



- NIH may be a part of a larger system
- to improve on existing techniques
- to improve NIH search
- ...





- creation of initial solutions
- · local improvement of candidate solutions
- intelligent decoders
- intelligent / heuristic variation operators

How to Hybridize?

parallel populations
 – same / different heuristic

- same / different metaheuristic _
- different parameter settings - different fitness functions
- · approximate models
 - costly fitness evaluation - use approximate model for some evaluations
 - use different levels of approximation for sub-populations
 - hierarchical model

How to Hybridize?

- modify problem instance - e.g. decrease search space size
- partition into sub-problems
- · interactive iterations
 - for local tuning
 - for constraints





Heuristics for Initializing Population

- *n*-way tournament among randomly created solutions
- multi-start local search: pick *N* points randomly to climb from
- · constructive heuristics often exist

Heuristics for Initialising Population

- · diversity is important
- advantage: good solutions found quickly
- disadvantage: possible to get stuck at local optima (strong bias)

Intelligent Operators

- incorporating problem or instance specific knowledge within operators
- usually with problem specific representations
- usually fast

Local Search Acting on Offspring

- to speed-up the NIH
- NIH for exploration, LS for eploitation
- makes search around good solutions more systematic
- · fast local optimizer needed
- · smoothes fitness landscape
- · introduces redundancy and plateaus
- · very successful in practise

Example: Memetic Algorithms

- combination of EAs with local search in EA loop: Memetic Algorithms
- also EAs using instance specific info in operators
- shown to be faster and more accurate than EAs on some problems
- are the "state of the art" on many problems: e.g. scheduling and timetabling problems

Local Search

- *neighbourhood* concept
- *N*(*x*): set of points that can be reached from *x* with one application of a move operator

 e.g. bit flipping search on binary problems



Local Search

- *degree* of graph: max. no of edges coming into/out of a single point
 - size of biggest neighbourhood
- local search look at points in neighbourhood of a solution
 - complexity related to degree of graph
 - bit-wise mutation on binary

Local Search

- is neighbourhood searched randomly, systematically or exhaustively ?
- does search stop as soon as a fitter neighbour is found (*Greedy Ascent*)
- or is whole set of neighbours examined and the best chosen (*Steepest Ascent*)
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Local Search

- local search in representation or solution space ?
- how many iterations of local search ?
- local search applied to whole population?
 - or just the best ?
 - or just the worst ?

Two Models of Lifetime Adaptation

- Lamarkian
 - traits acquired by individual during lifetime transmitted to offspring
 - e.g. replace individual with fitter neighbour
- Darwinian
 - traits acquired by individual during lifetime not transmitted to offspring
 - e.g. individual receives fitness (but not genotype) of fitter neighbour

Two Models of Lifetime Adaptation

- Baldwinian effect: individual learning improves evolutionary learning by changing fitness landscape (both models)
- Lamarckian good in stationary environments
- Darwinian good in dynamic environments

Intelligent Decoders

- · indirect representation
- use a decoding function •
 - decoding function uses problem specific info
- representations - permutations
 - random keys
 - weight codings
- ٠ good with handling constraints
- ٠ time consuming
- locality problem •

Hybrid Algorithms Summary

- hybridize especially for real world problems
- · hybridization may involve
 - use of operators from other algorithms
- incorporation of domain-specific knowledge hyrid algorithms
- shown to be much faster and more accurate on some problems - the "state of the art" on many problems
- more problem specific
- · requires more parameter settings
- · possible loss of creativity