



Universidad de Granada



# Diversidad en AG

**Manuel Lozano**

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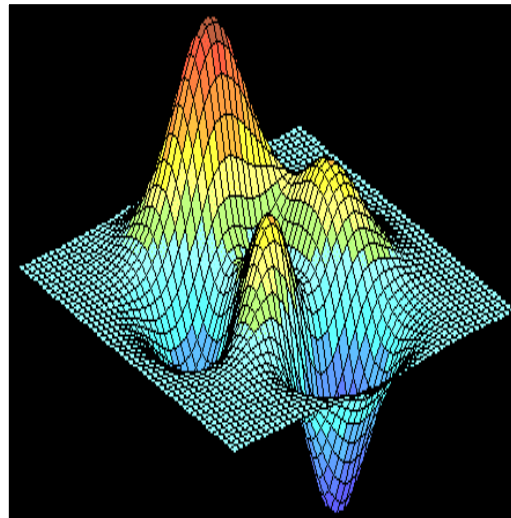
*Técnicas de Soft Computing  
para  
Aprendizaje y Optimización*



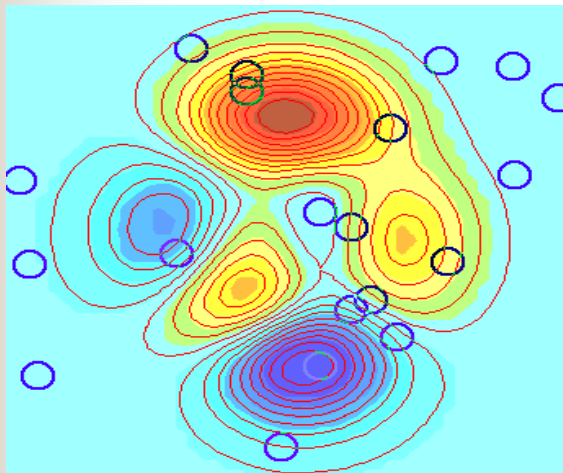
**Departamento de Ciencias de la  
Computación e Inteligencia Artificial**

# Diversity versus convergence

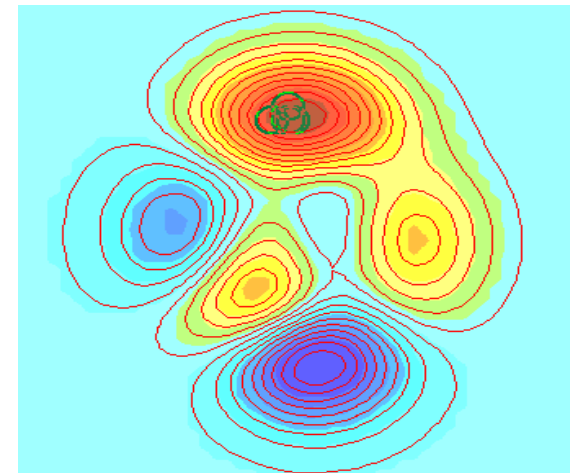
***Diversity***



***Convergence***



***Conflict***





# Diversity versus convergence

## **Exploration and Exploitation in Evolutionary Algorithms: A Survey**

MATEJ ČREPINŠEK, University of Maribor  
SHIH-HSI LIU, California State University, Fresno  
MARJAN MERNIK, University of Maribor

“Exploration and exploitation are the two cornerstones of problem solving by search.” For more than a decade, Eiben and Schippers’ advocacy for balancing between these two antagonistic cornerstones still greatly influences the research directions of evolutionary algorithms (EAs) [1998]. This article revisits nearly 100 existing works and surveys how such works have answered the advocacy. The article introduces a fresh treatment that classifies and discusses existing work within three rational aspects: (1) what and how EA components contribute to exploration and exploitation; (2) when and how exploration and exploitation are controlled; and (3) how balance between exploration and exploitation is achieved. With a more comprehensive and systematic understanding of exploration and exploitation, more research in this direction may be motivated and refined.

ACM Computing Surveys, Vol. 45, No. 3, Article 35, Publication date: June 2013.



# Diversity and convergence

## ➤ ***Mechanisms for Convergence***

### ➤ ***Competition***

- ✓ *Tournament Parent Selection*
- ✓ *Replace-Worst Replacement Mechanism*

## ➤ ***Mechanisms for Diversity***

### ➤ ***Producing Diversity***

- ✓ *Incest Prohibition*
- ✓ *HUX Crossover*
- ✓ *Restart Operator*

### ➤ ***Maintaining Diversity***

- ✓ *Crowding Methods*
- ✓ *Distributed GAs*
- ✓ *Multiploid Representations*





# Incest prohibition

- **MAIN IDEA:** *The application of the crossover to similar parents generates offspring that do not offer diversity*
- **Incest prohibition (Eshelman et al., 1991)**
  - $P_1$  and  $P_2$  are mated only if  $D_H(P_1, P_2)$  is above a *threshold*
  - The threshold *decreases* as evolution proceeds



# HUX crossover

- ***MAIN IDEA:*** Diversity may be introduced by generating offspring very different from their parents
- ***HUX (Eshelman, 1991)***
  - HUX flips exactly *half of the different* bits between two parents
  - Children have always the *maximum Hamming distance* from their *two* parents



# Restart operator

- ***MAIN IDEA:*** *To renew the population of GAs that have converged prematurely*
- ***Restart operator (Goldberg, 1989)***
  - *It restarts the search with a new population*
    - ✓ *New individuals generated randomly*
    - ✓ *Best individuals from the previous population*



# Diversity and convergence

## ➤ ***Mechanisms for Convergence***

### ➤ ***Competition***

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## ➤ ***Mechanisms for Diversity***

### ➤ ***Producing Diversity***

- ✓ *Incest Prohibition*
- ✓ *HUX Crossover*
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- ✓ *Crowding Methods*
- ✓ *Distributed GAs*
- ✓ *Multiploid Representations*



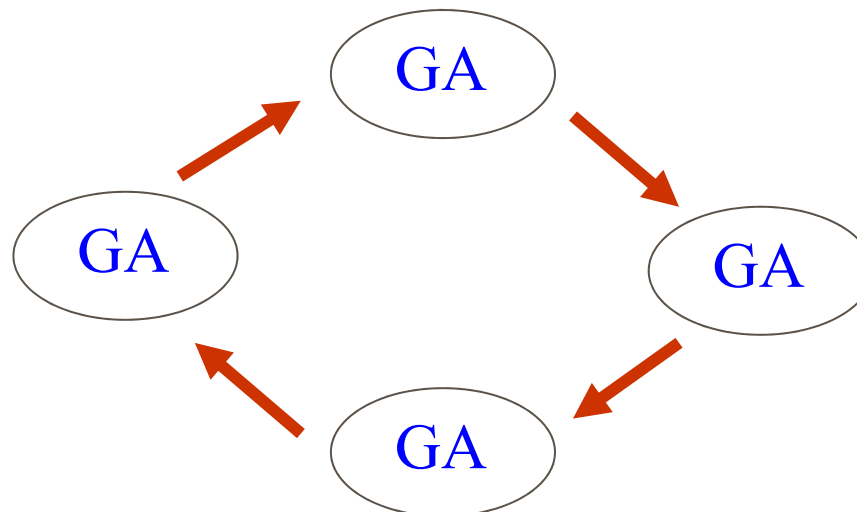


# Crowding methods

- **MAIN IDEA:** Diversity may be sustained by producing little changes with the introduction of new individuals
- **Crowding (De Jong, 1975)**
  - ✓ New individuals *replace* individuals that are *similar* to themselves
  - ✓ **Restricted tournament selection:** To replace the closest individual  $R$  to the one being inserted in the population,  $I$ , from a set of  $n_T$  randomly selected ones, if  $I$  is better than  $R$ .

# Distributed GAs

- **MAIN IDEA:** Diversity may be preserved by isolating individuals in different subpopulations (spatial separation)
- **Distributed GAs (Tanese, 87)**
  - ✓ Several subpopulations are processed by *independent* GAs
  - ✓ A migration operator produces a chromosome exchange between them





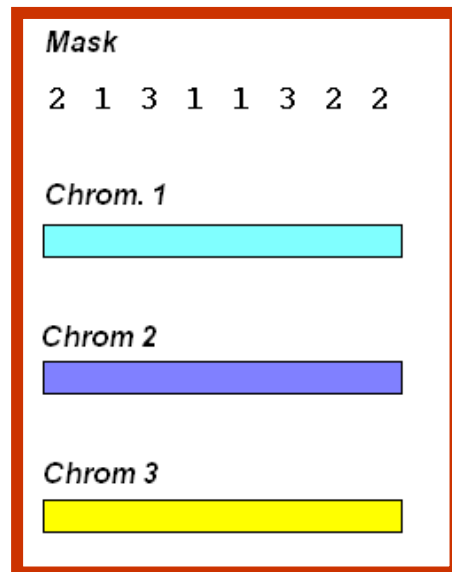
# Multiploid chromosomes

- **MAIN IDEA:** *In nature, many organisms have multiploid genotypes*
  - *Multiple chromosomes (genotype)*
  - *Mechanism for determining the phenotype*
    - ✓ *It determines which of the chromosomes has the dominant gene at each locus*
- *The use of multiploid genotypes enhances diversity*
  - *Unused genes remain in a multiploid genotype until they may later become useful (“latent diversity”)*

# Multiploid chromosomes

- **Multiploid chromosomes (Collingwood et al., 1996)**
  - A *solution* is represented by  $p$  chromosomes ( $p > 1$ )
  - A *mask* is used for obtaining the corresponding solution

*Genotype*



*Solution (Phenotype)*





# Diversity and convergence

## ➤ ***Mechanisms to Combine Diversity and Convergence***

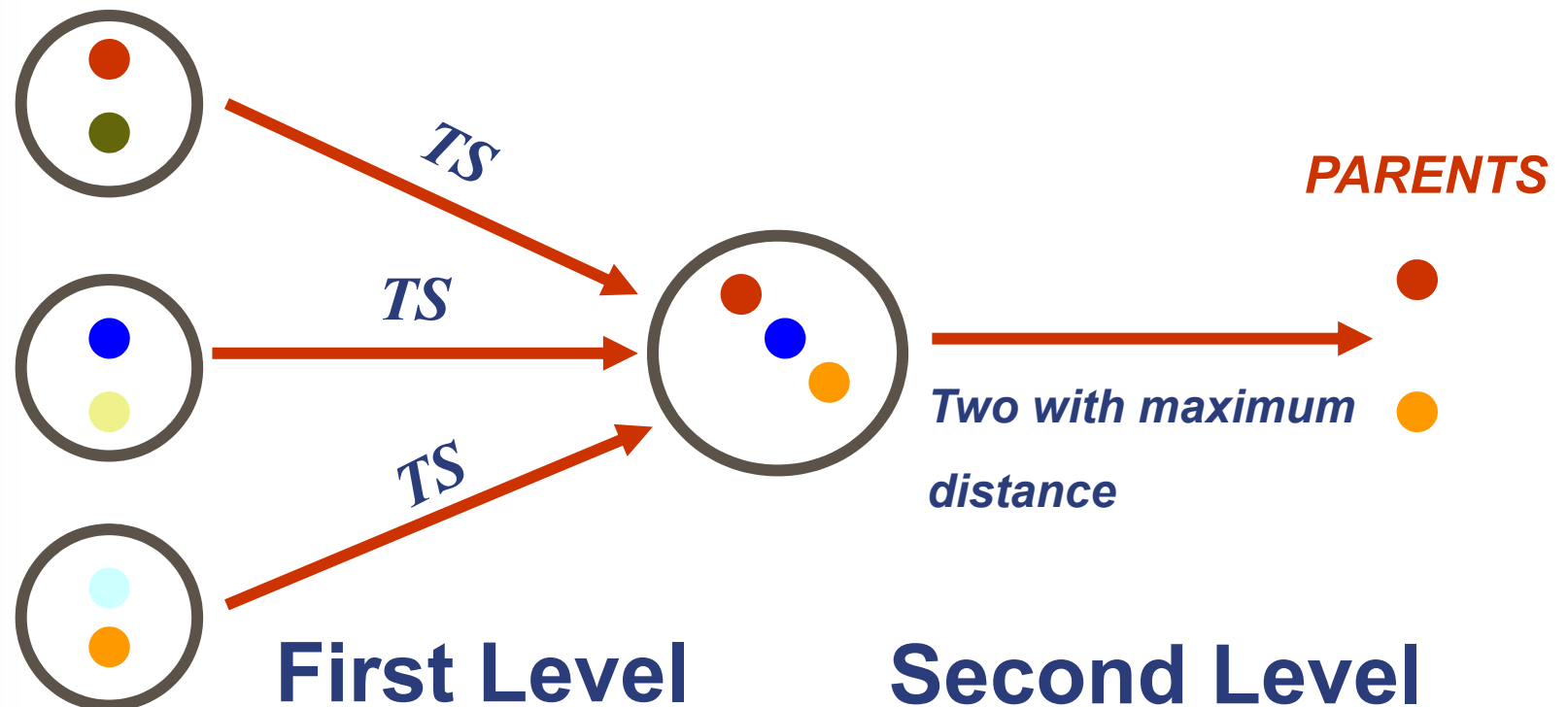
- *Two-level Tournament Parent Selection*

- *Evolutionary Algorithms*

- ✓ **CHC Algorithm**
- ✓ **Micro GAs**
- ✓ **Saw-tooth GA**
- ✓ **Gradual distributed GAs**

# Two-level TS

- **MAIN IDEA:** To select as parents the individuals that are both:
  - ✓ Fit, and
  - ✓ Diverse in relation to others (they provide diversity!)
- **Two-level TS (Brameier, 2002)**



# Favouring fit and diverse chroms.

- Fitness sharing (Goldberg et al., 1987)
- Thermodynamical selection (Mori et al., 1995)
- Disruptive selection (Kuo et al., 1996)
- Diversity control oriented GA (Shimodaira, 1996)
- Multi-objective methods (De Jong et al., 2001)
- Lineage selection (Burke et al., 2003)
- Entropy-boltzmann selection (Lee, 2003)

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Replacement strategies to preserve useful diversity in steady-state genetic algorithms

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# Diversity and convergence

## ➤ ***Mechanisms to Combine Diversity and Convergence***

- *Two-level Tournament Parent Selection*
- *Decreasing Mutation Probability*
- *BLX- $\alpha$  Real-Parameter Crossover*

## ➤ ***Evolutionary Algorithms***

- ✓ ***CHC Algorithm***
- ✓ ***Saw-tooth GA***
- ✓ ***Gradual distributed GAs***
- ✓ ***Memetic algorithms***



# CHC algorithm

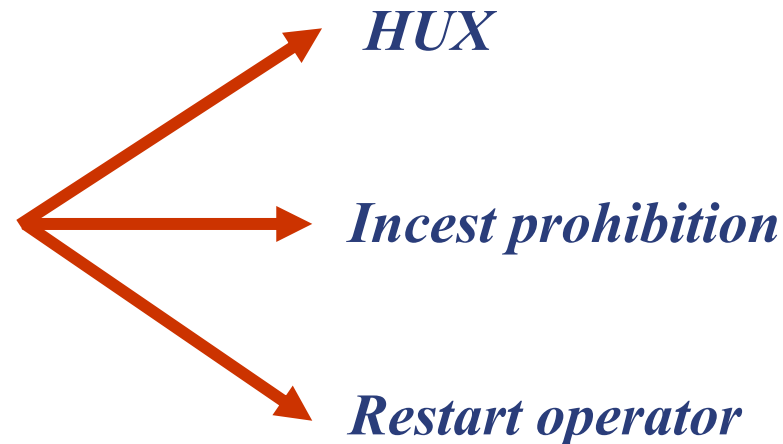
*Idea:*

*High diversity*

+

*High selection  
pressure*

(Eshelman et al., 1991)



→ *Conservative  
selection strategy*  
(It keeps the *N* best elements appearing so far)



# CHC algorithm

## ***References***

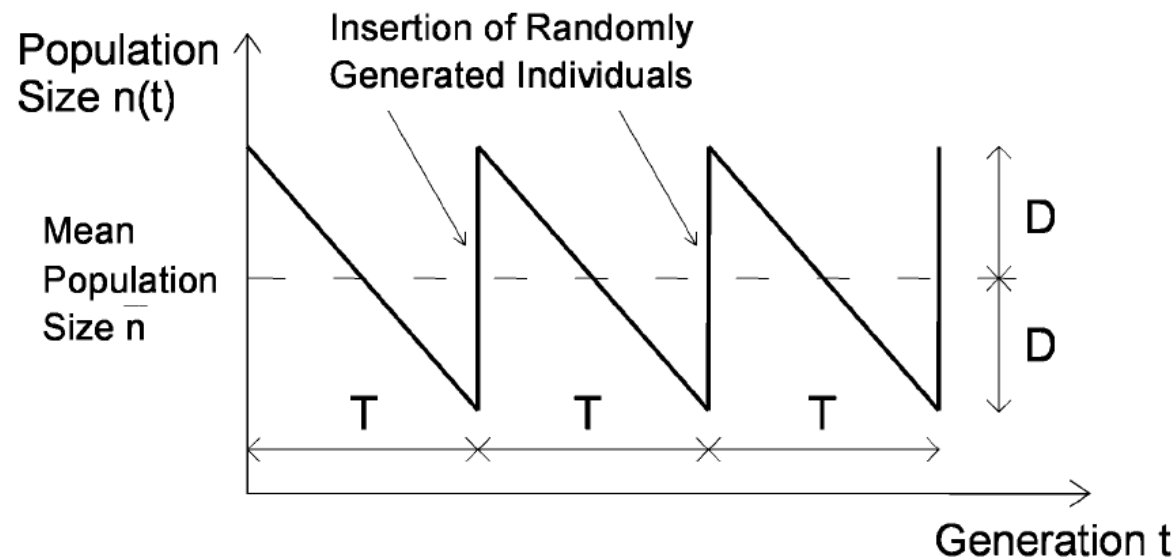
*L.J. Eshelman (1991). The CHC adaptive search algorithm: how to have safe search when engaging in non-traditional genetic recombination. In: Foundations of Genetic Algorithms 1, Morgan Kaufmann, San Mateo, California, pp. 265-283.*

*L.J. Eshelman, K.E. Mathias, J.D. Schaffer (1997). Convergence controlled variation. In: Foundations of Genetic Algorithms 4, Morgan Kaufmann, San Mateo, California, pp 203-224.*

# Saw-Tooth GA

V.K. Koumousis and C.P. Katsaras. IEEE  
TRANSACTIONS ON EVOLUTIONARY  
COMPUTATION, VOL. 10, NO. 1, FEBRUARY  
2006

- It uses variable population size with periodic reinitialization
- In each period, the **population size decreases** linearly
- At the beginning of the next period **randomly generated individuals** are appended to the population

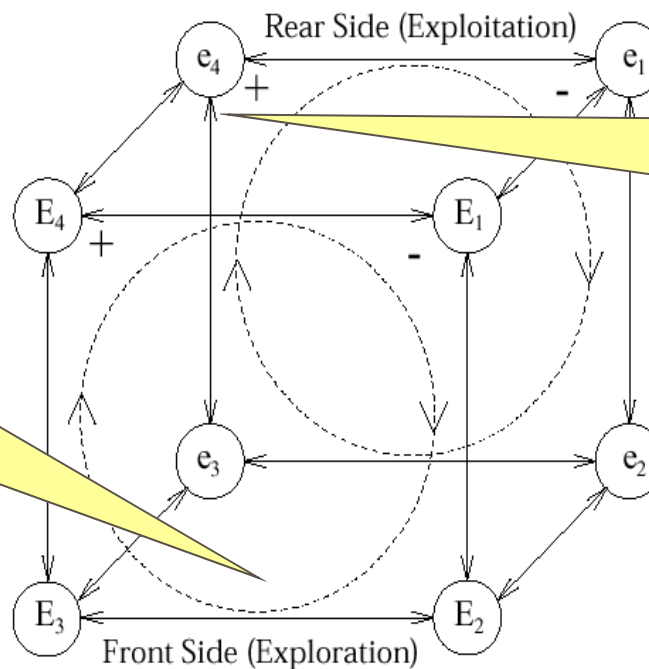


# Gradual distributed GAs

- **MAIN IDEA:** To provide different levels of diversity and convergence in a parallel way
- **Gradual DGAs (Herrera et al., 2000)**
  - ✓ DGAs that apply a *different crossover operator* to each subpopulation

## Front side:

CXs providing diversity at different levels



## Rear side:

CXs introducing convergence at different levels





# Gradual distributed GAs

## ***References***

*F. Herrera, M. Lozano (2000). Gradual distributed real-coded genetic algorithms. IEEE Transactions on Evolutionary Computation 4(1): 43-63.*