

# Temporal context as a posterior distribution over latent states

Stephanie C.Y. Chan Carl Nist-Lund Yael Niv

Kenneth A. Norman



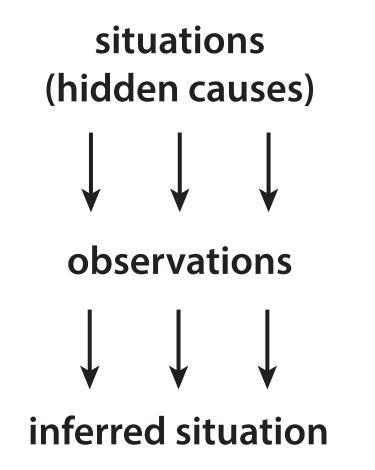
# Introduction

### Situation models and "schemas"

- · Serve to organize thoughts and experiences as we encode them into memory
- · The posterior-medial network (PM network) of brain regions might be involved in their construction and application

### Central question: How do we infer what situation we are in?

## Using Bayesian latent cause models?



- · Situations can be viewed as hidden causes that give rise to observable events
- · We can use Bayesian inference to infer the current situation, as the posterior probability distribution P(situation | observations)

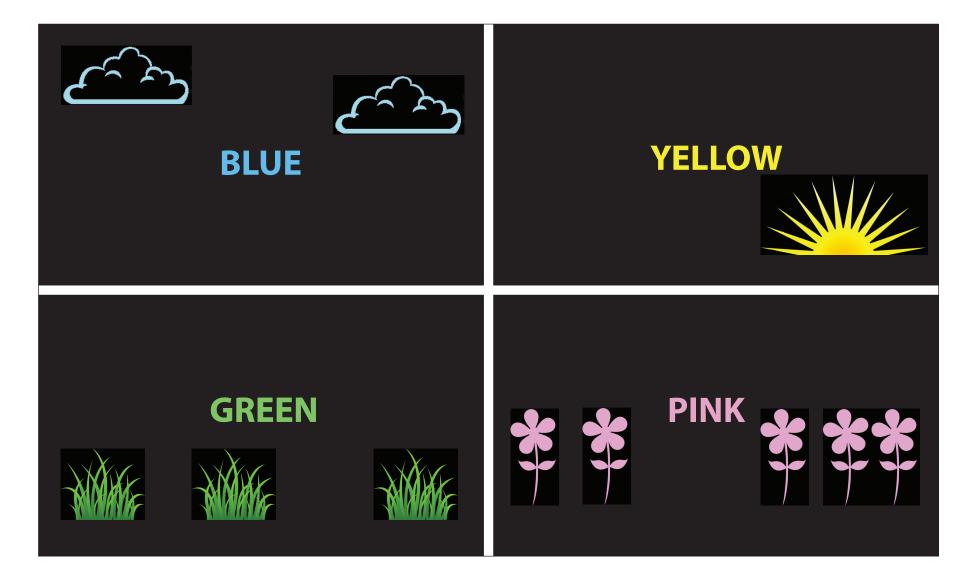
# The posterior-medial network

(Ranganath & Ritchey 2012)

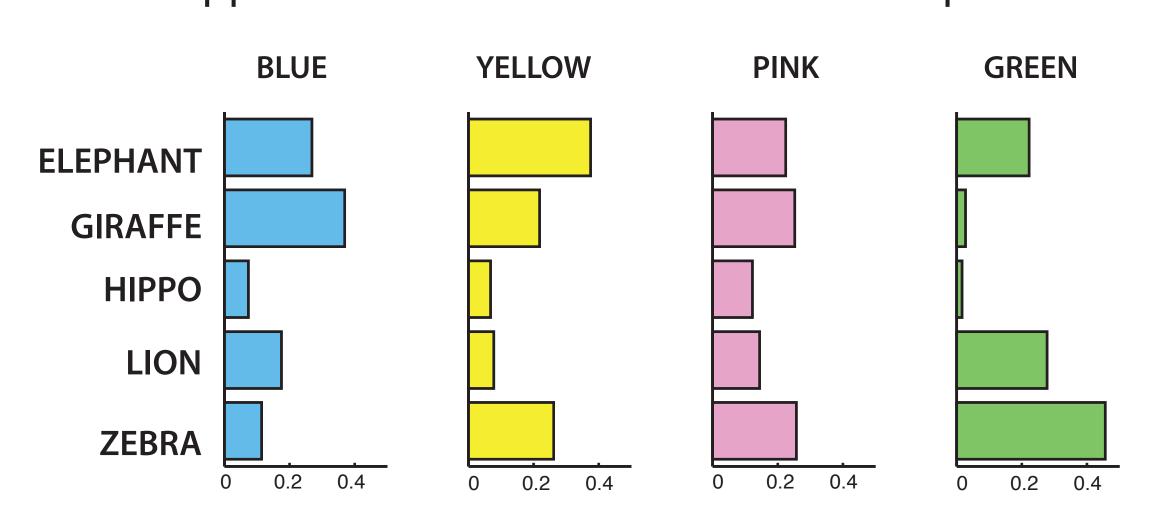
Hypothesis: Brain regions implicated in situation modelling (the PM network) represent the posterior distribution over situations, as computed by Bayesian latent cause models.

# 2 "Animal Safari" posteriors task

The safari is divided into 4 "zones"

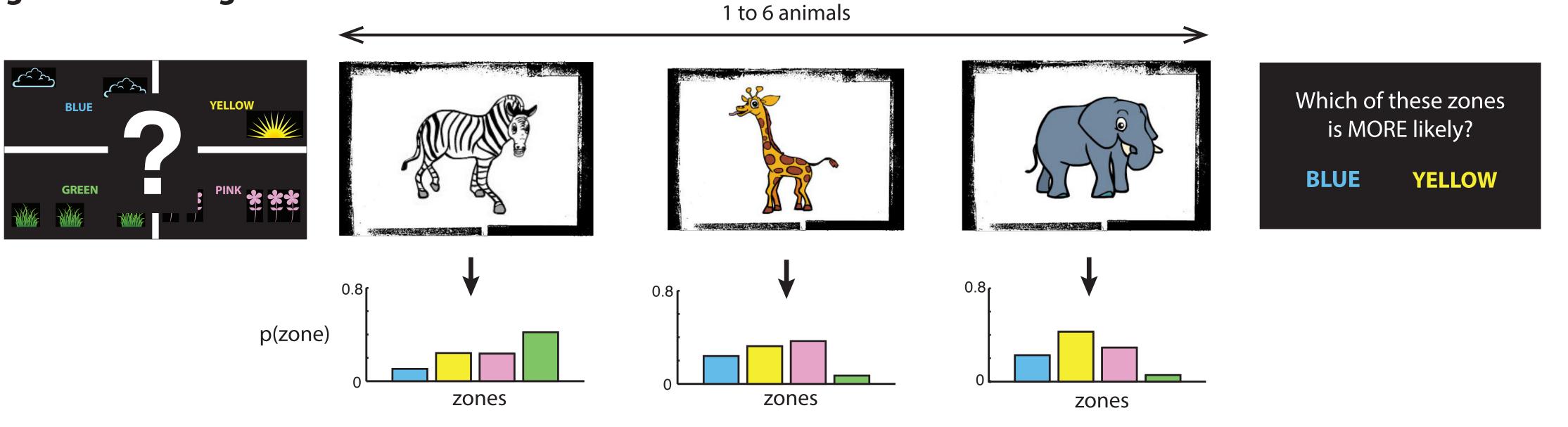


Animals appear in different zones with different probabilities:



Subjects are trained on these probabilities before scanning.

### **During fMRI scanning:**



Subjects must continuously update their beliefs about the posterior probability of each zone.

# 3 Representational similarity analysis

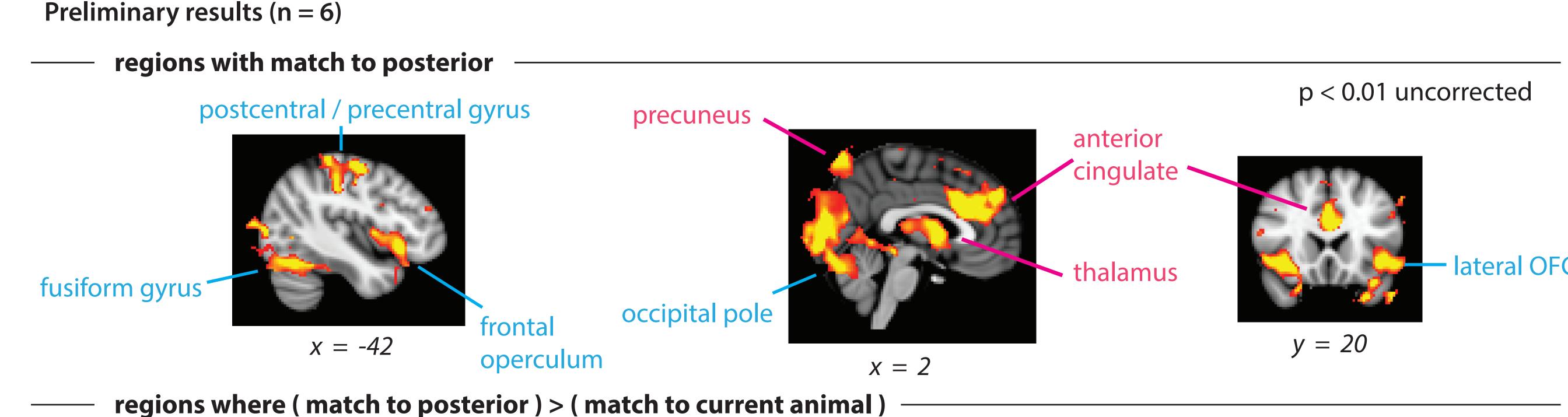


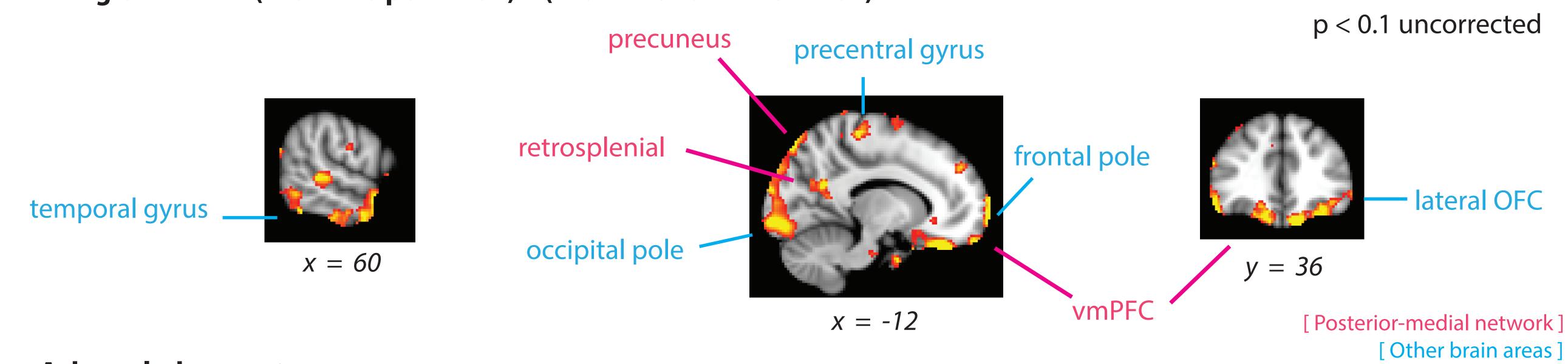
Take correlation of these two matrices to obtain the representational similarity match of ROI and posterior

We can also compute similarly compute the representational similarity for various **null models**:

• current animal = indicator vector, e.g. [ 0 1 0 0] if lion at timestep t

# 4 Whole-brain searchlight





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