# In search of a measure of organisedness for dynamical systems

Silvere Gangloff (LIP, ENS Lyon)

24 avril 2019

"Symbolic dynamics as an exploration field"

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 Clinical: detecting consciousness (ex: patients in coma state).

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#### **Motivations:**

- 1. **Clinical**: detecting consciousness (ex: patients in coma state).
- 2. Mathematical: sharper distinction of conjugation classes:

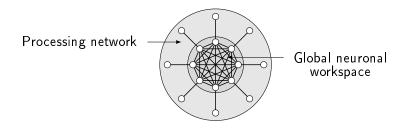
$$C[f] = \{g : \exists \sigma : g = \sigma \circ f \circ \sigma^{-1}\}.$$

#### I. Main theories of consciousness

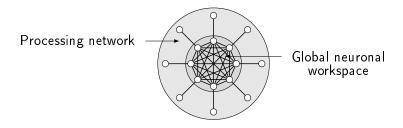
- 1. Global neuronal workspace
- 2. Integrated information theory

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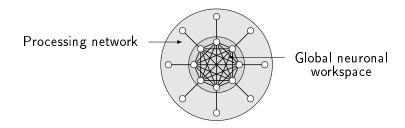


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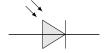
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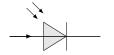


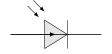
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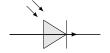
A cognitive theory of consciousness, B.J. Baars, 1988.

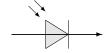






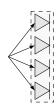


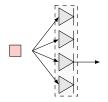


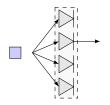


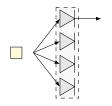


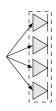
The knowledge of the output generates 1bit of information.



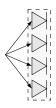






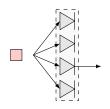


Discriminate more inputs

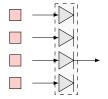


Discriminate more inputs  $\Rightarrow$  generates more information.

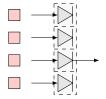
**2. Integrated information theory :** Non-integration of information : decomposition into independant smaller mechanisms :



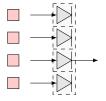
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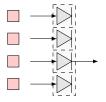


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In the human brain: balance between generation of information and integration.

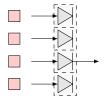
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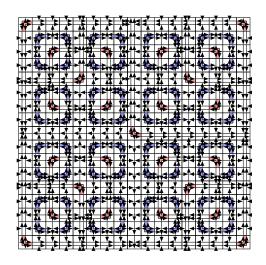
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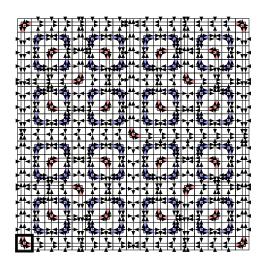
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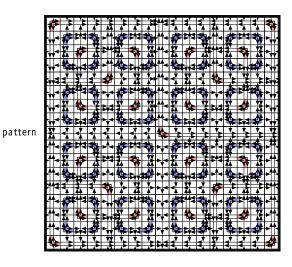
Question: quantity which measures this balance?

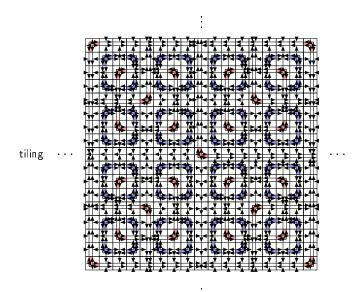
# II. Symbolic dynamics and universal computation

1. Hierarchical aperiodic tilings and further constructions
2. Minimal tile sets generating aperiodic tilings

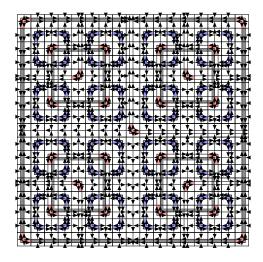




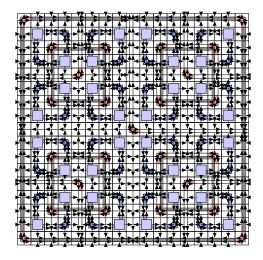




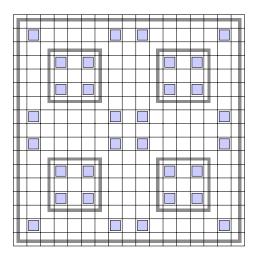
#### Computation boards:

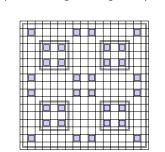


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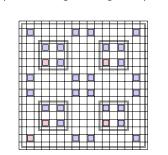


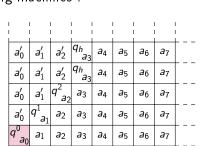
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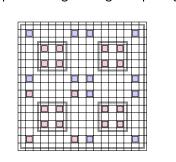


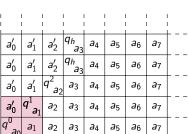


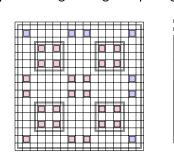


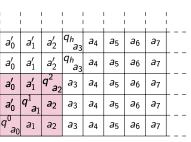


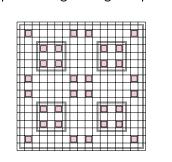


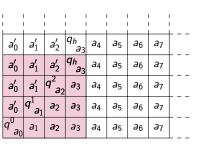












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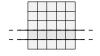
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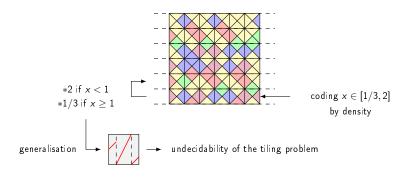
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3. B. Durand, A. Romaschchenko, A. Shen: **fixed-point** tilings constructions.

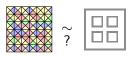
$$\boxed{ au} \quad \boxed{ au'} \equiv$$

#### Small aperiodic tile sets : Kari-Culik (13) :



Minimal: Jeandel-Rao (11).

#### Question:

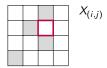


**No**: entropy of Kari-Culik > 0, and 0 for Robinson [Aperiodic tilings and entropy, B. Durand, G. Gamard, A. Grandjean].



# III. A starting point : neural complexity and intricacies

#### Neural complexity of a finite family of random variables :



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Synchronous or independant :  $\mathcal{N}(X) = 0$ .

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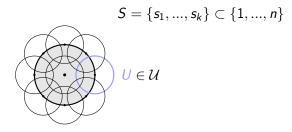
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- (Number of) maxima of intricacy?



$$\mathcal{I}(X, f, \mathcal{U}, c) = \lim_{n} \sum_{S \subset \{1, \dots, n\}} c_S^n \log_2 \left( \frac{|\mathcal{U}_S| \cdot |\mathcal{U}_{S^c}|}{|\mathcal{U}_{\{1, \dots, n\}}|} \right).$$

$$t=s_1$$
  $S=\{s_1,...,s_k\}\subset\{1,...,n\}$   $U\in\mathcal{U}$   $U_{s_1}$ 

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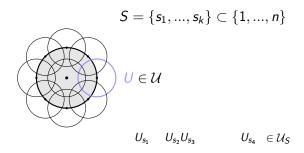
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#### One dimensional tiling sets: [Petersen, Wilson 2016]

$$X = \begin{bmatrix} 0 & 1 \\ 1 & 0 \end{bmatrix} \square$$

When  $M^2 > 0$ :

$$\mathcal{I}(X) = \sum_{k=1}^{+\infty} \frac{\log_2(N_k(X))}{k} - h(X),$$
$$h(X) = \lim_k \frac{\log_2(N_k(X))}{k},$$

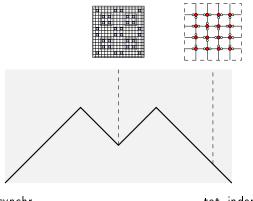
$$h(X) = \lim_{k} \frac{32(x+x)^{2}}{k}$$

where  $N_k(X)$ : number of size k observable patterns.

## IV. Symbolic dynamics as an exploration field

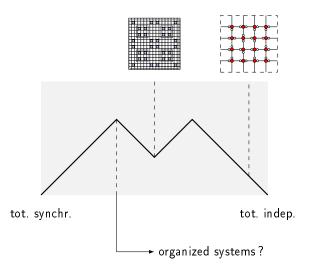
- 1. Strategy
- 2. A pool of existing formalisms
- 3. Organisedness and dynamical constraints

#### 1. Strategy: search a quantity with tractable maxima such that:



tot. synchr. tot. indep.

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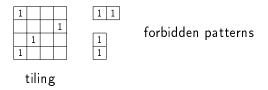
Kolmogorov complexity of sequences:



Kolmogorov complexity of sequences :

## Minimal number of forbidden patterns:

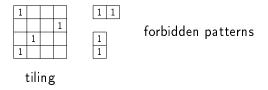
Ex: hard core model:



Kolmogorov complexity of sequences :

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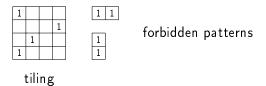
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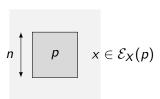
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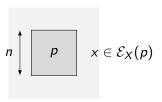


In practive, maxima difficult to apprehend.

### Number of extender sets: X tiling set

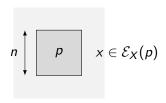


#### Number of extender sets: X tiling set



Set of extender sets of  $X : \mathcal{E}_n(X) = \{\mathcal{E}_X(p) : p \in \mathcal{L}_n(X)\}.$ 

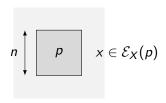
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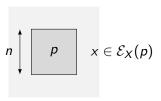


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Extender sets  $\equiv$  measure of how a pattern constrains a tiling.

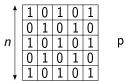
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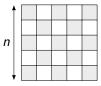


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Extender sets  $\equiv$  measure of how a pattern constrains a tiling. Principle of organisation : a lot of patterns exert a non-trivial constraint.





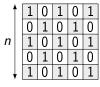












$$n = \begin{bmatrix} 1 & 0 & 1 & 0 & 1 \\ 0 & 1 & 0 & 1 & 0 \\ 1 & 0 & 1 & 0 & 1 \\ 0 & 1 & 0 & 1 & 0 \\ 1 & 0 & 1 & 0 & 1 \end{bmatrix}$$

$$S_n(X, p) = \{S \subset [1, n]^2 \text{ minimal } : p \text{ constructible from } S\}.$$

$$n = \begin{bmatrix} 1 & 0 & 1 & 0 & 1 \\ 0 & 1 & 0 & 1 & 0 \\ 1 & 0 & 1 & 0 & 1 \\ 0 & 1 & 0 & 1 & 0 \\ 1 & 0 & 1 & 0 & 1 \end{bmatrix}$$

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Related complexity notion:

$$\mathcal{S}_n(X) = \{\mathcal{S}_n(X,p) : p \in \mathcal{L}_n(X)\}.$$

$$S_n(X,p) = \{S \subset \llbracket 1,n 
rbracket^2 \mod S\}.$$

 $Related\ complexity\ notion\ :$ 

$$S_n(X) = \{S_n(X, p) : p \in \mathcal{L}_n(X)\}.$$



### 3. Organisedness and dynamical constraints:

Expressive power of multidimensional tiling sets under constraints : [G.,Sablik]

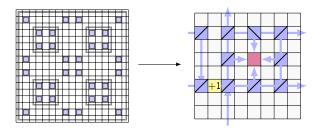
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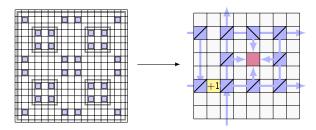


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#### Functional segregation:



Analysis of this kind of phenomenon in relation with organisedness?

