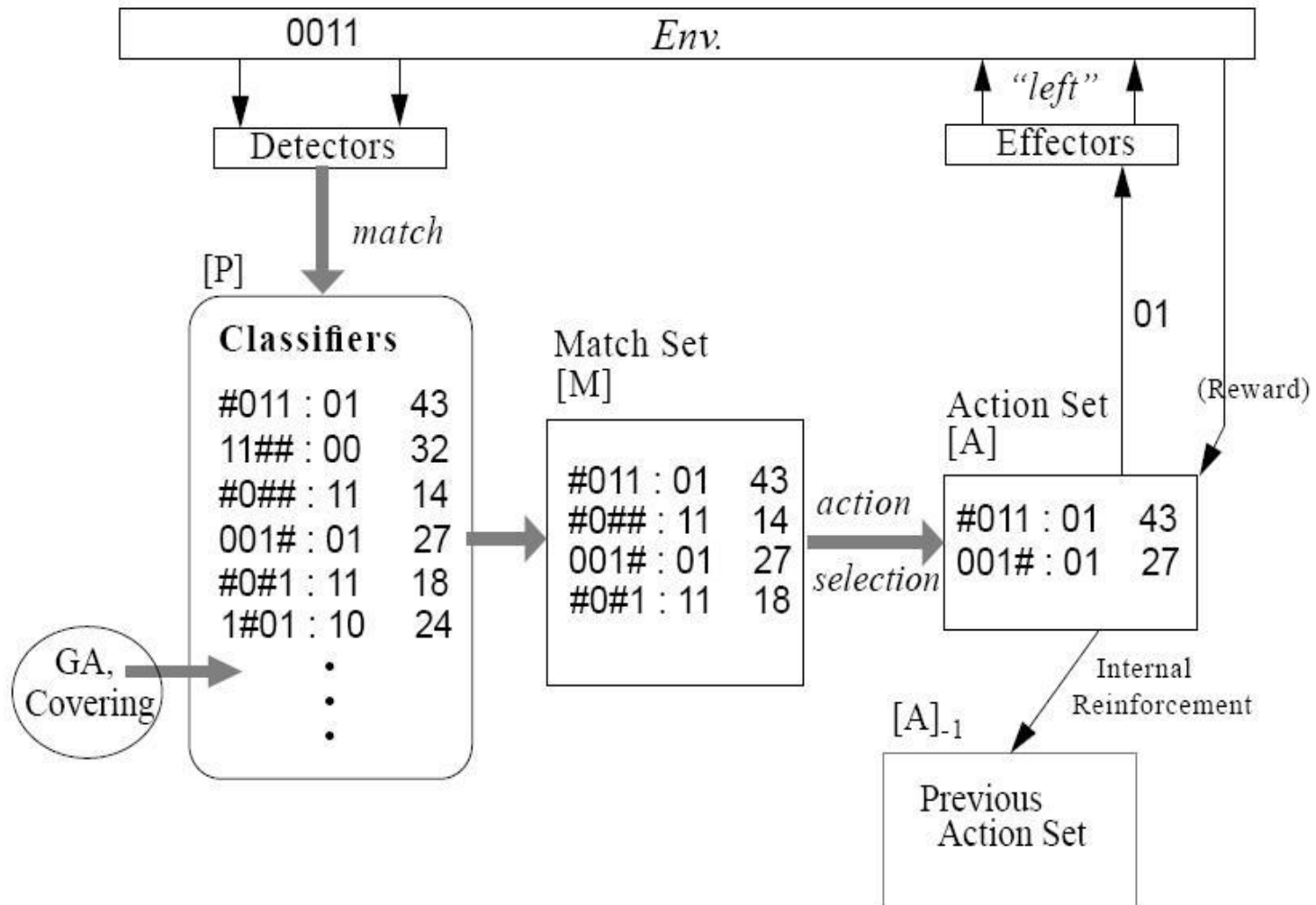


# Organic Computing

Dr. Sebastian von Mammen

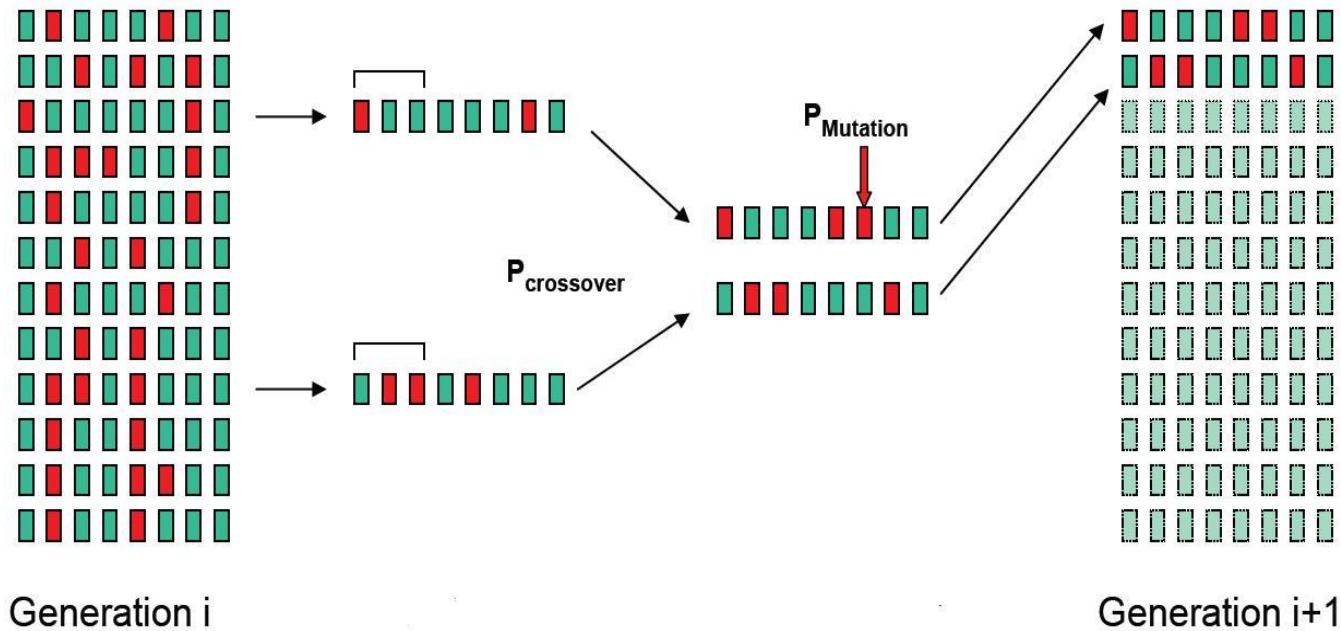
Summer Term 2015

Part 2: Learning



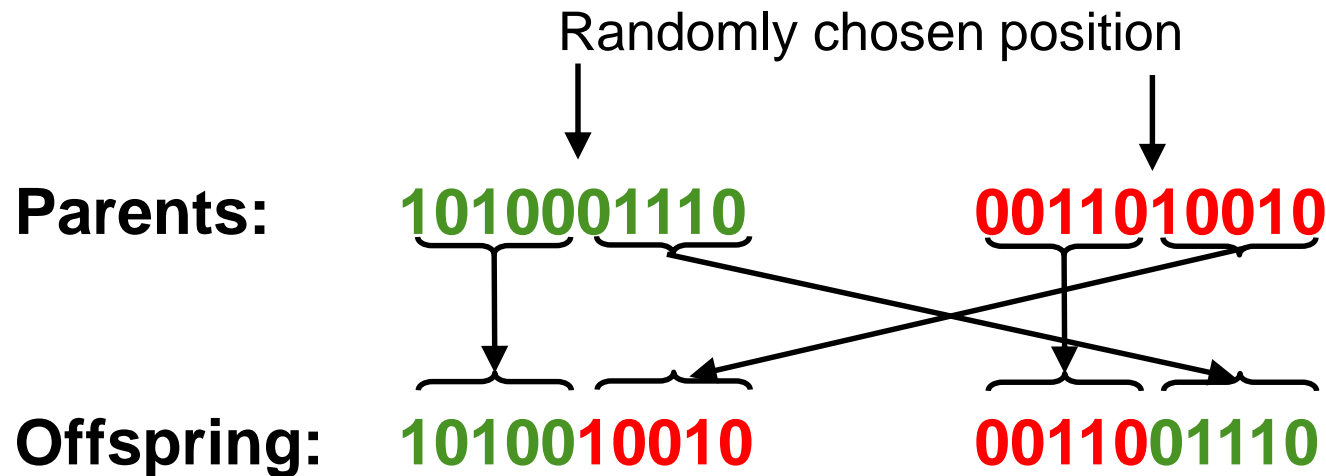
- inspired by Darwinian evolution:
  - “survival of the fittest”
- at each invocation **two classifiers are selected as parents**:  
roulette-wheel selection based on fitness
- with fixed probabilities, these “parents” are subject to “**genetic**” operations
  - crossover (recombination)
  - mutation
- resulting two classifiers get half of their parents’ fitness
- **rate at which GA is invoked** is application dependent
  - too frequent: noisy fitness
  - too seldom: slow development
  - classifiers should have been evaluated a “couple of times” before becoming parents

## Genetic operations: Illustrated example

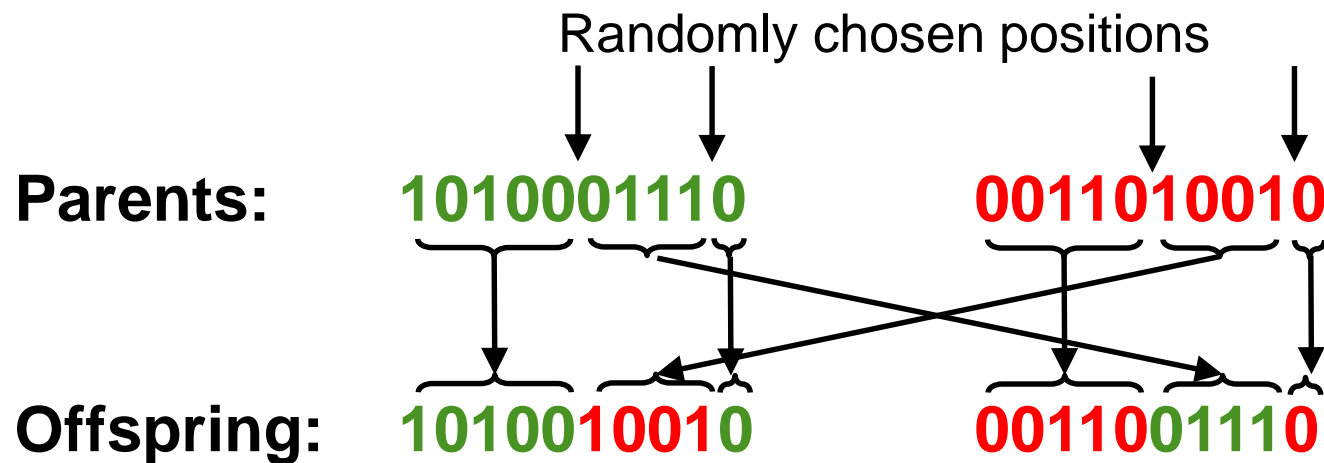


- **one-point crossover** in the example: 3/5 bits
- **multi-point crossover**: more than one “cut point” selected
- **mutation**: randomly flips a bit (typically very small probability)

- one position in the chromosomes is chosen **randomly**
- recombining the parents' genetic material ahead and after this position yields two offspring



- two positions in the chromosomes are chosen **randomly**
- avoids that genes at the head and at the tail are split



- a random mask is generated, e.g. ABAABABBAB
- the mask determines which bits are copied from which parent

**Parents:**      1010001110      0011010010

**Offspring:**    1010000010      0011011110

- *which crossover operator to choose?*  
→ trade-off between **exploration** due to the introduction of new combinations of features and **exploitation** by keeping the good feature combinations of existing solutions

- generate new offspring from single parents
- *alternatively*: mutate recombined offspring

**Parent:**      1010001110



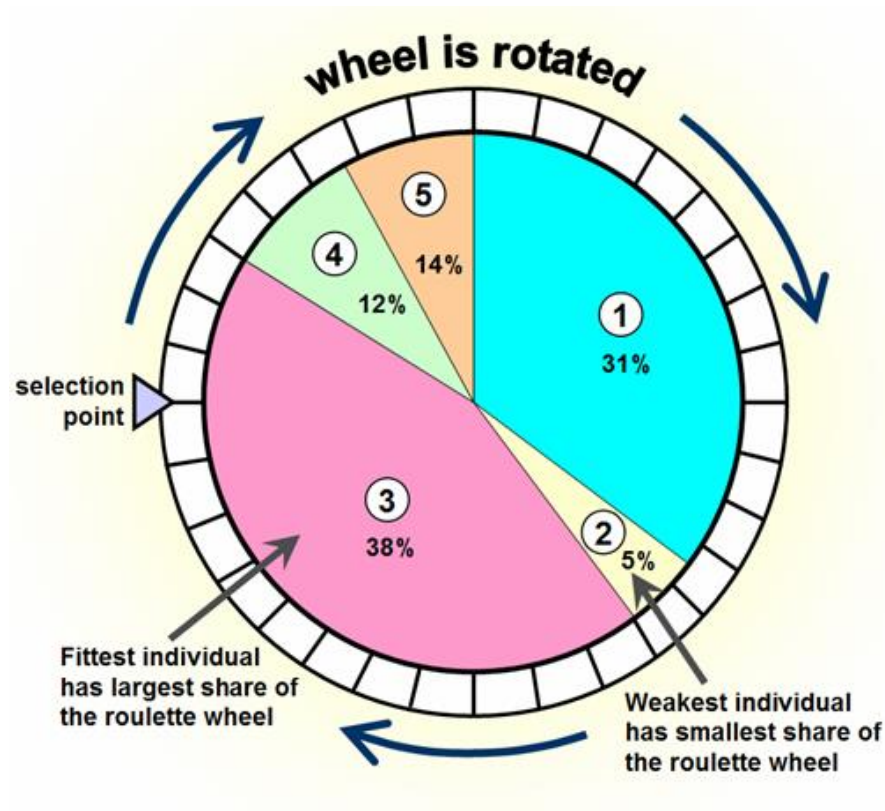
**Offspring:**    1011010010

- ensures diversity of the **gene pool**
  - crossover can only explore combinations of the existing genes
  - mutation can “generate” new genetic information



## Genetic operators: Roulette Wheel Selection

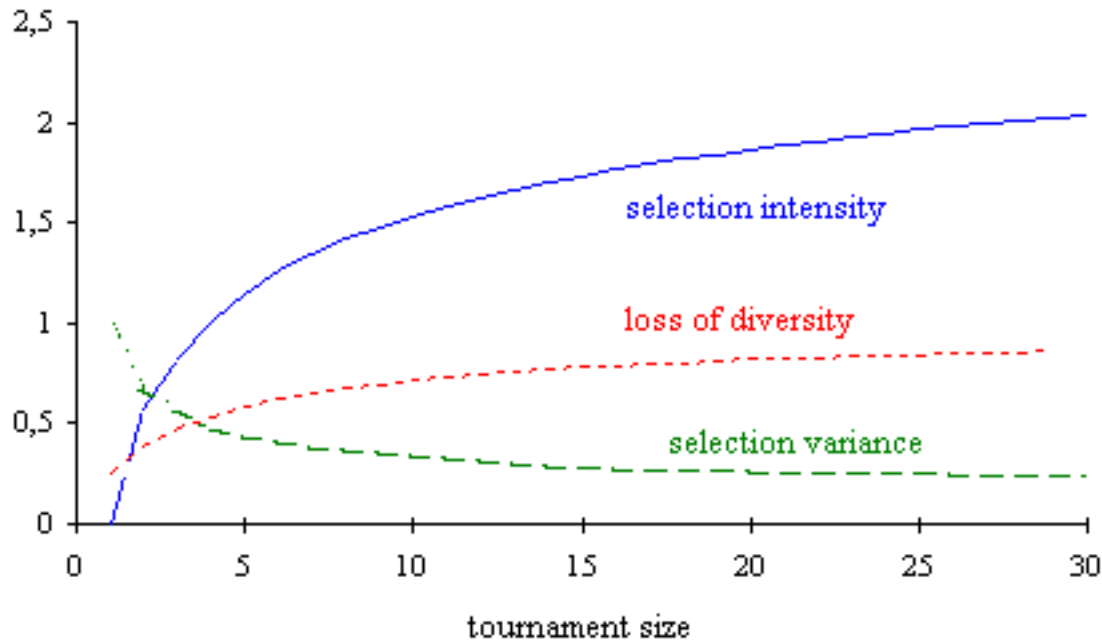
- *aka*: fitness-proportionate selection
- probability of selection is proportional to individuals' fitness values



- rank individuals according to their fitness values
  - each rank has a predefined selection probability
    - linear ranking vs. exponential ranking
- selection based on but not proportional to fitness



- randomly evaluate sets of individuals (2 or more)
- the one with higher fitness is selected as parent
- considers multiple objectives



- *selection intensity or selective pressure:*  
change in average fitness due to selection
- *niche count:*  
the number of points in the population within a certain distance  
→ the higher the niche count, the lower the fitness/rank
- *elitism:*  
automatically transfer the best genotype(s) to the next generation

- upfront, as always: problem-specific!
- greater population size:
  - increase in **diversity**
  - increase in **computation time**
- greater crossover probability:
  - increase the **exploration** of the search space
  - increase the **chance of disruption** of good combinations
- greater mutation probability:
  - helps to **introduce new genes** or reintroduce lost ones
  - closer to **random search**