

Neural Mechanisms of Hierarchical Planning during Navigation

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Reinforcement Learning (RL) is a general framework of Machine Learning that also describes mammalian behaviour $^{\rm 1}$.

Hierarchical Reinforcement Learning (HRL) ² is an extension of RL for temporal abstraction 3,4

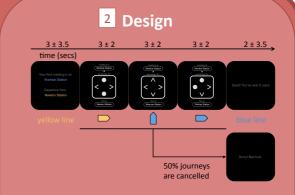
To achieve this, HRL exploits the concept of options (or contexts), each defined by its subgoal state.

Prediction error signals for interim goals have been observed in the striatum ⁵, but it remains unknown whether representations of contex

We predicted that subgoal states would be associated with activation in Anterior Cingulate Cortex (dACC), a region previously associated with task-switching ⁶ and tracking the reward value of alternative contexts 7.

Methods

N = 19 participants (age 00±00, 00 males) performed a navigation task while undergoing fMRI. There was a preliminary training session.

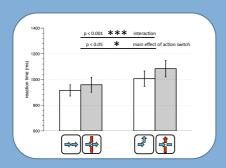


Participants navigated in a virtual subway network (stations and lines) with which they were familiar. The instruction was to minimise the length of their journey. The map was not shown during navigation. Possible directions (actions) were North/South/East/West.

Regular, exchange and elbow stations allowed us to disentangle between effects due to number of directions, direction switch and line switch.

3 Results

Behavioural



Neural



dACC was significantly more activated in exchange stations than in regular stations.



interaction **₽** Critically, this activation was strongest when

4 Conclusions

This results provide evidence that a unique network is activated when participants reach an interim goal during navigation. This network bears a striking resemblance to that previously implicated in response inhibition during speeded decisions⁸. **The dACC**, which forms a prominent part of this network, may be key for **switching to a new context** in pursuit of reward⁷.

Our findings suggest that

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