



Engineering, Operations & Technology  
Boeing Research & Technology

# Genetic Algorithm Applied to Multi-Agent War Gaming Simulation

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# Agenda

## **IR&D Project and Goal**

**Brief description of COTS/GOTS analysis software employed**

**Modeling of an urban rescue scenario**

**How “Darwin” Genetic Algorithm was used**

**Pareto-optimal solutions obtained + description**

**Summary and thoughts**

# IR&D Project and Goals

## **Develop a New Capability for Operations Research Design and Analysis**

### **Apply a Genetic Algorithm to a Multi-Agent Theatre-Level Simulation**

- Design optimal traits of networked interoperating agents
- Evolve agent/platform traits (capabilities) that must achieve a desired effect (EBO is implicit)
- Co-evolution of adversary = predict adversarial response

### **Provide Capability-Based Evaluation of System-of-Systems**

- Multiple interacting networked platforms/agents form a SoS
- No modeling of specific technologies, only the *capabilities* they bring
- Satisfaction not tied to a specific technology solution (but could be)

# Tools Applied

## SEAS

- GOTS Multi-agent theatre-level simulation
- Programmable rule-based behavior (actions & re-actions)
- <http://teamseas.com/>



## ModelCenter

- COTS framework for MDAO
- Provides “Darwin” genetic algorithm (among others)
- <http://phoenix-int.com/>



## MatLab

- Provided easy interface between ModelCenter and SEAS
- Provided statistical collation of MC trials and post-analysis
- <http://www.mathworks.com/>



## Excel

- Host application for a simple/notional cost model
- Hook for more complex cost models



# What is a Multi-Agent Simulation?

**Variety of entities that operate autonomously and independently in a simulated world**

**Continuously get input, interpret it, and make decisions**

**Communicate and negotiate with other agents to achieve a goal or get things done**

**Agents can be programmed to learn from experience**

**Outcome is not certain or entirely predictable**



**SEAS contains all of these properties**



# “Darwin” Genetic Algorithm

**Process analogous to natural selection**

**Most fit designs from a population will produce offspring**

- Crossover (recombination)
- Mutation

**Small set of less fit designs maintain diversity**

- prevent pre-mature convergence on poor solutions

**Continues until improvement stops**

**Effective in noisy or discontinuous design spaces (uncertain outcomes)**

**Better chance of finding global optimum and multiple near-optimal designs**



# Capabilities and Objectives

## Blue Agent Capabilities (Darwin Design Traits)

- Vehicle Speed
- Vehicle Armor Hardness (probability of being killed if fired upon)
- Sensor Range
- Weapon Range
- Weapon Lethality (probability of kill when firing upon red)



## Multi-Objective Optimization Goals (Effectiveness)

- Probability of Successful Rescue (maximize)
- Monetary Cost (minimize)
- Bloodshed / Lost Lives (minimize)

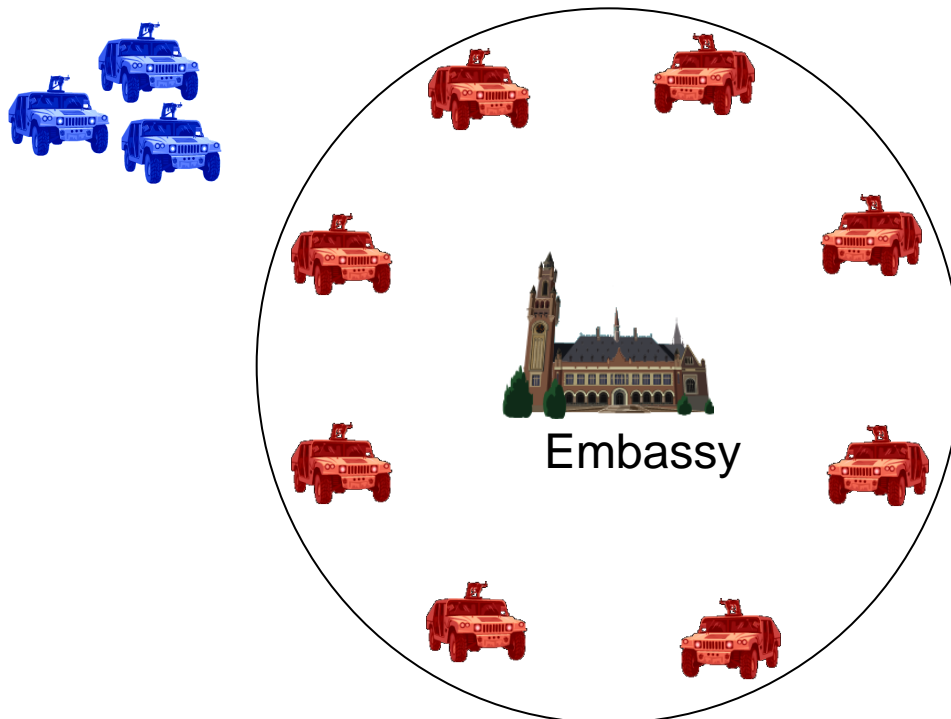


# SEAS “Embassy Rescue” Scenario

Officials are trapped in an embassy following a coup

Eight red agents patrol the central area of town

Three blue agents try to covertly reach the embassy to rescue trapped personnel and covertly get them out



- **Red and Blue agents are outfitted with different capabilities**
  - Vehicle, Sensors, Weapon, Communications
- **Blue agents attempt to avoid being spotted by red agents**
- **Each side has rules for engaging the enemy should they encounter each other**



# Process Overview & ModelCenter

Phoenix Integration ModelCenter 7.1 - [C:\Documents and Settings\mark\My Documents\SEAS\TestModel3.pxc] - [Model (Analysis)]

File Edit View Tools Component Project Window Help

Model: Model: Model

Name	Value
<b>Model</b>	
PerformanceComponent	
EyeRange	1.399
GunRange	1.324
Pk	0.9918
Speed	13.2
ProbRescue	0.99667
FractRedKilled	0.7675
Hardness	0.01
CostModel	
EyeRange	1.399
GunRange	1.324
Pk	0.9918
Speed	13.2
Cost	4.29115
Hardness	0.01
Darwin	

Capability Traits

Effectiveness

4) The best of the population is used to generate modified population

1) Population of solutions to test is generated

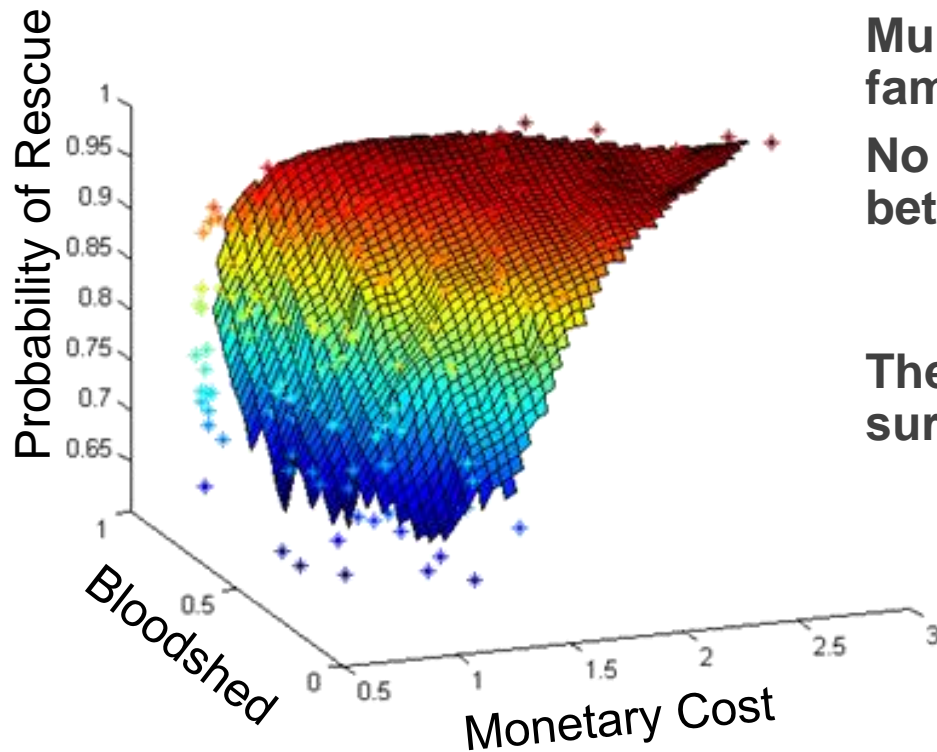
3) Each solution also produces a monetary cost

2) Each solution is tested in over 100 war games to obtain an average MOE score(s)  
"Probability of winning the war"

SEAS

Model Project

# Pareto-Optimal Set and The Trade Space

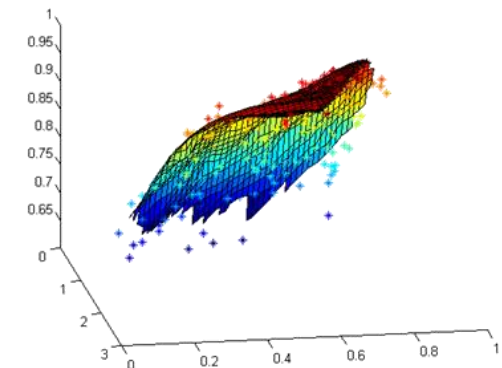
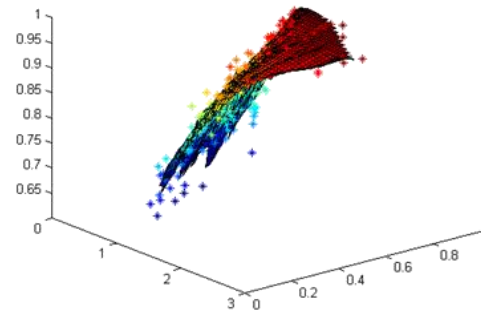
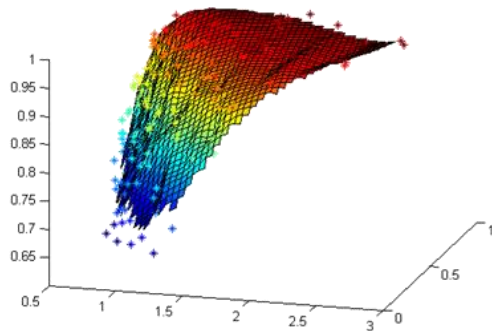


**Multi-Objective optimization leads to a family of optimal design solutions**

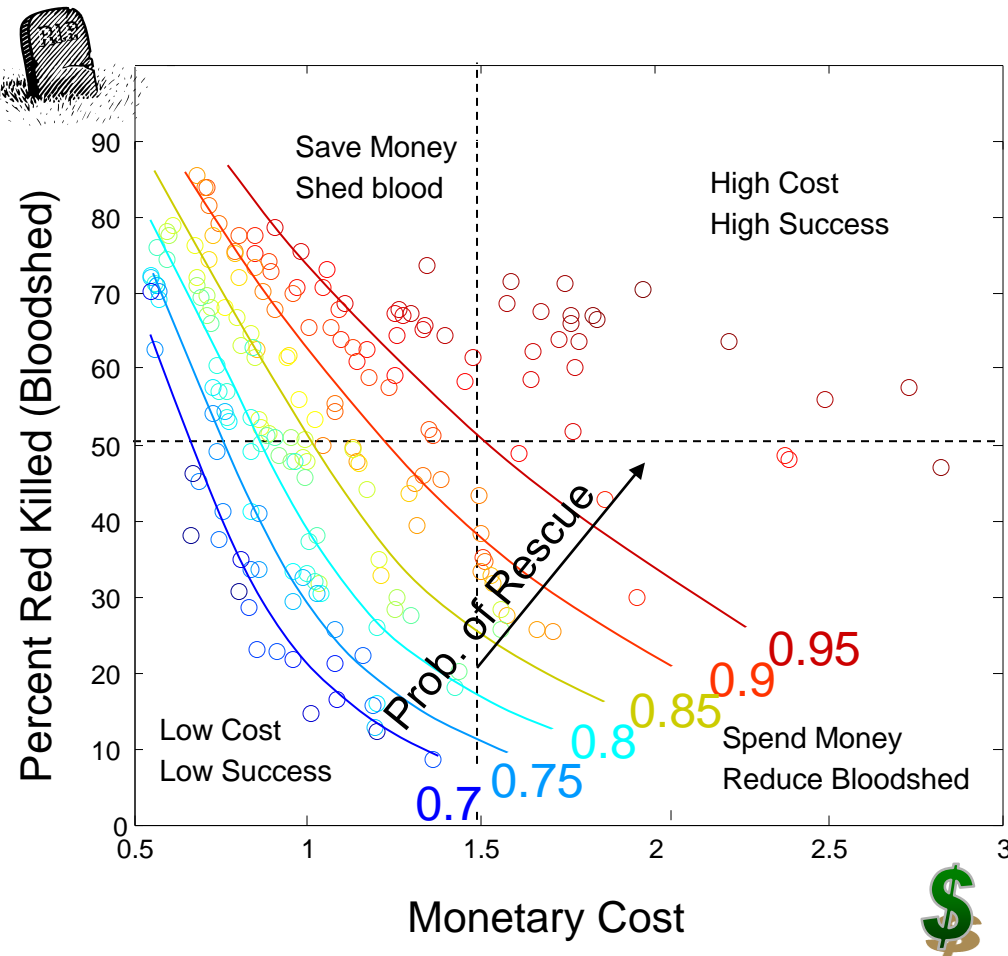
**No other designs were found that were better in *all* objectives**

- To continue to improve one objective, you must give up (trade) on another

**These solutions tend to lay along a surface representing the “trade space”**



# Identified "Solution Quadrants"

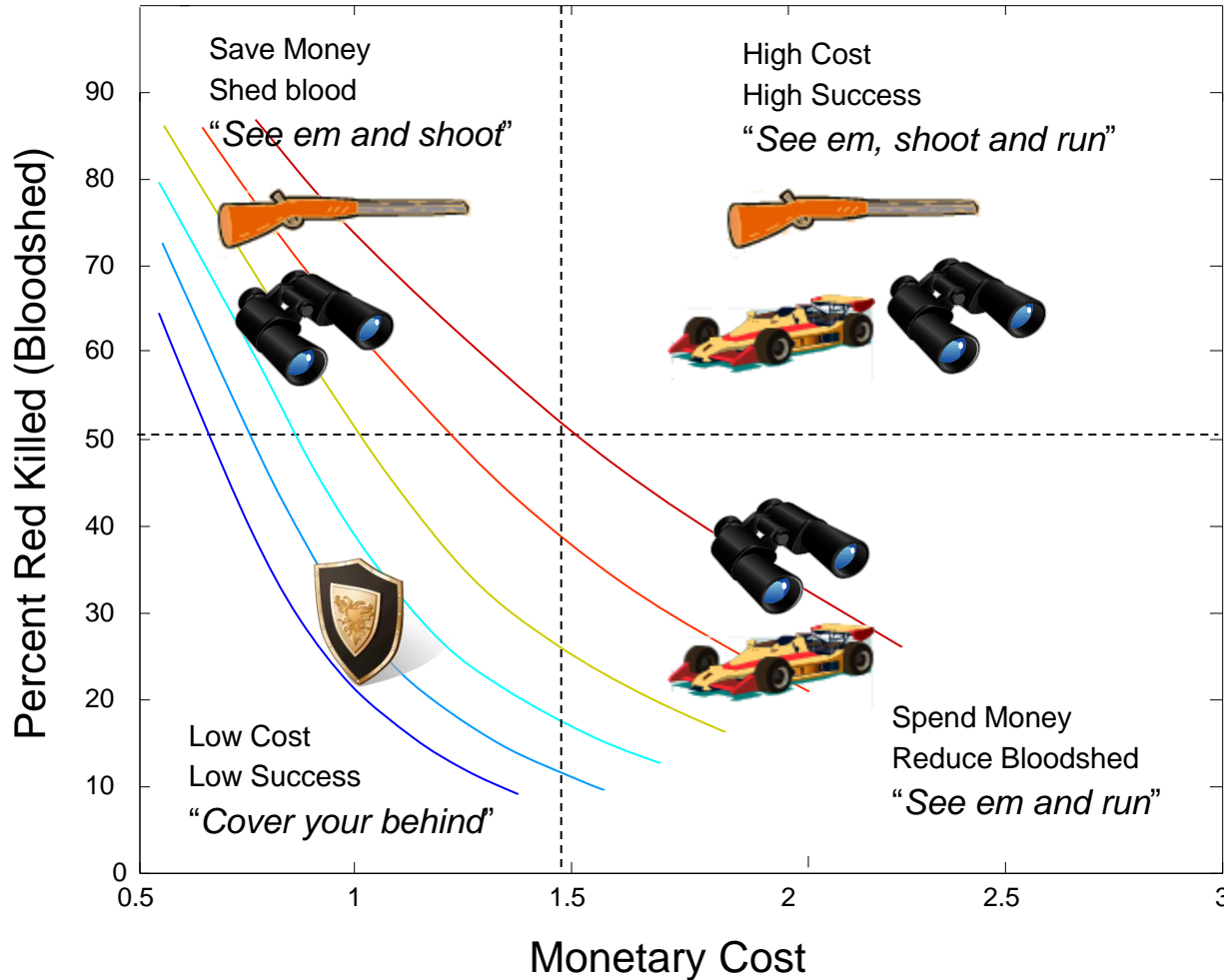


## Optimal design set identifies where one objective must be traded against another Solution "Quadrants" Emerge

- Highest probability of rescue comes at the cost of both money and lives
- Lowest probability of success comes at minimal cost to lives and money
- If you want to save money and have high probability of success, you must shed blood
- If you can afford to spend lots of money, you can achieve high probability of success with minimal bloodshed

**Design traits (capabilities) can be examined in each of these quadrants (which is actually a continuum)**

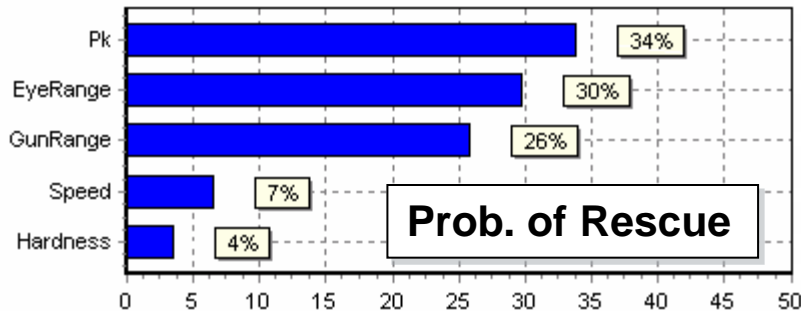
# Dominate Traits in Solution Quadrants



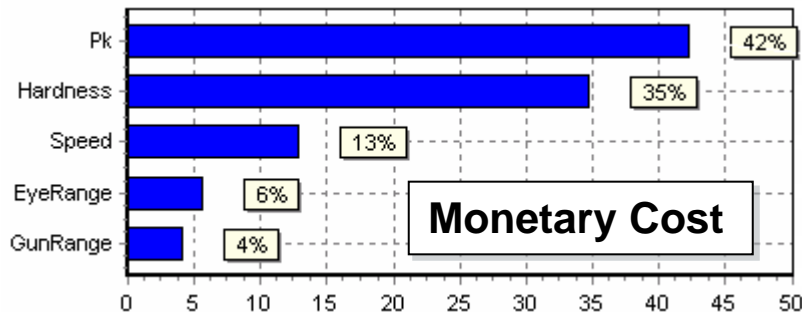
By examining dominate traits in each quadrant, one can infer the strategy evoked

# Sensitivity of Objectives to Capabilities

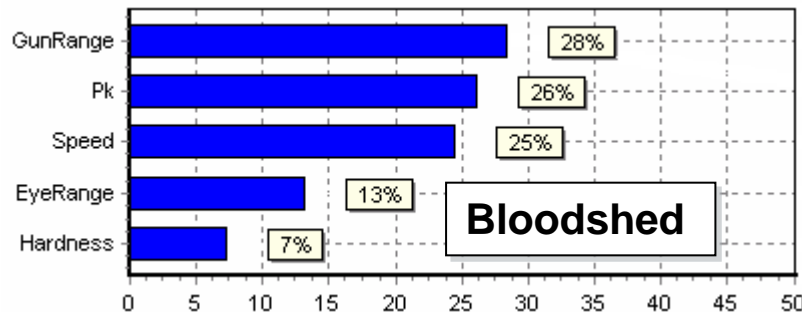
## “Driving Capabilities”



Rescue success is most sensitive to weapon lethality, sensor range, and weapon range. Speed and armor contribute less to rescue success.



Monetary cost is driven most by weapon lethality and armor. Speed, sensor range, and weapon range play a lesser roll.



Bloodshed is most sensitive to weapon range, weapon lethality and speed. Sensor range and armor play a lesser role.

\*Pareto Front (efficient) Solutions Only

# Summary and Thoughts

## **A GA has been successfully applied to a multi-agent war game simulation**

- Proof of concept: evolution of agents in a MAS is achievable

## **“Adaptation” happens over generations, not during war game scenario events**

- Basis of a “complex adaptive system”
- Behavior and strategies can also evolve over generations

## **Solutions become specified more as required capabilities, not specific technology solutions**

- Although the technological source of different capabilities can be modeled if a specific technology solution is sought.

## **Co-evolution of adversarial traits (capabilities) is possible, but not yet done.**

- Predict the likely evolutionary responses of an adversary



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**Questions?**

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### Biography:

M.S. in Physics from C.S.U. Long Beach, B.S. in Applied Physics from U.C. Irvine. Twenty eight (28) years experience in MS&A for ISR, remote sensing, and military operations. General skills include developing mathematical models and algorithms, developing software and simulations, planning and running simulation experiments for design and requirements analysis, and architectural and systems engineering design efforts. Specific areas include remote sensing for space and airborne EO/IR platforms, EO/IR phenomenology (natural backgrounds and targets), advanced image processing (including spectral and temporal domain processing), sensor performance, space mission analysis and orbital design. High interest in applying MDAO to multi-agent simulations.