Vectorial data use in a m:n-AC^k cellular automaton.

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GIS data classification used is shown in the next table:

<table>
<thead>
<tr>
<th>Layer</th>
<th>GPS integration</th>
<th>Description</th>
</tr>
</thead>
<tbody>
<tr>
<td>2DLayers</td>
<td>Geo referenced</td>
<td>Fixed population of elements over a matrix, and DEM.</td>
</tr>
<tr>
<td>Routes</td>
<td>Track points</td>
<td>Represent Objects movement.</td>
</tr>
<tr>
<td>2DObjects</td>
<td>Waypoints</td>
<td>a 2D object in an specific position</td>
</tr>
<tr>
<td>3DObjects</td>
<td>Waypoints</td>
<td>a 3D object in an specific position</td>
</tr>
</tbody>
</table>

In a simulation environment suitable data that can be represented in the m:n-AC^C layers are vectorial data (2DLayers) or raster data (3DLayers). Other elements (Routes, 2DObjects or 3DObjects) can be represented using common simulator elements. A two dimensional cellular automaton is represented by a 1:2-AC.

Example: 1:2-AC on Z^2

In this example GIS data in vectorial format must be used to evaluate a propagation of “something”. The data, represented in IDRISI format is shown in the next figures.

Example: 1:2-AC on Z^2
State definition E_1, and evolution function are the same that over R^2, but due to the vicinity and nucleus topology works in Z^2; vn and nc functions must be redefined.

11 iterations 1:2-AC on R^2
11 iterations 1:2-AC on Z^2

Conclusions
The m:n-CA^k cellular automaton is presented, a generalization of classical cellular automaton without dependency on implementation used to represent it.

This extension of cellular automaton allows vectorial GIS data use inside a discrete simulation model and the use of different layers in a single cellular automaton. This simplifies GIS data use in a complex simulation model due to all the GIS data can be represented in a single structure.

A comparison for the same evolution function is presented using different topological spaces. The two evolutions helps to remark first the independence of this new theoretical approach with the implementation method used and second, the importance of the topological space used for the results accuracy.

Cellular automata are discrete dynamical systems whose behavior is completely specified in terms of a local rule (Emmrich C., 1980). Cells represent automaton space; time advances in discrete steps following “the rules”, the laws of “automaton universe”, usually expressed in a small look-up table. At each step every cell computes its new state in function of its closer neighbours. Thus, system’s laws are both local and uniform. Figure shows one-dimensional cellular automaton initial state and successive two states after rules application.

Theory: m:n-AC^k
A multi n dimensional cellular automaton is a cellular automaton generalization composed by m layers of dimension n.

m:n = AC^k

Where
- m is the automaton number of layers.
- n is the layers dimension.
- k is the number of main layers (1 by default). A layer in a m:n-AC^k is a main layer if a transition function A is defined in order to modify its state. A m:n-CA automaton only presents one main layer, while a m:n=AC^k automaton presents k main layers.

Em is a function describing cell state in position x_1...x_n for layer m.

The global state of cellular automata depends on Em function in all automata positions. Combination functions Ψ allows the combination for the data that belongs to different cellular automaton layers.

In a common cellular automaton, evolution function allows global automata state change through cells value modification. In a m:n-AC^k vectorial layers use makes necessary to generalize the neighbourhood and later define a new function that determines something similar to cell size (nucleus function).

Evolution Function Λ_m
Function defined for the layer m to modify its state through the state of others layers using combination function Ψ, and vicinity and nucleus functions.

Viscosity topology
Topology defining the set of points (neighbourhood) for layer m, to be considered for Λ_m calculation.

Vicinity function vn(x_1...x_n)
Function returning minimum open set of nuclei topology containing point x_1...x_n, and including maximum points that accomplishes “the restriction” and minimum points not accomplishing “the restriction”.

Nucleus topology
Topology defining the set of points (neighbourhood) for layer m, to be modified by Λ_m calculation.

Vicinity function vn(x_1...x_n)
Function returning minimum open set of nuclei topology containing point x_1...x_n, and including maximum points that accomplishes “the restriction” and minimum points not accomplishing “the restriction”.

Nucleus function nc(x_1...x_n)
Function returning minimum open set of nuclei topology containing point x_1...x_n, and including maximum points that accomplishes “the restriction” and minimum points not accomplishing “the restriction”.

Vicinity function defines, from a position x_1...x_n, the points to be considered inside evolution function in new layer state calculus. Nucleus function allows to define, from a position x_1...x_n, the environment to be modified after evolution function calculation. To characterize vicinity and nucleus functions usually a metric must be defined, for instance the Euclidian distance:

\[ d(x,y) = \sqrt{(x_1 - y_1)^2 + (x_2 - y_2)^2} \]  

Distance d(x,y) allows the definition of neighbourhood bases as:

\[ B(x,r) = \{ y \in \mathbb{R}^n / d(x,y) < r \} \]

Restriction example: A typical vicinity and nucleus function can be defined for a point p for instance as the set containing maximum points that accomplish d(p,r)/2 and minimum points that accomplish d(p,r).

The representations for the nucleus and vicinity functions for a point p over R^2 depends on the distance selected (the green cells belongs to vn, while the yellow cells belongs to nc).

With (1) or with (2)