

Engineering Confluent Computational Fields

from Functions to Rewrite Rules

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Computational fields

A way to engineer self-organising applications in different scenarios

- Sensor networks (aggregating/diffusing data)^a
- Swarm and modular robotics (cooperative team work)^b
- Pervasive computing (crowd steering, coordinated visualisation)^c

^aJacob Beal and Jonathan Bachrach. "Infrastructure for Engineered Emergence on Sensor/Actuator Networks". In: *IEEE Intelligent Systems* 21.2 (2006).

^bJonathan Bachrach, Jacob Beal, and James McLurkin. "Composable continuous-space programs for robotic swarms". In: *Neural Computing and Applications* 19.6 (2010).

^cMirko Viroli, Matteo Casadei, Sara Montagna, and Franco Zambonelli. "Spatial Coordination of Pervasive Services through Chemical-inspired Tuple Spaces". In: *ACM Transactions on Autonomous and Adaptive Systems* 6.2 (2011).

Definition

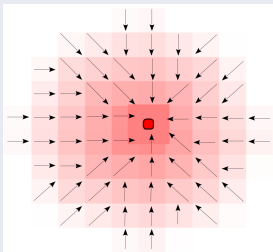
- A function ϕ mapping nodes of a (dense) network to structured values
- Defines a spatio-temporal structure situated in the (physical) world
- Paradigmatic case: a *gradient* (i.e., distance-to field)

Spreading: $x \Rightarrow (x + 1)^{\rightsquigarrow}$

Aggregation: $x + y \Rightarrow \min(x, y)$

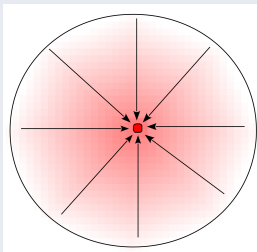
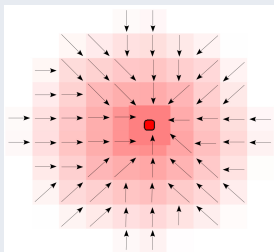
Towards a library of computational fields

Gradient (discrete, flat continuous, structured continuous)



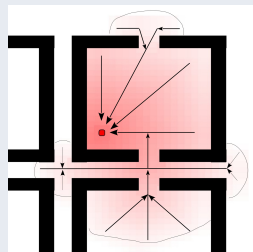
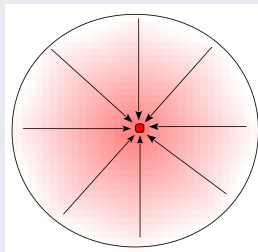
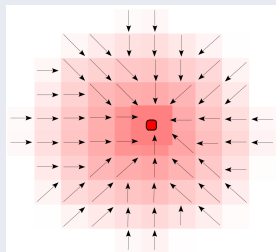
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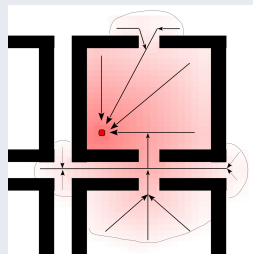
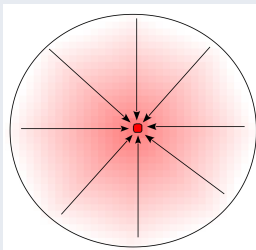
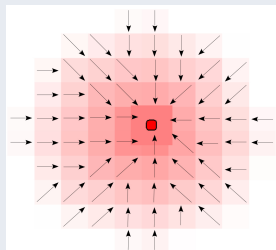
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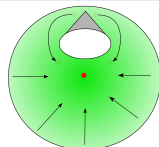
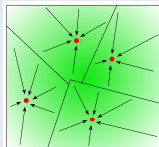
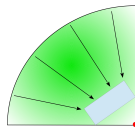
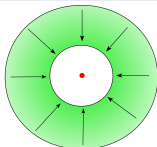
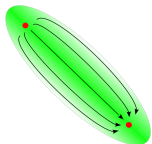


Towards a library of computational fields

Gradient (discrete, flat continuous, structured continuous)



Other structures (channel, shrinking crown, sector, partition, shadow, ...)



The quest for well-engineered computational fields

General problems we are interested in

- Find good constructs to program spreading-aggregation process
- Predict the final outcome of a field computation
- Find sufficient conditions for certain key properties – stabilisation

Two recent converging research threads

- Proto: the (functional) language of space-time
- ⇒ Isolating an expressive and tractable core of Proto^a
- SAPERE: chemical-inspired fields for pervasive computing^b
- ⇒ Seeking for “predictable” outcomes, taming nondeterminism

^aMirko Viroli, Jacob Beal, and Kyle Usbeck. “Operational Semantics of Proto”. In: *Science of Computer Programming* 78.6 (2013). Edited by Marjan Mernik, pages 633–656. DOI: [j.scico.2012.12.003](https://doi.org/10.1016/j.scico.2012.12.003).

^bSara Montagna, Mirko Viroli, Jose Luis Fernandez-Marquez, Giovanna Di Marzo Serugendo, and Franco Zambonelli. “Injecting Self-organisation into Pervasive Service Ecosystems”. In: *Mobile Networks and Applications* (2012). Online first.

Two approaches to be possibly reconciled

Functional fields (Proto): a programming language approach

```
(rep d (inf) (mux (sense 1) 0 (min-hood (nbr (+ d 1))))))
```

- High modularity and composability of specifications
- High expressiveness (up to space-time computation completeness)
- Proto programs difficult to analyse in the entire language

Chemical-like rewrite rules (SAPERE): a concurrency model approach

$$x + y \Rightarrow \min(x, y) \quad x \Rightarrow (x + 1)^{\sim}$$

- Not developed in a full-fledged programming language
- More tailored to the “coordination” aspects of spatial systems
- Seemingly more easy to analyse in terms of the structure of rules

Outline

Contributions

- Reconcile the functional and rewrite-based field approaches
- Accordingly give sufficient conditions for stabilisation (via confluence)

Progression

- 1 A core functional calculus of fields
- 2 Semantics by rewrite rules
- 3 Proving the confluence property
- 4 Implications on behaviour predictability



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Definition

Core Syntax

Let f, g be computable pointwise-applied function expressions:

$$F ::= \sigma \mid f(F_1, \dots, F_n) \mid [g(\{F\}, F_1, \dots, F_n)]$$

A field expression (or process) F is either:

- an environmental field σ (e.g. enacted by distributed sensors);
 - functional composition of fields $F_1, \dots, F_n$ ($n \geq 0$) by function f ;
 - a spreading-aggregation field $[g(\{F\}, F_1, \dots, F_n)]$:
 - originated by F
 - continuously spread to neighbours
 - locally contextualised by function $g(\cdot, F_1, \dots, F_n)$
 - reconciling multiplicity of values in a node by taking the minimum
- ⇒ rough Proto-equivalent:

```
(rep x NaN (min F (min-hood+ (g (nbr x) F1 .. Fn))))
```

Surface language

On functions

- f, g are denoted by functional expressions
- functions are partial, and can test for absence of a value
- allow non-recursive function definitions (arguments capitalised)
- values (0 – ary functions) are literals (numbers, strings) and tuples
- use built-in mathematical and tuple-like functions and operators
 $\Rightarrow +, -, *, 0, 1, >=, \text{not}, ?:, \text{tup}, \text{1st}, \text{2nd}, \text{min}$
- minimum function over tuples is custom

Examples of environmental fields

- #: constant field undefined everywhere
- #SRC, #DST: generic sensors (possibly undefined somewhere)
- #RNG: 0 on physical nodes, r on virtual nodes representing connections

Mini-tutorial

Basic spreading-aggregation fields

Assume $\#SRC$ is defined in some *sources*, where it holds 0:

$[\{\#SRC\}]$: creates a plateau of 0 out of sources

$[\{\#SRC\}+1]$: creates a hop-count gradient from sources

$[\{\#SRC\}+\#RNG]$: creates a (distance-to) gradient from sources

Restriction (if): can be used to block spreading-aggregation

$[\#OBS \ ? \ \# : \{\#SRC\}+\#RNG]$: gradient avoiding the domain of $\#OBS$
 this is different from expression $\#OBS \ ? \ \# : [\{\#SRC\}+\#RNG]$

GradCast (gradient propagating $\#VAL$ from $\#SRC$)

$[\{\langle\#SRC, \#VAL\rangle\}+\langle\#RNG, 0\rangle]$: partitioning out of $\#SRC$'s domain nodes



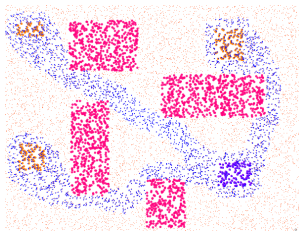
Example 1: Channel

Few definitions

```
def If(FIELD,WHERE) is WHERE ? FIELD : #
def Gossip(VAL) is [{VAL}]
def Grad(SRC) is [{SRC}+#RNG]
def Grad(SRC,OBS) is [If({SRC}+#RNG,not(OBS))]
def Dist(SRC,DST,OBS) is Gossip(If(Grad(SRC,OBS),DST))
```

Channel from #S to #D of width #W, avoiding #O

$\text{Grad}(\#S,\#O) + \text{Grad}(\#D,\#O) + \#W < \text{Dist}(\#S,\#D,\#O)$



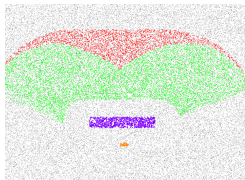
(Simulation of an equivalent model developed in Proto)



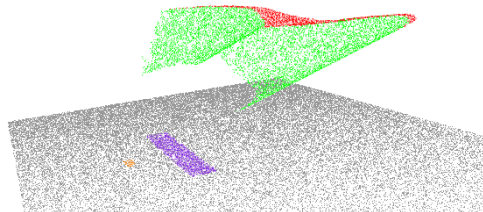
Example 2: Anticipative gradient [paper at SASO 2012]

AG from S, with future event F in between distance/time D1 and D2

```
def Sector(S,F) is 2nd( [ {<S,F>}+<#RNG,F> ] )
def Crown(S,D1,D2) is Grad(S)>=D1 and Grad(S)<=D2
def Warn(S,F,D1,D2) is Grad(S,F) min Grad(F)-Grad(S)+D2
def Ag(S,F,D1,D2) is Sector(S,F) and Crown(S,D1,D2) ? Warn(S,F,D1,D2)
                    : Grad(S)
```



(Simulation of an equivalent model developed in Proto)



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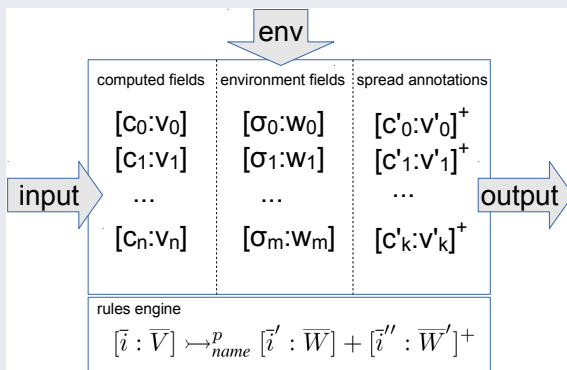
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Node architecture and rule model

Architecture



Computation round

Sleep, clean, environment update, input, exec rules, output

Rewrite Semantics of the Field Calculus

Universal rules

$\mathcal{F}(i, \bar{i}, f)$	$= [\bar{i} : \bar{x}] \mapsto_F [\bar{i} : \bar{x}] + [i : f(\bar{x})]$	<i>Functional comp.</i>
$\mathcal{C}(i_s, i_c, \bar{i}, g)$	$= [i_s : x] + [\bar{i} : \bar{x}] \mapsto_C^* [\bar{i} : \bar{x}] + [i_c : g(x, \bar{x})]$	<i>Contextualisation</i>
$\mathcal{A}(i_c)$	$= [i_c : x] + [i_c : x'] \mapsto_A^* [i_c : x \wedge x']$	<i>Min-aggregation</i>
$\mathcal{O}(i_c, i, i_o)$	$= [i_c : x] + [i : x'] \mapsto_O [i : x'] + [i_o : x \wedge x']$	<i>Output</i>
$\mathcal{S}(i_o, i_s)$	$= [i_o : x] \mapsto_S [i_o : x] + [i_s : x]^+$	<i>Spreading</i>

Deriving the rules needed for a given field expression

$$\frac{\frac{}{|\sigma| \stackrel{\sigma}{=}}}{\frac{\text{fresh}(i) \quad |\bar{F}| \stackrel{\bar{i}}{=} \bar{R} \quad R_F = \mathcal{F}(i, \bar{i}, f)}{|f(\bar{F})| \stackrel{i}{=} (\bar{R} \ R_F)}}$$

$$\frac{\text{fresh}(i_s, i_c, i_o) \quad |F| \stackrel{i}{=} \bar{R} \quad |\bar{F}| \stackrel{\bar{i}}{=} \bar{R}' \quad R_C = \mathcal{C}(i_s, i_c, \bar{i}, g) \quad R_A = \mathcal{A}(i_c) \quad R_O = \mathcal{O}(i_c, i, i_o) \quad R_S = \mathcal{S}(i_o, i_s)}{||g(\{F\}, \bar{F})|| \stackrel{i_o}{=} (\bar{R} \ \bar{R}' \ R_C \ R_A \ R_O \ R_S)}}$$

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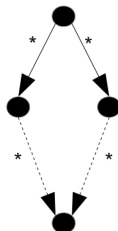


Motivations and Conditions for Confluence

Idea and consequence

- Idea: prove CF by local CF + absence of infinite transition sequences
- Consequence: it will suffice to show that one sequence of transitions leads to a stable field S , to prove that S will be inevitably reached

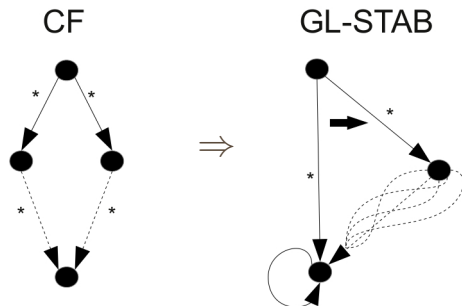
CF



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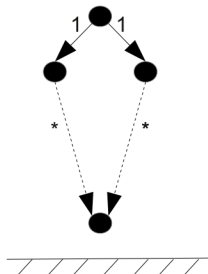


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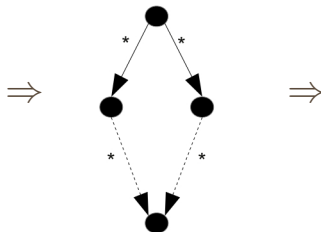
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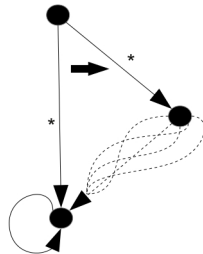
LCF + NOE



CF



GL-STAB

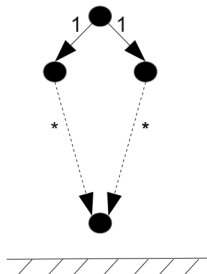


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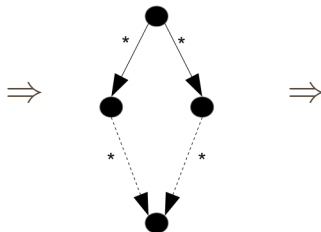
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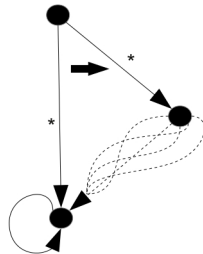
LCF + NOE



CF



GL-STAB



Our proof methodology based on confluence

Sufficient conditions for confluence

- Fair evolution (the whole network computes)
 - ⇒ at each step, each node n is given another chance to “fire”
- Transitory evolution to topology T (avoiding network partition issues)
 - ⇒ start with $T' \subseteq T$, can add/remove links, never escape T , reach T
 - ⇒ environmental fields eventually stabilise
- Well-monotonicity (ensuring “termination” of spreading-aggregation)
 - ⇒ $g(v, \bar{v}) \geq v$
 - ⇒ $g(v, \bar{v}) \geq g(v', \bar{v})$ if $v \geq v'$
 - ⇒ no infinite sequence built as $v_i = g(v_{i-1}, \bar{v})$ is strictly ascending

Generality of those conditions

- After ensuring values have an upper-bound..
- $[\{\#SRC\}]$, $[\{\#SRC\}+1]$, $[\{\#SRC\}+\#RNG]$, ... and all others, are ok!

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Result 1: Compositionality of stabilised fields

Theorem

- Consider field specifications $F_1, \dots, F_n$, each stabilising;
- let $\Downarrow F_i$ be the fields ϕ_i that F_i stabilises to;
- then, informally: $\Downarrow f(F_1, \dots, F_n) = f(\Downarrow F_1, \dots, \Downarrow F_n)$
- and, $\Downarrow [g(\{F_1\}, F_2, \dots, F_n)] = [g(\{\Downarrow F_1\}, \Downarrow F_2, \dots, \Downarrow F_n)]$

Rationale

- Field evolution processes are very nondeterministic
 - Though, we can think to field expressions in terms of their (stabilised) outcome, i.e., the resulting field ϕ constructed compositionally
- ⇒ Namely, we can forget about transient effects

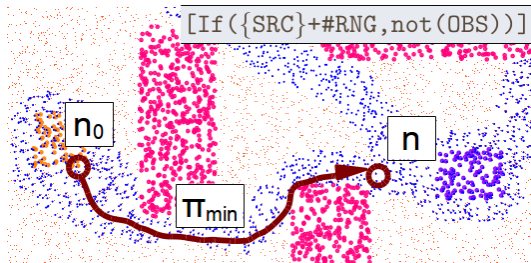


Result 2: Outcome of spreading-aggregation

What $[g(\{F_1\}, F_2, \dots, F_n)]$ stabilises to?

Computing stable value v in node n

- it is the same you would have considering a chain π_{min} in isolation
- π_{min} is the shortest path according to the metric imposed by g, F_i
- value v is easily iteratively computed out of g
- stabilisation-time is linear in the size of π_{min}



Conclusions

Main result

Proving that a whole set of fields are stabilising

⇒ Fostering production of a library of well-engineered selforg mechanisms

Future technical works

- enlarging applicability (e.g., releasing boundedness)
- enriching the spreading-aggregation construct
- studying further properties (soundness, convergence to continuum)

Future practical works

- formal mapping to Proto (towards better analysis of Proto programs)
- formal mapping to SAPERE (applying in pervasive computing)

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