

Fire Risk Management Conference 2010

Multi-Objective Decision Making for the Fire and Rescue Services – A Scoping Study

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Multi-Objective Decision Making for the Fire & Rescue Service

Contents



- Introduction
- Problem scale
- Evolutionary Algorithms
- Software Development
- Conclusions and Future Work

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Multi-Objective Decision Making for the Fire & Rescue Service

Introduction

- Strategic (long term) decision making for Fire and Rescue Service Resources
 - Fire Station location
 - Appliance location
 - Crewing types (wholetime or retained)
 - Specialised equipment locations
 - Response types
- Large problem
- Many potential solutions

Multi-Objective Decision Making for the Fire & Rescue Service

Introduction

- Current tools available for strategic decision making
 - Fire Service Emergency Cover toolkit (software) FSEC
 - Measures effectiveness of a particular scenario based on life and property loss
 - Based on statistical incident data
 - Only allows user to evaluate one option at a time
 - Run-times are long (approximately 20-30 minutes for a typical FRS area)
- Is this a problem?

Multi-Objective Decision Making for the Fire & Rescue Service

Contents

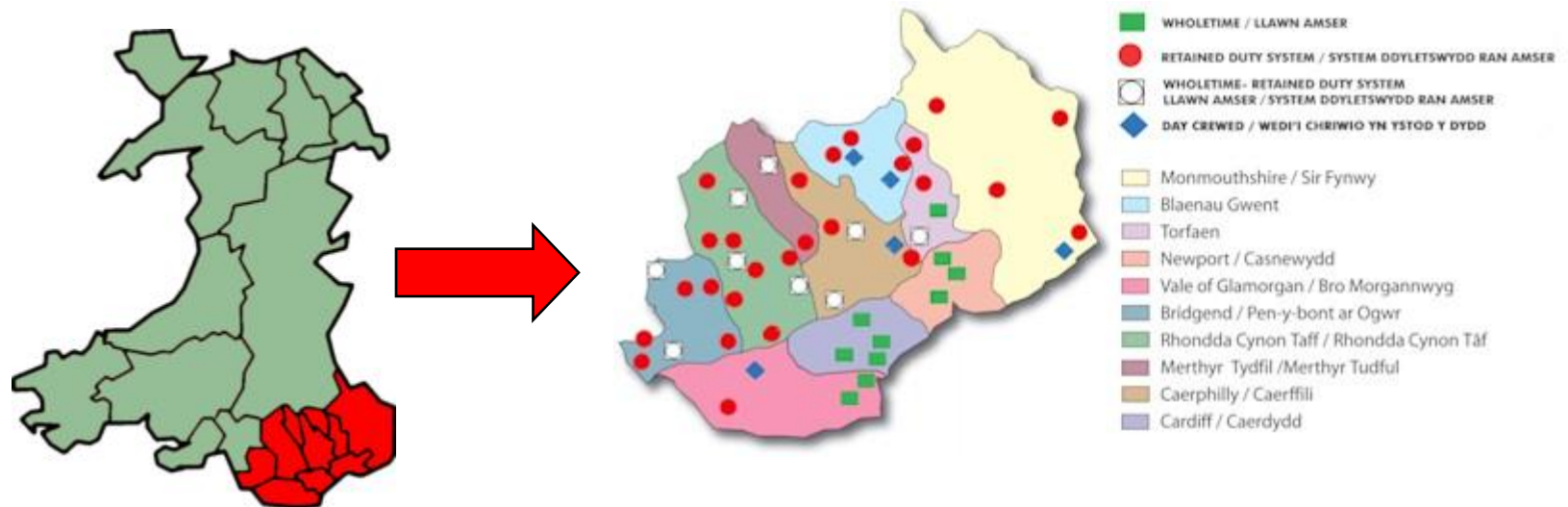


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Problem Scale

- Typical brigade – e.g. South Wales FRS
 - Approximately 50 stations
 - 19 wholetime, 5 day crewed, 26 retained
 - 1000 full time firefighters
 - 600 retained



Problem Scale

- Assuming 70 station sites (i.e. current 50 sites plus 20 potential sites)
- 50 stations to be placed in suitable locations – how many combinations of 50 stations can be selected from 70 sites?

$$N_s = {}_{70}C_{50} = \frac{70!}{50!(70-50)!} \approx 10^{17}$$

- But each station can have a variety of configurations based on crewing type, vehicle allocation etc
- Conservative estimate would suggest 6 different station configurations

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Problem Scale

- Each set of 50 stations can therefore be configured in 6^{50} ways ($\approx 10^{38}$)

- Thus total configurations is

$$N = N_s \times N_c = 10^{17} \times 10^{38} = 10^{55}$$

- Some of these configurations may not be feasible, or may include near-duplicates.
- However, it is impossible to even evaluate 10% of the total number of solutions manually.
- The only feasible way of finding good solutions is via the use of some form of search algorithm
- Evolutionary algorithms have been chosen for this work

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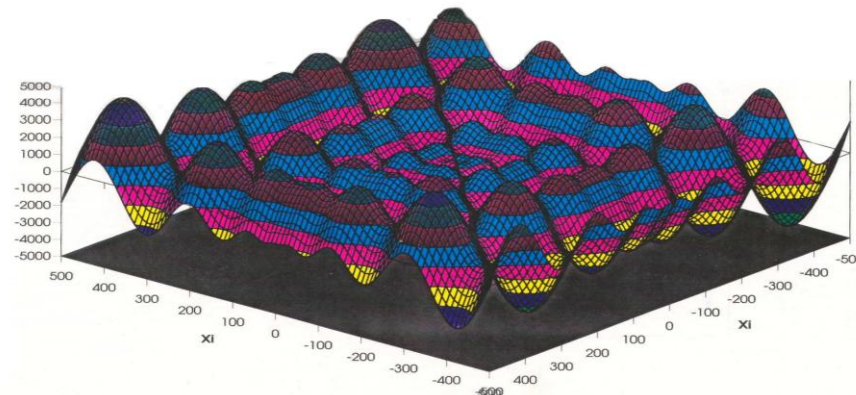
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Evolutionary Algorithms

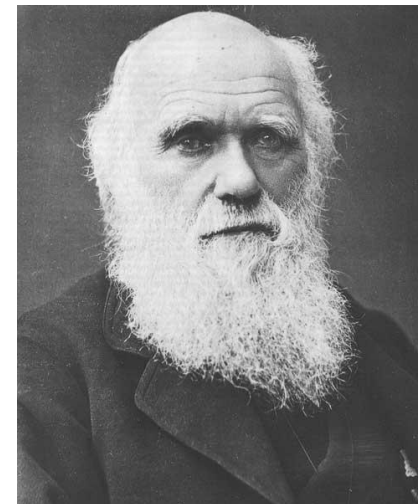
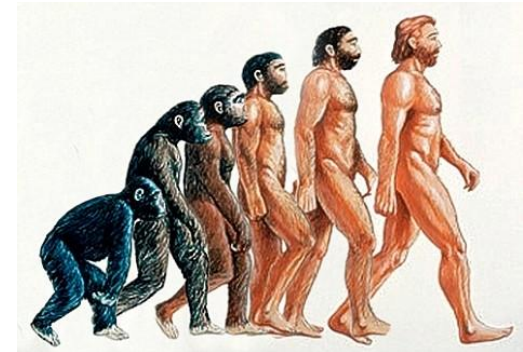
- Bio-inspired search algorithms
 - e.g. Ant colony, particle swarm analysis, genetic algorithms
- Ideal for complex problems
 - do not require fully-defined objective function
 - use a “fitness function” as a means of judging whether one solution is better than another
 - avoid getting stuck on local optima



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Evolutionary Algorithms

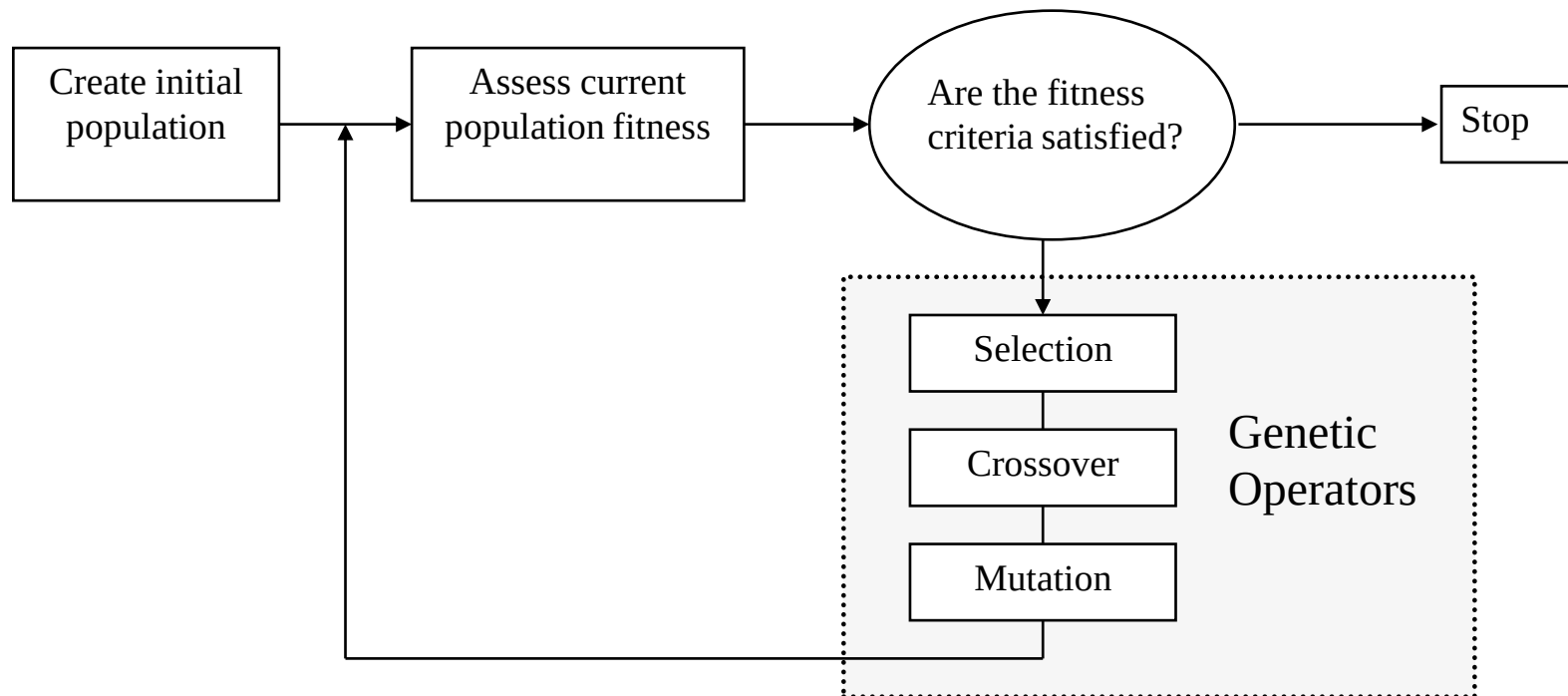
- Genetic Algorithms chosen for this work
- Mimic Darwinian evolution (i.e. survival of the fittest)
- Starts with a population of random solutions
- Population gradually evolves by selection, breeding and mutation of the best solutions at each generation
- Only has to sample a small proportion of the total possible solutions



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Evolutionary Algorithms

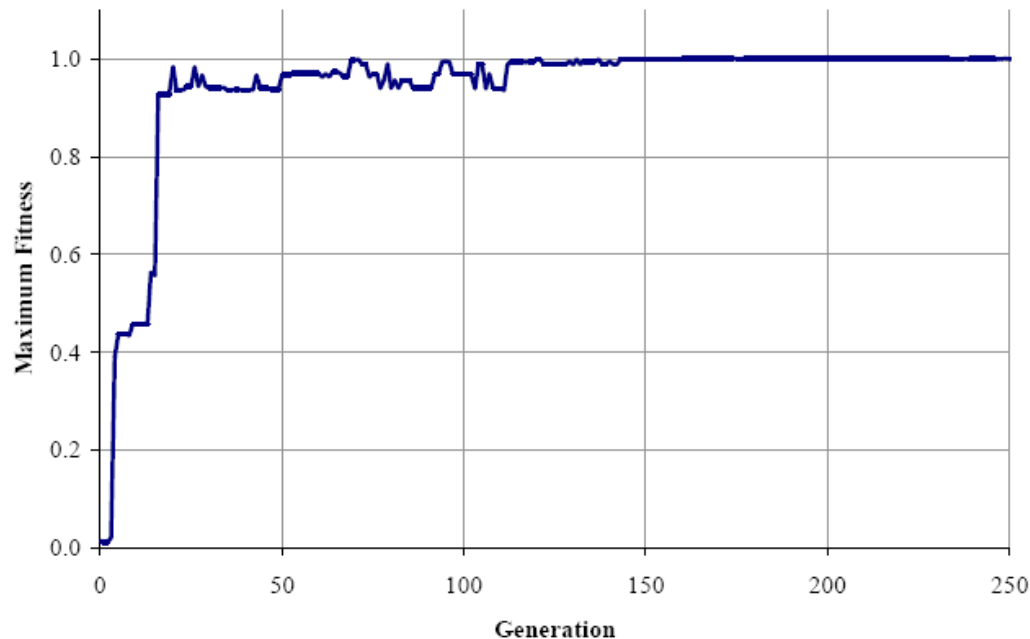
- Genetic algorithm (GA) architecture



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Evolutionary Algorithms

- GA developed for this project
- Tested using simple fitness functions with known solutions
- e.g. maximise $\sin(x)$ where $0 < x < \pi$



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Software Development

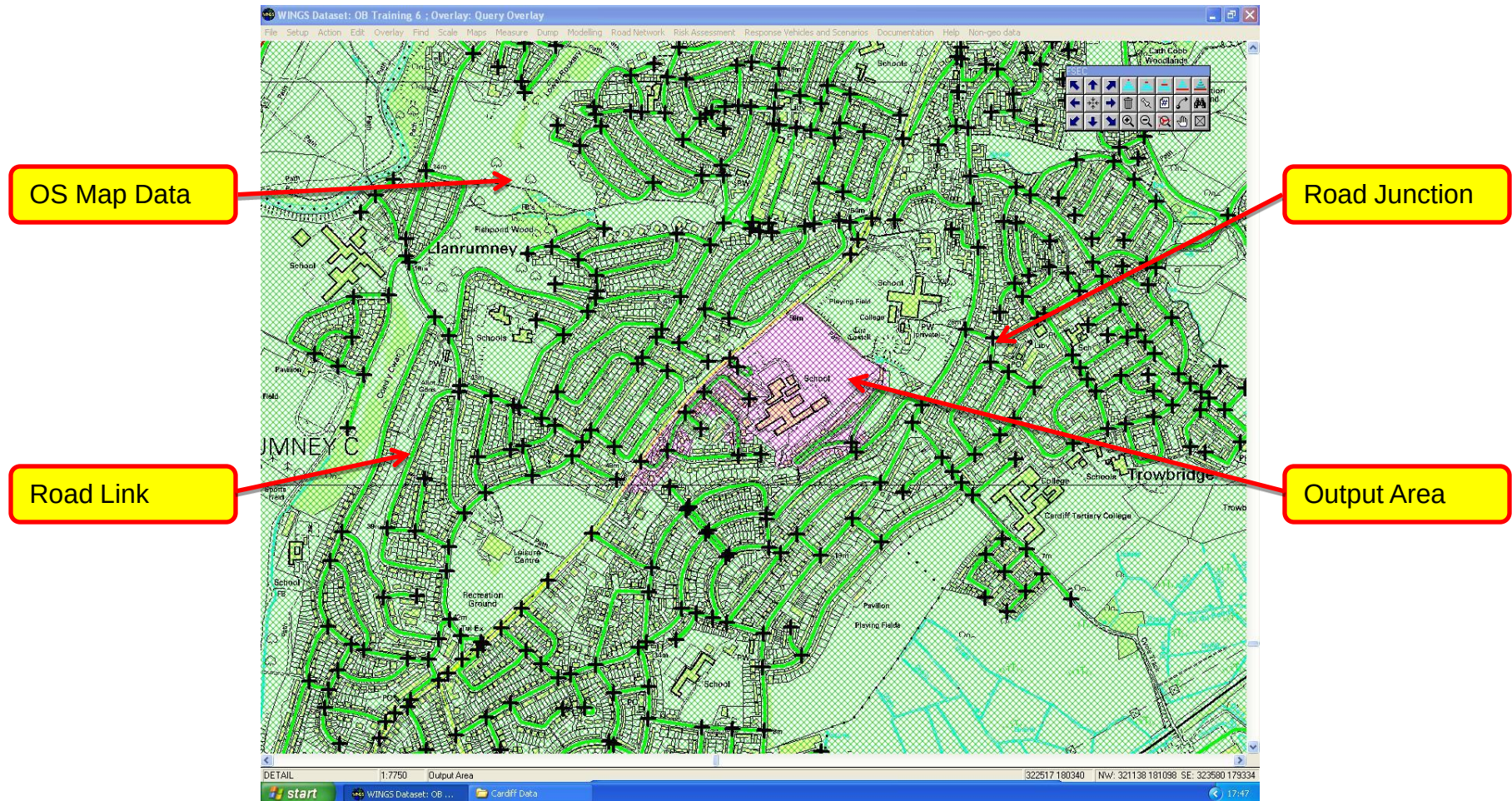
- Need a measure of the suitability of potential solutions within the Genetic Algorithm
- A “fitness function”
- Based on methodology from existing software (FSEC)
- Fire Service Emergency Toolkit (FSEC)
 - Based on a Geographical Information System (Wings32)
 - Run-times very long – 27 minutes for a typical brigade
 - Very graphics-intensive
 - Manual model configuration
 - Unsuitable for direct use as fitness function

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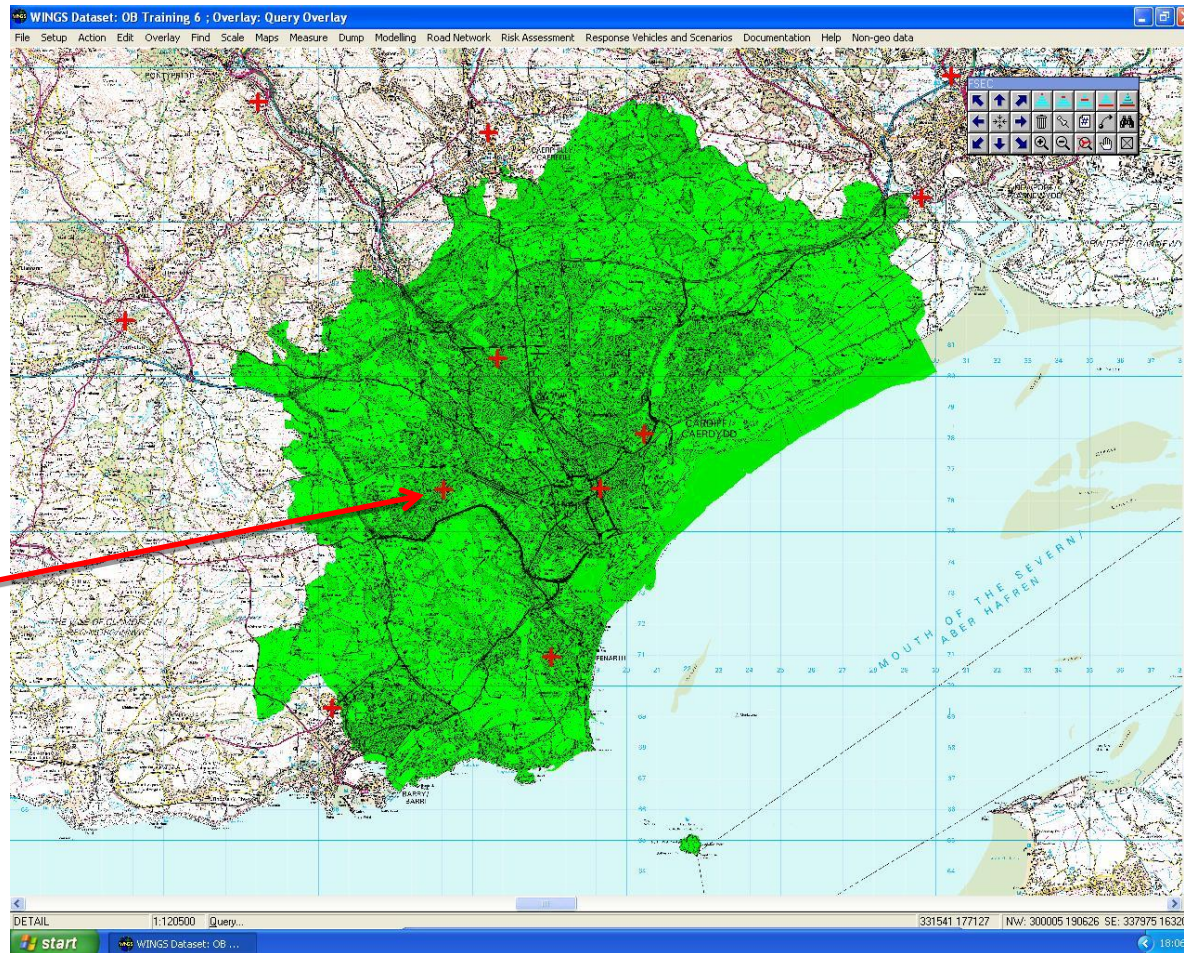
Software Development

- FSEC
 - Includes geographical relationships of brigade area
 - Road network
 - Census data
 - Incident data
 - Fire station locations and vehicle / staffing allocations
 - Calculates likely rates of four types of incident
 - Dwellings fires
 - Special Services Incidents (e.g. road traffic incidents)
 - Other buildings fires
 - Major Incidents (e.g. terrorist attack, major rail accidents)
 - Calculates fatalities and property damage based on mathematical relationship between response times and losses

Multi-Objective Decision Making for the Fire & Rescue Service Software Development

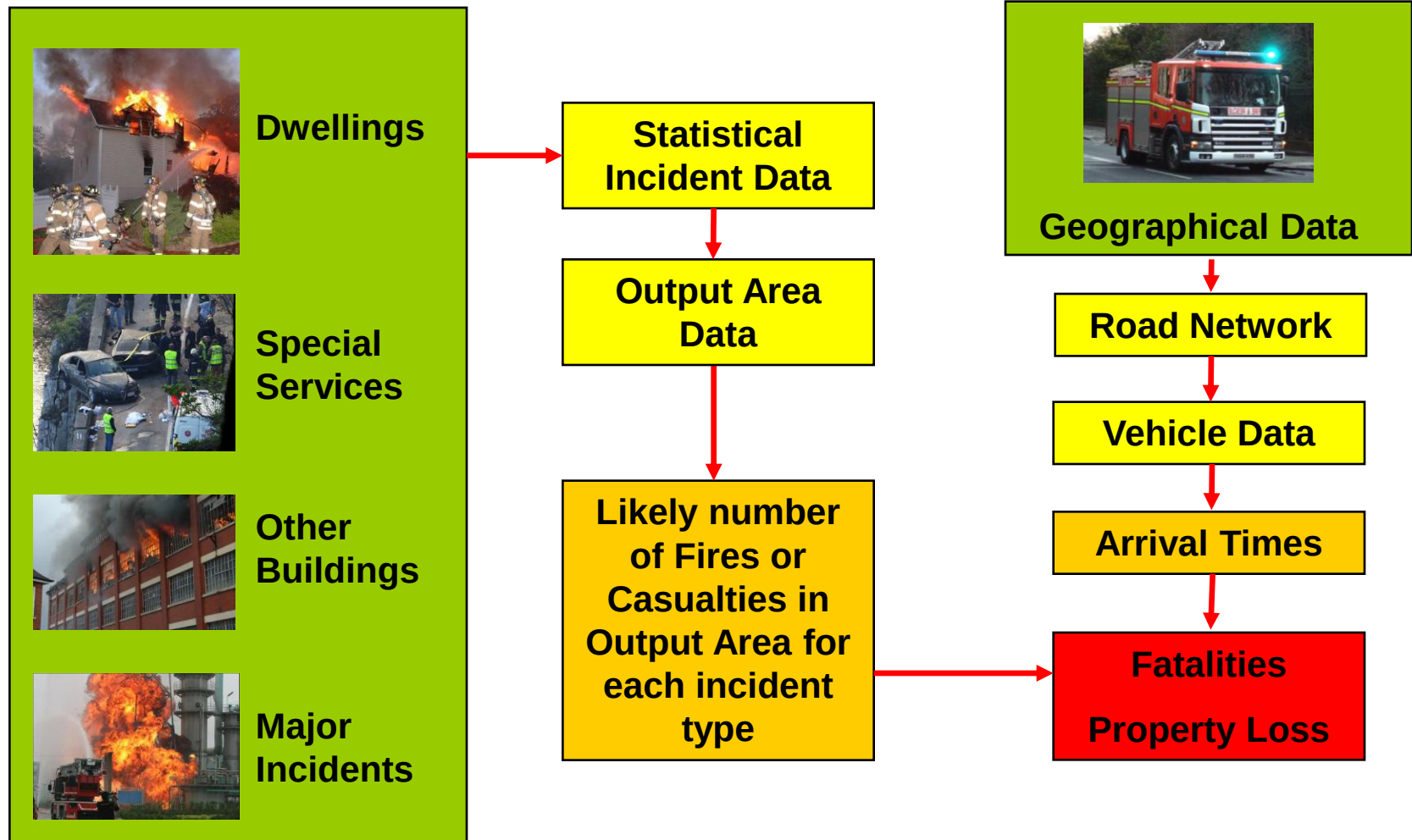


Multi-Objective Decision Making for the Fire & Rescue Service Software Development



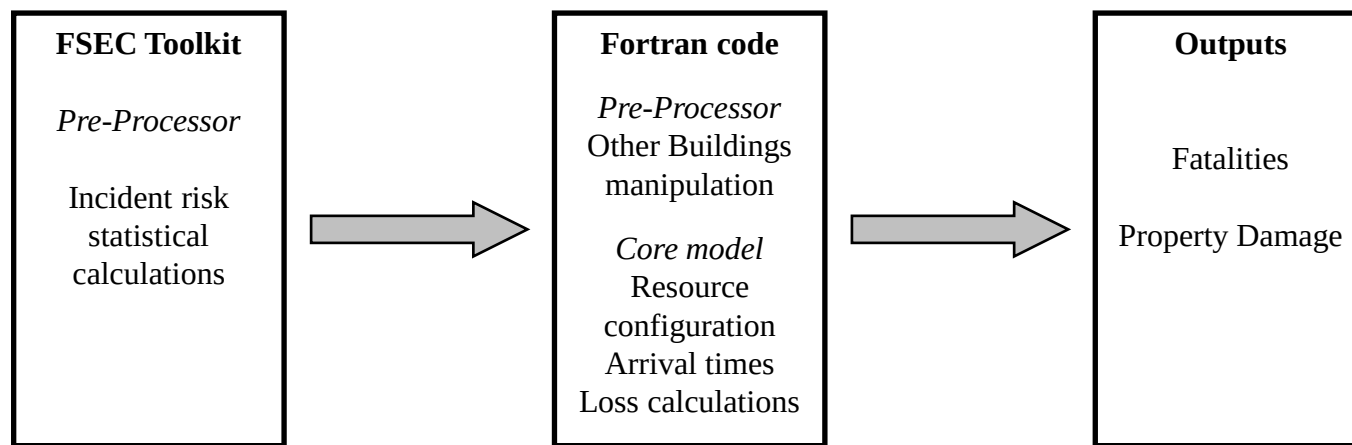
Fire Station
with vehicles

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Multi-Objective Decision Making for the Fire & Rescue Service Software Development

- Using FSEC as a fitness function
 - Fitness function is evaluated multiple times within each generation of the genetic algorithm
 - Need for significant reduction in execution times
 - Core FSEC calculations re-programmed in Fortran
 - Original FSEC used as pre-processor

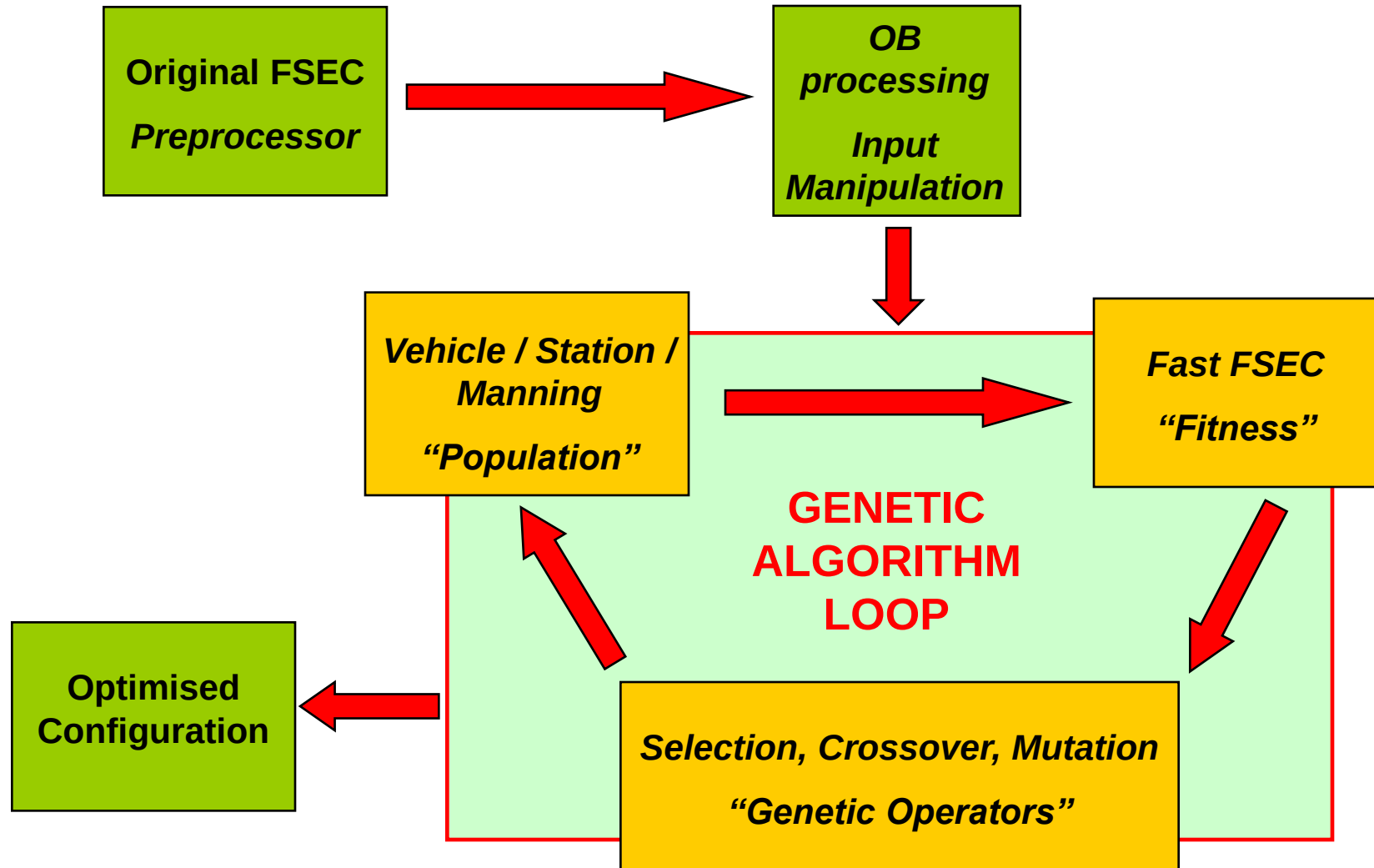


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Software Development

- Re-coded model achieves significant reduction in execution times:
 - Original full FSEC model for typical brigade 27 min
 - New Fortran FSEC code for same data 18 sec
- All configuration-specific calculations contained within Fortran code
- Pre-processor deals with statistical processing
- Time savings achieved for multiple runs – i.e. to evaluate 500 different resource configurations:
 - Original FSEC 227 hours
 - New Fortran FSEC 3 hours

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Future Work



- Continue work to link Genetic Algorithm to Fortran FSEC
- Test and experimentation.....

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Conclusions

- Optimising Fire Service resources is a highly complex problem
- There are a massive number of potential solutions
- It is impossible to manually evaluate all solutions
- Evolutionary algorithms offer many advantages in dealing with complex problems such as this
- A computationally more efficient version of FSEC has been developed for use as a fitness function
- A Genetic Algorithm has been written
- Work is ongoing to couple the two

Multi-Objective Decision Making for the Fire & Rescue Service Project Website

<http://fire.engineering.cf.ac.uk>



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Acknowledgements

