

# Handling Collusion in Dynamic Desktop Grids

Louis-Claude CANON, Emmanuel JEANNOT and Jon WEISSMAN

Project-Team AlGorille  
Loria/INRIA/Université Henri Poincaré  
University of Minesota  
ALEAE Project

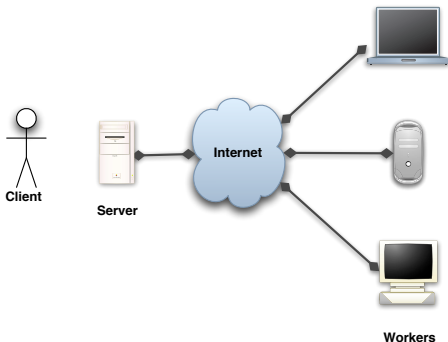
August 25, 2009

- 1 Collusion in Desktop Grids
- 2 Scheduling
- 3 Reputation system
- 4 Simulation
- 5 Conclusion

# Outline

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# Principles of Desktop Grid



## Distributed Computing

The server assigns **job** ( $j \in J$ ) to each active **worker** ( $w \in W$ ). Each worker returns a **result** ( $r \in R$ ) for the corresponding job. The client wants to have the *correct* result for each submitted job. Example: SETI@Home, BOINC-based projects.

# Cheating

## Byzantine fault

Unreliability or malicious isolated peer (worker).

35% of workers gives at least one wrong result (Derrick Kondo *et al.* 2007).

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## $k$ -majority

Assign a job to  $k$  distinct peers (redundancy) and select the result that has the majority.

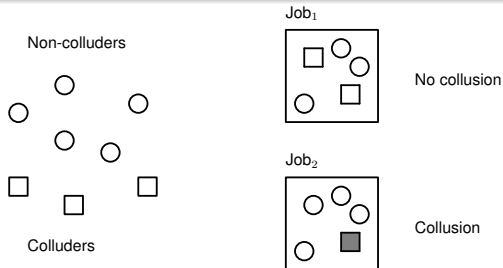
BOINC use a quorum-based system.

# Organized cheaters

## Collusion

A subset of the workers produces the same wrong result for some given job.

Against Sybil attack or library with platform specific bugs.



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# Mechanisms

## Generality

Generic and non-intrusive mechanism: no information available on the structure of the jobs and their results.

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## Main components

**Resource grouping** assures duplications of jobs by creating a set of workers for each job.

**Result certification** selects one of the results as the best one.

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## Objectives (Zhao and Lo, 2005)

**Overhead** minimum redundancy.

**Accuracy** maximum of correct certified results.

# Grouping strategies

## Detection orientation

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## Need for informations

In either cases, the probability to collude of a set of workers is needed. More generally, the colluding characteristics of the system.

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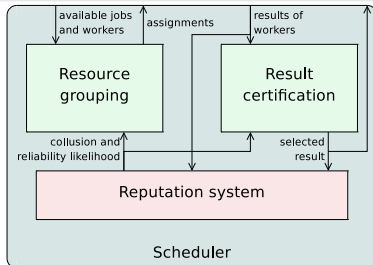
5 Conclusion

# Reputation system: an underlying problem

## Problem description

Based only on the generated results, what can we say?

Build a **reputation system** that will then be used at higher level.





# Objectives of the reputation system

## Goal

Determine the probability of collusion of any subset of workers.

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## Criteria (Jøsang, Ismail, and Boyd, 2007)

**Accuracy** Root Mean Square Deviation (RMSD) of all generated informations

**Adaptativeness or dynamicity** #iterations to acknowledge changes

**Robust** resistant to attacks

**Smooth** aggregated difference between 2 iterations

**Introspection** RMSD of error estimations

# Observations

## Input

$\langle t, w, j, r \rangle$  (timestamp, worker, job, result)

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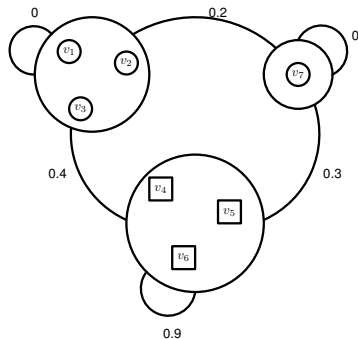
## Observations = group interactions

Consider that an observation informs us on the interaction between group(s) of workers. This involves to characterize the group to which each worker belongs.

# Architecture

## Data structure

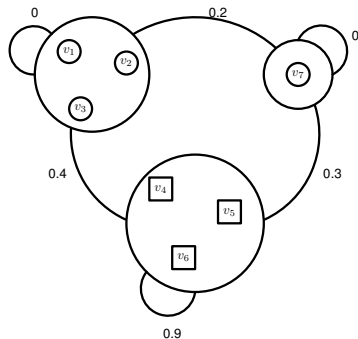
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## Algorithm

Initially, each worker is in a singleton.

We proceed by succession of merges and splits operations.

# Reputation representation

## Representation

- collusion likelihood between workers from set  $i$  and set  $j$ :  $c_{ij}$
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$$a_{ij} = 1 + 2 \times c_{ij} - c_{ij} - c_{jj}$$

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## Comparison

$\{c_{ij}\} \rightarrow \{a_{ij}\}$  but  $\{c_{ij}\} \not\leftarrow \{a_{ij}\}$ , then collusion likelihood has more informations.

Agreement likelihood is easier to manage, faster and more stable.

# Example of a merging criterion

Merging sets  $i$  and  $j$  with collusion representation.

Input:  $c_{ii}$ ,  $c_{ij}$  and  $c_{jj}$

$$c_{ii} \approx c_{ij} \approx c_{jj} \wedge \min(N(c_{ii}), N(c_{ij}), N(c_{jj})) > \max(\Sigma, \frac{n}{m}) \times \frac{\alpha}{(1-d_1)(1-d_2)(1-d_3)}$$

$N(c)$  number of observations with which  $c$  is estimated

$\Sigma$  sum of size of sets  $i$  and  $j$

$n$  number of workers

$m$  number of characterized groups

$d_k$  difference between  $c_{ii}$  and  $c_{ij}$

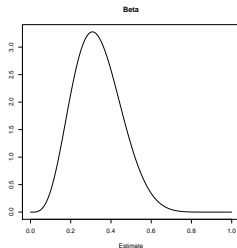
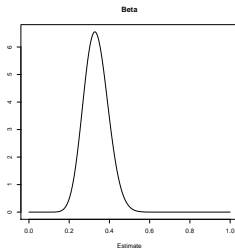
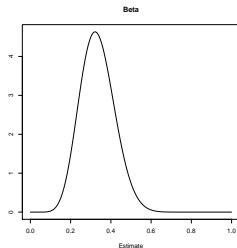
$$\alpha = \begin{cases} 2.5 & \Sigma > B \\ 1 & \text{otherwise} \end{cases} \quad B \text{ is the size of the biggest characterized set}$$

# Estimator

## How to deal with uncertainty (sparse observations)?

Values are modeled by random variables, *e.g.* Beta function (Jøsang, Ismail, 2002).

Arithmetic operation propagates errors (variance).



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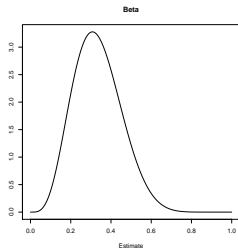
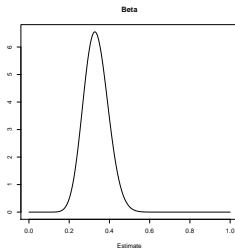
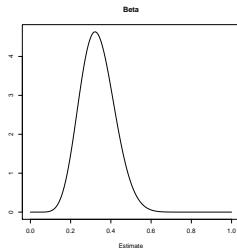
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## How to deal with dynamicity?

Bayesian inference is used to reinitialize each estimator when estimation is too far from recent observations.



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# Trace preparation (1)

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We want  $\langle t, w, j, r \rangle$ .

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We mix FTA and GWA trace together.

Number of needed cores for each job in GWA is eluded.

GWA is for HPC platform not Desktop grid (still better than arbitrary model).



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## Other adjustment

The same performance file is used for distinct availability trace because availability and power are uncorrelated on SETI@home.

# Trace preparation (2)

## Scheduling

Redundancy-based scheduler: achieve a quorum of  $q$  with initial and maximal duplication  $l$  and  $l_{\max}$ .

Jobs scheduled on workers when available (durations determined with the performance file and the normal duration of the job).

Result computations are based on reliability and colluding probabilities.

## Trace preparation (2)

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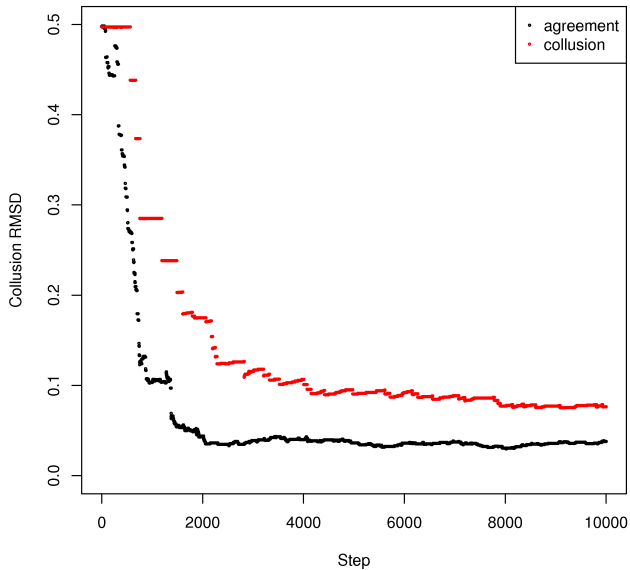
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### Trace that could have been useful

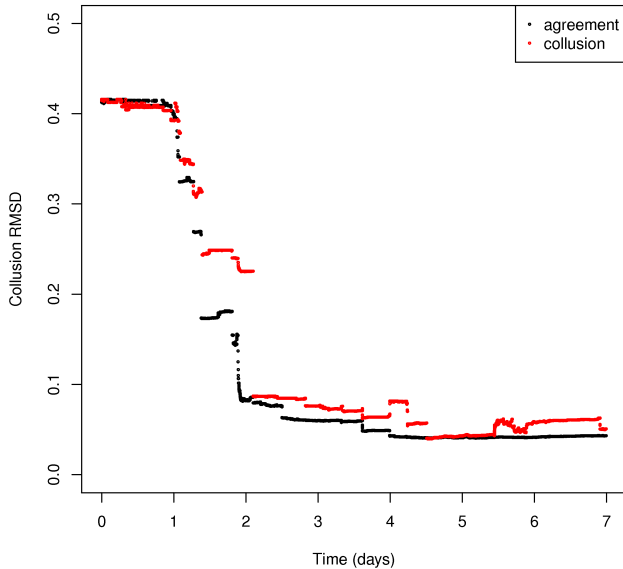
Job-centric traces on Desktop Grid: start and end times of the computation of a job; hash of the result.

# Results with synthetic input

Evolution of collusion error



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# Conclusion and future directions

## Main contributions

- Propose a reputation system
  - that observes interaction between groups of workers
  - with 2 representations (agreement, collusion)
- Use a realistic input based on existing traces

## Perspective

- Correct the dependance reliability/collusion (the system does not converge)
- Apply clustering techniques based on agreement distance