

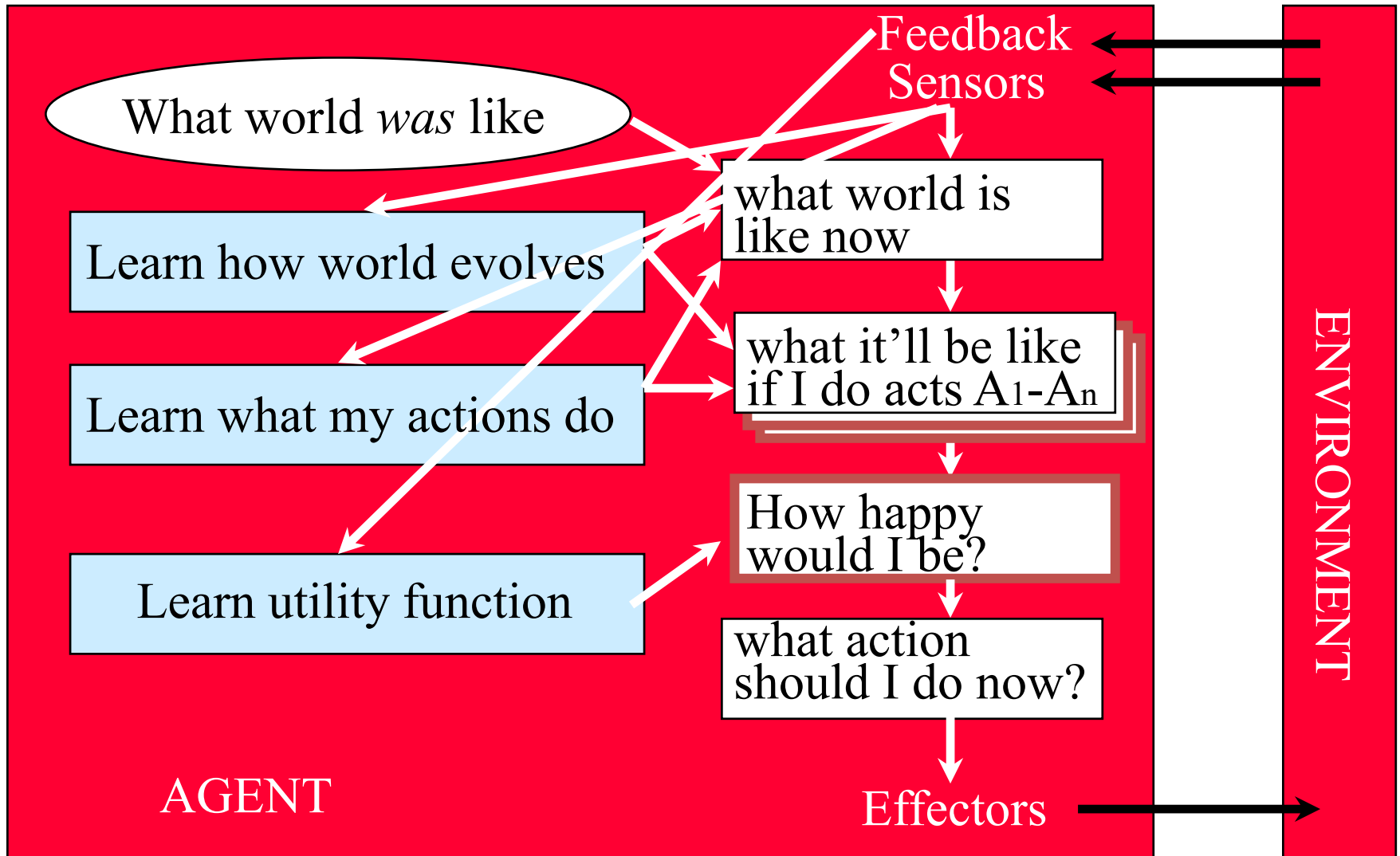
Artificial Intelligence Recap

Mausam

What is intelligence?

- (bounded) Rationality
 - We have a performance measure to optimize
 - Given our state of knowledge
 - Choose optimal action
 - Given limited computational resources
- Human-like intelligence/behavior

Learning agents



Search in Discrete State Spaces

- This is different from Web Search 😊
- Every discrete problem can be cast as a search problem.
- (states, actions, transitions, cost, goal-test)
- Types

– **uninformed systematic:** often slow

- DFS, BFS, uniform-cost, iterative deepening

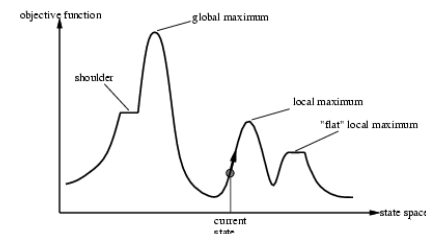
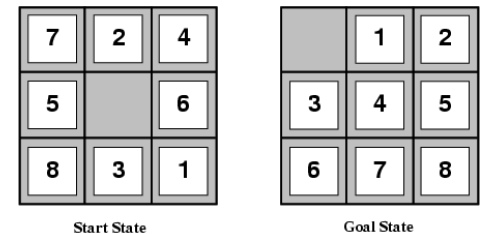
– **Heuristic-guided:** better

- Greedy best first, A*
- relaxation leads to heuristics

– **Local:** fast, fewer guarantees; often local optimal

- Hill climbing and variations
- Simulated Annealing: global optimal
- Genetic algorithms: somewhat non-local due to crossing over

– (Local) Beam Search



Search Example: Game Playing

- Game Playing
 - AND/OR search space (max, min)
 - minimax objective function
 - minimax algorithm (~dfs)
 - alpha-beta pruning
 - Utility function for partial search
 - Learning utility functions by playing with itself
 - Openings/Endgame databases
 - Secondary search/Quiescence search



Knowledge Representation and Reasoning

- Representing: what I know
- Reasoning: what I can infer

- Logic

- CSPs

- Bayes Nets

KR&R Example: Propositional Logic

- **Representation:** Propositional Logic Formula
 - CNF, Horn Clause,...
- **Reasoning:** Deduction
 - Forward Chaining
 - Resolution
- Model Finding
 - Enumeration
 - SAT Solving

Search+KR&R Example: CSP

- **Representation**

- Variables, Domains, Constraints

- **Reasoning:** Constraint Propagation

- Node consistency, Arc Consistency, k-Consistency

- **Search**

- Backtracking search: partial var assignments

- Heuristics for choosing which var/value next

- Local search: complete var assignments

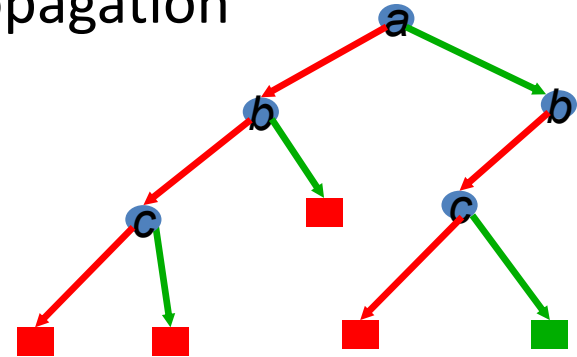
- Tree structured CSPs: polynomial time

- Cutsets: vars assigned \rightarrow converts to Tree CSP



Search+KR&R Example: SAT Solving

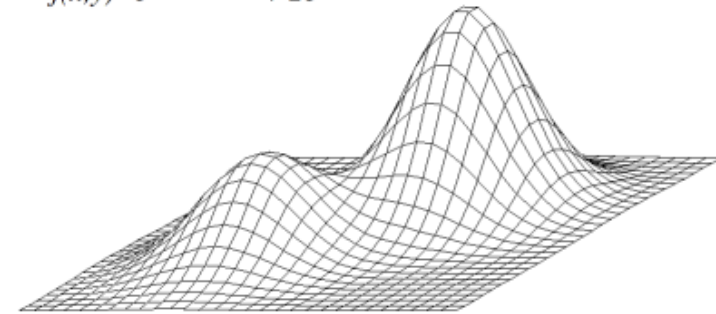
- **Representation:** CNF Formula
- **Reasoning**
 - pure literals; unit clauses; unit propagation
- **Search**
 - DPLL (~ backtracking search)
 - MOM's heuristic
 - Local: GSAT, WalkSAT
- **Advances**
 - Clause Learning: learning from mistakes
 - Restarts in systematic search
 - Portfolio of SAT solvers; Parameter tuning
- Phase Transitions in SAT problems



KR&R Part 2: Continuous Spaces

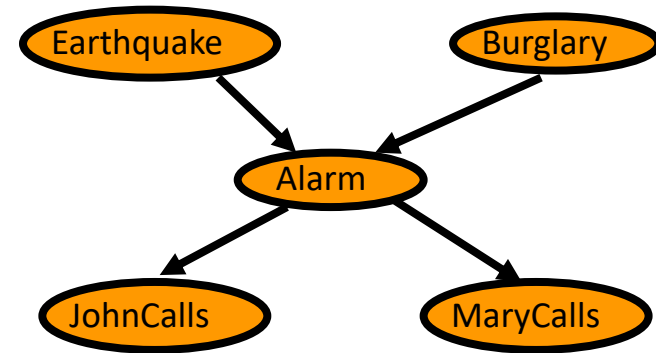
$$f(x,y) = e^{-(x^2+y^2)} + 2e^{-((x-1.7)^2+(y-1.7)^2)}$$

- Search
 - Gradient Descent
 - Newton Raphson
 - Optimization (convex/non-convex...)
- Constraint Optimization (we didn't study this)
 - Linear Programming
 - Integer Linear Programming
 - Mixed Integer Linear Programming



KR&R: Probability

- **Representation:** Bayesian Networks
 - encode probability distributions compactly
 - by exploiting conditional independences



- **Reasoning**
 - Exact inference: var elimination
 - Approx inference: sampling based methods
 - rejection sampling, likelihood weighting, Gibbs sampling

KR&R: One-step Decision Theory

- **Representation**

- actions, probabilistic outcomes, rewards

- **Reasoning**

- expected value/regret of action

- Expected value of perfect information

| Actions | States of Nature | |
|--------------------|-------------------------|---------------------------|
| | Favorable Market | Unfavorable Market |
| Large plant | \$200,000 | -\$180,000 |
| Small plant | \$100,000 | -\$20,000 |
| No plant | \$0 | \$0 |

- **Non-deterministic uncertainty**

- Maximax, maximin, eq likelihood, minimax regret..

- **Utility theory: value of money...**

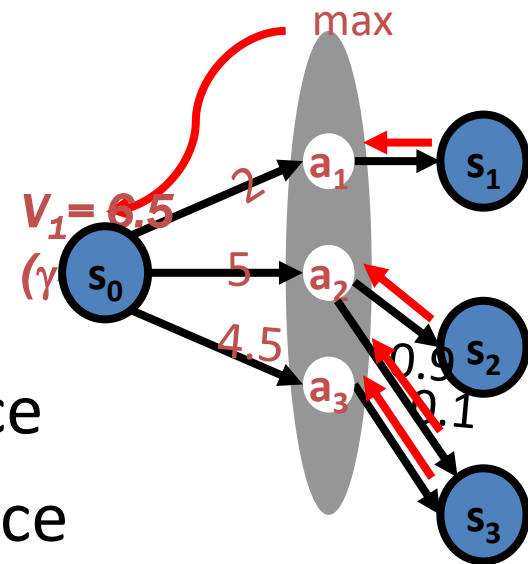
KR&R: Markov Decision Process

- Representation

- states, actions, probabilistic outcomes, rewards
- ~AND/OR Graph (sum, max)

- Reasoning: $V^*(s)$

- Value Iteration: search thru value space
- Policy Iteration: search thru policy space



Learning: BNs/NB

- ML estimation. $\max P(D | \theta)$
 - counting; smoothing
- MAP estimation $\max P(\theta | D)$..
 - Gradient descent

Learning: Neural Nets

- Representation

- Network of weighted sum + non-linearities

- Reasoning

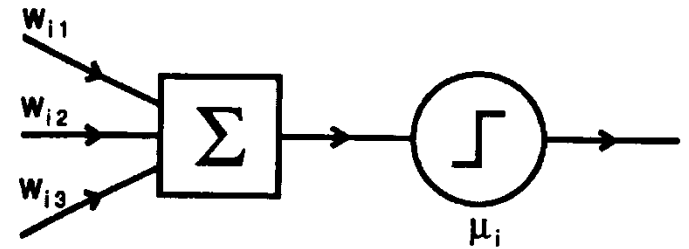
- Forward pass

- Learning

- Gradient descent

- Backpropagation through network

- Advantage: feature learning



Learning: Reinforcement Learning

- Learn model while taking actions
- What to learn
 - T and R: model based
 - Policy: Model free
- Which actions to take
 - Exploration – Exploitation
- Large state spaces
 - function approximation

AI we didn't cover

- Temporal models: HMMs, Kalman filters...
- Ontologies
- Robotics
- Vision
- Mechanism design
- Multi-agent systems
- Sensor Networks
- Computational Neuroscience
- ...

AI is about problems.

- It is an application-driven field
- Happy to beg, borrow, steal ideas from anywhere
- Traditionally discrete ... more and more cont.
- Traditionally logic... almost all probability
 - Recent close connections with EE/Stat due to ML
- HUGE field

Applications of AI

- Mars rover: planning
- Jeopardy: NLP, info retrieval, machine learning
- Puzzles: search, CSP, logic
- Chess: search
- Web search: IR
- Text categorization: machine learning
- Self-driving cars: robotics, prob. reasoning, ML...

AI-Centric World 😊

Graphics

Algorithms
Theory

Databases

Operations
Research

AI

Statistics

Linguistics

Robot
Design

Psychology
Neurosc.

...