

# ParadisEO-MO: Fitness Landscape Analysis and efficient local search implementation

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INSTITUT NATIONAL  
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The ParadisEO Team,  
including:

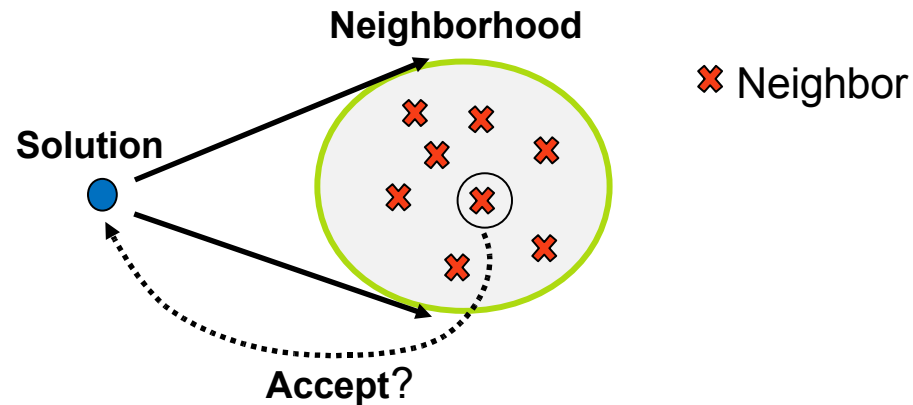
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# Concepts in ParadisEO-MO: “Neighbor”, “Neighborhood” and “Evaluation”

- “Neighbor”
  - Move + save neighbor informations (fitness and more)
- “Neighborhood”
  - Describe how to compute all the neighbors
  - Methods: init, next, cont and hasNeighbor
- “Evaluation”
  - “moEval”: set fitness (and more) of the neighbor
  - Can be incremental or full evaluation



# General schema of LS



## Search explorer:

- Generate neighbors from the neighborhood
- Select a neighbor
- Decide to replace solution by the selected neighbor

# Simplified schema of LS

Constructors:

- moNeighborhood(moNeighbor)
- moSearchExplorer(moNeighborhood, moEval)

```
do {
  searchExplorer(solution)
  if ( searchExplorer.accept(solution) )
    searchExplorer.move(solution)
}
while (searchExplorer.continue(solution))
```



# Completed schema of LS

```
searchExplorer.initParam(solution)
continuator.init(solution)
do {
    searchExplorer(solution)
    if ( searchExplorer.accept(solution) )
        searchExplorer.move(solution)
    searchExplorer.updateParam(solution)
}
while (continuator(solution) && searchExplorer.continue(solution))
searchExplorer.terminate(solution)
```

Statistics can be computed (and reported):

- at initialization: `continuator.init(solution)`
- at each iteration : `continuator(solution)`



# Improvements

- Hill-climbing like:
  - Simple HC (best improvement)
  - Simple HC neutral (random choice among several best solutions) NEW
  - HC neutral (+ allows move when best solution has equal fitness) NEW
  - First Improvement HC
- Walk like to sample search space: NEW
  - Random walk
  - Random neutral walk
  - Metropolis hasting
- Simulated Annealing
  - cooling schudeling: init, decreases, final* NEW
- Tabu Search
  - "moMemory" : diversification, intensification, tabu "list"* NEW



# Classical LS: improvements

- Hill-climbing like:
  - Simple HC (best improvement)
  - Simple HC neutral (random choice among several best solutions)
  - HC neutral (+ allows move when best solution has equal fitness)
  - First Improvement
- Walk like to sample search space:
  - Random walk
  - Random neutral walk
  - Metropolis Hastings
- Simulated Annealing  
*cooling schedule: init, decreases, final*
- Tabu Search

*"noMemory" : diversification, intensification, tabu "list"*

**Modularity improved**  
**Easy reuse of basic LS**

NEW

NEW

NEW

NEW

NEW



# Statistics and checkpointing : Fitness Landscape Analysis

NEW

- Checkpointing like in ParadisEO-EO is available
  - moCheckpoint
  - moContinuator
  - moStatBase
    - Statistics in the Neighborhood : min, max, mean, standard deviation, probability to increase, neutral degree...
    - Autocorrelation of fitness, fitness cloud, FDC...





# Fitness Landscape Analysis

Class to define a sampling:

- `moSampling<Neighbor>` (*paradiseo-mo/src/sampling*)

Constructor:

```
moSampling (eoInit< EOT > &_init,  
            moLocalSearch< Neighbor > &_localSearch,  
            moStat< EOT, ValueType > &_stat,  
            bool _monitoring=true)
```

Method:

```
add(moStat< EOT, ValueType > &_stat, bool _monitoring=true)
```



# Available Statistics

- `moFitnessStat` : the fitness of the current solution
- `moDistanceStat` : the distance between the current solution and a given solution
- `moSolutionStat` : the current solution
- `moCounterStat` : the number of iterations
- `moBestSoFarStat` : the best current solution found
- `moNeighborBestStat` : best fitness over  $k$  neighbors
- `moNeighborhoodStat` : to compute the statistics from the neighbors solutions : .....



# Available Statistics

- `moNeighborhoodStat` : to compute the statistics from the neighbors solutions :
  - `moAverageFitnessNeighborStat` : average fitness in the neighborhood
  - `moStdFitnessNeighborStat` : standard deviation of fitness
  - `moMaxNeighborStat` : maximum fitness
  - `moMinNeighborStat` : minimum fitness
  - `moSecondMomentNeighborStat` : average and standard deviation
  - `moSizeNeighborStat` : size of the neighborhood
  - `moNeutralDegreeNeighborStat` : number of neighbors with equal fitness
  - `moNbInfNeighborStat` : number of neighbors with lower fitness
  - `moNbSupNeighborStat` : number of neighbor with higher fitness

## Pre-defined classes NEW

- `moDensityOfStatesSampling` : density of states
- `moAutocorrelationSampling` : autocorrelation length and functions
- `moFDCsampling` : fitness distance correlation
- `moHillClimberSampling` : adaptive walks
- `moNeutralDegreeSampling` : neutral degree
- `moNeutralWalkSampling` : evolvability of neutral networks
- `moFitnessCloudSampling` : evolvability of the operator, and the neighborhood

(In directory *paradiseo-mo/src/sampling*)



# How to use ParadisEO-MO?

## What I need define?

### Only problem dependent components:

- Solution representation *EOT*
- “Random” initialisation operator *eoInit*
- Evaluation Functions
  - full Evaluation *eoEvalFunc*
  - incremental (if possible) *moEval*
- Neighbor *moNeighbor*
- Neighborhood
  - *moNeighborhood*
  - or a mapping (int -> Neighbor) using an *moIndexNeighborhood* NEW

### Predefined components for the classical search spaces: NEW

- Bit string
- Permutation



## Example of code (from tutorial lesson 6)

```
eoUniformGenerator<bool> uGen;  
eoInitFixedLength<Indi> random(vecSize, uGen);  
  
oneMaxFullEval<Indi> fullEval;  
  
eoHammingDistance<Indi> distance;  
  
Indi bestSolution(vecSize, true); // global optimum  
  
moFDCsampling<Neighbor> sampling(random, fullEval, distance, bestSolution, nbSol);  
  
sampling();  
  
sampling.fileExport(str_out);  
  
const std::vector<double> & fitnessValues = sampling.getValues(0);  
const std::vector<double> & distValues   = sampling.getValues(1);
```



# Conclusion

- More reusable design (“Neighbor”, “Neighborhood”, ...)
- 11 standard local search vs 6 before
- More than 133 classes ( $\approx$  13500 code lines) vs 40 before
- All classes are tested and documentation available
- fitness landscape analysis is now available
- Standard components added (Bit string, permutation and so...)
- ...A lot of <sup>NEW</sup>



# Perspectives

- Create tool boxes for automatics fitness landscape analysis
- Extend to multiobjective optimization: set-based local search
- Create problems repository (done in fact)
- Create tools for automatics parameters tunings and parameters control
- Connection with GPGPU

