

Instructors

Squillero and Tonda have been working on diversity in EAs since 2008, and in 2015 they published the paper "Divergence of character and premature convergence: A survey of methodologies for promoting diversity in evolutionary optimization". In 2016 and 2014 they gave introductory tutorials about the topic, respectively, at PPSN (Edinburgh) and IEEE WCCI (Beijing): the instructors also co-chaired *MPDEA*, a GECCO workshop on Measuring and Promoting Diversity in Evolutionary Algorithms in 2016 and 2017.

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Course Agenda



- Divergence of character in natural and artificial evolution
- Background (diversity and similarity, …)
- Mechanisms for promoting diversity
- Hints and tips
- Conclusion











Premature convergence

- I.e., the tendency of an algorithm to converge towards a point where it was not supposed to converge to in the first place
- Probably an oxymoron
- Holland's "Lack of speciation"
- EAs general inability to exploit environmental niches

Divergence of character

"The basic point of the principle of divergence is simplicity itself: the more the coinhabitants of an area differ from each other in their ecological requirements, the less they will compete with each other; therefore natural selection will tend to favor any variation toward greater divergence."



divergence of character VS. *b*_{remature} convergence 10











Exploration vs. Exploitation

Recombination

- mixes together two or more solutions to create the offspring
- associated with the idea of exploration

Mutation

- performs a (usually small) **change** in an individual
- · associated with the idea of exploitation



Exploration vs. Exploitation

- When all parents are very similar, the effectiveness of recombination is limited
- The ability to explore remote parts of the search space is impaired
- "Conventional wisdom suggests that increasing diversity should be generally beneficial"

Exploration vs. Exploitation

- When all parents are very similar, the effectiv what is the definition of "similar"? limited
- The ability to explore remote parts of the search space is impaired
- Conventional wisdom suggests that increasing diversity should be generally beneficial[®] and the definition of "diversity"?

Levels in biology

- Genotype: the genetic constitution of an organism
- Phenotype: the composite of the organism's observable characteristics or traits
- Fitness: individual's ability to propagate its genes (well, almost)



19













<text>







End goal vs. Means goal

- The end goal in optimization is reaching better solutions in less time
- Promoting diversity has often been seen as the key factor to improve performances
- Promoting diversity is a mere means goal (yet a quite important one)
- No distinction is made here whether the means goal is
 - preserve existing diversity
 - increase diversity

How diversity is promoted (practice) Fitness scaling Fitness holes Tweaking selection mechanism Adding selection mechanism Multiple populations

- Population topologies
- *****...

In theory there is no difference between theory and practice







Lineage-based methodologies

- The value of ξ(°) does not depend on individual structure nor behavior, but it can be determined considering circumstances of its birth (e.g., time, position)
- LBMs can be applied to any kind of problem, even in addition to other diversity preservation methods

Genotype-based methodologies

- Particularly effective when it is possible to define a sensible distance between genotypes
- Often used to
 - avoid overexploitation of peaks in the fitness landscape
 - promote the generation of new solutions very far from the most successful ones
 - preserve variability in the gene pool

Phenotype-based methodologies

- ✤Usually impractical
- Sometimes fitness distance can be used as a proxy for phenotype distance (multi objective EAs, or many objective EAs)

Type of selection

- ◆ Parent selection (α)
 - Usually non-determinstic
- ♦ Survival selection (ω)
 - · Usually deterministic









- · Same as island models
- The selective pressure decreases during evolution









Deterministic crowding

Recipe [LIN ω]

· Offspring compete against parents for survival

Rationale

- Niching with implicit neighborhood
- Parents and offspring occupy the same niche
- · No need for evaluating the similarity

Allopatric selection

Recipe [LIN ω]

- · The whole offspring compete for survival
- Rationale
 - Niching with implicit neighborhood
 - · No need for evaluating the similarity
 - Genetic operators that create large offspring can be exploited without the risk for the offspring to invade the population

Fitness Sharing

Recipe [GEN αω]

Scale down individual fitness

•
$$\bar{f}(I_k) = \frac{f(I_k)}{\sum_i sh(I_k, I_i)}$$

• with sh(x, y) depending on the distance between the individuals, and is 0 beyond a fixed radius

Rationale

- Niching with explicit neighborhood
- · Reduce attractiveness of densely populated area

Clearing

Recipe [GEN αω]

- Inside niches of a certain radius, the best k individuals retain their fitness while the rest are zeroed
- Rationale
 - Niching with explicit neighborhood
 - · Set a hard limit to population density

49

Standard crowding

Recipe [GEN ω]

 New individuals replace the most similar individual in a random niche of size CF

Rationale

- · Niching with explicit neighborhood
- Favor novelty (generational approach)

Crowded-comparison operator

Recipe [PHE α]

- Estimate the free territory around solutions and favor solutions less crowded regions
- Rationale
 - · Smart implementation of artificial niches
 - Requires a strong correlation between phenotype and fitness
- NSGA-III introduces ε-domination (adaptive discretization)

Reference points partitioning

Recipe [GEN ω]

- Population is partitioned using in clusters centered around a set of reference points
- Reference points are initially chosen by the user, then can be dynamically updated
- New individuals compete for survival inside their own niche

Rationale

Niching with explicit neighborhood

Vector evaluated genetic algorithm

Recipe [PHE α]

- Divide the mating pool in N parts, each one filled with individual selected on their i-th component of the fitness
- Alternative: select on a weighted sum, but use different weight sets for the different parts
- Rationale
 - · Increase the push towards specialization
- Caveats
 - Only applicable to MOEAs, or when using an aggregate fitness

53

Airplane tickets

Ticket

Lexicase selection

Recipe [PHE α]

- Before selection, re-arrange the components of the fitness
- Compare individual fitnesses lexicographically

Rationale

• Increase the push towards specialization

Caveats

Only applicable when using an aggregate fitness

57

Restricted tournament selection

Recipe [GEN ω]

- New individuals compete with the most similar individual in a random niche of size CF
- ✤ Rationale
 - Niching with explicit neighborhood

Sequential niching

Recipe [GEN αω]

• The most promising points in the search space after each run are altered so to become less interesting in further executions

Rationale

· Avoid over exploitation

Gender

Recipe [LIN/GEN α]

- Add gender to individual and enforce sexual reproduction
- More than two sexes are possible, with different mutation probabilities
- · Gender might be part of the genome or not

✤ Rationale

- Prevent crossover between clones
- · Limit interactions between related individuals

Tarpeian method

Recipe [PHE αω]

Randomly kill individual who don't adhere to given standards

Rationale

- Note: originally used to prevent bloat
- Creating dynamic and non-deterministic fitness holes may have several beneficial effects, including to promote diversity

Diversifiers

Recipe [GEN αω]

- Detect less populated areas in the search space and try to generate random inhabitants
- ✤ Rationale
 - Increase variability in the gene pool regardless the fitness
 - Require a reliable distance metric

62

Random immigrants

Recipe [PHE αω]

• Periodically insert random individuals in the population

✤ Rationale

• Try to introduce novelty

Caveats

• Newborns may need to be artificially kept alive when competing against already optimized individuals

Extinction

Recipe [PHE ω]

- Upon convergence (or periodically) remove a significant part of the population
- Then fill up the population with the offspring of the survivors and/or random individuals

Rationale

 A gust of fresh air: already optimized individuals are not enough to occupy the whole population and newborns may start exploring new regions

Caveat

• Fitness variability used as phenotype variability



GDEM – Genetic Diversity Evaluation Method

- Recipe [GEN αω]
 - · Add diversity as an explicit goal and go MO
- Rationale
 - · Modify the domination criteria
 - · Need a reliable diversity metric
- Historical note
 - See: Find Only and Complete Undominated Sets (FOCUS)

Delta entropy and pseudo entropy

Recipe [GEN α]

• With a certain probability select individuals on their ability to increase the global entropy of the population instead of fitness

Rationale

- Not-so-fit individual with peculiar traits should be preserved
- Measuring the entropy of the population is easier than defining a distance function

67

Outline

- ♦ Generic EA
- Divergence of character in natural and artificial evolution
- Background (diversity and similarity, …)
- Mechanisms for promoting diversity
- Hints and tips
- Conclusion

Hints and Tips

- Do you really need to promote diversity?
 - Several problems in EA are caused by ill-designed fitness functions
 - Check whether the locality principle holds true
 - · Check what happen with multistart



Hints and Tips

- Do you really need to promote diversity?
- ♦ Use extinction (20m)
 - Simple n' easy



Hints and Tips

- Do you really need to promote diversity?
- ✤ Use extinction (20m)
- Use lexicase selection (20m)
 - Simple n' easy
 - Only useful for aggregate fitness (combination of several components)



Hints and Tips

- Do you really need to promote diversity?
- ✤Use extinction (20m)
- ♦ Use lexicase selection (20m)
- ♦ Use an island model (2h)
 - Far better than multistart (if migrations are properly handled)
 - Only useful if different experiments yield different results





Hints and Tips

- Do you really need to promote diversity?
- ♦ Use extinction (20m)
- Use lexicase selection (20m)
- ♦ Use an island model (2h)
- Use fitness holes (20h)
- Use real niching (2-20d)
 - Only useful if the distance between genotypes is meaningful



Outline

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