

## PROJECT PRESENTATION # 3 - Planning and Search

**Notebook:** Artificial Intelligence NanoDegree

**Created:** 10/16/2017 8:52 AM

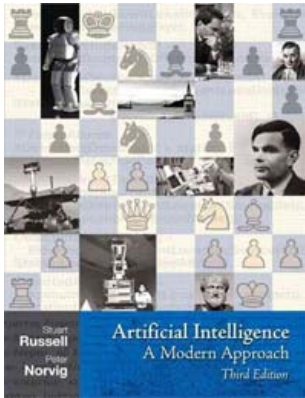
**Updated:** 10/17/2017 9:06 AM

**Author:** Matthew R. Versaggi

**URL:** <https://stripsfiddle.herokuapp.com/>

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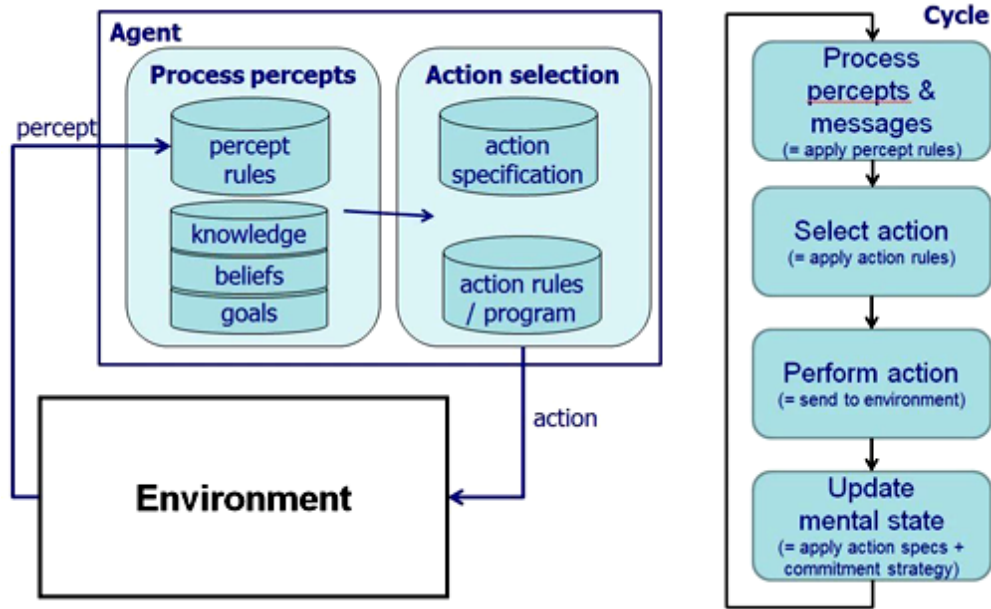
## PROJECT PRESENTATION # 3 - Planning and Search



### Overview of the book

The **main unifying theme** is the idea of an **intelligent agent**. We define AI as the study of agents that receive percepts from the environment and perform actions.

**GOAL:** Cognitive Intelligent Agents



**SITE:**

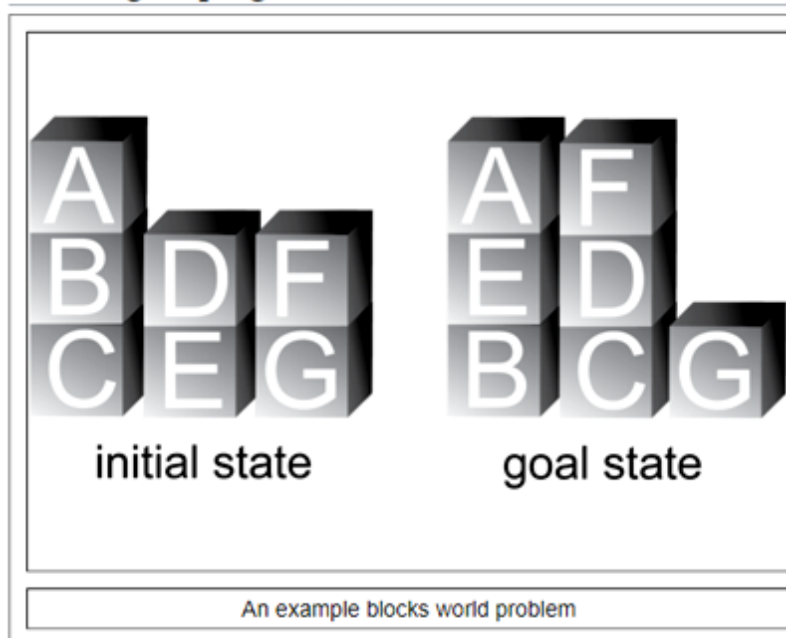
<https://goalapl.atlassian.net/wiki/spaces/GOAL/overview?mode=global>

**Wiki:**

[https://en.wikipedia.org/wiki/GOAL\\_agent\\_programming\\_language](https://en.wikipedia.org/wiki/GOAL_agent_programming_language)

- **Planning is central to Intelligent Agents**

GOAL agent program [ edit ]



## Part 1 - Planning problems

### THEORY:

<https://medium.com/towards-data-science/ai-planning-historical-developments-edcd9f24c991>

Example of applying the **STRIPS** language to an **Air Cargo transport system** using a planning search agent. Suppose we have an initial state of Cargo 1 at SFO, Cargo 2 at JFK, Plane 1 at SFO and Plane 2 at JFK.

We want to formulate an optimal plan to transport Cargo 1 to JFK and Cargo 2 to SFO. Summarizing this problem description, we have:

```
Init(At(C1, SFO) ∧ At(C2, JFK)
    ∧ At(P1, SFO) ∧ At(P2, JFK)
    ∧ Cargo(C1) ∧ Cargo(C2)
    ∧ Plane(P1) ∧ Plane(P2)
    ∧ Airport(JFK) ∧ Airport(SFO))

Goal(At(C1, JFK) ∧ At(C2, SFO))
```

### PYTHON:

```
def air_cargo_p1() -> AirCargoProblem:
    cargos = ['C1', 'C2']
    planes = ['P1', 'P2']
    airports = ['JFK', 'SFO']
    pos = [expr('At(C1, SFO)'),
            expr('At(C2, JFK)'),
            expr('At(P1, SFO)'),
            expr('At(P2, JFK)'),
            ]
    neg = [expr('At(C2, SFO)'),
            expr('In(C2, P1)'),
            expr('In(C2, P2)'),
            expr('At(C1, JFK)'),
            expr('In(C1, P1)'),
            expr('In(C1, P2)'),
            expr('At(P1, JFK)'),
            expr('At(P2, SFO)'),
            ]
    init = FluentState(pos, neg)
    goal = [expr('At(C1, JFK)'),
            expr('At(C2, SFO)'),
            ]
    return AirCargoProblem(cargos, planes, airports, init, goal)
```

## PDDL and STRIPS Fiddle URL:

<https://stripsfiddle.herokuapp.com/>

## AIR CARGO PROBLEM: UDACITY Style

### Inspiration:

<https://medium.com/towards-data-science/ai-planning-historical-developments-edcd9f24c991>

### Udacity GitHub:

<https://github.com/udacity/AIND-Planning>

## Project Submission Result: (P1, S1)

```
> python run_search.py -p 1 -s 1
```

Solving Air Cargo Problem 1 using **breadth\_first\_search...**

Expansions	Goal Tests	New Nodes
43	56	180

Plan length: 6 Time elapsed in seconds: 0.03145325632251838

1. Load(C1, P1, SFO)
2. Load(C2, P2, JFK)
3. Fly(P2, JFK, SFO)
4. Unload(C2, P2, SFO)
5. Fly(P1, SFO, JFK)
6. Unload(C1, P1, JFK)

## FIDDLE Air Cargo Planner Result:

### Solution found in 6 steps!

1. LOAD C1 P1 SFO
2. LOAD C2 P2 JFK
3. FLY P1 SFO JFK
4. UNLOAD C1 P1 JFK
5. FLY P2 JFK SFO
6. UNLOAD C2 P2 SFO

## CODE REVIEW:

### MY\_AIR\_CARGO\_PROBLEMS.PY

**TODO:** Implement methods and functions in `my_air_cargo_problems.py`

- `AirCargoProblem.get_actions` method including `load_actions` and `unload_actions` sub-functions
- `AirCargoProblem.actions` method
- `AirCargoProblem.result` method
- `air_cargo_p2` function
- `air_cargo_p3` function

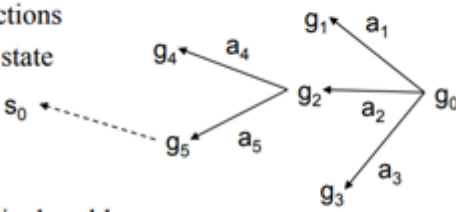
## Part 2 - Domain-independent heuristics

### PLANNING GRAPHS and HEURISTICS

#### THEORY:

##### Motivation:

#### Motivation

- A big source of inefficiency in search algorithms is the *branching factor*
    - ◆ the number of children of each node
  - e.g., a backward search may try lots of actions that can't be reached from the initial state
- 
- One way to reduce branching factor:
  - First create a *relaxed problem*
    - ◆ Remove some restrictions of the original problem
      - » Want the relaxed problem to be easy to solve (polynomial time)
    - ◆ The solutions to the relaxed problem will include all solutions to the original problem
  - Then do a modified version of the original search
    - ◆ Restrict its search space to include only those actions that occur in solutions to the relaxed problem

## Inspiration Links: - Graph Planning Techniques

- <http://pages.mtu.edu/~nilufer/classes/cs5811/2009-fall/lecture-slides/cs5811-ch11b-graphplan.pdf>
- <https://www.cs.umd.edu/~nau/planning/slides/chapter06.pdf>

## CODE REVIEW:

### MY\_PLANNING\_GRAPH.PY

**TODO:** Implement a Planning Graph with automatic heuristics in `my_planning_graph.py`

- `PlanningGraph.add_action_level` method
- `PlanningGraph.add_literal_level` method
- `PlanningGraph.inconsistent_effects_mutex` method
- `PlanningGraph.interference_mutex` method
- `PlanningGraph.competing_needs_mutex` method
- `PlanningGraph.negation_mutex` method
- `PlanningGraph.inconsistent_support_mutex` method
- `PlanningGraph.h_levelsum` method

## Part 3 - Heuristic Analysis

- See Heuristic Analysis PDF file.
- See Test Results Spreadsheet file.

## Part 4 - Research Review

[Tip: The book *Artificial intelligence: A Modern Approach* by Norvig and Russell is chock full of references in the Bibliographical and Historical notes at the end of Chapter 10.]

**AI Planning Historical Developments:**

<https://medium.com/towards-data-science/ai-planning-historical-developments-edcd9f24c991>

**Recent Advances in AI Planning:**

<https://homes.cs.washington.edu/~weld/papers/pi2.pdf>

**Progress in AI Planning Research and Applications:**

[https://www.researchgate.net/publication/242415929\\_Progress\\_in\\_AI\\_Planning\\_Research\\_and\\_Applications](https://www.researchgate.net/publication/242415929_Progress_in_AI_Planning_Research_and_Applications)

**An overview of recent algorithms for AI Planning:**

<http://www.cs.toronto.edu/~sheila/2542/w06/readings/RintanenHoffmann01.pdf>