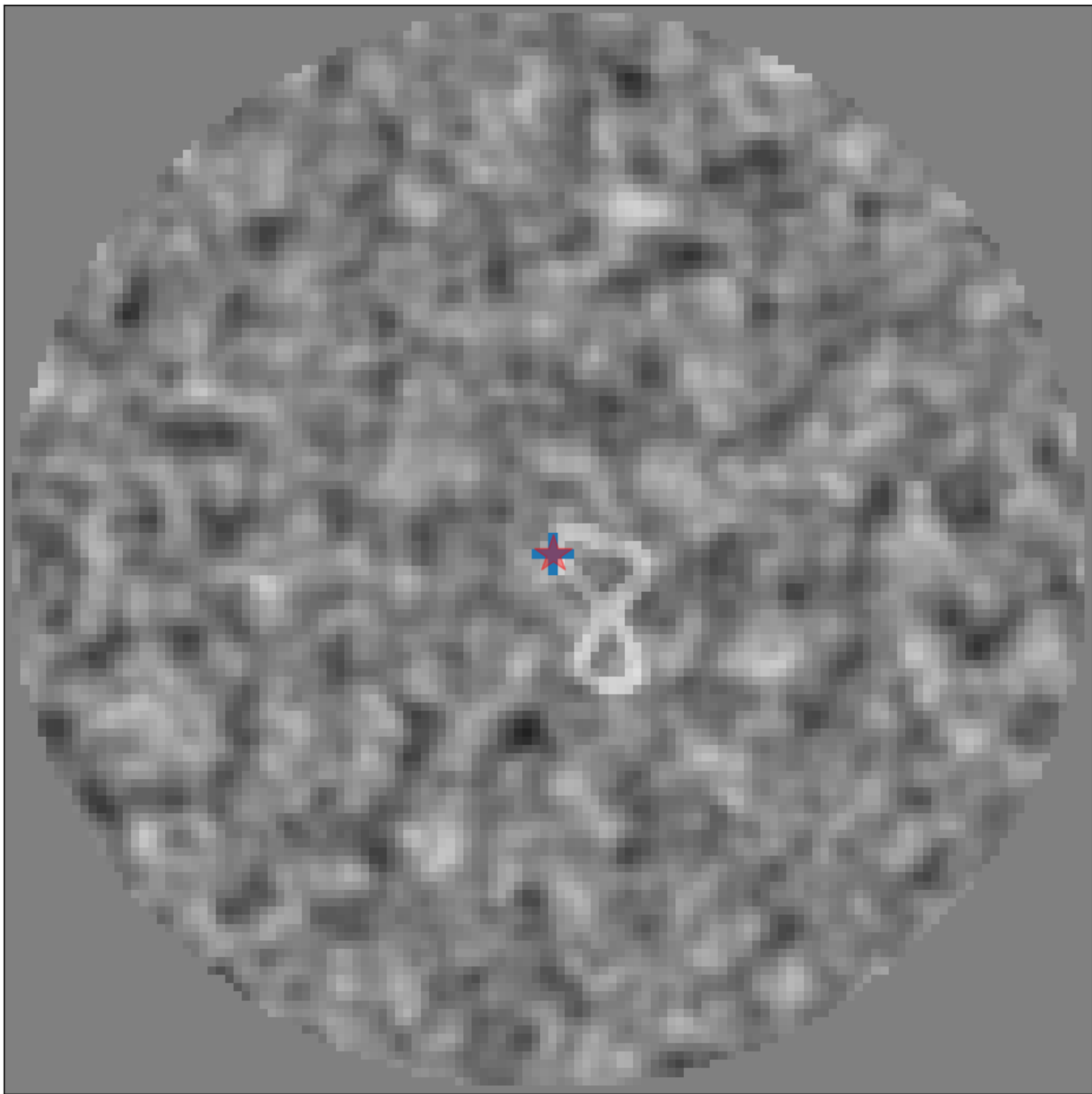
Central processing

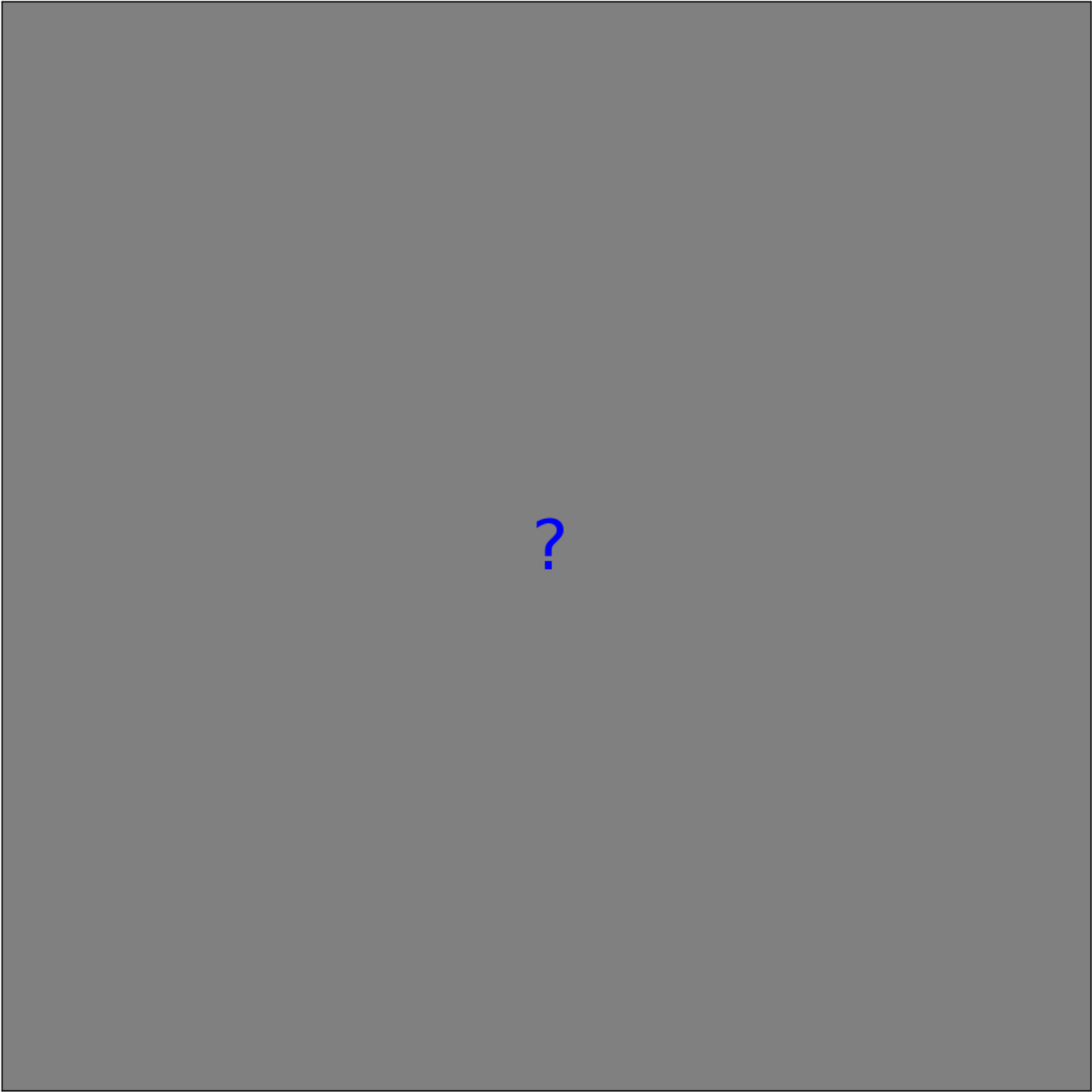


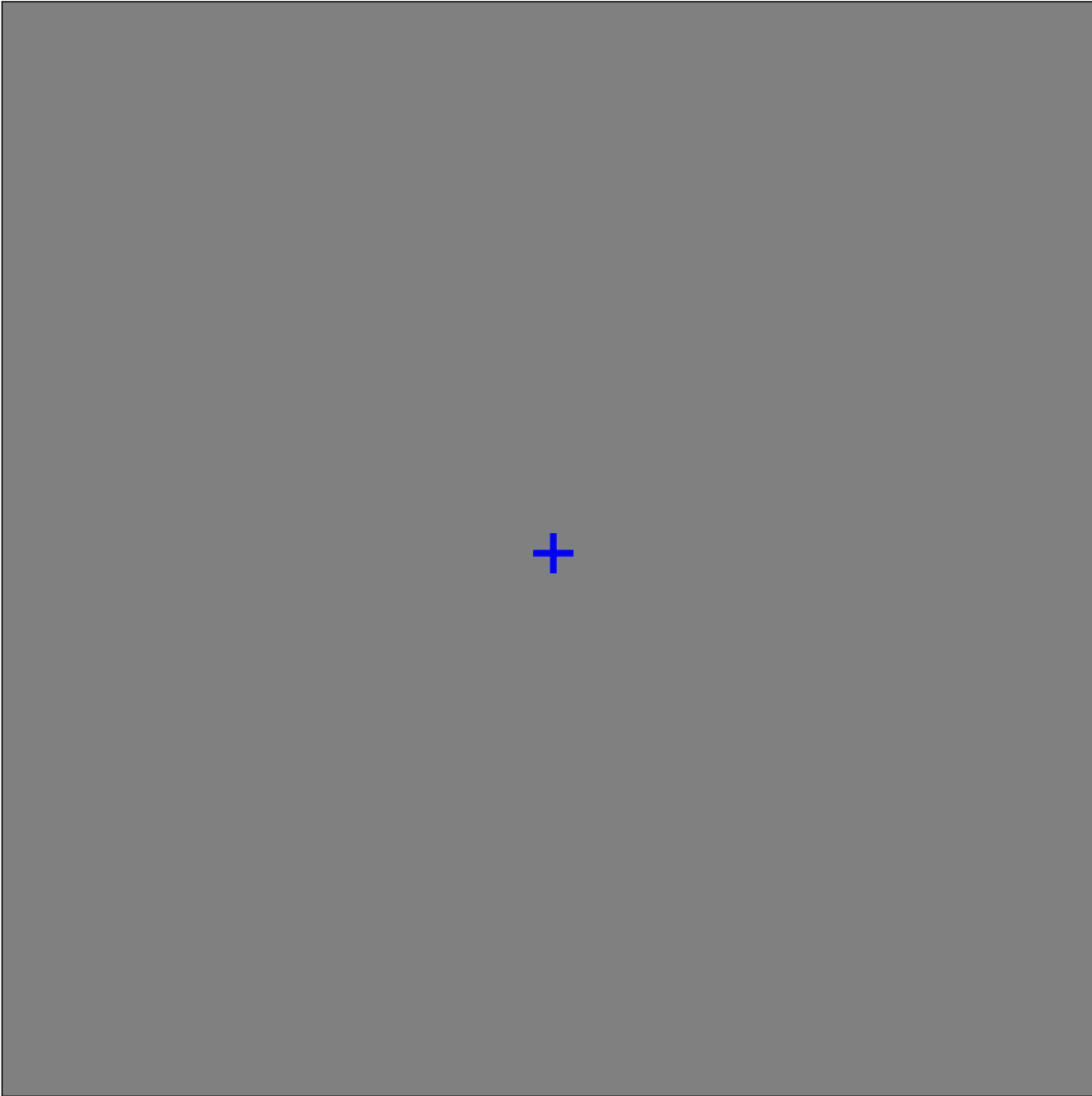


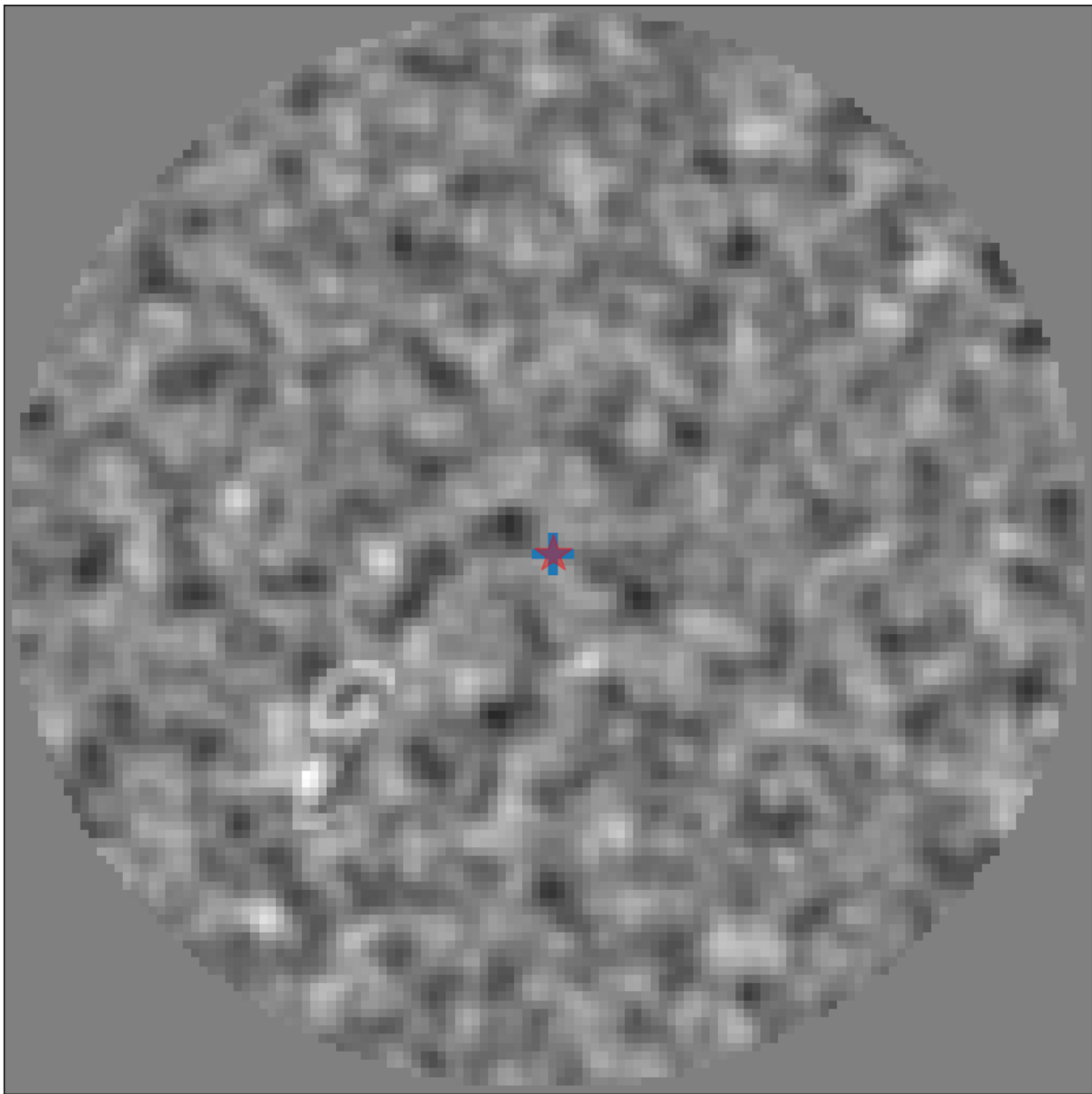


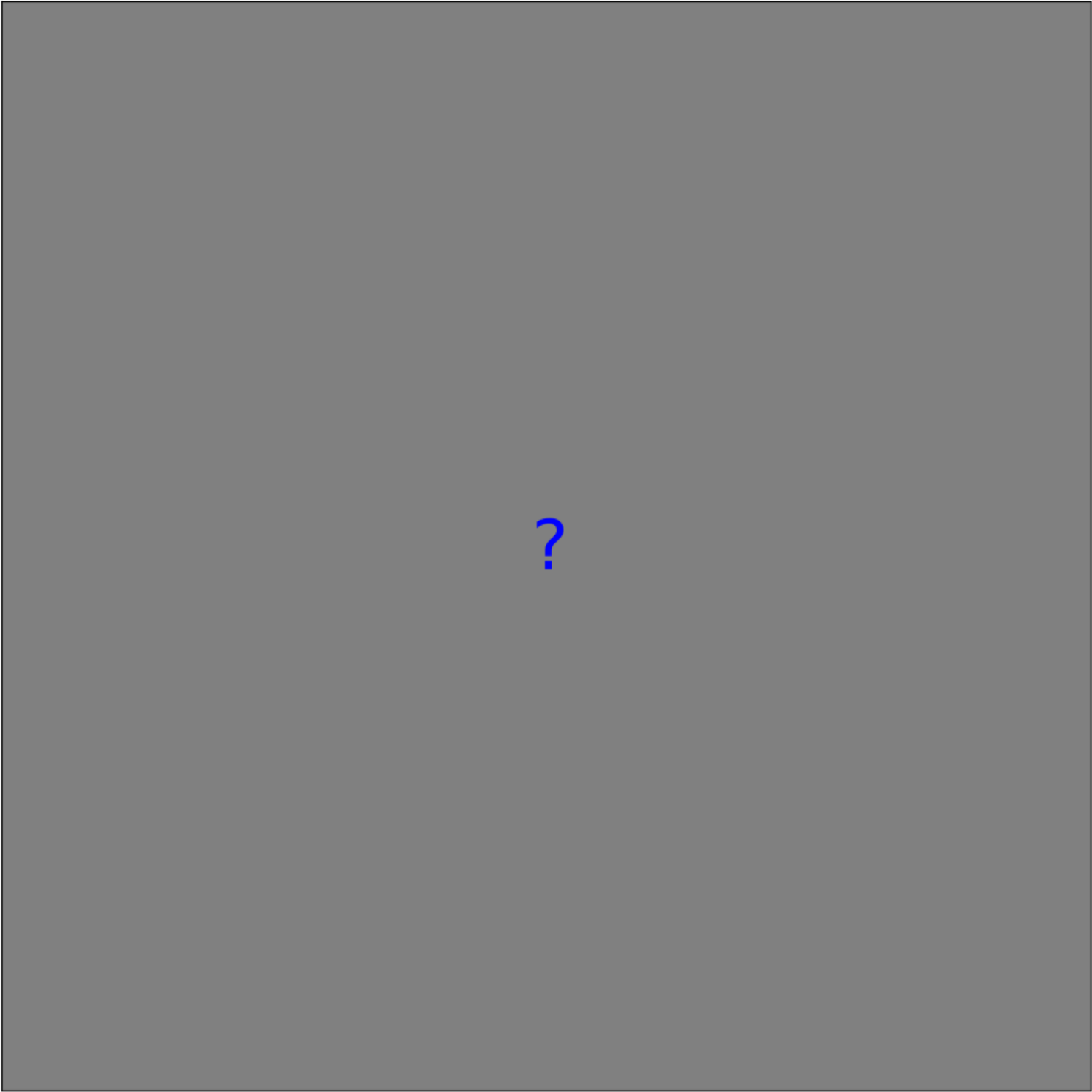


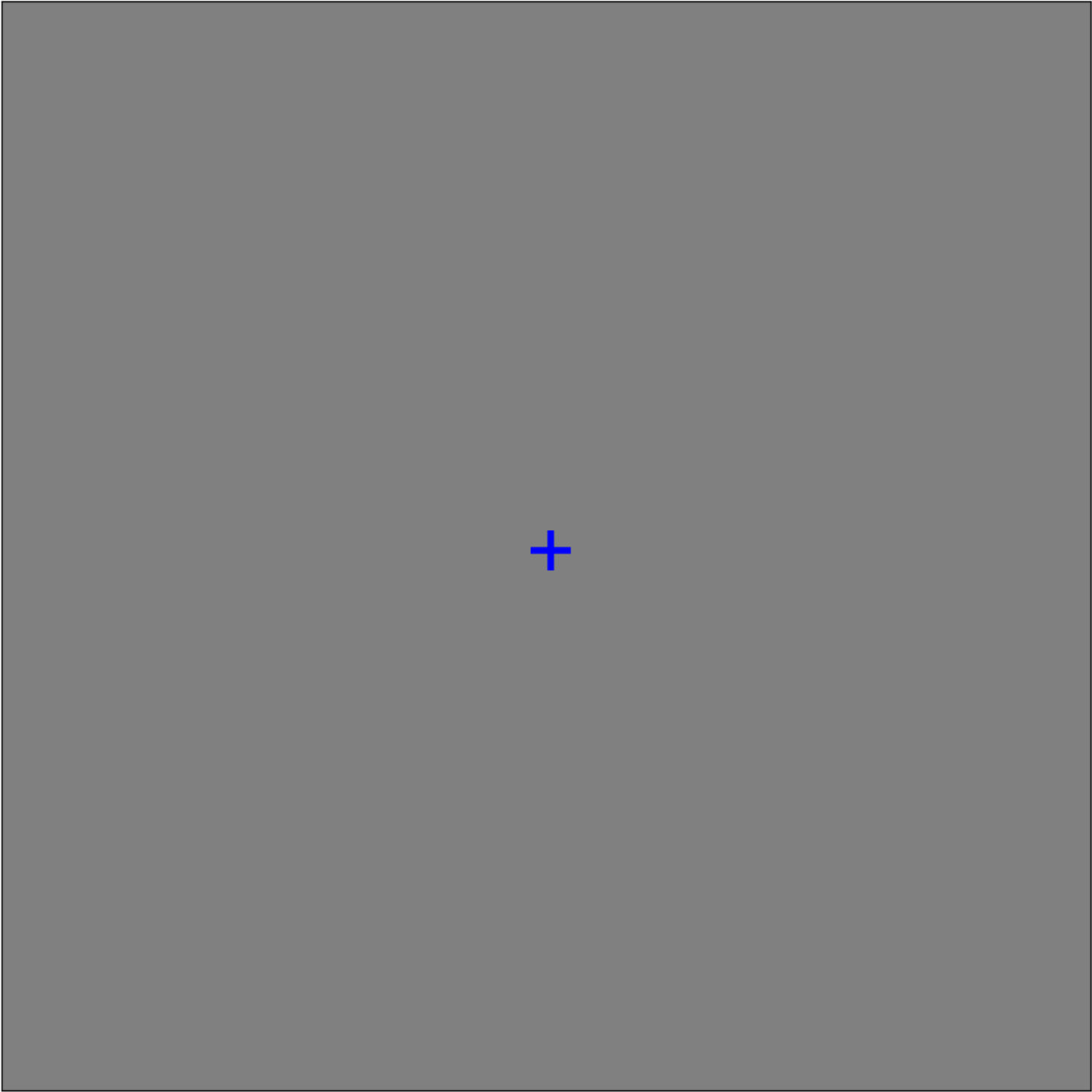


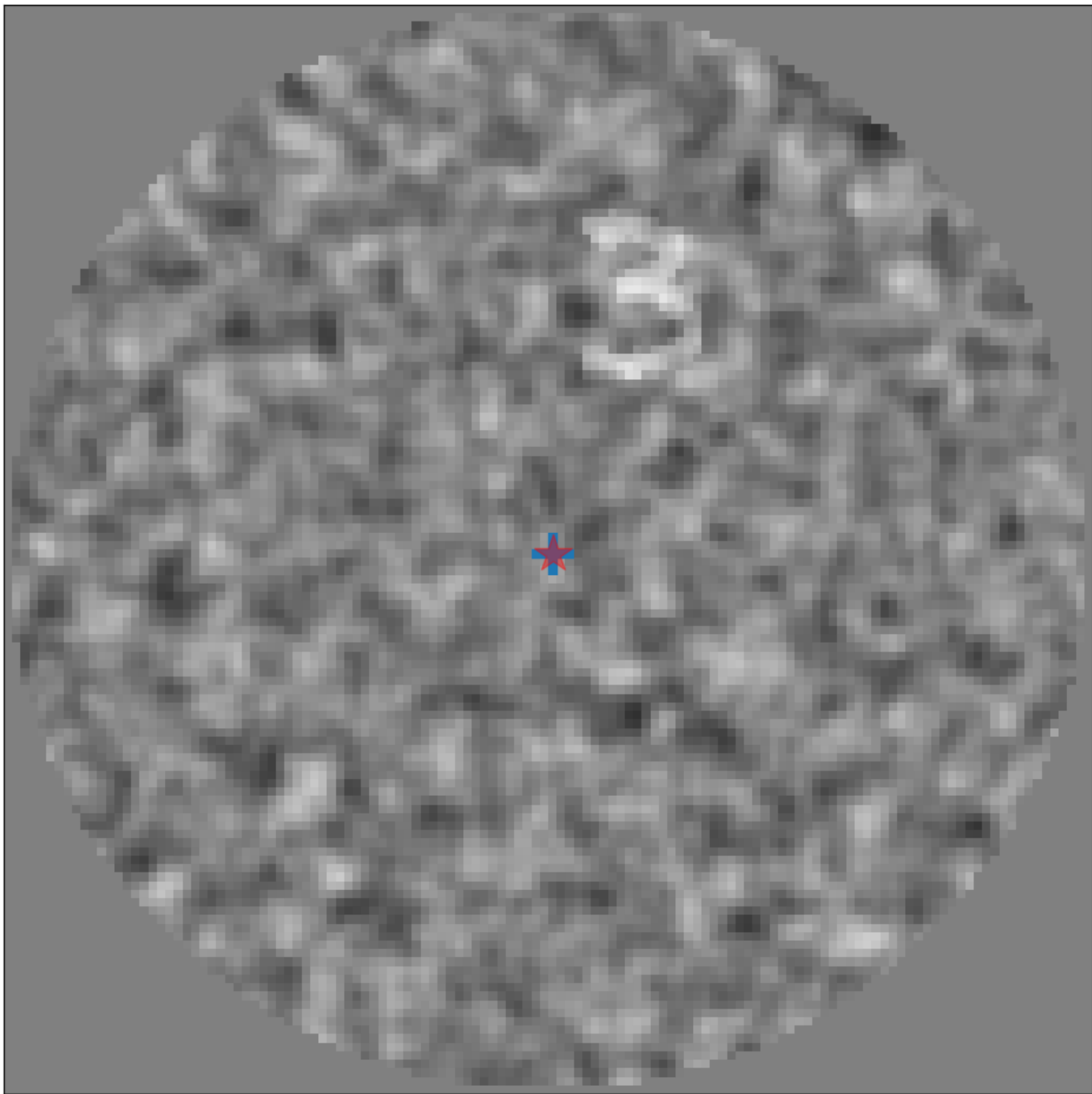


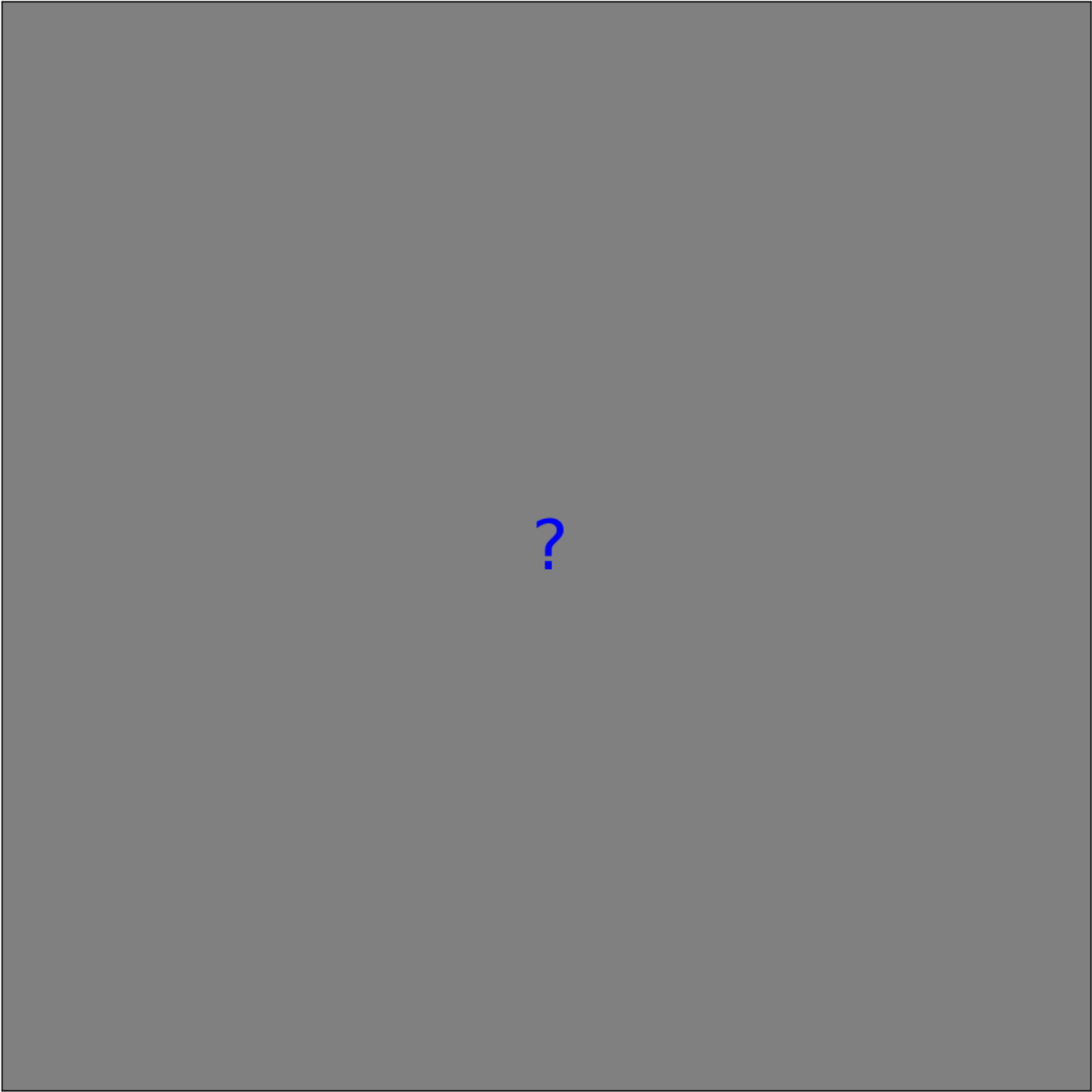




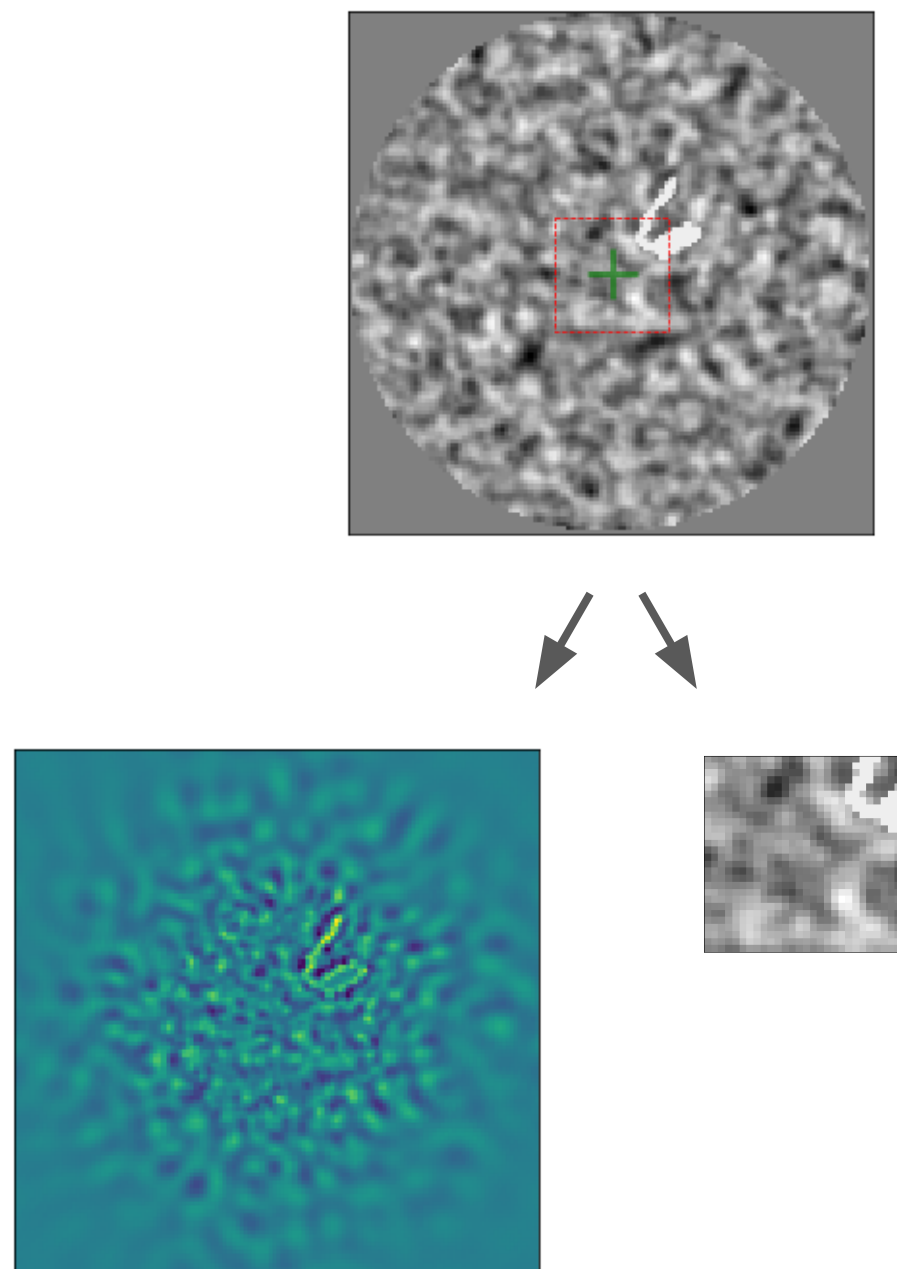








Methods: What/Where separation



Approximate Information Gain :

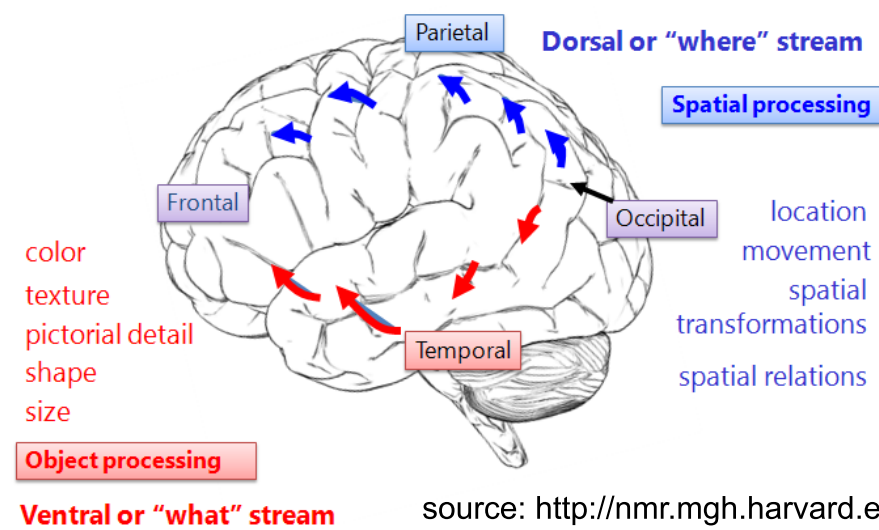
$$E_y [\log P(Y|x, u) - \log P(Y)] \\ \approx \\ \log P(\hat{y}|x, u) - \log P(\hat{y})$$

Future Central
Accuracy
(after saccade)

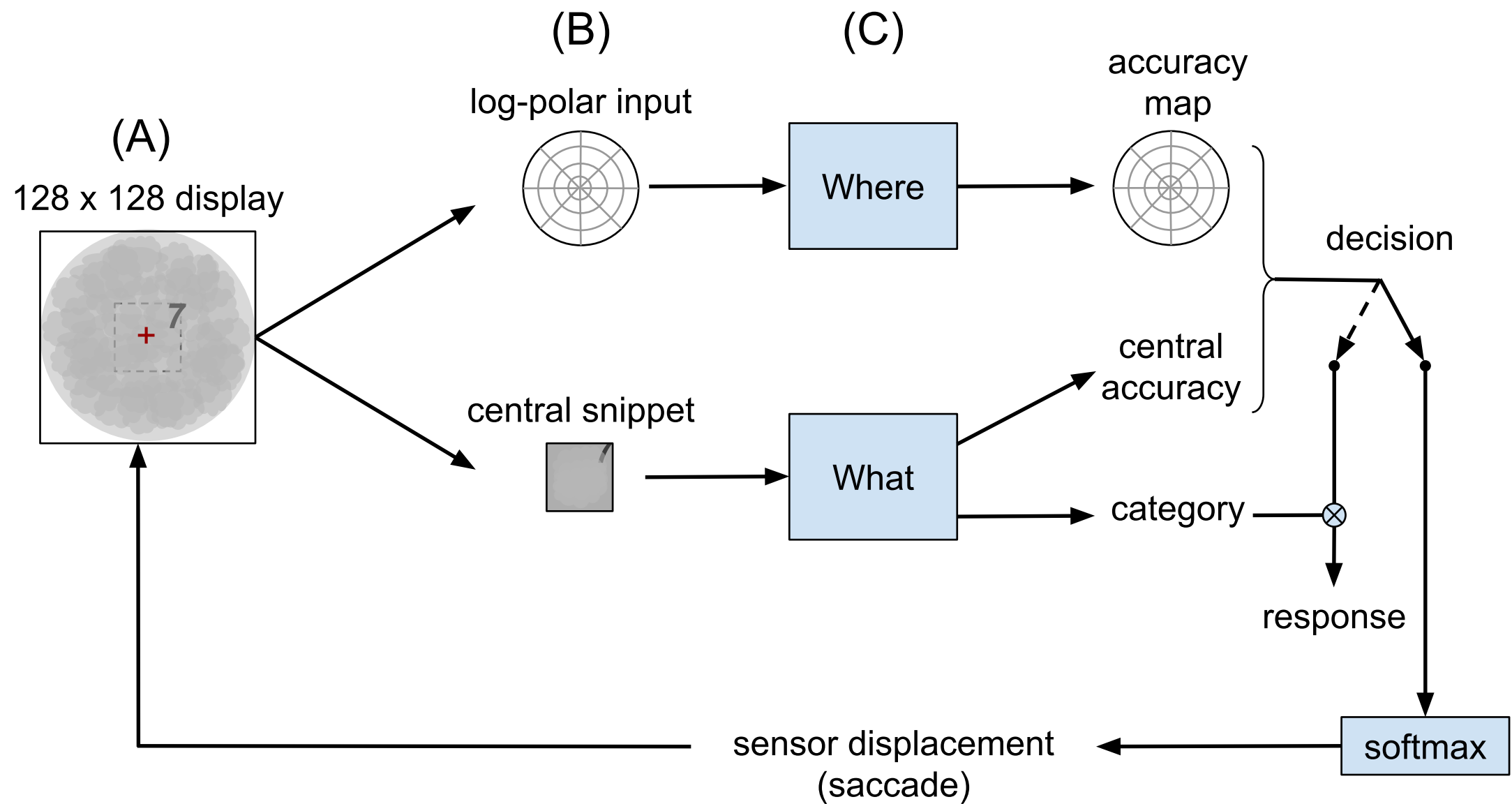
Central Accuracy

What/where pathways in visual processing

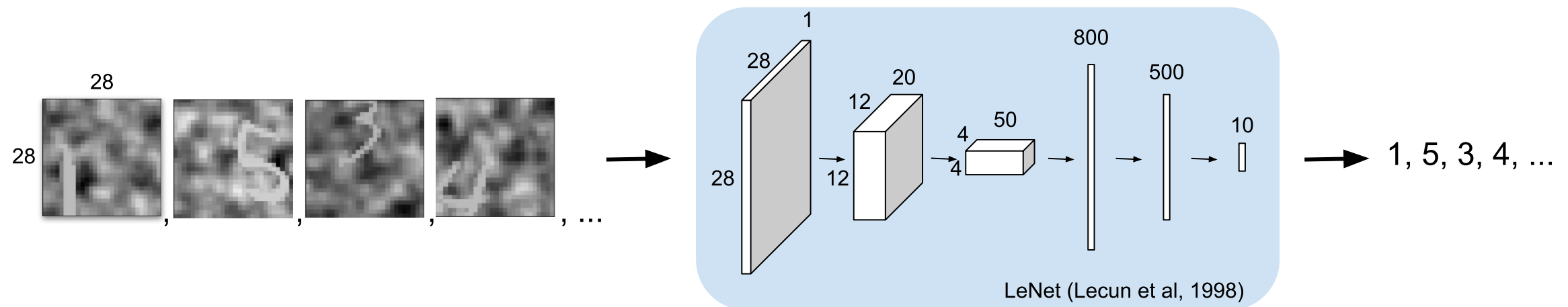
Mishkin, M., Ungerleider, L. G., & Macko, K. A. (1983).



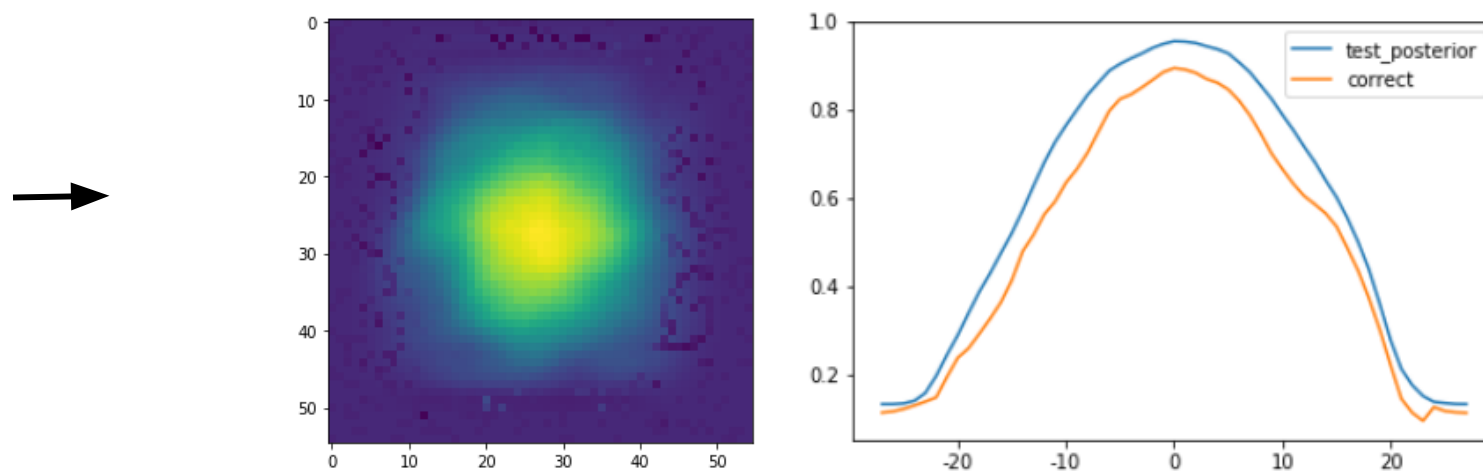
Methods: Computational Graph



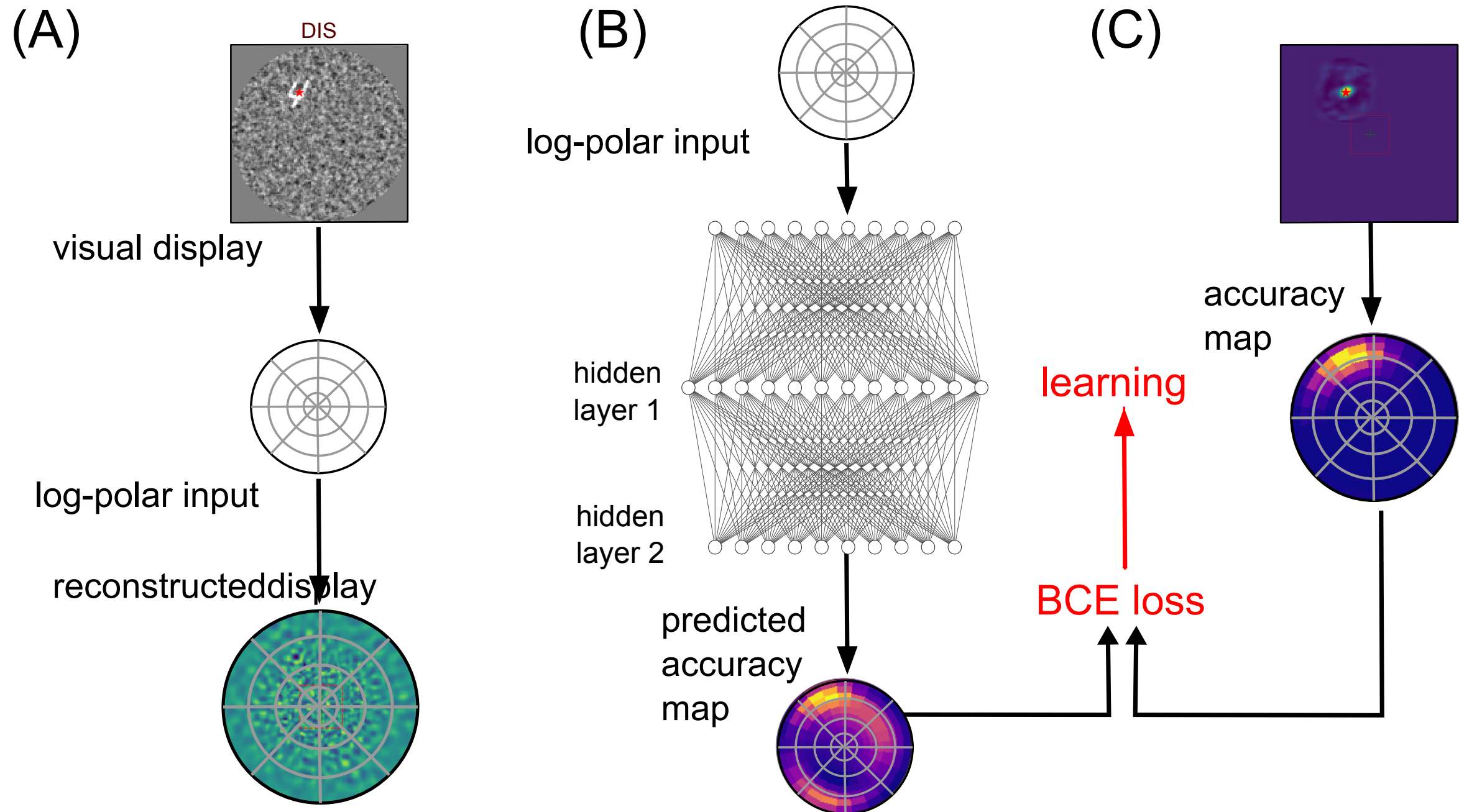
Methods: What



Shift-dependent Accuracy Map



Methods: Where



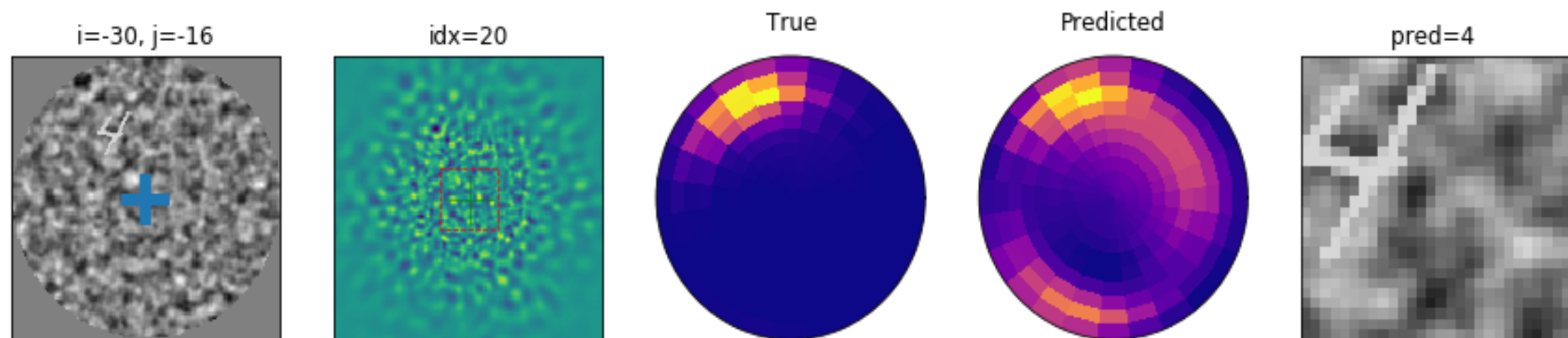
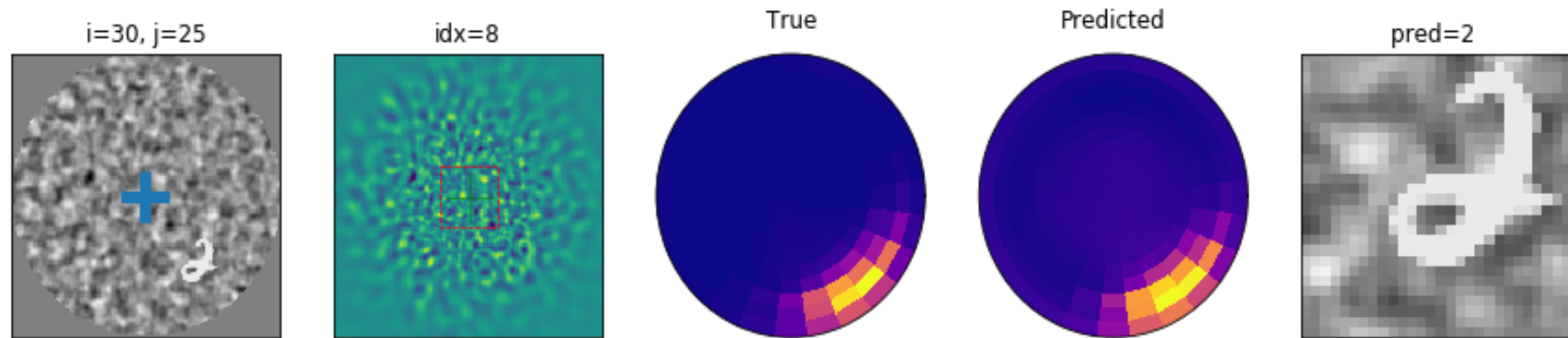
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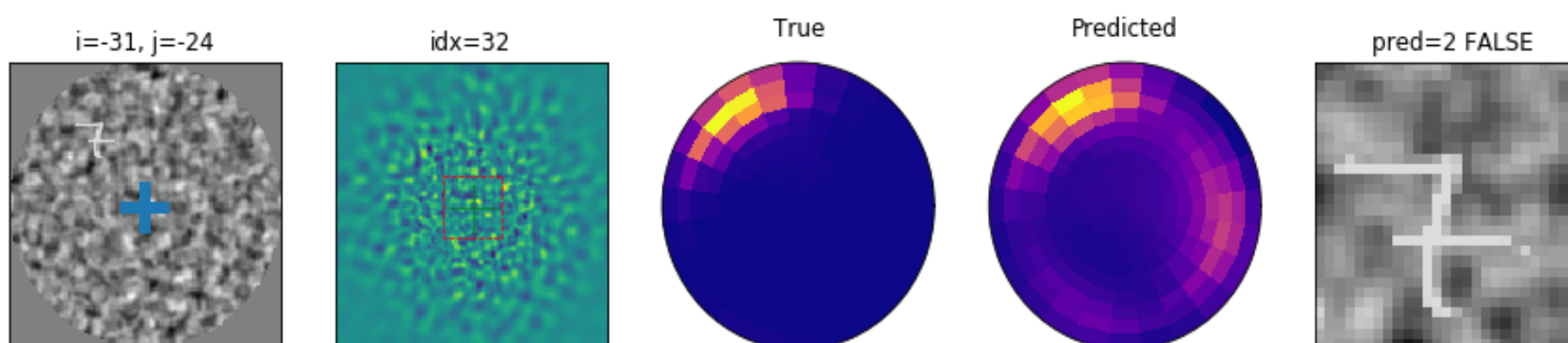
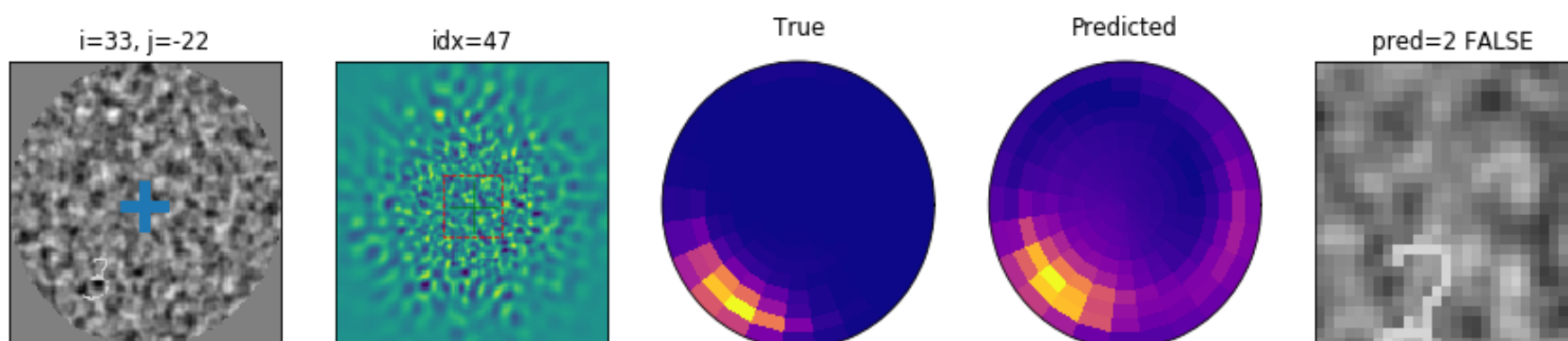
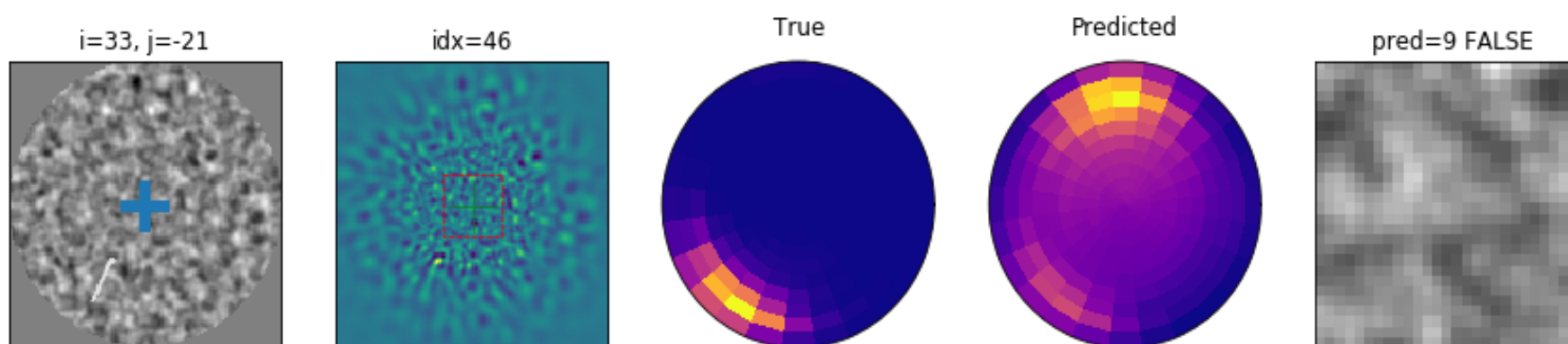
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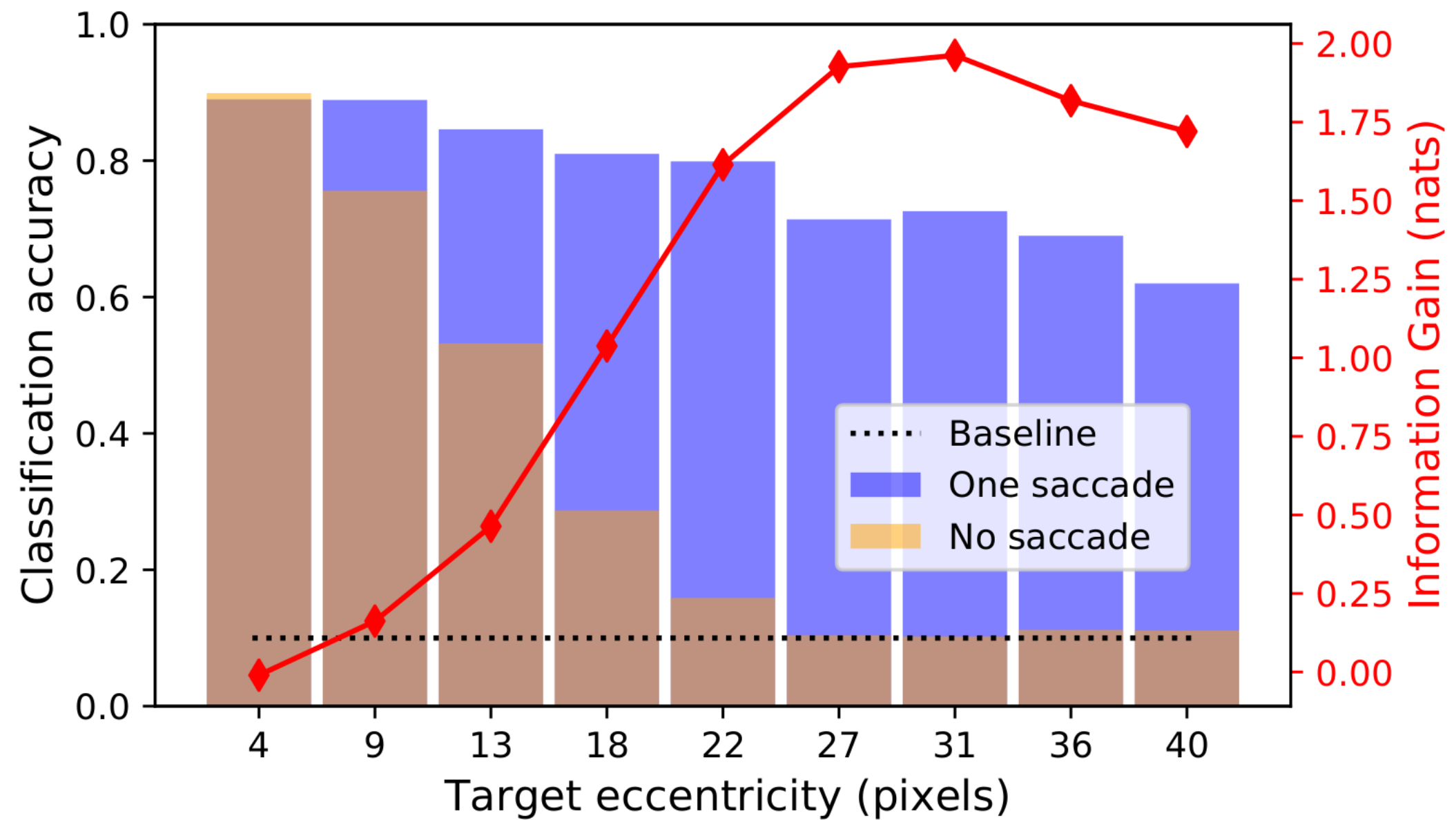
Results: success



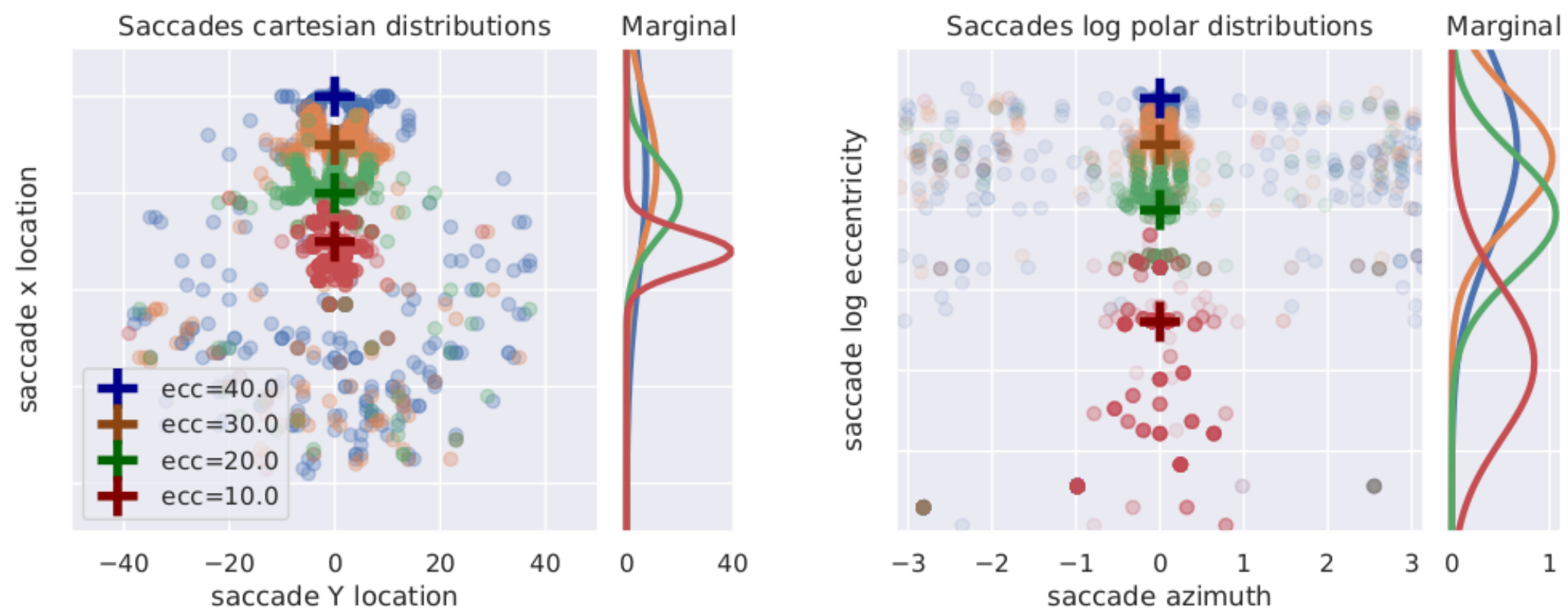
Results: failure



Effect of eccentricity



Saccades distribution



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Main results:

- A new interpretation of Information Gain in visuo-motor action selection :
 - Center-surround interpretation
 - An effective decoding scheme with strong bandwidth reduction
 - Information-gain based selection of action (actor/critic)
- A sub-linear object detection for image processing:
 - A full log-polar processing pathway (from early vision toward action selection)
 - Ready for up-scaling
- The combination of accuracy prediction and accuracy-seeking policies can be formally recast in an active inference framework (see paper)

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<https://laurentperrinet.github.io/talk/2020-09-14-iwai>