### **A Grid-oriented Genetic Algorithm**

José Herrera Sanz Eduardo Huedo Cuesta Rubén Santiago Montero Ignacio Martín Llorente







Advanced Computing Laboratory Associated to NASA Astrobiology Institute CSIC-INTA





Distributed Systems Architecture and Security Group Universidad Complutense de Madrid



# Outline



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  API
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# **Motivation**



- **GAs** are stochastic search methods that have been successfully applied in many search, optimization, and machine learning problems.
- **PGAs** offers many advantages over the traditional GAs (speed, wok in a larger search space, and less likely to run into a local optimum).
- With the advent of Grid computing, the computational power that can be deliver to the applications have substantially increased.
- PGAs can potentially benefit from this new Grid technologies.
- Implementation and execution of **PGAs** in a Grid involve challenging issues.
- Our research: PGA across the Grid using the DRMAA standard API and the GridWay framework.
- The efficiency and reliability of before schema to solve the One Max problem is analyzed in a globus-based research testbed.

# The GridWay Framework



**GridWay** provides an easier and more efficient execution (*submit & forget*) on **heterogeneous** and **dynamic** Grid.



#### **Characteristics**

- **Dynamic Scheduler**: GridWay periodically adapts the scheduler to the available resources
- **Resource Selector**: Reflects the applications demands, in terms of requirements and preferences.
- Adaptive Job Execution: To migrate running applications to more suitable resources.
- Fault tolerance (callbacks) and Job exit codes (Job-manager).

### DRMAA



### **Distributed Resource Management Application API**

- The DRMAA specification constitutes a homogenous interface to different DRMS to handle job submission, monitoring and control, and retrieval of finished job status. Moreover, DRMAA has been developed by DRMAA-WG within the Global Grid Forum (GGF).
- The DRMAA standard represents a suitable and portable framework to express this kind of distributed computations.
- Some DRMAA interface routines:
  - Initialization and finalization routines: drmaa\_init and drmaa\_exit.
  - Job submission routines: drmaa\_run\_job and drmaa\_run\_bulk\_jobs.
  - Job control and monitoring routines: drmaa\_control, drmaa\_synchronize, drmaa\_wait and drmaa\_job\_ps.

• DRMAA interface routines has been implemented within the **GridWay** framework.

### **Development Model**





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### **Parallel Genetic Algorithms (1/2)**



### Single Population (Panmitic GA)

- Usually implemented using a Master/Worker paradigm.
- Can be efficiently used when evaluation function requires a considerable amount of computational work.
- Main advantage: the search behavior of the sequential GA is not altered.
- **Disadvantage**: This approach **is not well** suited for a **Grid** because of the high network requirements of its communication patters.

#### **Single Population (Fine Grain GA)**

- Only **one population** and its spatial structure limits the interactions between individuals.
- This limit can be imposed:
  - Chromosome level: each member can only interact with their neighbors
  - **Population** level: **only member** of the **same subpopulation** may mate during crossover.

# Parallel Genetic Algorithms (2/2)



#### **Coarse Grain GA**

- Main population is divided into subpopulations (demes) each one independently evaluated in a different node.
- **Tree** possible communication patters:
  - **Ring model**: processes can only interact with their neighbors in a ring topology
  - Master-slave model: slave processes swap best individuals with the master.
  - All-to-all model: All processes swap best individuals with the others.
- **Disadvantage**: Introduce fundamental **changes** in the implementation of a **simple GA**.
- Advantage: It is more tolerant to the high latencies and dynamic bandwidths that can be expected in the Internet, unlike the single population alternatives.
- In this research we use a modified version of the coarse grain approach, since this algorithm does not imply a tightly coupled deme topology

# **Algorithm Description**



#### **Main Characteristics**

- We use a **fully connected multi-deme** genetic algorithm, **all demes exchange** individuals every generation.
- Not imply any overhead since the population of each deme is used as checkpoint files.

#### **Algorithm Execution**

- Initial **population** is uniformity **distributed** among available number of **nodes**.
- Sequential **GA** is locally **executed** over each **subpopulations**.
- Worst individuals of each subpopulation are exchanged with the best ones of the rest.
- New population is generated to perform the next iteration.

#### **Algorithm Optimization**

- Previous algorithm may incur in **performance losses**, since the **iteration time** is determined by the **slowest machine**.
- Solution ⇒ Dynamic Connectivity:
  - We allow an asynchronous communication pattern between a fixed number of demes.
  - Minimum number of demes in each iteration depends on the numerical characteristics of the problem.

# **Algorithm Schema**





# **Experiences (1/5)**



#### **Testbed description**

Host	Model	Hz	OS	Memory	Nodes	GRAM
babieca	Alpha DS10	466Mhz	Linux 2.2	256MB	5	PBS
hydrus	Intel Pentium 4	2.5 Ghz	Linux 2.4	<b>512MB</b>	1	fork
cygnus	Intel Pentium 4	2.5 Ghz	Linux 2.4	<b>512MB</b>	1	fork
aquila	Intel Pentium III	666 Mhz	Linux 2.4	128 MB	1	fork

#### **Objectives**

- We evaluate the functionality and efficiency of the Grid-oriented Genetic Algorithm, in the solution of the **One-Max problem**.
- One-Max problem is a **classical benchmark** problem for genetic algorithm computations, and it tries to evolve an initial matrix of **zeros** in a matrix of **ones**.

# **Experiences (2/5)**



#### **Experience description**

- We will consider an initial population of **1000** individuals each one a **20x100** zero matrix.
- Sequential GA:
  - Iterations: 50
  - Mutation Probability: 0.1%
  - Crossover Probability: 60%
- The exchange probability of best individuals between demes is **10%**
- Final score must be a **1800** ones **matrix**.

#### **Two experiments**

- Execution profile of **4 generations** of the **GOGA**, with a **5-way** *dynamic connectivity*.
- 5 different executions of GOGA, with different degrees of *dynamic connectivity*: 2-way, 3-way, 5-way, 6-way and 8-way.

# **Experiences (3/5)**





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# **Experiences (4/5)**





### **Experiences (5/5)**





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- We have presented an efficient Grid-oriented genetic algorithm.
- Our approach uses a fully connected multi-deme GA, with a dynamic connectivity between subpopulations to deal with the heterogeneity of the Grid.
- The **optimum degree** connectivity depends on:
  - The computational characteristics of the Grid nodes.
  - The computational problem.
- The **GOGA** has been developed taking advantage of the **GridWay** framework features and the **DRMAA API**.
- It have been shown that **DRMAA** can aid the rapid development and **distribution** across the **Grid** of typical genetic algorithm strategies.

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