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# Diversified Virtual Camera Composition

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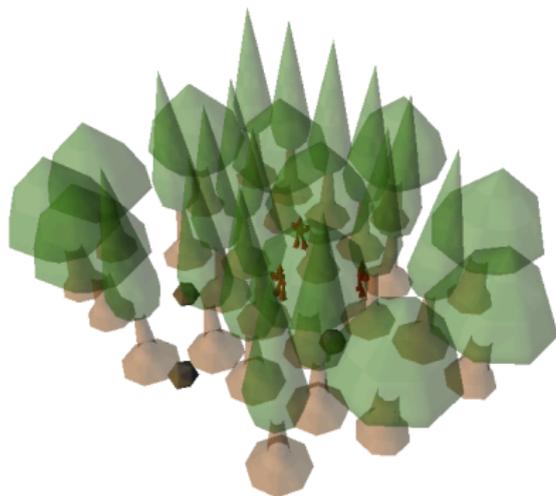
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## VCC: What is it about?



growing need to automatically determine good camera position

- in games and related 3D applications
- influenced by cinematographic techniques, e.g. tracking shots
- in response to growing content complexity (e.g. PCG!)

# Desired visual properties

- visibility
- projection size
- shot angle

5 dimensions:

- 3 spatial coordinates
- 2 angles

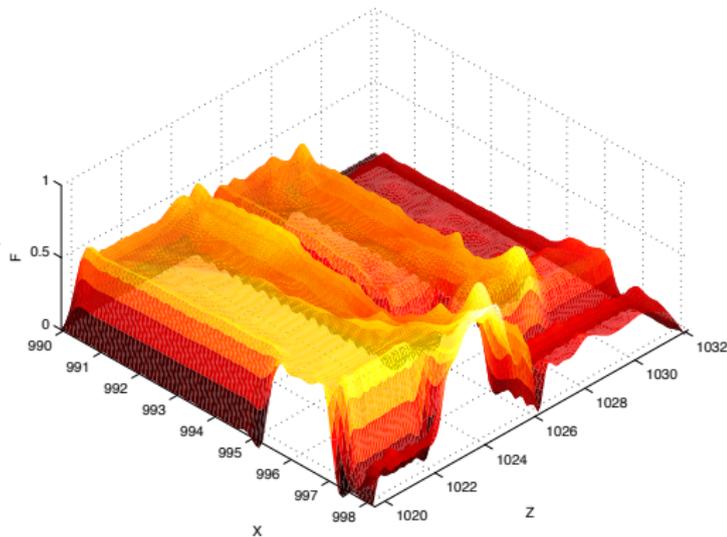
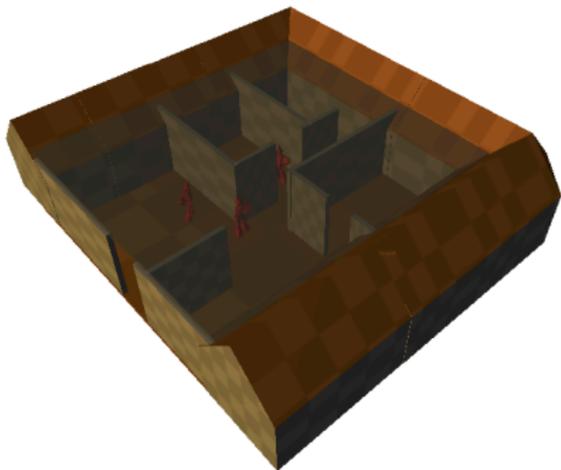


# Objectives

main objectives of this work:

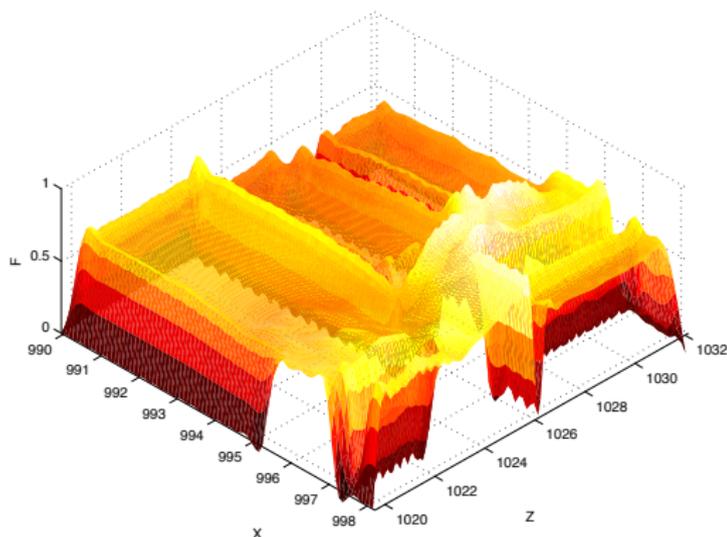
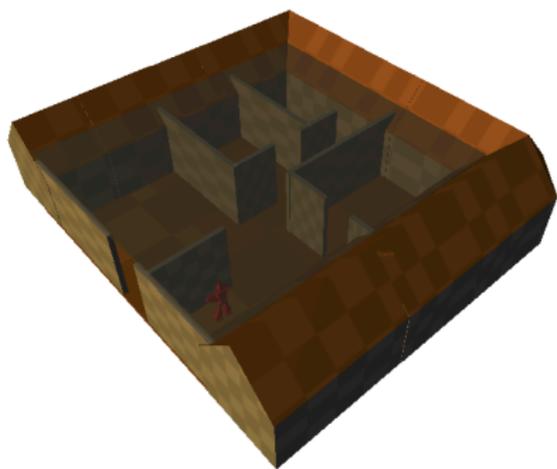
- explore diversity vs quality tradeoffs: how difficult is it to obtain good but different shots? runtimes?
- compare with existing solutions (optimization algorithms): can we do quicker and/or more reliable?
- obtain problem knowledge (landscape structure)

# Eavesdropping



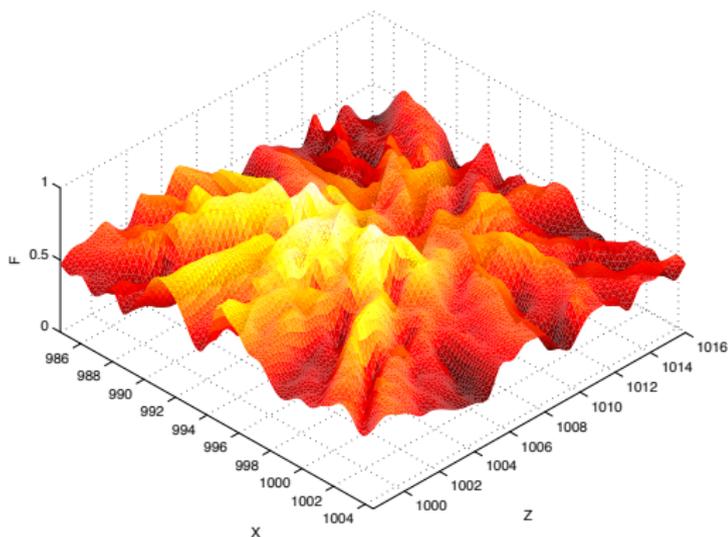
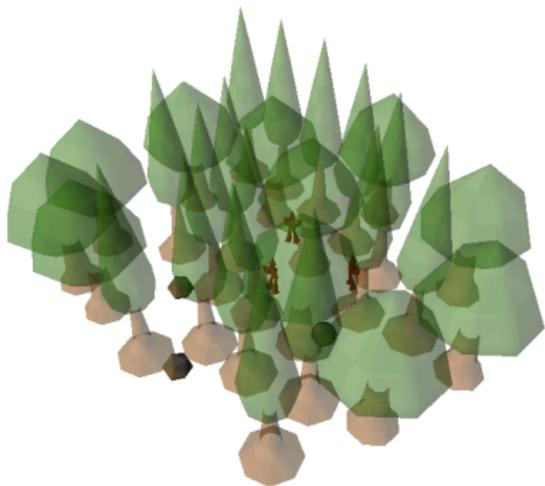
- birdseye view, (scanned) maximum over  $x$  and  $z$  axis
- includes three characters, two of them facing each other (chatting), one eavesdropping
- demanded properties: full visibility of all characters, projection size  $1/3$  of the screen for all characters

# Ambush



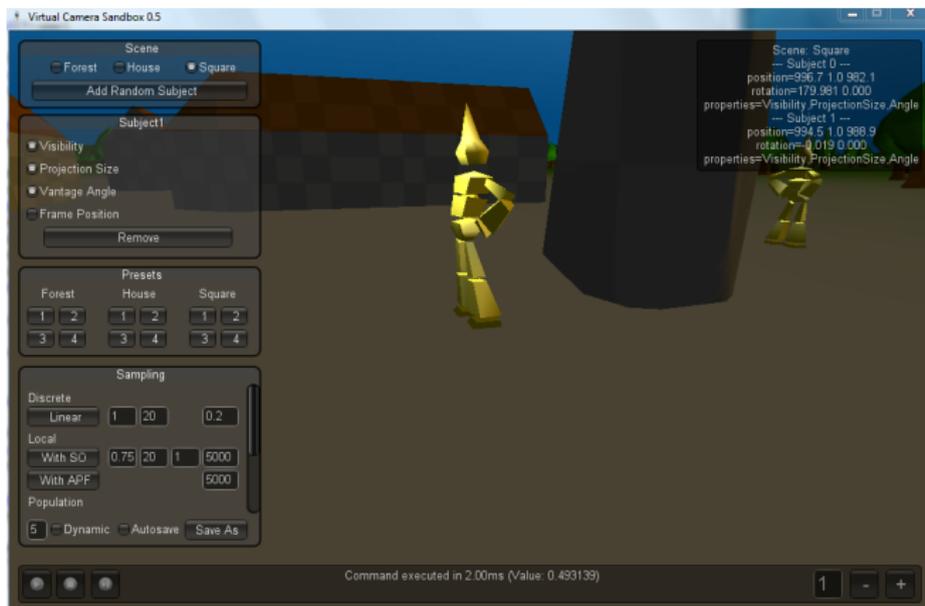
- two characters on two sides of a wall
- demanded properties: full visibility of all characters, projection size  $1/2$  for all characters, horizontal angle of 90 degrees to the right of each character

# Chat



- based on chat scene by Thawonmas et al.
- three characters with one ideally chatting to the other two
- demanded properties: visibility and projection size equal to problem 1, camera on the back of listening characters

# Evaluation



- sandbox by Paolo, implemented for the Unity engine
- communication via TCP/IP, no realtime (realtime: 16ms)

# Algorithms

three 'default' solutions (have been used in the context before):

- Particle Swarm Optimisation (PSO)
- Differential Evolution (DE)
- Sliding Octree (SO)

two 'new' solutions (not used yet):

- CMA-ES without any population heuristic (short runs!)
- related niching method (on base of CMA-ES): NEA2

and NEA1 (old version) as reference for NEA2

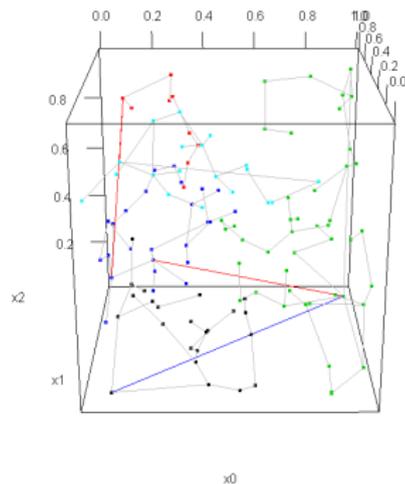
# Restart/niching based methods

we apply a single parameter change (expert knowledge):

- TolFun stopping criterion parameter for restarts set to  $10^{-3}$
- this triggers restarts earlier (short runs!)
- does not affect other 3 methods (no restarts)

differences CMA-ES / NEA2:

- NEA2 does initial LHS sample and detects clusters
- CMA restarts located randomly, NEA2 works down clusters

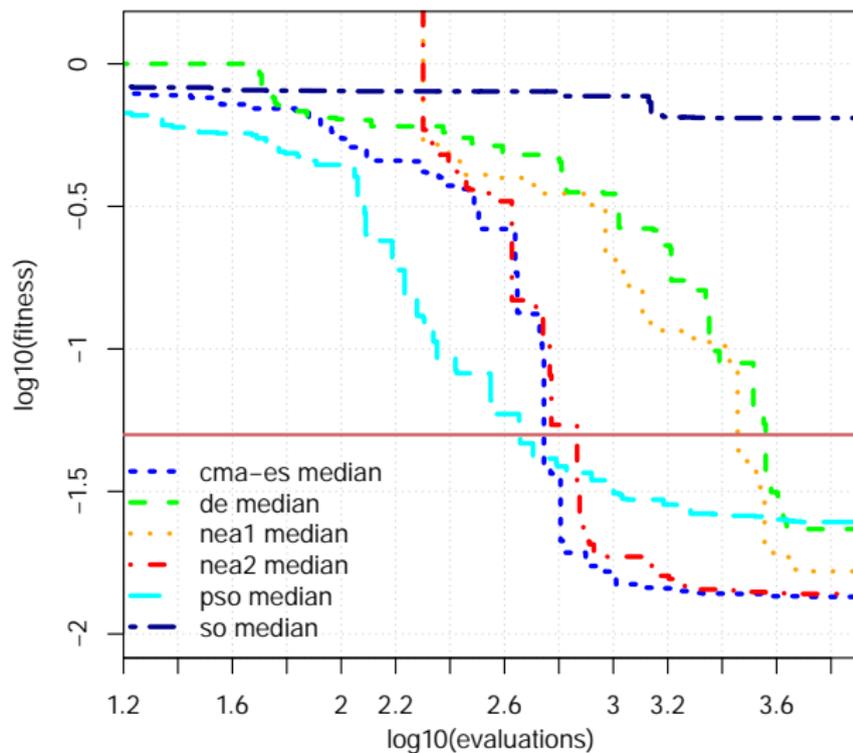


## Experimental setup

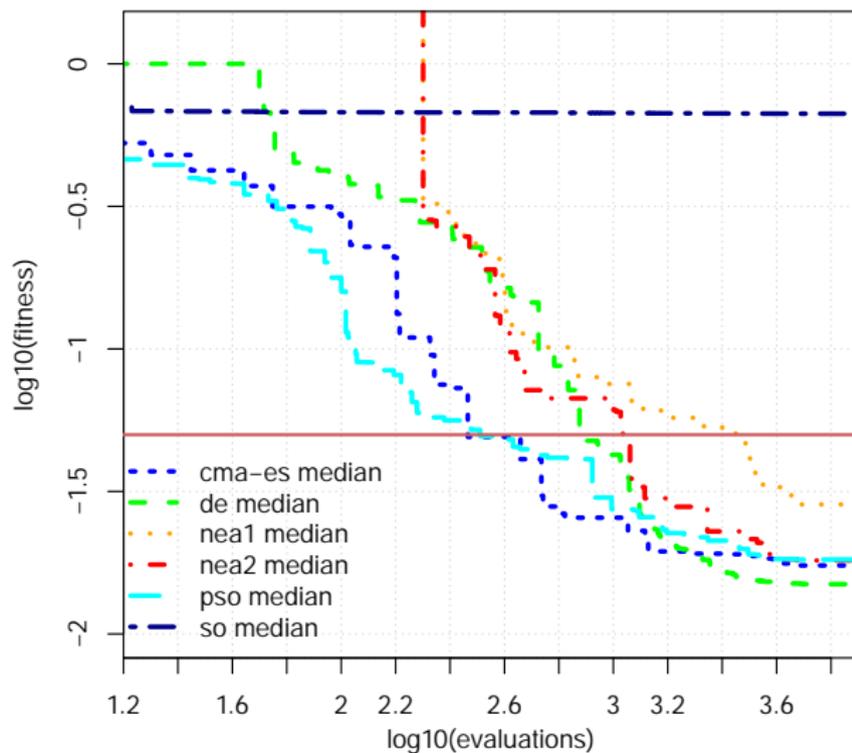
- 5000 evaluations per run
- 20 runs per algorithm on each problem
- all default parameters (except TolFun)
- we measure the ERT until less than 5% error to optimal value
- times to 2nd and 3rd diverse solution  
(min Euclidean distance 1 in 3 spatial coordinates)

$$ERT = \frac{\#fevals > f_{target}}{\#succ}$$

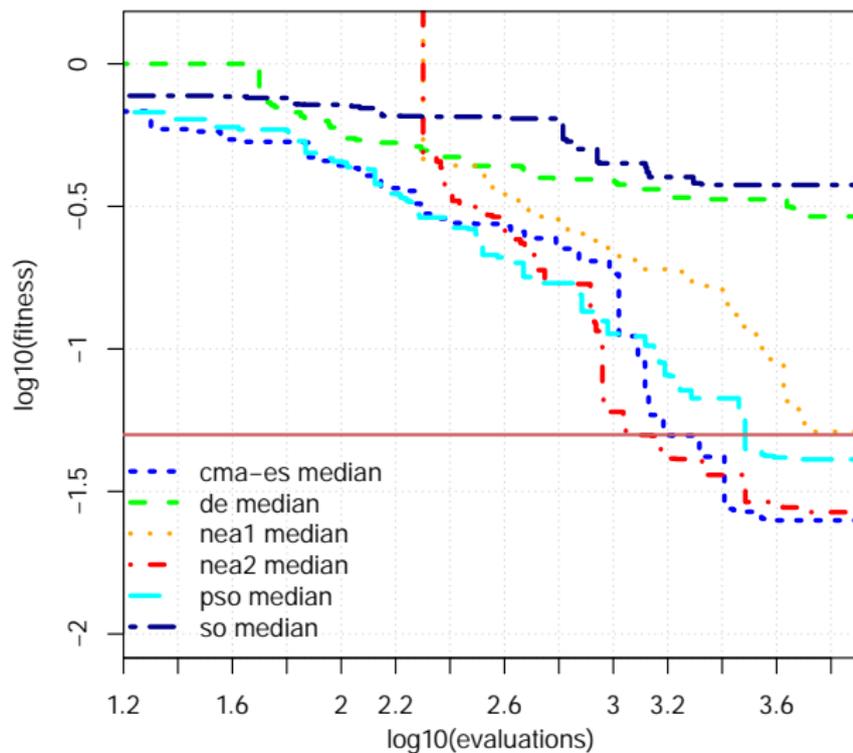
# Problem 1: Eavesdropping



## Problem 2: Ambush



## Problem 3: Chat



## Some numbers...

| divers. | ERT1  | ERT2  | ERT3  | sd1  | alg    | inst. |
|---------|-------|-------|-------|------|--------|-------|
| 0.151   | 1580  | 4710  | -     | 417  | pso    | 1     |
| 5.458   | 5868  | 7503  | 8250  | 1313 | de     | 1     |
| -       | -     | -     | -     | -    | so     | 1     |
| 7.370   | 740   | 1437  | 2018  | 524  | cma-es | 1     |
| 2.237   | 4266  | 5881  | 11314 | 1047 | nea1   | 1     |
| 8.968   | 1031  | 1444  | 2286  | 599  | nea2   | 1     |
| 4.568   | 1095  | 3004  | 8509  | 806  | pso    | 2     |
| 44.755  | 989   | 1238  | 1395  | 526  | de     | 2     |
| 0.131   | 95290 | 95501 | -     | -    | so     | 2     |
| 15.286  | 851   | 1266  | 2020  | 917  | cma-es | 2     |
| 7.351   | 3807  | 6094  | 10809 | 1226 | nea1   | 2     |
| 10.648  | 1338  | 2509  | 5276  | 1051 | nea2   | 2     |
| 0.150   | 5752  | 8203  | -     | 750  | pso    | 3     |
| 4.018   | 18566 | 19596 | 49899 | 414  | de     | 3     |
| 0.141   | 95354 | 96265 | -     | -    | so     | 3     |
| 4.650   | 2433  | 3937  | 11109 | 1069 | cma-es | 3     |
| 0.745   | 10252 | 11902 | 99635 | 718  | nea1   | 3     |
| 4.501   | 1587  | 3564  | 10687 | 1013 | nea2   | 3     |

## Observations/Discussion

- PSO good on problems 1 and 2
- CMA-ES always good
- NEA2 quite good, best on problem 3
- problem 3 seems to be more difficult than 1 and 2 (due to softer basin shapes?)
- SO and DE are not reliably able to provide valid solution
- CMA-ES and NEA2 most reliable for providing second solution

## What did we learn?...

- for diverse solutions we need some kind of restart
- multiple solutions require a bit more time (linear factor?)
- overall still too slow for realtime ( $\approx 1500$  evaluations)
- formerly applied algorithms (without restart) not reliable
- problem landscape topologies 'reflect' the visual impression
- huge differences, multimodal, ridges, plateaus