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Local search (optimization)

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In [computer science](#), **local search** is a [metaheuristic](#) method for solving computationally hard [optimization](#) problems. Local search can be used on problems that can be formulated as finding a solution maximizing a criterion among a number of [candidate solutions](#). Local search algorithms move from solution to solution in the space of candidate solutions (the *search space*) by applying local changes, until a solution deemed optimal is found or a time bound is elapsed.

Local search algorithms are widely applied to numerous hard computational problems, including problems from [computer science](#) (particularly [artificial intelligence](#)), [mathematics](#), [operations research](#), [engineering](#), and [bioinformatics](#). Examples of local search algorithms are [WalkSAT](#) and the [2-opt algorithm for the Traveling Salesman Problem](#).

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Examples [\[edit\]](#)

Some problems where local search has been applied are:

1. The [vertex cover problem](#), in which a solution is a [vertex cover](#) of a [graph](#), and the target is to find a solution with a minimal number of nodes
2. The [travelling salesman problem](#), in which a solution is a [cycle](#) containing all nodes of the graph and the target is to minimize the total length of the cycle
3. The [boolean satisfiability problem](#), in which a candidate solution is a truth assignment, and the target is to maximize the number of [clauses](#) satisfied by the assignment; in this case, the final solution is of use only if it satisfies *all* clauses
4. The [nurse scheduling problem](#) where a solution is an assignment of nurses to [shifts](#) which satisfies all established [constraints](#)
5. The [k-medoid](#) clustering problem and other related [facility location](#) problems for which local search offers the best known approximation ratios from a worst-case perspective

Description [\[edit\]](#)

Most problems can be formulated in terms of search space and target in several different manners. For example, for the travelling salesman problem a solution can be a cycle and the criterion to maximize is a combination of the number of nodes and the length of the cycle. But a solution can also be a path, and being a cycle is part of the target.

A local search algorithm starts from a candidate solution and then [iteratively](#) moves to a [neighbor](#) solution. This is only possible if a [neighborhood relation](#) is defined on the search space. As an example, the neighborhood of a vertex cover is another vertex cover only differing by one node. For boolean satisfiability, the neighbors of a truth assignment are usually the truth assignments only differing from it by the evaluation of a variable. The same problem may have multiple different neighborhoods defined on it; local optimization with neighborhoods that involve changing up to *k* components of the solution is often referred to as **k-opt**.

Typically, every candidate solution has more than one neighbor solution; the choice of which one to move to is taken using only information about the solutions in the [neighborhood](#) of the current one, hence the name *local* search. When the choice of the neighbor solution is done by taking the one locally maximizing the criterion, the

metaheuristic takes the name [hill climbing](#). When no improving configurations are present in the neighborhood, local search is stuck at a [locally optimal point](#). This local-optima problem can be cured by using restarts (repeated local search with different initial conditions), or more complex schemes based on iterations, like [iterated local search](#), on memory, like [reactive search optimization](#), on memory-less stochastic modifications, like [simulated annealing](#).

Termination of local search can be based on a time bound. Another common choice is to terminate when the best solution found by the algorithm has not been improved in a given number of steps. Local search is an [anytime algorithm](#): it can return a valid solution even if it's interrupted at any time before it ends. Local search algorithms are typically [approximation](#) or [incomplete algorithms](#), as the search may stop even if the best solution found by the algorithm is not optimal. This can happen even if termination is due to the impossibility of improving the solution, as the optimal solution can lie far from the neighborhood of the solutions crossed by the algorithms.

For specific problems it is possible to devise neighborhoods which are very large, possibly exponentially sized. If the best solution within the neighborhood can be found efficiently, such algorithms are referred to as [very large-scale neighborhood search](#) algorithms.

See also [\[edit\]](#)

Local search is a sub-field of:

- [Metaheuristics](#)
- [Stochastic optimization](#)
- [Optimization](#)

Fields within local search include:

- [Hill climbing](#)
- [Simulated annealing](#) (suited for either local or global search)
- [Tabu search](#)
- [Reactive search optimization](#) (combining [machine learning](#) and local search heuristics)

Real-valued search-spaces [\[edit\]](#)

Several methods exist for performing local search of [real-valued](#) search-spaces:

- [Luus–Jaakola](#) searches locally using a [uniform distribution](#) and an exponentially decreasing search-range.
- [Random optimization](#) searches locally using a [normal distribution](#).
- [Random search](#) searches locally by sampling a [hypersphere](#) surrounding the current position.
- [Pattern search](#) takes steps along the axes of the search-space using exponentially decreasing step sizes.

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v · t · e	Major subfields of optimization	[show]
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