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a generic distributed parallel genetic algorithm development platform

Nicolas Kruchten 4th year Engineering Science (Infrastructure Option)

Last Week

- Introduction to Galapagos:
 - Background
 - Motivation
 - Capabilities
 - What but not how...

Major Points

- Galapagos is a platform with which we can develop distributed parallel genetic algorithms
- Galapagos is very flexible
 - Many (GA-appropriate) problems
 - Many GA types
 - Operators
 - Population structure
 - Many distribution schemes

Today

- Technical explanation of Galapagos
- Approach: 3 views of Galapagos' flexibility
 - Problem-domain
 - GA types & parallel population structure
 - GA distribution architecture

Problem-Domain

- GAs are optimization tools
- Need objective function

 Not always real function of real variables
- Galapagos can be used with a wide variety of objective functions

Object Oriented

- Galapagos defines and executes a set of 'black-box' steps in a flowchart
- Internal workings of boxes, and data types they operate on are up to the developer
- Galapagos provides templates for building new data types and GA operators





GA Types: Galapagos GA Flow Diagram



Initializer

- picks the initial population
- default: random, specified min/max
- other uses: seeding the population



evaluation

- handled by Galapagos
- usually handed to LightGrid for distributed processing
- can be handled on local machine (single-computer version)



Generator

- Creates a generation of children
- GA *usually* works with:
 - A Mutator
 - A Recombiner (compare: EP)

each child = mutate(recombine())

(mutate the output of recombination operation)



evaluation

• same as above



Assembler

• Creates a new population, using an existing population and an incoming generation

- GA falls roughly in:
 - Simple ('canonical')
 - Crowding ('steady-state', 'incremental', 'truncation')
- Crowding *usually* implemented using a Selector
- This is where elitism occurs

Convergence

- Defines some end condition for the run
- Usually based on:
 - A real time measure
 - An 'algorithm time' measure
 - Some parameter (mean, std etc.)

end

• Run is over

• Whatever post-processing, logging etc happens here

Epoch

- Asynchronously evaluated
- Defines some timeframe:
 - Real time
 - 'algorithm time'
 - Some parameter (mean, std, etc.)

Migrator

- Required for Parallel GA (multiple populations)
- On sending side, defines:
 - migrant destination
 - migrant number
 - migrant selection
- On destination side, defines:
 - replacement policy
- usually implemented with:
 - Topology
 - Selector
 - Assembler

Supporting Operators

- Generator:
 - Mutator
 - Recombiner
- Assembler:
 - Selector
- Migrator:
 - Topology
 - Selector
 - Assembler (already covered)

Mutator

- Basic EA staple
- Input: a chromosome
- Output: a chromosome
- Function: performs some mutation
- Usually implemented using a GeneMutator, in which case Mutator selects which genes are mutated, GeneMutator governs how that occurs

GeneMutator

- Usually called from a Mutator
- Input: a gene
- Output: a gene
- Function: mutate that gene according to some rule

Recombiner

- Input: none
- Output: a chromosome
- Function: uses an internal Selector to pick parents, creates n children according to some rule
- If no recombination/crossover occurs, this is EP, not GA

Selector

- Shows up everywhere!
- EA staple
- Input: selection pool, n
- Output: n chromosomes
- Function: selects n chromosomes according to some rule

Topology

- Required in Parallel GA
- Defines migrant destination, number per destination
- Defines nature of PGA:
 - Coarse-grained/Fine-grained
 - Deterministic /Stochastic

Operators

- Initializer
- Generator
- Assembler

- Mutator
- GeneMutator
- Recombiner

- Convergence
- Epoch
- Migrator

- Selector
- Topology

Summary

- Meta-operators and supporting operators provide a well-defined yet flexible GA base
- Migrator/Epoch/Topology operators allow for wide variety of PGA types

evaluation

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LightGrid explained

- Clients and Resources
- Dispatcher
- Generic: can be used to implement any distributed application

Grid Computing

Master / Worker

Hybrid: Peered Masters with Worker Pool

High-Level Process Diagram

Process view

- 4 major process types:
 - Container (LightGrid Clients)
 - Evaluator (LightGrid Resources)
 - Controller
 - LightGrid Dispatcher
- Each process holds more than one thread of execution (sub-process)

Process View II

- Containers (LightGrid Clients):
 - Contain and manage populations
 - Interface between populations and evaluators
 - Interface between individual populations
- Evaluators (LightGrid Resources):
 - Compute the fitness value for a given chromosome

Process View III

- Controller:
 - 'where the user sits'
 - Input/Output, control process
 - Issues start/stop/reset messages
- LightGrid Dispatcher:
 - Dispatches jobs from Clients to Resources and passes the results back

High-Level Process Diagram

Controller Process Diagram

Container Process Diagram

Dispatcher Process Diagram

Dispatcher

Evaluator Process Diagram

Physical View

- Given N computers, M populations, what goes where?
 - Nature of fitness
 - Nature of operators
 - Nature of computers
 - Nature of network
- Galapagos is flexible enough to be deployed almost anywhere

Summary

- Distributed PGA design is very complex:
 - Problem-domain specifics
 - GA type & population structure design
 - Available hardware
- Galapagos doesn' t provide easy answers but does provide a flexible common basis to work from

Questions?

nicolas@kruchten.com

this presentation is available at: http://nicolas.kruchten.com/

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