

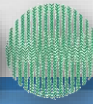
Promoting Diversity in Evolutionary Optimization: Why and How

Giovanni Squillero

`giovanni.squillero@polito.it`

Alberto Tonda

`alberto.tonda@grignon.inra.fr`



INRA

1

Copyright © 2016 by Giovanni Squillero and Alberto Tonda.

Permission to make digital or hard copies for personal or classroom use of these files, either with or without modification, is granted without fee provided that copies are not distributed for profit, and that copies preserve the copyright notice and the full reference to the source repository. To republish, to post on servers, or to redistribute to lists, contact the Authors. This file is offered as-is, without any warranty.

2

Saturday, September 17th 2016



squillero@polito.it PPSN16 3

3

Outline

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

squillero@polito.it PPSN16 4

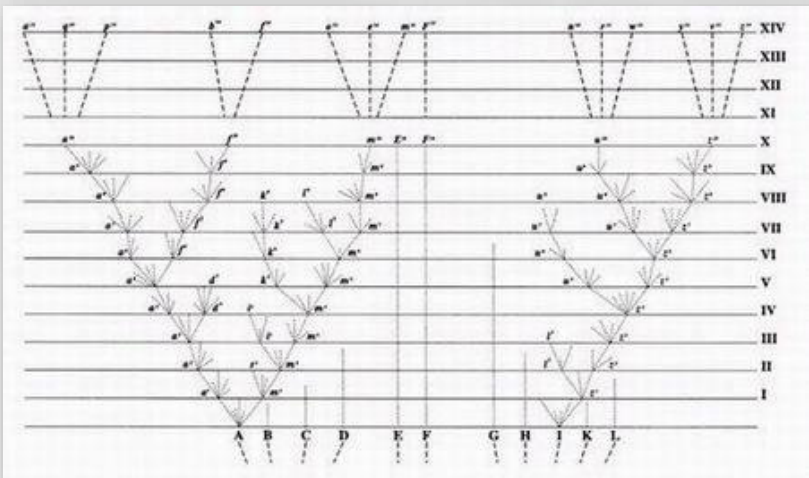
4

Real world (Galapagos)



5

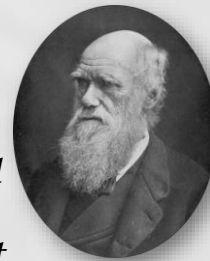
Darwin's tree of life



The only illustration in *On the Origin of Species by Natural Selection* (1859)

6

Divergence of character

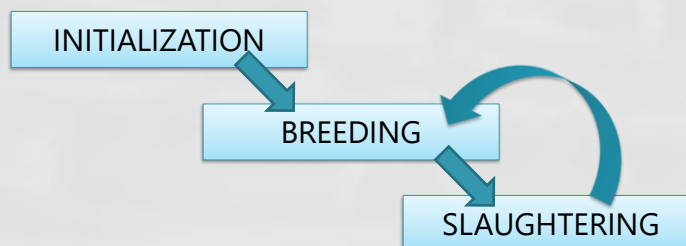


- *"Great diversity of forms in nature"*
- *"The principle, which I have designated by this term, is of high importance, and explains, as I believe, several important facts"*
 - *"The principle of divergence causes differences, at first barely appreciable, to steadily to increase, and the breeds to diverge in character, both from each other and from their common parent"*
 - *"The varying descendants of each species try to occupy as many and as different places as possible in the economy of nature"*

7

Evolutionary computation


- A rough idea about "what" an evolutionary algorithm is



- Note: Optimization, not artificial life!

8

Evolutionary algorithms



squillero@polito.it PPSN16 9

9

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

squillero@polito.it PPSN16 10

10



divergence of character
vs.
premature convergence

squillero@polito.it PPSN16 11

11

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.”




squillero@polito.it PPSN16 12

12

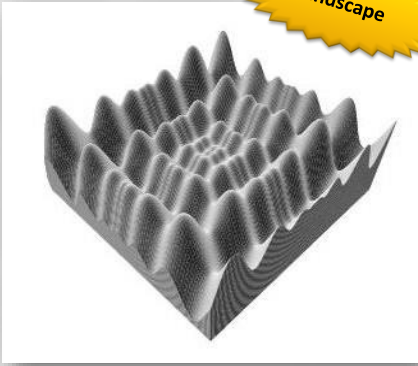
Environment vs. Fitness function

ecosystem



environment

landscape

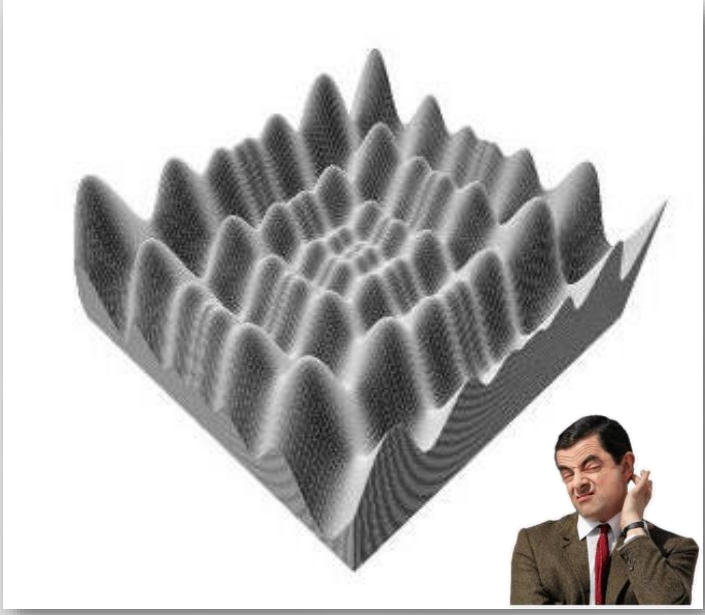


squillero@polito.it

PPSN16

13

13



squillero@polito.it

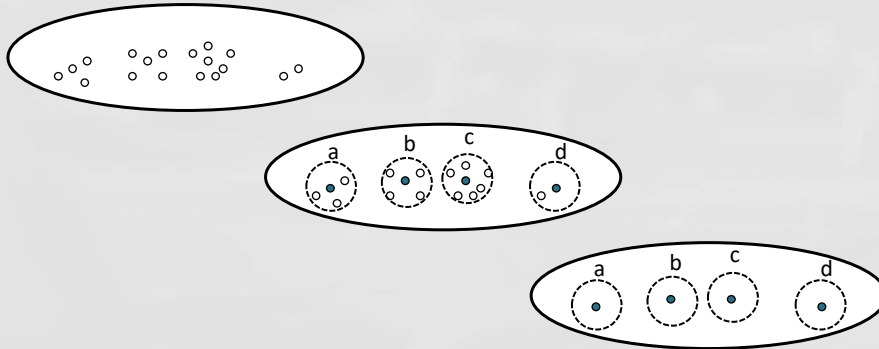
PPSN16

14

14

Niches

- Niche: subspace in the environment with a finite amount of physical resources that can support different types of life



squillero@polito.it

PPSN16

15

15

Niches

- Niches favor the divergence of character
- Niches and speciation
- How to create “niches” in EAs since the environment is missing?

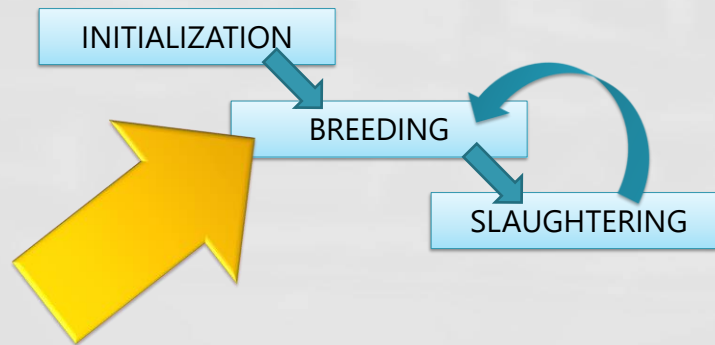
squillero@polito.it

PPSN16

16

16

A Very Generic EA



squillero@polito.it

PPSN16

17

17

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*



squillero@polito.it

PPSN16

18

18

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”



squillero@polito.it

PPSN16

19

19

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”

what is the definition of “similar”?

and the definition of “diversity”?



squillero@polito.it

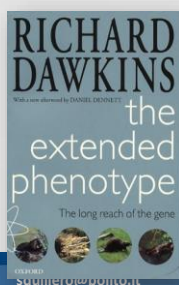
PPSN16

20

20

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)



Richard Dawkins

The Extended Phenotype: The Long Reach of the Gene
Oxford University Press, 1982 (revised ed. 1999)

PPSN16

21

21

Levels in EC (a modest proposal)

- **Fitness:** how well the candidate solution is able to solve the target problem
- **Genotype:** the internal representation of the individual, i.e., what is directly manipulated by genetic operators
- **Phenotype:** the candidate solution that is encoded in the genotype
 - the intermediate form in which the genotype needs to be transformed into for evaluating fitness
 - if genotype can be directly evaluated: genotype and phenotype coincide

squillero@polito.it

PPSN16

22

22

Levels in EC (GA)

The diagram illustrates the levels in Evolutionary Computation (EC) for a Genetic Algorithm (GA). It shows a sequence of three levels: Genotype, Phenotype, and Fitness, connected by dashed vertical lines.

Genotype: Represented by a horizontal row of six boxes. The first four boxes contain the values 0, 1, 1, and 0. The fifth box contains an ellipsis (...), and the sixth box contains the value 1. To the left of this row is the label I .

Phenotype: Represented by a single box containing the value 1. To the left of this box is the label P .

Fitness: Represented by a single box containing the mathematical expression $\sum_{k=0}^n I(k)$. To the left of this box is the label F .

The three levels are connected by dashed vertical lines, indicating a sequential process: Genotype leads to Phenotype, which leads to Fitness.

23

Levels in EC (GP & LGP)

Genotype

$$f(x) = \left(0.2 - \frac{x}{42}\right) \cdot \ln(x)$$

Phenotype

$$\text{Fitness} \approx \int_{x=0}^E |f(x) - g(x)|$$

Fitness

squillero@polito.it

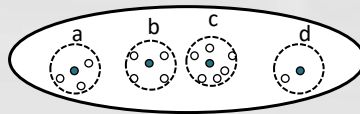
PPSN16

24

24

Niches in EA

- Niching: grouping similar individual
 - similar spatial positions (i.e., islands)
 - similar genotypes (i.e., niching)
 - similar phenotypes
- Several approaches are based on niching



25

Detecting clones

- Detecting whether two individuals are clones, i.e., identical, is often an easy task at any level



26

Measuring diversity

- Diversity \Rightarrow distance metric: *how far* the individual is
 - from (a subset of) the whole population
 - from a single individual
- Diversity \Rightarrow property of the population
- But, at what level?
 - Phenotype
 - Genotype
 - Fitness



27

Measuring diversity

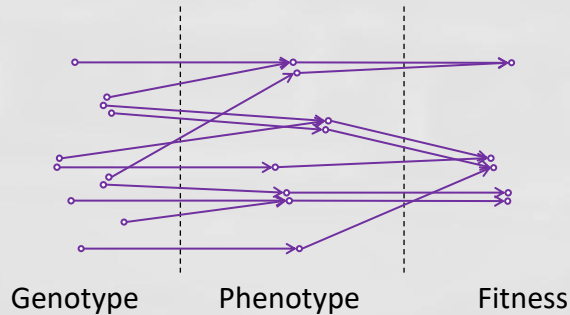
- Different fitness values imply different phenotypes, different phenotypes imply different genotypes

$$F_x \neq F_y \Rightarrow P_x \neq P_y \Rightarrow G_x \neq G_y$$

28

Measuring diversity

- What about “diversity”?
- Locality principle
- Rechenberg’s *strong causality*



squillero@polito.it

PPSN16

29

29

Measuring diversity

- Fitness
 - Usually trivial
- Phenotype
 - Usually ad-hoc
- Genotype
 - Different genotypes in the population
 - GP subtree frequency
 - Edit distance (a.k.a., Levenshtein distance)
 - Entropy and free energy

squillero@polito.it

PPSN16

30

30

Outline

- Generic EA
- Divergence of character evolution
- Background (diversity and similarity, ...)
- Mechanisms for promoting diversity
- Hints and tips
- Conclusion

What has been proposed to alleviate it?

31

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

32

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



squillero@polito.it PPSN16

33

How diversity is promoted (theory)

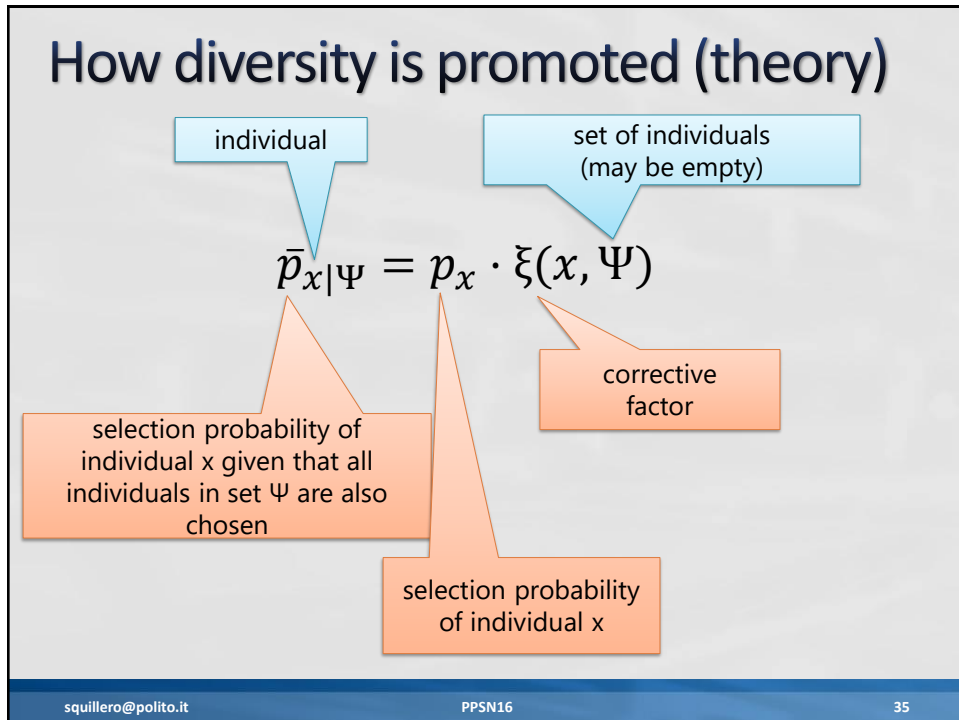
- A methodology for promoting diversity alters the selection probability of individuals

$$\bar{p}_{x|\Psi} = p_{x|\Psi} \cdot \xi(x, \Psi)$$

- **Mere definition:** we do not imply that a mechanism operates *explicitly* on the selection operators
- **But** the *effects* on selection probabilities are assessed to classify it

squillero@polito.it PPSN16 34

34



35

Relevant characteristic

- Lineage (**LIN**)
- Phenotype (**PHE**)
- Genotype (**GEN**)
- ~~Fitness~~ (used as a proxy for either phenotype or genotype)

$$\xi(x, \Psi)$$

squillero@polito.it PPSN16 36

36

Lineage-based methodologies

- The value of $\xi(\circ)$ 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

squillero@polito.it

PPSN16

37

37

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

squillero@polito.it

PPSN16

38

38

Phenotype-based methodologies

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

39


Type of selection

- Parent selection (α or ϵ)
 - Usually non-deterministic
- Survival selection (ω or ω)
 - Usually deterministic

$$\bar{p}_{x|\Psi}$$

40


Diversity Promotion in EA







Information Sciences

Volume 329, 1 February 2016, Pages 782–799


Special issue on Discovery Science



Divergence of character and premature convergence: A survey of methodologies for promoting diversity in evolutionary optimization ☆

Giovanni Squillero  , Alberto Tonda  

DOI: [10.1016/j.ins.2015.09.056](https://doi.org/10.1016/j.ins.2015.09.056)




squillero@polito.it
PPSN16
41

41

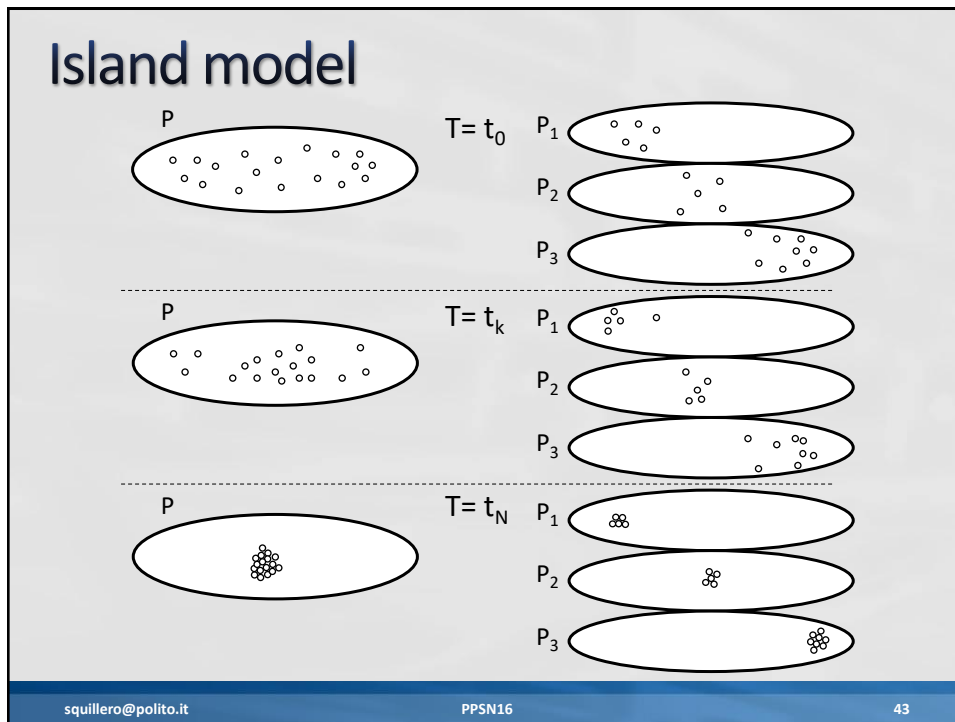
Island model

- Recipe [**LIN** $\alpha\omega$]
 - The population is partitioned into sub-populations
 - Only local interactions are allowed
 - Periodically, individuals are moved between sub-populations (**migrants**)
- Rationale
 - Different populations may explore different parts of the search space
 - ... but global interactions can be useful



squillero@polito.it
PPSN16
42

42



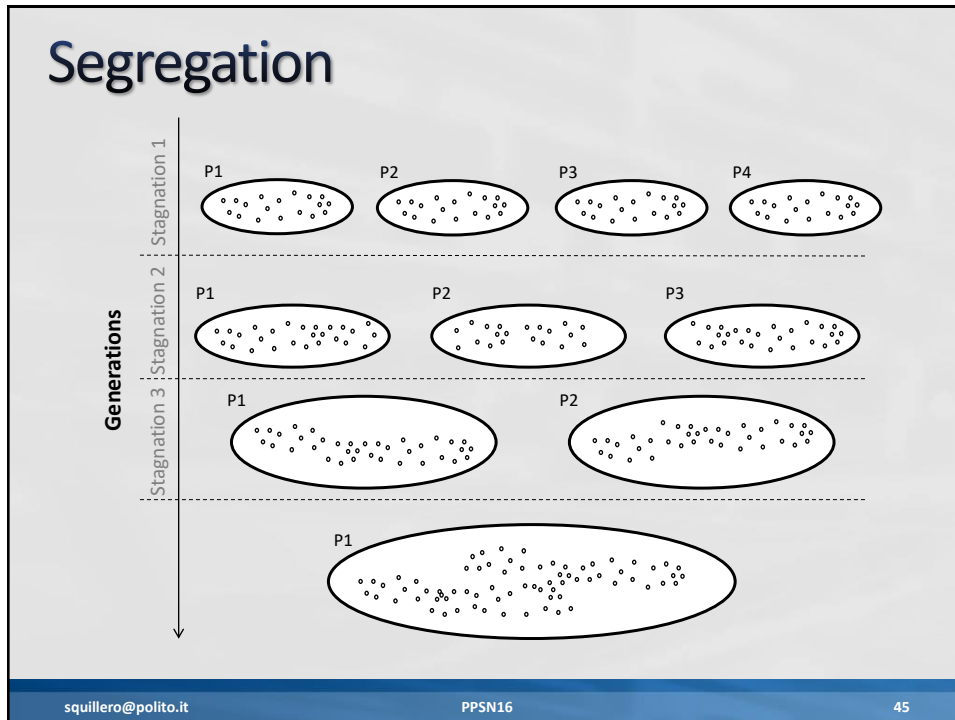
43

Segregation

- Recipe [**LIN** $\alpha\omega$]
 - The population is partitioned into N sub-populations
 - Only local interactions are allowed
 - Upon stagnation, the N sub-populations are merged into $N-1$ sub-populations
- Rationale
 - Same as island models
 - The selective pressure decreases during evolution

squillero@polito.it PPSN16 44

44



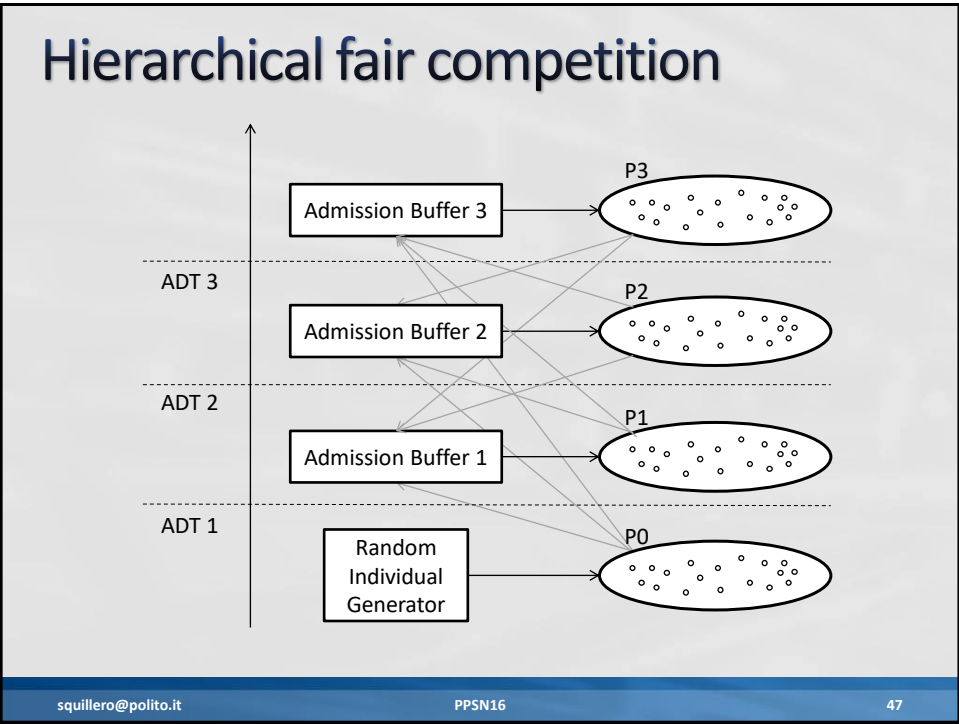
45

Hierarchical fair competition

- Recipe [**PHE** $\alpha\omega$]
 - The population is partitioned into sub-populations with similar fitness
 - Only local interactions are allowed
 - The offspring is promoted or demoted according to fitness
 - New random individuals are constantly generated
- Rationale
 - Hard niching with implicit neighborhood
 - Reduce competition between newborns and already optimized individuals (ladder)

squillero@polito.it PPSN16 46

46



47

Cellular EA

- Recipe [**LIN** $\alpha\omega$]
 - Fixed topology (lattice)
 - Only interactions between neighbors are allowed
- Rationale
 - Limiting interaction could defer the takeover of the population by clones of the fittest individual

Linear-5 (L5)

Compact-9 (C9)

squillero@polito.it PPSN16 48

48

Deterministic crowding

- Recipe [LIN ~~ew~~]
 - Offspring compete against parents for survival
- Rationale
 - Flexible niching with implicit neighborhood
 - Parents and offspring occupy the same niche
 - No need for evaluating the similarity

49

Allopatric selection

- Recipe [LIN ~~ew~~]
 - The whole offspring compete for survival
- Rationale
 - Flexible 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

50

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

- Flexible 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

- Flexible niching with explicit neighborhood
- Set a hard limit to population density

Standard crowding

- Recipe [**GEN** ~~α~~ ω]
 - New individuals replace the most similar individual in a random niche of size CF
- Rationale
 - Flexible niching with explicit neighborhood
 - Favor novelty (generational approach)

squillero@polito.it

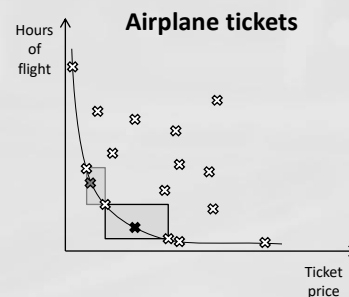
PPSN16

53

53

Crowded-comparison operator

- Recipe [**PHE** ~~α~~ ω]
 - 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)



squillero@polito.it

PPSN16

54

54

Reference points partitioning

• Recipe [GEN ~~aw~~]

- 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

- Flexible niching with explicit neighborhood

Vector evaluated genetic algorithm

• Recipe [PHE ~~aw~~]

- 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

Lexicase selection

- Recipe [**PHE** ~~aw~~]
 - 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

squillero@polito.it

PPSN16

57

57

Restricted tournament selection

- Recipe [**GEN** ~~aw~~]
 - New individuals compete with the most similar individual in a random niche of size CF
- Rationale
 - Flexible niching with explicit neighborhood

squillero@polito.it

PPSN16

58

58

Sequential niching

- Recipe [~~GEN~~ $\alpha\omega$]
 - 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

59

Gender

- Recipe [~~LIN~~/~~GEN~~ $\alpha\omega$]
 - 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

60

Tarpeian method

- Recipe [**PHE** $\alpha\omega$]
 - 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** $\alpha\omega$]
 - 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

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

squillero@polito.it

PPSN16

63

63

Extinction

- Recipe [**PHE** αω]
 - 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

squillero@polito.it

PPSN16

64

64

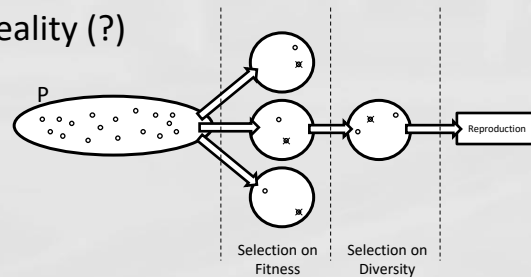
Two-level diversity selection

- Recipe [**GEN** $\alpha\omega$]

- Select three individuals using fitness, then pick the two with maximum distance for reproduction

- Rationale

- Exploit a reliable distance metric to increase the efficacy of crossover
- Not so far from reality (?)



squillero@polito.it

PPSN16

65

65

GDEM – Genetic Diversity Evaluation Method

- Recipe [**GEN** $\alpha\omega$]

- 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)

squillero@polito.it

PPSN16

66

66

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

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



squillero@polito.it

PPSN16

69

69

Hints and Tips

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



squillero@polito.it

PPSN16

70

70

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)



squillero@polito.it

PPSN16

71

71

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



squillero@polito.it

PPSN16

72

72

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)
 - Tweak selection operator(s)
 - Only useful if a global (and efficient) diversity measure is available



squillero@polito.it

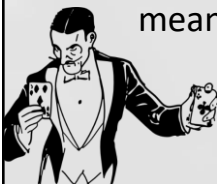
PPSN16

73

73

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



squillero@polito.it

PPSN16

74

74

Outline

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

75

Conclusion



76

More Materials & Bibliography

- **MPDEA**: GECCO Workshop on Measuring and Promoting Diversity in Evolutionary Algorithms
- mpdea@polito.it
- <https://github.com/squillero/mpdea>



squillero@polito.it

PPSN16



77

77