

A Simulated Annealing Algorithm For Cartographic Map Generalization With Multiple Operators

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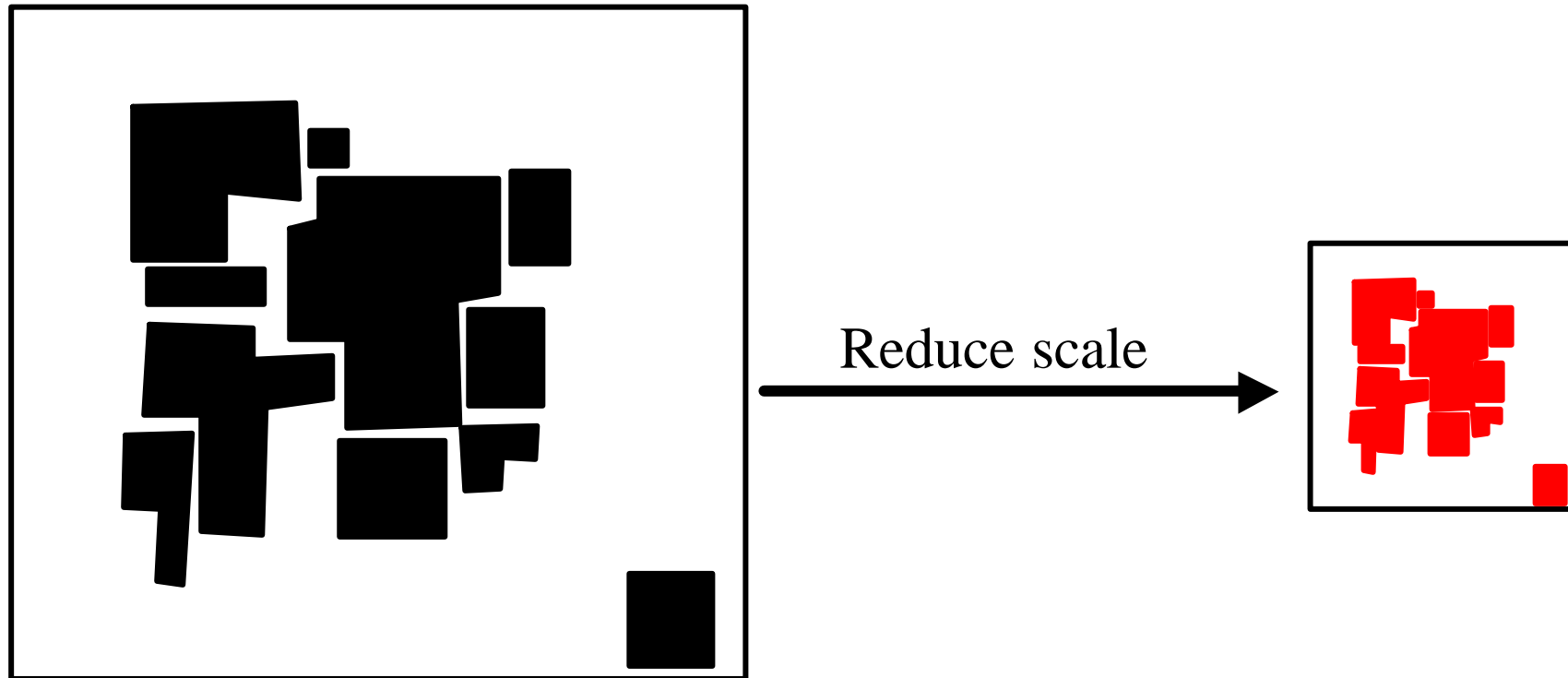
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Ordnance Survey

THE PROBLEM

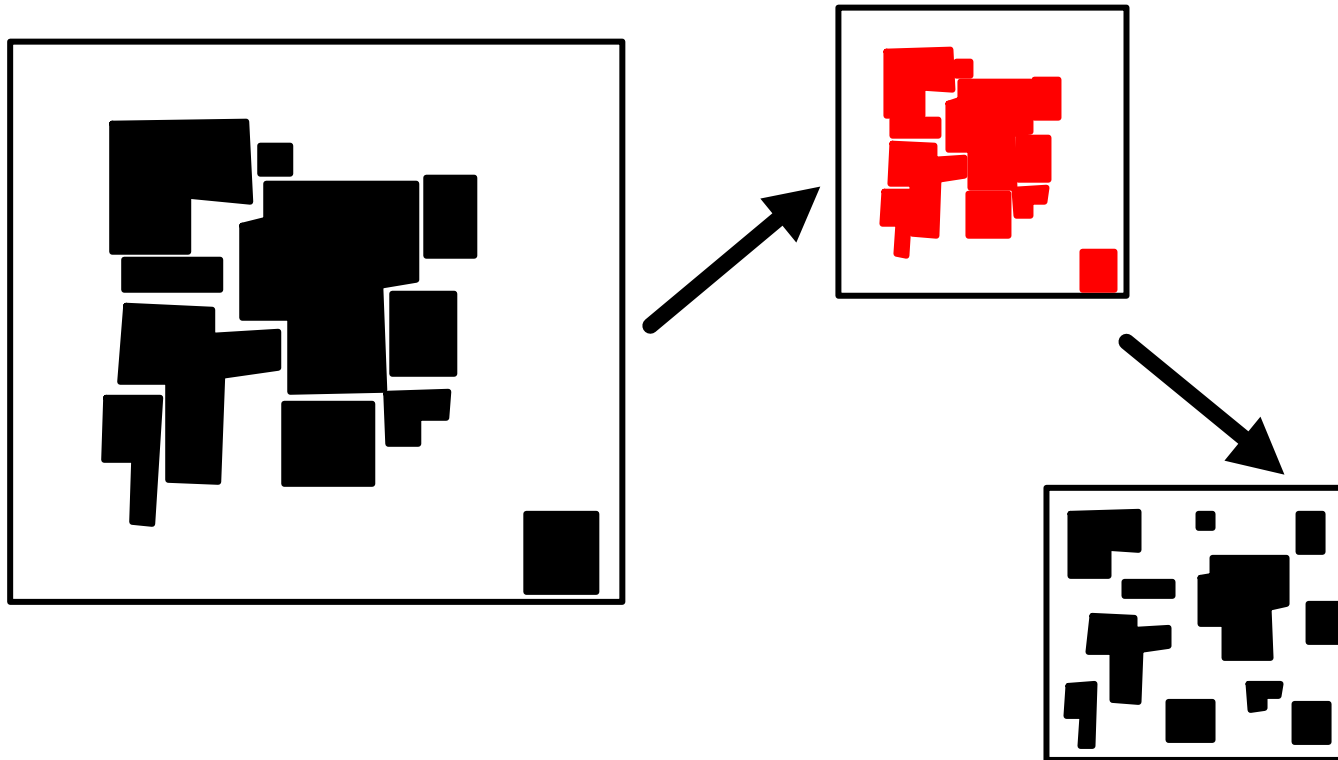


spatial conflict due to
– objects lying too close to each other

MAP GENERALIZATION - THE SOLUTION

- simplification
- amalgamation
- reduction
- typification
- deletion
- displacement
- resize

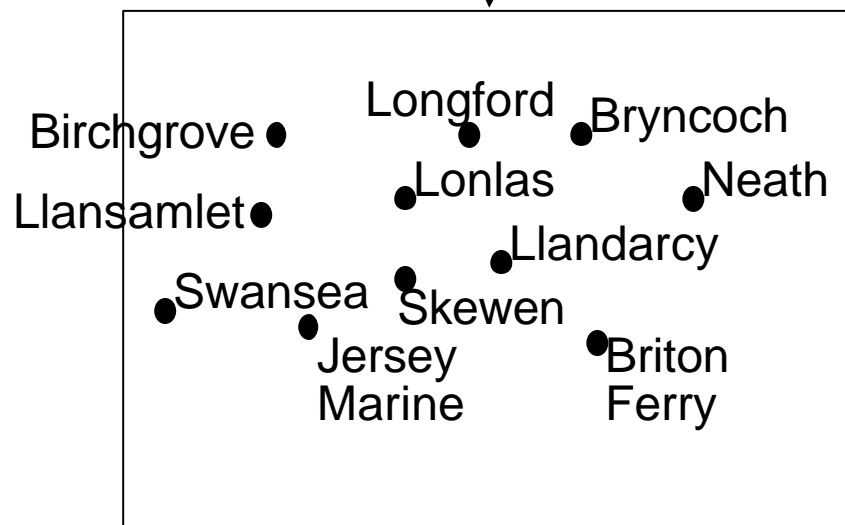
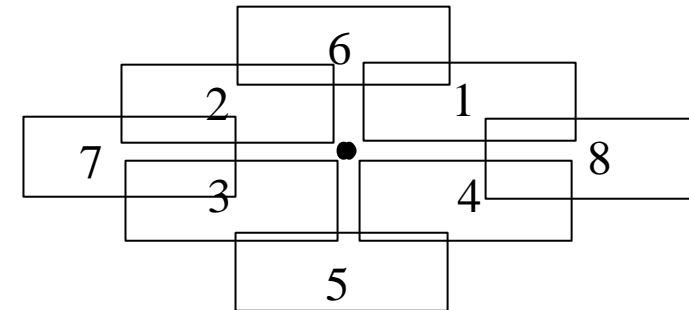
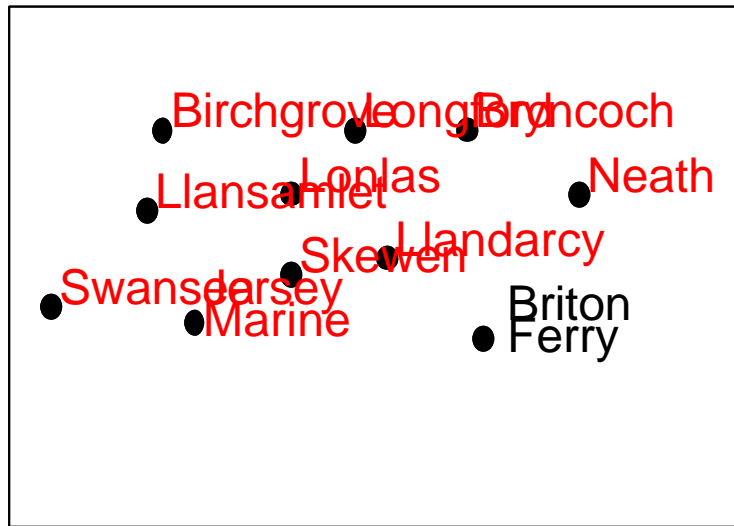
CONFLICT RESOLUTION BY OBJECT DISPLACEMENT



resolve by displacing one or more objects
(assume it is permissible to move each object up to a
predefined maximum distance from its origin)

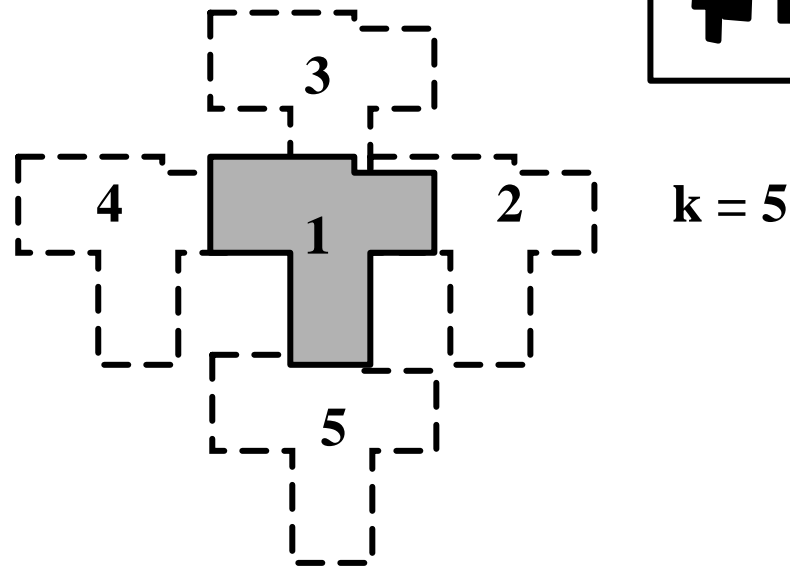
OBJECT DISPLACEMENT USING TRIAL POSITIONS

Point Feature Label Placement Using Trial Positions

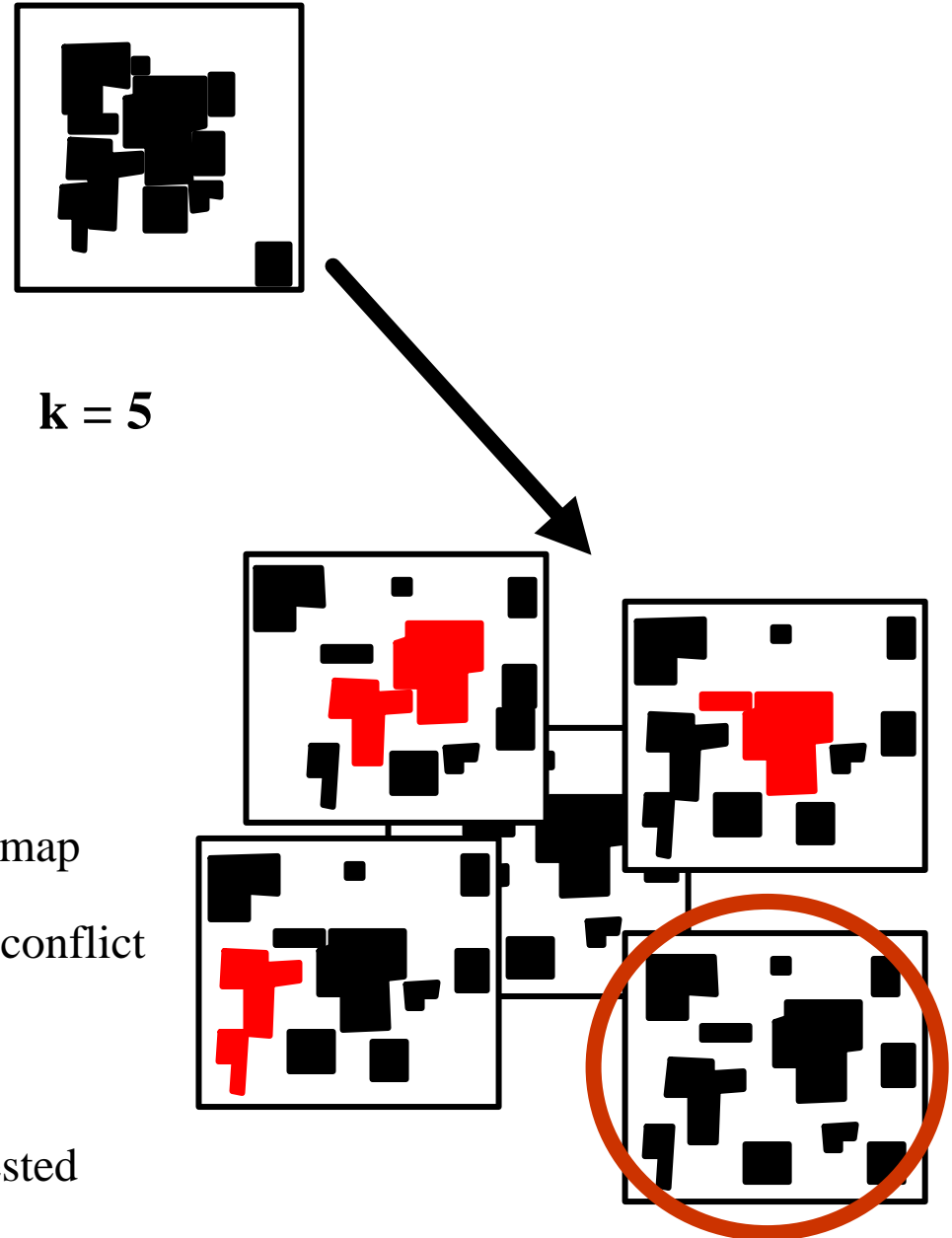


Apply PFLP trial position techniques to Object Displacement

- we have map, n objects, containing conflict
- assign each object k trial positions



- there will be k^n alternative realisations of the map
- hopefully some will contain reduced levels of conflict
- too many to generate and test all
(e.g. $k=8$, $n=10$, > 1 billion configurations)
-need some strategy for limiting number tested



A SIMULATED ANNEALING APPROACH

function **SimulatedAnnealing**

input: D_{initial} , Schedule, Stop_Conditions

$D_{\text{current}} \leftarrow D_{\text{initial}}$

$T \leftarrow \mathbf{initialT}(\text{Schedule})$

while **NotMet**(StopConditions)

$D_{\text{new}} \leftarrow \mathbf{RandomSuccessor}(D_{\text{current}})$

$\Delta E \leftarrow C(D_{\text{current}}) - C(D_{\text{new}})$

if $\Delta E > 0$ *then* $D_{\text{current}} \leftarrow D_{\text{new}}$

else

$P = e^{-\Delta E / T}$

$R = \text{Random}(0,1)$

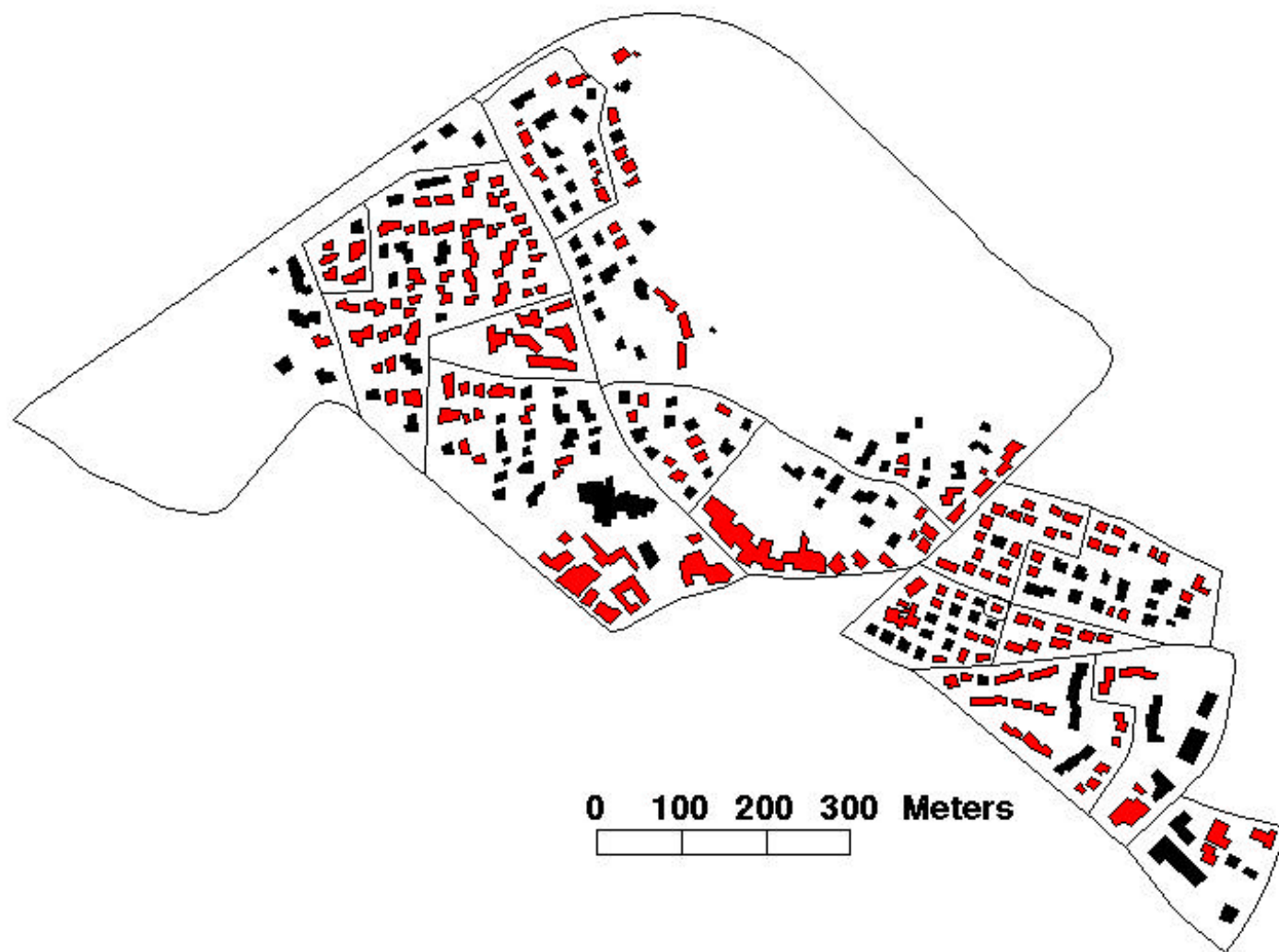
if $(R < P)$ *then* $D_{\text{current}} \leftarrow D_{\text{new}}$

end

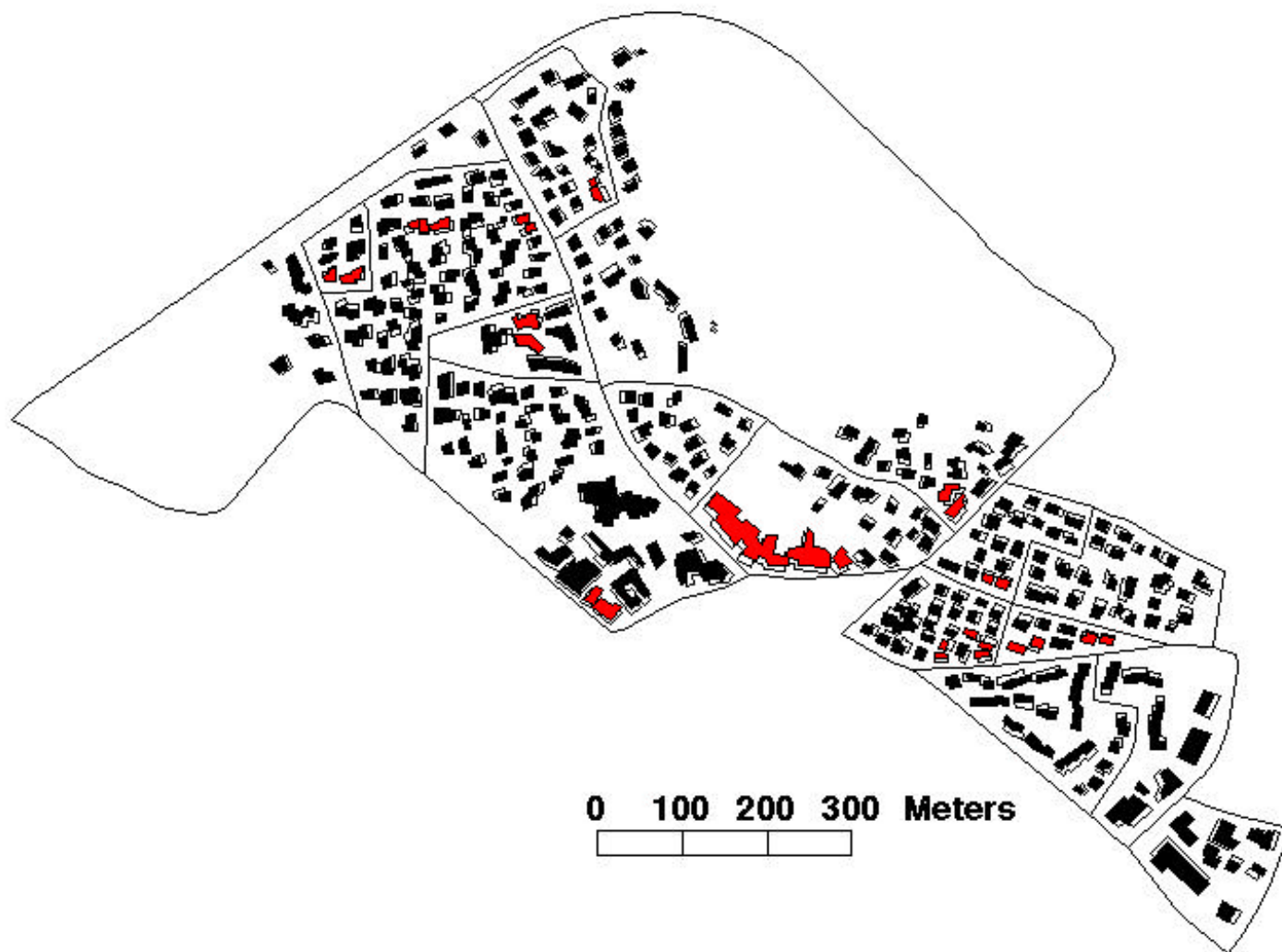
$T \leftarrow \mathbf{UpdateT}(\text{Schedule})$

end

Return(D_{current})

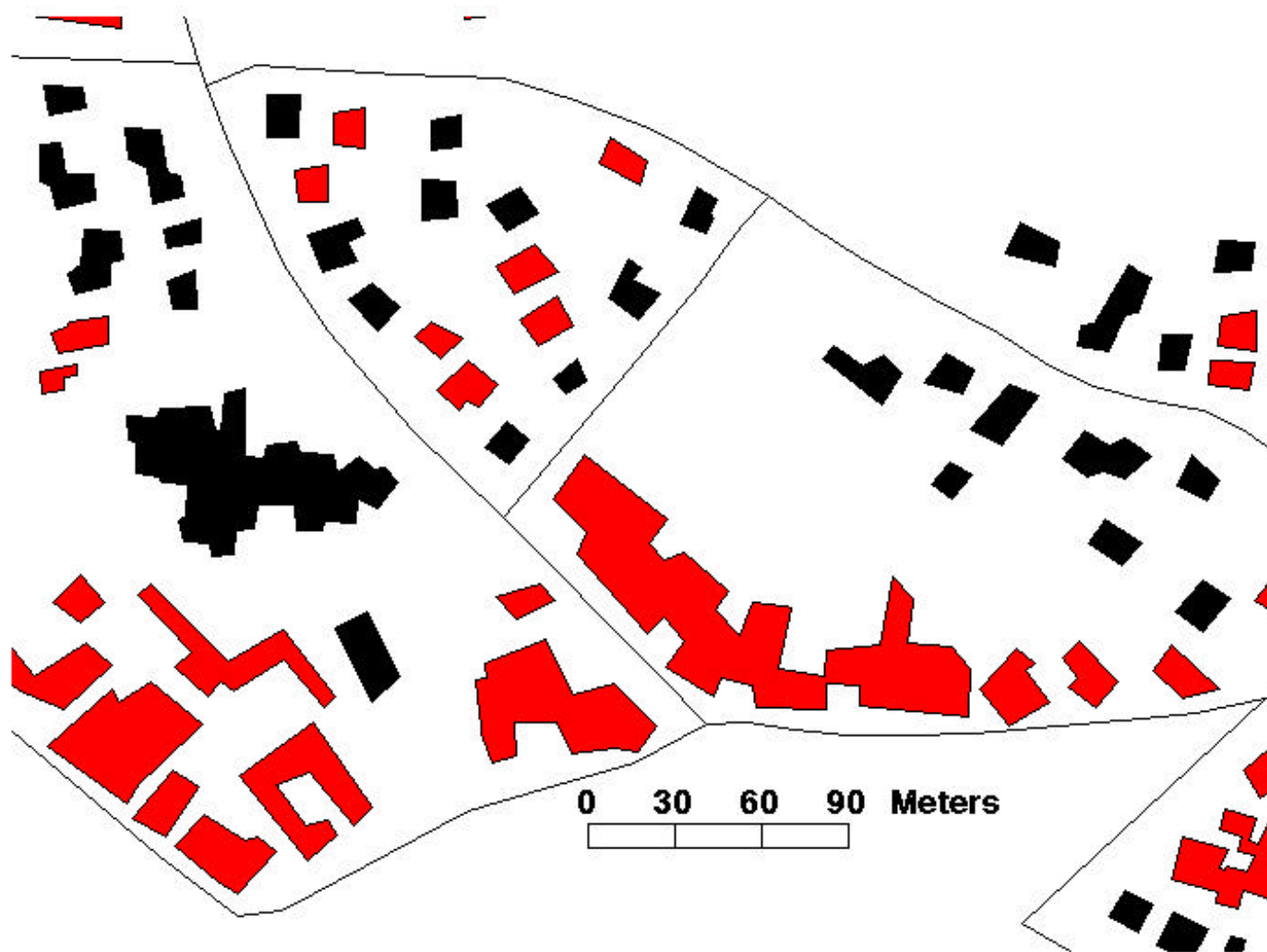








DISPLACEMENT COST

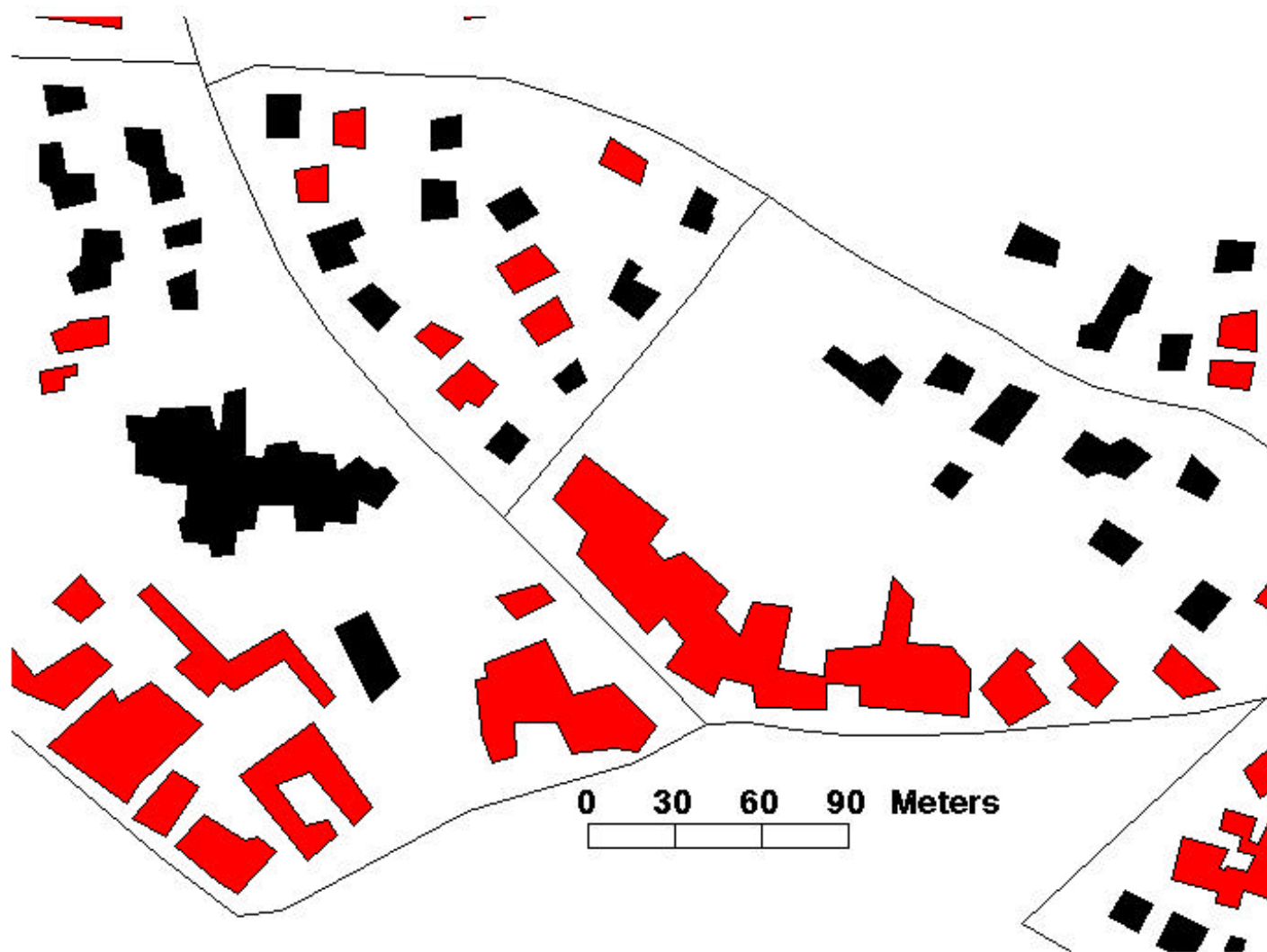






ADDITIONAL OPERATORS

DELETION



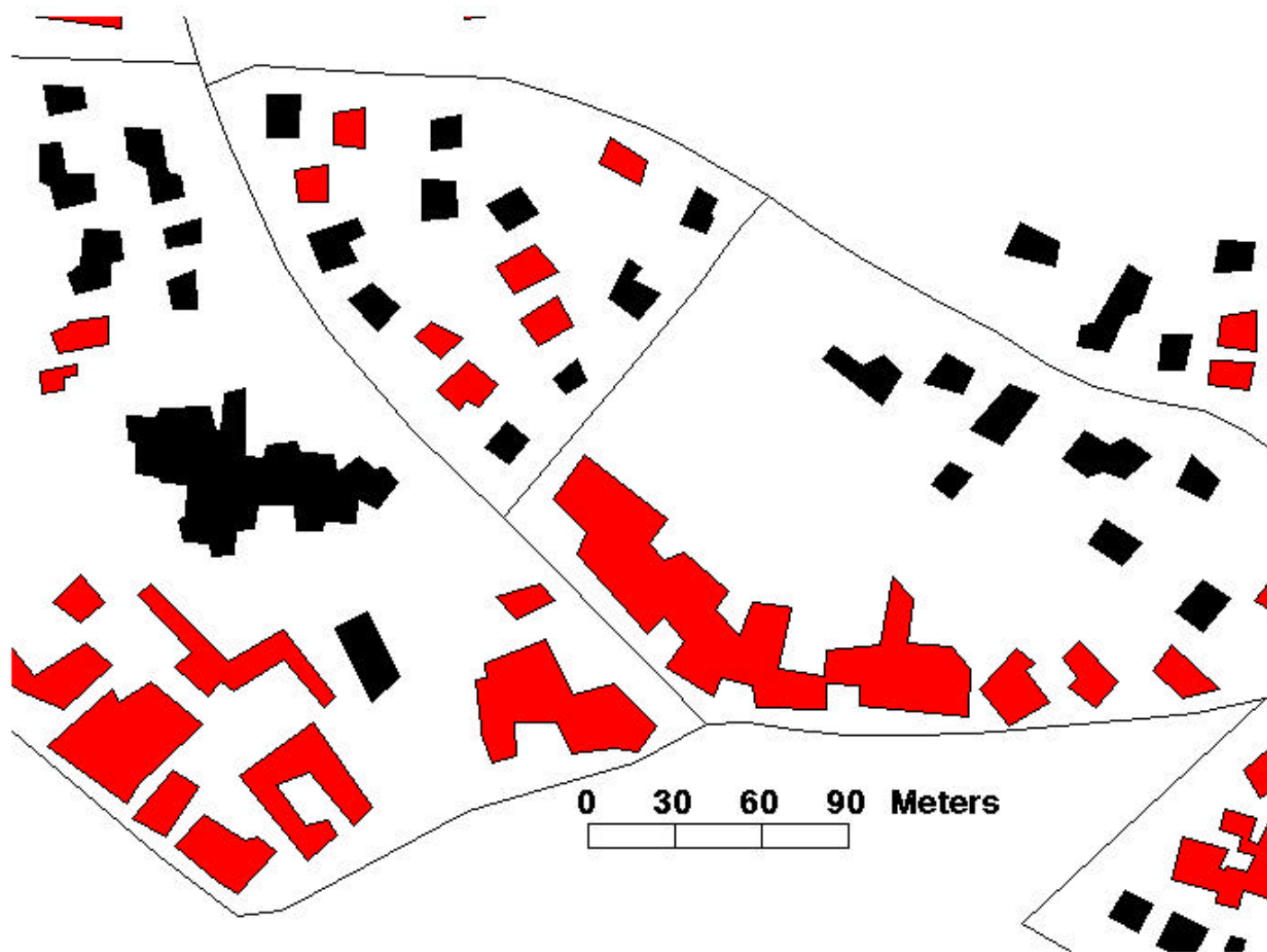


IMPORTANCE WEIGHTING





SCALE/RESIZE





COST SETTING











CURRENT AND FUTURE WORK

- **Additional operators**
e.g. amalgamation
- **Additional feature types**
e.g. lines
- **Constraints**
e.g. feature alignment
- **Higher/global level control**
e.g. staggered use of operators
- **Alternative optimisation**
e.g. tabu

EXECUTION TIME IMPROVEMENT

Original Simulated Annealing Results

- Total number of configurations = 29^{321}
 - Number of configurations evaluated = 342000
 - Average Cost = 27 (best result = 22)
 - Time taken = 40s (as reported in GeoInformatica 1998)
- too slow !

Improvement 1 - run on a faster machine

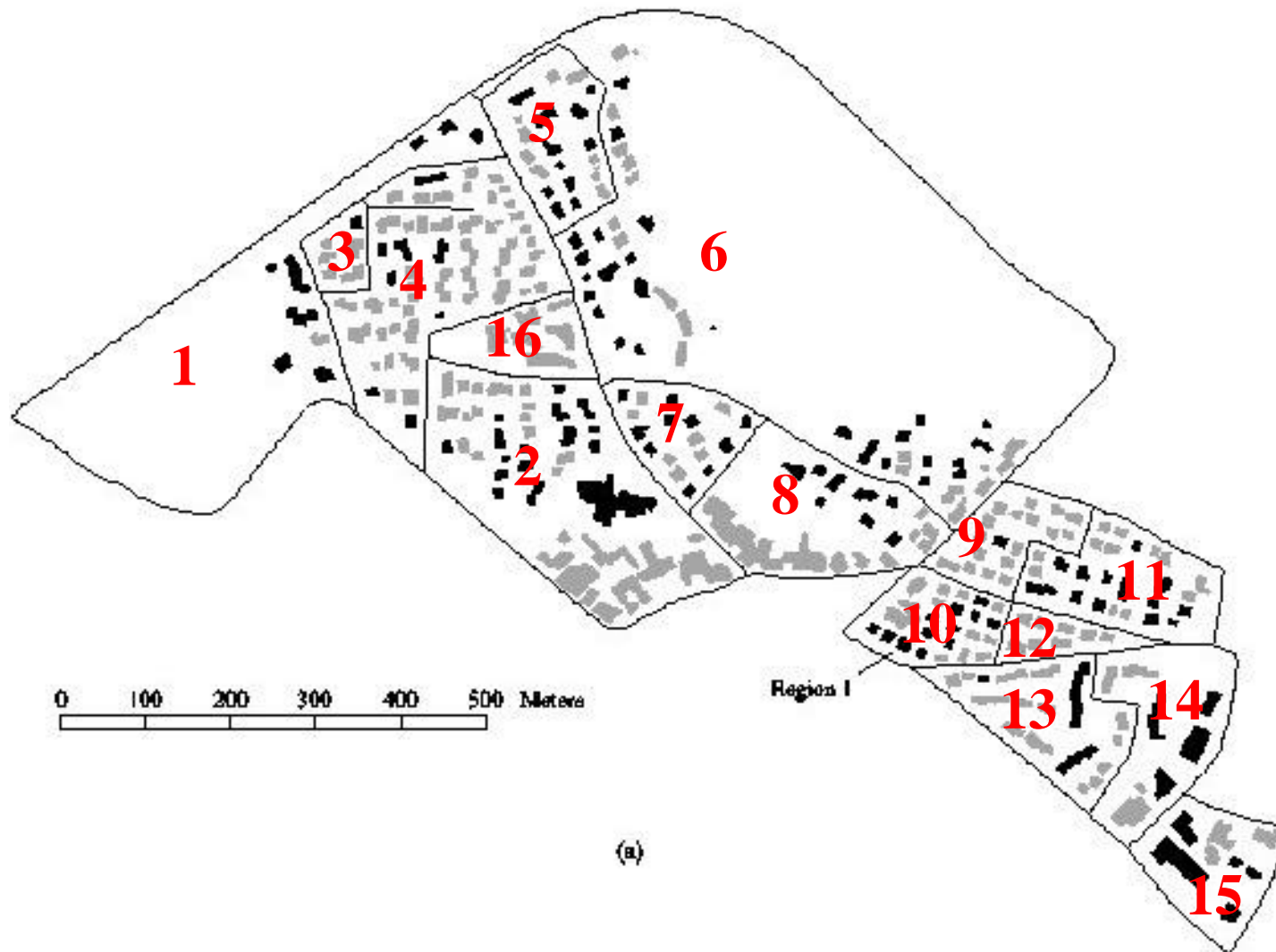
- Total number of configurations = 29^{321}
- Number of configurations evaluated = 341000
- Cost = 26
- Time taken = 13.5s

Improvement 2 - segment data

- Number of configurations evaluated dictated by annealing schedule and problem complexity
- Annealing schedule -
 - Initial temperature T
 - Number of evaluations at each temperature
 - Temperature reduction factor
- Difficult problem will require many configurations
 - i.e. annealing schedule
 - high initial T
 - many evaluations at each temperature
 - small reductions in T
- Simple problem will require few configurations
 - i.e. annealing schedule
 - low initial T
 - few evaluations at each temperature
 - large reductions in T

- If data is processed as a whole, annealing schedule must be set so as to be able to deal with most difficult part of data - leading to processing redundancy in parts where problem is simple
- Segment data - a separate, appropriate, annealing schedule for of each data subset
- Data segmented into autonomous regions
i.e. an object in a particular region can never come into conflict with object belonging to any other region

Data segmented into 16 autonomous regions



Segmentation Results

- Total number of configurations = 29^{321}
- Number of configurations evaluated = 79000
- Cost = 27
- Time taken = 3.2s
 - 75% saving

Problem

- Each of 16 annealing schedules arrived at via experimentation
- Need some method for automating the setting of annealing parameters (lots of work on this in general SA literature, such as automated setting of initial temperature T)

Improvement 3 - two stage annealing

- Simulated annealing

- high temp gets you to a low cost area in solution space
- low temp gets you to the local minimum

- Many authors suggest low temperature start annealing

- Need some method to stop solution from immediately getting caught in a local minimum

- Two stage annealing - replace annealing actions taking place at higher temperatures with a faster heuristic algorithm

- fast heuristic algorithm - locates low cost area in solution space
- simulated annealing (low initial T) - locates local minimum

TSSA algorithm

Stage 1

simulated annealing

- high initial temperature
- rapid cooling

Stage 2

simulated annealing/sintering

- low initial temperature
- gradual cooling

TSSA Results

- Total number of configurations = 29^{321}
- Number of configurations evaluated = 74000
- Cost = 26
- Time taken = 3.1s
 - 75% saving

Improvement 4 - combine segmentation & TSSA

(i.e. apply TSSA to each of the 16 regions in turn)

- Total number of configurations = 29^{321}
- Number of configurations evaluated = 37000
- Cost = 26
- Time taken = 1.6s
 - 88% saving