CSE/ECE 848 Introduction to Evolutionary Computation

Module 3 - Lecture 14 - Part 1

Comparison of EC Methods:

Motivation

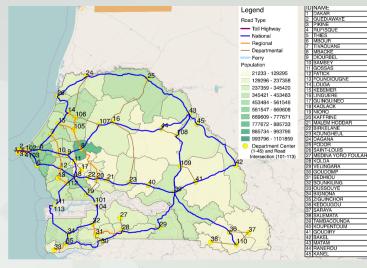
Wolfgang Banzhaf, CSE
John R. Koza Chair in Genetic Programming

What do we want?

- What do we want an EA to do for us?
 - We want them to solve our problem
- Applications
 - Design (one-off) problems
 - Repetitive problems, including online problems
- Academic Research
 - Method exploration

A Design Application

- Example: Optimization of extensions of an existing road network
- Complex, multi-objective problem, subject to many constraints
- One excellent solution created once is sufficient
 - Solution quality is priority
 - Time to solution not important
 - OK, if algorithm does not reach solution for some initial conditions
- Specific aspects allow to research special operators/representations not transferable to other problems



From: Y. Wang et al., "National and Regional Road Network Optimization for Senegal using Mobile Phone Data", NetMob 2015



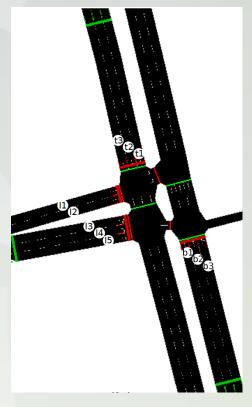
- Example: Domestic transportation firm with several trucks and drivers, given daily schedule every morning
- Pick-up and delivery plan for each truck and driver
- EA must find a good solution quickly, reliably, and repeatedly, i.e. for different instances
 - Speed is very important
 - Solutions must be good, but not necessarily optimal
 - There might be only time for one run, therefore, quality should have low variance
- Algorithm might be used in many different companies



From: caliper.com

A Online Application

- Example: Traffic light controller
- Set up traffic signals to optimize traffic flow under different conditions (weather, daytime, etc.)
- Sensors allow to calculate density
- Online mode, optimizing a baseline schedule to adapt to current (hour to hour) conditions
- Consecutive problems are similar
- Repetitive problem solving
- Must be fault tolerant and robust (e.g., noise in the data or missing data should not prohibit solution)



-rom: SUMO traffic control simulator illustration as taken from E. Ricalde, PhD thesis, 2019



- Get a good solution for a given challenging problem
- Show EAs are applicable in a novel problem domain
- Show that an EA with some new features is better than another EA

Show that the EA is better than an existing non-evolutionary

algorithm

- Find best parameter set-up, given conditions
- Obtain insight into the workings of the algorithm
- See how algorithm scales with problem size



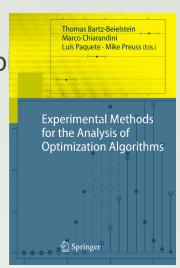
Design of Computational Experiments

- A generic problem (e.g. TSP) and its instances (particular numerical cases)
- Experiments involve running a set of algorithms on a number of instances of the problem
- It starts with a set of questions about the heuristics under study
 - Problem characteristics, e.g. size, #constraints ...
 - Algorithm components, e.g. stopping criteria, move selection

•	Often: Pro		Algorithm 1	Algorithm 2	 Algorithm n	d,
	instances	Instance 1				es)
		Instance 2				
		Instance m				



- Randomized algorithms (like GAs) require on top of that: Repeated runs under same parameter conditions, as results are different from run to run
- Often experimental factors are run with just 2 conditions: high and low, but to get clarity whether there is a linear relationship, we need at least 3 conditions!
- Example:
 - 8 problem factors, with only 2 conditions -> 256 comb
 - 3 replicates -> 768 combinations
 - 3 algorithms factors with only 2 conditions -> 6144 combinations
 - 5 times with different random seeds -> 30,720 runs !!



Experimental Analysis Myths

(from Wineberg and Christensen)

- Results from 1 run is all that is needed
 - No, shows only proof of concept
- The best value achieved in a set of runs tells you something about the population distribution
 - No
- Using the same RNG seed for both systems provides a fairer comparison
 - It doesn't it's the statistical properties of the system that we are looking for
- One system is obviously better than the other when looking at the data/graphs no statistics necessary
 - If it is so obvious, one might as well show with statistics
- My average is better than your's means my algorithm is better than your's
 - In the best case, you would need to take variance into account