

Cooperative and Adaptive Algorithms

Instructor: Alaa Khamis, Ph.D, SMIEEE

Senior Research Scientist  **vestec**
voice enabling systems technology

TAs: Sepehr Eghbali (ECE)

Keyvan Golestan Irani (ECE)

Course Presentation – Tuesday May 6, 2014

Outline

- Course Description
- Course Objectives
- Course Topics
- Course Outline
- Teaching Methodology
- Course Policy
- Resources

Outline

- **Course Description**
- Course Objectives
- Course Topics
- Course Outline
- Teaching Methodology
- Course Policy
- Resources

Course Description

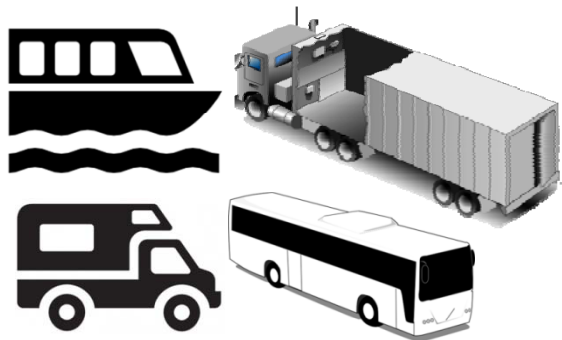
The course addresses the **ill-structured problems** and need for computational intelligence methods.

The course introduces the concepts of **heuristics** and their use in conjunction with search methods, solving problems using heuristics and metaheuristics.

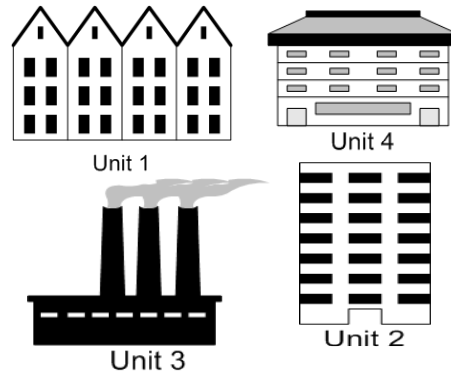
The course also introduces the concepts of **cooperation and adaptation** and how they are influencing new methods for solving **complex problems**.

Course Description

- Complex problems are everywhere



Vehicle routing



Plant layout



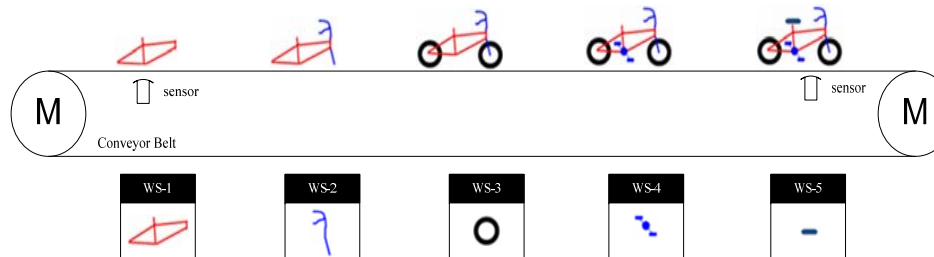
Airline scheduling



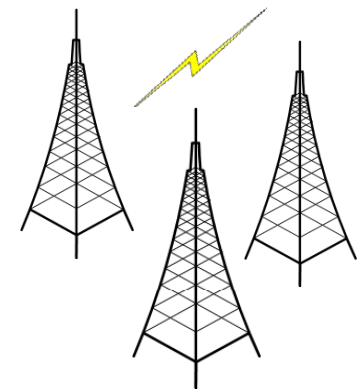
Railway scheduling



Elevator dispatching

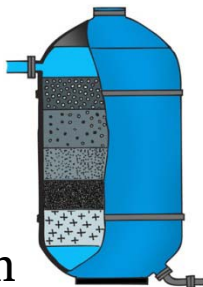


Assembly line balancing

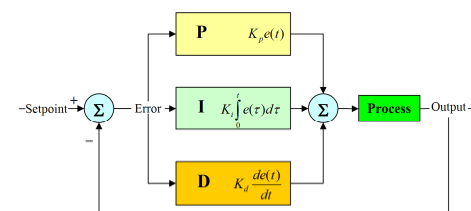


Cell assignment

Filter design



Path Planning

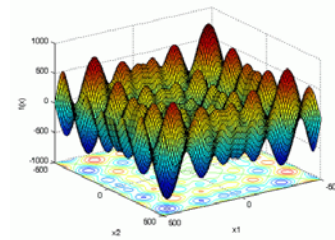


PID Auto-Tuning

Course Description

- Some Sources of Complexity

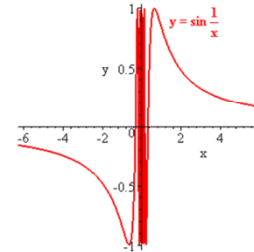
Multimodality



Lack of exact mathematical model

$$\begin{aligned} \min \quad & \sum_{i,j} c_{ij} x_{ij} \\ \text{s.t.} \quad & \sum_i x_{ij} = 1 \quad \forall j, \quad \sum_j x_{ij} = 1 \quad \forall i, \\ & 0 \leq x_{ij} \leq 1 \quad x_{ij} \text{ integer}, \\ & u_1 = 1, \\ & 2 \leq u_i \leq n \quad \forall i \neq 1, \\ & u_i - u_j + 1 \leq (n-1)(1 - x_{ij}) \quad \forall i \neq 1, \forall j \neq 1. \end{aligned}$$

Non-Differentiability



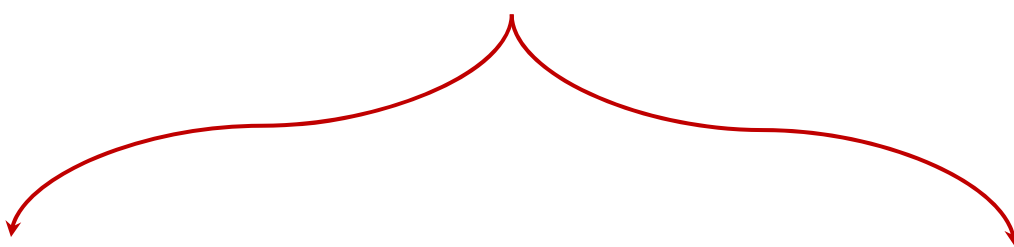
Combinatorial nature



Distributed nature

Course Description

Engineering Problems



well-structured Problems

- Can be stated in terms of numerical variables.
- Goals can be specified in terms of a well defined objective function.
- There exists an algorithmic solution.
- Correct answers required

ill-Structured Problems[1]

- Those which do not meet one or more of the left.
- Division between well and ill-structured is less precise.
- Some problem even they may be well structured in principle, they may be ill-structured in practice [2].

Course Description

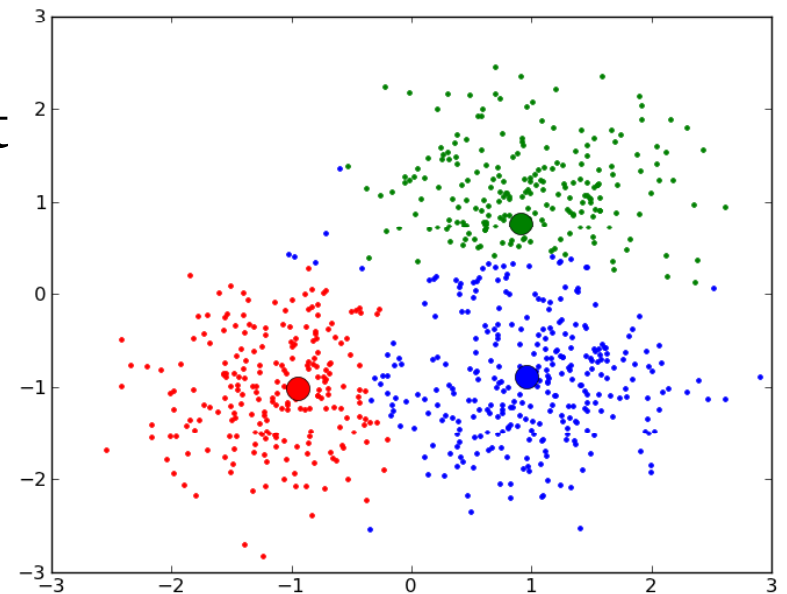
- **Solution strategies for ill-structured problems**
 - ◇ Retrieving a solution or an answer.
 - ◇ Starting from a guess and then improve on it .
 - ◇ Searching among alternatives.
 - ◇ Searching forward from the problem to the answer.
 - ◇ Searching backward from a goal to the situations of the problem.

Course Description

- **Example of well-structured problem but practically ill-structured**

The Clustering Problem

- ◇ Given n objects, group them in c groups (clusters) in such a way that all objects in a single group have a “natural” relation to one another, and objects not in the same group are somehow different.

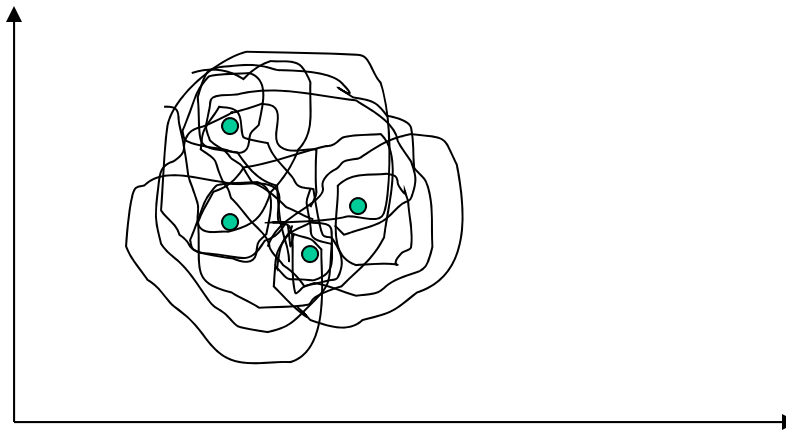


[3]

Course Description

- **Example of well-structured problem but practically ill-structured**

The Clustering Problem



Course Description

- **Example of well-structured problem but practically ill-structured**

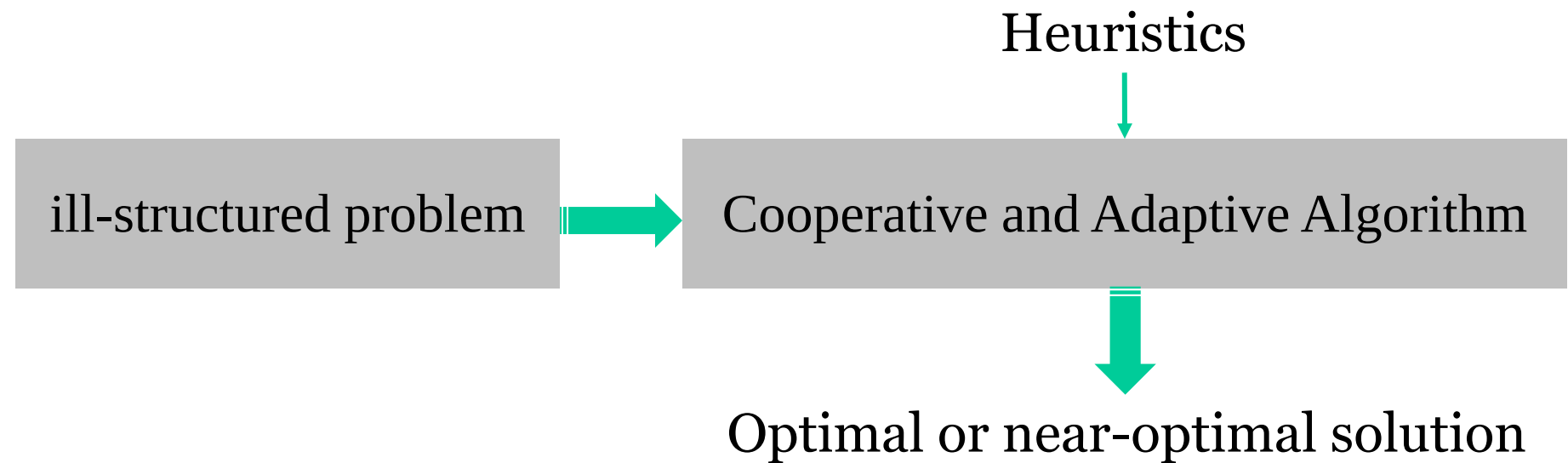
The Clustering Problem: Exponential Growth

$$N(n, c) = \left(\frac{1}{c!} \right) \sum_{m=1}^c (-1)^{c-m} \binom{c}{m} m^n$$

[See [partition theory](#)]

- ◇ Example: If **$n=50$** and **$c=4$** , the number is **5.3×10^{28}**
- ◇ If **n** is increase to **100** , the number becomes **6.7×10^{58}** .
- ◇ So enumerating all possible partitions for large problems is practically not feasible

Course Description



Course Description

- **Adaptive Algorithm**

- ◇ To adjust to new or different situations
- ◇ Ability to improve behaviour
- ◇ Evolution
- ◇ Learning from instructor, example or by discovery
- ◇ Generalization

Course Description

- **Adaptive Algorithm: Types of adaptation**

- ◇ **Deterministic:** when the control parameter is changed according to some deterministic update rule without taking into account any information from the search algorithm,
- ◇ **Adaptive:** when the update rule takes information from the search algorithm and changes the control parameter accordingly,
- ◇ **Self-adaptive:** when the update rule itself is being adapted.

Course Description

- **Cooperative Algorithm**

- ◇ A group to solve a joint problem or perform a common task(s) based on sharing the responsibility for reaching the goal.
- ◇ Direct cooperation
- ◇ Indirect/Stigmergic cooperation
- ◇ Independent actions

Course Description

- **Heuristics**

Heuristics is a solution strategy or rules by **trial-and-error** to produce acceptable (**optimal or near-optimal**) solutions to a complex problem in a reasonably practical time.

Examples: FIFO, LIFO, earliest due date first, largest processing time first, shortest distance first, etc.

Course Description

- **Heuristics:** 8-Puzzle/sliding-block puzzle/tile-puzzle Problem

1	4	2
3	7	5
6		8

Current state

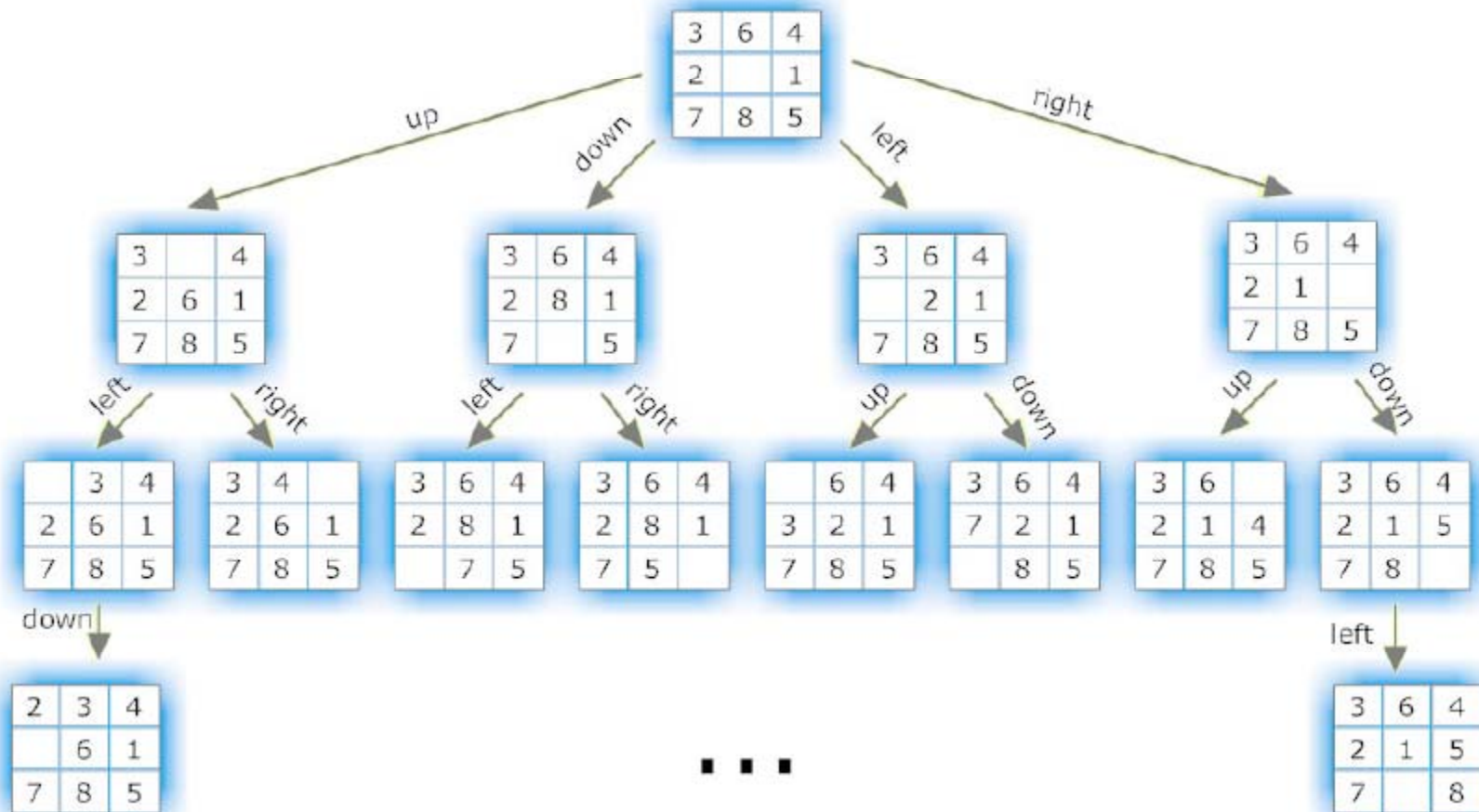


	1	2
3	4	5
6	7	8

Goal state

Course Description

- **Heuristics:** 8-Puzzle/sliding-block puzzle/tile-puzzle Problem



The 8-puzzle can be solved in 31 moves or fewer.

[4]

Course Description

- **Heuristics:** 8-Puzzle/sliding-block puzzle/tile-puzzle Problem

1	4	2
3	7	5
6		8

Current state

	1	2
3	4	5
6	7	8

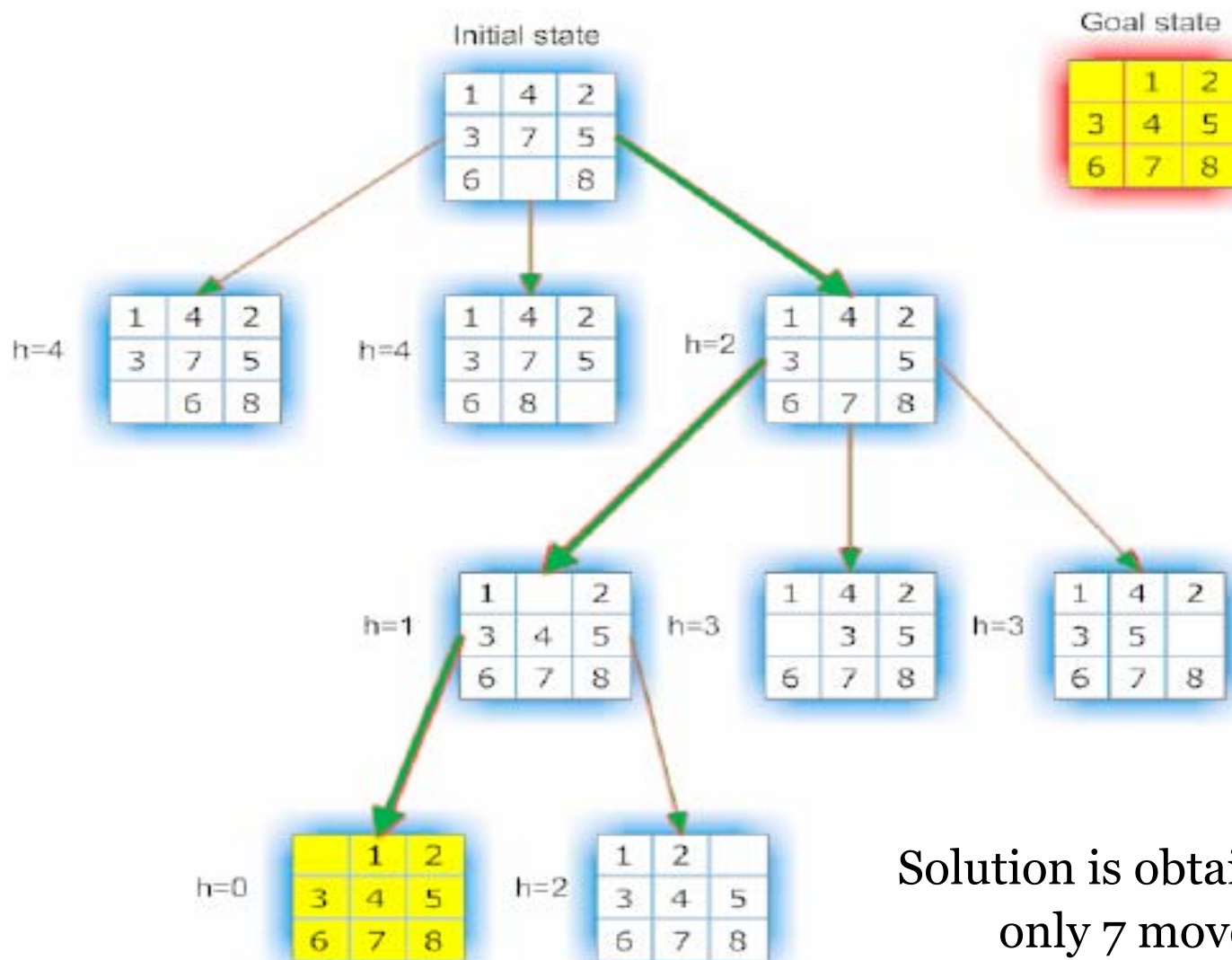
Goal state

- ◇ We can define the following heuristics :
- ◇ $h_1(n)$ = the number of **misplaced tiles** in the current state relative to the goal state.
- ◇ In this example, $h_1(n) = 4$ because tiles 1, 4, 7 and blank are out of place. Obviously, lower values for h_1 are preferred.

[4]

Meta-Heuristic Optimization

- **Meta-Heuristics: 8-Puzzle Problem**



Solution is obtained in
only 7 moves

[4]

Outline

- Course Description
- **Course Objectives**
- Course Topics
- Course Outline
- Teaching Methodology
- Course Policy
- Resources

Course Objectives

- Introduce the concepts of **cooperation** and **adaptations** and how they are influencing new methods for **solving complex problems**.
- Study **meta-heuristics, evolutionary computing methods, swarm intelligence, ant-colony algorithms, particle swarm methods**.
- Illustrate the use of these algorithms in solving **continuous and discrete problems** that arise in engineering applications.

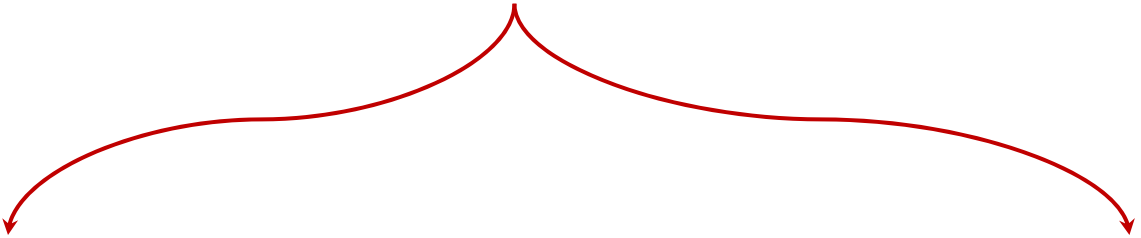
Outline

- Course Description
- Course Objectives
- **Course Topics**
- Course Outline
- Teaching Methodology
- Course Policy
- Resources

Course Topics

Optimization techniques are **search methods**, where the goal is to find a solution to an optimization problem, such that a given quantity is optimized, possibly subject to a set of constraints.

Optimization Algorithms



Exact Algorithms

- Find the optimal solution
- High computational costs.

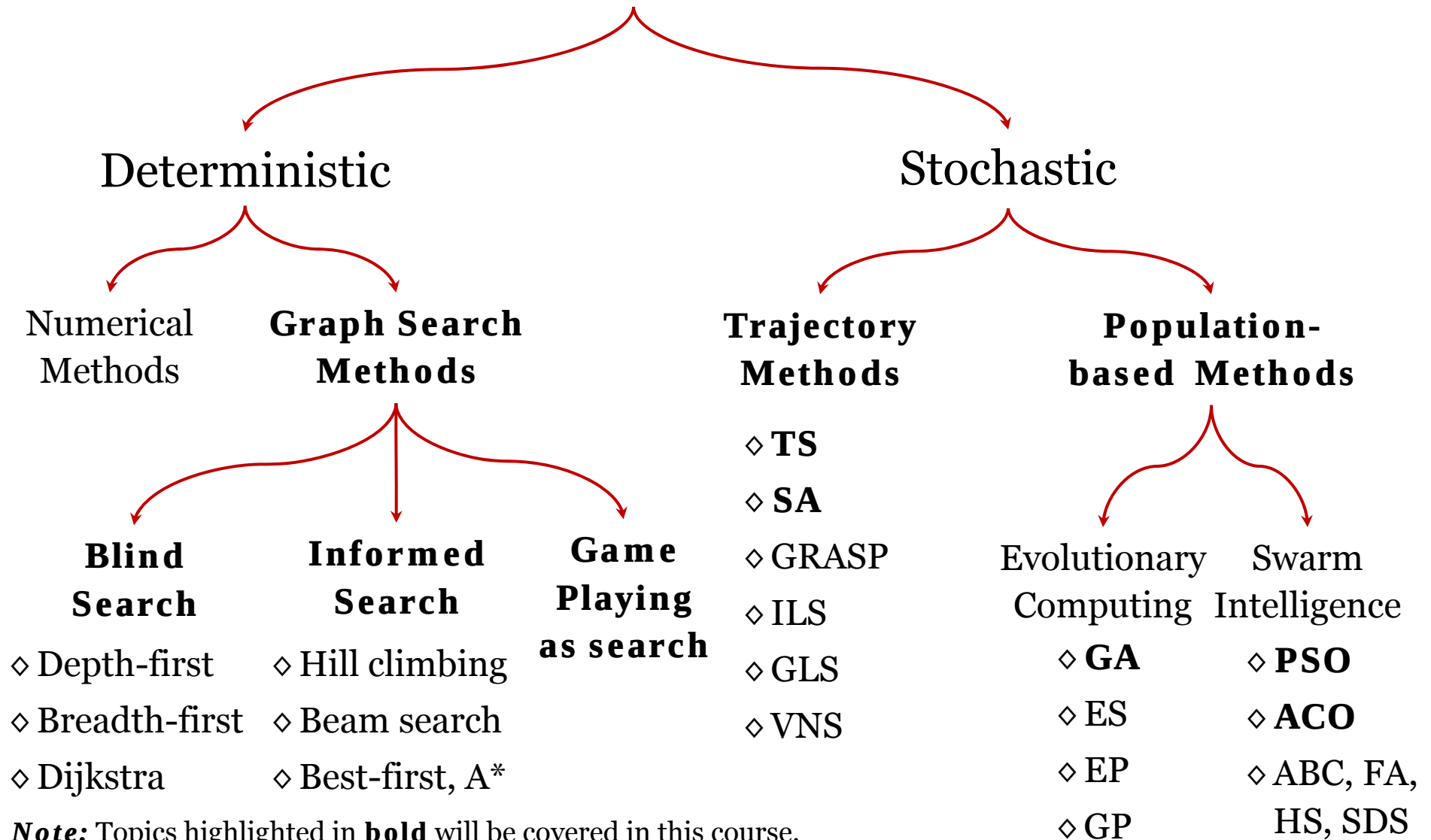
Approximate Algorithms

- Find near-optimal solutions
- Low computational costs.

In most of the real applications, **quickly finding near-optimal** solution is better than **spending large amount of time** in search for optimal solution.

Course Topics

Search Methods



Note: Topics highlighted in **bold** will be covered in this course.

Course Topics

Introductory Concepts

- Optimization Theory and Combinatorics
- Graph-search Algorithms
- Game Playing as Search
- Metaheuristics

Trajectory-based Metaheuristic Algorithms

- Tabu Search
- Simulated Annealing

Population-based Metaheuristic Algorithms

- P-Metaheuristics
- Evolutionary Computation
- Genetic Algorithms
- Swarm Intelligence
- Particle Swarm Optimization
- Ant Colony Optimization

Outline

- Course Description
- Course Objectives
- Course Topics
- **Course Outline**
- Teaching Methodology
- Course Policy
- Resources

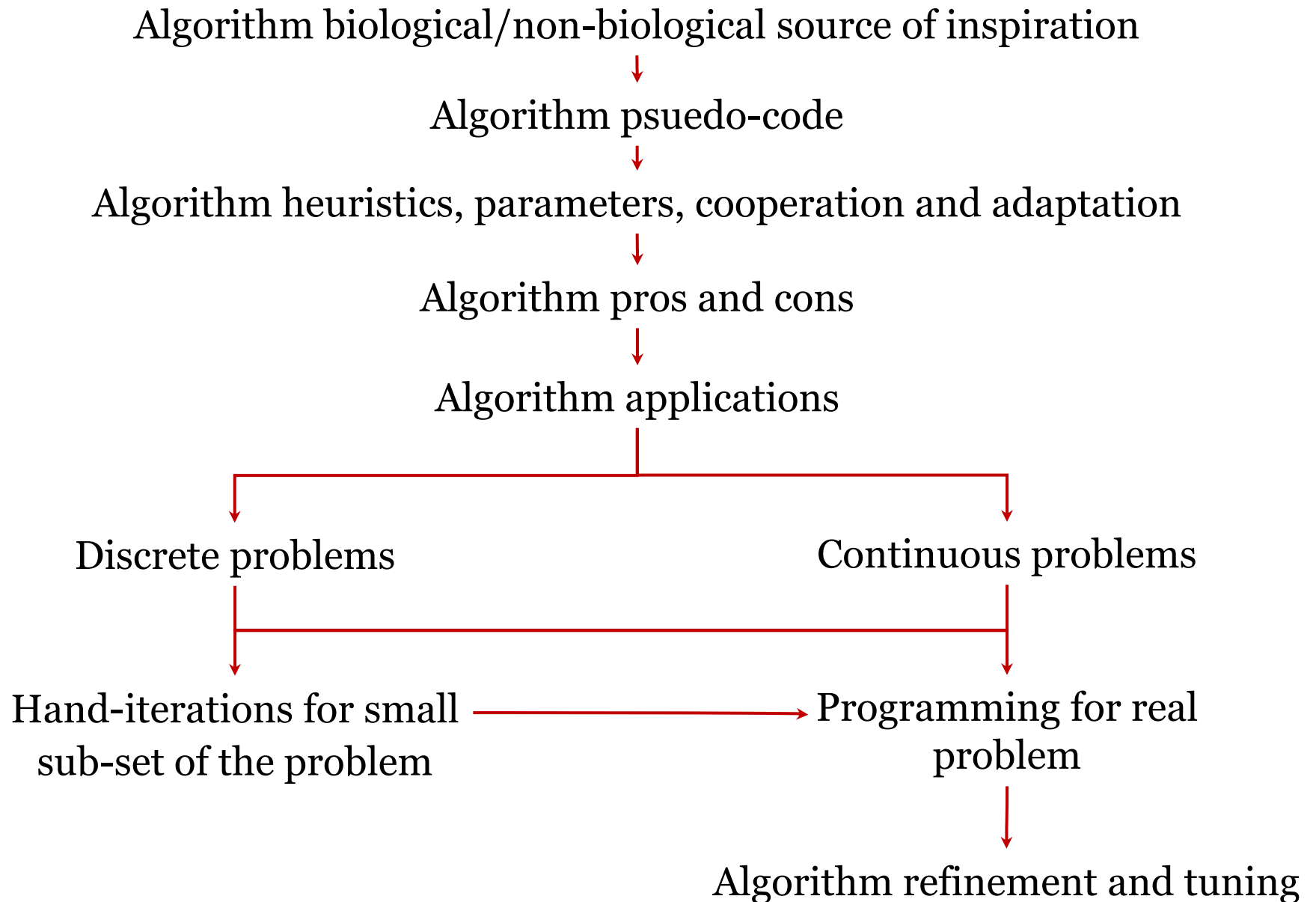
Course Outline

1. Optimization Theory and Combinatorics
2. Graph Search Algorithms
3. Game Playing as Search
4. Trajectory-based Metaheuristic Optimization
5. Tabu Search
6. Simulated Annealing
7. Population-based Metaheuristic Optimization
8. Evolutionary Computation
9. Genetic Algorithms
10. Swarm Intelligence
11. Particle Swarm Optimization (PSO)
12. Ant Colony Optimization (ACO)

Outline

- Course Description
- Course Objectives
- Course Topics
- Course Outline
- **Teaching Methodology**
- Course Policy
- Resources

Teaching Methodology



Outline

- Course Description
- Course Objectives
- Course Topics
- Course Outline
- Teaching Methodology
- **Course Policy**
- Resources

Course Policy

Evaluation Method	Weight
Assignments	20%
Projects	30%
Final Exam	50%

Course Resources

- **Course Website**

Lecture slides, tutorials, assignments, code will be posted on the course webpage:

<http://www.alaakhamis.org/teaching/ECE457A/index.html>

Course Resources

- **Textbook (Optional)**

- ◇ A. Engelbrecht. *Fundamentals of Computational Swarm Intelligence*. Wiley, 2006.
- ◇ Xin-SheYang. *Engineering Optimization: An Introduction with Metaheuristic Applications*. A JOHN WILEY & SONS, INC., 2010.
- ◇ Singiresu S. Rao. *Engineering Optimization: Theory and Practice*. John Wiley & Sons, INC., 2009.
- ◇ C. Revees. *Modern heuristic techniques for combinatorial problems*. Halsted Press, New York, 1993.

Course Resources

- **Books put on Reserve in the Library**

- ◇ Andries P. Engelbrecht. *Fundamentals of Computational Swarm Intelligence*. John Wiley and Sons, 2006.
- ◇ James F. Kennedy, Russell C. Eberhart. *Swarm Intelligence*. Yuhui. Shi Published 2001 by Morgan Kaufmann
- ◇ Eiben and A, Smith. *Introduction to Evolutionary Computing*. J.E, Springer, Berlin, 2003.
- ◇ Revees, C. *Modern heuristic techniques for combinatorial problems*. Halsted Press, New York, 1993.

References

1. Newell. Heuristic Programming: Ill-Structured Problems. Book Chapter in Progress in Operation Research, Vol. III, Aronofsky, ed. – 1969.
2. Simon, H.A. (1973), “Structure of ill structured problems”, Artificial Intelligence, Vol. 4 Nos 3-4, pp. 181-201.
3. M. Kamel, ECE457A: Cooperative and Adaptive Algorithms. Spring 2011, University of Waterloo.
4. Crina Grosan and Ajith Abraham. *Intelligent Systems: A Modern Approach*. Springer, 2011.
5. A. E. Eiben, R. Hinterding and Z. Michalewicz. “Parameter Control in Evolutionary Algorithms”. IEEE Transactions on Evolutionary Computing, vol. 3, issue 2, pp. 124-141, 1999.

Questions?