

VU Machine Learning WS 2020

Exercise 3.3 Automated Machine Learning Nysret Musliu

This is one of possible topics for exercise 3. See other possible topics in tuwel. You have to select only one topic for exercise 3

- Automated Machine Learning
 - Implementation of a metaheuristic algorithm for automated selection/configuration of machine learning algorithms
 - Comparison with other state of the art approaches
 - Group work (like in the first two assignments)
 - Presentations: after the submission

- Implement a metaheuristic algorithm (local search, genetic algorithms...) that searches for the best machine learning technique (and best hyperparameters) for a particular classification/regression data set
- *Simulated annealing and tabu search can be used, but they should be combined with other algorithms (e.g., iterated local search)*

- Search space:
 - At least five available machine learning algorithms
 - Most important hyperparameters that should be tuned for each of these algorithms. You can specify for each hyperparameter a reasonable range of possible values
 - The aim is to find a solution (the best algorithm/hyperparameters) in the search space that optimizes an evaluation score (e.g., classification accuracy or RMSE)
- Please write me an email if you have any questions

Comparison with other approaches

- Compare your approach with two state of the art AutoML systems (e.g. auto-sklearn, TPOT...)
- Use for comparison four classification or regression data sets (you can also use the data sets from the previous assignments)
- Time limit: you should use at least 1h per data set

- Your implementation
- More than 20 slides with this structure
 - Main information for your implementation: representation of solution, neighborhoods (or crossover operators), evaluation function, parameters used for implemented technique...
 - Selected state of the art AutoML systems for comparison
 - Discussion of results
- No report needed for this assignment



Presentations/Discussion of assignment

- Discussion of code
- Implementation issues
- Discussion of results and your findings

Iterated Local Search (ILS)

http://www.econ.upf.edu/~ramalhin/PDFfiles/2001_MIC_FILS.pdf

As a local search algorithms technique within ILS you should either use **simulated annealing or tabu search (see appendix)**

Genetic algorithms

See appendix or chapters 3 in <http://www.cs.vu.nl/~gusz/ecbook/ecbook-course.html>

or Memetic algorithms (combination of genetic algorithms and local search)

Implementation of other algorithms is also possible

- Please write me an email if you are interested to implement some other algorithm: nysret.musliu@tuwien.ac.at

Appendix A

A short introduction to local search
techniques/genetic algorithms

Definition of search problem

- Given a search space S together with its feasible part

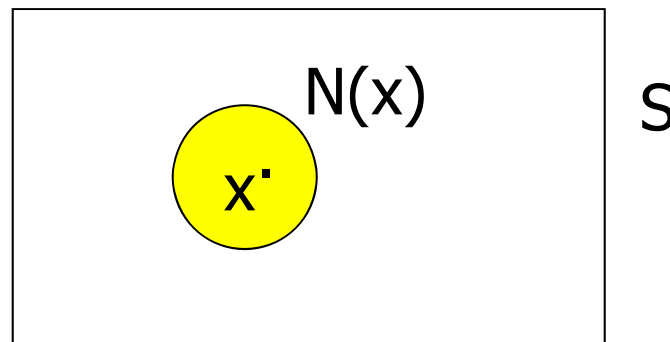
$F \subseteq S$, find $x \in F$ such that

$$eval(x) \leq eval(y) \quad \text{for all } y \in F$$

- x that satisfies the above condition is called global optimum (for minimization problem)

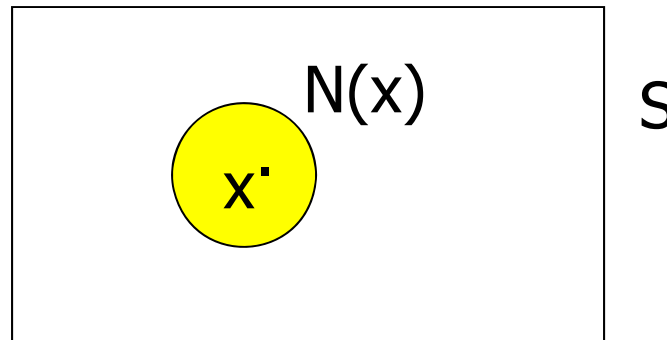
Neighbourhood and local optima

- Region of the search space that is near particular point in the space



- A potential solution $x \in F$ is a local optimum with respect to the neighborhood N , if and only if
 $eval(x) \leq eval(y)$,
for all $y \in N(x)$

- Are based on the neighbourhood of the current solution



- The solution is changed iteratively with so called neighbourhood relations (moves) until an acceptable or optimal solution is reached

1. Pick a random point in the search space
2. Consider all the neighbours of the current state
3. Choose the neighbour with the best quality and move to that state
4. Repeat 2 through 4 until all the neighbouring states are of lower quality
5. Return the current state as the solution state

- Is based on the analogy from the thermodynamics
- To grow a crystal, the raw material is heated to a molten state
- The temperature of the crystal melt is reduced until the crystal structure is frozen in
- Cooling should not be done too quickly, otherwise some irregularities are locked in the crystal structure

Simulated Annealing

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Procedure simulated annealing

begin

$t=0$

Initialize T

select a current solution v_c at random

evaluate v_c

repeat

repeat

select a new solution v_n in the neighborhood of v_c

if $eval(v_c) < eval(v_n)$ **then** $v_c = v_n$

else if $random[0,1) < e^{\frac{eval(v_n) - eval(v_c)}{T}}$ **then** $v_c = v_n$

until (termination-condition)

$T = g(T, t)$

$t = t + 1$

until (halting-criterion)

end

SA – problem specific questions

- What is a solution?
- What are the neighbors of a solution?
- What is a cost of a solution
- How do we determine the initial solution

- How do we determine the initial “temperature” T ?
- How do we determine the cooling rate $g(T,t)$?
- How do we determine the termination condition?
- How do we determine the halting criterion?

- STEP 1: $T = T_{max}$
select v_c at random
- STEP 2: pick a point v_n from the neighborhood of v_c

 if $eval(v_n)$ is better than the $val(v_c)$
 then select it ($v_c = v_n$)
 else select it with probability $e^{\frac{-\Delta eval}{T}}$
 repeat this step k_T times
- STEP 3: set $T = rT$
 if $T \geq T_{min}$
 then goto STEP 2
 else goto STEP 1

.....

Procedure Tabu-Suche

begin

Initialize tabu list

Generate randomly Initial Solution s_c

Evaluate s_c

repeat

 Generate all neighborhood solutions of the solution s_c

 Find best solution s_x in the neighborhood

 if s_x is not tabu solution then $s_c = s_x$

 else if 'aspiration criteria' is fulfilled then

$s_c = s_x$

 else

 find best not tabu solution in the neighborhood s_{nt}

$s_c = s_{nt}$

 Update tabu list

until (terminate-condition)

end

See

[https://dbai.tuwien.ac.at/staff/musliu/ProblemSolvingAI/Cla
ss7TabuSearch.pdf](https://dbai.tuwien.ac.at/staff/musliu/ProblemSolvingAI/Cla
ss7TabuSearch.pdf)

Simple Genetic Algorithm

{

initialize population;

evaluate population;

while TerminationCriteriaNotSatisfied

{

select parents for reproduction;

perform recombination and mutation;

evaluate population;

}

}

Genetic Algorithms: A Tutorial, Wendy Williams:

https://www.csd.uwo.ca/~mmorenom/cs2101a_moreno/Class9GATutorial.pdf