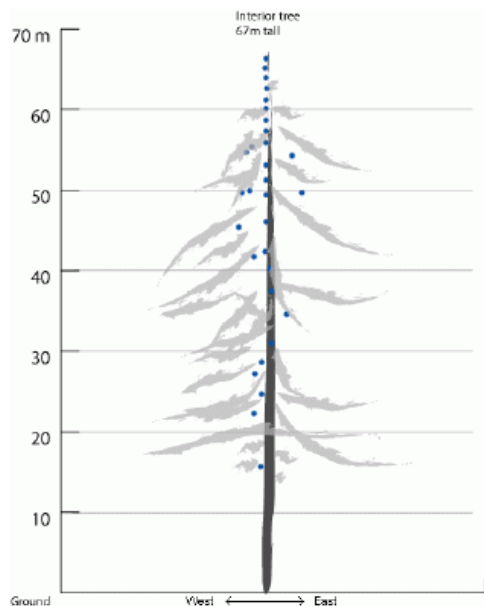


Programming Spatial Computers

Jonathan Bachrach & Jacob Beal
October, 2006

Space-filling Computers



Sensor networks



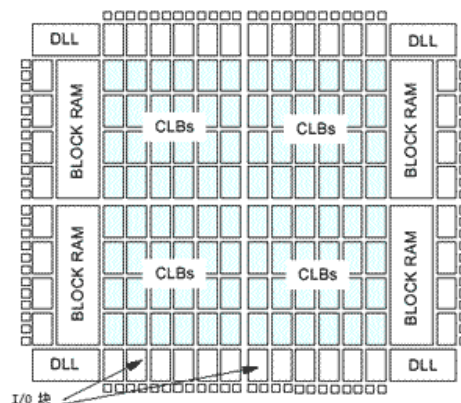
Distributed Control Systems



Biological Computing



Robot Swarms

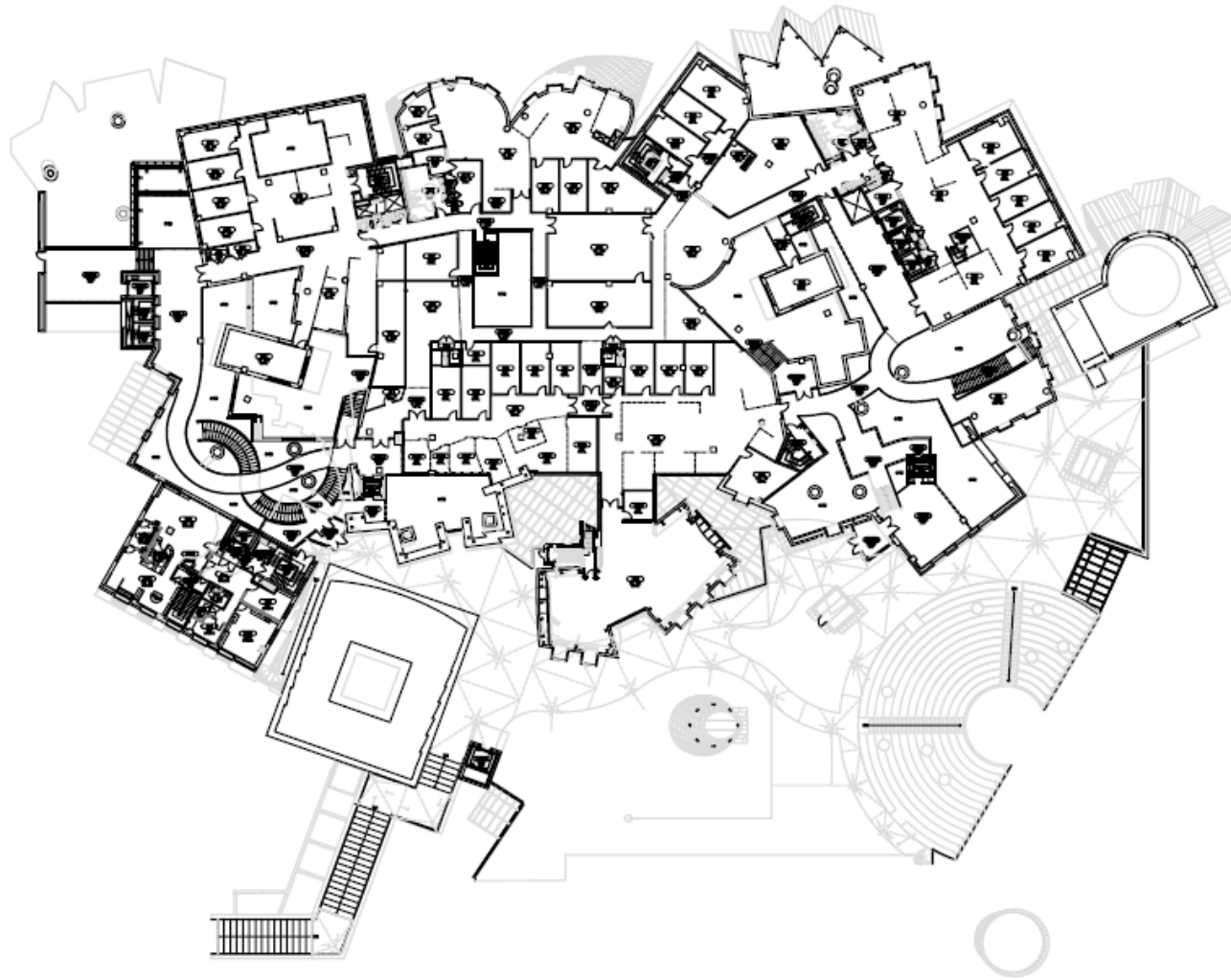


FPGAs

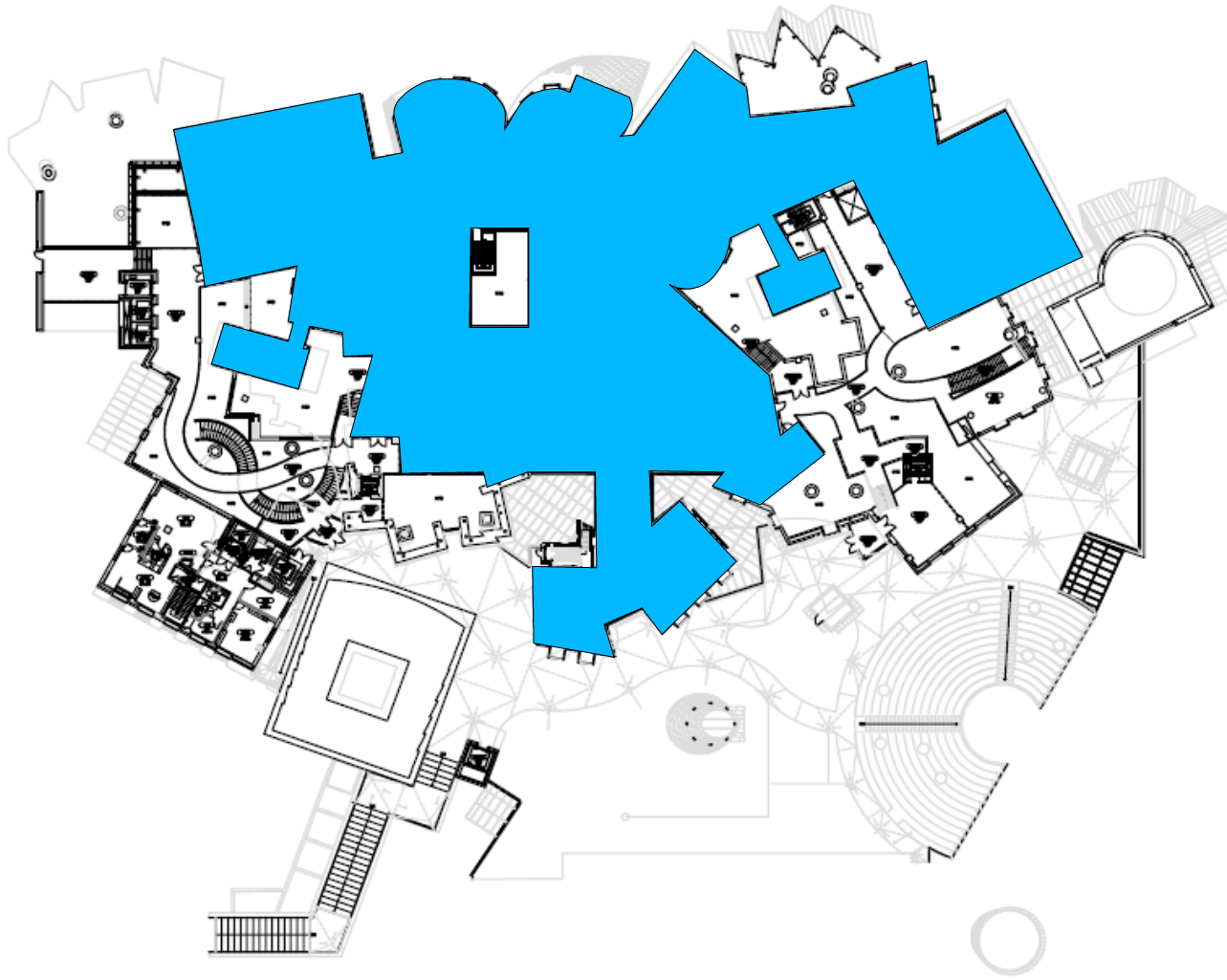


Programmable Matter

Amorphous Medium Approach



Amorphous Medium Approach

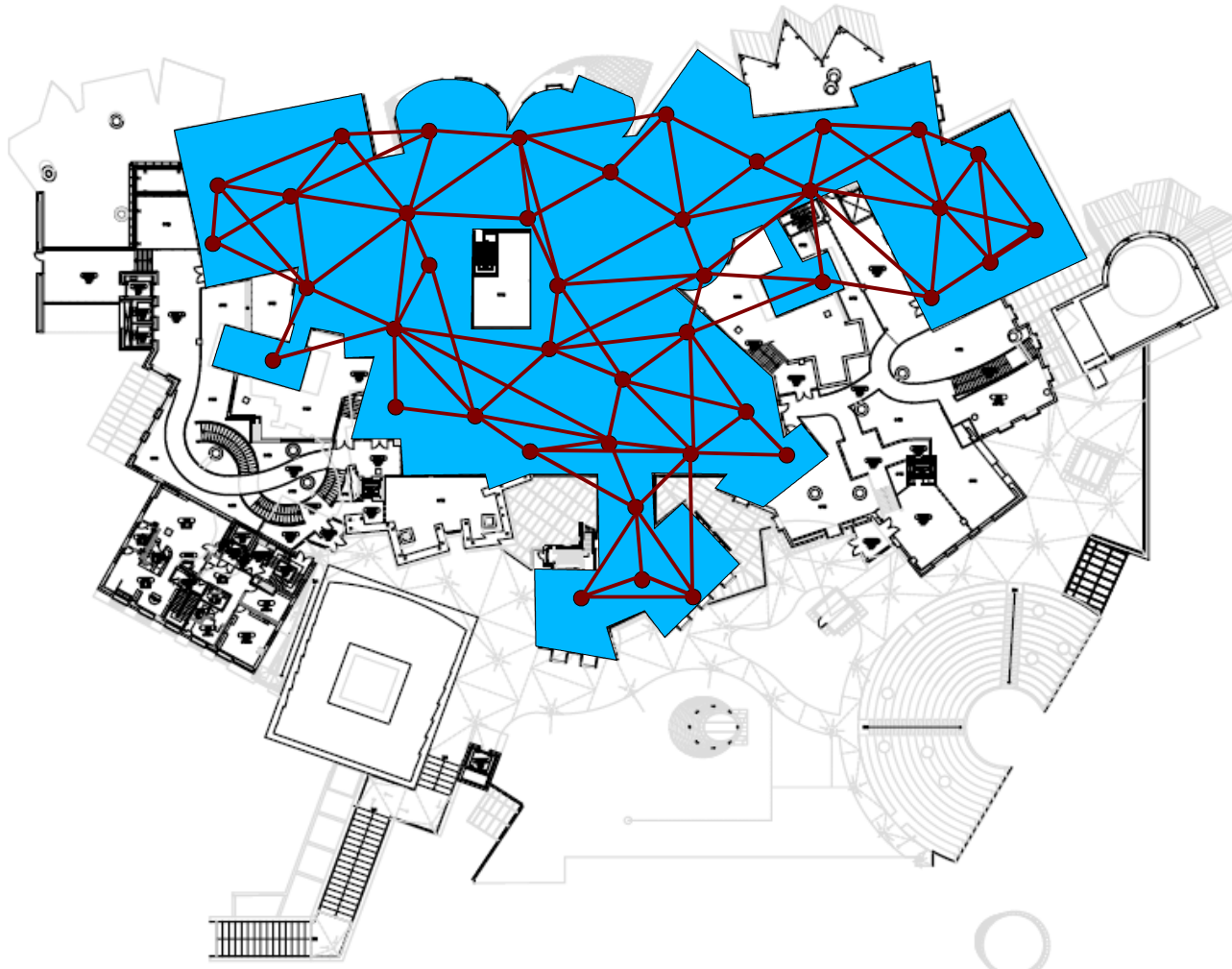


Amorphous Medium Approach



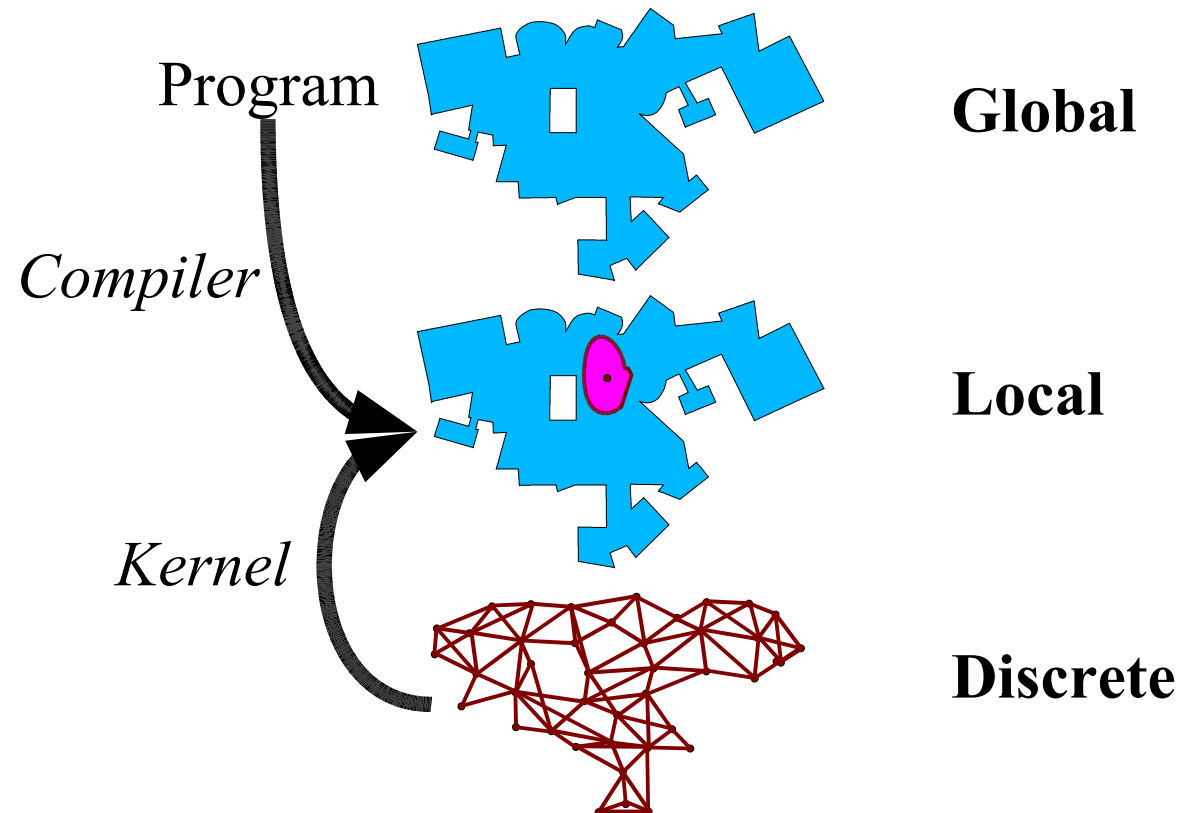
Program the space... approximate with a network

Amorphous Medium Approach

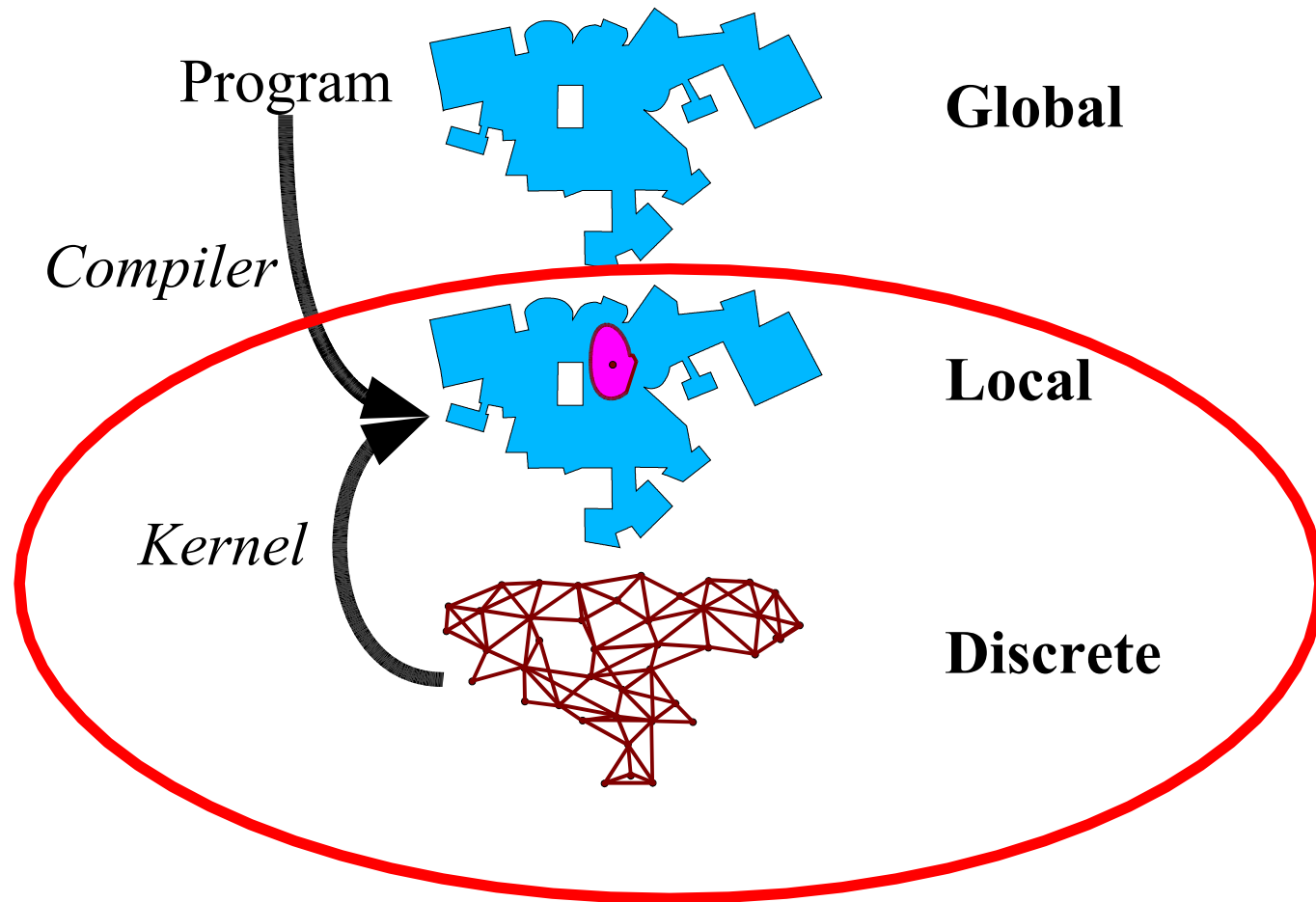


The discretization hardly matters!

Global v. Local v. Discrete



Global v. Local v. Discrete



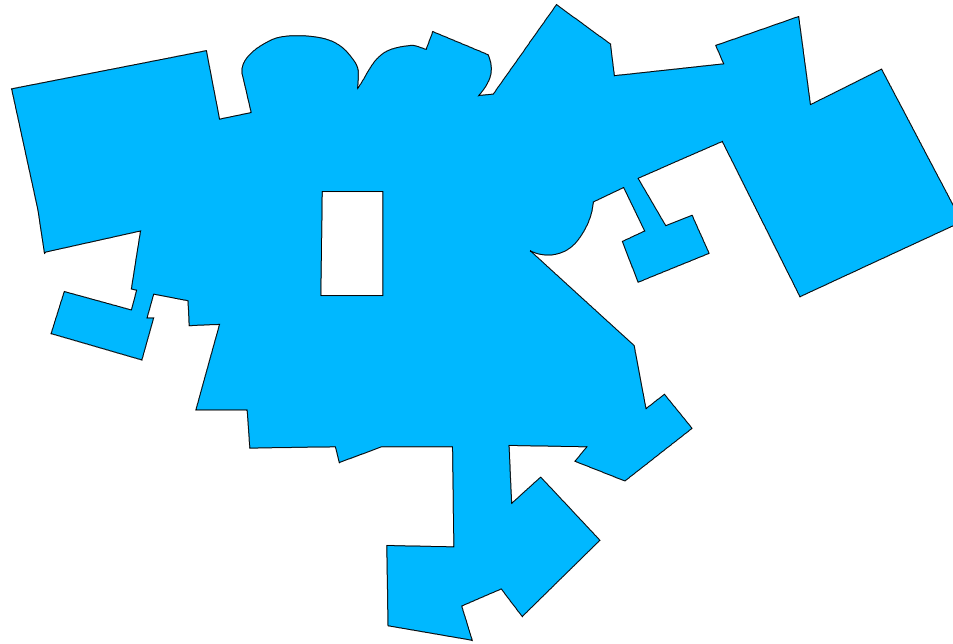
Discrete Model

- Dozens to billions of simple, unreliable agents
- Distributed through space, communicating by local broadcast
- Agents may be added or removed
- No guaranteed global services (e.g. time, naming, routing, coordinates)
- Relatively cheap power, memory, processing
- Partial synchrony

Kernel

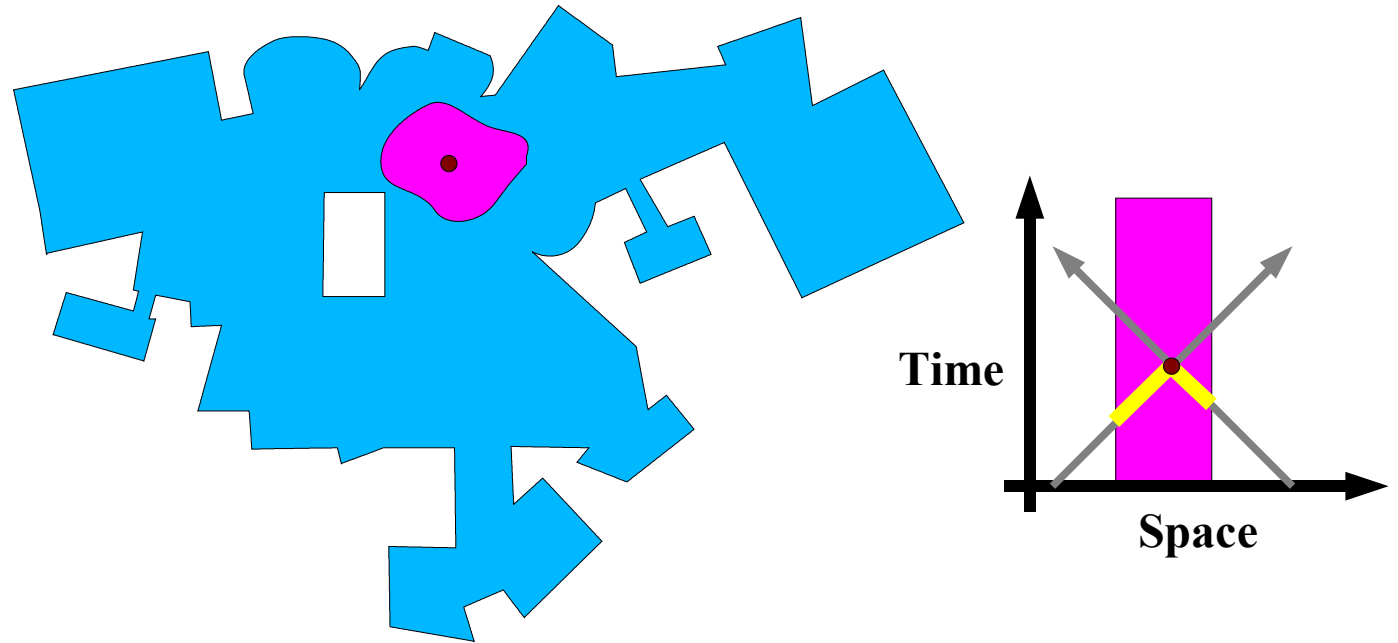
- Responsibilities:
 - Emulate amorphous medium
 - Time evolution
 - Interface with sensors, actuators
 - Viral reprogramming
- Current platforms:
 - Simulator
 - Mica2 Mote
 - McLurkin's Swarm

Amorphous Medium



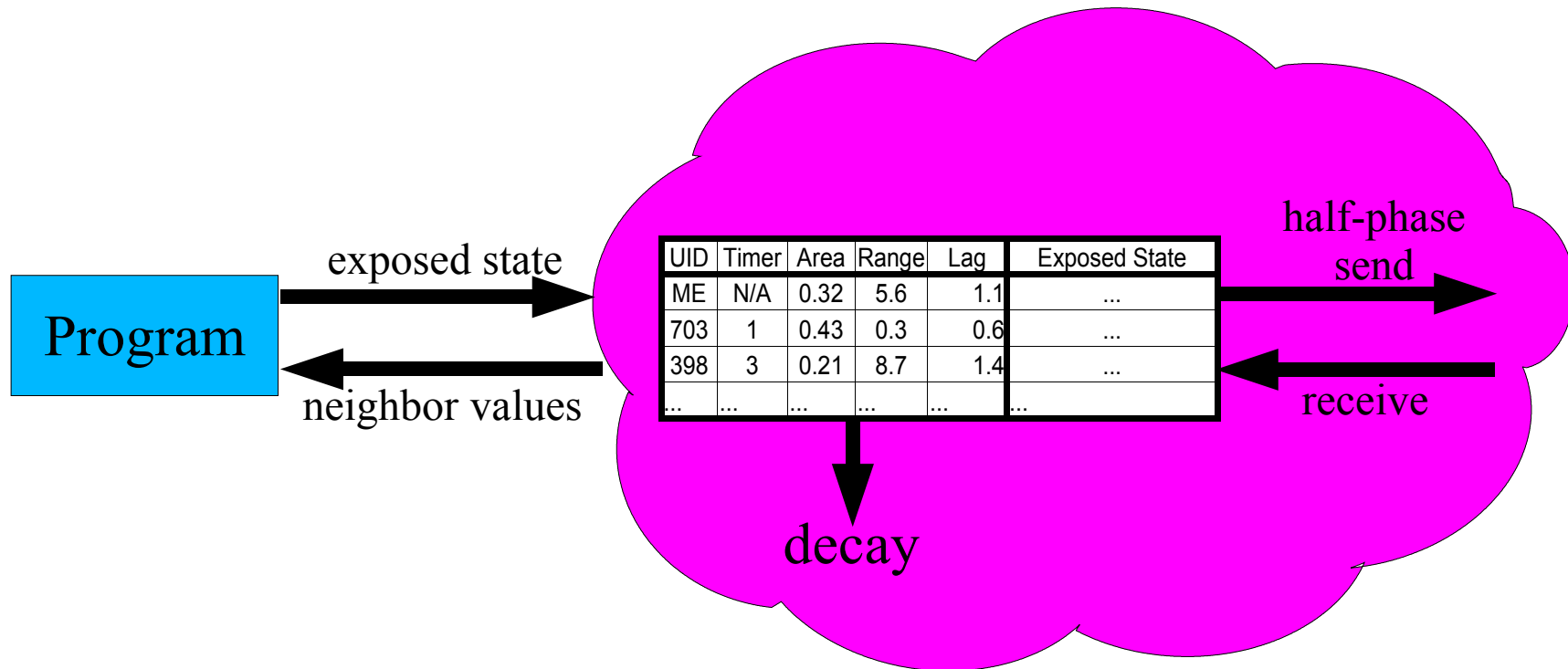
- Manifold (locally Euclidean space)
 - Assume Riemannian, smooth, compact
 - Simple locally, complex globally

Amorphous Medium



- Points access past values in their neighborhood
 - Information propagates at a fixed rate c
- Evaluation is repeated at fixed intervals

Neighborhood Abstraction

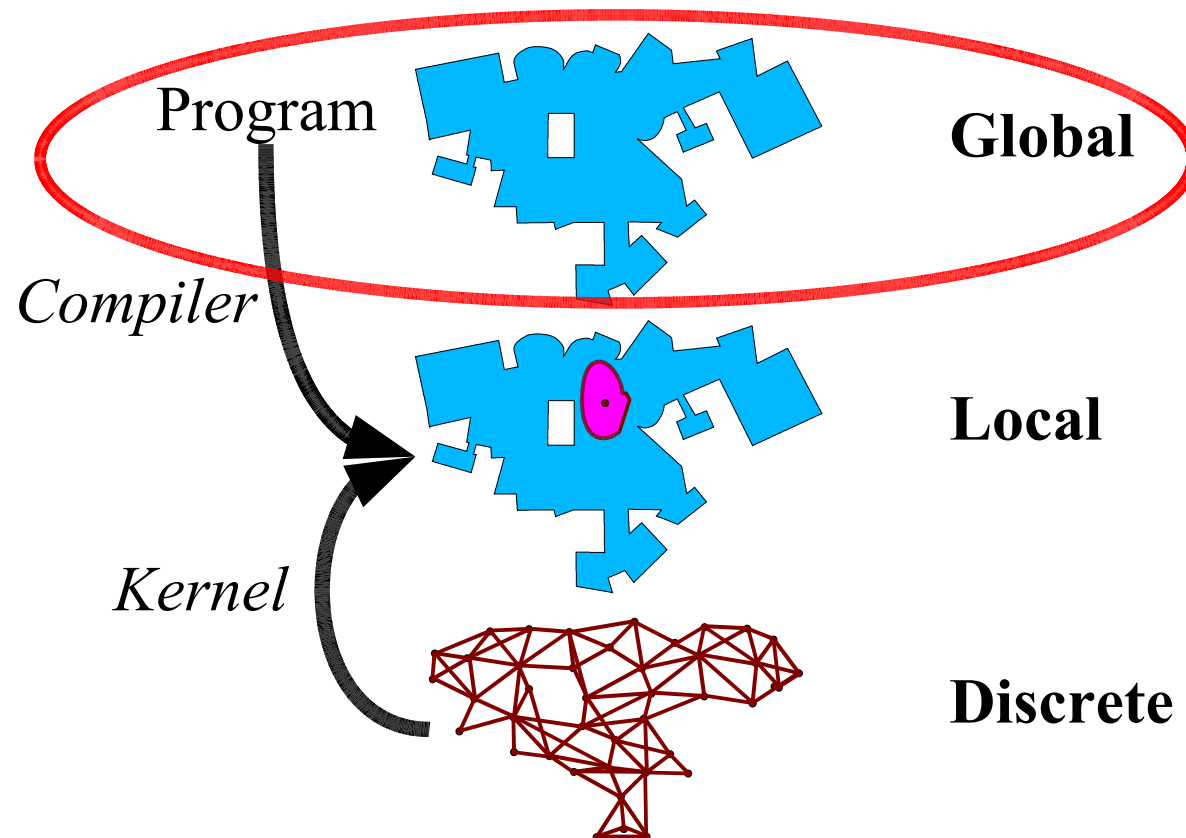


- Aggregate access to best-effort estimate of neighbor state, space-time properties
- Neighbors decay without updates

Kernel Open Questions

- What is the optimal replacement policy when there are more neighbors than table memory?
- What is the optimal decay rate?
- How much energy can be saved by throttling update and decay rates?
- What are good ways to expose the cost/responsiveness tradeoff to the programmer?

Global v. Local v. Discrete



Expressions

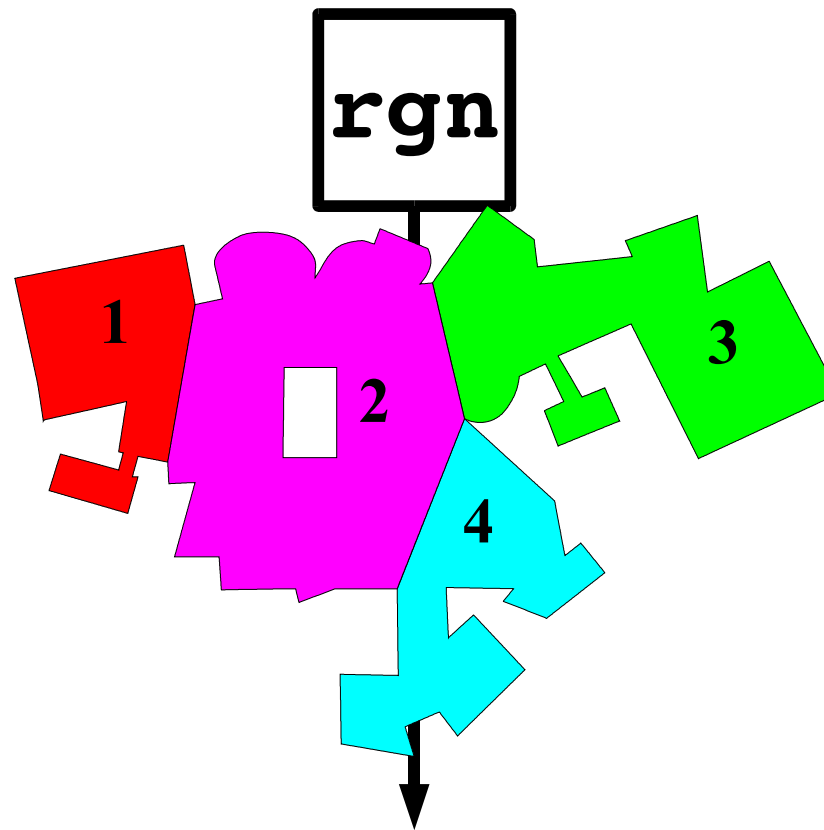
rgn



- An expression maps a manifold to a field

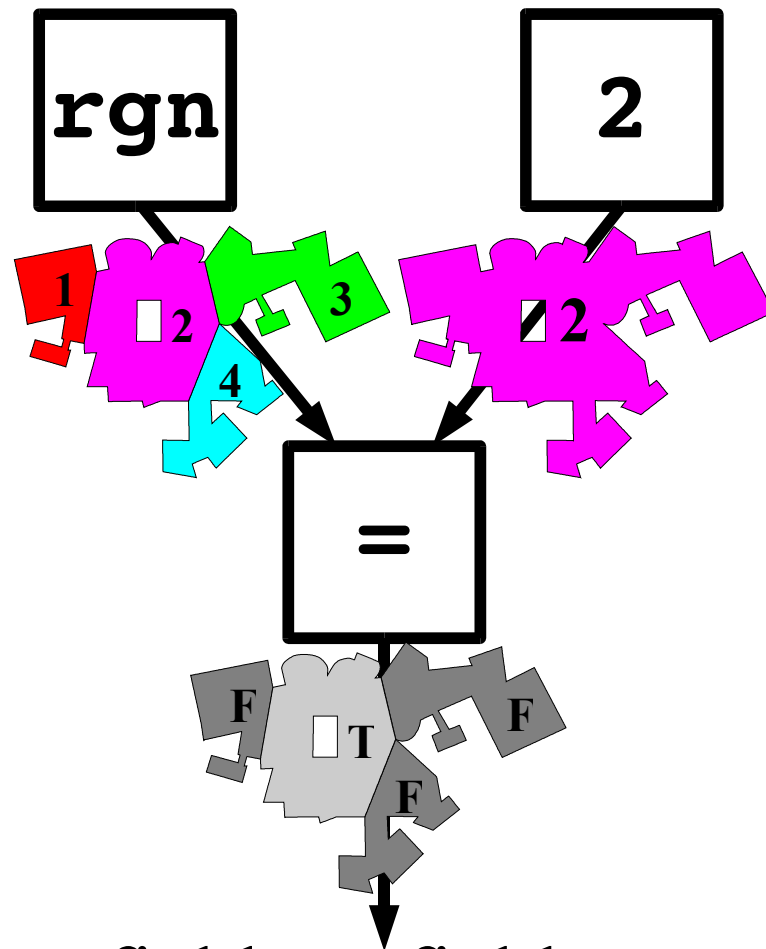
$$\mathbf{rgn}: \mathbf{M} \rightarrow (\mathbf{M} \rightarrow \mathbf{R})$$

Expressions



- An expression maps a manifold to a field
 $\mathbf{rgn}: \mathbf{M} \rightarrow (\mathbf{M} \rightarrow \mathbf{R})$

Operators



- Operators map fields to fields (**= rgn 2**)

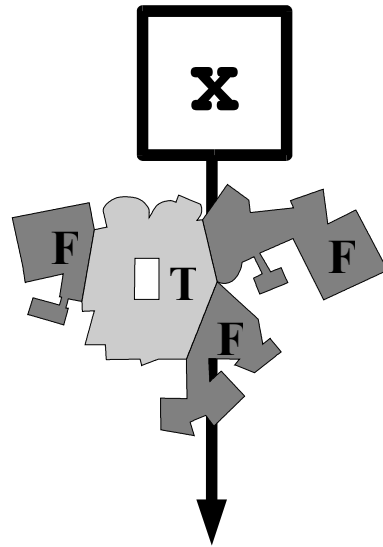
$$=: (\mathbf{M} \rightarrow \mathbf{R}) \times (\mathbf{M} \rightarrow \mathbf{R}) \rightarrow (\mathbf{M} \rightarrow \mathbf{B})$$

Composition & Abstraction

- Functional composition:
 - $\text{operator} \circ \text{expressions} = \text{expression}$
 - $\text{operator} \circ \text{operators} = \text{operator}$
 - $\text{scope} \circ \text{expression} = \text{operator}$
- } **Lambda!**

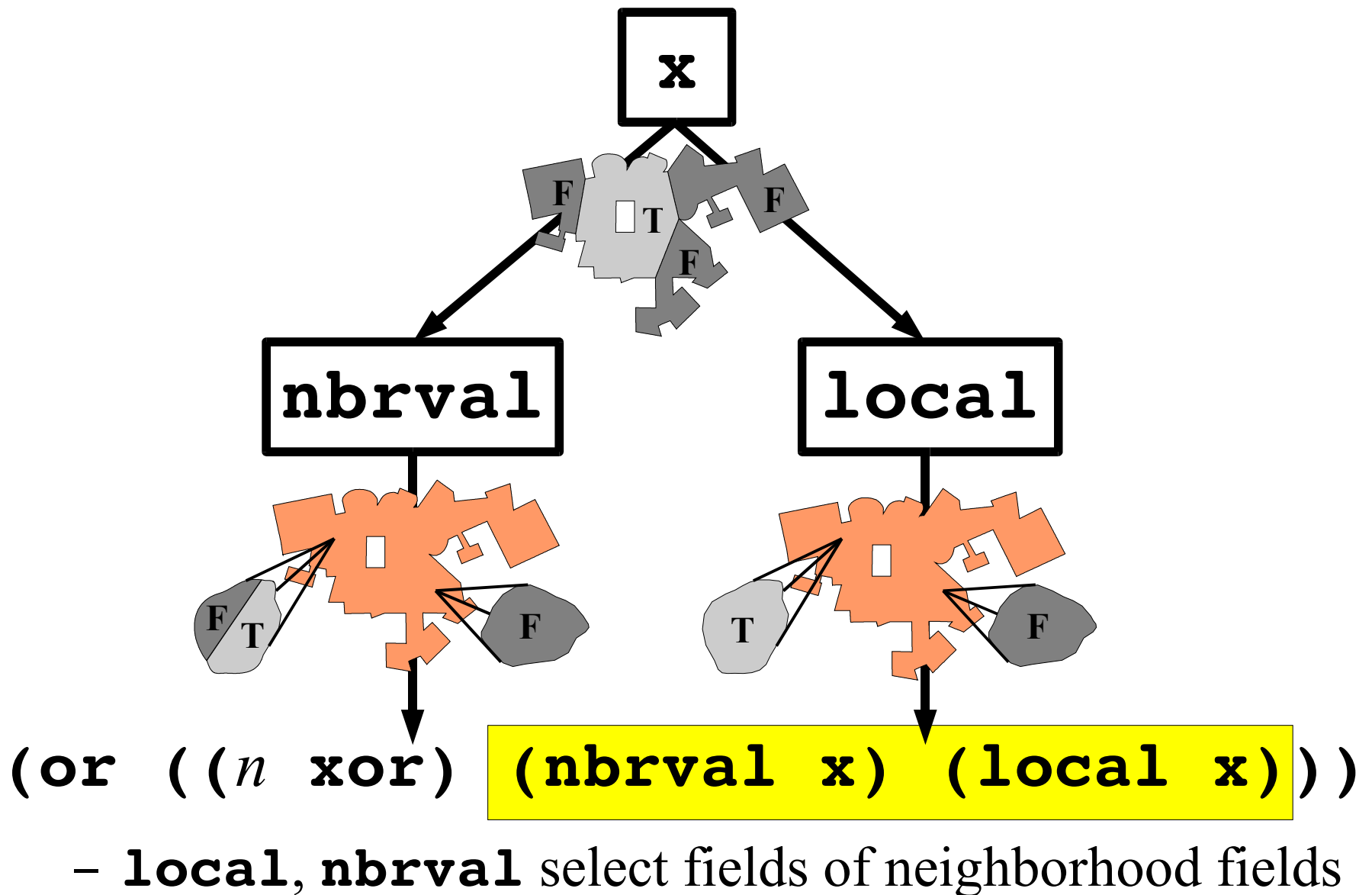
Purely functional pointwise computation ✓

Computation over Neighborhoods

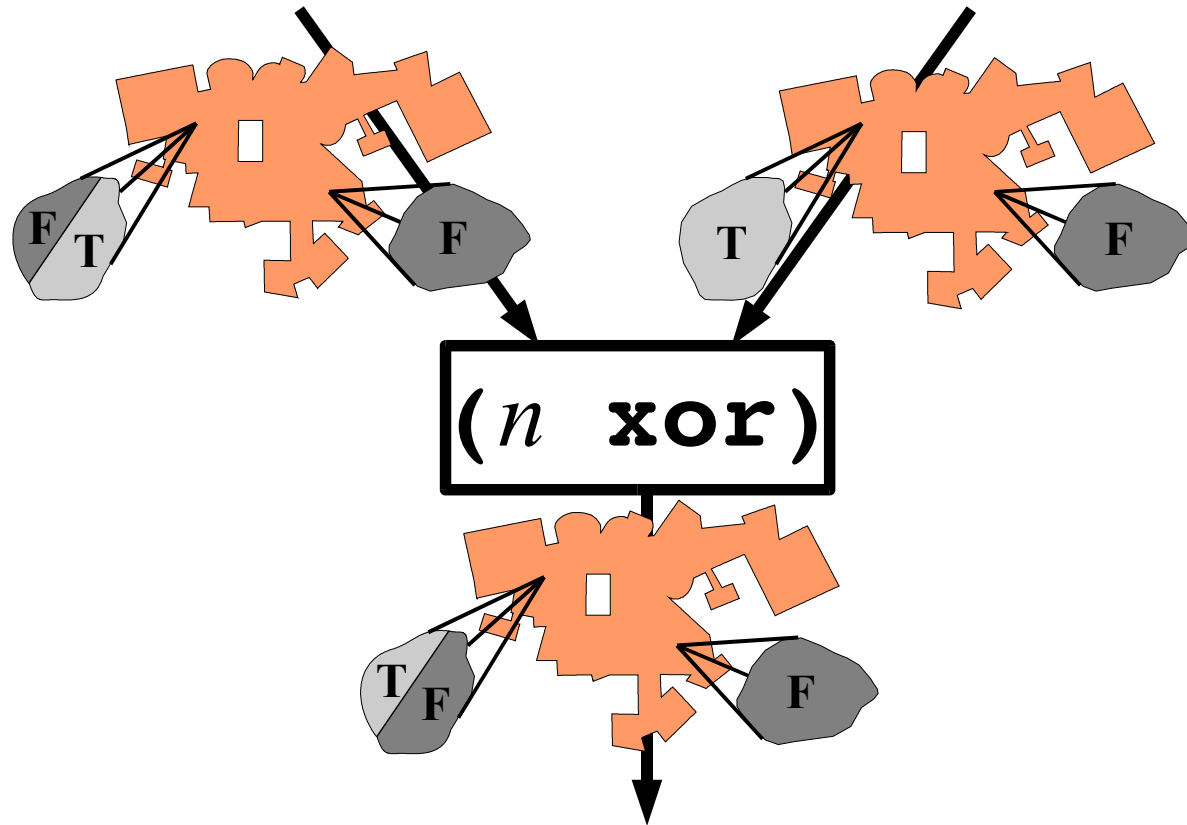


(or ((*n* xor) (nbrval x) (local x)))

Computation over Neighborhoods



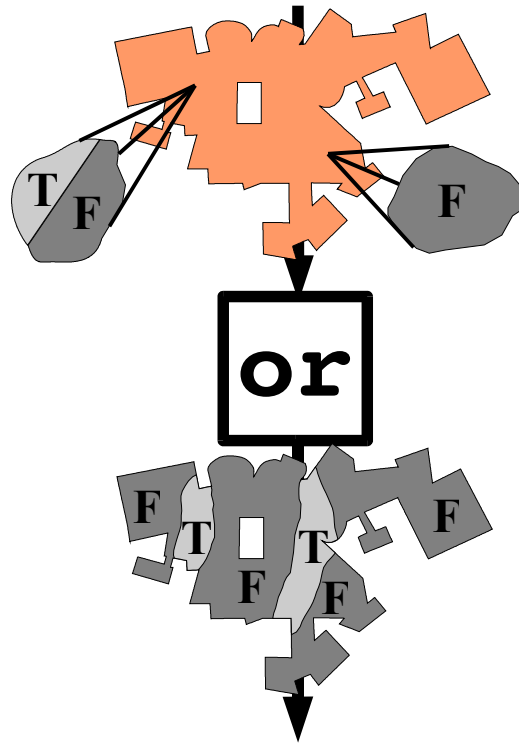
Computation over Neighborhoods



(or ($(n \text{ xor})$ (nbrval x) (local x)))

– n applies an operator to neighborhood fields

Computation over Neighborhoods



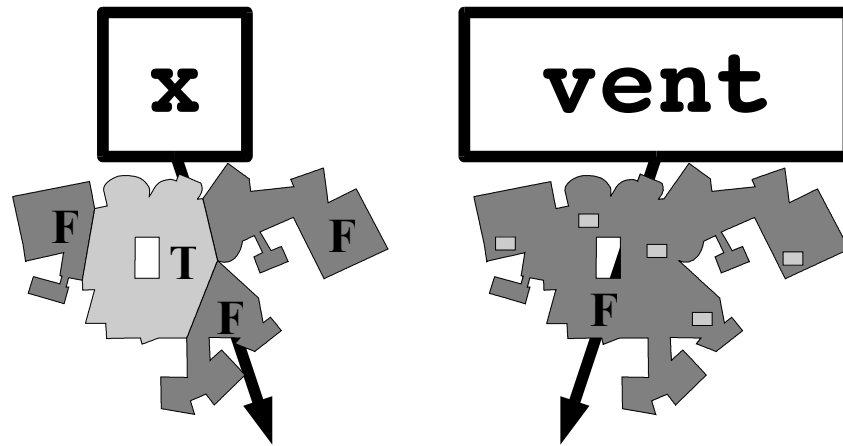
(**or** ((*n* xor) (nbrval x) (local x)))

- Measures (e.g. **or**, **integral**) reduce fields to values
- Sugar: (**reduce-nbrs or (xor x (local x))**)

Neighborhood Open Questions

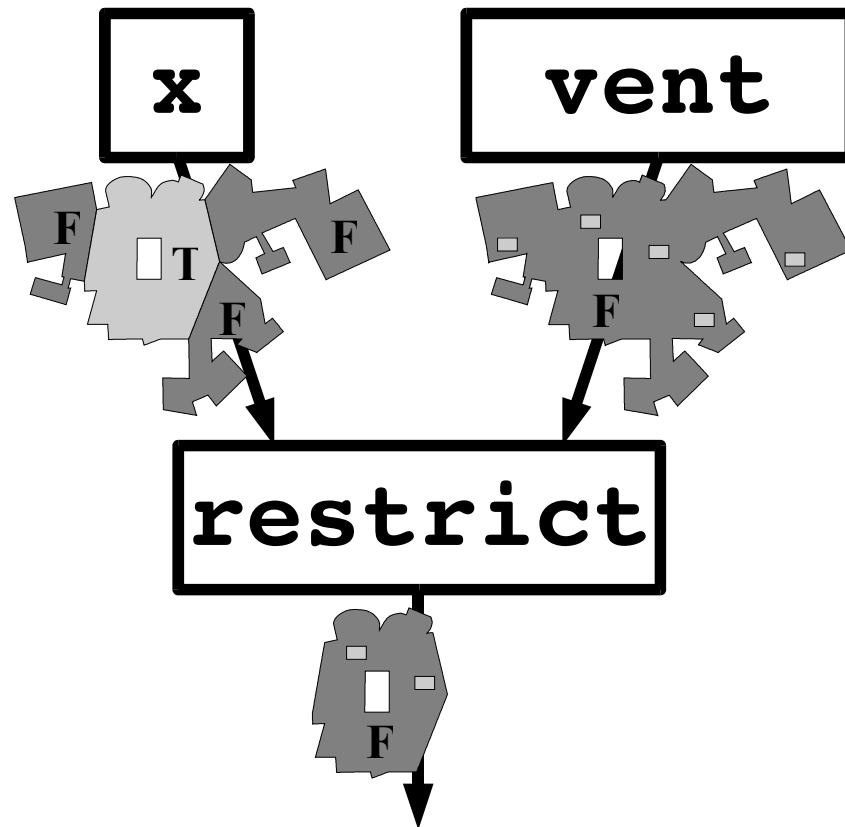
- Are the summary operations **and**, **or**, **min**, **max**, and **integral** sufficient for all approximatable continuous neighborhood computations?
- Are the field primitives **local**, **nbrval** **nbr-range**, **nbr-lag**, and **nbr-bearing** sufficient sources of neighborhood data?
- What is the discretization error of arbitrary composite neighborhood computations?

Conditional Computation



```
(mux x (or (nbrval (restrict x vent)) #F))
```

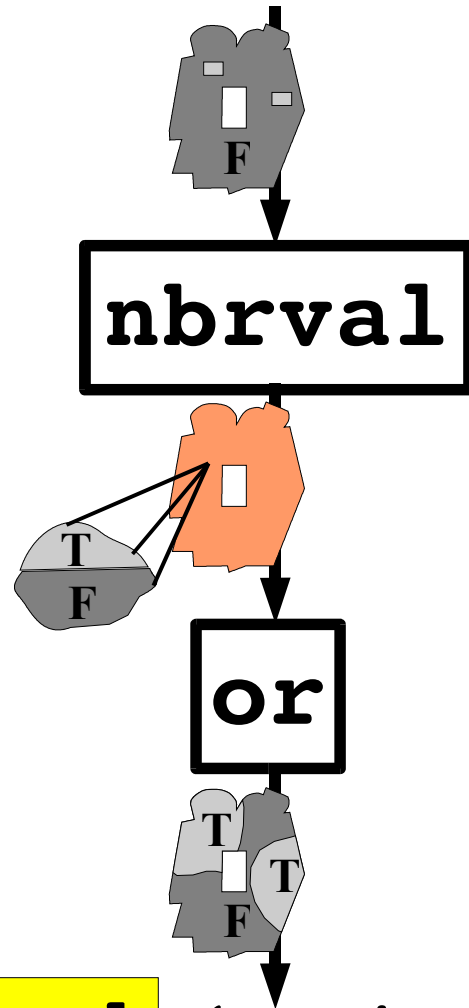
Conditional Computation



```
(mux x (or (nbrval (restrict x vent)) #F))
```

– **restrict** limits the domain of a field

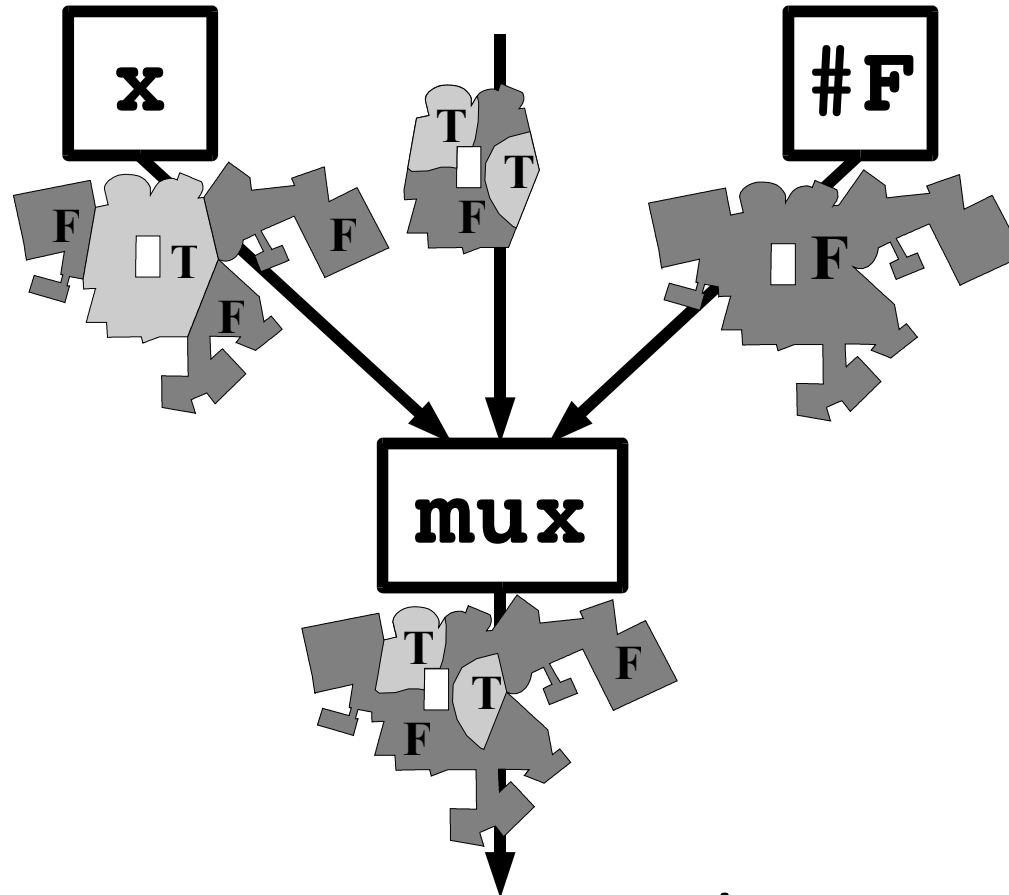
Conditional Computation



`(mux x (or (nbrval (restrict x vent)) #F))`

- operations proceed normally in the restricted field

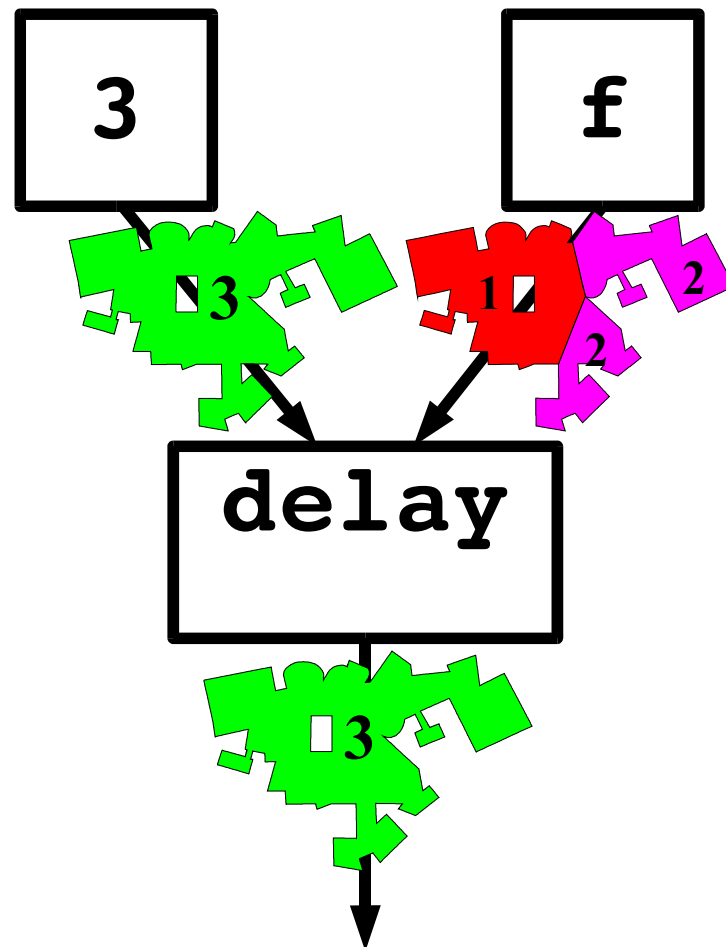
Conditional Computation



(mux x (or (nbrval (restrict x vent)) #F))

- **mux** constructs a field piecewise from inputs
- Sugar: **(if x (or (nbrval vent)))**

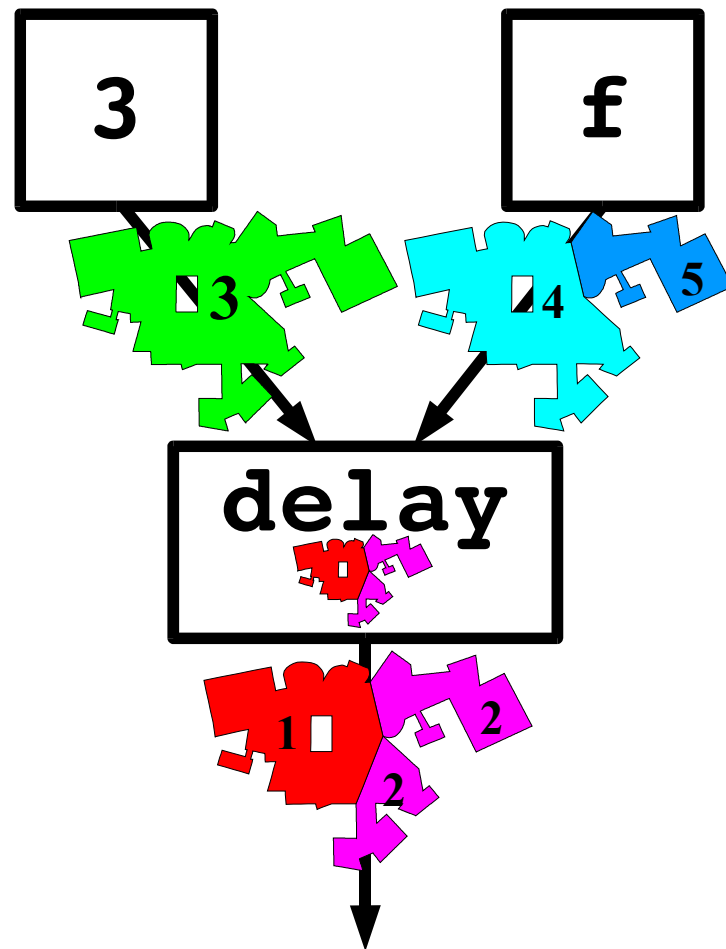
Computation with State



`(delay default init)`

- Previous values, current domain

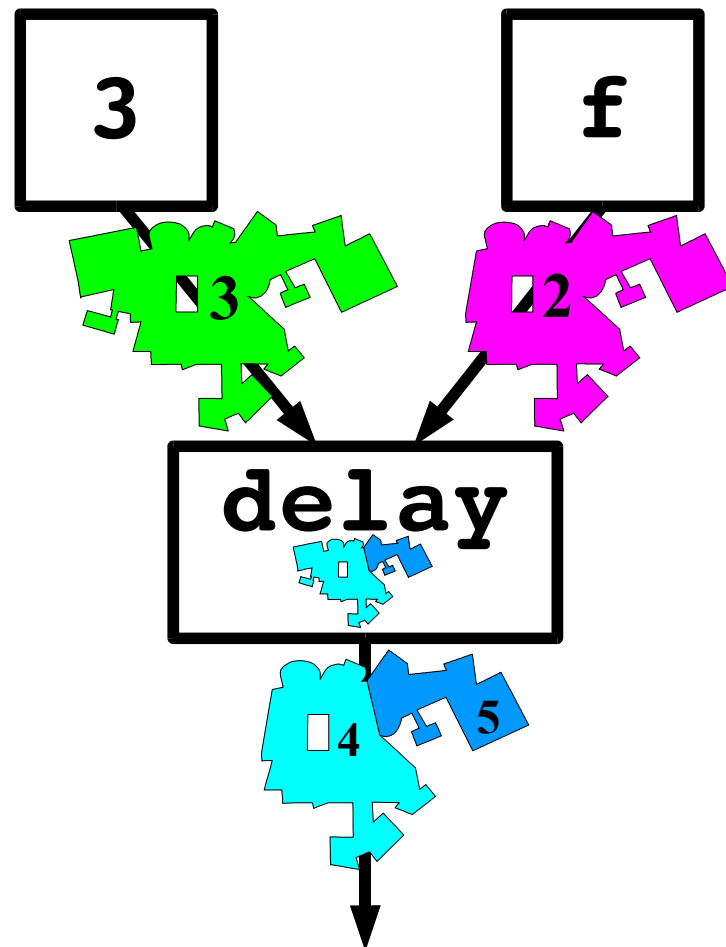
Computation with State



(delay default init)

- Previous values, current domain

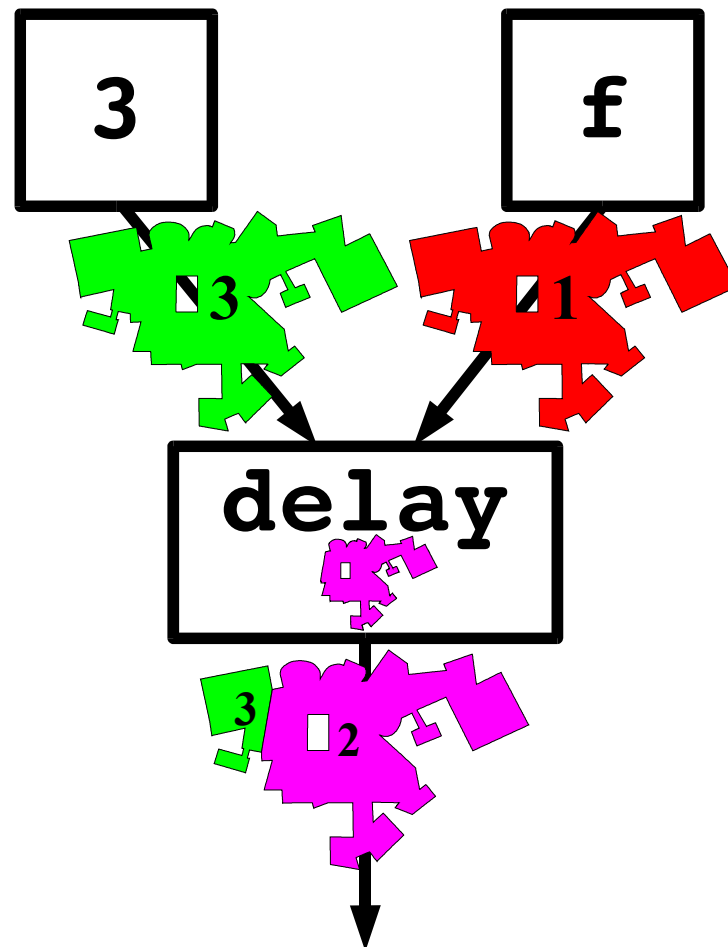
Computation with State



(delay default init)

- Previous values, current domain

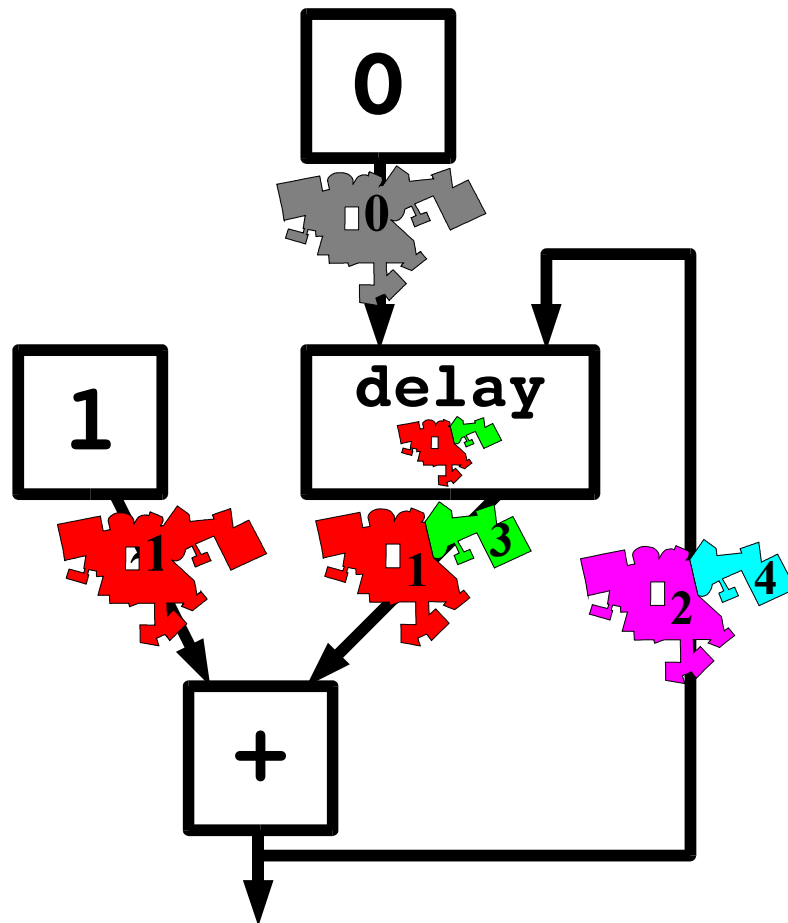
Computation with State



`(delay default init)`

- Previous values, current domain

Computation with State

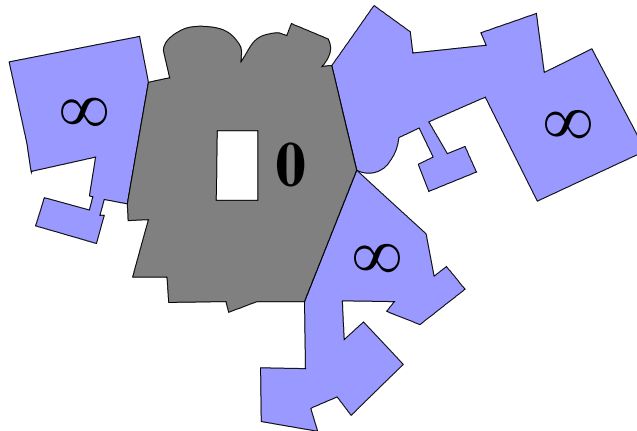


```
(letfed ((n 0 (+ n 1))) n)
```

Putting it all together

- State chains neighborhoods to arbitrary regions
 - Example: relaxation to calculate distance

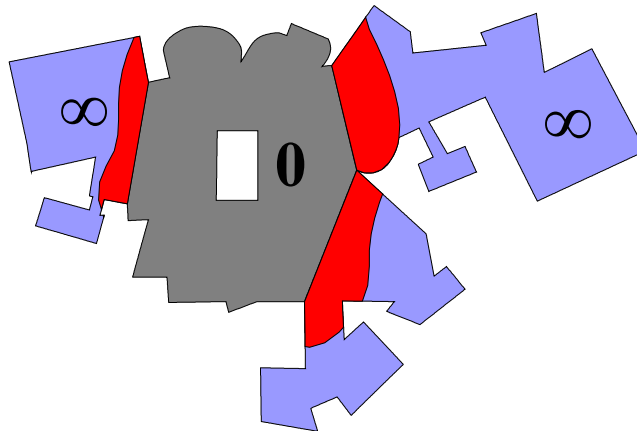
```
(lambda (src)
  (letfed
    ((d ∞ (mux src 0
                 (reduce-nbrs min (+ d nbr-range))))))
  d))
```



Putting it all together

- State chains neighborhoods to arbitrary regions
 - Example: relaxation to calculate distance

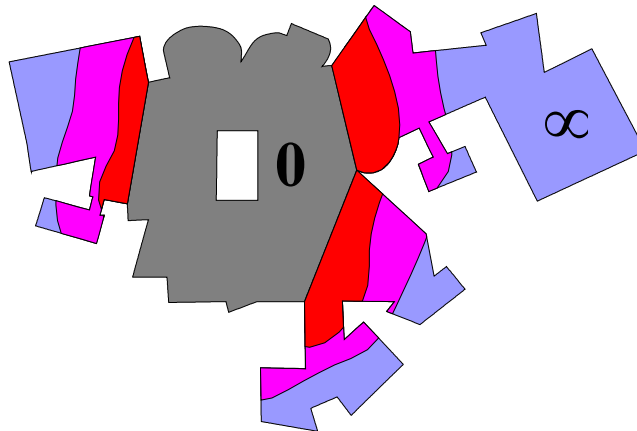
```
(lambda (src)
  (letfed
    ((d ∞ (mux src 0
      (reduce-nbrs min (+ d nbr-range))))))
    d))
```



Putting it all together

- State chains neighborhoods to arbitrary regions
 - Example: relaxation to calculate distance

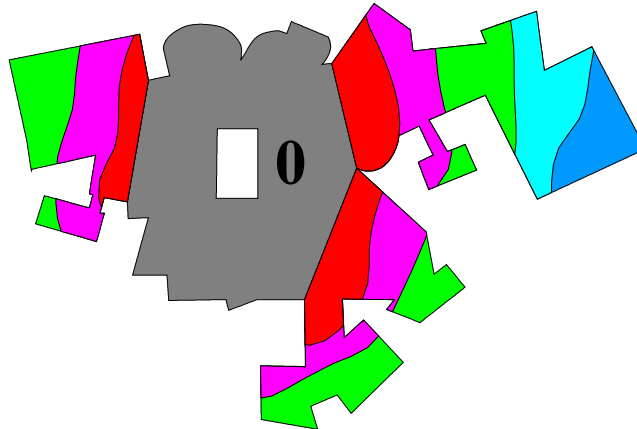
```
(lambda (src)
  (letfed
    ((d  $\infty$  (mux src 0
      (reduce-nbrs min (+ d nbr-range))))))
    d))
```



Putting it all together

- State chains neighborhoods to arbitrary regions
 - Example: relaxation to calculate distance

```
(lambda (src)
  (letfed
    ((d ∞ (mux src 0
                 (reduce-nbrs min (+ d nbr-range))))))
  d))
```

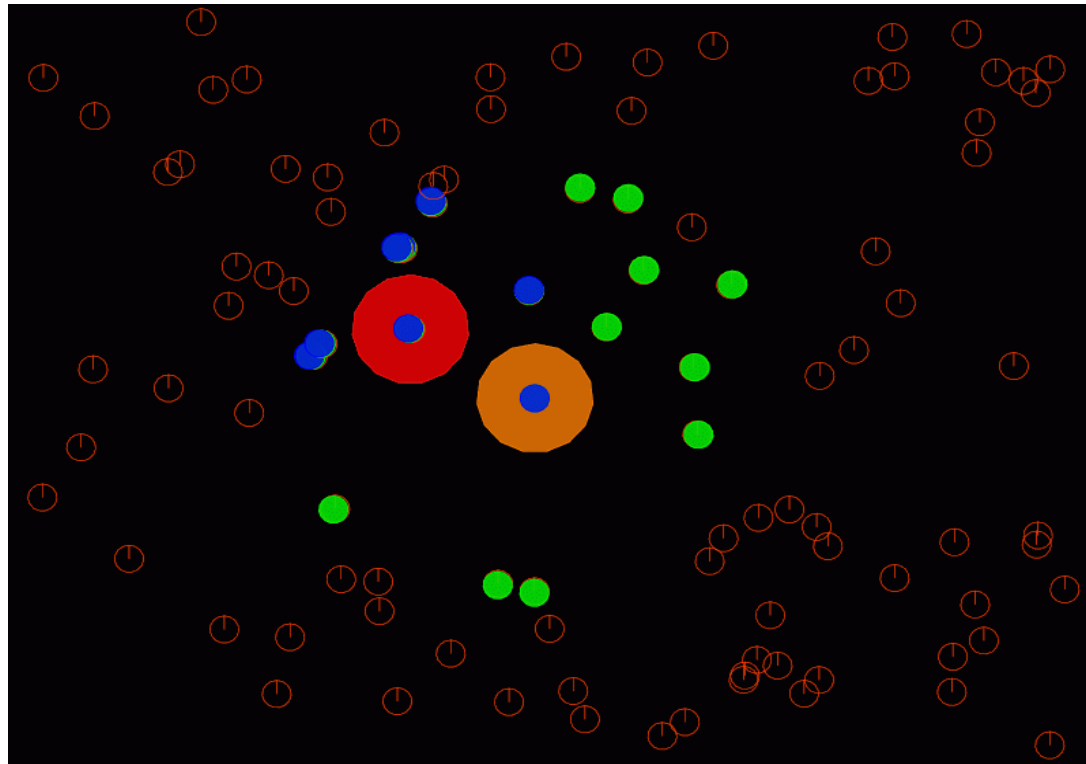


Program Open Questions

- Under what conditions does continuous convergence imply discrete convergence?
- How do convergence properties compose?
- Given a continuous program and desired error bounds, what discretization will suffice?
- Given a continuous program and a discretization, what will the error bounds be?

Programs scale gracefully

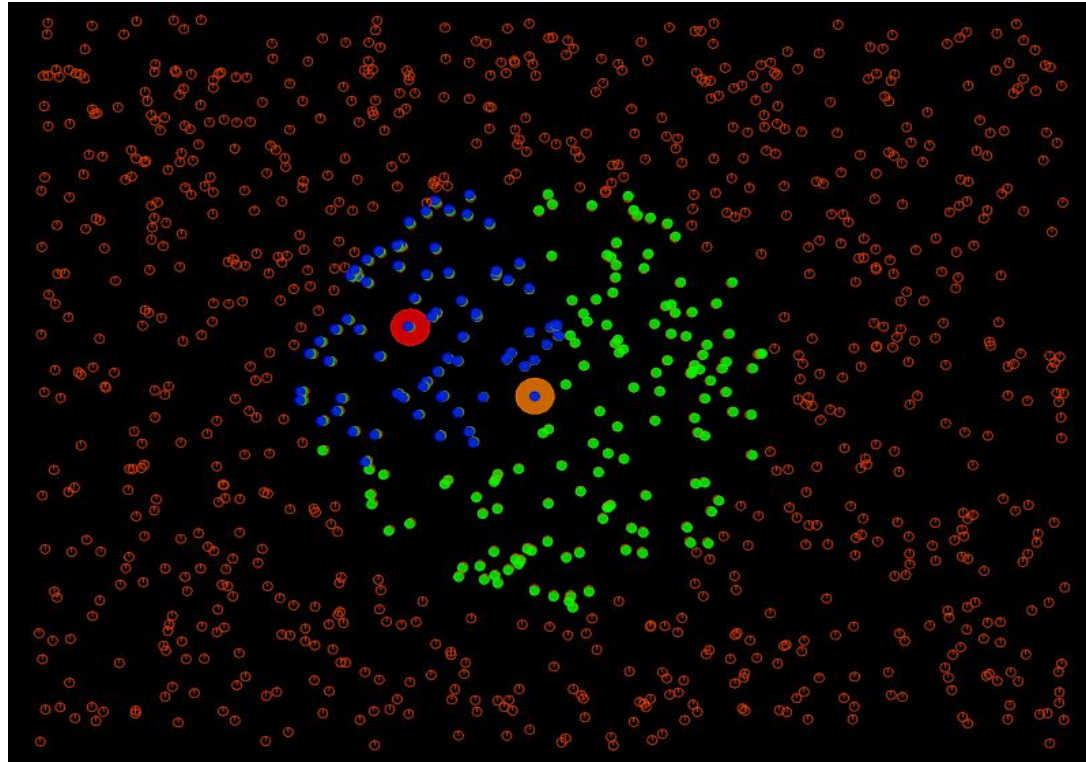
100 nodes



```
(and (green (dilate (sense 1) 30))  
      (blue (dilate (sense 2) 20)))
```

Programs scale gracefully

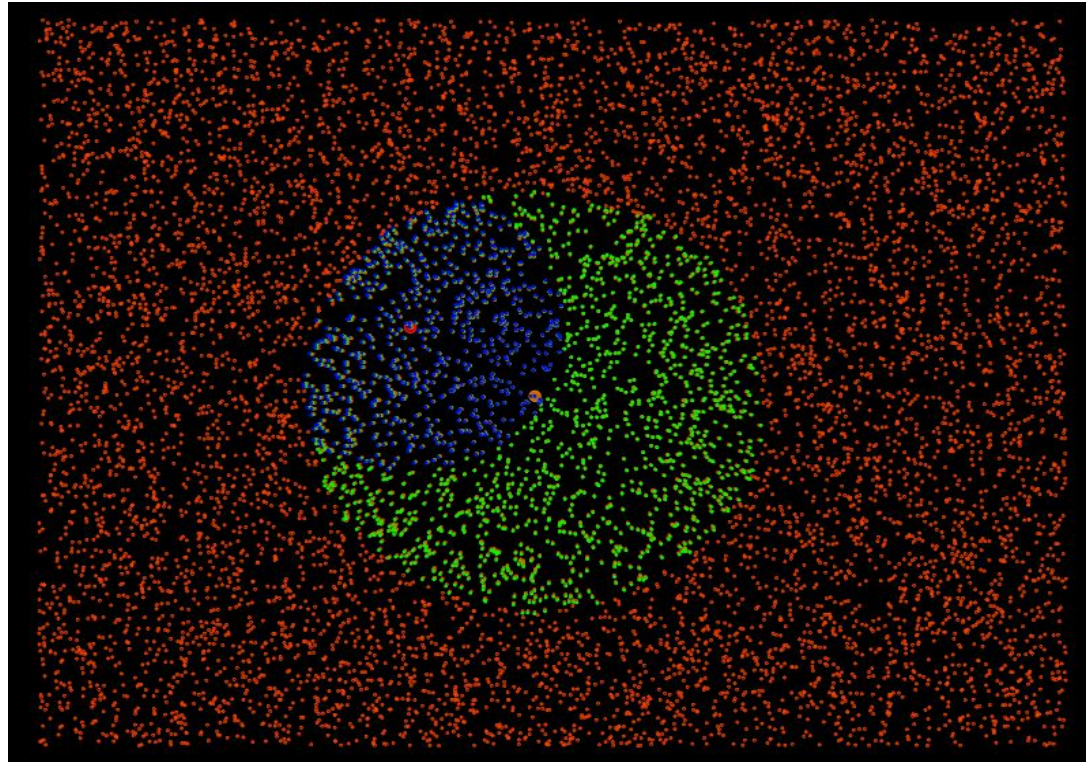
1,000 nodes



```
(and (green (dilate (sense 1) 30))  
      (blue (dilate (sense 2) 20)))
```

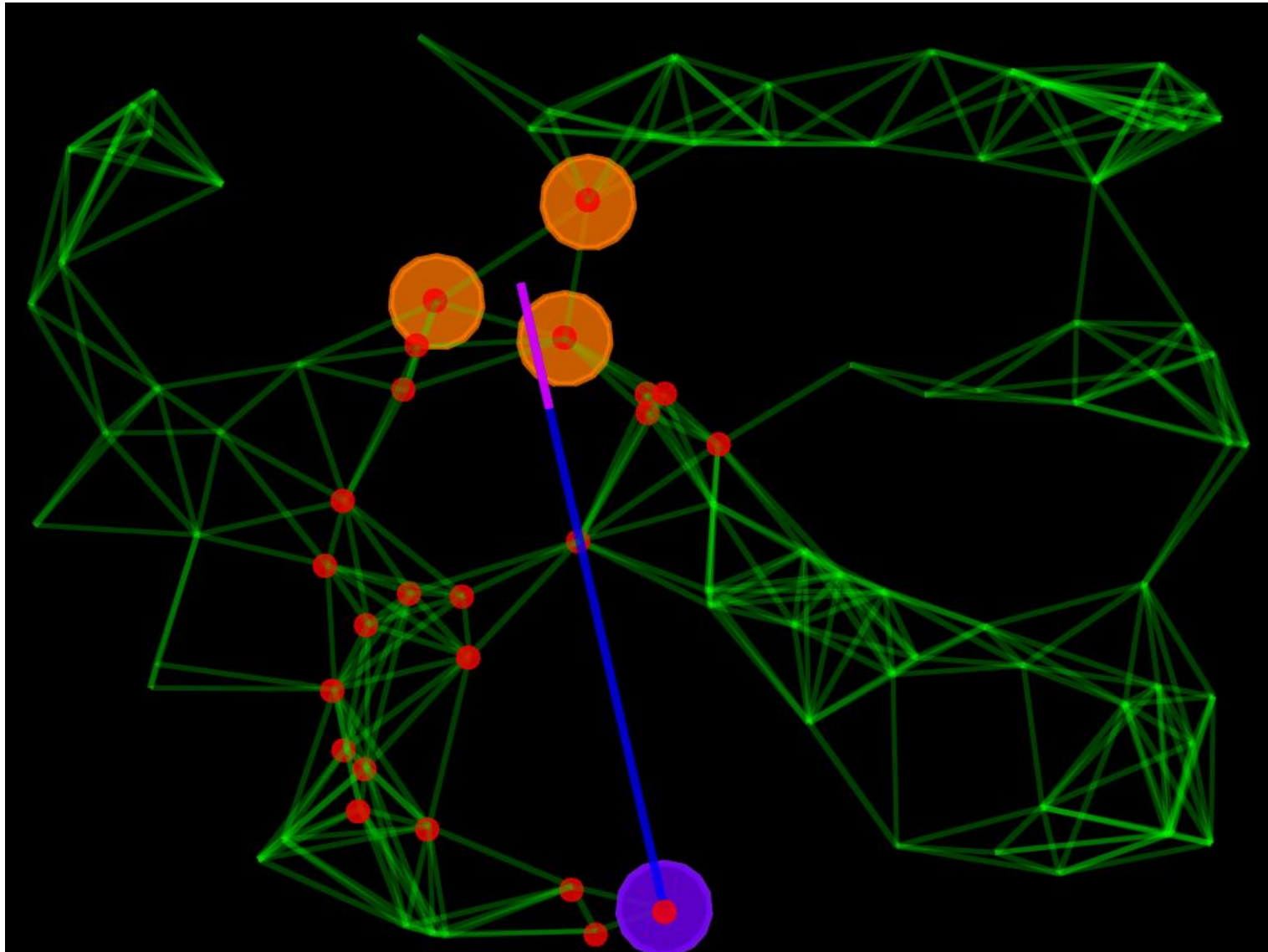
Programs scale gracefully

10,000 nodes



```
(and (green (dilate (sense 1) 30))  
      (blue (dilate (sense 2) 20)))
```

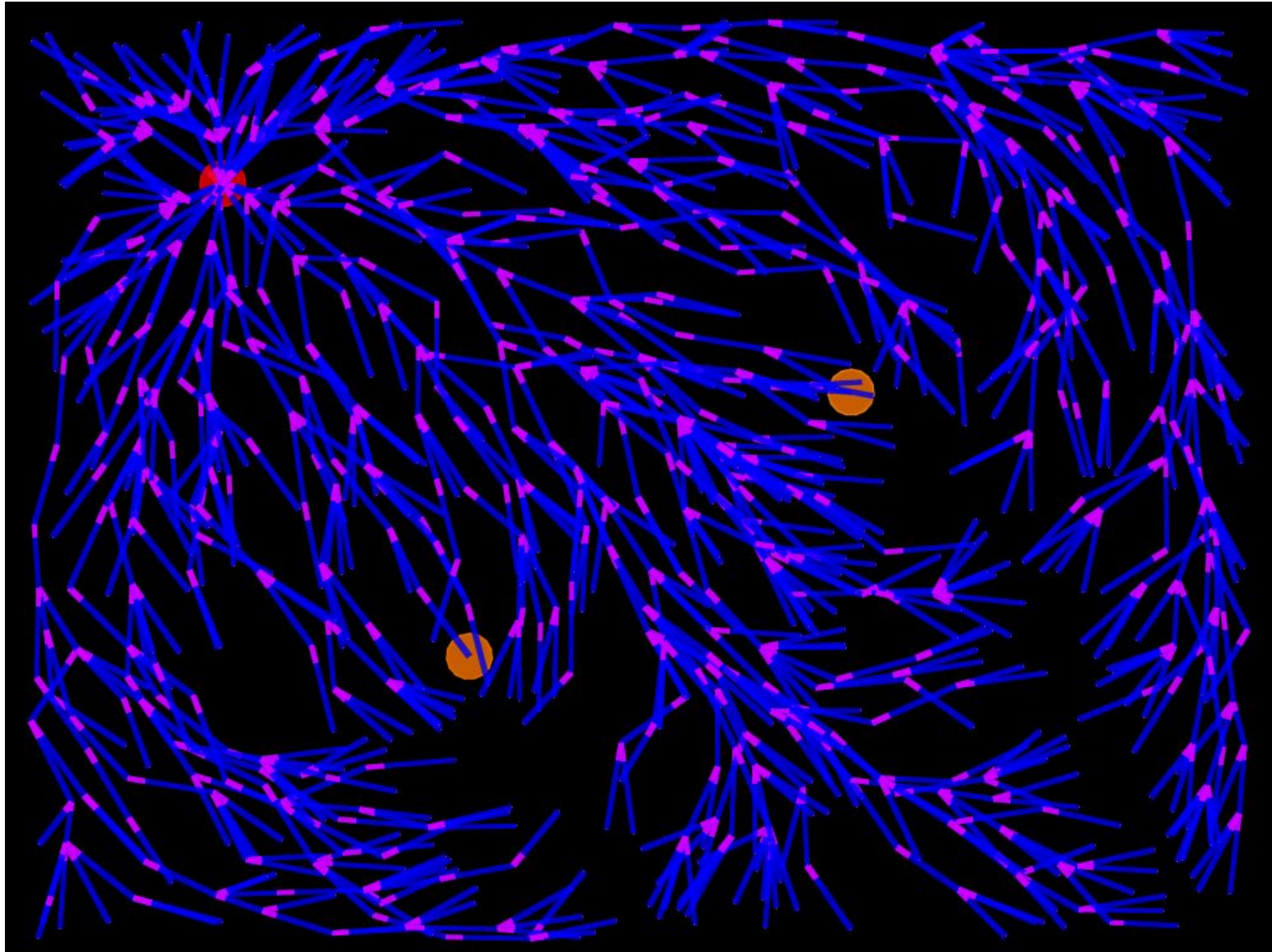
Target Tracking



Target Tracking

```
(def local-average (v) (/ (reduce-nbrs v integral) (reduce-nbrs integral 1)))
(def gradient (src)
  (letfed ((n infinity
            (+ 1 (mux src 0 (reduce-nbrs min (+ n nbr-range))))))
    (- n 1)))
(def grad-value (src v)
  (let ((d (gradient src)))
    (letfed ((x 0 (mux src v (2nd (reduce-nbrs min (tup d x))))))
      x)))
(def distance (p1 p2) (grad-value p1 (gradient p2)))
(def channel (src dst width)
  (let* ((d (distance src dst))
         (trail (<= (+ (gradient src) (gradient dst)) d)))
    (dilate width trail)))
(def track (target dst coord)
  (let ((point
        (if (channel target dst 10)
            (grad-value target
              (mux target
                (tup (local-average (1st coord))
                     (local-average (2nd coord)))
                (tup 0 0)))
            (tup 0 0))))
    (mux dst (vsub point coord) (tup 0 0))))
```

Threat Avoidance



Threat Avoidance

```
(def exp-gradient (src d)
  (letfed ((n src (max (* d (reduce-nbrs max n)) src)))
    n))
(def sq (x) (* x x))
(def dist (p1 p2)
  (sqrt (+ (sq (- (1st p1) (1st p2)))
            (sq (- (2nd p1) (2nd p2))))))
(def l-int (p1 v1 p2 v2)
  (pow (/ (- 2 (+ v1 v2)) 2) (+ 1 (dist p1 p2))))
(def max-survival (dst v p)
  (letfed
    ((ps 0 (mux dst
                  1
                  (reduce-nbrs max (* (l-int p v (local p) (local v)) ps))))
      ps))
  ps))
(def greedy-ascent (v coord)
  (- (2nd (reduce-nbrs max (tup v coord))) coord))
(def avoid-threats (dst coords)
  (greedy-ascent
    (max-survival
      dst
      (exp-gradient (sense :threat) 0.8) coords) coords))
```

Future Directions

- Continuous time evaluation
- Analysis of distortion from space discretization
- Evaluation on a changing manifold
- Actuation of the manifold
- Applications!