

# A neural coarse graining theory of consciousness

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## 2 ABSTRACT

3 Neural systems process information through different levels  
4 of organisation in a hierarchical manner. Information at  
5 lower levels is finer-grained and can be coarse-grained  
6 for higher level computation. However, one is aware of  
7 information processed only at specific levels. Theorists have  
8 addressed this issue. For example, the intermediate level  
9 theory of consciousness suggests that the intermediate level  
10 seems to be privileged with respect to consciousness. It is  
11 true that we do not experience information processed by  
12 individual neurons which is always highly noisy. Besides, we  
13 have no conscious experience from interpersonal activities  
14 albeit massive interactions among individuals. Instead,  
15 neurophysiological evidence has been showing that conscious  
16 experience tends to covary with information encoded in  
17 coarse-grained neural states such as neural population codes.  
18 We argue that the neural states within the scope of the  
19 information closure determine the contents of consciousness  
20 and brain processes outside apart from the representations  
21 of that level remain unconscious. This argument suggests a  
22 distinction between conscious and unconscious processing  
23 and provides a generic computational framework. Finally,

24 using the deep learning network, we can measure information  
25 closure in deep hidden layers. Our preliminary results  
26 show that information closure representation emerged after  
27 learning. We further decoded the information from the  
28 representation and compared it with human conscious  
29 perception.

30 **Keywords:** theory of Consciousness, information closure, neural coarse-graining,  
31 level of analysis, keyword, keyword, keyword, keyword

## 1 INTRODUCTION

## 2 NEURAL COARSE-GRAINING

32 \* *"Neural population code: the set of response features of*  
33 *a population of neurons that carry all information about the*  
34 *considered stimuli. These features consist of spatio-temporal*  
35 *sequences of action potentials distributed across neurons and/or*  
36 *time."*

37 *¿ The diverse response selectivity of sensory neurons ¿ "How a*  
38 *neural population represents information is partly determined by the*  
39 *diverse selectivity of individual neurons Shamir (2014)"*

40 Stanley (2013) Quian Quiroga and Panzeri (2009)

41 \* Sensors can only receive partial information \* Through coarse-  
42 graining, higher level cortical areas can integrate those partial  
43 information to infer hidden causes.

### 44 2.1 The advantage of NCG

45 2.1.1 Resist to noise

46 2.1.2 Operate on deterministic and abstract level

47 2.1.3 Predictive power

48 2.1.4 Reduce energy cost

### 3 NON-TRIVIAL INFORMATION CLOSURE

49 *"Our theoretical interest concerns the type of system that is a*  
 50 *unity for and by itself and not only for an external observer*  
 51 *distinguishing some entity from the rest of the world. This requires*  
 52 *a system that can be described as a whole without reference to its*  
 53 *environment. In systems theory, this property is usually referred to*  
 54 *as closure."*(BERTSCHINGER et al., 2006)

55 *"These concepts of closure play an important role in the*  
 56 *architecture of systems theory, because they are used to 1. define*  
 57 *the system (in distinction to its environment) and to 2. explain the*  
 58 *autonomy of the system. "*

59 Informational closure: The higher process is informationally  
 60 closed, i.e., there is no information flow from the lower to  
 61 the higher level. Knowledge of the microstate will not improve  
 62 predictions of the macrostate (PFANTE et al., 2014, p. 4).

63 *"A system that is independent from its environment trivially*  
 64 *achieves informational closure."* (BERTSCHINGER et al., 2006)

#### 65 3.1 Mathematical definition of information closure

66 *"The notion of informational closure refers to a situation where*  
 67 *the information flow between the environment and the system tends*  
 68 *to zero. "*

69 BERTSCHINGER et al. (2006) suggests that information closure  
 70 can be defined as the degree of information flow between  
 71 environment and the system.

$$\begin{aligned}
 J_n(E \rightarrow S) &= MI(S_{n+1}; E_n | S_n) \\
 &= H(S_{n+1} | S_n) - H(S_{n+1} | S_n, E_n) \\
 &= H(E_n | S_n) - H(E_n | S_n, S_{n+1}) \\
 &= H(E_n | S_n) - H(E_n | S_n, S_{n+1})
 \end{aligned} \tag{1}$$

72 **3.2 Non-trivial information closure**

73 Information closure could be trivial. That is, when a system is fully  
74 independent from the environment, no information can flow into the  
75 system from the environment and also leak to the environment from  
76 the system.

77 Equa. xxxx shows the mathematical description of trivial  
78 information closure. When mutual information between the  
79 environment and system future state close to 0, and the system  
80 transition is independent from the environment, the system can  
81 reach informational closure. However, such systems do not have any  
82 functional meaning and evolutionary advantages.

83 Therefore, it's important to ensure the informational closure that the  
84 system reach is non-trivial.

85

86 [[ cite 2006 and nic's paper ]]

87 To achieve non-trivial information closure,

88 (BERTSCHINGER et al., 2006)

89 (Guttenberg et al., 2016)

90 **3.3 how to achieve information closure**

91 *"This demonstrates that a system exhibiting certain internal*  
92 *regularities as measured by  $A^* = MI(S_{n+1}; S_n)$  can*  
93 *achieve informational closure either by gaining information about*  
94 *the environment or by increased autonomy, i.e. by becoming*  
95 *unpredictable or uncontrollable from the (13) environment.*  
96 *Therefore, information about the environment, i.e. modeling, and*  
97 *autonomy can be considered as complementary strategies for*  
98 *achieving informational closure."*

99 **3.4 information closure and Level Identification**

## 4 MY CLAIM OF CONSCIOUSNESS

100 \* We claim that the coarse-grained state in the information closure  
101 determines contents of consciousness. \* It's important that not the  
102 neural states but the coarse-grained state determine contents of  
103 consciousness.

### 104 4.1 Information and system theory

105 [[ Closure can define a system, consciousness is information ]]  
106 [[ Could be linked to system theory, but may need help ]]  
107

108 *"This self-referential distinction from its environment therefore*  
109 *gives rise to the specific autonomy of such a system. Consequently,*  
110 *in systems theory, closure properties and autonomy are considered*  
111 *to be closely related concepts which are both at the heart of defining*  
112 *the system itself. "*

## 5 BIOLOGICAL EVIDENCE OF INFORMATION CLOSURE IN NEURAL SYSTEM

113 Sederberg et al. (2018)

## 6 RELATION TO EMPIRICAL FINDINGS

114 6.1 Explain normal and abnormal  
115 conscious/non-conscious experience

116 6.2 Deterministic vs probabilistic

117 Dehaene et al. (2017) Vul et al. (2008); Moreno-Bote et al. (2011);  
118 Asplund et al. (2014); Vul et al. (2009)

## **7 COMPARISON WITH OTHER THEORIES**

119 **7.1 Intermediate Level Theory**

120 **7.2 Global Workspace**

121 7.2.1 Ignition in GWT

122 [[ Explain ignite ]]

123

124 **7.3 Sensorimotor contingency**

125 **7.4 IIT**

126 **7.5 Internal simulation and self-modeling**

## **8 COUNTERINTUITIVE PREDICTION**

## **9 EVOLUTION OF CONSCIOUS MIND**

127 Dennett (2008)

## **10 CONCLUSION**

128 NCC: To find NCC, it's not about where and when, it's about scale  
129 and level of discription

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## **SUPPLEMENTAL DATA**

136 Supplementary Material should be uploaded separately on  
137 submission, if there are Supplementary Figures, please include the  
138 caption in the same file as the figure. LaTeX Supplementary Material  
139 templates can be found in the Frontiers LaTeX folder.

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## FIGURE CAPTIONS



**Figure 1.** Enter the caption for your figure here. Repeat as necessary for each of your figures