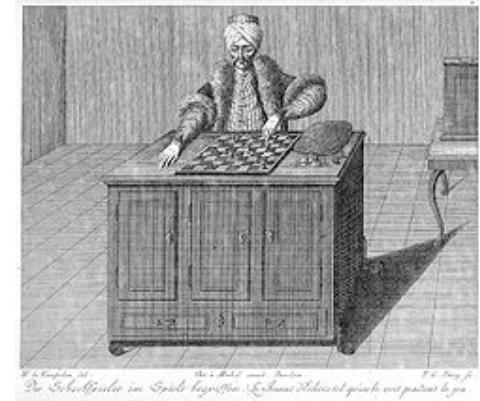


Artificial intelligence: a primer

Mayank Kejriwal

What is AI?

- Logicians:"
 - Can we define 'the laws of thought'? (Ancient Greece, also India, China)"
 - Can we automate the laws of thought (since the Industrial Revolution)
- Mechanical Turk (1770): Highly influential hoax ('automatic' chess player); arguably first hyped-up 'AI' that didn't work out
- Now, IBM's DeepBlue beat Kasparov in 1997, and our phones can beat us
 - Numerous other games 'conquered' (AlphaZero)



Other examples

Vacuum-cleaning robots (Roomba)

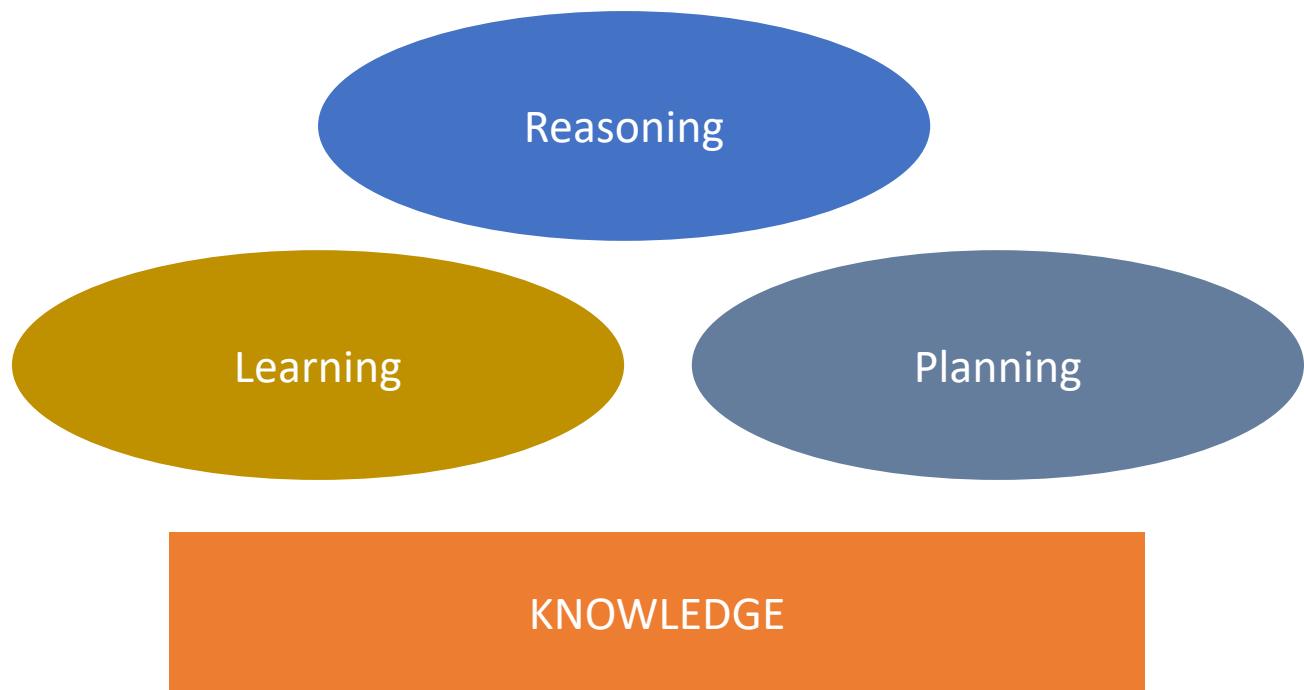
NASA's Mars exploration rovers

Autonomous vehicles

(EUREKA's Prometheus Project, DARPA's Grand Challenge, Google's Driverless car)

What is ‘intelligence’?

- We’ll take a computational perspective



AI ‘tasks’

Reasoning:

Solve sudoku; play a game of chess

Robotics:

Move towards a goal, avoiding obstacles

Natural language processing:

Understand/produce sentences

Computer vision:

Recognize faces in an image

Areas of AI

- Reasoning/problem solving
- Knowledge representation
- Machine learning
- Planning
- Computer perception
(vision, audio/speech)
- Natural language processing
- Robotics

Reasoning

Simple problem

- Modus Ponens:
 - Given (1) IF A THEN B, (2) A, we can conclude (3) B
 - (1) and (2) are premises, (3) is the conclusion
- Example:
 - If today is Monday, then it is the first day of the week
 - Today is Monday
 - Modus Ponens allows us to conclude that it is the first day of the week

Is it really that obvious?

Either Shakespeare or Hobbes wrote *Hamlet*.

If either Shakespeare or Hobbes wrote *Hamlet*, then if Shakespeare didn't do it, Hobbes did.

Therefore, if Shakespeare didn't write *Hamlet*, Hobbes did it.

What if we make it probabilistic?

What if we are given B, and not A (*abductive reasoning*)?

What other 'laws' can we work with?

Learning

Mainly taken to mean machine learning

- Supervised machine learning
 - Classification problems (we'll see much more on this on Wed.)
- Unsupervised machine learning
 - Usually taken to mean clustering
 - We just saw an example with K-means!
- Reinforcement learning
 - Considered its own special branch of machine learning
 - Typically applicable to gameplaying, robotics and any application that lies at the intersection of control and AI



AlphaZero: Shedding new light on chess, shogi, and Go

06 DEC 2018



How research funders profit
from hidden investments p. 200

New books for budding
scientists p. 204

Drug leads for malaria
pp. 2122 & 2129

Science

\$15
7 December 2018
science.org
AAAS

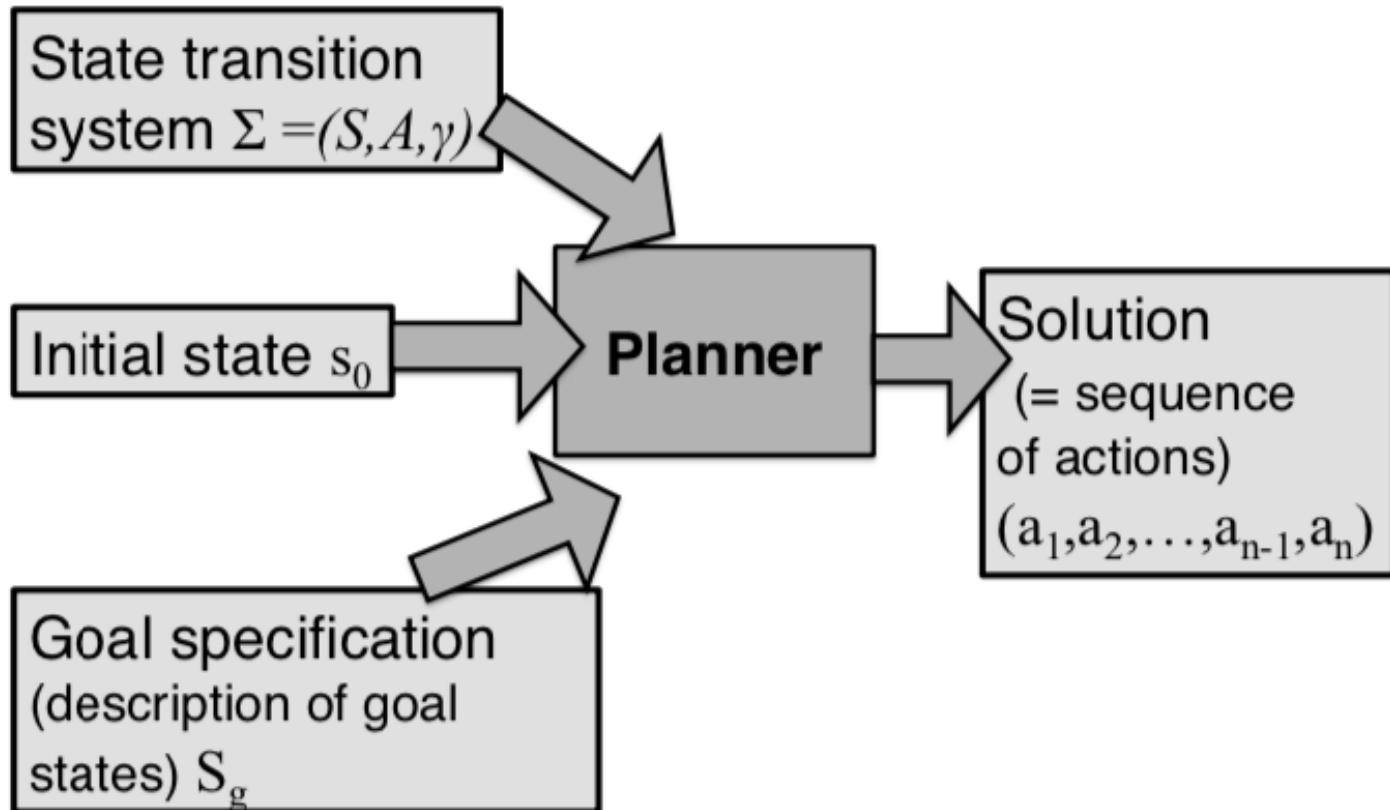


MATTHEW SADLER & NATASHA REGAN

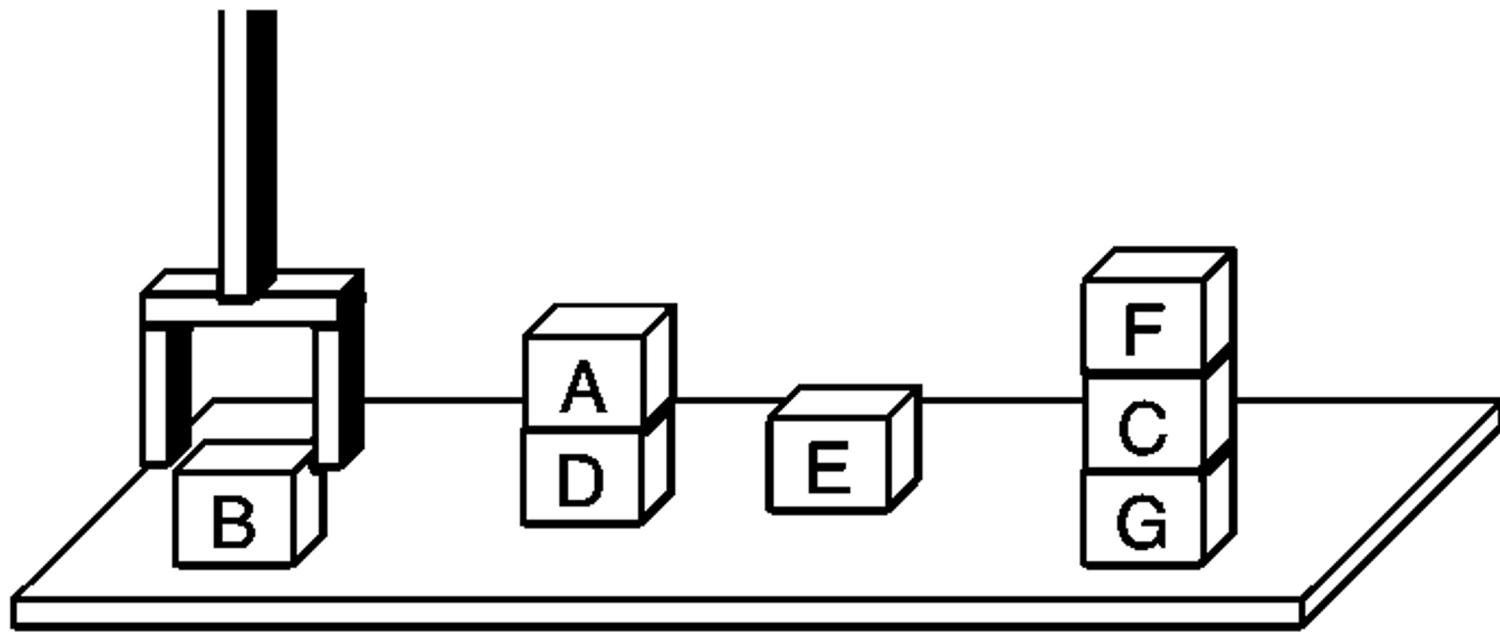
GAME CHANGER

Planning

Classical planning



EXAMPLE: BLOCKS WORLD



Goals:

- Build a tower of A,B,C,...
- Get block G,
-

Silly domain, but concisely illustrates many general planning issues

Planning as Theorem Proving

Planning: given the initial state, find a sequence of actions that yield a situation where goal holds.
= Use operators to *derive* goal from initial state.

Operator: carry(x)

General knowledge of one kind of action:
preconditions and effects

Action: carry(BlockA)

Ground instance of an operator

Key questions and challenges

How do we represent states?

- What information do we need to know?
- What information can we (safely) ignore?

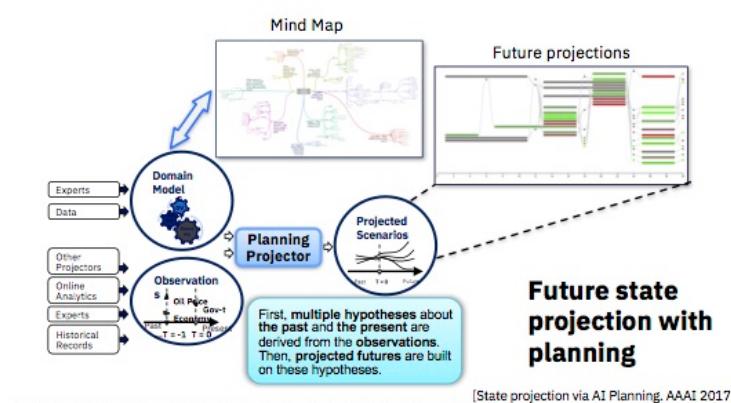
How do we represent actions?

- When can we perform an action?
- What changes when we perform an action?
- What stays the same?
- What level of detail do we care about?

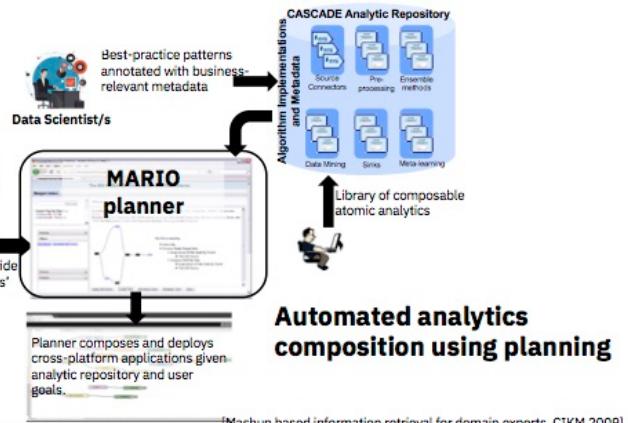
Planning: Summary

- Planning is the task of finding a procedural course of action for a declaratively described system to reach its **goals** while **optimizing overall performance measures**
- General features of planning problems:
 - Declarative
 - You want to find a procedural course of action for a declaratively described system to reach its goals while optimizing overall performance measures.
 - Domain Knowledge can be elicited or learned over time
 - Existing domain knowledge can/should be exploited for building the model
 - Human involvement controllable. Humans build the model and can contribute to the solution by introducing knowledge.
 - Favor consistency over learning transient behaviors
 - There is a structure of the problem that cannot be learned just training
 - When no large training data is available
 - Changes in the problem can make previous data irrelevant

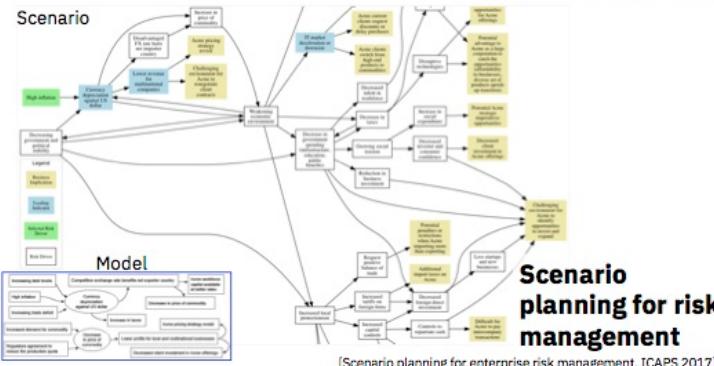
Example planning projects in IBM



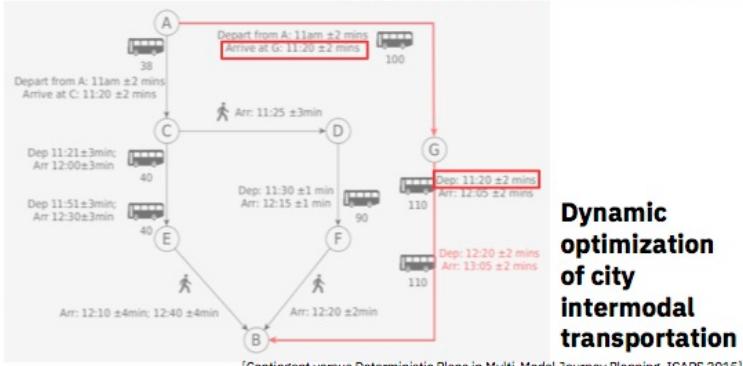
Future state projection with planning



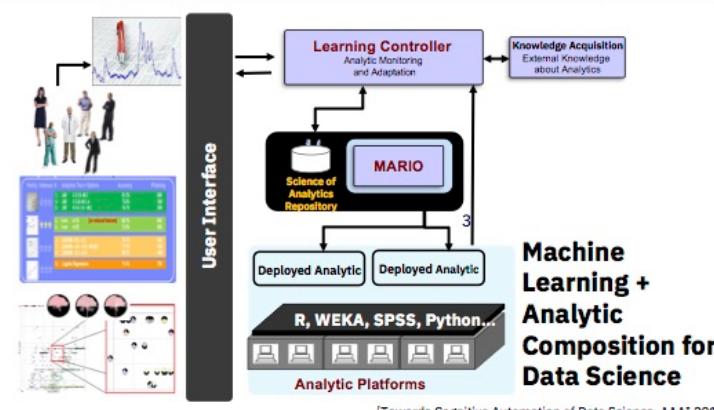
Automated analytics composition using planning



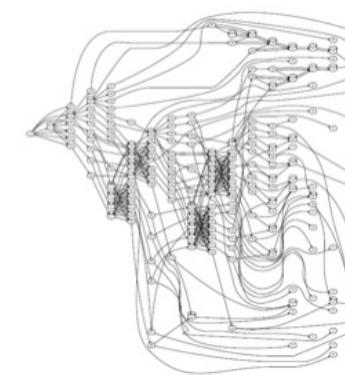
Scenario planning for risk management



Dynamic optimization of city intermodal transportation



Machine Learning + Analytic Composition for Data Science



Planners and tools developed at IBM

- Top-K planners
- Semi-black box for transportation
- Planner4J
- DOCIT (multi-modal transportation)
- Plan recognition as planning