

# An Experimental Assessment of a Stochastic, Anytime, Decentralized, Soft Colourer for Sparse Graphs

Stephen Fitzpatrick & Lambert Meertens

Kestrel Institute

3260 Hillview Avenue, Palo Alto, California, U.S.A.

fitzpatrick@kestrel.edu & meertens@kestrel.edu

<http://ants.kestrel.edu/>

## Outline:

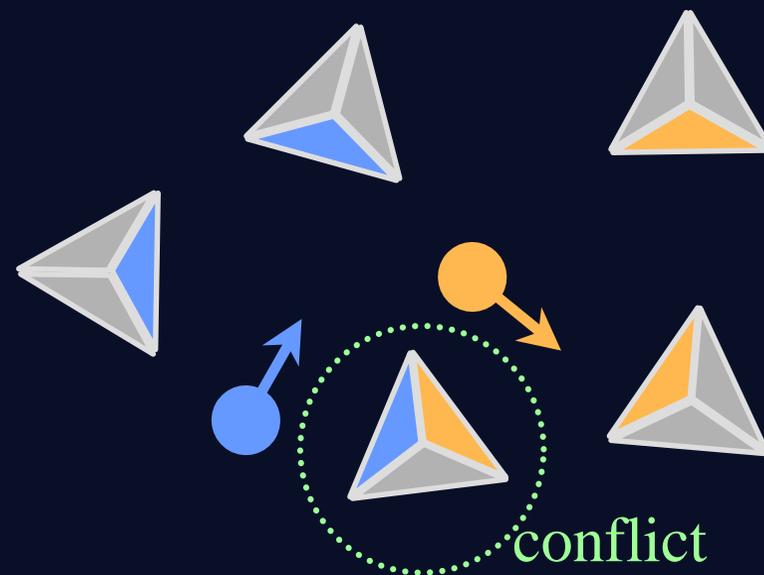
- Motivation: real-time coordination of sensors in a high-latency network
- Modeling coordination as graph colouring
- Soft graph colouring for real-time responsiveness
- A class of stochastic, distributed, anytime algorithms for soft colouring
- Convergence
- Tightness of constraints: deterministic & conservative variants
- Scalability
- Robustness

SAGA 2001, 13-14 December, Berlin

This work is sponsored in part by DARPA through the 'Autonomous Negotiating Teams' program under contract #F30602-00-C-0014, monitored by the Air Force Research Laboratory. The views and conclusions contained in this document are those of the authors and should not be interpreted as representing the official policies, either expressed or implied, of the Defense Advanced Research Projects Agency or the U.S. Government.

# Motivation: Large Networks of Short-Range Sensors

- **Short-range, directional radars**
  - each can scan 1 of its 3 sectors at a time
  - each scan acquires range & radial velocity
  - battery-operated – conservation important
- **Collaboration needed for tracking**
  - 3 approximately-simultaneous scans needed for trilateralization
- **Low-power radio communication**
  - low bandwidth, high latency
  - reveals positions of radars – minimize

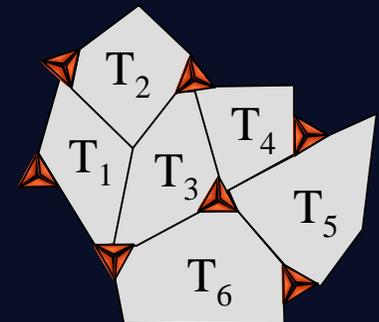


- **Coordination mechanism organizes collaboration**
  - optimizes simultaneous scanning, minimizes costs
- **Must be:**
  - scalable (e.g., to  $10^5$  sensors)
  - real-time adaptive (e.g., new targets are detected, existing targets disappear)
  - robust (e.g., hardware may fail)

# Inter-Sensor Collaboration

- **Main requirement: scan each target simultaneously with 3 radars**
  - define virtual resources: *trackers*
  - each tracker is comprised of 3 sectors on nearby radars
    - $T_i \equiv \{R_{i1}:S_{i1}, R_{i2}:S_{i2}, R_{i3}:S_{i3}\}$
  - each tracker can track a single target over some contiguous region
- **Main constraint: each radar can scan only 1 sector at a time**
  - if two trackers use different sectors on the same radar, they are mutually exclusive
    - $\text{mutually\_exclusive}(T_1, T_2) \Leftrightarrow \exists j, k \in \{1, 2, 3\}: R_{1j} = R_{2k} \wedge S_{1j} \neq S_{2k}$
- **Compute a cyclic schedule of tracker usage**
  - worst-case assumption: all trackers need to be used
  - mutually exclusive trackers cannot be used in the same time slot
  - number of time slots determined by target speed, scan time & revisit period

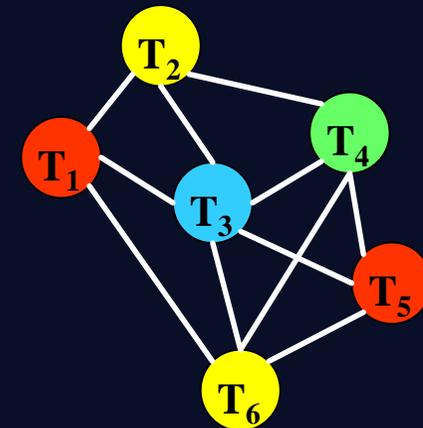
timeslot #	scan start time (seconds)	scan end time (seconds)	T1	T2	T3	T4	T5	T6
1	0.0	2.0	X				X	
2	2.0	4.0		X				X
3	4.0	6.0			X			
4	6.0	8.0				X		



# Modeling Coordination as Graph Colouring

- Each tracker can be mapped to a *node* in an undirected graph
- Each mutual exclusion constraint then maps to an *edge*
  - nodes that are *adjacent* in the graph are mutually exclusive/cannot be used simultaneously
  - two nodes are said to be neighbors iff they are adjacent
- A *proper k-colouring* of the graph's nodes maps to a feasible schedule
  - time slot  $\Leftrightarrow$  integer in  $Z_k \Leftrightarrow$  colour

timeslot #	scan start time (seconds)	scan end time (seconds)	T1	T2	T3	T4	T5	T6
1	0.0	2.0						
2	2.0	4.0						
3	4.0	6.0						
4	6.0	8.0						



# Soft Graph Colouring

- An edge connecting nodes of the same colour represents a *conflict*
  - some radar has been scheduled to scan two sectors simultaneously
- For real-time adaptation, the number of conflicts must be quickly reduced
  - fast reduction to acceptable levels is more important than total elimination
- Define the *degree of conflict* as the fraction of edges that are conflicts
  - let  $E$  be the set of edges and  $C_v$  the colour of node  $v$

$$g \equiv \frac{|\{\{u, v\} \in E \mid C_u = C_v\}|}{|E|}$$



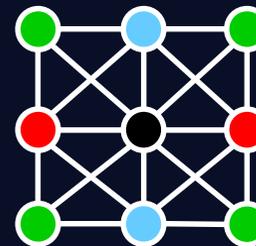
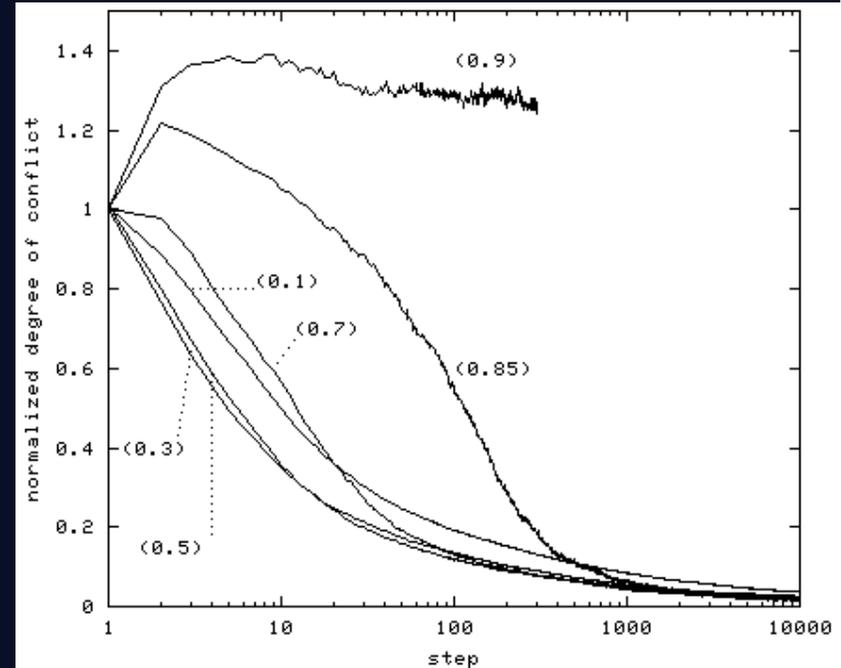
- Normalize:  $\Gamma \equiv k\gamma$ 
  - random  $k$ -colouring has an expected  $\Gamma$  of 1
- Assessment of coordination mechanism is based on how quickly it reduces  $\Gamma$  after random initialization

# A Class of Distributed, Min-Conflict Algorithms

- Main idea: each node repeatedly chooses its own colour to minimize its conflicts with adjacent nodes
- Fixed Probability algorithm FP(p) ...
- Initialization:
  - each node chooses a random colour and informs its neighbours
- Synchronized loop:
  - probabilistic activation
    - a node activates if a randomly generated number falls below some fixed activation level  $p$
  - if a node activates, it non-deterministically chooses its next colour
    - it computes a histogram of colour usage among its neighbours, based on what they last told it
    - it then chooses *any* colour that is least used in the histogram
    - if the chosen colour differs from its current colour, it informs its neighbours
- Convergence?
  - under the right conditions, the total number of conflicts reduces over time and *may* converge to 0 ...

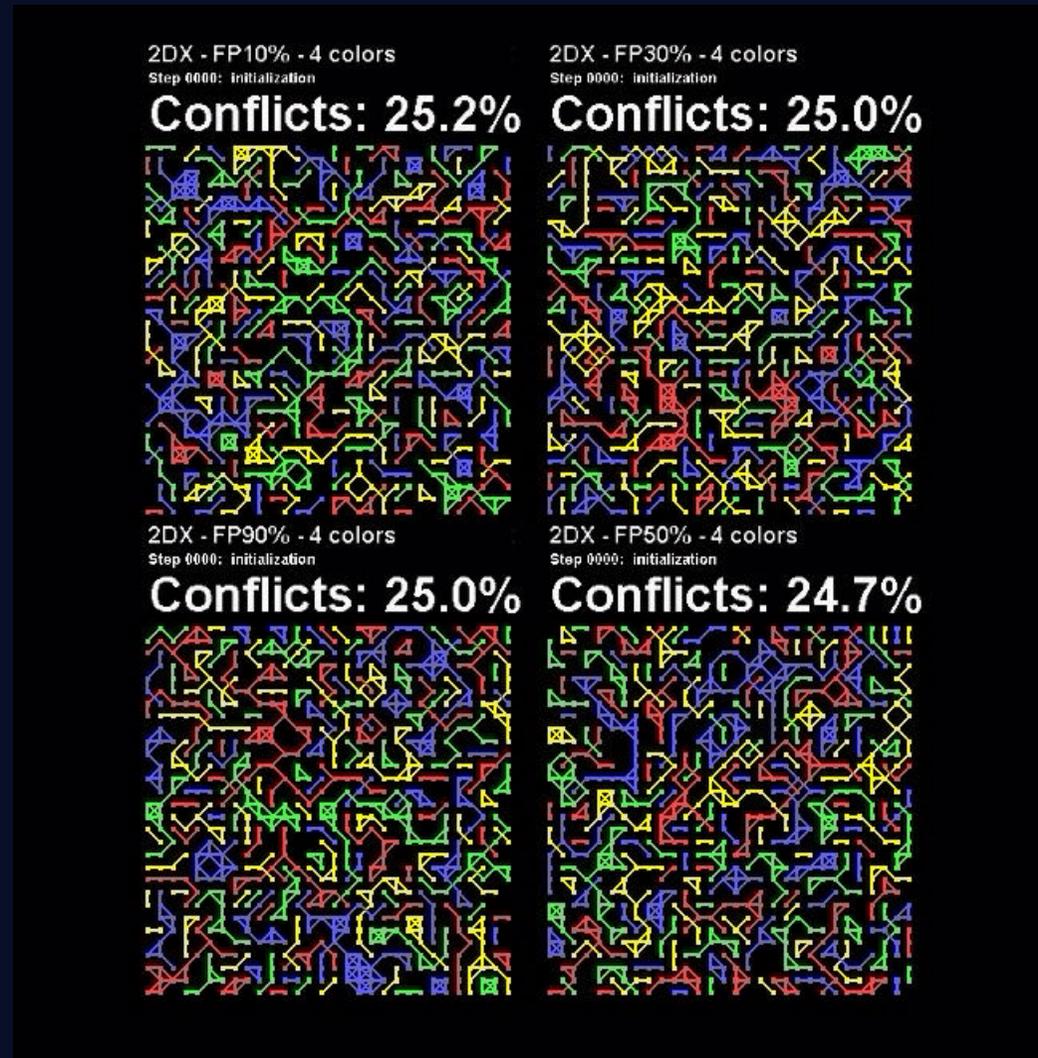
# Effect of Activation Level on Convergence of FP

- Measure (normalized) degree of conflict after each synchronous step
  - experiment performed in simulator
- When activation level is too high, thrashing occurs
  - too many neighbours are simultaneously updating colours
  - because of out-of-date information, they make mutually harmful decisions
- When activation level is too low, adaptivity is hindered
  - extreme case is sequential execution
- Need compromise between speed and coherence
  - an activation level of 0.3 seems to be reasonable for sparse graphs
  - this level was used for experiments reported in following slides



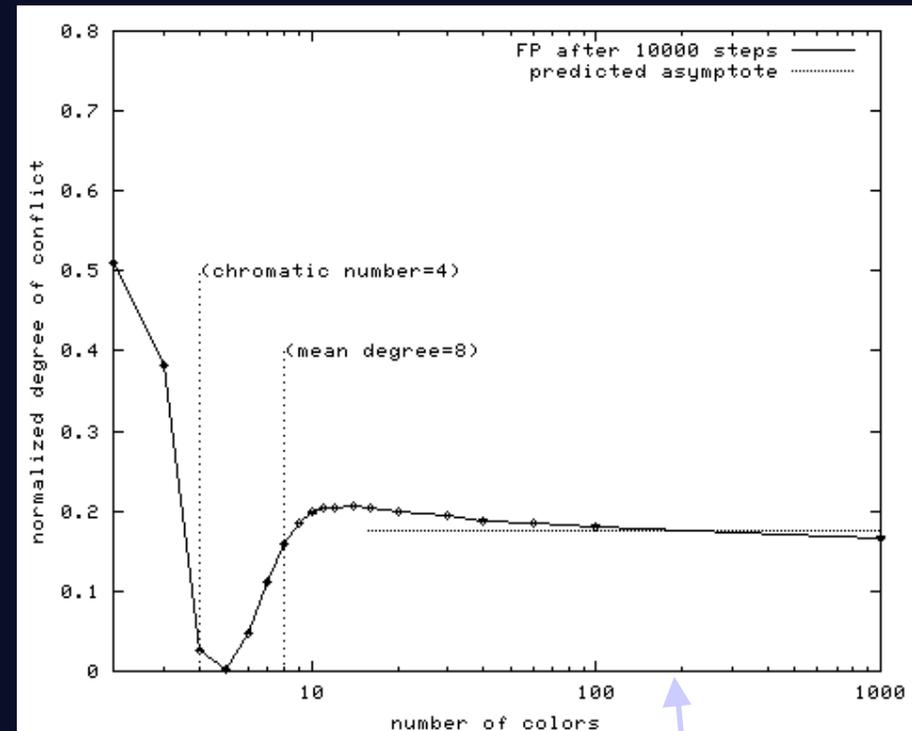
- experimental results shown for 2D grids
  - number of colours = chromatic number = 4
  - 500-5000 nodes
- experiments also performed with random graphs having higher, known chromatic numbers

# Animation: Activation Threshold



# Effect of Tightness of Constraints

- Performance of FP is good on over-constrained problems
  - where  $\#colours < chromatic\ number$
  - for 2D & 3D grids, observed convergence value of degree of conflict is close to theoretical minimum
- Performance of FP is poor on loosely constrained problems
  - where  $\#colours \gg chromatic\ number$
  - intuitively, these are easy problems
- When loosely constrained, each colour choice is essentially random
  - for each given node, most colours are not used by any neighbour
  - FP chooses randomly from among the unused colours
  - asymptotic value predicted as  $\alpha/(2-\alpha)$  where  $\alpha$  is the activation level



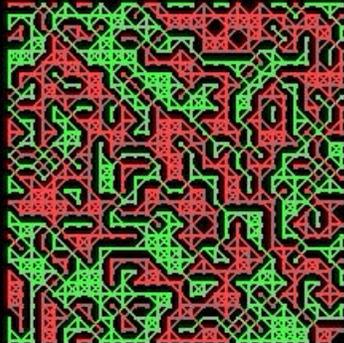
this is *not*  
a time axis

- experimental results shown for 2D grids
- chromatic number = 4

# Animation: Tightness of Constraints

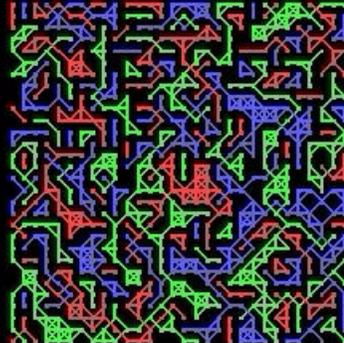
2DX - FP30% - 2 colors  
Step 0000: initialization

**Conflicts: 49.7%**



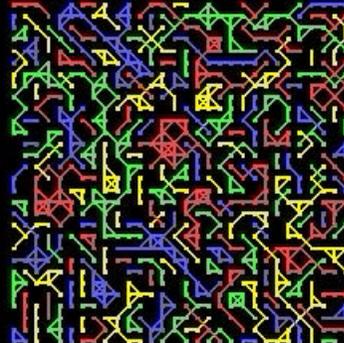
2DX - FP30% - 3 colors  
Step 0000: initialization

**Conflicts: 33.0%**



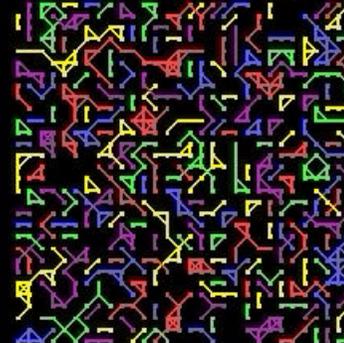
2DX - FP30% - 4 colors  
Step 0000: initialization

**Conflicts: 24.7%**



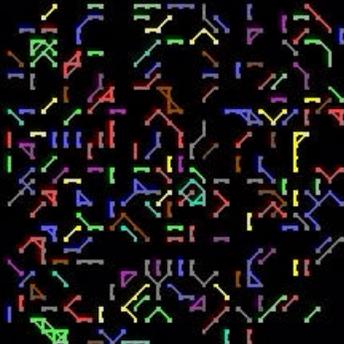
2DX - FP30% - 5 colors  
Step 0000: initialization

**Conflicts: 19.3%**



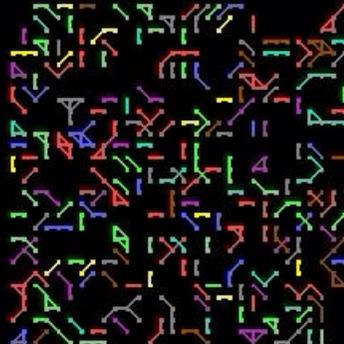
2DX - FP30% - 12 colors  
Step 0000: initialization

**Conflicts: 08.6%**



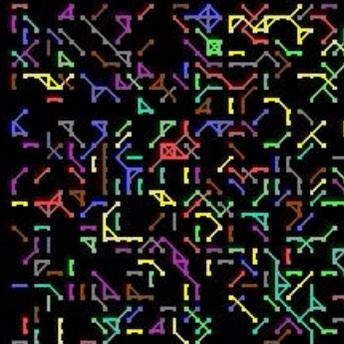
2DX - FP30% - 10 colors  
Step 0000: initialization

**Conflicts: 09.8%**



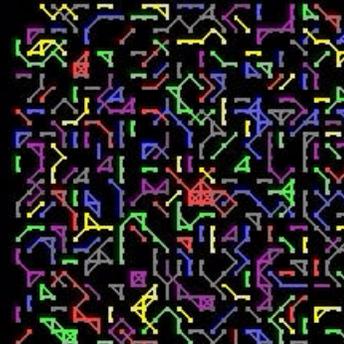
2DX - FP30% - 8 colors  
Step 0000: initialization

**Conflicts: 12.5%**



2DX - FP30% - 6 colors  
Step 0000: initialization

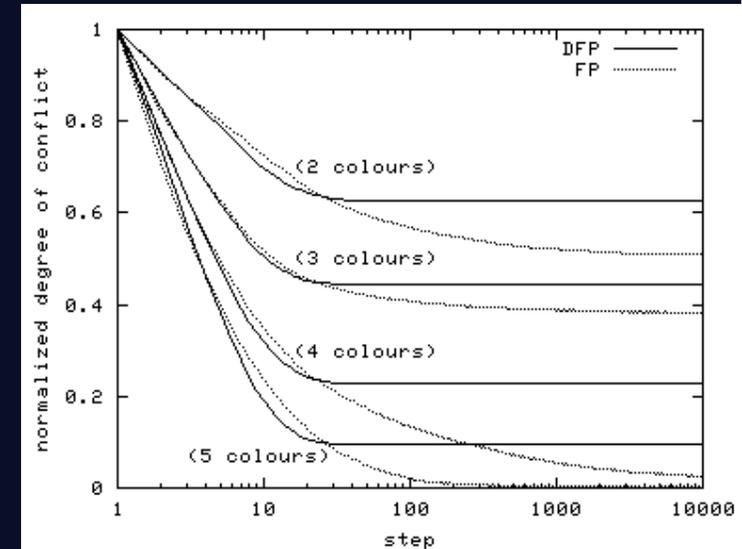
**Conflicts: 16.8%**



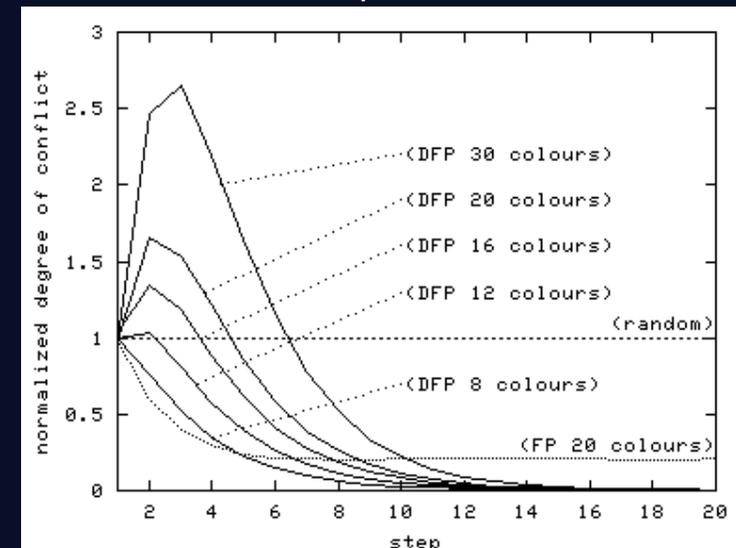
# Deterministic Variant

- **Possible solution: deterministic colour choice**
  - each node chooses the *lowest* colour (in  $Z_k$ ) that minimizes its conflicts
- **Long-term performance:**
  - better than FP when loosely-constrained
  - worse than FP otherwise
    - converges to local minimum
    - randomization techniques can improve convergence values, but at the cost of poor short-term performance
- **Short-term performance is poor**
  - extreme spike in degree of conflict when loosely constrained
    - random initialization causes many neighbours to have the same, unused colour
    - in the next step, those that activate all change to that colour, causing numerous conflicts
  - non-uniform, deterministic choice reduces but does not eliminate this problem

long-term performance



short-term performance



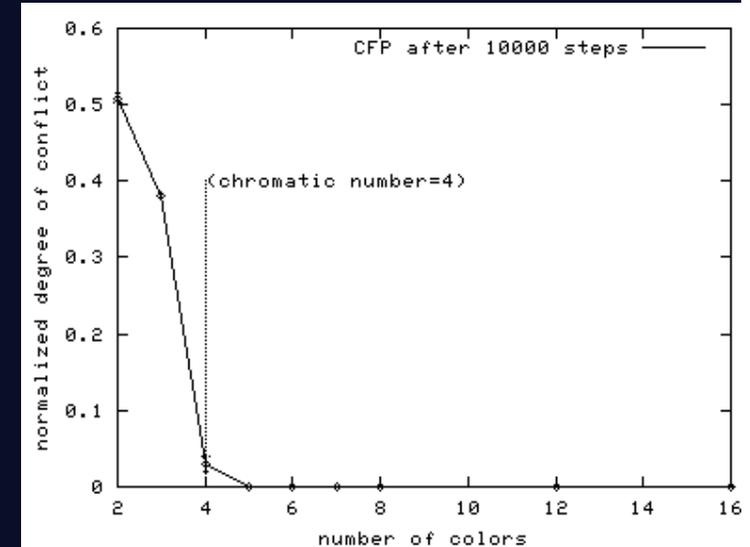
- chromatic number = 4

# Conservative Variant

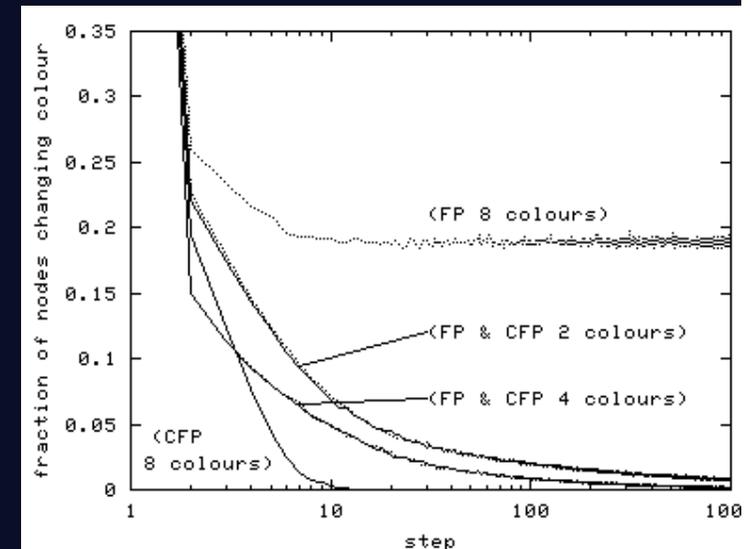
- Colour choice is non-deterministic
- But activation is restricted
  - in addition to passing the test for random number < activation level
  - a node may activate only if it has a conflict with any neighbour
- Conservative variant has good performance overall
  - communication costs are also better than FP's for loosely constrained problems
    - under FP, node activity continues unabated forever
    - under CFP, node activity decreases with the degree of conflict

- experimental results shown for 2D grids
- chromatic number = 4

conflicts



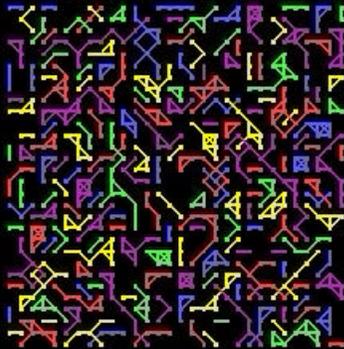
communication rate



# Animation: FP vs. CFP

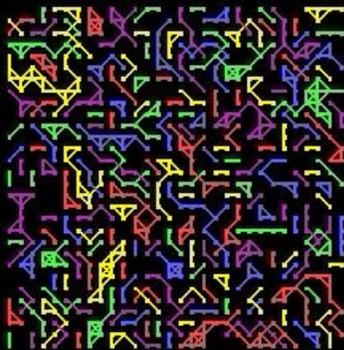
2DX - FP30% - 5 colors  
Step 0000: initialization

**Conflicts: 19.7%**



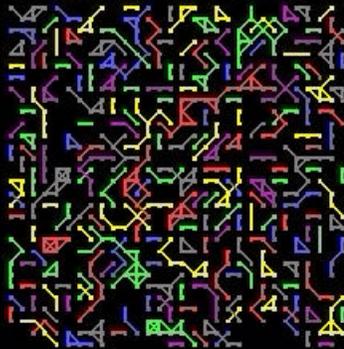
2DX - CFP30% - 5 colors  
Step 0000: initialization

**Conflicts: 20.3%**



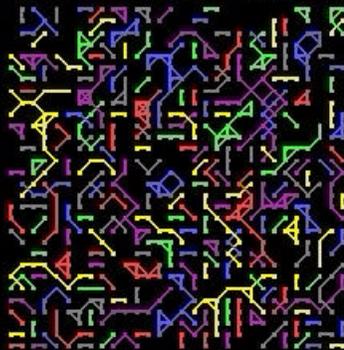
2DX - FP30% - 6 colors  
Step 0000: initialization

**Conflicts: 16.7%**



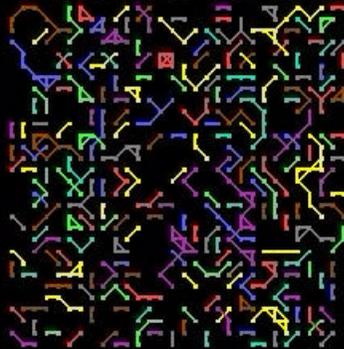
2DX - CFP30% - 6 colors  
Step 0000: initialization

**Conflicts: 16.2%**



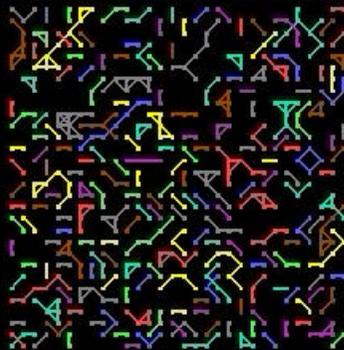
2DX - FP30% - 8 colors  
Step 0000: initialization

**Conflicts: 12.2%**



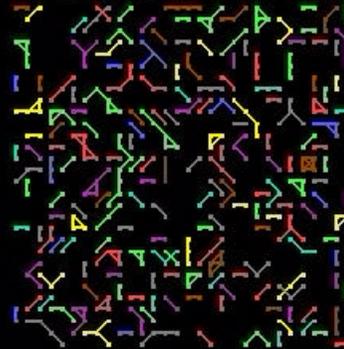
2DX - CFP30% - 8 colors  
Step 0000: initialization

**Conflicts: 12.6%**



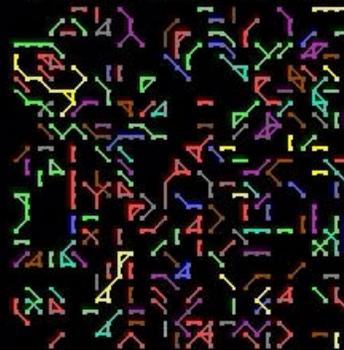
2DX - FP30% - 10 colors  
Step 0000: initialization

**Conflicts: 09.4%**



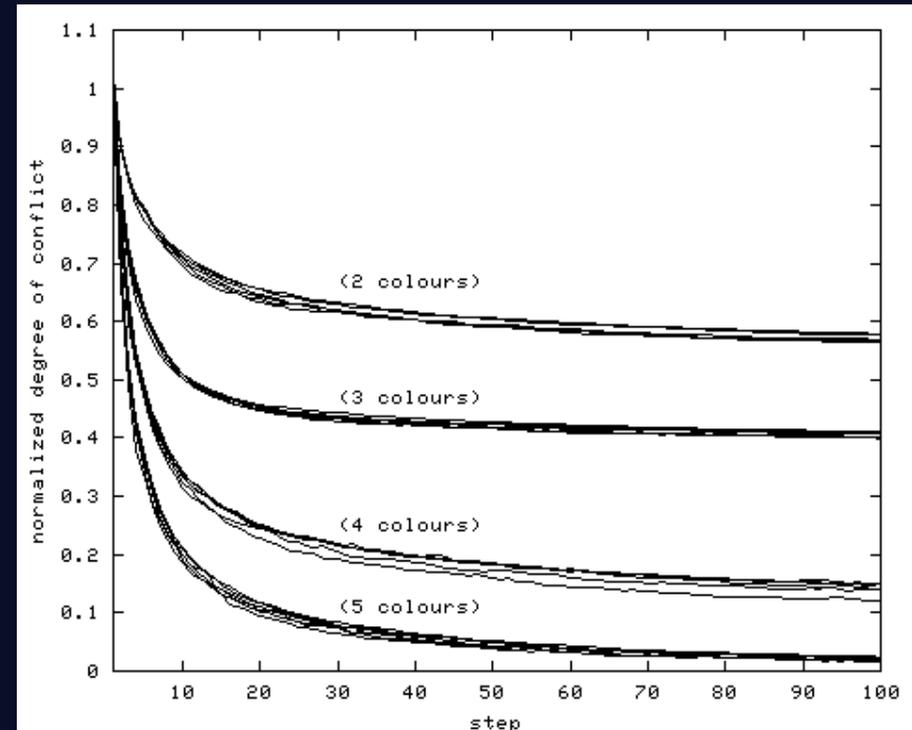
2DX - CFP30% - 10 colors  
Step 0000: initialization

**Conflicts: 09.9%**



# Scalability

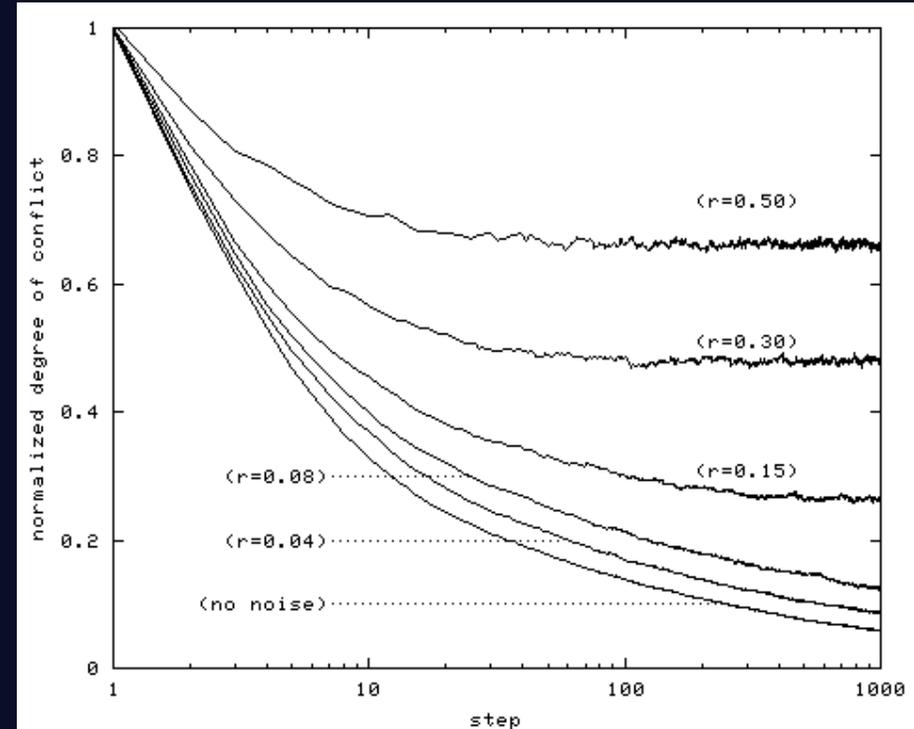
- The algorithm is scalable in cost
  - per node, per step costs depend on (mean) degree of the graph
  - they do not depend on the number of nodes
    - to the extent that the mean degree is independent of the number of nodes
- The algorithm is scalable in performance
  - for large graphs, the reduction in normalized degree of conflict over steps shows little variation for graphs of different sizes



- results shown are for CFP(0.3)
- 6 graphs of different sizes (500-5000 nodes)
  - each graph has chromatic number 4
  - each was coloured using 2, 3, 4 & 5 colours

# Robust against Communication Noise

- Each colour-change message subjected to random process:
  - probability  $r$ , colour randomized
  - probability  $d$ , message lost
  - otherwise, message unchanged
- For small amounts of noise, incremental increases in degree of conflict are observed
  - no catastrophic failure



- results shown are for CFP(0.3) on 2D grids with 4 colours subject to various amounts of message randomization
- similar results were obtained for small amounts of message loss

# Robust against Topology Change

- Simulate the effects of dynamic hardware availability by varying the topology

- initially,  $R$  nodes (and their incident edges) are removed at random and recorded
- then, every  $P$  steps:
  - another  $R$  nodes (and their incident edges) are removed and recorded
  - of the pool of  $2R$  removed nodes,  $R$  are selected at random and restored
  - any removed edges whose end nodes are now present in the graph are restored
- not a complete simulation
  - it does not address the need to reassign tasks that were supposed to be handled by hardware that failed

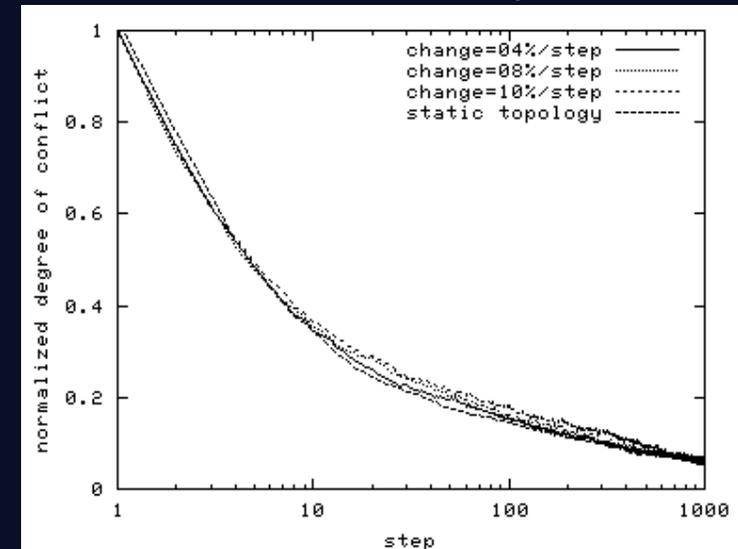
- Continuous change:  $P=1$ , small  $R$

- little effect

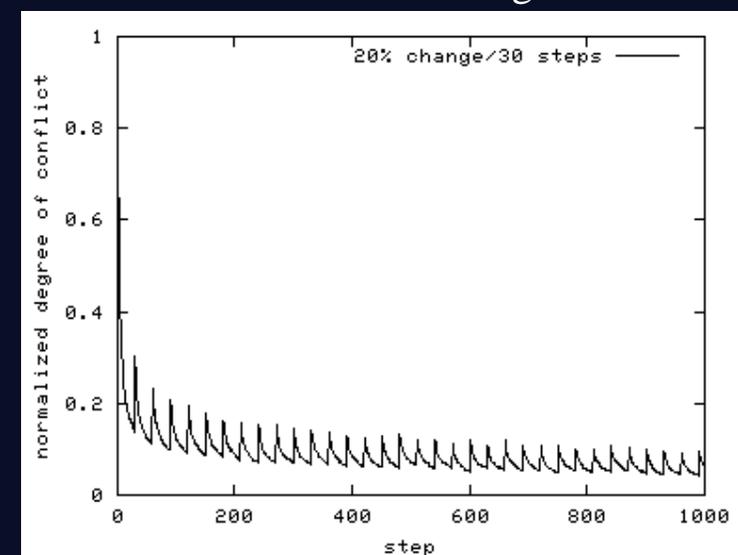
- Intermittent change:  $P=30$ , large  $R$

- spikes in the number of conflicts

Continuous Change



Intermittent Change



# Conclusion

- The CFP algorithm is simple but appears to be effective for distributed, real-time, approximate colouring of sparse graphs
  - scalable, low-cost, robust
  - effective on under-, critically- and over-constrained problems
- Basic framework of stochastic activation & local optimization seems appropriate for other distributed constraint problems
  - graph colouring serves as a clean, archetypal problem
- The algorithm has also been tested with dense, random graphs
  - interesting, but different, results
  - proper k-colourings quickly obtained for very dense k-colourable graphs
    - local constraints sufficient to guide colouring to a unique, proper colouring
- Further work on experiments
  - other types of graphs and/or constraints
  - lower bounds for over-constrained problems on random graphs
- Further work on algorithm
  - non-uniform activation levels, perhaps determined dynamically from local metrics