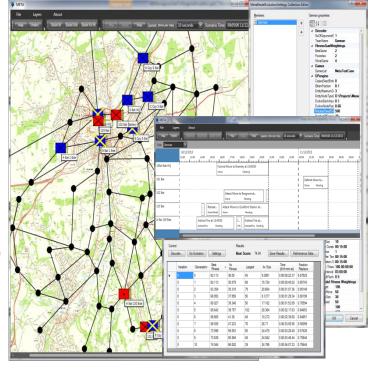


Overview

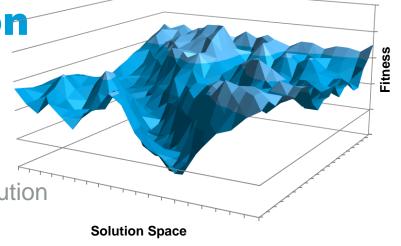
- Mission Planner: combat decision-making (AI) toolset
 - Supports Dstl high intensity warfighting simulations, to reduce/eliminate:
 - Complex pre-scripting
 - Human-in-the-loop
- Stochastic Optimisation Al
 - Genetic Programming, Simulated Annealing (novel approach)
 - Generic algorithms & architecture Plug and Play
 - Simple application to different problems
- Formulate the problem: Military-like syntax
 - Al algorithms efficiently generate plans for tactical problems
 - Resemble human-like decision making
- META model Al generates plans against a reduced problem set
 - Representing essential elements of the full problem.
 - Resulting solution evaluated against the full problem set
 - SimBrig assessing brigade level land engagements
 - Overcomes some of the limitations of AI techniques used





Stochastic Optimisation

- Family of techniques for solving any generalised problem
- Complex problem
 - Finds good but not guaranteed best solution
 - Explores whole solution space not just locally good solution
- Explore solution space in controlled way
 - Based on fitness measure
 - Definition of "good" can vary with user requirement
- Wide range of problems
 - Timetabling/Scheduling (travelling salesmen)
 - Game solutions (chess/soccer bots)

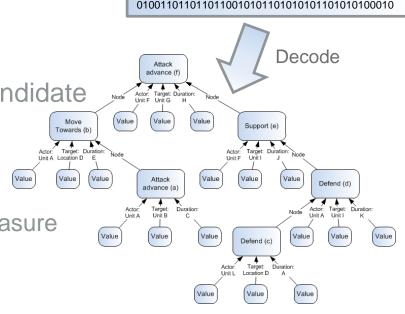




Genetic Algorithms

Entity – abstract representation of a candidate solution to a problem

- Typically bit stream
- Decoded to a solution
- Solution evaluated to obtain fitness measure
- Population of Entities
 - Initialise randomly
 - Evolve in generations, mutation, parent crossover & selection effects mimic survival of fittest
- End with a 'best' solution
- Genetic Programming
 - Ensures efficient decoding





Fitness measure

- Core of all Stochastic Optimisation algorithms
- Good measure of fitness
 - Allows algorithm to correctly apply selection pressure
 - Ensures fittest elements of population are evolved
- Each entity represents order set
 - Run through game & assess results
 - Losses
 - Achievements (positions held or denied from the enemy, or enemy losses or neutralisation)
 - Risk (enemy proximity, own units mutually supporting)
 - Efficiency (minimum resource consumption)
- SO algorithms notorious for finding loopholes

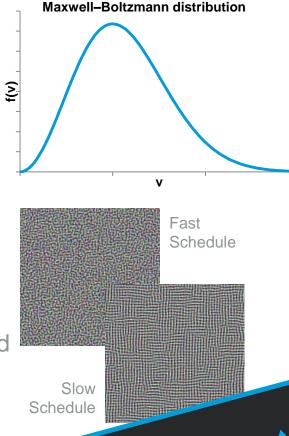


Simulated Annealing

- Well understood & efficient optimisation technique
- Candidate solution
 - Randomly perturbed for new solution
 - Probability new solution accepted:

$$e^{-\frac{F_C-F_N}{T}}$$

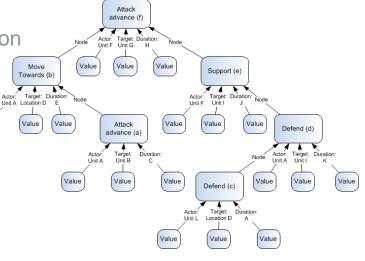
- T: "Temperature"
- F: Fitness measure Well understood/constrained
- Annealing Schedule
 - Initially large T explore solution space
 - As progress reduce T to "polish" good solution





SA: Solution Perturbation

- Problem in application to gaming problems
 - How to perturb candidate solution for new solution
 - Efficient algorithm should consider T schedule
 - Large perturbations for high T
 - Small perturbations for low T
 - Node/Input tree of GP solution easy to do this
 - Perturbation
 - Node: select node in tree replace with randomly generated node tree
 - Input: change input value(s)
 - High T
 - Favour Node perturbation
 - Favour Nodes with many descendants
 - Change multiple Inputs





Stochastic Optimisation: limitations

- Slow consider many solutions
- Solutions are problem specific:
 - Only optimises based on fitness criteria
 - Excellent at novel solutions tailored to detail of problem
 - Not doctrinally correct solution found in Staff Officers Handbook
 - Exploits loopholes
 - In fitness criteria
 - In evaluation model
 - Good test for model

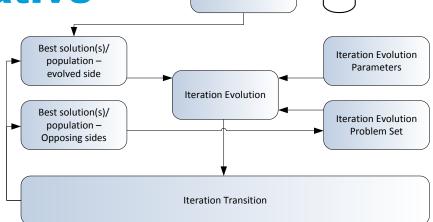




Mission Planner: Iterative

Seed Solutions Solution Library

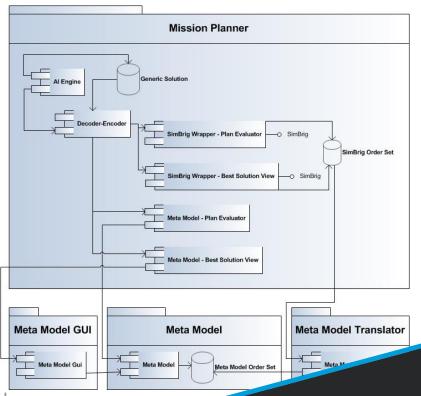
- Generate/evolve best solution
- Next iteration, use last iteration best solution &
 - Change side
 - Change problem scenario
 - Change Al algorithm control parameters
- Solution can be evolved against multiple scenarios
 - Must be good against each
- Solution can be seeded
 - Library of solutions





Mission Planner: Generic

- Al algorithms Generic Solution
 - Nodes and inputs have no meaning
 - No concept of problem applied to
 - Simply randomly changed
- Decoder
 - Only problem specific part
 - Translates generic node tree to order set
 - Runs evaluation model to get fitness score
- Architecture: plug and play decoders
 - Proof of Concept
 - SimBrig
 - META
 - Iterations can use different decoder
 - Generate end solution using different decoder





Military Syntax

- Military Synch Matrix
- Nodes correspond to:
 - Areas
 - Timelines
 - Orders
 - Seize, Hold etc.
 - Linked to Areas & Timelines
- OBJ BAILEY IN ORDER TO ALLOW FORCE PRELIM OP BRAVO TO COMPLETE DEFEAT OF en IN DTG: OBJ STEWART. CONCEPT OF OPS BRAVO OP Div Comd's intent is for fast attack to defeat en in Obj BAILEY, secure PL GIN and PLUM so set conditions for Force BRAVO attack before enemy res fmns can react deciseively On completion of prelim ops by Cx to secure PL SODA, Dx OBJ and Ex attack to seize obi GRANT **STEWART** BAILEY in close coordination with Div deep ops to isolate and attack en in Objs GRANT BAILEY and STEWART and protect open flanks. BRAVO PL SODA TIME (Estimated) -10 H Hour +12 Deep Ops Fight in Defend Fight in Defend СВ **ENEMY ACTION** Recce Sy Zone Fire Sy Zone Main Posn Fire OWN DECISION PTD Os to move to Launch Avn O deep Op to D and E Bdes Sp BRAVO Attack Air and Arty Air and Arty Air Attack Air and Arty Attack in sp Overwatch DEEP OPS Attack GRANT Attack BAILÉY RAG Attack STEWART of BRAVO Assy Area Div Secure C BDE OPS PLUM Obi GRANT Res BRAVO LD Assv Area Move on Cross Secure D BDE OPS APPLE Obi BAILEY Routes 1 + 2LD V Assy Area Move on Cross Secure E BDE OPS APPLE Routes 3 + 4 Obi BAILEY Res Assv Area Risk as Bde Take on sv Ε **REAR OPS** Preps Attack of GRANT Prep Fwd Attack RAG Attack Attack Posn + Ammo Obi GRANT C Bde Attack fire SEAD AIR DEFENCE Protect Protect Cover PL Mov with Cover PL Protect BRAVO Sp BRAVO Attack on Assy Area C Bde Mov Bde Movs SODA Lead Bde ENGINEER SP Clear BRAVO Route Sv Sp to D and E Bdes BRAVO MSR Passage of Lines to Corp Tps Routes 1 - 4 Refurb Refuel Refurb Resup Bdes D + E Bdes C. Bde on Obi BAILEY BRAVO Tps COMD FWD HQ Main to FWD MAIN HQ FWD to with C Bde Moves GRANT with E Bde BAILEY BRAVO MAIN Moves REMARKS Sp to BRAVO op
- Units naturally co-operate in time and space

ON Os, DIV ATTACKS TO SECURE

Efficiently generate "human-like" orders

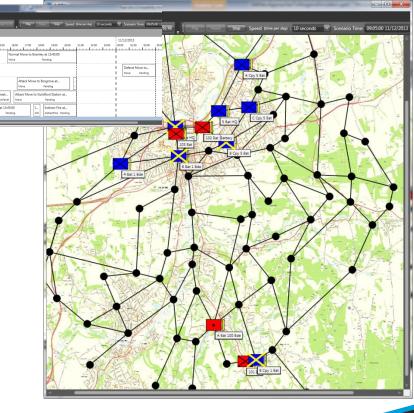


NORTH

ORIGINATOR:

META

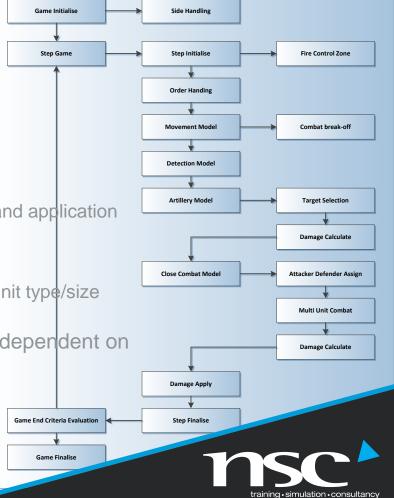
- Model for EngagemenT Analysis
- Al algorithms requirements for plan evaluation model
 - Fast
 - Robust
 - No logic loopholes
 - Evaluate nonsense order sets
 - Al considers bad solutions
 - Need good measure of fitness
- META Simple
 - Representing ONLY the essential elements of the full problem
 - Simple and Robust algorithms
 - Quick and cheap (3 months, 2 man team)

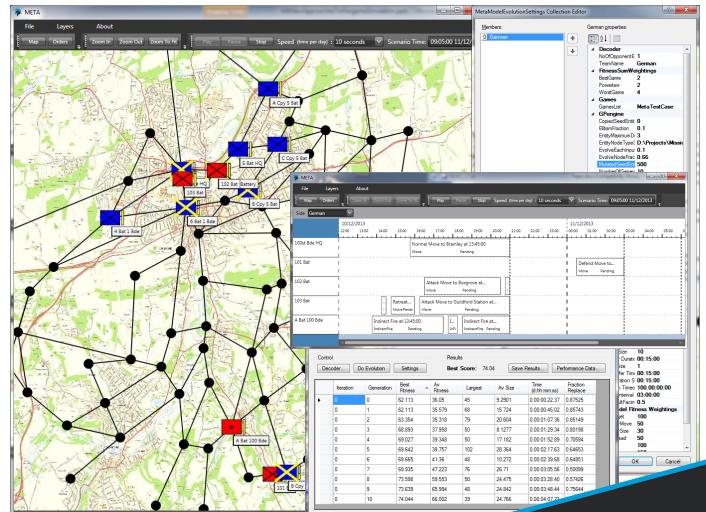




META

- Aggregated land model
 - Brigade level (SimBrig)
 - Algorithms at an aggregated level
 - Quickest execution speed
 - Expose the fewest loop holes in algorithm logic and application
 - Combat:
 - Lanchester-like
 - BAMS-like 2-D cross matrix of attrition rates by unit type/size
 - Handles units of differing sizes/capabilities
 - Movement, combat, detection, artillery models dependent on
 - Terrain, unit types, postures, suppression
 - Arc, node movement network (SimBrig)
 - Zones of control
 - Internode visibility



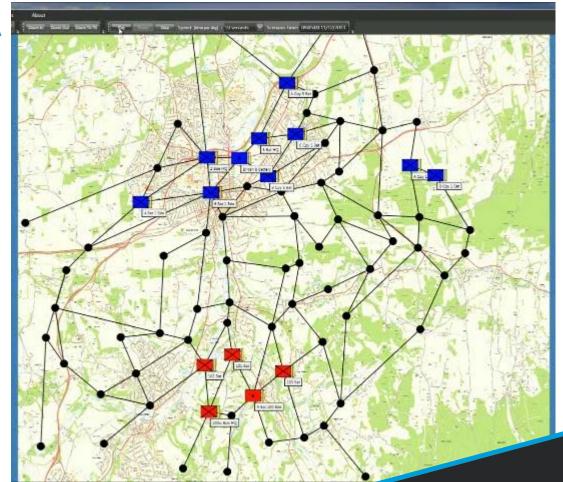


Results

- MP Demo
- META Demo



META Demo





Conclusions

- "Simplicity is the ultimate sophistication"
 - Leonardo da Vinci
- Al Algorithms
 - Generic architecture Plug and Play
 - Simple application to different problems
- Formulate the solution: Military-like syntax
 - Al algorithms efficiently generate plans for tactical problems
 - resemble human-like decision making
- Meta model Al generates plans against a reduced problem set
 - Representing essential elements of the full problem.
 - Simple, Fast, Robust
 - Overcomes limitations of AI techniques



Dstl Conclusions

- Mission planner will allow
 - Greater range of potential solution space to be examined
 - More reactive Red allowing more robust testing of plans
 - Rapid generation of variations, examining changes in
 - Force Structure
 - Constraints
 - Improved testing of complex models

