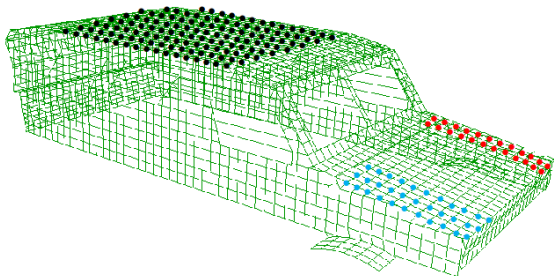


# A Comparison of Antenna Placement Algorithms

Abhinav Jauhri, Jason D. Lohn, Derek S. Linden

# Antenna Placement Example

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- ▶  $A_1$  has 24 possible antenna placements
- ▶  $A_2$  has 33 possible antenna placements
- ▶  $A_3$  has 136 possible antenna placements

Size of search space =  $m^n$

Goal: Find a placement for each of three antennas.

# The Problem

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Given:

- platform  $P$  with its surface gridded so that grid points represent possible antenna placement
- set of  $m(m > 1)$  antennas  $A = A_1, A_2, \dots, A_m$
- for each  $A_i$ , a set of  $n$  possible placement locations ( $n > 1$ );  
 $A_i = \{(x_{i1}, y_{i1}, z_{i1}), (x_{i2}, y_{i2}, z_{i2}) \dots (x_{in}, y_{in}, z_{in})\}$

Find: A set of  $m$  optimal antenna locations on  $P$

# Antenna Placement Objectives

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## Antenna Placement Issues

- Coupling among antennas
- Parasitic effects and reflections from the host platform
- Loss of efficiency
- Difficulty conforming to aerodynamic, thermal, other environment factors

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## Desired Antenna Placement Objectives

- Gain in radiation pattern
- Minimize coupling

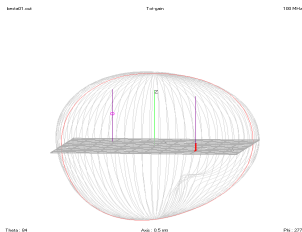
# Fitness Functions - I

To minimize difference in gain:

$$F_{RP}(A_i) = \sum_{\theta} \sum_{\phi} \|ISG_i(\theta, \phi) - FSG_i(\theta, \phi)\|^2, \quad (1)$$

where

- ▶  $\theta, \phi$  spherical and cylindrical coordinates
- ▶  $ISG(\cdot)$  returns in-situ gain pattern
- ▶  $FSG(\cdot)$  returns free-space gain pattern



## Fitness Functions - II

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To minimize coupling:

$$F_{MC} = \sum_{i=1}^{m-1} \sum_{j=i+1}^m CP(A_i, A_j), \quad (2)$$

where

- $CP(\cdot)$  computes the coupling between two antennas

# Overall Fitness Function

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For an individual/hypothesis, fitness is defined as:

$$F = \alpha F_{MC} + \beta \sum_i F_{RP}(A_i), \quad (3)$$

where  $\alpha + \beta = 1$

# Experimental Setup

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- ▶ Create individual(s) such that each individual is defined by a placement for each of the  $m$  antennas
- ▶ Run all individuals through *NEC* simulator \* to get fitness parameters
- ▶ Apply EA operators
- ▶ Repeat. . .

Algorithms explored: Simple GA, Simulated Annealing, Evolutionary Strategy, and Hill Climber †

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\*<http://www.nec2.org>

†<https://github.com/ajauhri/evol-ant-placement>



# Experiments: Test Cases

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ID	Antennas	Total allowable placements
tc1	2	7,056 (83x83)
tc2	3	50,625 (45x45x25)
tc3	3	126,025 (71x71x25)
tc4	4	20,736 (12x12x12x12)

\*Allowable placements for each antenna are provided within parenthesis

# Results: Mean Evaluations

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Mean number of evaluations to reach the best solution (over 10 runs):

test case \ method	GA	ES	SA	HC
tc1(7056) <sup>‡</sup>	2350	1728	667	164
tc2(50,625)	31 680	11 165	1653	174
tc3(126,025)	45 900	26 880	4809	227
tc4(20,736)	6150	4466	423	90

- ▶ Simulated Annealing was the fastest
- ▶ Evolutionary Strategy always found optimal, but was relatively slow

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<sup>‡</sup>Total number of possible evaluations within parenthesis

# Results - Mean Best Fitness

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- Lower fitness is better
- Mean taken over 10 runs
- In all test cases, antennas were subjected to same frequency

method \ test case	Exhaustive( $H^*$ )	GA	ES	SA	HC
test case 1	0.4968	0.4993	0.4968	0.4994	0.5015
test case 2	0.4968	0.4979	0.4968	0.5042	0.5138
test case 3	0.4974	0.4976	0.4974	0.4974	0.4987 <sup>§</sup>
test case 4	0.4992	0.4992	0.4992	0.4992	0.4992

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<sup>§</sup>Best performing individuals lie within a small fitness range

# Equivalence of fitness to efficiency

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For a particular test case, fitness change of 0.01 is equivalent to either the corresponding value under expected gain ( $\mathbb{E}_g$ ) column, or difference in coupling ( $\Delta_c$ ).

ID	$\mathbb{E}_g$	$\Delta_c$ (dB)
tc1	872.277	0.5474
tc2	862.082	1.3034
tc3	861.845	1.5180
tc4	871.049	0.5693

$$\mathbb{E}_g = \frac{1}{N \cdot m} \sum_i^m F_{RP}(A_i), \text{ where } N = |\theta| \cdot |\phi|$$

Thanks!