

# Foundations of Artificial Intelligence

## 20. Combinatorial Optimization: Introduction and Hill-Climbing

Malte Helmert

University of Basel

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## 20.1 Combinatorial Optimization

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# 20.1 Combinatorial Optimization

## Introduction

previous chapters: **classical state-space search**

- ▶ find **action sequence** (path) from initial to goal state
- ▶ difficulty: large number of states (“**state explosion**”)

next chapters: **combinatorial optimization**

↔ similar scenario, but:

- ▶ no actions or transitions
- ▶ don't search for path, but for **configuration** (“state”) with low cost/high quality

**German:** Zustandsraumexplosion, kombinatorische Optimierung, Konfiguration

## Combinatorial Optimization: Overview

### Chapter overview: combinatorial optimization

- ▶ 20. Introduction and Hill-Climbing
- ▶ 21. Advanced Techniques

## Combinatorial Optimization Problems

### Definition (combinatorial optimization problem)

A **combinatorial optimization problem** (COP) is given by a tuple  $\langle C, S, opt, v \rangle$  consisting of:

- ▶ a set of (solution) **candidates**  $C$
- ▶ a set of **solutions**  $S \subseteq C$
- ▶ an **objective sense**  $opt \in \{\min, \max\}$
- ▶ an **objective function**  $v : S \rightarrow \mathbb{R}$

**German:** kombinatorisches Optimierungsproblem, Kandidaten, Lösungen, Optimierungsrichtung, Zielfunktion

### Remarks:

- ▶ “problem” here in another sense (= “instance”) than commonly used in computer science
- ▶ practically interesting COPs usually have too many candidates to enumerate explicitly

## Optimal Solutions

### Definition (optimal)

Let  $\mathcal{O} = \langle C, S, opt, v \rangle$  be a COP.

The **optimal solution quality**  $v^*$  of  $\mathcal{O}$  is defined as

$$v^* = \begin{cases} \min_{c \in S} v(c) & \text{if } opt = \min \\ \max_{c \in S} v(c) & \text{if } opt = \max \end{cases}$$

( $v^*$  is undefined if  $S = \emptyset$ .)

A solution  $s$  of  $\mathcal{O}$  is called **optimal** if  $v(s) = v^*$ .

**German:** optimale Lösungsqualität, optimal

## Combinatorial Optimization

The basic algorithmic problem we want to solve:

### Combinatorial Optimization

Find a **solution** of good (ideally, optimal) quality for a combinatorial optimization problem  $\mathcal{O}$  or prove that no solution exists.

**Good** here means **close to  $v^*$**  (the closer, the better).

## Relevance and Hardness

- ▶ There is a huge number of practically important combinatorial optimization problems.
  - ▶ Solving these is a central focus of **operations research**.
  - ▶ Many important combinatorial optimization problems are **NP-complete**.
  - ▶ Most “classical” NP-complete problems can be formulated as combinatorial optimization problems.
- ↪ **Examples:** TSP, VERTEXCOVER, CLIQUE, BINPACKING, PARTITION

German: Unternehmensforschung, NP-vollständig

## Search vs. Optimization

Combinatorial optimization problems have

- ▶ a **search aspect** (among all candidates  $C$ , find a solution from the set  $S$ ) and
- ▶ an **optimization aspect** (among all solutions in  $S$ , find one of high quality).

## Pure Search/Optimization Problems

Important special cases arise when one of the two aspects is trivial:

- ▶ **pure search problems:**
  - ▶ all solutions are of equal quality
  - ▶ difficulty is in finding a solution **at all**
  - ▶ **formally:**  $v$  is a constant function (e.g., constant 0);  $opt$  can be chosen arbitrarily (does not matter)
- ▶ **pure optimization problems:**
  - ▶ all candidates are solutions
  - ▶ difficulty is in finding solutions of **high quality**
  - ▶ **formally:**  $S = C$

## 20.2 Example

## Example: 8 Queens Problem

### 8 Queens Problem

How can we

- ▶ place **8 queens** on a chess board
- ▶ such that **no two queens threaten each other?**

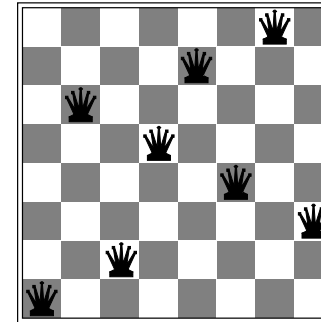
German: 8-Damen-Problem

- ▶ originally proposed in 1848
- ▶ **variants:** board size; other pieces; higher dimension

There are **92 solutions**, or **12 solutions** if we do not count symmetric solutions (under rotation or reflection) as distinct.

## Example: 8 Queens Problem

**Problem:** Place 8 queens on a chess board such that no two queens threaten each other.



Is this candidate a solution?

## Formally: 8 Queens Problem

How can we formalize the problem?

idea:

- ▶ obviously there must be exactly one queen in each file ("column")
- ▶ describe candidates as 8-tuples, where the  $i$ -th entry denotes the rank ("row") of the queen in the  $i$ -th file

formally:  $\mathcal{O} = \langle C, S, opt, v \rangle$  with

- ▶  $C = \{1, \dots, 8\}^8$
- ▶  $S = \{ \langle r_1, \dots, r_8 \rangle \mid \forall 1 \leq i < j \leq 8 : r_i \neq r_j \wedge |r_i - r_j| \neq |i - j| \}$
- ▶  $v$  constant,  $opt$  irrelevant (pure search problem)

## 20.3 Local Search: Hill Climbing

## Algorithms for Combinatorial Optimization Problems

### How can we algorithmically solve COPs?

- ▶ formulation as classical state-space search  
↪ previous chapters
- ▶ formulation as constraint network ↪ Wednesday
- ▶ formulation as logical satisfiability problem ↪ later
- ▶ formulation as mathematical optimization problem (LP/IP)  
↪ not in this course
- ▶ local search ↪ this and next chapter

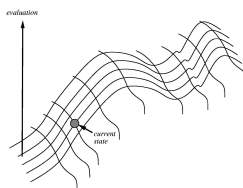
## Search Methods for Combinatorial Optimization

- ▶ main ideas of heuristic search applicable for COPs  
↪ states  $\approx$  candidates
- ▶ main difference: no “actions” in problem definition
  - ▶ instead, we (as algorithm designers) can choose which candidates to consider neighbors
  - ▶ definition of neighborhood critical aspect of designing good algorithms for a given COP
- ▶ “path to goal” irrelevant to the user
  - ▶ no path costs, parents or generating actions  
↪ no search nodes needed

## Local Search: Idea

### main ideas of local search algorithms for COPs:

- ▶ heuristic  $h$  estimates quality of candidates
  - ▶ for pure optimization: often objective function  $v$  itself
  - ▶ for pure search: often distance estimate to closest solution (as in state-space search)
- ▶ do not remember paths, only candidates
- ▶ often only one current candidate ↪ very memory-efficient (however, not complete or optimal)
- ▶ often initialization with random candidate
- ▶ iterative improvement by hill climbing



## Hill Climbing

### Hill Climbing (for Maximization Problems)

*current* := a random candidate

**repeat:**

*next* := a neighbor of *current* with maximum  $h$  value

**if**  $h(\text{next}) \leq h(\text{current})$ :

**return** *current*

*current* := *next*

Remarks:

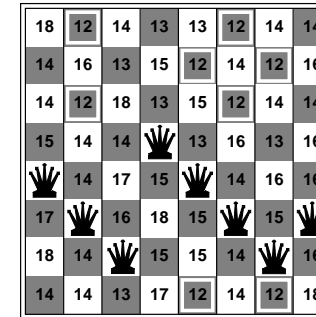
- ▶ search as walk “uphill” in a landscape defined by the neighborhood relation
- ▶ heuristic values define “height” of terrain
- ▶ analogous algorithm for minimization problems also traditionally called “hill climbing” even though the metaphor does not fully fit

## Properties of Hill Climbing

- ▶ always terminates if candidate set is finite (Why?)
- ▶ no guarantee that result is a solution
- ▶ if result is a solution, it is **locally optimal** w.r.t.  $h$ , but no global quality guarantees

## Example: 8 Queens Problem

- Problem:** Place 8 queens on a chess board such that no two queens threaten each other.
- possible heuristic:** no. of pairs of queens threatening each other (formalization as minimization problem)
- possible neighborhood:** move one queen within its file



## Performance of Hill Climbing for 8 Queens Problem

- ▶ problem has  $8^8 \approx 17$  million candidates (reminder: 92 solutions among these)
- ▶ after random initialization, hill climbing finds a solution in around 14% of the cases
- ▶ only around 4 steps on average!

## 20.4 Summary

## Summary

### combinatorial optimization problems:

- ▶ find **solution** of good **quality** (objective value) among many **candidates**
- ▶ special cases:
  - ▶ pure search problems
  - ▶ pure optimization problems
- ▶ differences to state-space search:  
no actions, paths etc.; only “state” matters

### often solved via **local search**:

- ▶ consider **one candidate** (or a few) at a time;  
try to improve it iteratively