DEPARTMENT OF COMPUTER SCIENCE



# CSCI-564 CONSTRAINT PROCESSING AND HEURISTIC SEARCH

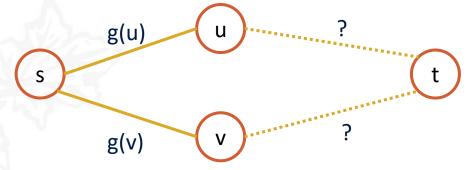
**LECTURE 6 - INFORMED SEARCH** 

**Dr. Jean-Alexis Delamer** 



### Recap

- Uninformed search
  - Know the value of one action/edge.
  - No information on the estimated cost to reach the goal.



- Algorithms:
  - Algorithms explore all the nodes.
  - BFS, DFS, Dijkstra, Dynamic programming



# **Informed search**

Heuristic search

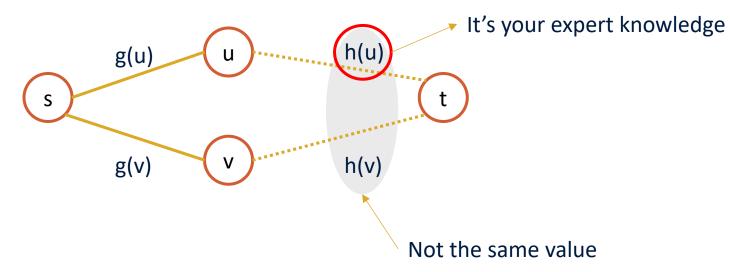
- An estimate of the remaining distance/cost to reach the goal.
- Use the estimate to prioritize the node expansion.
- It's a way to exploit domain knowledge to prune the search tree.
  - Informed search

# STFX

# **Informed search**

• The estimate of a node u is noted h(u).



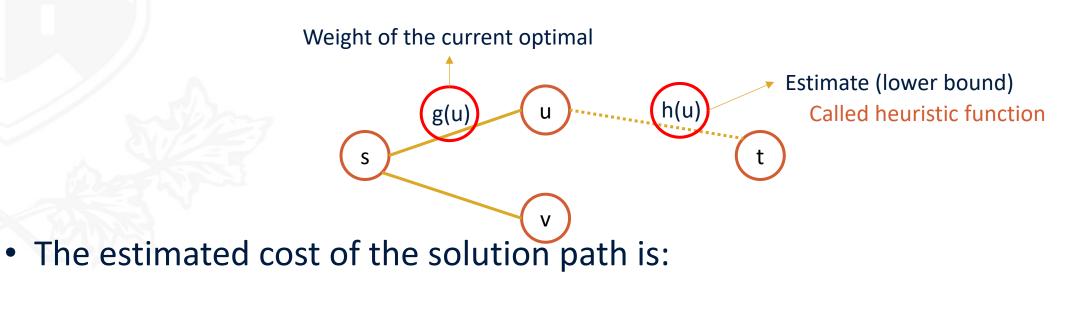


#### Do you know any algorithm that use heuristic search?



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#### • The most prominent heuristic search algorithm is A\*.

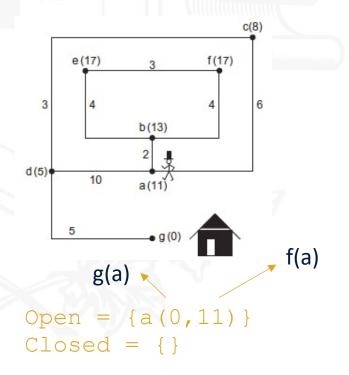


f(u) = g(u) + h(t)

```
Closed \leftarrow \emptyset; Open \leftarrow \{s\}; f(s) \leftarrow h(s)
While (Open \neq \emptyset)
       Remove u from Open with minimum f(u)
       Insert u into Closed
       If u \in T return Path(u)
       Else
              Succ(u) \leftarrow Expand(u)
              Foreach v in Succ(u)
                     If v in Open
                             If q(u) + w(u, v) < q(v)
                                    parent(v) \leftarrow u
                                    f(v) \leftarrow q(u) + w(u, v) + h(v)
                     Else if v in Closed
                             If g(u) + w(u, v) < g(v)
                                    parent(v) \leftarrow u
                                    f(v) \leftarrow q(u) + w(u, v) + h(v)
                                    Remove v from Closed
                                    Insert v in Open
                     Else
```

```
parent(v) \leftarrow u
Initialize f(v) \leftarrow g(u)+w(u,v)+h(v)
Insert v in Open
```

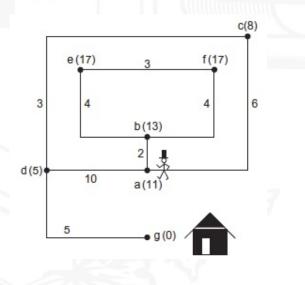




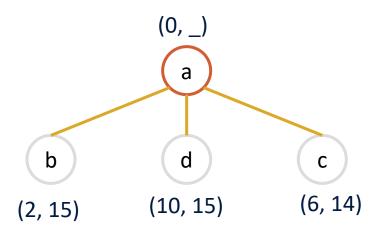






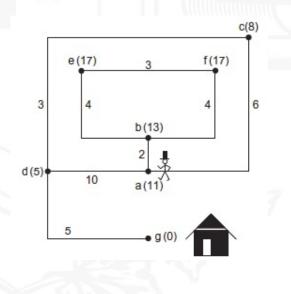


Open = {c(6,14), b(2,15),
 d(10,15)}
Closed = {a}

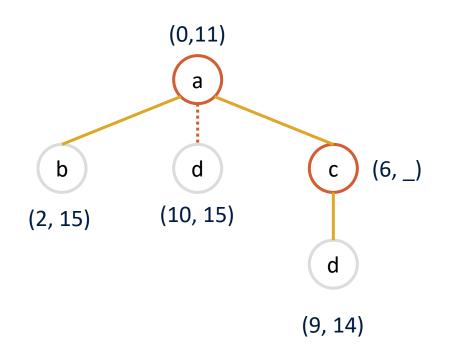




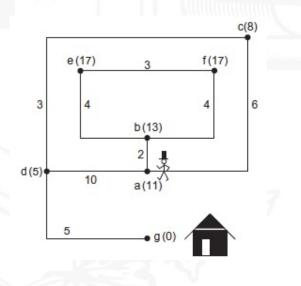




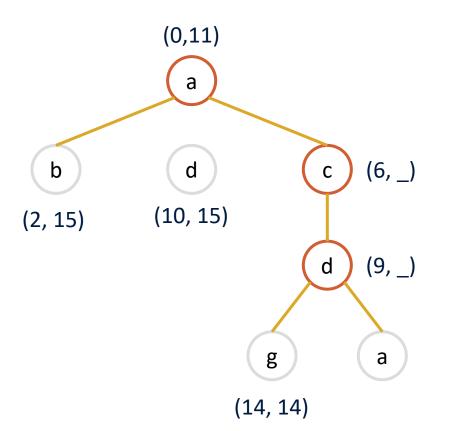
Open = {b(2,15),
 d(9,14) }
Closed = {a, c}



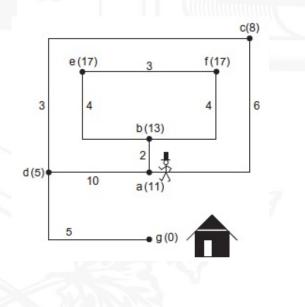




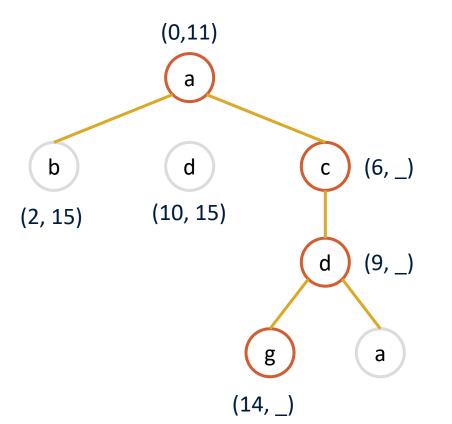
Open = {b(2,15), g(14,14) }
Closed = {a, c, d}







Open = {b(2,15) Closed = {a, c, d, g}

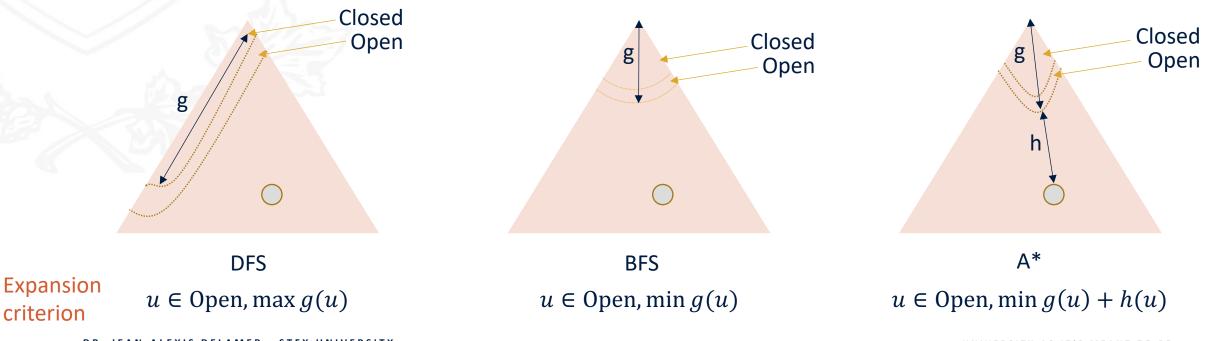


#### We don't need to visit every nodes (e and f are avoided).





• Difference between A\* and other algorithms:



# Optimality

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- We often say that A\* has optimal efficiency:
  - It gives an optimal solution.
  - Expands the minimal number of nodes.

Is it true?

# Optimality

- We often say that A\* has optimal efficiency:
  - It gives an optimal solution.
  - Expands the minimal number of nodes.
- It is true for consistent heuristics.
- But not always for admissible heuristics.

**StFX** 

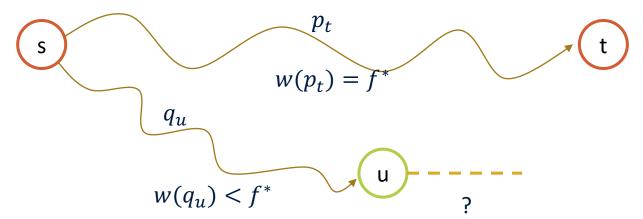
- Definition (Consistent heuristic):
  - A goal estimate h is a consistent heuristic if h(u) ≤ w(u, v) + h(v) for all edges e = (u, v) ∈ E.

#### Theorem (Efficiency Lower Bound):

• Let G be a problem graph with nonnegative weight function, with initial node s and final node set T, and let  $f^* = \delta(s, T)$  be the optimal solution cost. Any optimal algorithm has to visit all  $u \in V$  nodes with  $\delta(s, u) < f^*$ .

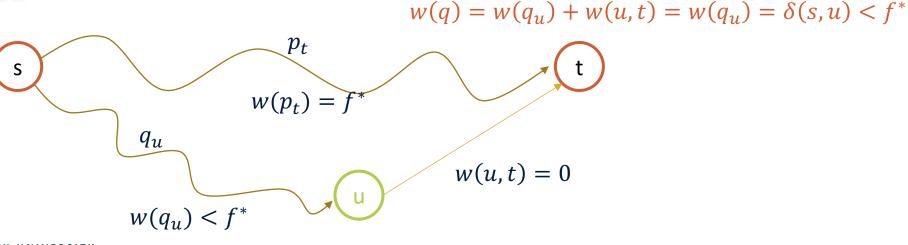
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- Proof:



### • Theorem (Efficiency Lower Bound):

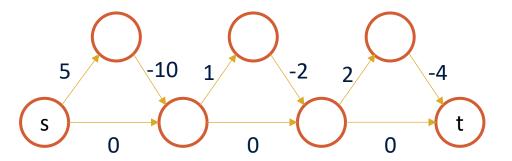
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- Theorem (Efficiency Lower Bound):
  - Let G be a problem graph with nonnegative weight function, with initial node s and final node set T, and let  $f^* = \delta(s, T)$  be the optimal solution cost. Any optimal algorithm has to visit all  $u \in V$  nodes with  $\delta(s, u) < f^*$ .
- The number of nodes that any algorithms expands will have to be larger than or equal to the number of nodes that A\* expands.

# **Admissible heuristics**

- Definition (Admissible Heuristic):
  - An estimate h is an admissible heuristic if it is a lower bound for the optimal solution costs; that is, h(s) ≤ δ(s,T) for all s ∈ V.
- If we have admissibility but not consistency, A\* will reopen nodes.
- Worse! A\* might reopen nodes exponentially many times.
  - This behavior does not appear frequently in practice.





# **Admissible heuristics**

- In these cases, we use other algorithms:
  - Bellman-Ford algorithm, that deal with negative edge costs.
- But A\* is not optimal with non consistent heuristics.



### **A\***

- A\* can find a shortest path even though it expands every state at most once. It does not need to reexpand states that it has expanded already.
- A\* is at least as efficient as every other search algorithm (that has the same heuristic values as A\*). They needs to expand at least the states that A\* expands.



# Exercise

- The Missionary and Cannibals:
  - At one side of a river there are three missionaries and three cannibals.
  - They have a boat that can transport at most two persons. The goal for all persons is to cross the river. The boat requires someone to cross the river.
  - At no time should the number of cannibals exceed the number of missionaries.
- 1. Draw the problem graph and provide its adjacency list representation.
- 2. Solve the problem via DFS and BFS by annotating the graph with numbers.
- 3. Consider a heuristic function that counts the number of people on the other side of the river. Do you observe an inconsistency?