### Modeling managment of access to working memory as a self-evalution process for intrinsically motivated prediction

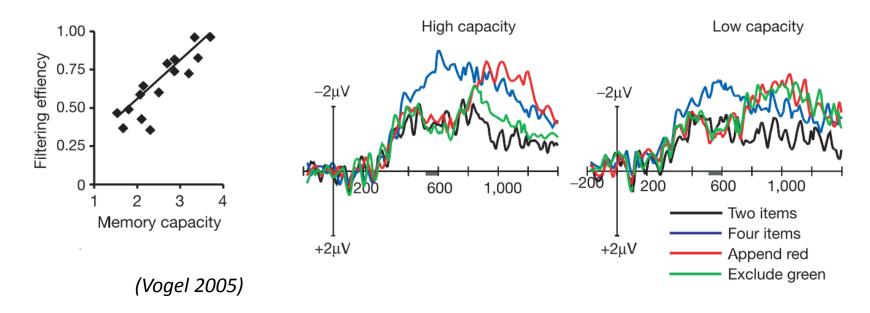
## Catherine Wacongne ICON 2014

#### Introduction

• Working memory has a limited capacity (Vogel 2004)

➔ Necessity to manage working memory content

linked to the capacity to filter out irrelevant items



• Implication of the basal ganglia and valuation system (Mc Nab 2008, O'Reilly & Frank 2006)

Main question

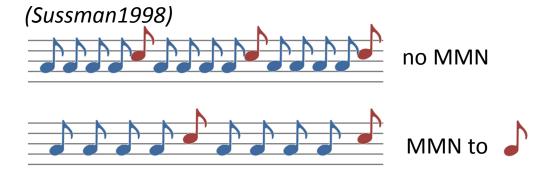
# What is working memory used for in the absence of an external task ?

#### limits of automatic processing

#### Auditory processing :

(Bekinschtein 2009)





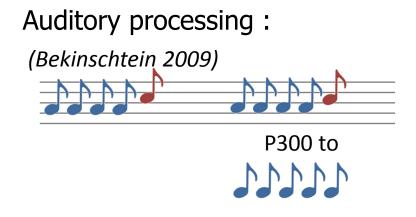
#### Serial reaction time tasks :



- 1st order transitions = independent of attention.
- higher order transitions is affected by attention demanding secondary tasks (Curran and Keele 1993)

Evidence for predictive systems in the brain Preattentive processing is strongly affected by temporal gaps

#### Attentive processing and working memory



#### Serial reaction time tasks :



The attention-related P300 is not affected by variable or long time intervals

Dependencies at distance up to 6 items can be learned (in several thousands of trials) if the subject is **attentive**, although the learning remains **implicit** (*Remillard*, 2008, 2010)

Some responses to violation depend on attention and working memory

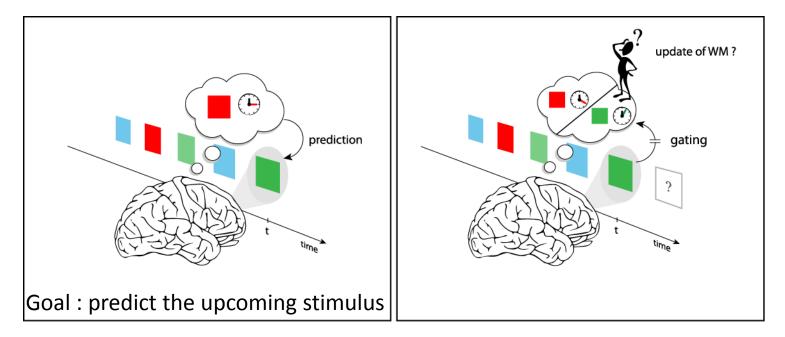
- → Working memory can bridge temporal gaps
- $\rightarrow$  Even in the absence of a task

What is working memory used for in the absence of an external task ?

Predicting future events

How can we manage working memory content to predict future events ?

#### **Computational Problem**



#### "Rules":

- working memory can hold **one** item for arbitrary long time
- Past items that are not kept in working memory cannot be retrieved
- working memory slot encodes identity and number of stimuli since the current content was stored

• Probability estimate of the upcoming item is computed based on the learned P(stimulus|**Slot content**)

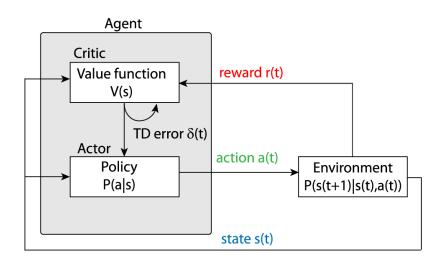
Updating can be seen as an internal decision

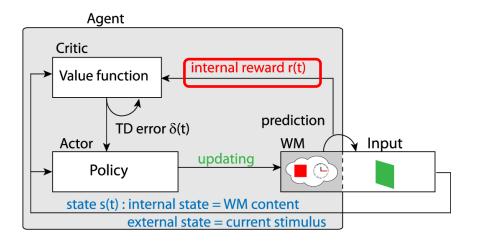
The implication of the value system implies the existence of a reward

Monkeys prefer informative options even when they do not change the objective reward => consistent with the idea of a rewarding value of predictive information *(Bromberg-Martin & Hikosaka, 2009)* 

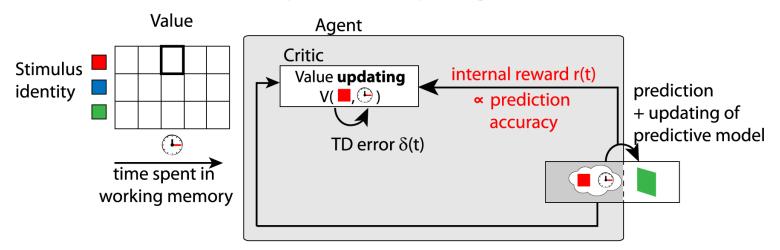
Classical reinforcement learning

Working memory updating model

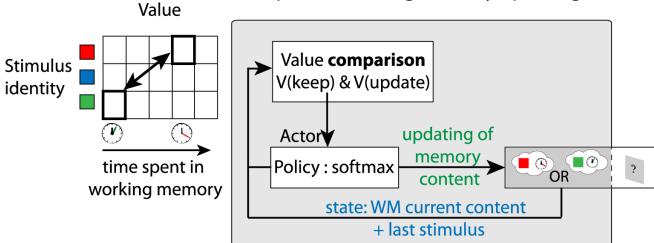


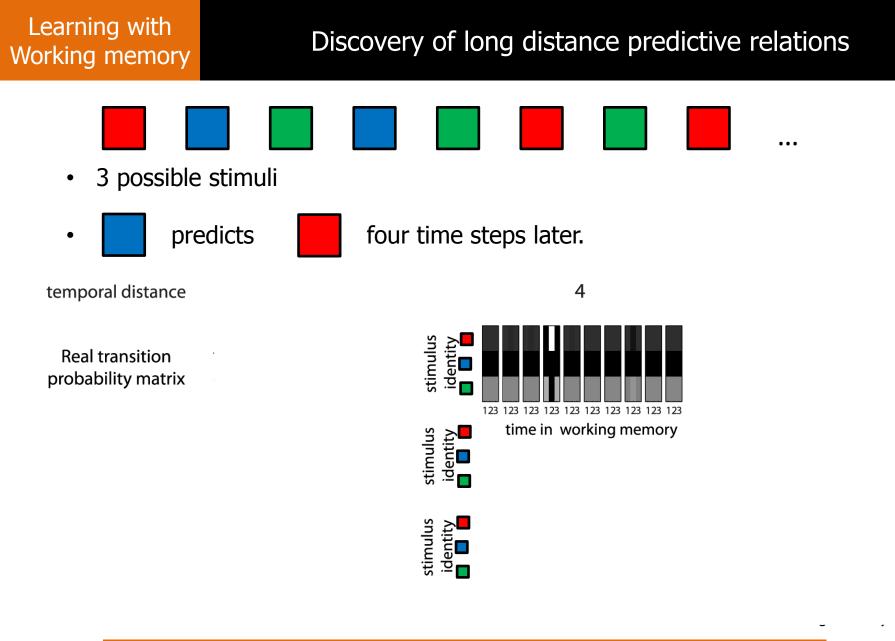


#### "Critic" phase : value updating

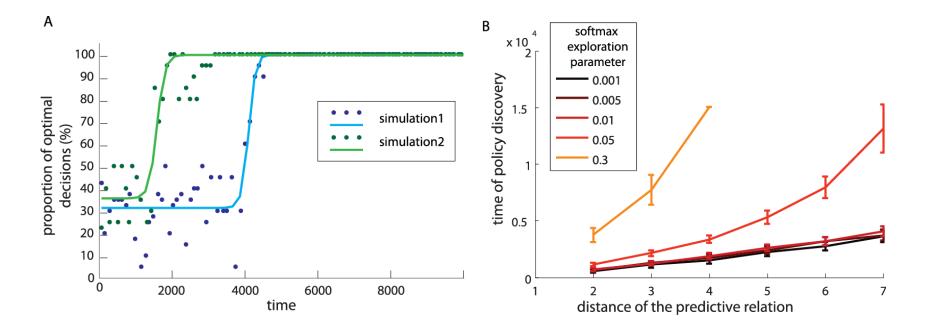








Discovery of long distance dependencies Working memory dependent but not necessarily explicit Optimal strategy: store the predictive item until it is not predictive any more



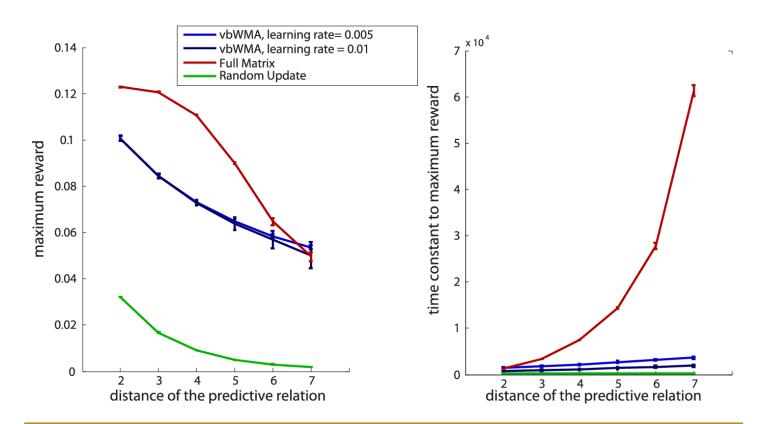
Sudden discovery of the policy Stabilization of the policy before the stabilization of the conditional transition probability estimate

#### Learning with Working memory

Speed

Comparison of the predictive performances of the reinforcement learning model with 2 learning rates compared to

- A model that tracks the entire relevant history (last past 4 stimuli)
- A model that has one slot but updates at random
  Performance

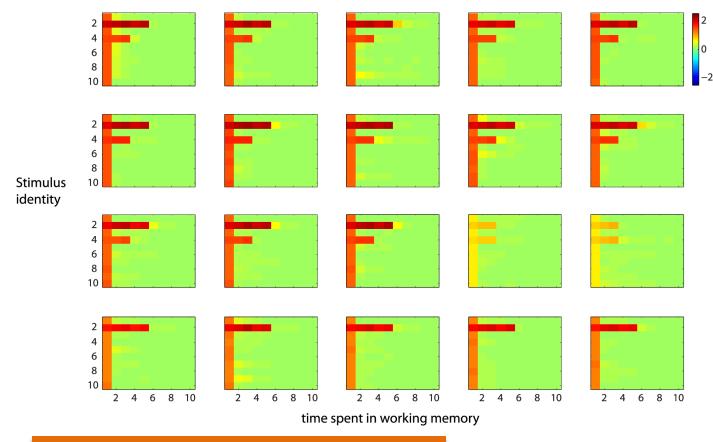


#### → Good tradeoff between final performance and speed of learning

#### Competition between temporal regularities

20 simulations with the same dependencies:

- "stim2" is predictive 3 and 5 time steps later
- "stim4" is predictive 3 time steps later



→ Exploration/exploitation tradeoff

- Predicting accurately future events is one of the brain's intrinsic goals
- Limitations in working memory capacity imposes to select relevant information
- Working memory can be managed to achieve this goal in the absence of extrinsic rewards
- A reinforcement learning mechanism is suitable for the discovery and exploitation of long distance dependencies
- Having a small working capacity can be advantageous in terms of learning speed
- The model predicts impairment of long distance dependencies learning in patients with striatal dysfunction
- It also predicts a competition between external tasks, and intrinsic goals like prediction

#### Acknowlegments

- Stanislas Dehaene
- Jean-Pierre Changeux





