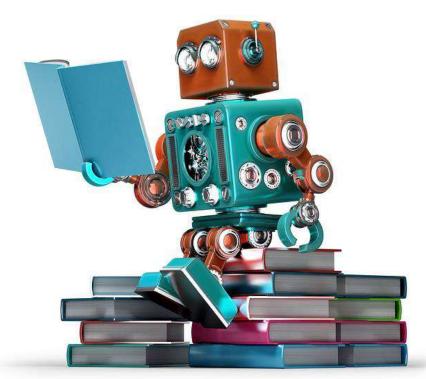




# REASONING SYSTEMS DAY 1







https://robohub.org/wp-content/uploads/2016/11/bigstock-Retro-Robot-Reading-A-Book-Is-110707406.jpg

#### DAY 1 AGENDA





1.1 Reasoning Systems Overview

1.2 Uninformed Search Techniques

1.3 Informed Search Techniques (part 1/2)

1.4 Search Representation Workshop

#### **DAY 1 TIMETABLE**





No	Time	Topic	By Whom	Where
1	9 am	Welcome and Introduction	GU Zhan (Sam)	Class
2	9.30 am	1.1 Reasoning Systems Overview	GU Zhan (Sam)	Class
3	10.10 am	Morning Break		
4	10.30 am	1.2 Uninformed Search Techniques	GU Zhan (Sam)	Class
5	12.10 pm	Lunch Break		
6	1.30 pm	1.3 Informed Search Techniques (part 1/2)	All	Class
7	3.10 pm	Afternoon Break		
8	3.30 pm	1.4 Search Representation Workshop	All	Class
9	4.50 pm	Summary and Review	All	Class
10	5 pm	End		



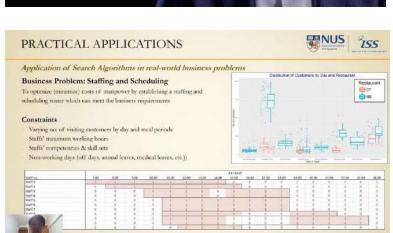


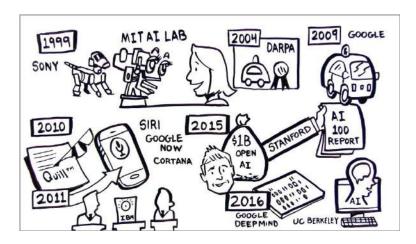




# Al History



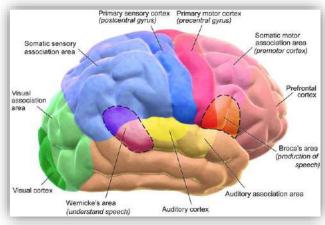












notor
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Prefrontal
cortex

roca's area
oduction of
speech)

Thinking vs Acting Human vs Rational

(acting = behaviour)

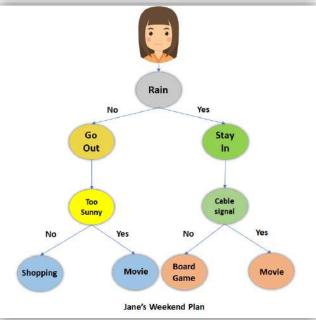
(rationality = doing the right thing)

Systems that think like humans (cognitive science)

Systems that think rationally (logic/laws of thought)

Systems that act like humans (c.f. Turing test)

Systems that act rationally (agents)





Source https://slideplayer.com/slide/4644026/15/images/20/Systems+that+think+like+humans+%28cognitive+science%29.jpg





Question Answering System: IBM Watson

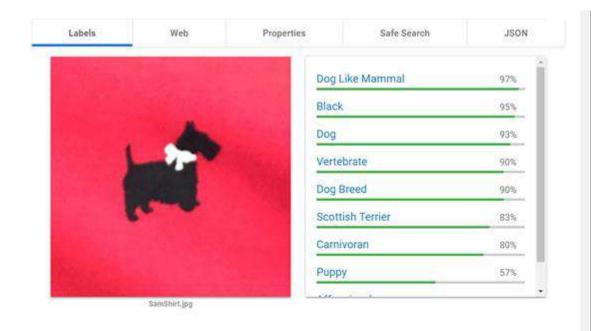


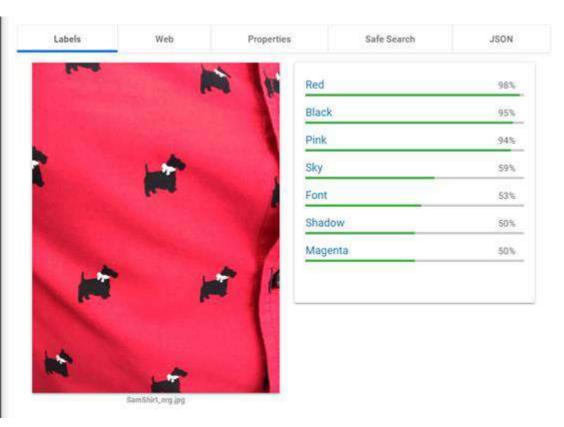






# Image Object Recognition: Google Vision



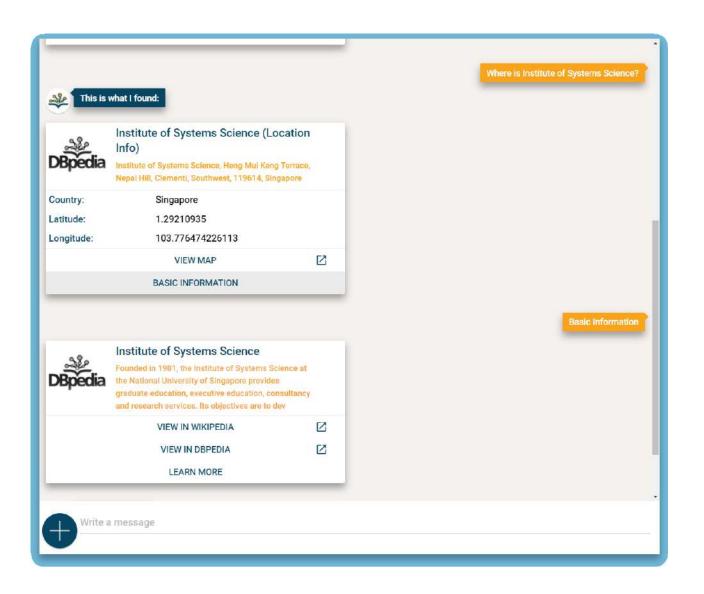






Chat-Bot: DBpedia

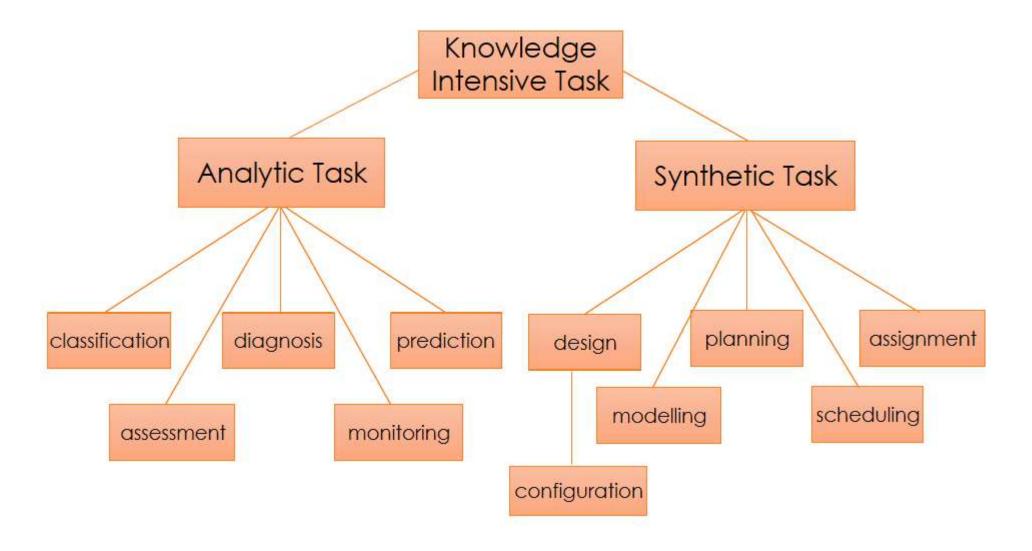








**Problem Solving Task Hierarchy** 







#### **Problem Solving Task Types**

### Analytic Tasks

- System/Solution to be analysed pre-exists, but usually not completely "known".
- Input: some data to trigger the system (e.g. patient symptoms)
- Output: some characterization or behaviours about the system (e.g. cause of illness)

### Synthetic Tasks

- System/Solution does not yet exist.
- Input: requirements about system to be constructed
- Output: constructed system description





#### **Problem Solving of Analytic Tasks**

### Analytic Tasks

Identification, Classification, Prediction, Clustering/Grouping, ...

#### Techniques (Machine Reasoning)

Heuristic Business Rules

**Decision Trees** 

Case Based Reasoning

Fuzzy Logic

Rule Induction

Machine Learning

. . .





#### **Problem Solving of Synthetic Tasks**

## Synthetic Tasks

Planning, Scheduling, Optimisation, Design, ...

#### Techniques (Reasoning Systems)

Uninformed (brute force / blind) Search Informed (heuristic) Search

Simulations

Genetic Algorithms

Reinforcement Learning

Data Mining

• • •









- Solving Problem by Search
- Search Tree Representation
- Depth First Search (DFS)
- Breadth First Search (BFS)



https://modernmarketingtoday.com/wpcontent/uploads/2013/02/searchmarketing.jpg

Solving Problem by Search





- Synthesis of a new valid solution is performed by searching through the (search/solution) space, which contains all possible solutions
- Each possible solution is evaluated to see whether it is valid and/or the optimum (best solution found by now), e.g. a valid employee schedule, a valid vehicle delivery route, an optimal (shortest) vehicle delivery route,
- Validity of solution involves satisfaction of a set of constraints on the solution variables
- Optimality is measured by a user-defined function which measures the "goodness" of the solution, e.g. the shorter delivery route the better.





- Solving Problem by Search
- (1) Create a pool of solution candidates (search space)
- (2) Pick up one candidate solution from pool
- (3) Check whether this candidate is valid (constraints satisfied?)
  - (3)=True If valid, continue
  - (3)=False If not valid, go to (2)
- (4) Check whether this candidate is the best till now (optimal solution?)
  - (4)=True If best, save this solution as the best then continue
  - (4)=False If not best, discard this solution then continue
- (5) Go to (2). Repeat the cycle until a stopping criteria is met.





- **Search Tree Representation**
- Search is illustrated using a search space with a particular restricted structure
- Solutions (search space) can be represented as a Tree
  - Nodes in tree represent

an initial state

an intermediate state

a final state (feasible solution, or failure)

Connection between nodes represents a search step





# Depth First Search (DFS)

 Always prefers to search deeper in the search tree rather than wider.

# 1.2 UNINFORMED SEARCH TECHNIQUES Depth First Search (DFS)





#### Algorithm Pseudo Code

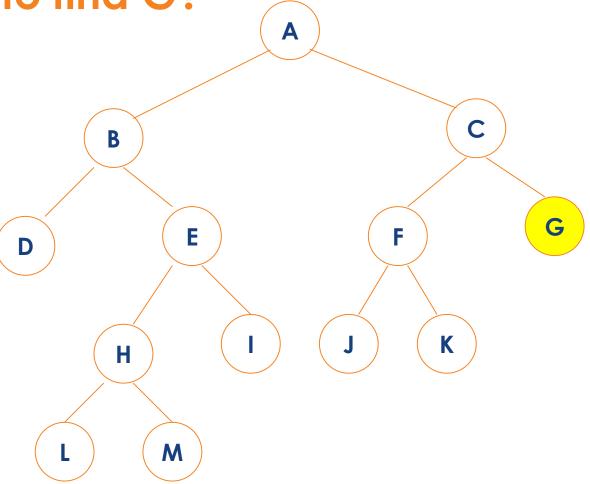
- (1) Set N to be a list of initial nodes
- (2) If N is empty, then exit and signal failure
- (3) Set n to be the first node in N, and remove n from N
- (4) Check n:
  - (4.1) If n is a goal node, then exit and signal success
  - (4.2) Otherwise, add the children of n to the <u>front</u> of N then go to step (2)





Depth First Search (DFS)

Visit order to find G?

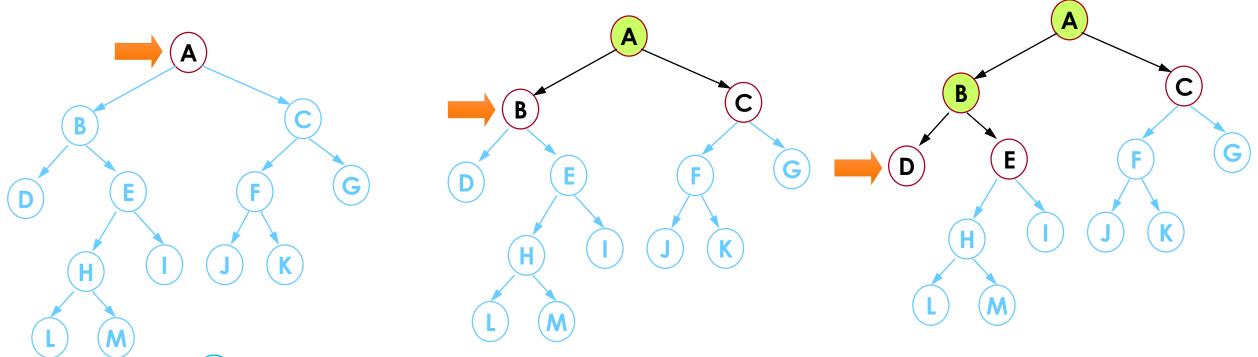






Depth First Search (DFS)

# Keep track of nodes



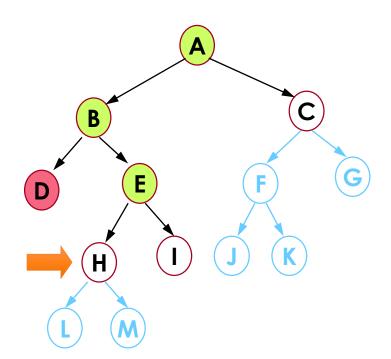
- Node unexplored
- Node waiting to be explored
- Node already explored

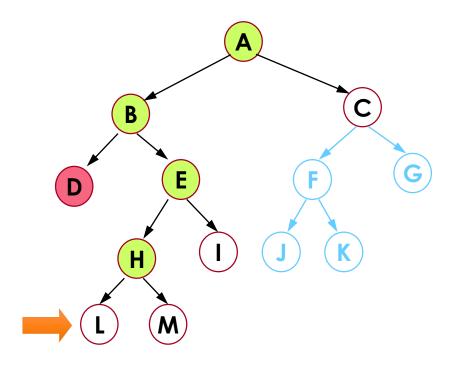




Depth First Search (DFS)

# Keep track of nodes







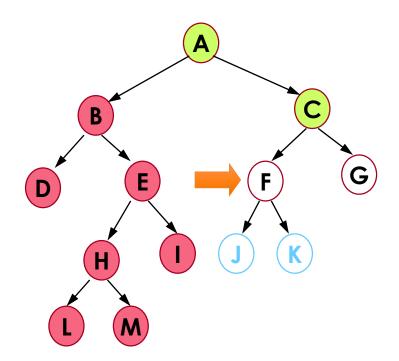
Explored non-solution node can be removed

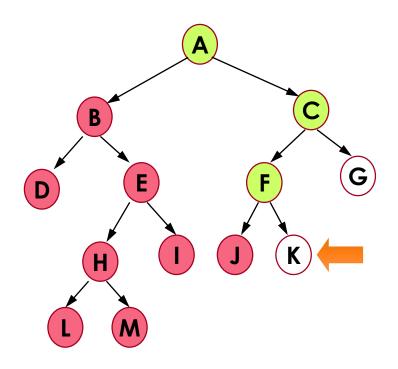




Depth First Search (DFS)

## Keep track of nodes







Explored non-solution node/branch can be removed





Depth First Search (DFS)

 Exercise: Find G B H K M





# **Breadth First Search (BFS)**

• Explores all the nodes at a given depth before processing deeper in the search tree.

# 1.2 UNINFORMED SEARCH TECHNIQUES Breadth First Search (BFS)





#### Algorithm Pseudo Code

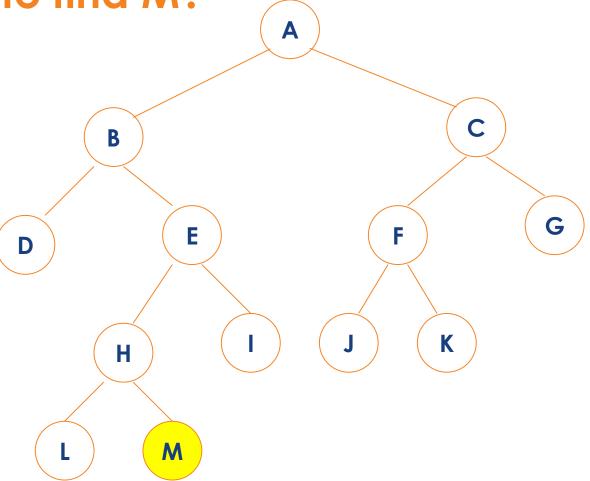
- (1) Set N to be a list of initial nodes
- (2) If N is empty, then exit and signal failure
- (3) Set n to be the first node in N, and remove n from N
- (4) Check n:
  - (4.1) If n is a goal node, then exit and signal success
  - (4.2) Otherwise, add the children of n to the <u>end</u> of N then go to step (2)





**Breadth First Search (BFS)** 

Visit order to find M?







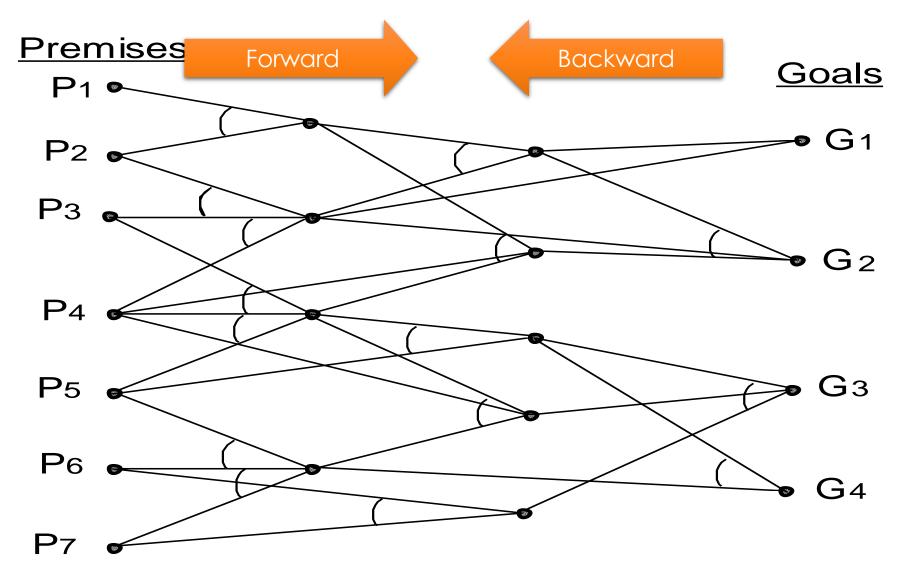
**Breadth First Search (BFS)** 

 Exercise: Find M B G H K





Forward Chaining (BFS) vs. Backward Chaining (DFS)







(PART 1/2)

# 1.3 INFORMED SEARCH TECHNIQUES (1/2)





- Use Heuristics
- Hill Climbing Search (HC)
- A Star Search (A\*)
- Tabu Search (TS)
- Simulated Annealing (SA)
- Informed Search Use Case



https://modernmarketingtoday.com/wpcontent/uploads/2013/02/searchmarketing.jpg

# 1.3 INFORMED SEARCH TECHNIQUES (1/2) Use Heuristics





#### Basic Idea

 It works by firstly sorting the list of nodes, then explore them orderly, according to their **optimality** (best score) determined by an evaluation function f(n)

#### Typical Best-first Strategies

Use heuristics only : Hill Climbing, Tabu search

Use heuristics and past cost : A\*, Late Acceptance

Use heuristics and randomness: Simulated Annealing, GA

# 1.3 INFORMED SEARCH TECHNIQUES (1/2) Use Heuristics





- A key component of evaluation function f(n) is a heuristic function: h(n) = estimated cost of the cheapest path from node n to a goal node. Or estimated degree of difference between the current states/solutions and ultimate goal state.
- By convention, the lower the heuristic value the more promising the node: better to check first. h(n) = 0 when n is goal
- $\odot$  When there is no other information available but only the heuristics, f(n) = h(n)

# 1.3 INFORMED SEARCH TECHNIQUES (1/2) Use Heuristics





- Heuristic Function (Knowledge)
  - h(n) = Correct or Incorrect positions
     Initial=

1	3	2
8		4
5	6	7

•  $h_1(n)$  = number of tiles in their correct goal state positions

•  $h_2(n)$  = number of tiles in their incorrect goal state positions

$$\begin{bmatrix}
 1 & 3 & 2 \\
 8 & 4 & \\
 5 & 6 & 7
 \end{bmatrix}
 = 4$$

# 1.3 INFORMED SEARCH TECHNIQUES (1/2) Use Heuristics





# Heuristic Function (Knowledge)

• h(n) = Manhattan distance (sum of the horizontal & vertical distance each tile is away from its goal state position)

Goal= 
$$\begin{bmatrix} 1 & 2 & 3 \\ 8 & 4 \\ 7 & 6 & 5 \end{bmatrix}$$
  $= (1+1) + 1 + 2 + 2 = 7$ 

Manhattan distance gives a better estimate of the distance to the Goal state

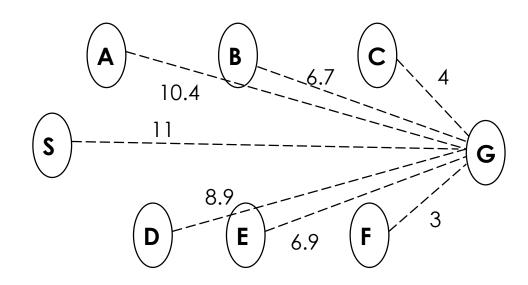
# 1.3 INFORMED SEARCH TECHNIQUES (1/2) Use Heuristics



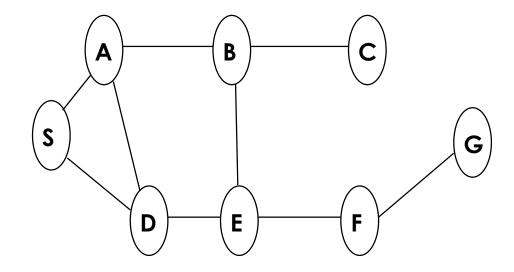


# Heuristic Function (Knowledge)

h(n) = "straight-line" distance between each city & the goal
 (This is useful estimation or heuristic.)



Heuristic Distances



**Actual Roads** 



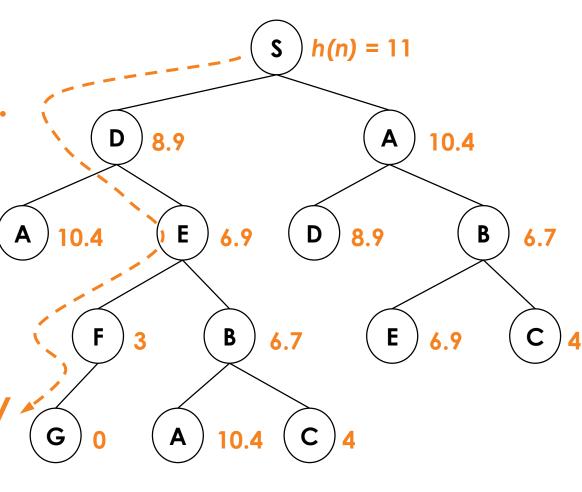


Hill Climbing (Greedy Best First Search)

- Minimize the cost to reach the goal state by expanding the node that is closest to the goal.
- Using only the heuristic values for evaluation function:

$$f(n) = h(n)$$

- Select search node with min(f(n)) at each step.
- Follow a single path all the way to a goal, but can back track when it hits a dead end.

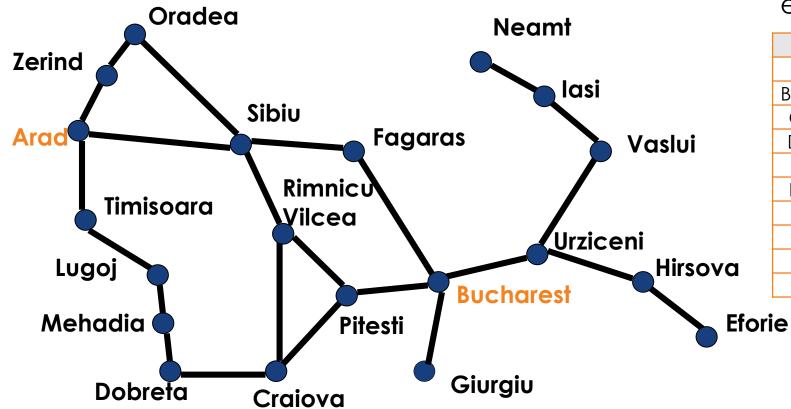






Hill Climbing (Greedy Best First Search)

# Arad → Bucharest



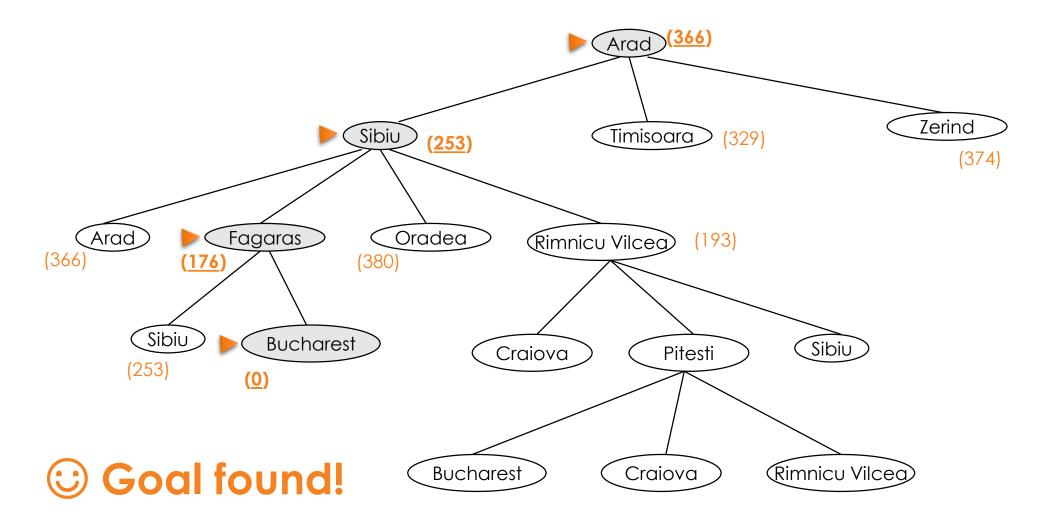
#### h(n) =Straight-line distance from each city to Bucharest

City	Distance	City	Distance
Arad	366	Mehadia	241
Bucharest	0	Neamt	234
Craiova	160	Oradea	380
Dobreta	242	Pitesti	100
Eforie	161	Rimnicu Vilcea	193
Fagaras	176	Sibiu	253
Giurgiu	77	Timisoara	329
Hirsova	151	Urziceni	80
lasi	226	Vaslui	199
Lugoj	244	Zerind	374





Hill Climbing (Greedy Best First Search)

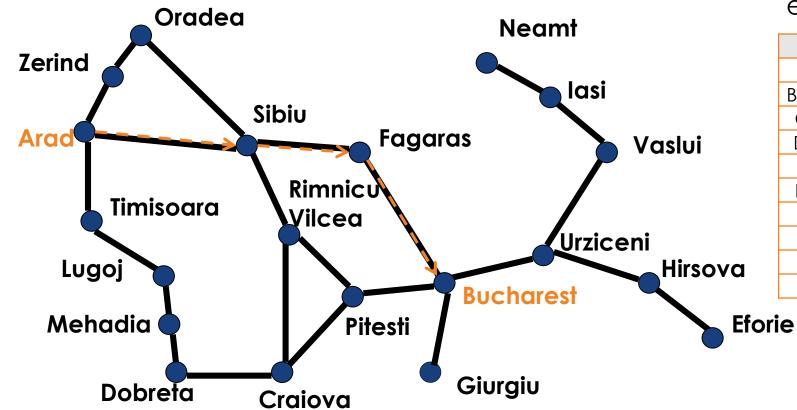






Hill Climbing (Greedy Best First Search)

# Arad → Bucharest



#### h(n) =Straight-line distance from each city to Bucharest

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lasi	226	Vaslui	199
Lugoj	244	Zerind	374

# 1.3 INFORMED SEARCH TECHNIQUES (1/2) A\* Search





- A\* search is the most widely-known form of bestfirst search
  - This strategy evaluates each search node by combining g(n), the past (path) cost from the start node to current node n, and h(n), the estimated future (path) cost: the cheapest path/cost from current node n to a goal node
  - Estimated total cost of the cheapest solution through n

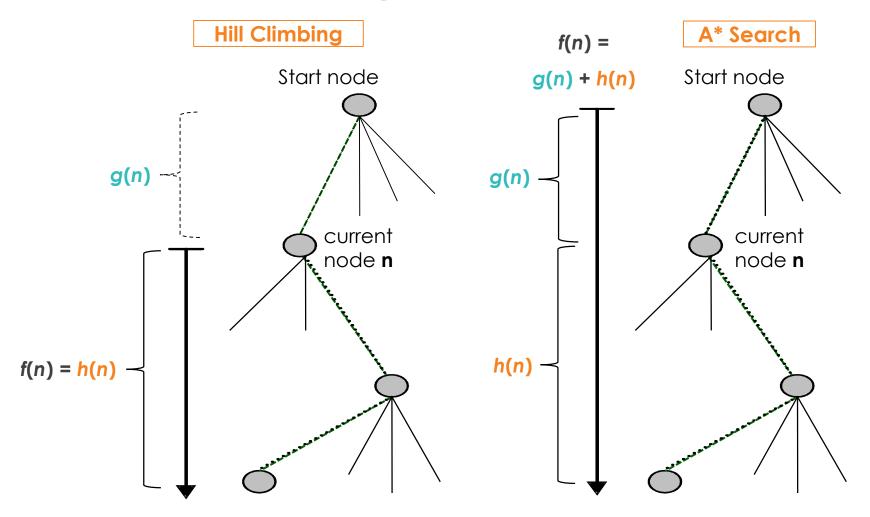
$$f(n) = g(n) + h(n)$$

 $\odot$  g(n) is exactly known, but h(n) is only an estimation with possible error.





#### A\* Search vs. Hill Climbing



Select then expand "best-path-**from-n**-to-goal" child-node at each layer

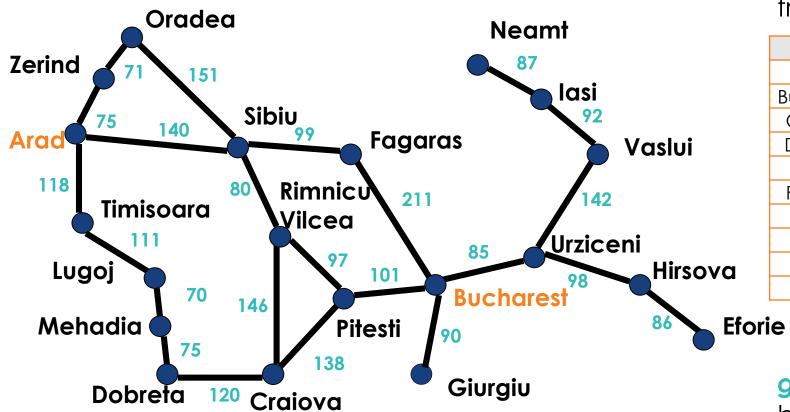
Select then expand "best-path-**from-start**-to-goal" child-node at each layer

# 1.3 INFORMED SEARCH TECHNIQUES (1/2) A\* Search





# Arad → Bucharest



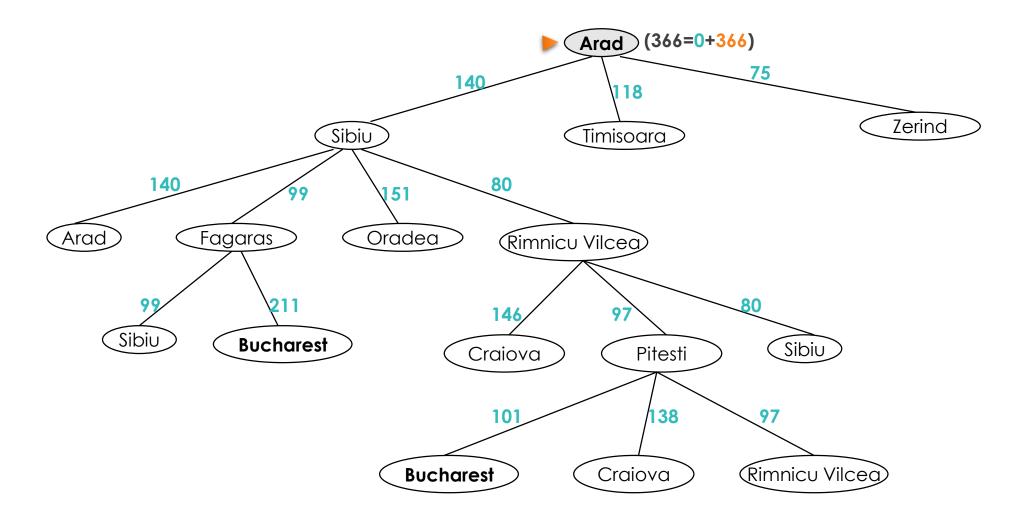
h(n) = Straight-line distance
from each city to Bucharest

City	Distance	City	Distance
Arad	366	Mehadia	241
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Craiova	160	Oradea	380
Dobreta	242	Pitesti	100
Eforie	161	Rimnicu Vilcea	193
Fagaras	176	Sibiu	253
Giurgiu	77	Timisoara	329
Hirsova	151	Urziceni	80
lasi	226	Vaslui	199
Lugoj	244	Zerind	374

g(n) = Actual path distance
between different cities

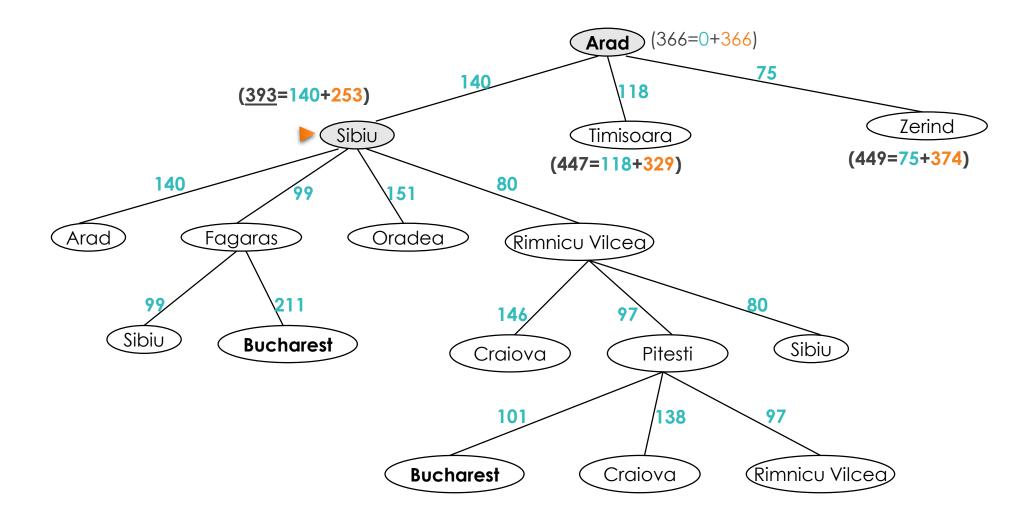






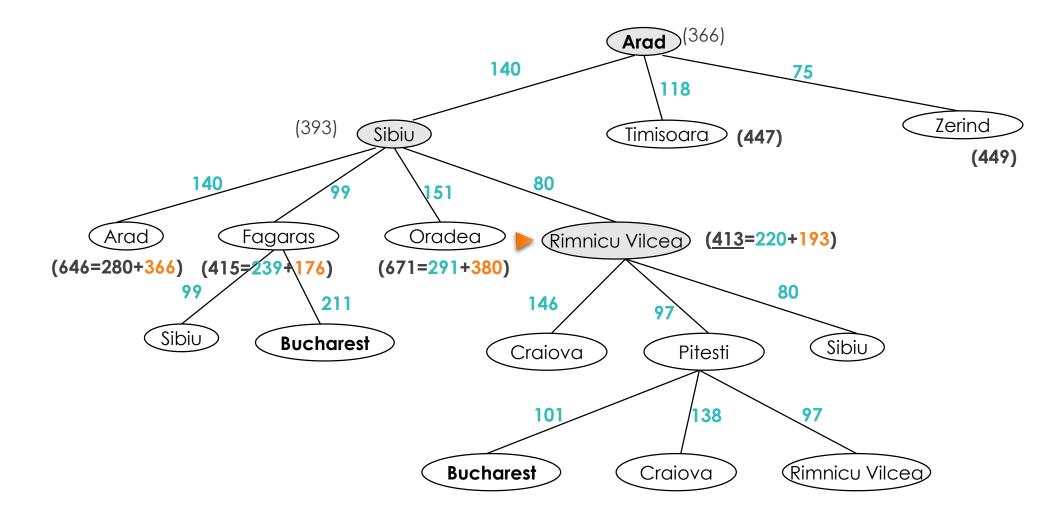






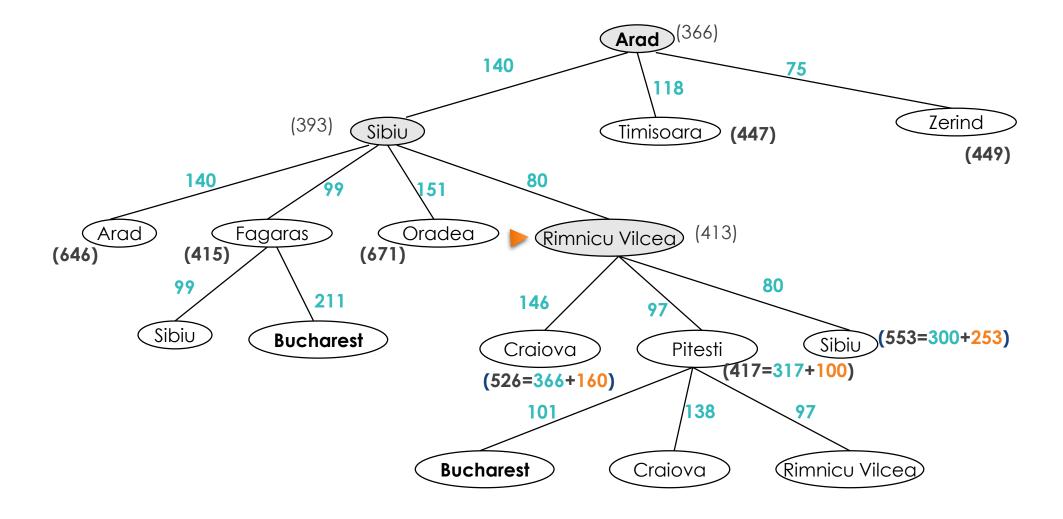






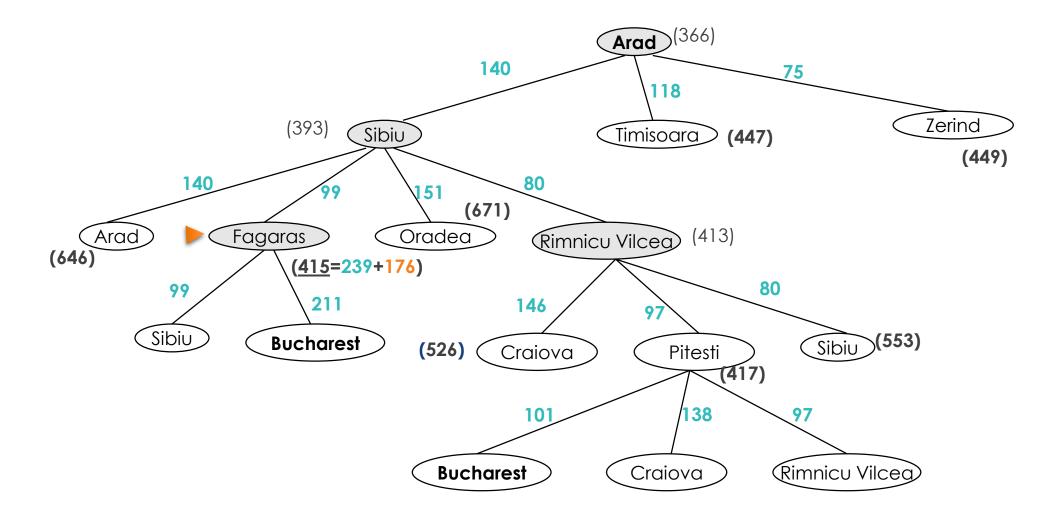






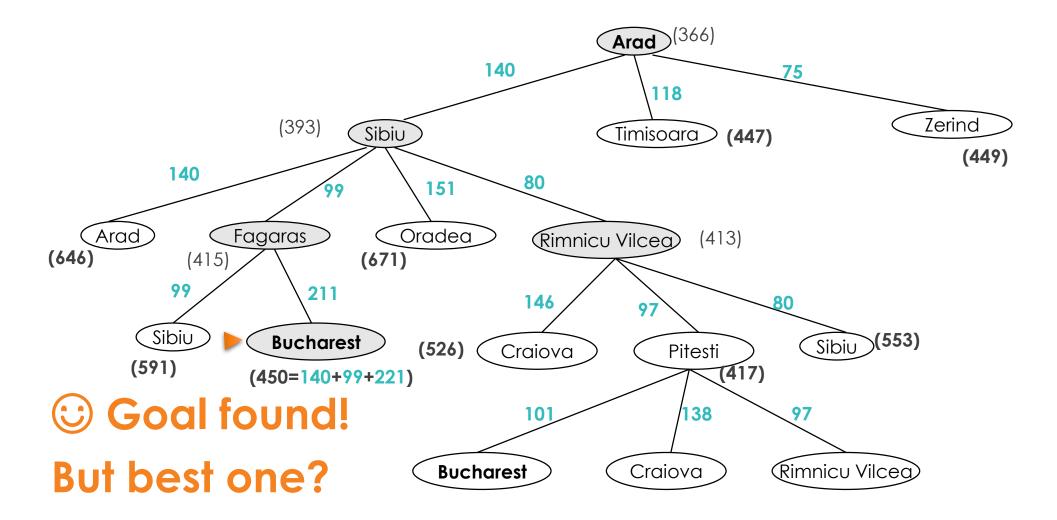






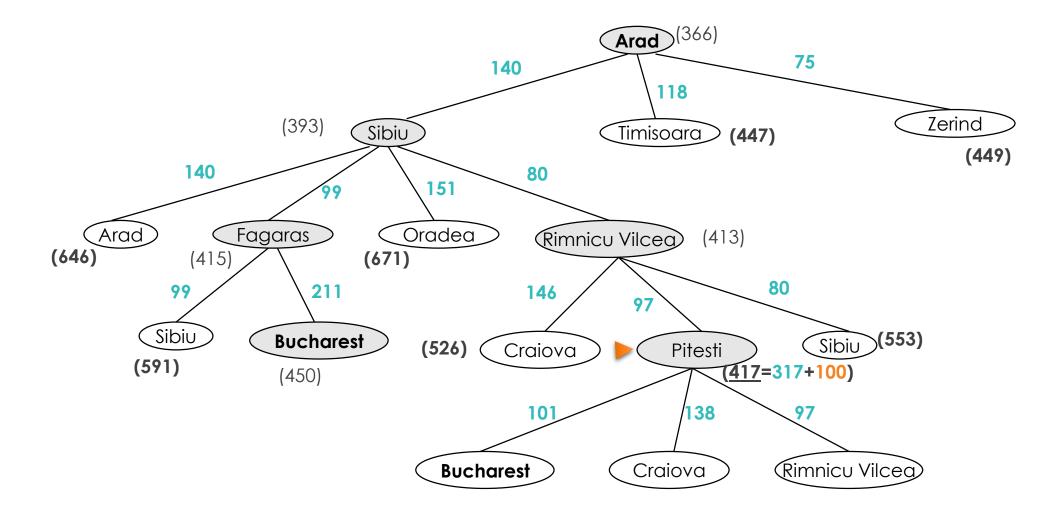






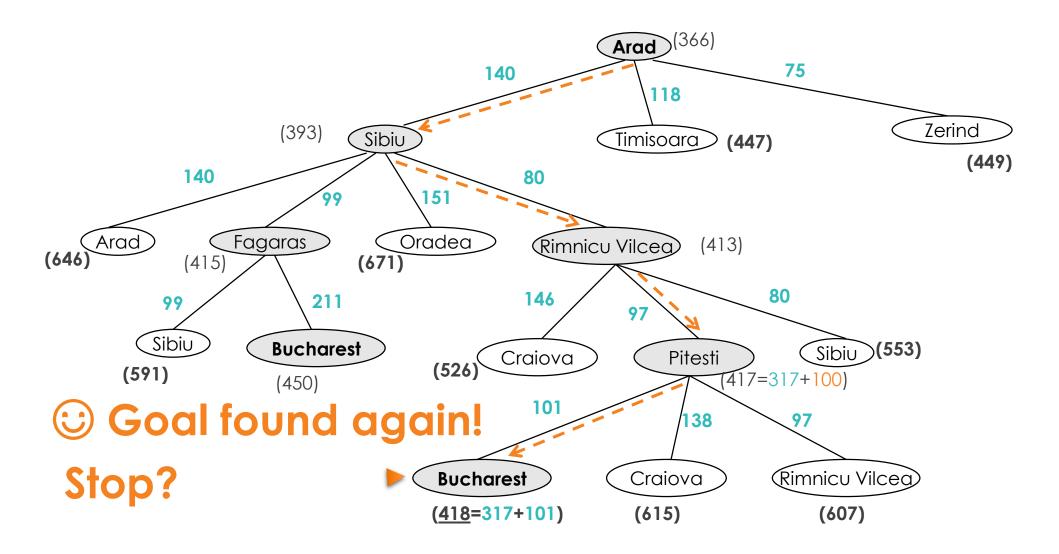












# 1.3 INFORMED SEARCH TECHNIQUES (1/2) A\* Search vs. Hill Climbing





#### **Exercise**

Compare the results between A\* Search & Hill Climbing.

Are they the same? If not, why?









# Search Modelling & Representation

- Pen & Paper Planning
- Robot Navigation

# KIE OptaPlanner Tutorial

- Optimizing Vehicle Route Planning (VRP)
- Optimizing Europe Travelling Sales Person (TSP)

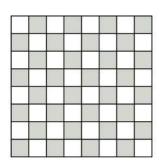


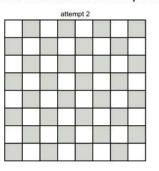


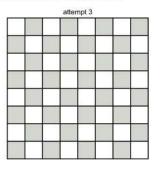
**Search Modelling & Representation** 

# Pen & Paper Planning

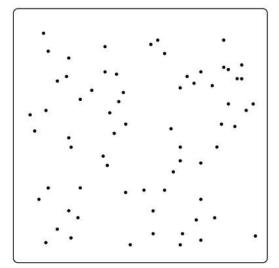
1) Place 8 queens on this chessboard so no 2 queens can attack each other.

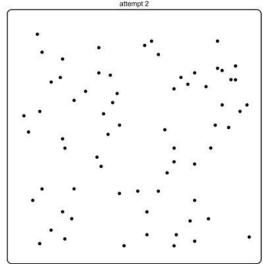


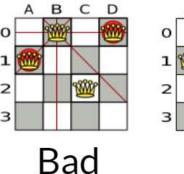




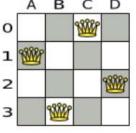
2) Draw the shortest line that connects all dots and returns to its origin.











Good

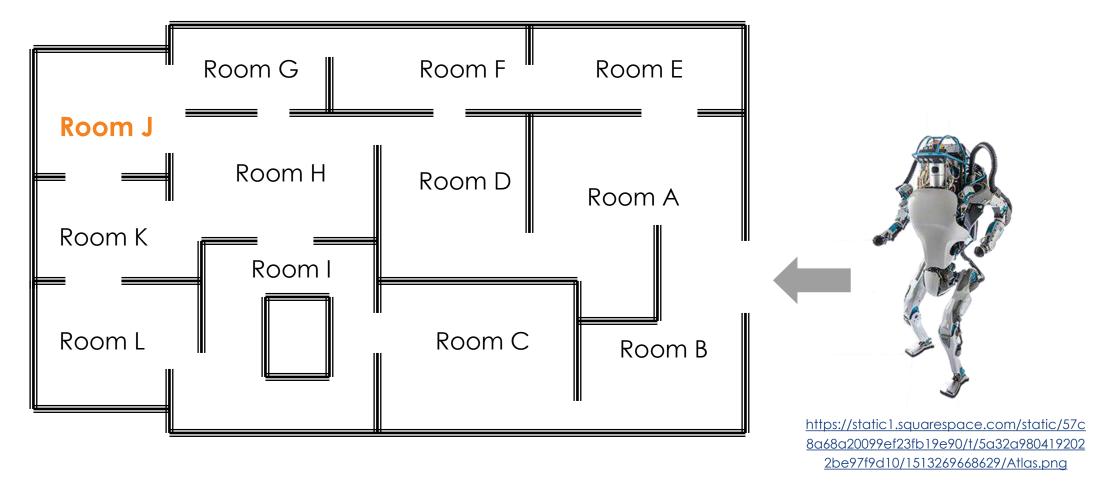






**Search Modelling & Representation** 

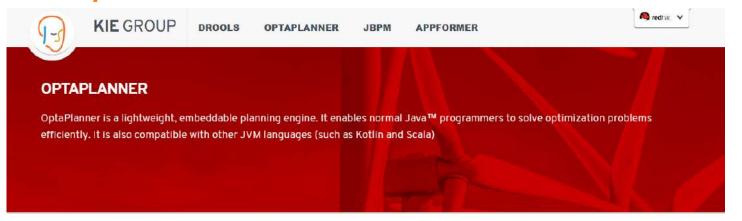
Robotics: How to rapidly navigate to Room J?







#### **KIE OptaPlanner Tutorial**



JBoss KIE

http://www.kiegroup.org/

#### DROOLS

Drools is a business rule management system with a forwardchaining and backward-chaining inference based rules engine, allowing fast and reliable evaluation of business rules and complex event processing.

Read more →

#### **OPTAPLANNER**

OptaPlanner is a constraint solver that optimizes use cases such as employee rostering, vehicle routing, task assignment and cloud optimization.

Read more →

#### JBoss KIE OptaPlanner

http://www.optaplanner.org/

#### **JBPM**

jBPM is a flexible Business Process Management suite allowing you to model your business goals by describing the steps that need to be executed to achieve those goals.

Read more --

#### **APPFORMER**

AppFormer is a low code platform to develop modern applications. It's a powerful tool for developers that can easily build applications by mashing up components and connect them to other Red Hat modules and software.

We make building apps looks easy.

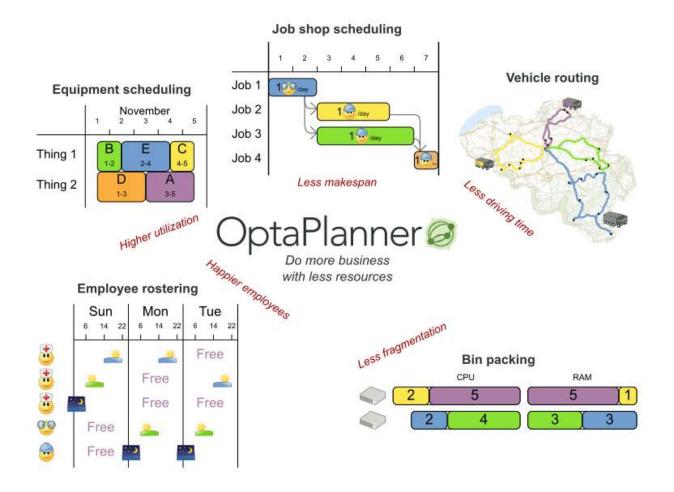
Read more →





**KIE OptaPlanner Tutorial** 

# Constrain Satisfaction: Business Resource Optimizer

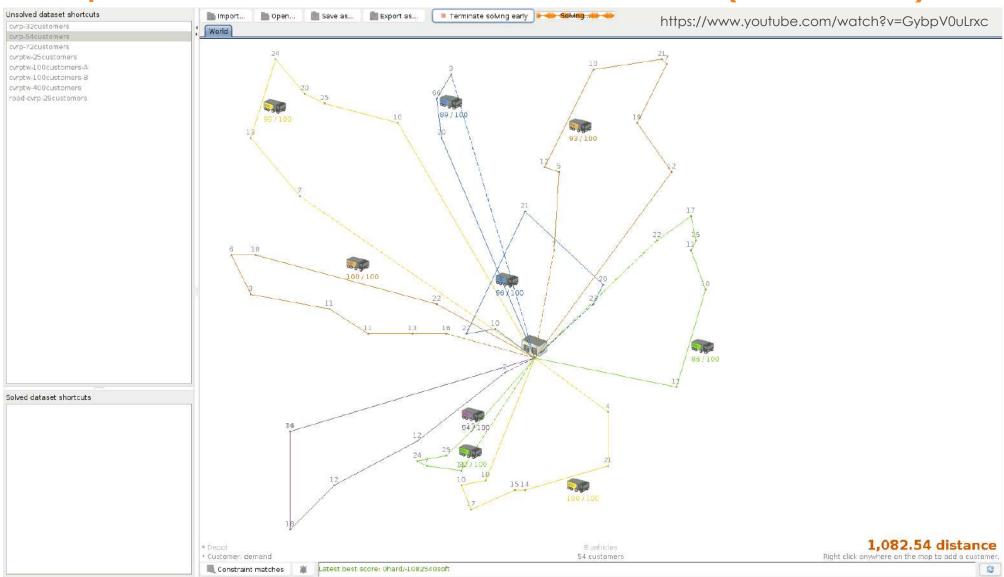








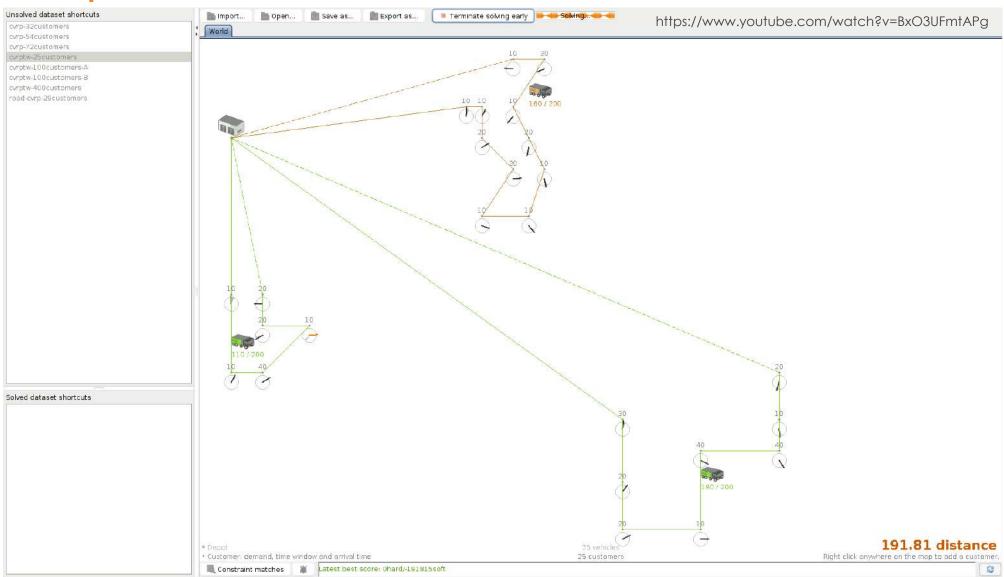
#### KIE OptaPlanner Tutorial – VRP: Customer demand (vehicle load)







#### KIE OptaPlanner Tutorial – VRP: Customer demand, Time window

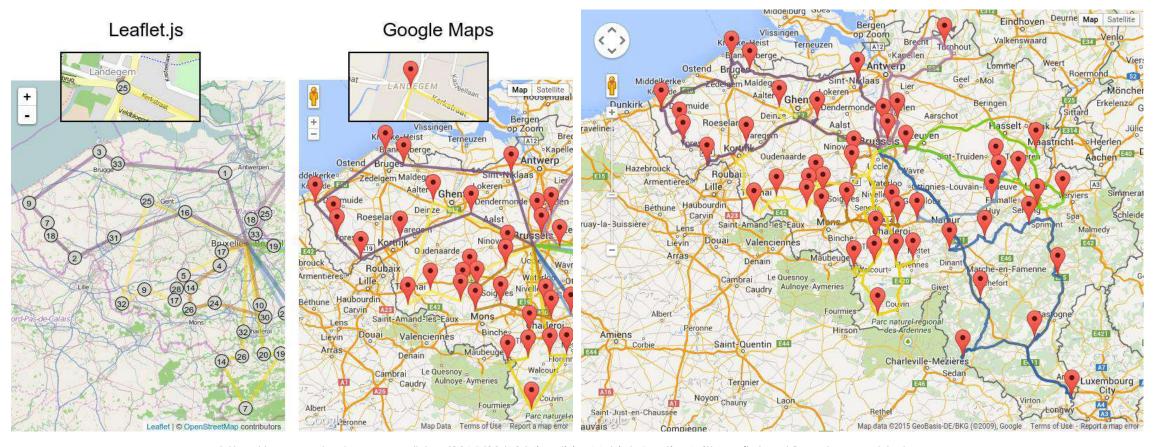






KIE OptaPlanner Tutorial – VRP with map integration

# Visualizing Vehicle Routing with Leaflet and Google Maps



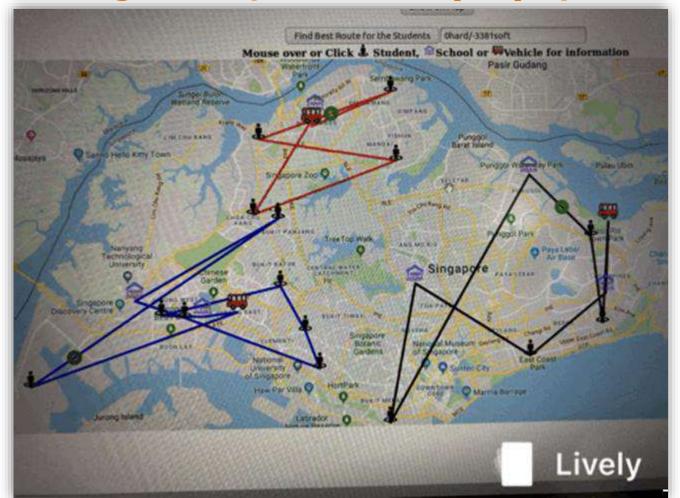
https://www.optaplanner.org/blog/2015/03/10/VisualizingVehicleRoutingWithLeafletAndGoogleMaps.html





#### Past project – VRP with map integration

# Intelligent Rapid Shuttle (IRS) System



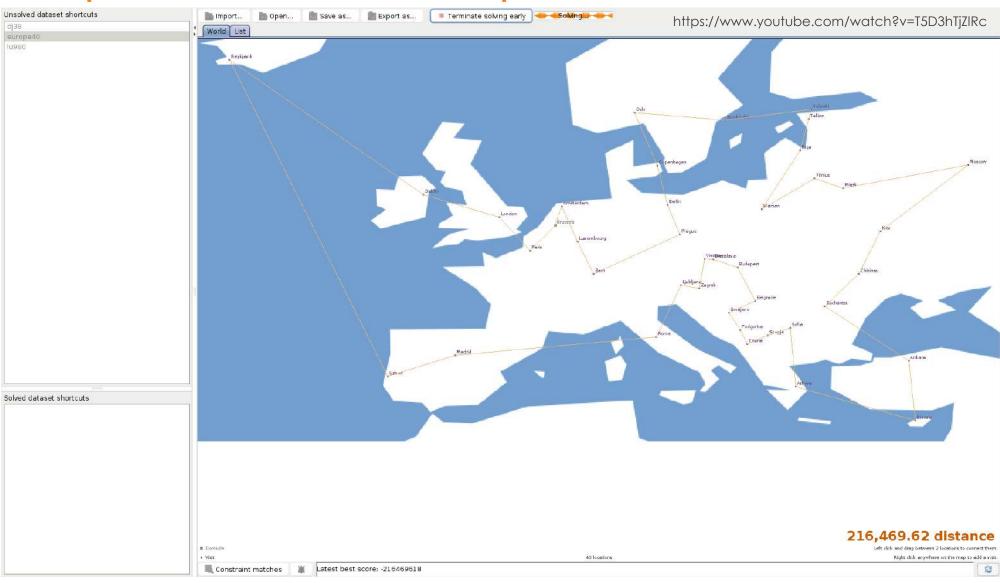


<u>Source</u> https://github.com/IRS-RS/IRS-RS-2019-03-09-IS1PT-GRP-aiVoyagers-irs-Intelligent-Rapid-Shuttle





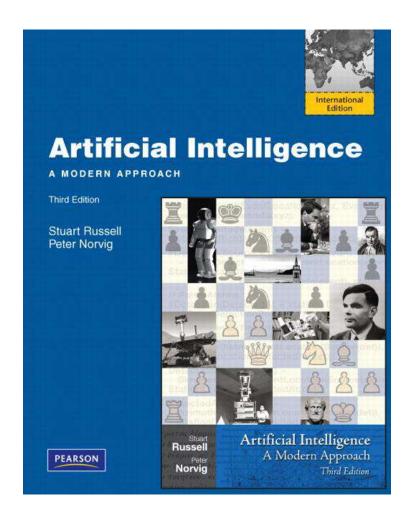
#### KIE OptaPlanner Tutorial – TSP: Europe cities



#### DAY 1 REFERENCE







1. OptaPlanner: Do more business with less recourses

http://www.optaplanner.org/learn/slides/optaplanner-presentation/index.html#/1

- OptaPlanner
   https://www.optaplanner.org/
- 3. OptaPlanner Use Cases & Demo Videos

  http://www.optaplanner.org/learn/useCases/index.html
- 4. OptaPlanner Video Tutorials

  <a href="http://www.optaplanner.org/learn/video.html">http://www.optaplanner.org/learn/video.html</a>
  <a href="https://www.youtube.com/user/ge0ffrey2">https://www.youtube.com/user/ge0ffrey2</a>
- Onne Beek. (2011). Efficient Local Search Methods For Vehicle Routing

https://lib.ugent.be/fulltxt/RUG01/001/788/544/RUG01-001788544\_2012\_0001\_AC.pdf

#### **DAY 1 SUMMARY**





#### 1.1 Reasoning Systems Overview

- History of Artificial Intelligence
- Question Answering System; Image Object Recognition; Chat-Bot
- Problem Solving: Analytic Tasks vs. Synthetic Tasks
- Synthetic Techniques: Search; Simulations; Genetic Algorithms; Data Mining

#### 1.2 Uninformed Search Techniques

- Search Representation
- Depth First Search (DFS)
- Breadth First Search (BFS)

#### 1.3 Informed Search Techniques (1/2)

- Heuristic Knowledge
- Evaluation (Scoring) Function; Heuristic Function; Past Cost Function
- Hill Climbing Search (HC); A Star Search (A\*)

#### 1.4 Search Representation Workshop





# **END OF LECTURE NOTES**





# **APPENDICES**





# **KIE System Architecture**

# KIE functionality overview

What are the KIE projects?



Rule engine and Complex Event Processing

Example: insurance rate calculation



Planning engine and optimization solver

Example: employee rostering

Drools Workbench

Design rules, decision tables, ... Drools Execution Server

REST/JMS service for business rules

OptaPlanner Workbench

Design solvers, benchmarks, ... OptaPlanner Execution Server

REST/JMS service for optimization





Workflow engine

Example: mortgage approval process

jBPM Workbench

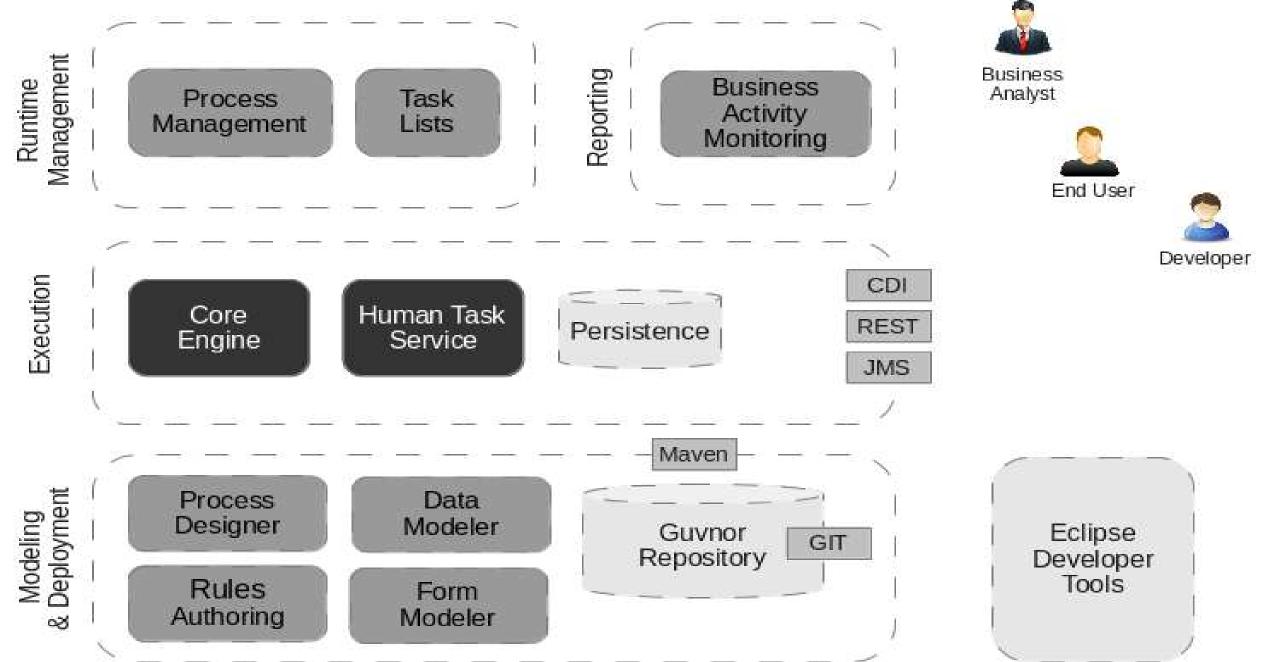
Design workflows, forms, ... jBPM Execution Server

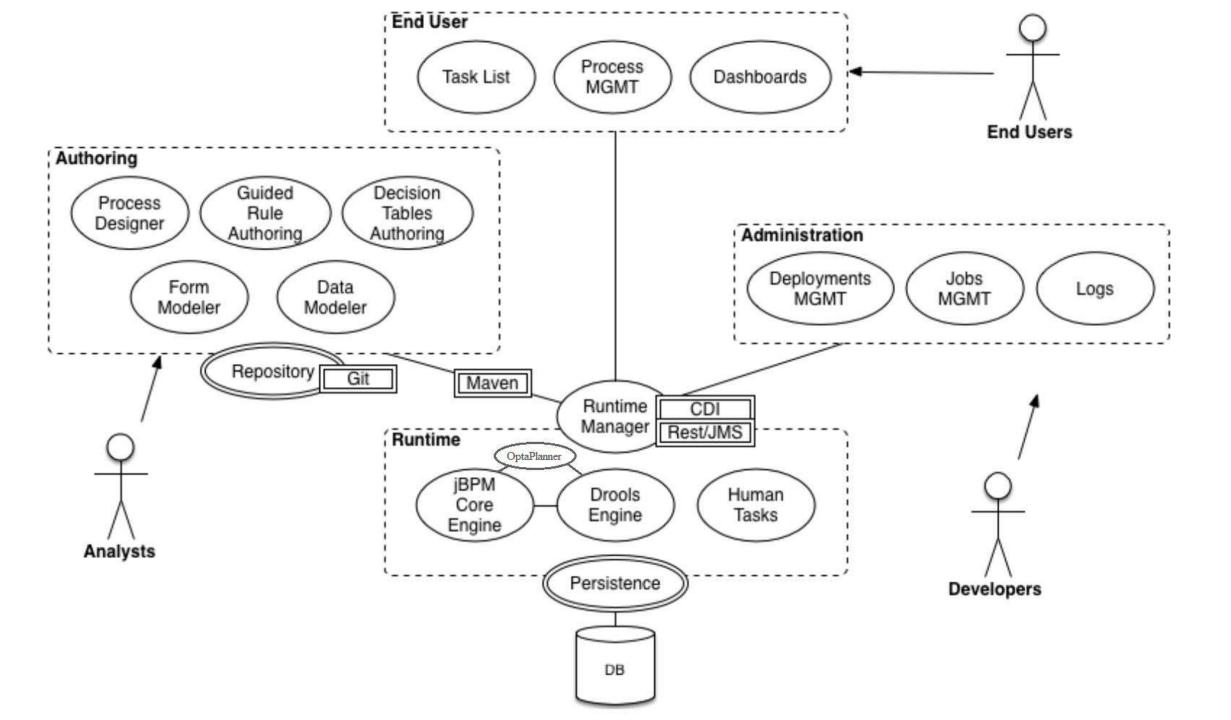
REST/JMS service for workflows

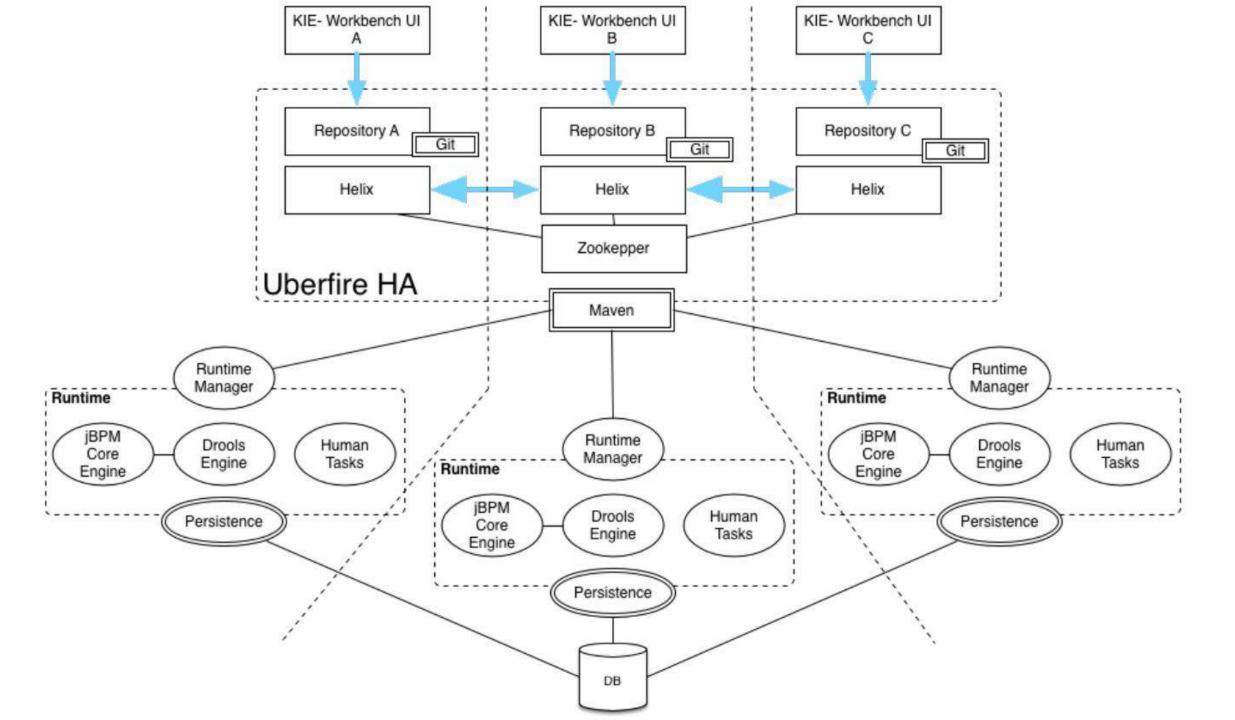


Lightweight, embeddable engines (jars) which run in a Java VM

Web applications (wars) which run on a Java Application Server











# **END OF APPENDICES**