Schaql af Artificial Intelligence，Nanding University

## Heuristic Search and Evolutionary Algorithms启发式搜索与演化算法

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## 课程相关信息

课程时间地点：周五下午16：00－18：00，逸A－117
课程主页：
https：／／www．lamda．nju．edu．cn／HSEA22／
课程讨论QQ群：881805350
助教：薛轲，刘丹璇（15800040583）
每个ppt的最后附有相关参考文献
答疑时间：周五下午14：00－16：00，逸A－502

## Outline of this course

DPart 1: Traditional heuristic search algorithms
(Assignment 1: 15\%)
DPart 2: Evolutionary algorithms (Assignment 2: 15\%)
DPart 3: Theoretical analysis of evolutionary
algorithms (Assignment 3: 15\%)
$\square$ Part 4: Design of evolutionary algorithms
(Assignment 4: 15\%)
Final exam: 40\% Schaql af Artificial Intelligence，Nanding University

## Heuristic Search and Evolutionary Algorithms

## Lecture 1：Search

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## Search example - route finding


http://www.lamda.nju.edu.cn/qianc/

## Search problem

A search problem can be defined formally by five components:

- Initial state
- Actions
- Transition model
- Goal test
- Path cost

```
e.g., In(Arad)
e.g., Go(Sibiu), Go(Timisoara), Go(Zerind)
e.g., Result(In(Arad), Go(Zerind))=In(Zerind)
e.g., Is(In(Bucharest))
e.g., the sum of action costs
```



## Search problem

A search problem can be defined formally by five components:

- Initial state
- Actions
- Transition model
- Goal test
- Path cost

$$
\begin{aligned}
& \text { e.g., In(Arad) } \\
& \text { e.g., Go(Sibiu), Go(Timisoara), Go(Zerind) } \\
& \text { e.g., } \operatorname{Result(} \operatorname{In}(\text { Arad }), G o(\text { Zerind }))=I n(\text { Zerind }) \\
& \text { e.g., Is(In(Bucharest)) } \\
& \text { e.g., the sum of action costs }
\end{aligned}
$$



## More examples - vacuum world



Actions: Left (L), Right (R), Suck (S)
Transition model: e.g., Result(Initial state, $L$ ) = Initial state
Path cost: the number of actions on the path

## More examples - 8-puzzle



Actions: movements of blank space, i.e., Left, Right, Up and down
Path cost: the number of actions on the path

## More examples - integer construction

Problem: starting with the number 4, apply a sequence of factorial, square root, and floor operations to reach any desired positive integer

- Initial state: 4
- Actions: factorial, square root, and floor operations
- Transition model: e.g., Result(4, factorial)=24
- Goal test: Is(the desired positive integer)
- Path cost: the number of actions on the path


Infinite state space: positive numbers

## More examples - route finding

Problem: find the shortest path between two cities


State: e.g., In(Oradea)

- Initial state
- Actions
- Transition model
- Goal test
- Path cost
e.g., In(Arad)
e.g., Go(Sibiu), Go(Timisoara), Go(Zerind)
e.g., $\operatorname{Result(In(Arad),~Go(Zerind))}=I n($ Zerind $)$
e.g., Is(In(Bucharest))
e.g., the sum of action costs


## More examples - touring

Problem: find the shortest route to visit each city at least once, starting and ending in the same city


## State: e.g., In(Oradea), <br> Visited(\{Arad, Zerind, Oradea\})

- Initial state
e.g., In(Arad), Visited(\{Arad\})
- Actions
e.g., Go(Sibiu), Go(Timisoara), Go(Zerind)
- Transition model
- Goal test
e.g., Is(In(Arad), Visited(\{all the cities\}))
- Path cost
e.g., the sum of action costs


## More examples - traveling salesman

Problem: find the shortest route to visit each city exactly once, starting and ending in the same city


State: e.g., In(Oradea),
Visited( $\{$ Arad, Zerind, Oradea\})

- Initial state e.g., In(Arad), Visited(\{Arad\})
- Actions: can go to non-visited cities, and return to the origin city at last
- Transition model e.g., Result(In(Arad), Visited (\{Arad\}), Go(Zerind)) $=I n($ Zerind $), \operatorname{Visited}(\{$ Arad, Zerind $\})$
- Goal test
e.g., Is(In(Arad), Visited(\{all the cities\}))
- Path cost
e.g., the sum of action costs


## Search problem

A search problem can be defined formally by five components:

- Initial state
- Actions
- Transition model
- Goal test
- Path cost

Solution: a path (i.e., an action sequence) from the initial state to the goal state

Optimal solution: a path with the lowest cost

## Are search problems difficult?

## Complexity classes

- Classify problems according to their complexities
- Class: a set of problems
- P, NP, NP-complete, NP-hard

A decision problem is a mapping from all possible inputs into the set \{yes, no\}

$$
f: I \rightarrow\{1,0\}
$$

## Example of decision problems

- Travelling salesman problem

Given a weighted graph, a specific vertex (i.e., city), and a positive number $k$, is there a tour with cost at most $k$ ?
starting and ending at the specific vertex after having visited each other vertex exactly once

- Graph coloring

Given a undirected graph $G$ and a positive integer $k$, is there a coloring of $G$ using at most $k$ colors?

## assigning colors to each vertex of $G$ such that no adjacent vertices get the same color

## Decision and optimization problems

There are standard techniques for transforming optimization problems into decision problems

Travelling salesman problem
Optimization version: find the shortest route to visit each city exactly once, starting and ending in the same city


Decision version: given a positive number $k$, is there such a route with cost at most $k$ ?

## The class P

## The class P contains decision problems that can be

 solved in polynomial time by a deterministic algorithm- For any input, the algorithm runs for polynomial time
- For any positive input, the algorithm output "yes"
- For any negative input, the algorithm output "no"


## The class NP

## Nondeterministic algorithm

```
void nondetA(String input)
    String s=genCertif();
    Boolean CheckOK=verifyA(input,s);
    if (checkOK)
        Output "yes";
    return;
```

Step 1: guess a solution
Step 2: verify the solution
If yes, output "yes"
Otherwise, no output

Given the same input, the algorithm may behave differently in different executions

## Nondeterministic traveling salesman

- Travelling salesman problem

Given a weighted graph, a specific vertex (i.e., city), and a positive number $k$, is there a tour with cost at most $k$ ?


Guess:


Verify: not a tour
$k=14$
No output

## Nondeterministic traveling salesman

- Travelling salesman problem

Given a weighted graph, a specific vertex (i.e., city), and a positive number $k$, is there a tour with cost at most $k$ ?


Guess:


Verify: a tour with cost 15
$k=14$
No output

## Nondeterministic traveling salesman

- Travelling salesman problem

Given a weighted graph, a specific vertex (i.e., city), and a positive number $k$, is there a tour with cost at most $k$ ?


Guess:


Verify: a tour with cost 14
$k=14$
Output: "yes"

## The class NP

## The class NP contains decision problems for which

 there is a polynomial bounded nondeterministic algorithm- For any positive input, there is some execution of the nondeterministic algorithm which outputs "yes" in polynomial time



## The class NP-hard

- Let $T$ be a function mapping from the input set of a decision problem $P$ into the input set of $Q$
- A decision problem $P$ is polynomially reducible to $Q$ if there exists a function $T$ satisfying:
$\checkmark T$ can be computed in polynomial time
$\checkmark x$ is a "yes" input for $P$ iff $T(x)$ is a "yes" input for $Q$



## The class NP-hard

- A decision problem $P$ is polynomially reducible to $Q$ if there exists a function $T$ satisfying:
$\checkmark T$ can be computed in polynomial time
$\checkmark x$ is a "yes" input for $P$ iff $T(x)$ is a "yes" input for $Q$ $Q$ is at least as hard as $P$
- A problem $Q$ is in NP-hard if every problem $P$ in NP is polynomially reducible to $Q$
$Q$ is at least as hard as any problem in NP
- A problem is in NP-complete if it is in both NP and NP-hard the hardest problems in NP


## P, NP, NP-complete and NP-hard


$\mathrm{P} \neq \mathrm{NP}$

$P=N P$

## Hard search problem

Many search problems are NP-hard, e.g.,

- n-puzzle: NP-complete


Start State


Goal State

- Travelling salesman problem: NP-hard



## Search algorithms

Route finding: the shortest path from Arad to Bucharest


Search tree: the possible action sequences starting from the initial state

Branch: action Node: state


## Tree-search algorithms

function Tree-search(problem) returns a solution or failure initialize the frontier using the initial state of problem loop do
if the frontier is empty then return failure
choose a leaf node and remove it from the frontier
if the node contains a goal state, return the corresponding solution expand the chosen node, adding the resulting nodes to the frontier


The chosen node: Arad
Frontier: Sibiu, Timisoara, Zerind


The chosen node: Sibiu Frontier: Arad, Fagaras, Oradea, Rimnicu Vilcea, Timisoara, Zerind

## Graph-search algorithms

function Graph-search(problem) returns a solution or failure initialize the frontier using the initial state of problem

## loop do

if the frontier is empty then return failure
choose a leaf node and remove it from the frontier
if the node contains a goal state, return the corresponding solution
add the node to the explored set
expand the chosen node, adding the resulting nodes to the frontier only if not in the frontier or explored set


The chosen node: Arad
Explored set: Arad
Frontier: Sibiu, Timisoara, Zerind


The chosen node: Sibiu
Explored set: Arad, Sibiu
Frontier: Fagaras, Oradea,
Rimnicu Vilcea, Timisoara, Zerind

## Search algorithms

- Different search algorithms: how to choose a node from the frontier for expansion
$\checkmark$ Breadth-first search: expand the shallowest node
$\checkmark$ Depth-first search: expand the deepest node
- Each search algorithm has two implementations
$\checkmark$ Tree-search
$\checkmark$ Graph-search


## Some notes on implementation

- Data structure of a node of the search tree

- The frontier and explored set can be implemented with a queue and a hash table, respectively


## Performance evaluation criteria

A search algorithm's performance can be evaluated in four ways:

- Completeness

Is the algorithm guaranteed to find a solution when there is one?

- Optimality

Is the solution found by the algorithm optimal?

- Time complexity

How long does the algorithm find a solution? measured by the number of nodes generated during the search

- Space complexity

How much memory is needed until finding a solution?
measured by the maximum number of nodes stored in memory

## Performance evaluation criteria

- Time and space complexity are usually characterized by three quantities:
$\checkmark$ The branching factor $b$, i.e., the maximum number of successors of any node
$\checkmark$ The depth $d$ of the shallowest goal node
$\checkmark$ The maximum length $m$ of any path



## Asymptotic notations

- Let $f$ and $g$ be two positive functions defined on integers, i.e., $f, g: N \rightarrow \mathrm{R}^{+}$
- $f \in O(g)$ if there exist positive constants $c$ and $n_{0}$ such that

$$
\forall n \geq n_{0}: f(n) \leq c \cdot g(n)
$$

$$
\lim _{n \rightarrow \infty} \frac{f(n)}{g(n)}<\infty
$$

- $f \in o(g)$ if for any positive constant $c$, there exists positive constant $n_{0}$ such that

$$
\forall n \geq n_{0}: f(n)<c \cdot g(n) \quad \lim _{n \rightarrow \infty} \frac{f(n)}{g(n)}=0
$$

## Asymptotic notations

- Let $f$ and $g$ be two positive functions defined on integers, i.e., $f, g: N \rightarrow \mathrm{R}^{+}$
- $f \in \Omega(g)$ if $g \in O(f)$

$$
\lim _{n \rightarrow \infty} \frac{f(n)}{g(n)}>0
$$

- $f \in \omega(g)$ if $g \in o(f)$

$$
\lim _{n \rightarrow \infty} \frac{f(n)}{g(n)}=\infty
$$

- $f \in \Theta(g)$ if $f \in O(g)$ and $f \in \Omega(g)$

$$
0<\lim _{n \rightarrow \infty} \frac{f(n)}{g(n)}<\infty
$$

## Asymptotic notations

- Let $f$ and $g$ be two positive functions defined on integers, i.e., $f, g: N \rightarrow \mathrm{R}^{+}$

| $f \in O(g)$ | $f \leq g$ |
| :--- | :--- |
| $f \in o(g)$ | $f<g$ |
| $f \in \Omega(g)$ | $f \geq g$ |
| $f \in \omega(g)$ | $f>g$ |
| $f \in \Theta(g)$ | $f=g$ |

## Asymptotic notations - example

$$
\forall \alpha>0: \log n \in o\left(n^{\alpha}\right)
$$

$$
\begin{aligned}
& \lim _{n \rightarrow \infty} \frac{\log n}{n^{\alpha}}=\frac{1}{\ln 2} \lim _{n \rightarrow \infty} \frac{\ln n}{n^{\alpha}}=\frac{1}{\ln 2} \lim _{n \rightarrow \infty} \frac{1}{n \cdot \alpha n^{\alpha-1}}=0 \\
& \text { L'Hospital's rule }
\end{aligned}
$$

For any positive integer $k, \forall c>1: n^{k} \in o\left(c^{n}\right)$

$$
\lim _{n \rightarrow \infty} \frac{n^{k}}{c^{n}}=\frac{k}{\ln c} \lim _{n \rightarrow \infty} \frac{n^{k-1}}{c^{n}}=\frac{k!}{(\ln c)^{k}} \lim _{n \rightarrow \infty} \frac{1}{c^{n}}=0
$$

## Asymptotic notations - example

$$
\begin{gathered}
\lim _{n \rightarrow \infty} \frac{n!}{2^{n}}=\lim _{n \rightarrow \infty} \frac{2^{n} \in o(n!)}{\sqrt{2 \pi n}\left(\frac{n}{e}\right)^{n}} \cdot \frac{\sqrt{2 \pi n}\left(\frac{n}{e}\right)^{n}}{2^{n}} \\
=\lim _{n \rightarrow \infty} \frac{\sqrt{2 \pi n}\left(\frac{n}{e}\right)^{n}}{2^{n}}=\lim _{n \rightarrow \infty} \sqrt{2 \pi n}\left(\frac{n}{2 e}\right)^{n}=\infty \\
\text { Stirling's approximation } \\
n!\in \omega\left(2^{n}\right)
\end{gathered}
$$

## Asymptotic notations - properties

- Transitivity
$f(n) \in O(g(n)) \wedge g(n) \in O(h(n)) \quad \square f(n) \in O(h(n))$
- Reflexivity
$f(n) \in O(f(n)) \quad f(n) \in \Omega(f(n)) \quad f(n) \in \Theta(f(n))$
- Order of sum functions
$O(f(n)+g(n))=O(\max \{f(n), g(n)\})$


## Summary

- What is search
- Problem complexity: P, NP, NP-hard, NP-complete
- Tree-search and graph-search
- Performance evaluation criteria
- Asymptotic notations


## References

- S. J. Russell and P. Norvig. Artificial Intelligence: A Modern Approach. Chapter 3.1-3.3, Third edition.
- T. H. Cormen, et al. Introduction to Algorithms. Chapter 3.1 and 34 , Second edition.

