

Solving Complexity & Search

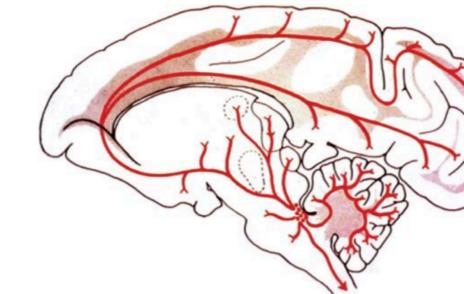
2025年1月25日 星期六 11:56

LC-NE System Solves the Best Computation Issue

Adaptive is the key to the Challenging World.

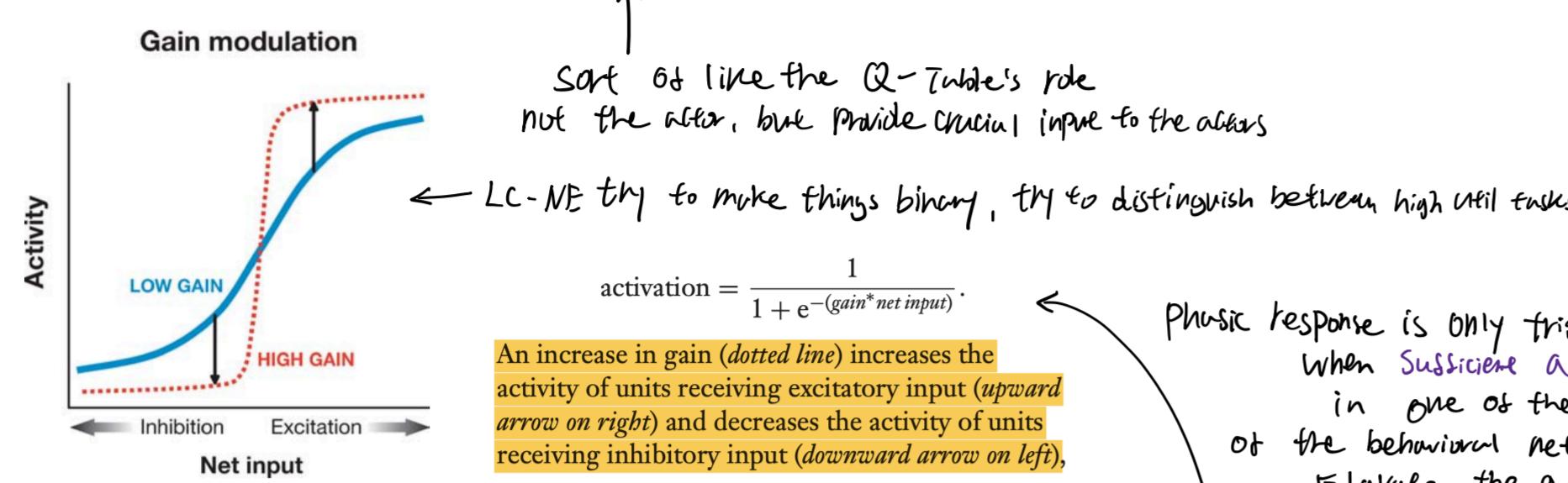
How to balance the trade off between **exploiting** known source of reward
and **exploring** potentially more valuable targets

The LC-NE system is widely distributed and ascending projections all the way to Neo Cortex



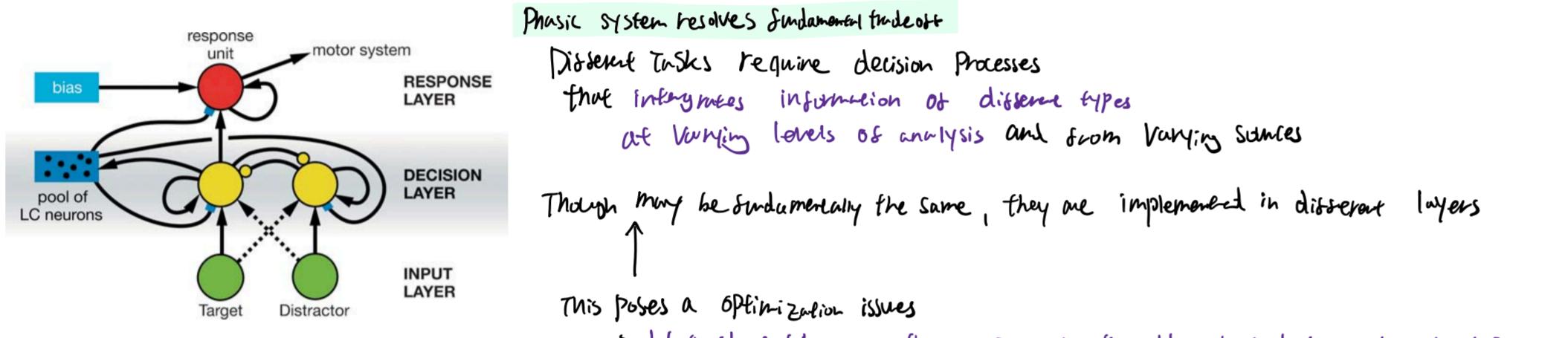
Such system exhibits 2 types of working mode, which has been thought to solve this problem of exploitation vs. exploration

LC-NE System is responsible for ongoing evaluation of task utility, provided by the input from frontal structure



① **Phasic**: Driven by task-relevant decision process
→ Ensuring behaviors to optimize task performance
→ only aroused by highly salient signals → NE release → NE is found in Signals in all sensory brain areas
→ LC is the brain's adrenal gland, augmenting the process of motivating relevant stimuli
→ Really high attention focus, hard to be distracted (like me reading this now)

② **Tonic**: When no task is at hand
→ Disengagement from the current task and search for alternative behaviors
→ Baseline is elevated but not bursting



Tonic System Solves Adaptability

On the other hand, tonic activity ensures that we are also sensitive to task irrelevant stimulus

From this perspective, optimization involves not only determining how to best perform the current task, but also considering its utility against alternative courses of action and pursuing these if they are more valuable. This is, of course, a more complex and less well-defined problem, which presents significant challenges to formal analysis. Reinforcement

→ Classical Conundrum faced in RL, do we sample more or commit more

In RL we deal with this issue by regulating the amount of random behaviors
→ Annealing in thermodynamics (melt metal slowly cool down to ensure achieving good thermal dynamics optimal equilibrium)

However, this is not adaptive to the environment

LC solves this: Adaptive gain theory modulated by OFC/ACC:

adaptive gain theory: Increased baseline release of NE increases the gain of units in the network indiscriminately, making them more responsive to any stimulus. This uniform increase in responsiveness is tantamount to increasing noise and favoring exploration. The

- If there are high tonic firing:
Phasic firing must be really really strong for it to have a good effect
(Kicking into exploitation)
- Intermediate tonic level tends to help "push" exploitation a bit
(When you explore moderately, You can exploit really good)

From a Bayesian modeling perspective, tonic system can also determine whether a failure of prediction reflects variability inherent in the task OR an underlying change in the environment
↑
Mediated by Ach
↑
Revision of expectations, mediated by increase baseline NE firing

All of the above would only work base on the assumption that they are responsive to such performance evaluative information from the Frontal Cortex Projecting to LC

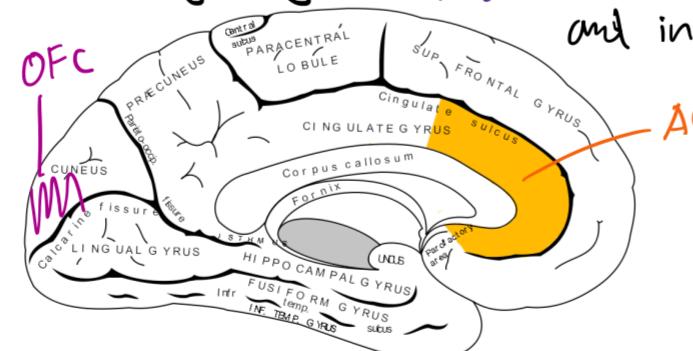
Determinants of Errors

Both Orbital Frontal Cortex and Anterior Cingulate Cortex have been finding having great projection upon LC-NE system

OFC plays a role in the evaluation of reward.
The OFC receives input from all modalities of higher-order sensory cortices, in particular areas processing information with strong appetitive significance, such as taste and olfaction, as well as primary limbic structures such as the ventral striatum and amygdala (Baylis et al. 1995; Carmichael et al. 1994; Carmichael & Price 1995a,b; Ongur & Price 2000; Rolls

and ins- serve or the hole for both

1. Giving Evaluation outcome and init Phasic mode
2. Regulation of the balance in LC-NE system



The evaluation need to incorporate both long / short term utility (Computed in ACC and OFC)

by a transition to the LC tonic mode. Importantly, the determination of when to promote exploration over exploitation requires that evaluative mechanisms take account of both short- and long-term changes in utility. There are many ways of doing so. The following equation describes one simple means (shown graphically in Figure 10):

$$\text{Engagement in current task} = [1 - \text{logistic(short term utility)}] * \text{logistic(long-term utility)}.$$

(Equation 1)

where logistic refers to the sigmoid function $1/(1 + e^{-\text{utility}})$, and high values of the equation favor the LC phasic mode, whereas low values favor the tonic mode.

High Long-term but Low short-term (Phasic / Exploit)

High Long-term and high short-term (Phasic / exploit while tonic / explore)

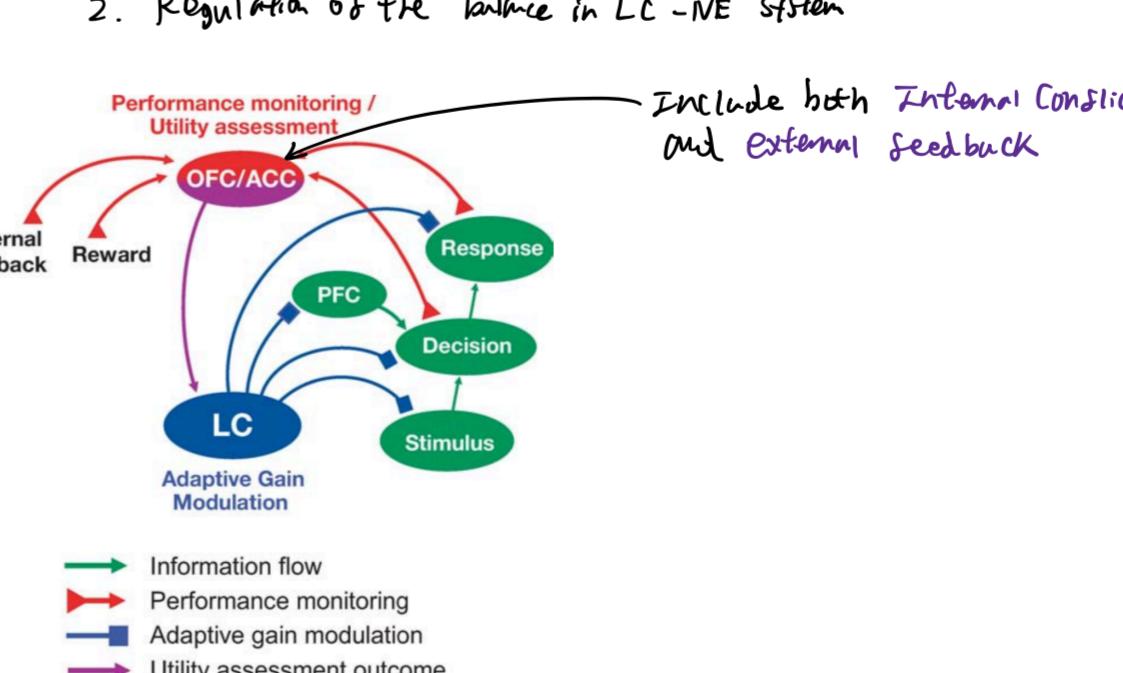
High short-term + Low Long-term Effect
Try to explore a little bit more (Tonic)

All low long term → Then Let's just exploit more !!!

It's a Continuum, once Phasic starts to take over
on long term utility modeling → Tonic starts kicking in more

Raising baseline, so when next Phasic comes, it must be really high reward / low cost
for it to kick out of the tonic mode

(Can we design an algorithm where ↑ in phasic ↑ in tonic?)



Activation of binary not only in LC system but also existed in a layer of abstraction above when deciding the signal output to LC-NE

Figure 10
Plot of the relationship between engagement in the current task and task-related utility integrated over relatively brief (e.g., seconds) and longer (e.g., minutes) timescales given by Equation 1 (see text). The adaptive gain of the LC phasic mode favors the LC mode, whereas low values favor the tonic mode. Accordingly, low values of long-term utility favor the LC phasic mode (exploitation), whereas high values favor the LC-tonic mode (exploration). Note that when long-term utility is high, a decrease in short-term utility augments the LC phasic mode, implementing an adaptive adjustment that serves to restore performance.

It's a Continuum, once Phasic starts to take over
on long term utility modeling → Tonic starts kicking in more

Raising baseline, so when next Phasic comes, it must be really high reward / low cost
for it to kick out of the tonic mode

(Can we design an algorithm where ↑ in phasic ↑ in tonic?)