



# Active Inference, homeostatic regulation and adaptive behavioural control



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## ABSTRACT

We review a theory of homeostatic regulation and adaptive behavioural control within the Active Inference framework. Our aim is to connect two research streams that are usually considered independently; namely, Active Inference and associative learning theories of animal behaviour. The former uses a probabilistic (Bayesian) formulation of perception and action, while the latter calls on multiple (Pavlovian, habitual, goal-directed) processes for homeostatic and behavioural control. We offer a synthesis these classical processes and cast them as successive hierarchical contextualisations of sensorimotor constructs, using the generative models that underpin Active Inference. This dissolves any apparent mechanistic distinction between the optimization processes that mediate classical control or learning. Furthermore, we generalize the scope of Active Inference by emphasizing interoceptive inference and homeostatic regulation. The ensuing homeostatic (or allostatic) perspective provides an intuitive explanation for how priors act as drives or goals to enslave action, and emphasises the embodied nature of inference.

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## Contents

1. Introduction	18
2. Homeostatic regulation and adaptive behavioural control in associative learning theories	18
2.1. A taxonomy of behavioural controllers in the associative learning literature	18
3. Homeostatic processes and adaptive behavioural controllers in Active Inference	19
3.1. Homeostatic regulation through autonomic reflexes and unconditioned responses	19
3.2. Towards more sophisticated forms homeostatic regulation: the hierarchical architecture supporting adaptive behavioural control	22
3.2.1. Pavlovian responses	22
3.2.2. Instrumental responses	23
3.2.3. Goal-directed behaviour	23
3.2.4. The interaction between controllers in the Active Inference framework	24
3.3. Learning the generative models required for hierarchical inference	25
4. Summary: motivated behaviour from the Active Inference perspective	26
5. Simplified functional anatomy of hierarchical Active Inference	29
5.1. Empirical evidence supporting the proposed framework and novel predictions	31
6. Conclusions	32
Acknowledgements	33
References	33

**Abbreviations:** CS, conditioned stimuli; US, unconditioned stimulus; PFC, prefrontal cortex; SMA, supplemental motor area; IC, inferotemporal cortex; AIC, anterior insular cortex; PMC/MC, premotor/motor cortex; ipoT, ipothalamus; ANS, autonomous nervous system; ACC, anterior cingulate cortex; PPC, posterior parietal cortex; VTA/SN, the dopaminergic ventral tegmental area and substantia nigra.

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## 1. Introduction

*The animal must respond to changes in the environment in such a manner that its responsive activity is directed towards the preservation of its existence. This conclusion holds also if we consider the living organism in terms of physical and chemical science. Every material system can exist as an entity only so long as its internal forces, attraction, cohesion, etc., balance the external forces acting upon it. [...] Being a definite circumscribed material system, it can only continue to exist so long as it is in continuous equilibrium with the forces external to it.* **Ivan Pavlov**

Current associative learning theories in psychology and neuroscience assume that animal behaviour depends on multiple forms of control (i.e., Pavlovian, goal-directed, and habitual processes). These control schemes are based on associations between stimuli, actions and outcomes and are either innate or learned through experience.

Our aim is to offer an integrative perspective by contextualizing classical formulations of adaptive behaviour within the Active Inference framework, which extends predictive coding from the domain of perception to cover action (Friston et al., 2009). Active Inference assumes that organisms act to fulfil prior expectations that encode the (evolutionarily) values of their states (e.g., having access to food). The mathematical foundation of Active Inference rests on the notion of *free energy minimization*, where the long-term average of free energy approximates the entropy of sensory states. Minimizing free energy (and therefore entropy) enables an organism to resist the dispersive effects of “external forces acting upon it” to ensure “it is in continuous equilibrium with the forces external to it” (Pavlov, 2010). Crucially, free energy can also be interpreted in a statistical sense as an approximation to Bayesian model evidence. This means Active Inference can be described normatively as maximizing (a negative free energy bound on the logarithm of) Bayesian model evidence. In other words, minimizing free energy reduces the discrepancy (e.g., prediction error) between sensations and their predictions. This discrepancy can be reduced by changing predictions – through perception – or by selectively sampling sensory inputs that were predicted – through action (Friston, 2010).

The basic premise of this article is that the ontology of behavioural paradigms in associative learning can be seen as a successive contextualisation of more elemental sensorimotor constructs, within generative models of increasing hierarchical depth. This formulation explains how the primitive sensorimotor architecture of homeostatic control – of our early evolutionary ancestors – evolved towards goal-directed and prospective forms of control. This phylogenetic progression rests on the hierarchical elaboration of more primitive architectures (Cisek and Kalaska, 2010; Pezzulo and Castelfranchi, 2009). Furthermore, this hierarchical elaboration dissolves any apparent mechanistic distinction between the optimization processes that underlie different control or learning schemes, suggesting that they are all manifestations of Active Inference – under various contexts or conditions (Friston et al., 2009). This novel hypothesis contrasts with the standard (associative learning) view that the computations underlying different forms of behavioural control are fundamentally different and appeal to different (optimization) principles.

We first review associative learning theories of homeostatic and behavioural control. We then offer an Active Inference formulation of adaptive behaviour that fulfils homeostatic imperatives in increasingly sophisticated ways – building upon peripheral (somatic and autonomic) reflexes to explain simple Pavlovian and instrumental motor responses and, finally, complex goal-directed behaviour. A crucial aspect of this hierarchical perspective is that higher-level hierarchical representations contextualize

lower levels and predict longer sequences of cues and responses. This is accommodated by predictions about transitions over increasingly protracted time scales (Friston, 2008; Pezzulo, 2012).

## 2. Homeostatic regulation and adaptive behavioural control in associative learning theories

Cannon (1929) first proposed that the evolutionary function of physiology and behaviour is to restrict homeostatic states to a physiologically tenable range. Homeostatic regulation therefore allows animals to maintain a “continuous equilibrium” between the internal milieu and environmental states, which (Pavlov, 2010) considered the *raison d'être* for our brains.

The regulation of homeostatic states – or of allostatic processes (Sterling and Eyer, 1988) – has long been described in terms of control-theoretic and cybernetic mechanisms of error cancellation and feedback control (Ashby, 1947). At the neurobiological level, one hypothesis is that homeostatic control requires *interoceptive signals* that report current homeostatic levels (e.g., the current level of glucose in the blood) (Craig, 2010; Damasio and Carvalho, 2013; Gu et al., 2013). A hungry animal can be described as an animal whose homeostatic condition departs significantly from a level that is ‘good’ for survival. With some simplifications, the ‘good’ level of glucose is used as a reference for the controller to steer action (e.g., ingest food to restore the level of glucose).

This form of autonomic regulation involves triggering autonomic reflexes that control bodily processes such as heart rate, blood pressure and peristalsis. Under some conditions, autonomic reflexes can restore homeostatic levels (e.g., a hyperthermic animal can cool down by perspiring). Although elemental, autonomic reflexes are not sufficient to fully support homeostasis: to satisfy hunger or thirst, the animal must act on the external world. Early theories of homeostatic regulation focused on simple (e.g., approach or avoidance) constituents of an innate behavioural repertoire. However, higher animals learn to achieve their goals in complex and flexible ways that go well beyond approach and avoidance. To do this they must acquire an adequate behavioural repertoire and learn to select from currently available actions or sequences of action (policies). This is the main focus of associative learning theories in psychology and biology.

### 2.1. A taxonomy of behavioural controllers in the associative learning literature

Contemporary associative learning theories assume that action selection depends on the continuous cooperation and competition of several behavioural controllers, which can be divided into “Pavlovian” and “instrumental” (goal-directed and habitual) (Balleine and Dickinson, 1998; Daw et al., 2005; Dayan, 2009).

*Behavioural reflexes* represent a basic form of controller that constitutes the innate repertoire of most animals. This controller is rather limited, as it calls on a limited set of *unconditioned responses* (e.g., approaching and ingesting food or withdrawing from a painful stimulus) in response to a circumscribed and predefined class of stimuli (called *unconditioned stimuli*). Still, this controller is sufficient for most animals to survive, even without any experience-dependent learning.

*Pavlovian (classical) conditioning* is the process by which an unconditioned stimulus (say, food), that triggers an unconditioned reflex (say, salivation), is repeatedly paired with a neutral stimulus (say, a bell). The pairing is thought to produce stimulus-stimulus associations (between the food and the bell). After several pairings, the (formerly) neutral stimulus is able – on its own – to trigger a reflex called a *conditioned response* (salivation when the bell rings). An evolutionary imperative for acquiring a conditioned response is that the stimulus-stimulus associations capture (ecologically)

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