

Sequence generation

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Sequences

- all behavior and thinking consist of sequences of physical or mental acts
- sometimes in a fixed order as in action routines, or highly trained action patterns
- but potentially highly flexible ... as in language, thinking, problem solving ...

DFT challenge for sequences

- DFT postulates that all neural states underlying behavior/mental process are attractors that resist change...
- but generating sequences of such states require change of state! Conflicting constraints!
- answer: instabilities are induced systematically to enable switching to a next/new attractor



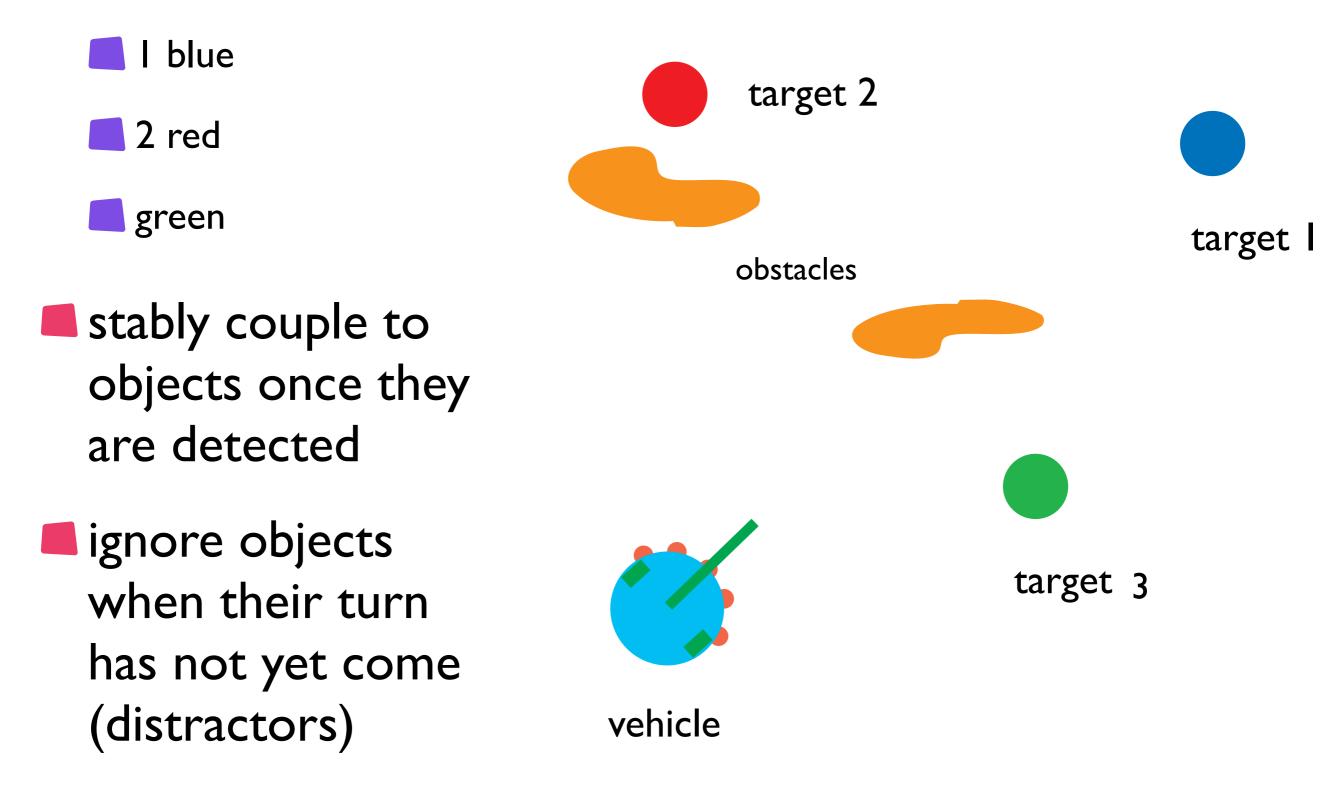
Sequence generation

an illustrative example

the neural/mathematical mechanism

Sequence of physical acts

task: search for objects of a given color in a given order

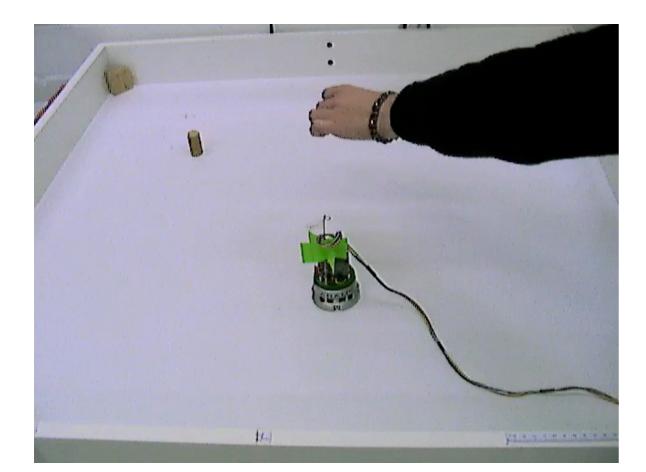


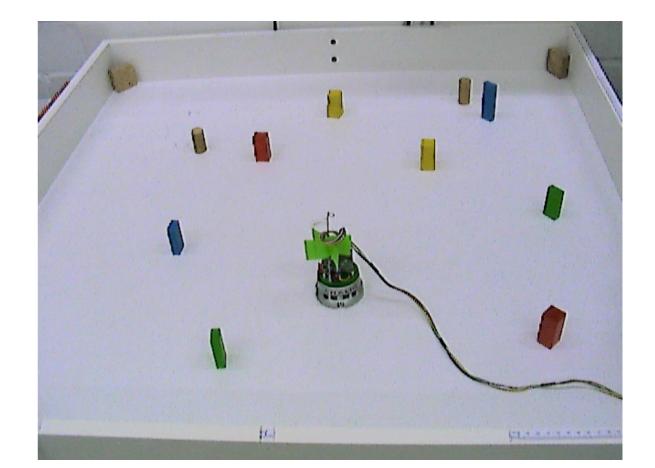
Implementation as an imitation task

- learn a serially ordered sequence from a single demonstration
 - yellow-red-green-blue-red

perform the serially ordered sequence with new timing

yellow-red-green-blue-red

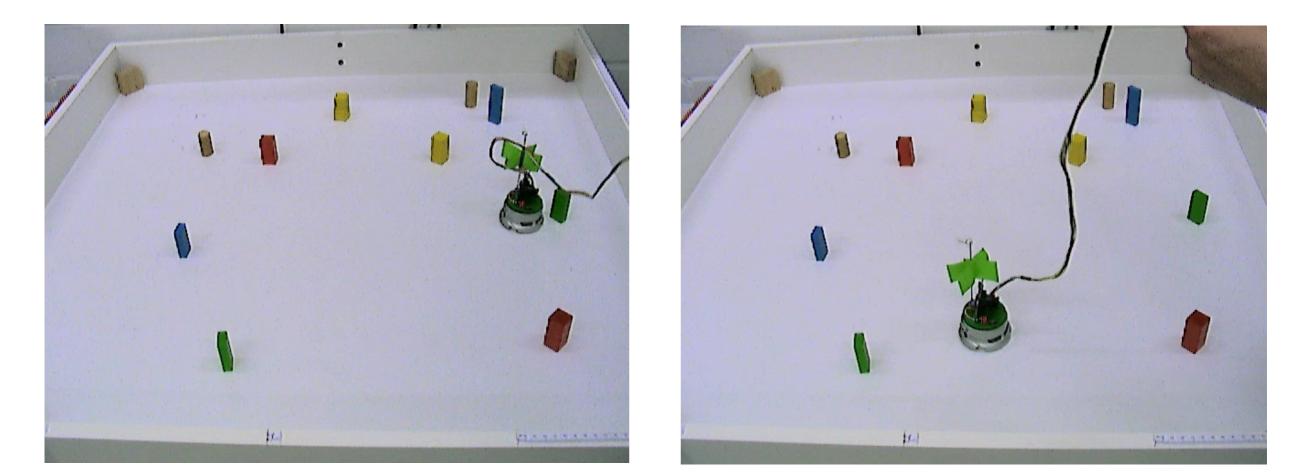




[Sandamirskaya, Schöner: Neural Networks 23:1163 (2010)]

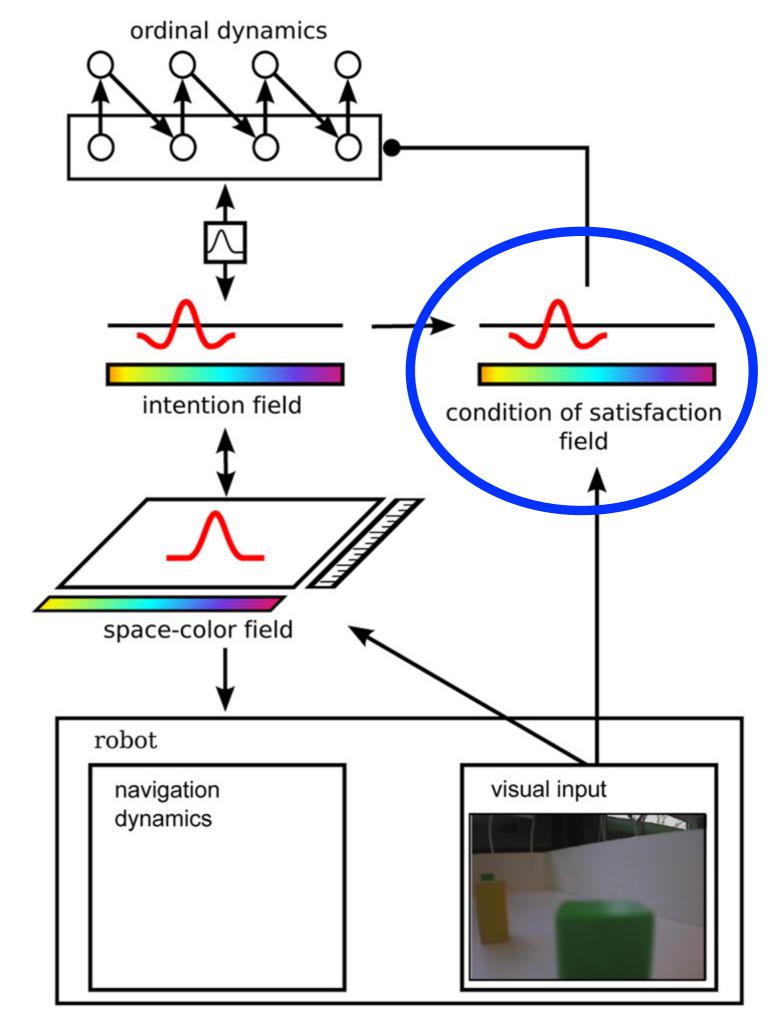
red a distractor

red a target



[Sandamirskaya, Schöner: Neural Networks 23:1163 (2010)]

Condition of Satisfaction (CoS)

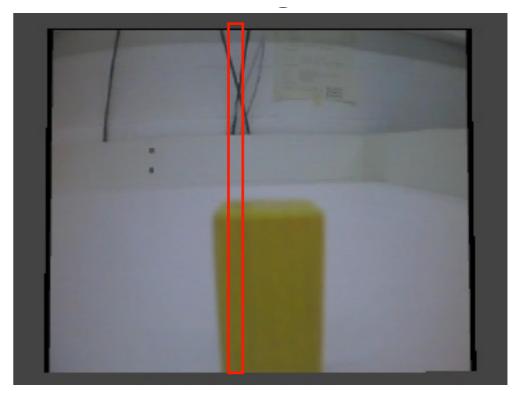


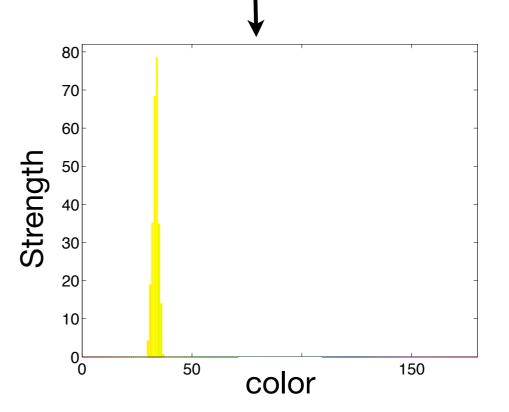
[Sandamirskaya, Schöner: Neural Networks 23:1163 (2010)]

Visual search

Camera image

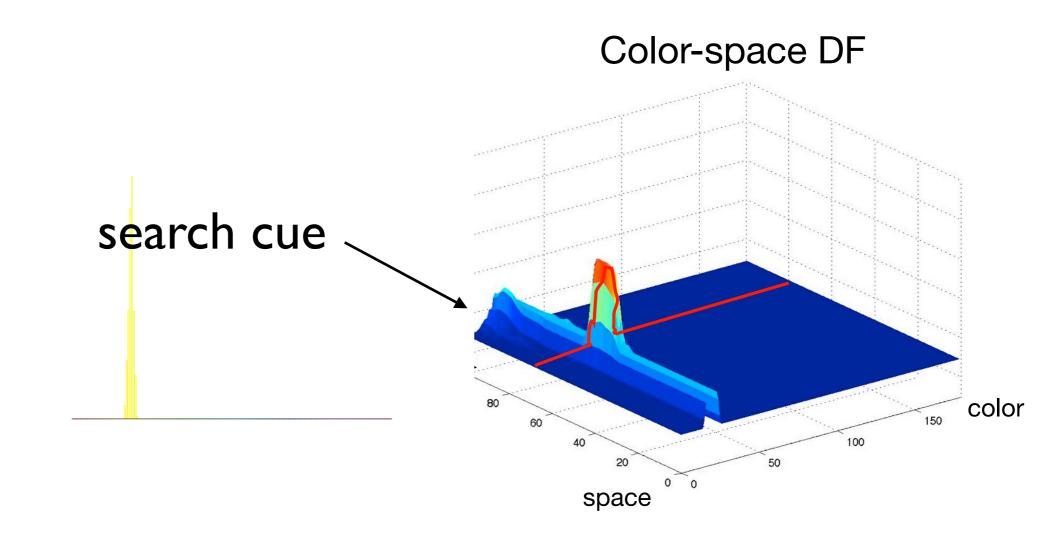
- 2D visual input color vs. horizontal space
- intensity of input from a color histogram within each horizontal location

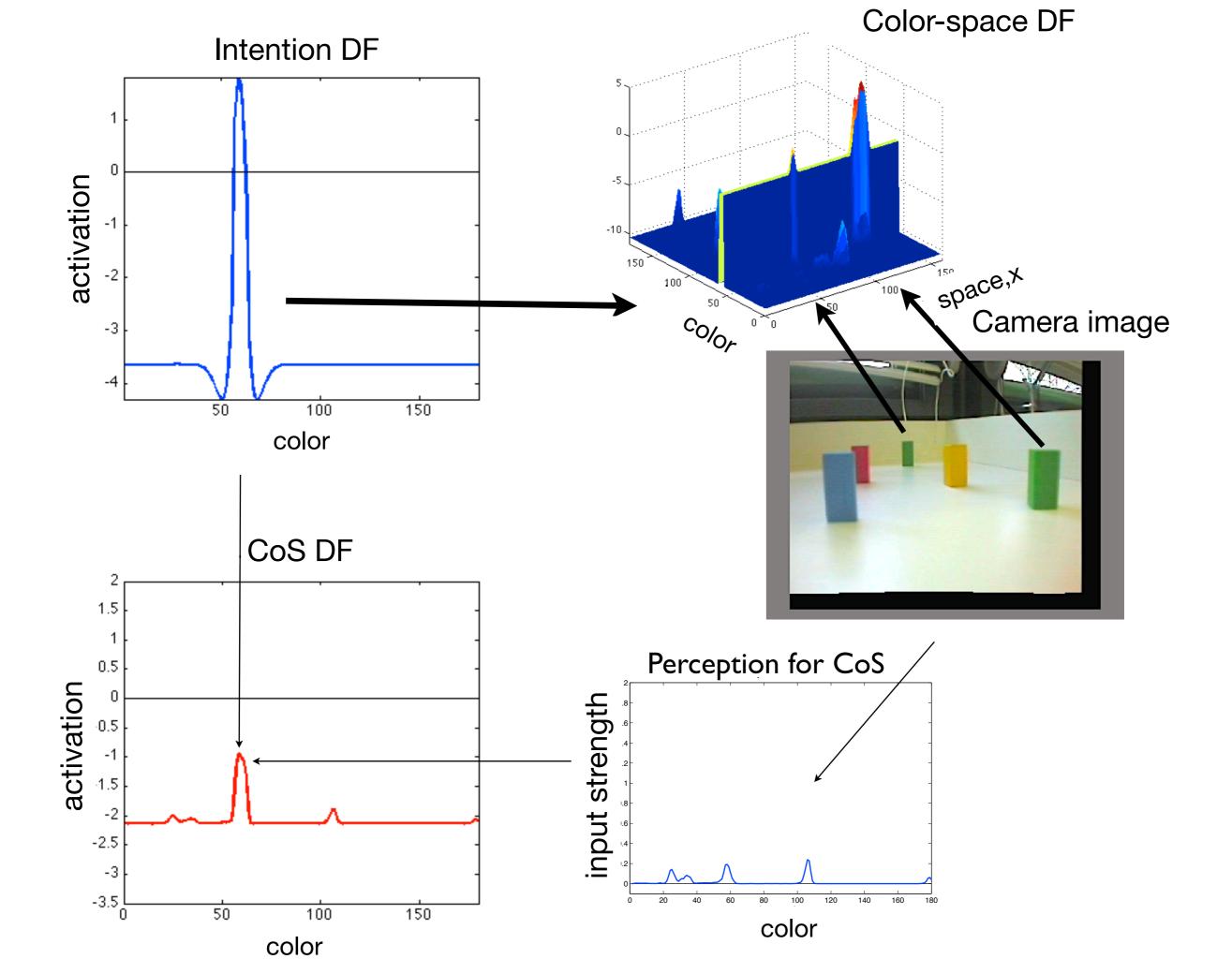




Visual search

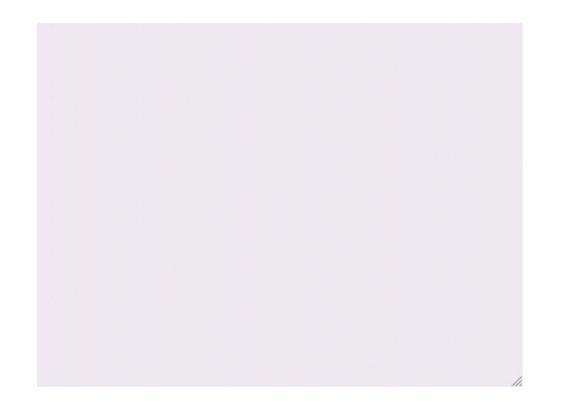
current color searched provides ridge input into a color-space field



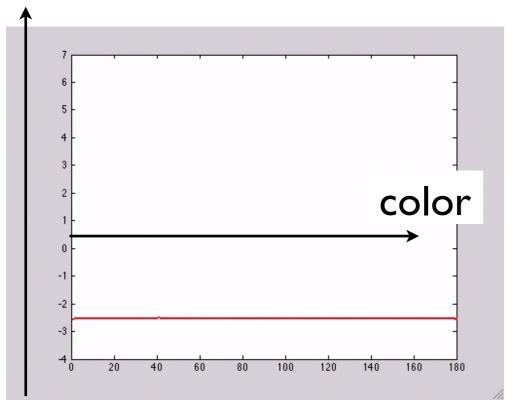


ordinal stack

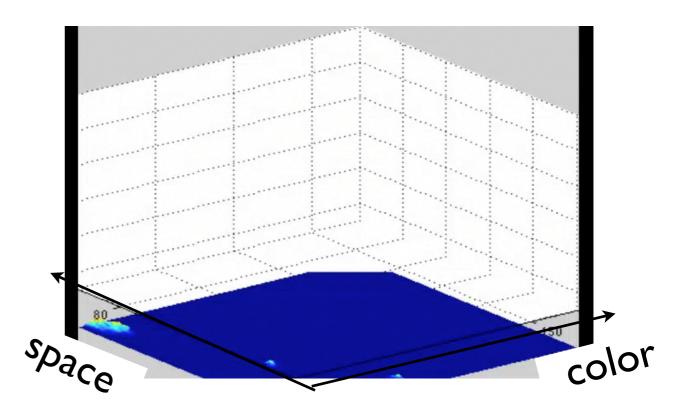
condition of satisfaction (CoS)

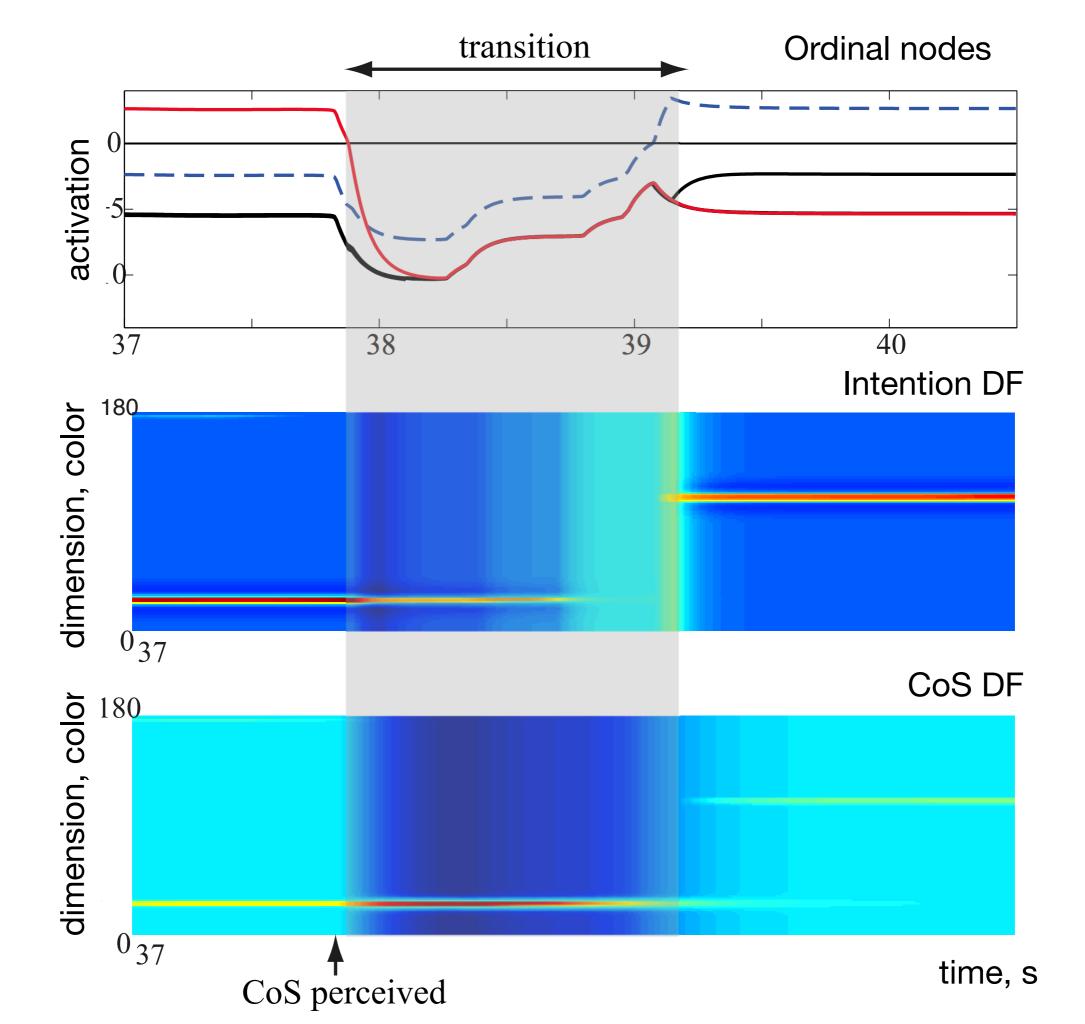


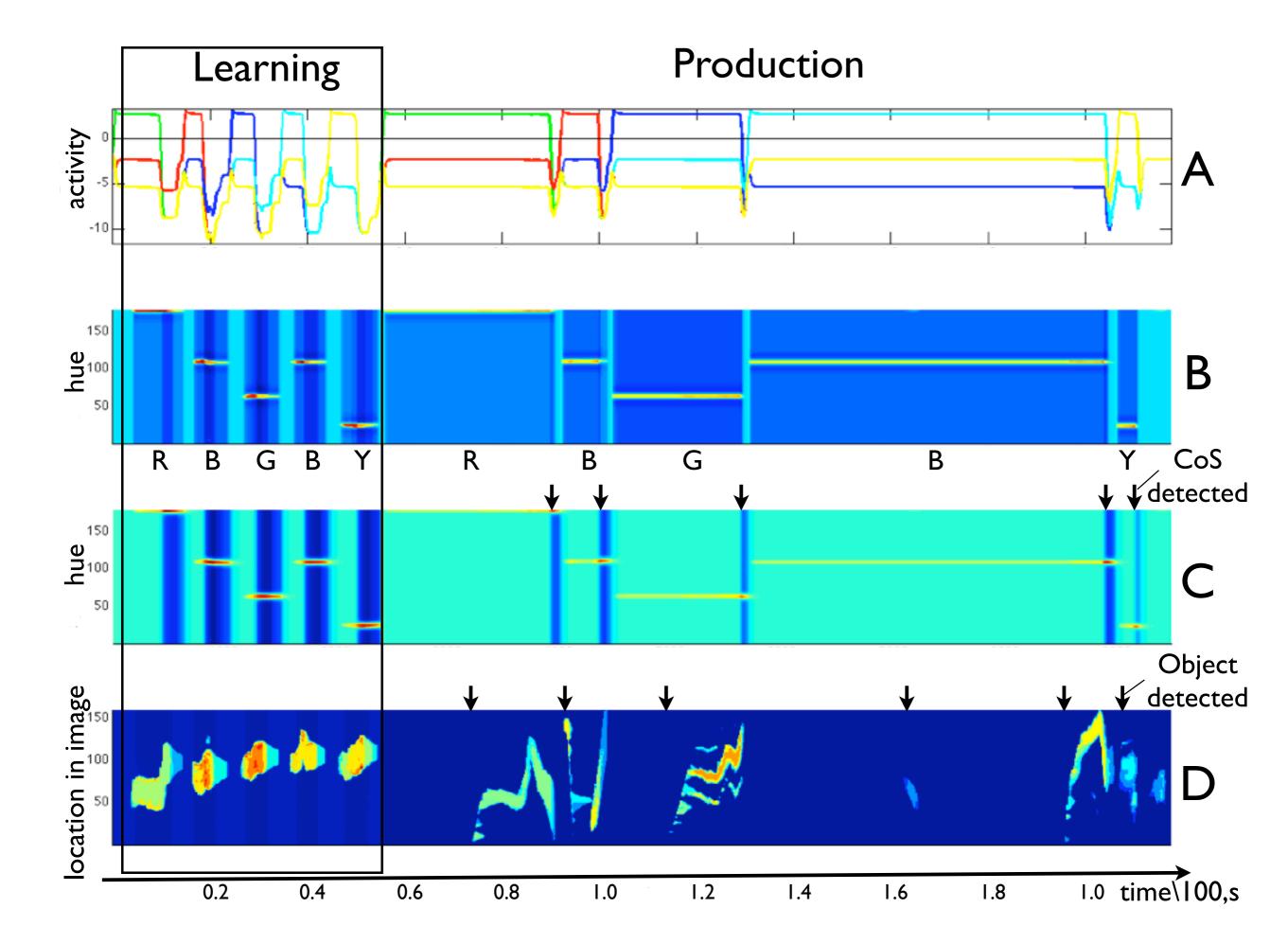
intentional state



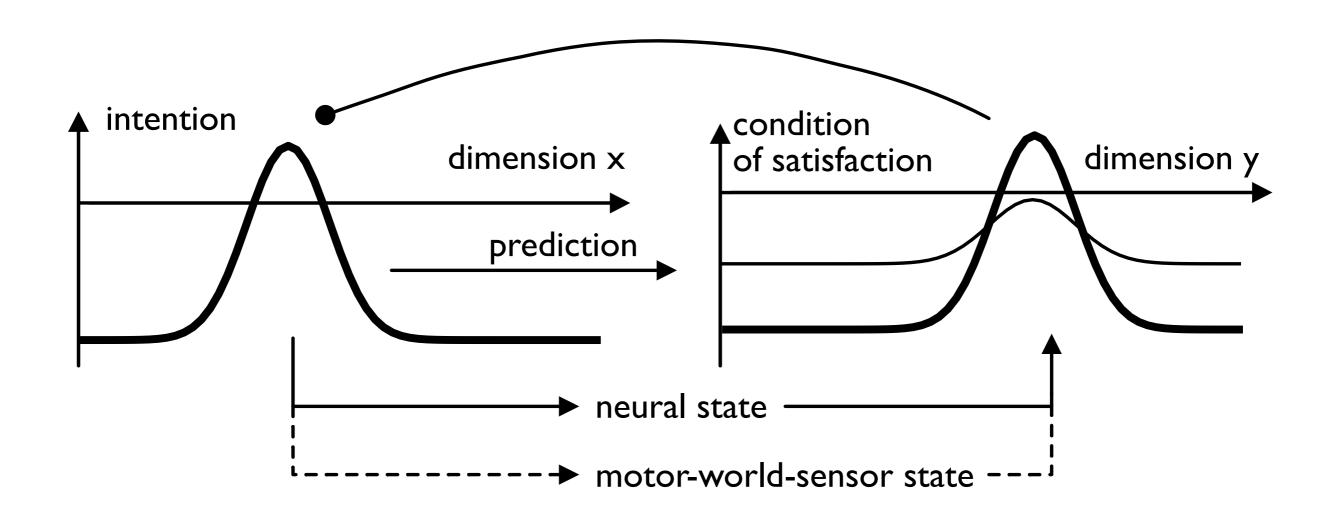
2D color-space field





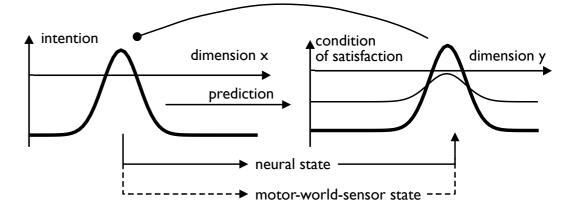


Mathematical mechanism



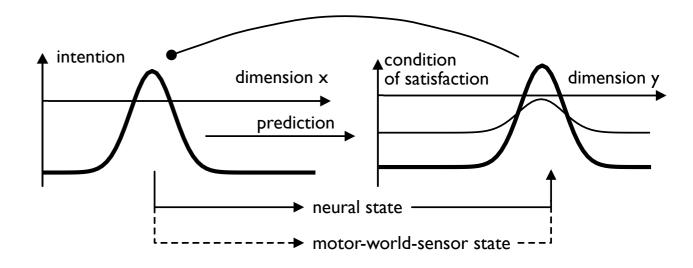
Sequence of instabilities

- the CoS is pre-shaped by the intention field, but is in the sub-threshold state
- until a matching input pushes the CoS field through the detection instability
- the CoS field inhibits the intention field that goes through a reverse detection instability
- the removal of input from the intention to the CoS field induce a reverse detection instability
- both fields are sub-threshold



CoS and efference copy

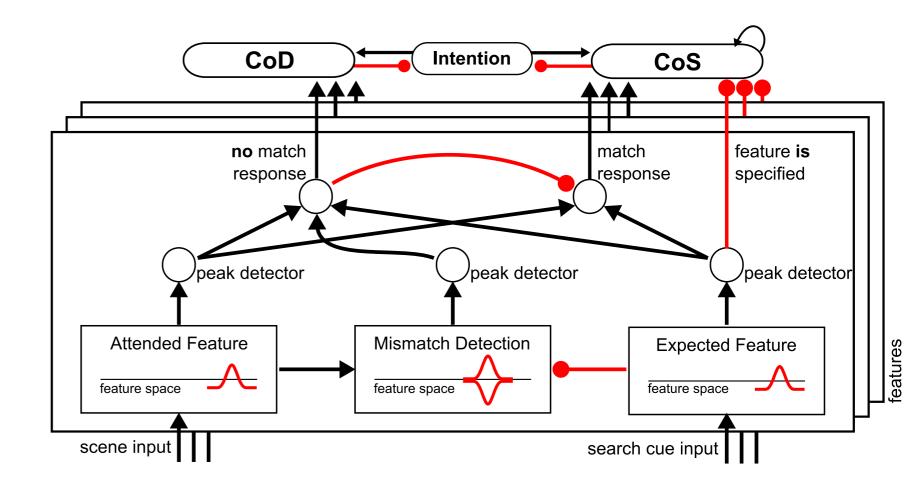
- one could think of the "prediction" implied in the CoS as being a form of efference copy
- that does act inhibitorily...
- but it does so on the (motor)intention, not on the perception of the outcome that is predicted!



Generalization

match-detection => CoS

mis-match (or change) detection => CoD (condition of dissatisfaction)



[Grieben, Schöner, CogSci 2021]

Roadmap How is the next state selected?

once the current state has been de-activated...

three notions

gradient-based selection

📕 chaining

positional representation

an illustration

How is the next state selected?

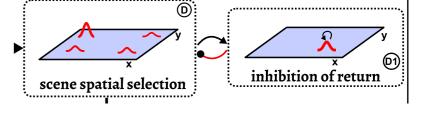
once the current state has been deactivated...

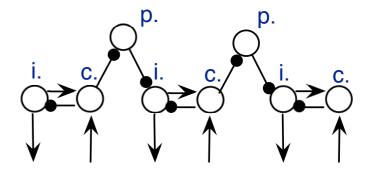
3 notions (~Henson Burgess 1997)

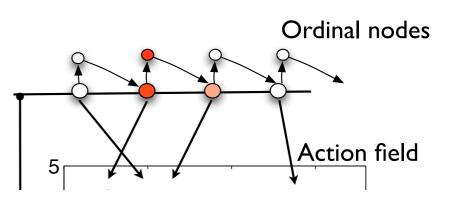


2 chaining

3 positional representation







Gradient-based

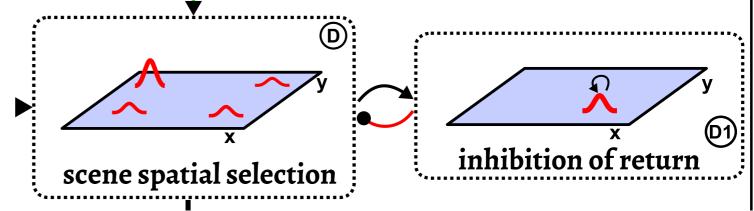
a field/set of nodes is released from inhibition once the current state is deactivated...

a new peak/node wins the selective competition based on inputs...

e.g. salience map for visual search

e.g. overlapping input from multiple fields..

return to previous states avoided by inhibition of return



[Grieben, Schöner, CogSci 2021]

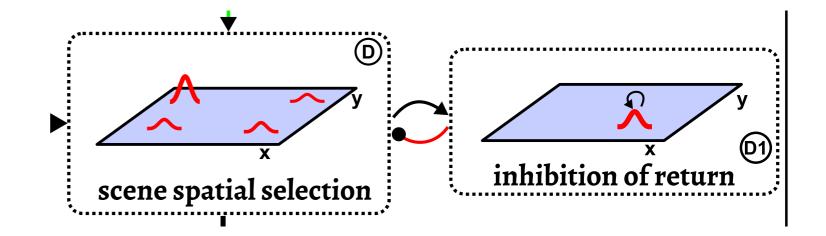
Gradient-based

this is used in many of the DFT architectures

visual search

relational grounding

mental mapping

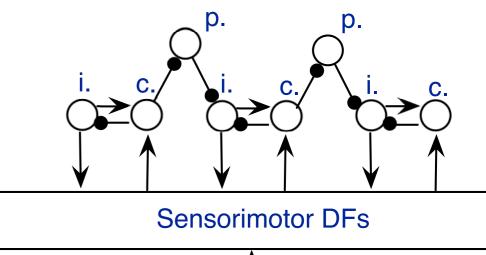


[Grieben, Schöner, CogSci 2021]

Chaining

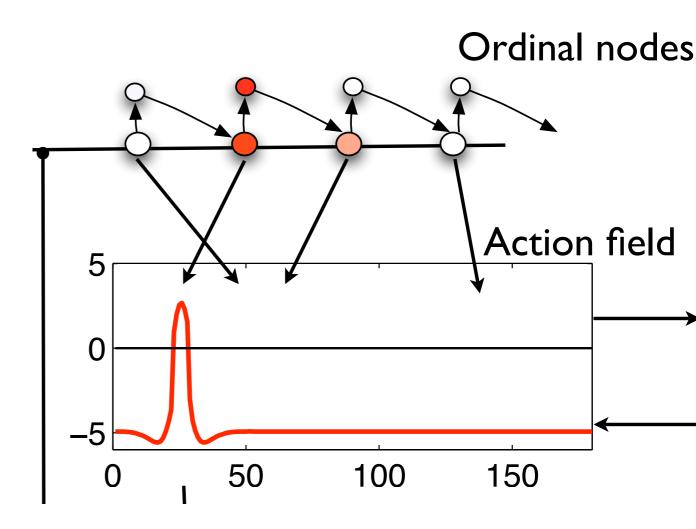
for fixed sequences...

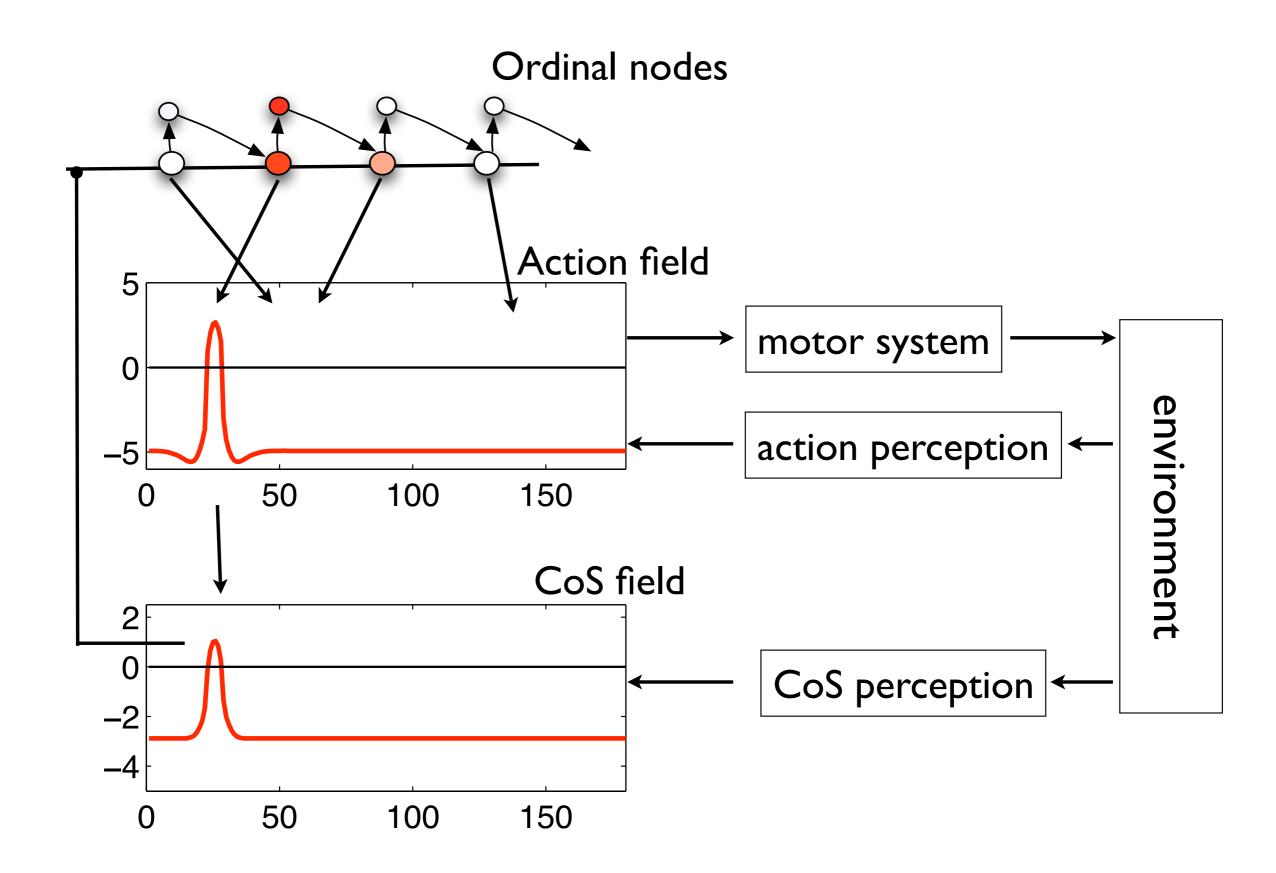
- e.g. reach-grasp
- fixed order of mental operations... e.g. ground reference object first, then target object
- less flexible (e.g., when going through the same state with different futures)
- could be thought to emerge with practice/habit from the positional system



Positional representation

- a neural representation of ordinal position is organized to be sequentially activated...
- the contents at each ordinal position is determined by neural projections from each ordinal node...





[Sandamirskaya, Schöner: Neural Networks 23:1163 (2010)]

Positional representation

essentially chaining with flexible contents

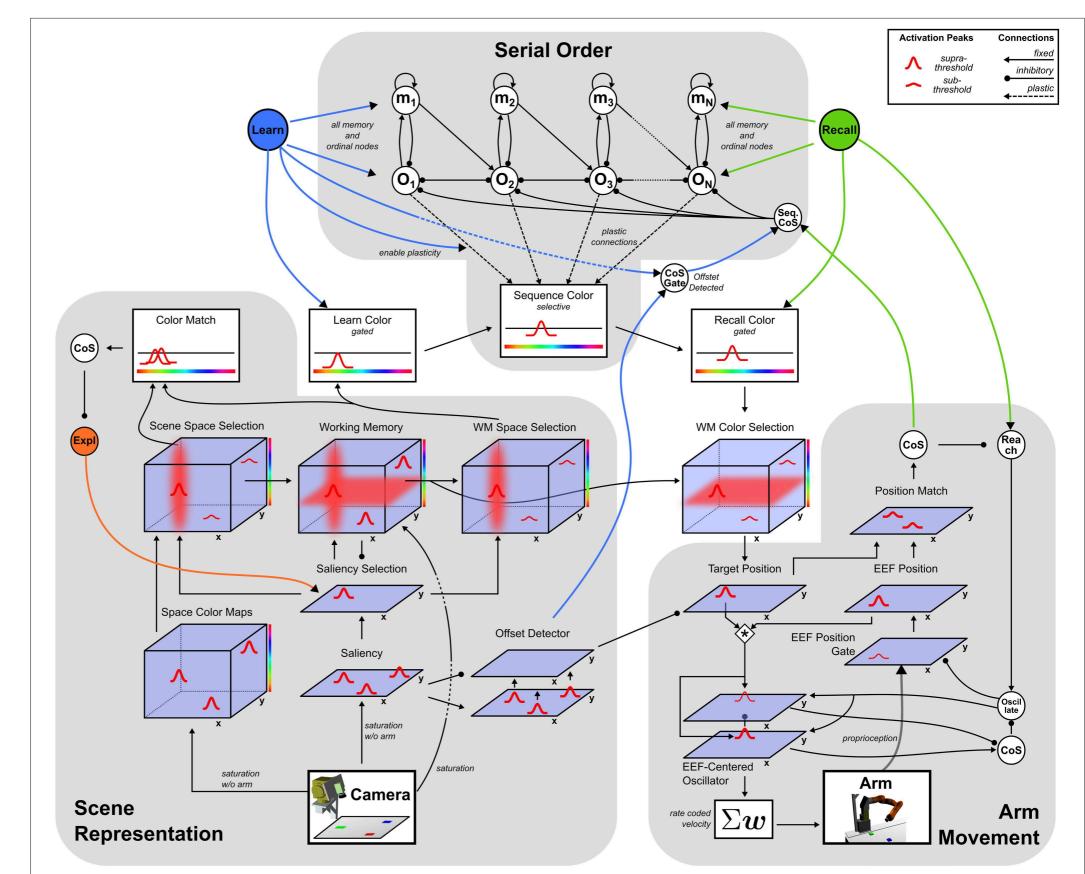
good for fast learning of sequences...

e.g. imitation

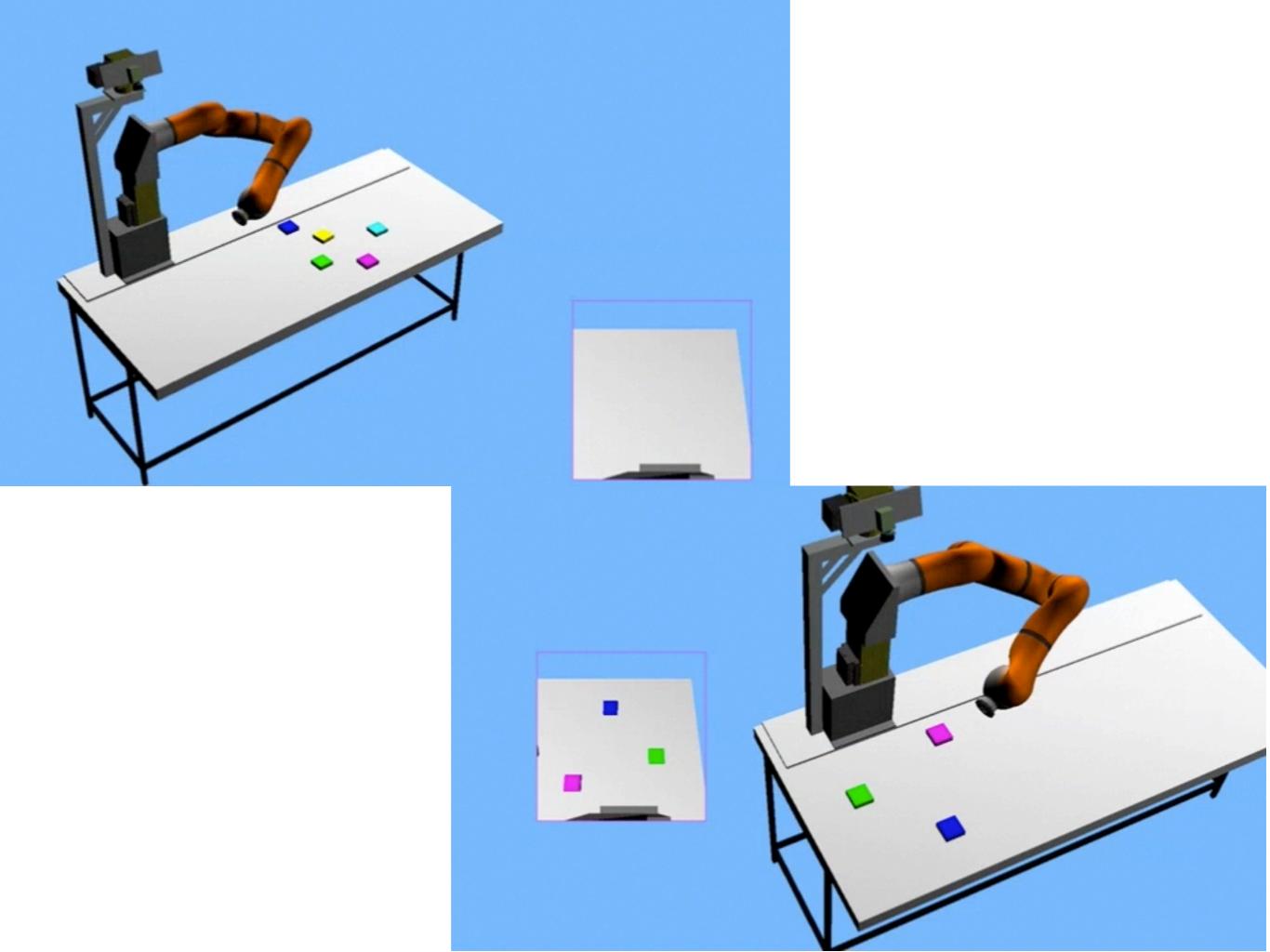
a Hippocampus function?

- but: must have potential synaptic links to many representations...
- => such ordinal systems must exist for subrepresentations... embodiment effects...

Serial order demonstrated/enacted



[Tekülve et al., Frontiers in Neurorobotics (2019)]



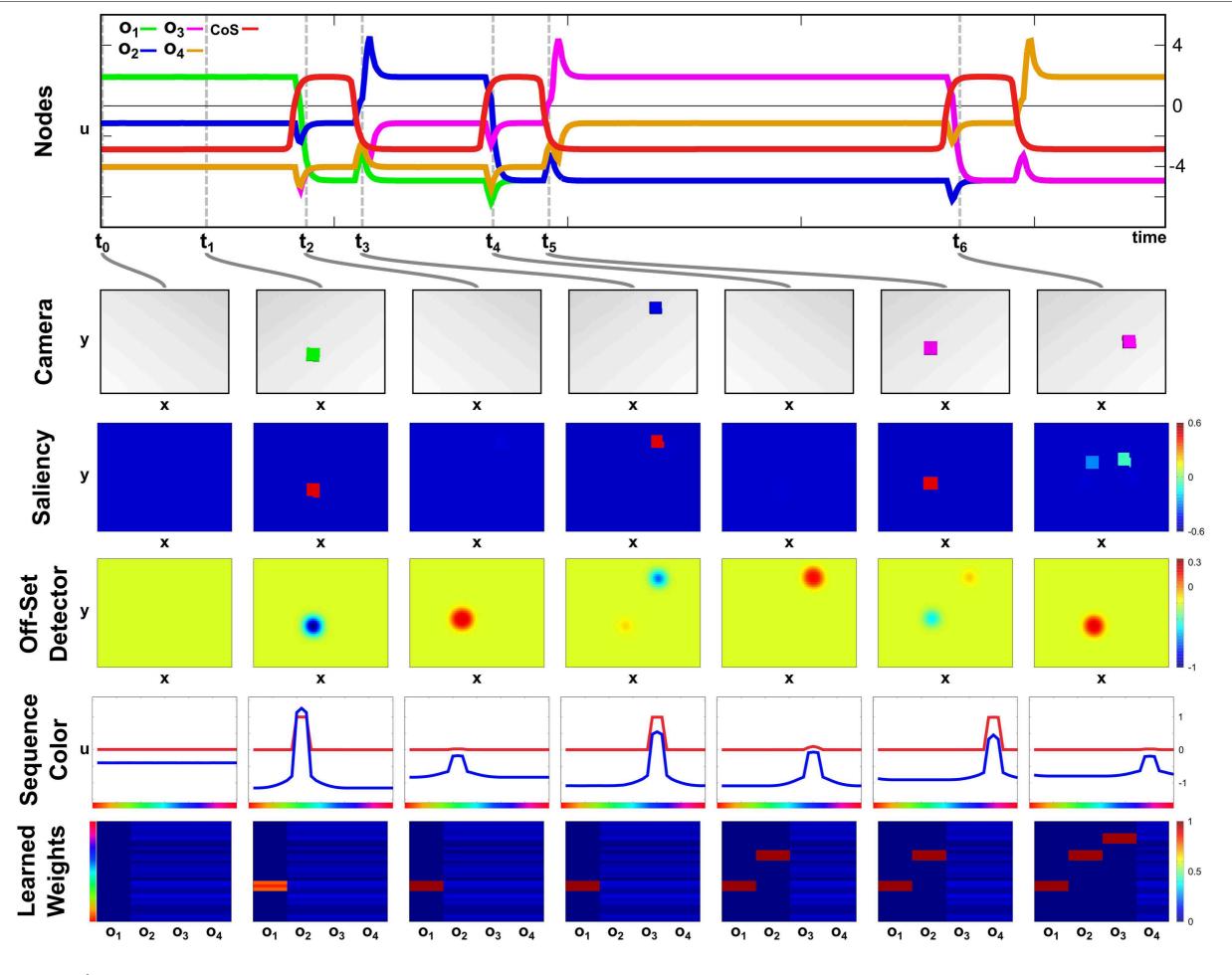


FIGURE 5 | Time course of learning a three element sequence with varying presentation time.

Time course of attention selection and building of scene memory

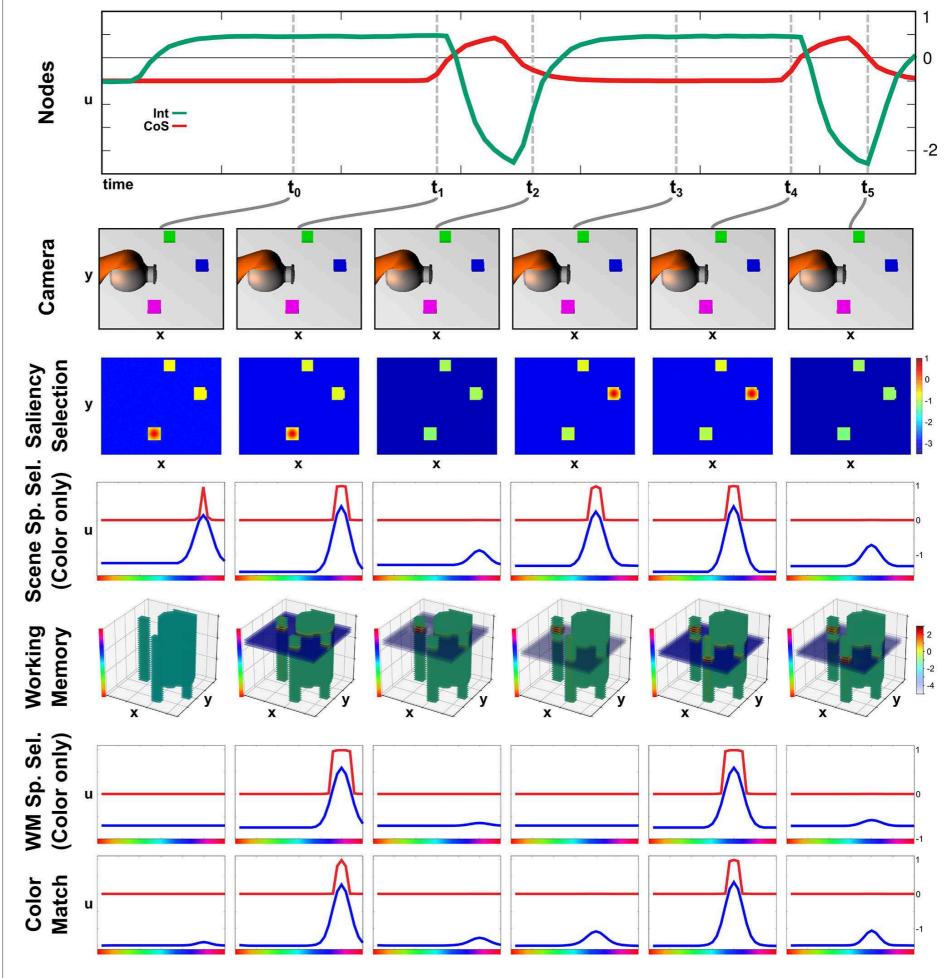


FIGURE 4 | Time course of building a scene memory.

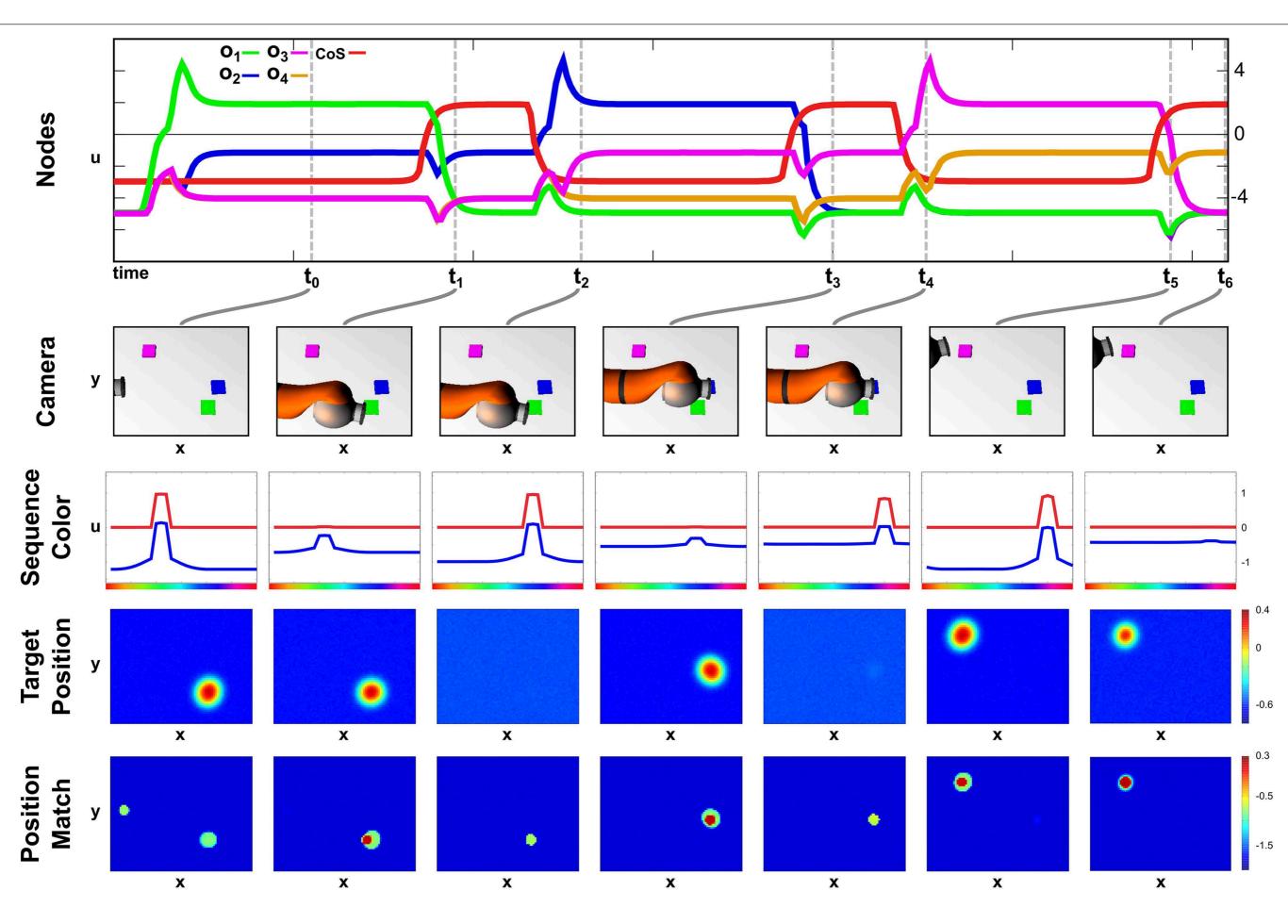
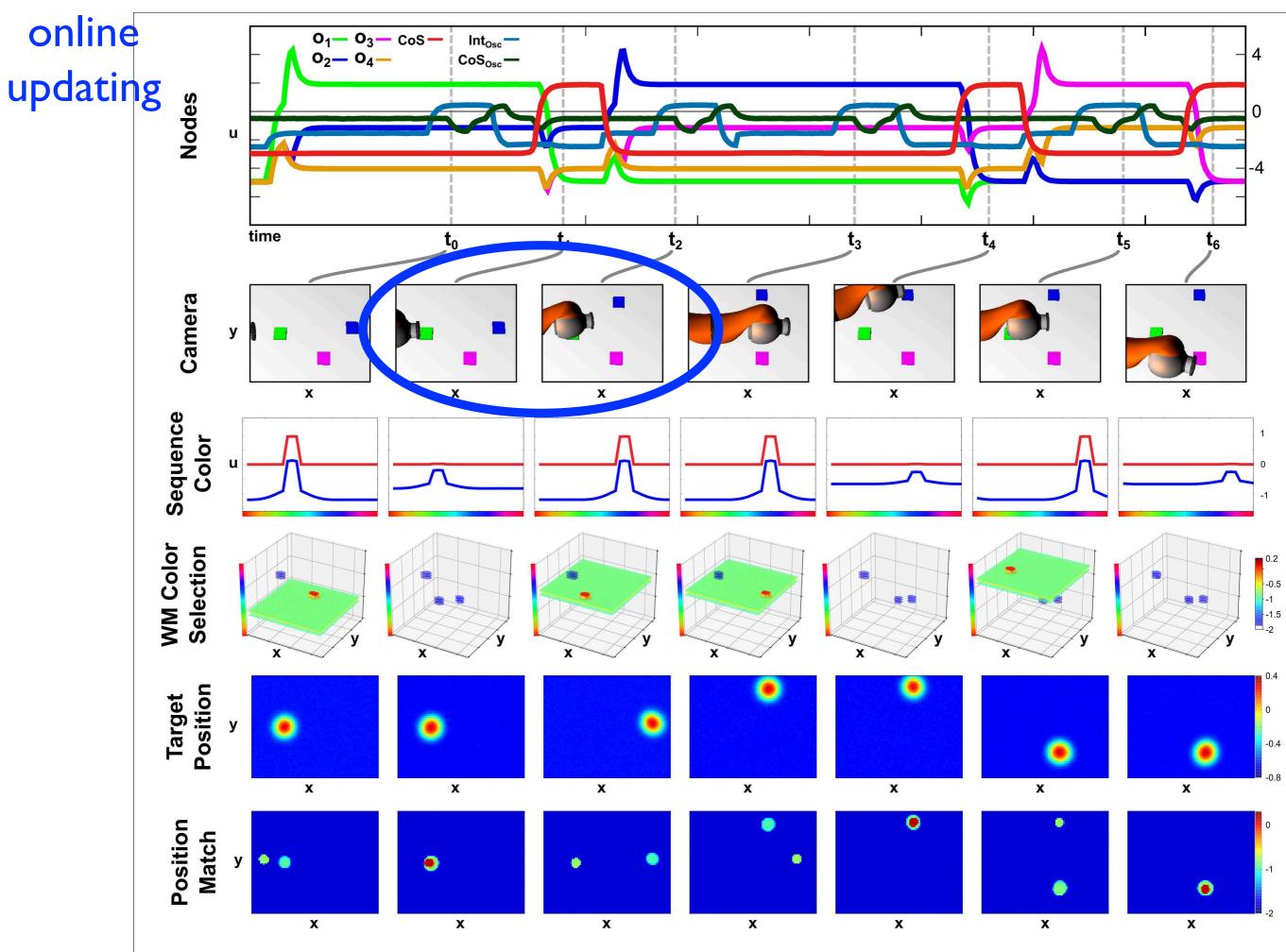


FIGURE 6 | Time course of recalling a three element sequence through pointing at colored objects.



Conclusion

the principles of DFT

localist representations form stable states

that may made unstable in a controlled way

through the "condition of satisfaction"

enable the autonomous generation of sequences of mental motor states

=> a fundamental first step toward higher cognition