

# Department of Computer Science

## COLLEGE OF SCIENCE AND MATHEMATICS

```

38 int Move(string& board, int dir){
39     int row, col;
40     LocBlank(board, row, col);
41     if ((dir == 1 || dir == -3) && row > 0){
42         Swap(board, row, col, row-1, col);
43     }
44     else if ((dir == 2 || dir == -4) && col < MAXDIM-1){
45         Swap(board, row, col, row, col+1);
46     }
47     else if ((dir == 3 || dir == -1) && row < MAXDIM-1){
48         Swap(board, row, col, row+1, col);
49     }
50     else if ((dir == 4 || dir == -2) && col > 0){
51         Swap(board, row, col, row, col-1);
52     }
53     else {
54         return -1;
55     }
56     return 0;

```



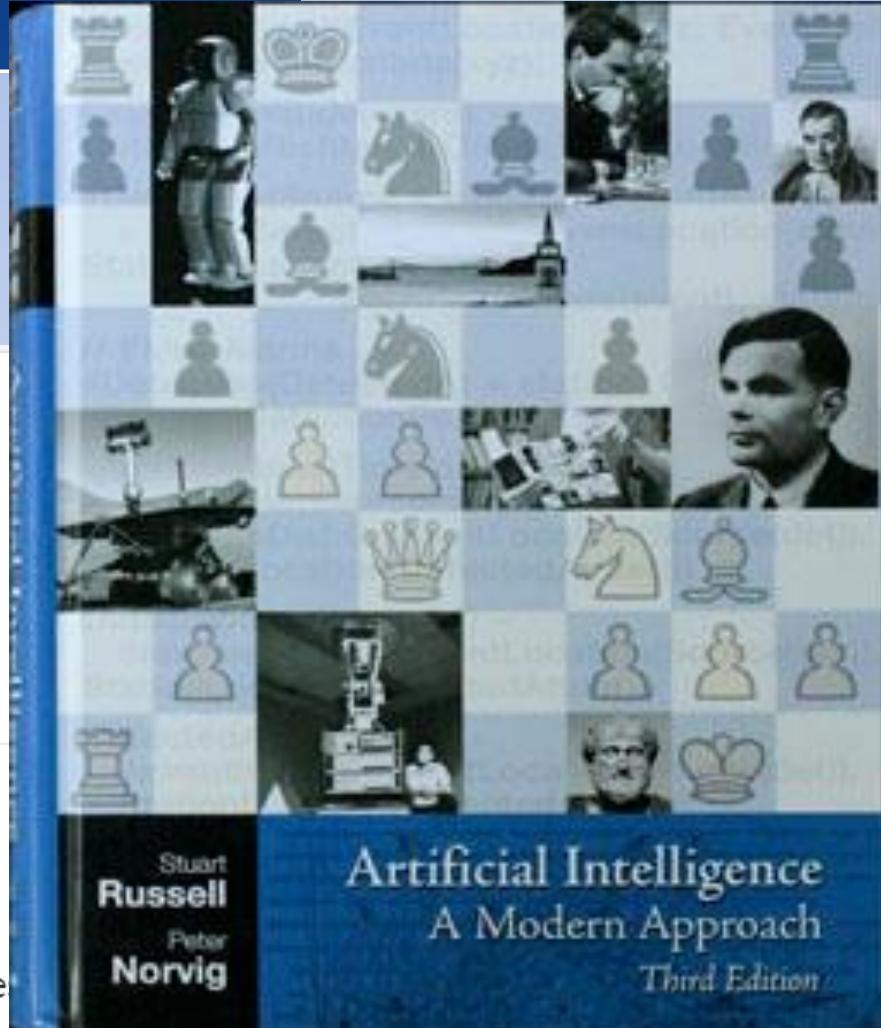
david ruby

Lecturer at California State University, Fresno

California State University, Fresno • UC Irvine

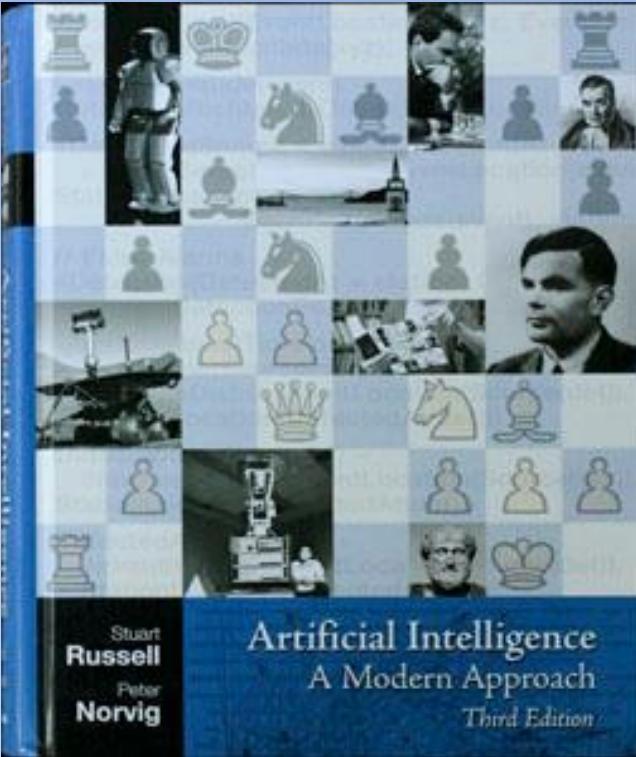
# WELCOME

## Spring '20



# CSCI 164

## AI Programming



← AI Platform: Qwik Start

**Start Lab** 01:00:00

Student Resources

- What is Machine Learning?
- Harness the Power of Machine Learning with AI Platform
- AI Platform: Qwik Start - Qwiklabs Preview



## AI Platform: Qwik Start



AUTOPLAY

Up next

The 7 Steps of Machine Learning (AI Adventures) Google Cloud Platform 1.5M views 10:36

What is Hadoop?: SQL Comparison Intricity101 506K views 6:14

Google Interview Experience (2019) Accepted... then... Keep On Coding 82K views 17:23

Top 10 Technologies To Learn In 2020 | Trending Technologi... edureka! 430K views 10:03

AI Adventures Google Cloud Platform 44

What is Machine Learning? (AI Adventures)

0:07 / 5:22

Pictures Music Text Videos

6.1K 90 SHARE SAVE ...

417,059 views • Aug 24, 2017

# Tech Topics

- Artificial Intelligence
  - It's everywhere

# Current Context: Information Age Arrival

- Machine Learning/Artificial Intelligence
- Data Insanely Important
- Computers Driving Change

# Recent Advances

- Smart Speakers
  - Amazon Echo
  - Google Home
  - Apple Siri
- Speech Recognition
  - Speech-To-Text
  - Trigger Word/Wake-Up Word
    - Siri!
  - Speaker ID
  - Speech Synthesis

# AI @ Fresno State

## BULLDOG AUTONOMOUS BOT



AWS  
RoboMaker

FRESNO STATE®

Digital Transformation

DX.FRESNOSTATE.EDU

# Areas of Advances

- Natural Language Processing
  - Text Classification
  - Information Retrieval
  - Name Entity Recognition
  - Machine Translation
  - Parsing, Parts of speech tagging

# Areas of Advances

- Self Driving Cars
- Computer Vision
  - Image Classification/Object Recognition
  - Object Detection
  - Image Segmentation
  - Tracking

# AI/ML Research Projects @ Fresno State

- **Automating pistachio canopy temperature identification: A machine learning approach**
  - Funding body: Agricultural Research Institute
  - Value: \$64,825
  - Investigator(s): **Athanasisos Aris Panagopoulos (PI)**, Shawn Ashkan (co-PI)  
CSU student impact: More than 10 CSU (grad and undergrad) students
- **Customer-controlled, Price-mediated, Automated Demand Response for Commercial Buildings**
  - Funding body: UC Berkeley (California Energy Commission Passthrough)
  - Value: \$30,000
  - Investigator(s): **Athanasisos Aris Panagopoulos (PI)**  
CSU student impact: 1 grad students

# Areas of Advances

- Robotics
  - Perception
  - Motion Planning
  - Control

# Smart Campus with AI, IoT, and Cloud

We have successfully completed several adoptions of IoT innovation toward a Smart Fresno State Campus with the partnership of DISCOVERe, Library, and PD, and Auxiliary. We received direct sponsorship from Amazon Enterprise Button program as the first Higher Ed institution highlighted in AWS re:Invent conference. Technology Services/Innovation is in collaboration with Lyles Center for further innovation and entrepreneurship opportunities.

California State University, Fresno - Technology Services/Innovation

## IoT@ Fresno State

### Internet of Things

"The Internet of things (IoT) is the inter-networking of physical devices (also referred to as "connected devices" and "smart devices") embedded with electronics, software, sensors, actuators, and network connectivity which enable these objects to collect and exchange data."

### Pilots

Guides On-The-Go! Pay for Print Workspace Climate Entrance Assistance Press to Report Awareness Button

### Learning

#### IoT Readiness Assessment

- Infrastructure (for MTC/M2M)
- Device Management (and Security)
- Service Architecture & Cloud
- Big Data and Analytics

#### Establish a campus IoT Framework of

- solution design;
- technology selection guidelines;
- full scope of infrastructure support;
- device management;
- security policies;
- user support and training;
- the best practices



Highlighted at AWS re:Invent Conference

Fresno State is the First Higher Ed partner sponsored by Amazon to join the AWS Enterprise IoT Button.

Presented by

Max J. Tsai and Jonathan Castro  
Technology Services  
innovate.fresnostate.edu

**FRESNO STATE**

Discovery. Diversity. Distinction.

President's  
**SHOWCASE**  
*of Excellence 2018*

# Algorithms Feeding on Data

- Data, Data, Data

Regulating the internet giants

# The world's most valuable resource is no longer oil, but data

*The data economy demands a new approach to antitrust rules*



[Print edition | Leaders >](#)

May 6th 2017



# Salesforce Tower San Francisco

- Tallest Building in SF when completed in 2018.

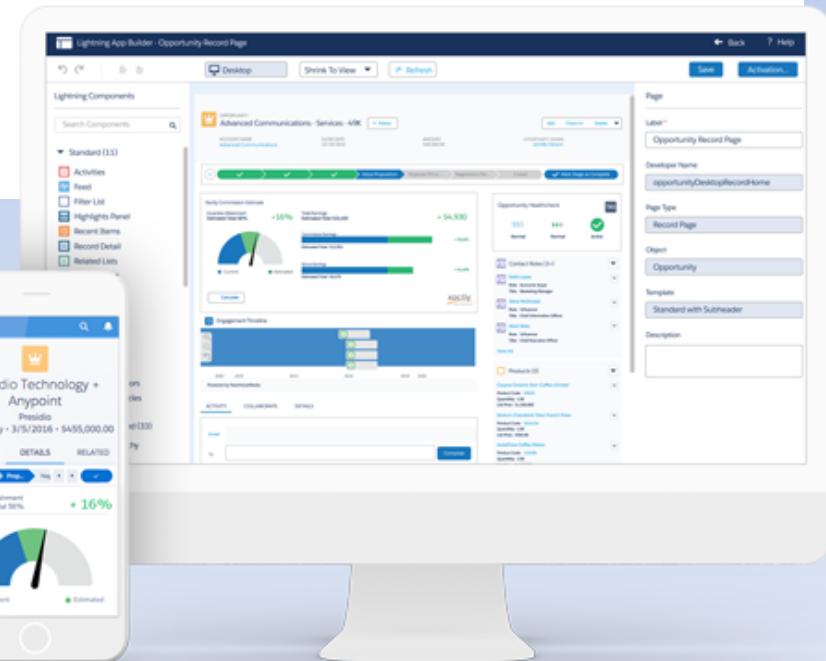




# Connect to your customers in a whole new way with the world's #1 CRM platform.

Integrate every part of your company that interacts with customers – including marketing, sales, service, and more – onto one CRM platform, Salesforce Customer 360 :

- ✓ Gives teams a shared view of every customer
- ✓ Automates workflows using artificial intelligence
- ✓ Integrates easily with existing data and systems
- ✓ Keeps your business data safe and secure
- ✓ Customizes and scales for any size company



# Salesforce Completes Acquisition of Tableau

PUBLISH DATE: AUGUST 1, 2019 - 5:30AM

*World's #1 CRM and #1 analytics platform come together to supercharge customers' digital transformations*

*Combination to accelerate Salesforce's opportunity in the \$1.8 trillion digital transformation space*

*Customers will be able to unlock even greater value from their data to drive smarter business decisions and more intelligent customer experiences*

*Tableau will be positioned to further its mission to help the world see and understand data*

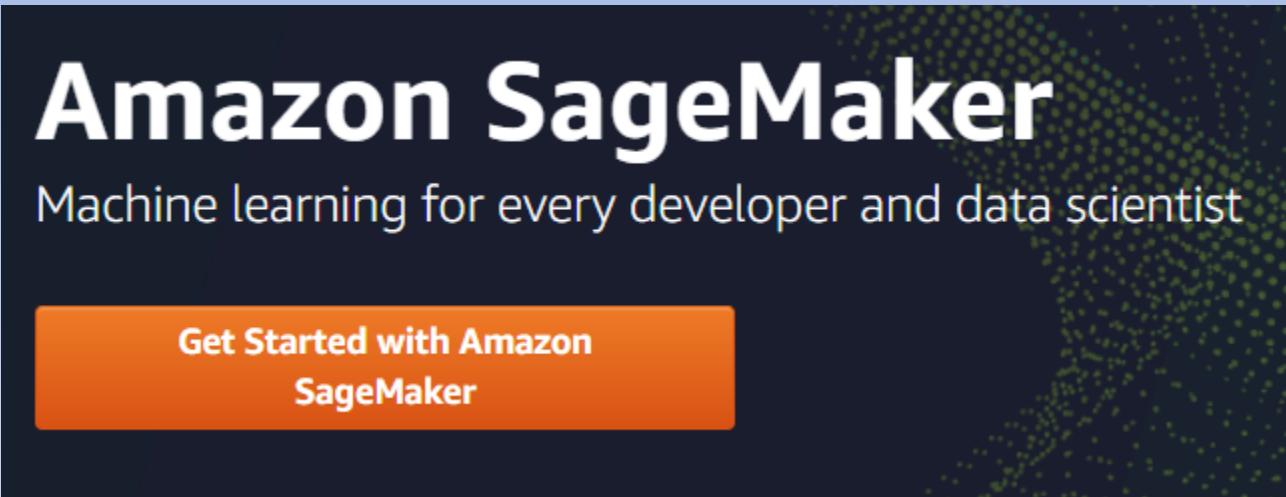
# AI/Data In Cloud

- Amazon Web Services (AWS)
- Microsoft Web Services (Azure)
- Google Cloud Services (GCP)

# AI w/ Amazon AWS

The screenshot shows the official AWS website homepage. At the top, there's a dark navigation bar with the AWS logo on the left and links for Contact Sales, Support, English, and My Account on the right. Below the navigation bar is a secondary horizontal menu with links for Products, Solutions, Pricing, Documentation, Learn, Partner Network, AWS Marketplace, Customer Enablement, Events, Explore More, and a search icon. The main content area features a large, stylized graphic of red and purple dots forming a wave pattern. Overlaid on this graphic is a prominent banner with the text "Machine Learning on AWS" in large, bold, white letters. Below the banner, a subtext reads "More organizations choose AWS for machine learning than any other cloud".

# AI w/ Amazon AWS



The image shows the landing page for Amazon SageMaker. The background features a dark blue gradient with a subtle dotted pattern. At the top, the text "AI w/ Amazon AWS" is displayed in white. Below this, the title "Amazon SageMaker" is prominently shown in large, bold, white letters. A subtitle "Machine learning for every developer and data scientist" follows in a smaller, white font. At the bottom left, there is an orange button with the text "Get Started with Amazon SageMaker" in white.

# Amazon SageMaker

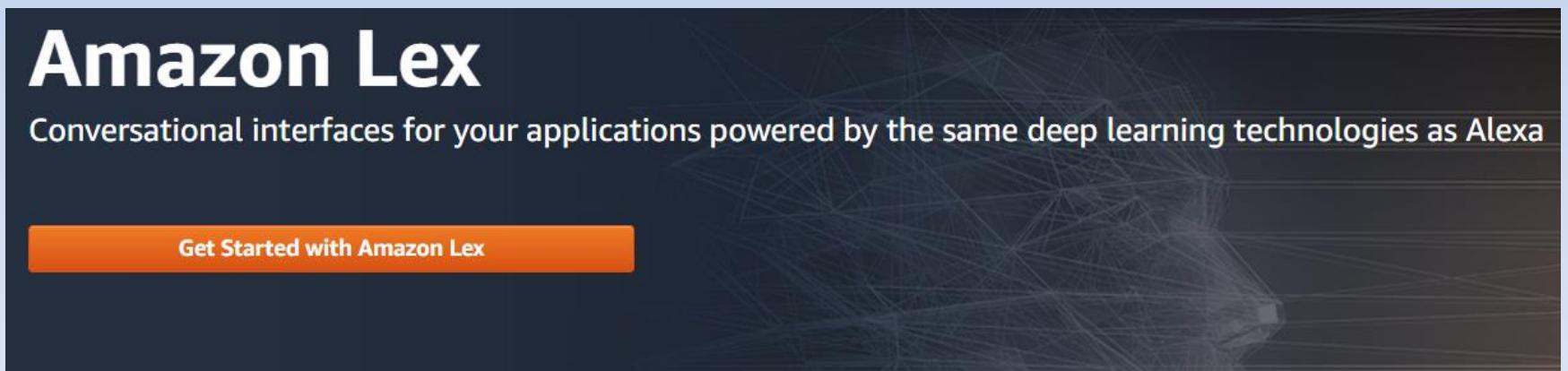
Machine learning for every developer and data scientist

Get Started with Amazon SageMaker

## Amazon Lex

Conversational interfaces for your applications powered by the same deep learning technologies as Alexa

Get Started with Amazon Lex

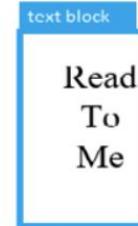


The image shows the landing page for Amazon Lex. The background features a dark blue gradient with a complex, abstract wireframe or network structure in the center. At the top, the title "Amazon Lex" is displayed in large, white, sans-serif font. Below the title, a subtitle explains the service: "Conversational interfaces for your applications powered by the same deep learning technologies as Alexa". At the bottom left, there is an orange button with the text "Get Started with Amazon Lex" in white.

# DeepLens w/ AWS

## Meet the AWS DeepLens Hackathon Winners

### First Place



#### ReadToMe

This is a Deep Learning enabled application which is able to read books to kids.

### Second Place



#### Dee

A fun, interactive device for children that asks them to answer questions by showing the right things.

## The world's first deep learning enabled video camera for developers

AWS DeepLens helps put machine learning in the hands of developers, literally, with a fully programmable video camera, tutorials, code, and pre-trained models designed to expand deep learning skills.

The new AWS DeepLens (2019 Edition) is available to purchase in the US and in seven new countries: **UK, Germany, France, Spain, Italy, Canada, and Japan**. We have improved the hardware and software to make the device even easier to setup, allowing you to get started with machine learning more quickly.

[Buy Now](#)

[Register your DeepLens](#)



# AI w/ Microsoft Azure

The image shows the Microsoft Azure website header. It features the Microsoft logo (four colored squares) followed by the text "Microsoft Azure". Below this, there is a navigation bar with four items: "Overview" (which is underlined, indicating it's the current page), "Solutions", "Products", and "Documentation".

## Azure AI

Make artificial intelligence (AI) real for your business today.

[Start free >](#)

Microsoft first to achieve  
human parity in:



Vision

Object recognition



Speech

Speech recognition



Language

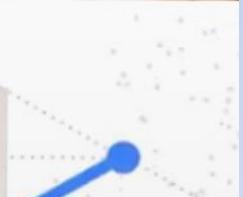
Machine translation

# ai.google

## Advancing AI for everyone



Our vision



# Google Dialogflow

Google Cloud   Why Google   Solutions   **Products**   Pricing   Getting started

AI & Machine Learning Products

## Dialogflow

Create conversational experiences across devices and platforms.

[Go to console](#)

[View documentation for this product.](#)

### Natural conversational experiences

Dialogflow is an end-to-end, build-once deploy-everywhere development suite for creating conversational interfaces for websites, mobile applications, popular messaging platforms, and IoT devices. You can use it to build interfaces (such as chatbots and conversational IVR) that enable natural and rich interactions between your users and your business. Dialogflow...

A 3D-style diagram showing a central cloud icon containing a blue gear-like logo. Green lines connect this central node to various devices: a laptop, a smartphone, a tablet, a desktop computer, and a person's head (representing a user). Each device has a small green speech bubble icon next to it, symbolizing communication or conversation. This visualizes how Dialogflow integrates with multiple platforms to facilitate conversational experiences.

23

# Google Cloud Vision

The screenshot shows the Google Cloud website's navigation bar at the top, featuring links for Google Cloud, Why Google, Solutions, Products (which is underlined), Pricing, and Getting started. To the right of the navigation are search, documentation, support, language selection, and console links. Below the navigation, a sub-navigation bar for 'AI & Machine Learning Products' includes a 'Contact' button. The main content area has a sidebar titled 'VISION AI' with links for Overview, Benefits, Features, Customers, Use cases, Pricing, Resources, Training, and Get started. The main content area features a large title 'Vision AI' and a description about deriving insights from images using AutoML Vision or pre-trained Vision API models. A blue 'Go to console' button is located below the description. To the right, there is a large image of wind turbines in a field, with a YouTube play button icon overlaid. A caption at the bottom of this image reads: 'AES, a Fortune 500 global power company, is using drones and AutoML Vision to accelerate a safer, greener energy future.'

## Vision AI

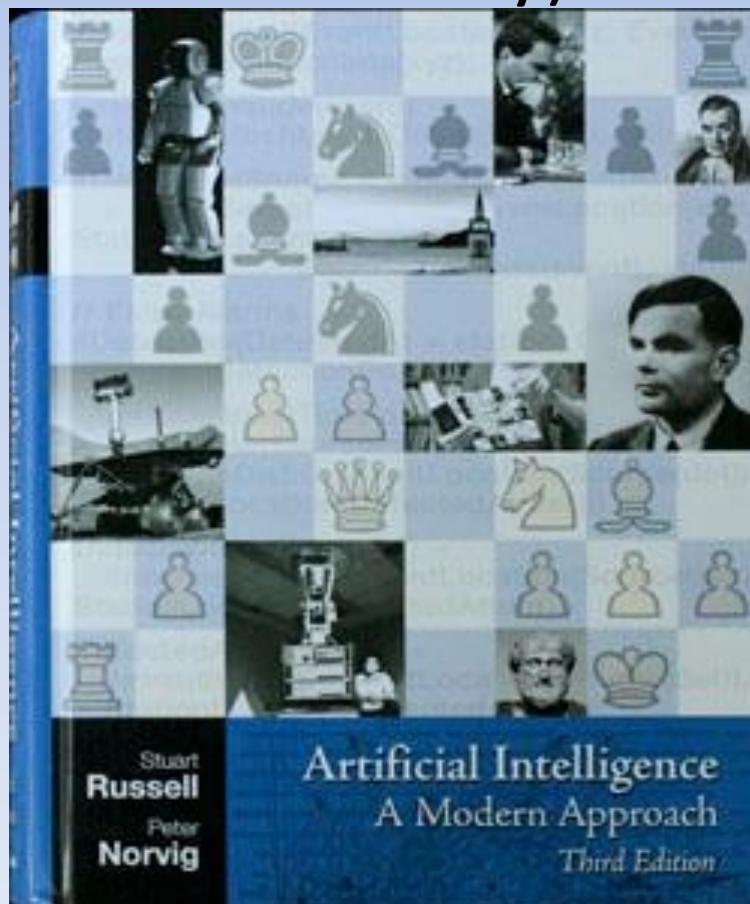
Derive insights from your images in the cloud or at the edge with AutoML Vision or use pre-trained Vision API models to detect emotion, understand text, and more.

[Go to console](#)

AES, a Fortune 500 global power company, is using drones and AutoML Vision to accelerate a safer, greener energy future.

# Textbook

- The 22nd most cited computer science publication on Citeseer (and 4th most cited publication of this century).



# Dr. Stuart Russell

Professor of Computer Science and Michael H. Smith and Lotfi A. Zadeh Chair in Engineering,  
Computer Science Division  
University of California  
Berkeley, CA 94720

Adjunct Professor of Neurological Surgery, University of California, San Francisco

**Tel.** (510) 642-4964

**Fax** (510) 642-5775

**Email** russell@cs.berkeley.edu

**Home page** <http://www.cs.berkeley.edu/~russell>

## Education

B.A. (Hons.) 1st Class, Physics, Wadham College, University of Oxford, 1979--82.

Ph.D., Computer Science, Stanford University, 1982--86.

## Employment history

2012--present, Professeur Invité, Université Pierre et Marie Curie, Paris

2012--present, Professeur, Fondation de l'École Normale Supérieure, Paris

2008--present, Adjunct Professor, Department of Neurological Surgery, University of California, San Francisco

2008--2010, Chair, Department of Electrical Engineering and Computer Sciences, University of California, Berkeley

2006--2010, Chair, Computer Science Division, University of California, Berkeley

1996--present, Professor, Computer Science Division, University of California, Berkeley

1991--96, Associate Professor, Computer Science Division, University of California, Berkeley

1986--91, Assistant Professor, Computer Science Division, University of California, Berkeley

1986, Summer employee, MCC, Austin, Texas, Machine learning research in the Large Scale KB Project (CYC)

1985--86, Research Assistant, Computer Science Dept., Stanford University

1983, Teaching Assistant, Computer Science Dept., Stanford University

1981, Programmer, graphics research project, IBM Los Angeles Scientific Center

1978--80 (1 year total), Programmer, IBM Systems Engineering Centre, Warwick, UK



# Dr. Peter Norvig

## Professional Employment (Full-Time)

2001-now	<a href="#">Google</a>	Director of Research (2006-now); formerly Director of Search Quality (2002-2006) and Machine Learning (2001).
1998-2001	<a href="#">NASA Ames Research Center</a>	Division Chief, Computational Sciences
1996-1998	<a href="#">Jungle Corp.</a>	Chief Scientist
1994-1996	<a href="#">Harlequin, Inc.</a>	Chief Designer
1991-1994	<a href="#">Sun Microsystems Labs</a>	Senior Scientist
1986-1991	<a href="#">University of California, Berkeley</a>	Research Faculty Member
1985-1986	<a href="#">University of Southern California</a>	Assistant Professor
1978-1980	<a href="#">Higher Order Software, Inc.</a>	Member of Technical Staff
1977-1977	<a href="#">Woods Hole Oceanographic Institute</a>	Summer Programming Intern

I also have served as an advisory board member for various companies, including: [Root-1](#), [Fetch](#), [CleverSet](#), [Ask Jeeves](#), [Thinking Software](#), [PersonalGenie.com](#).

## Education

1980-1985	Ph.D. Computer Science	<a href="#">University of California, Berkeley</a>
1974-1978	B.S. Applied Mathematics	<a href="#">Brown University</a>

## Personal Information

Citizen: U.S.  
Raised: RI, MA, CA.  
Status: Married with 2 children.  
Erdos #: 3 (Erdos to Peter Cameron to Stuart Russell to me)



# Personal Computer Science Focus

## CSCI 164. Artificial Intelligence Programming

Prerequisite: CSCI 117. Introduction to problem-solving methods from artificial intelligence. Production systems. Knowledge-based systems. Machine learning. Topics chosen from fuzzy logic, neural network models, genetic algorithms. Verification, validation, testing.

Units: 3

## CSCI 166. Principles of Artificial Intelligence

Prerequisite: CSCI 164. Analysis of knowledge-based and neural models, including self-organization, sequential learning models, neurally inspired models of reasoning and perception. Integration of different paradigms.

Units: 3

## CSCI 174. Design and Analysis of Algorithms

Prerequisites: CSCI 115, CSCI 119. Models of computation and measures of complexity, algorithms for sorting and searching, set representation and manipulation, branch and bound, integer and polynomial arithmetic, pattern-matching algorithms, parsing algorithm, graph algorithm, NP-complete problems.

Units: 3

## CSCI 126. Database Systems

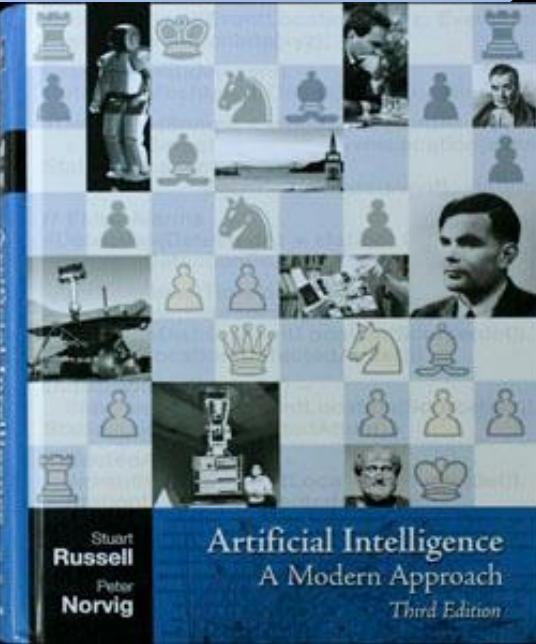
Prerequisites: CSCI 124. Database concepts; hierarchical and relational network models; object-oriented data models. Data normalization, data description languages, data manipulation languages, and query design.

Units: 3

Course Typically Offered: Spring

# Personal Computer Science Focus

- Data
  - Machine Learning, Algorithms, RDMS
- Search
  - AI, Algorithms, RDMS
- Puzzles
  - AI, Algorithms
- Books...



## CSCI 174. Design and Analysis of Algorithms

Prerequisites: CSCI 115, CSCI 119. Models of computation, measures of complexity, algorithms for sorting, representation and manipulation, branch-and-bound, polynomial arithmetic, pattern-matching algorithms, graph algorithm, NP-complete problems

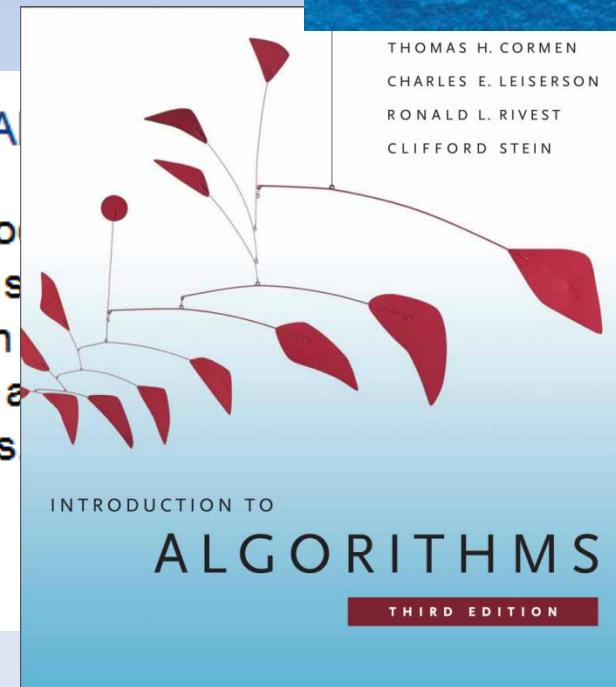
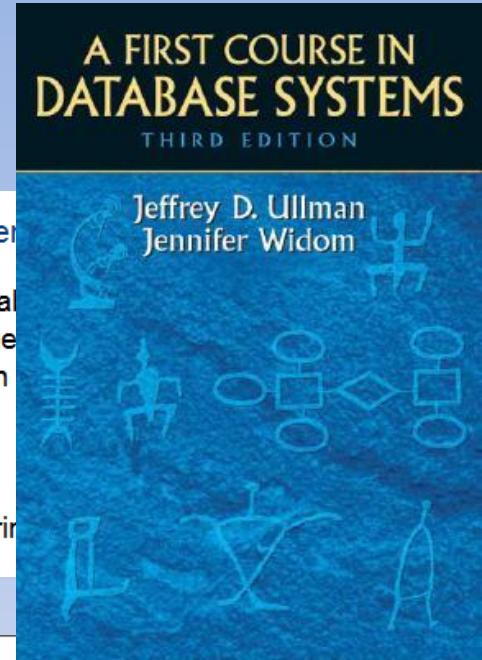
Units: 3

## CSCI 126. Database Systems

Prerequisites: CSCI 124. Data structures, relational algebra, relational network models; object-oriented modeling, normalization, data description languages, and query design.

Units: 3

Course Typically Offered: Spring



# Grading

Period	Presentations/Examinations/Assignments	Points
Various	Attendance	100
	In-Class	100
	Quizzes	100
	Qwiklabs	200
	Midterm	200
	Final	300

# Tests

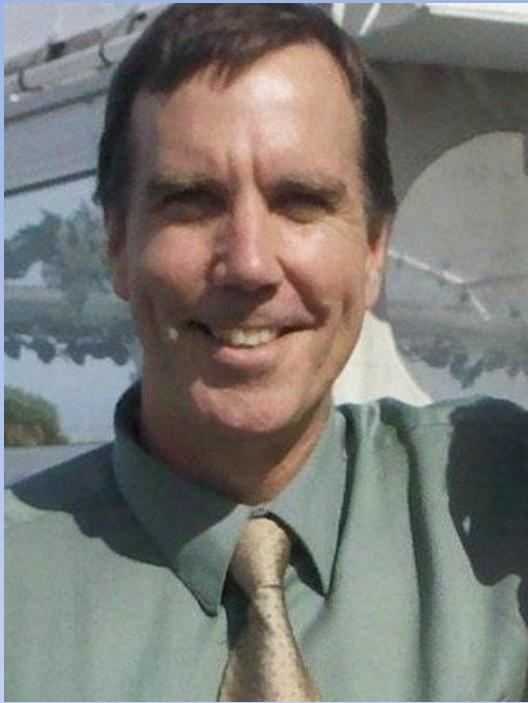
- Midterm (20%)
- Final (30%)

# Laptops

- In-Class testing using laptops.

# Assignments: Qwiklabs

- A chance to see AI Tools



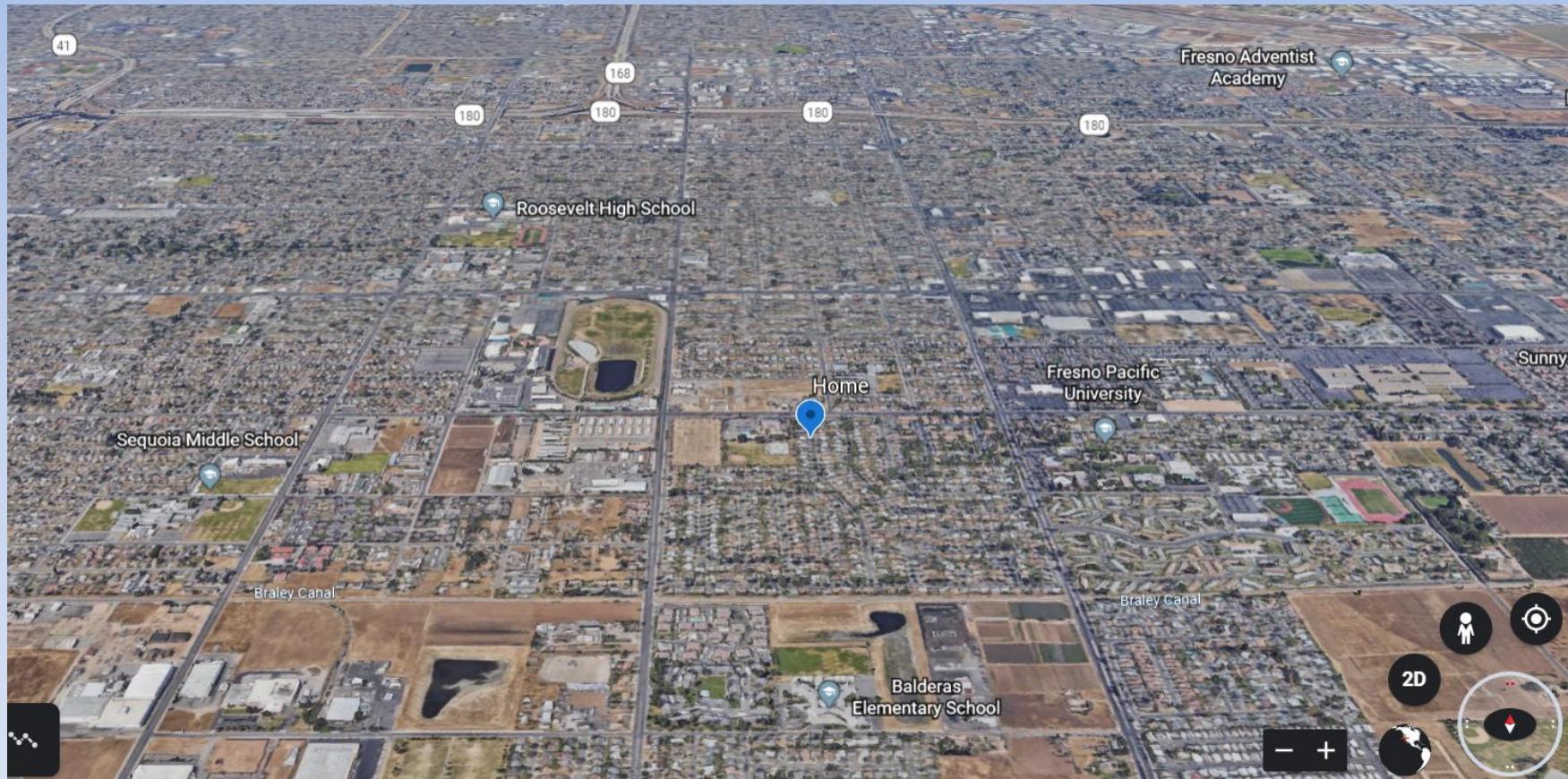
David Ruby

# Class Instructor

- Office
  - Science II – 273
- Email:
  - druby@csufresno.edu

- First-Generation College Student
- How PhD?

# Father Floyd Fresno Career Custodian Ending @ Sequoia Middle School



# North to CSUF



# Compute Science Focus: Jobs/Degrees

- Students want...
  - Jobs!
  - Advanced Degrees!!
- How???



Antony Ross

## Antony Ross

Machine Learning/Financial Advisor @ Principal Financial Group

Antony Ross has been a consultant with SportInsight in Los Angeles, specializing in data science and machine learning, and has worked with USC and UCLA analyzing sports performance. He has also worked with the Recurse Center in New York researching deep learning, music recommendations and voice recognition. Antony presently works for the Principal Financial Group in Fresno and applies machine learning in financial technology.



# Carlos Moreno



[Home](#)

[Posts](#)

[CV](#)

## Carlos Moreno

(last updated on 09/11/18)

Carlos Moreno, M.S.

Ph.D. Student, Computer Science, Cornell University

Office: Gates 322

Email: [cmoreno \(AT\) cs \(DOT\) cornell \(DOT\) edu](mailto:cmoreno@cs.cornell.edu)

**Cornell CIS**  
COMPUTING AND INFORMATION SCIENCE



Welcome to the website for Carlos Moreno! I will try my best to keep this as updated as possible throughout my professional career. At the moment, I am a new Ph.D. student at Cornell University, in the Department of Computer Science. Beyond graduation, I aspire to one day become a faculty member, to give back in the way many countless professors have in the past.

In terms of research interests, I am currently intrigued by two general areas. The first is computer science education. What are good approaches to teaching computer science concepts to students? How do these techniques vary in regards to the demographics of the students? Specifically, I am interested in looking at computer science education in regards to under-represented minorities (URMs).

Second, I am interested in machine learning. How can we teach computers to learn about their environment, whether supervised or not? Moreover, what are the moral and ethical implications of machine learning, and artificial intelligence in general? Specifically, I am looking at how machine learning can be used to solve problems in the world of computational sustainability. Can we use computer science to make the world more sustainable? I believe the answer to this question is paramount for our future as a species.



## EDUCATION

### **Cornell University**

*Ithaca, NY*

Aug. 2018 – Ongoing

- Ph.D. student in Computer Science
- Research interests: computer science education, specifically for URMs; and machine learning, specifically for computational sustainability

### **California State University, Fresno**

*Fresno, CA*

Aug. 2015 — May 2017

- Master of Science (M.S.) in Computer Science with Distinction
- 4.0 GPA
- Dean's Medalist for the College of Science and Mathematics
- Outstanding Graduate Award for the Department of Computer Science
- Thesis: Supporting Quality of Service in Real-Time Video Streaming Using the Kinect

### **California State University, Fresno**

*Fresno, CA*

Aug. 2011 — May 2015

- Bachelor of Science (B.S.) in Computer Science with University Honors
- Magna Cum Laude, 3.82 GPA
- President's Honors Scholar
- Standard Bearer for the College of Science and Mathematics
- Outstanding Undergraduate Award for the Department of Computer Science

# Dr. Joy Goto, Professor Biochemistry, Fresno State

- [https://youtu.be/FXUiEPrK\\_II](https://youtu.be/FXUiEPrK_II)



# Dr. Joy Goto, Professor Biochemistry, Fresno State



- [https://youtu.be/FXUiEPrK\\_II](https://youtu.be/FXUiEPrK_II)
- Engaging story of discovering joy in science growing up here in the central valley.

# Grad School Colleagues



**David W. Aha** ([Bio Vitae](#))

Naval Research Laboratory  
Navy Center for Applied Research in Artificial Intelligence

Washington, DC

david.aha@nrl.navy.mil



+1 (202) 404-4940

## Education

- 1992-93 NSERC International Post-Doctoral Fellowship, University of Ottawa, Canada
- 1991-92 APL/Johns Hopkins University Post-Doctoral Fellowship, Laurel, MD
- 1990-91 SERC Post-Doctoral Fellowship, Turing Institute, Glasgow, Scotland
- 1990 Ph.D. (Computer Science), University of California, Irvine
- 1985 M.S. in Computer Science, Syracuse University
- 1983 B.S. in Computer Science, Syracuse University, Magna Cum Laude with Honors, University Scholar (the highest undergraduate honor bestowed by SU)
- 1979 Walt Whitman High School; South Huntington, New York

## Employment (Non-Educational)

- (1993-) Naval Research Laboratory
- (1982) PAR Technology Corporation, Summer
- (1981) Fairchild Republic Company, Summer

# Goal Reasoning: Emerging Applications, Foundation, and Prospects

David W. Aha

Head, Adaptive Systems Section

Navy Center for Applied Research in Artificial Intelligence  
Naval Research Laboratory (Code 5514); Washington, DC  
[david.aha@nrl.navy.mil](mailto:david.aha@nrl.navy.mil)



## Goal Reasoning (GR): Examples



*It's a Wonderful Life, 1935 (Jimmy Stewart as George Bailey)*



*Goal: Suicide*



*Goal: Life*

*The Force Awakens, 2015 (John Boyega as Finn)*



*Goal: Serve Empire*



*Goal: Defeat Empire*

# Goal Reasoning (GR): Examples (cont.)



*United States, 1941*



*Goal: Peace*



*Goal: War*

*A Change of Research Focus*



*Goal: CBR*



*Goal: GR*

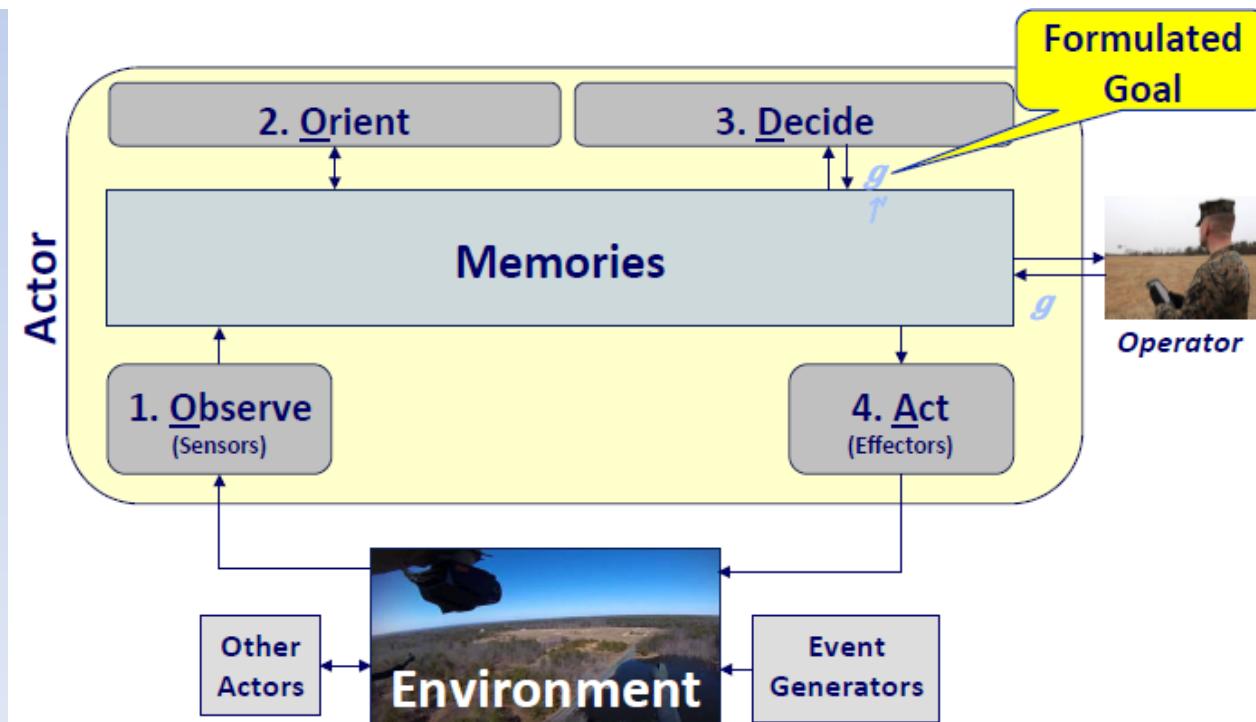
# An Actor Methodology for Continuously Reasoning on Goals

## Requirements

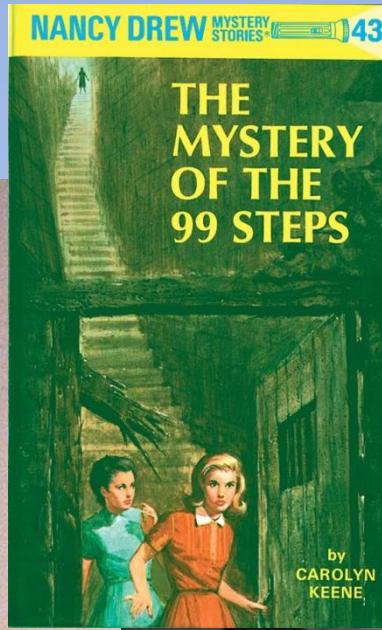
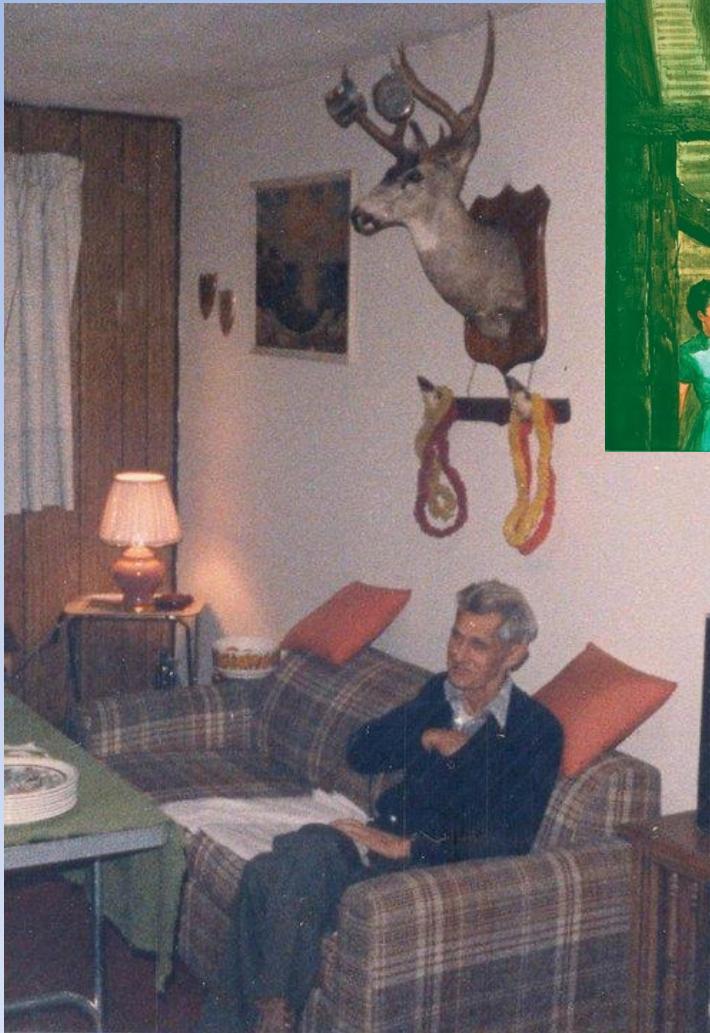
- Goal representation (explicit)
- Reasoning process

**Benefit:** Deliberate on changing one's *goal*

- When an impasse occurs
- When an affordance occurs
- After goal reprioritization



# My Story



- Family Memories



Interest  
In  
Puzzles

# Puzzles: Pacman Projects



## [Home & Projects Overview](#)

[Instructor's Guide](#)

[UNIX/Python Tutorial](#)

[Search Project](#)

[Multiagent Search Project](#)

[Reinforcement Learning Project](#)

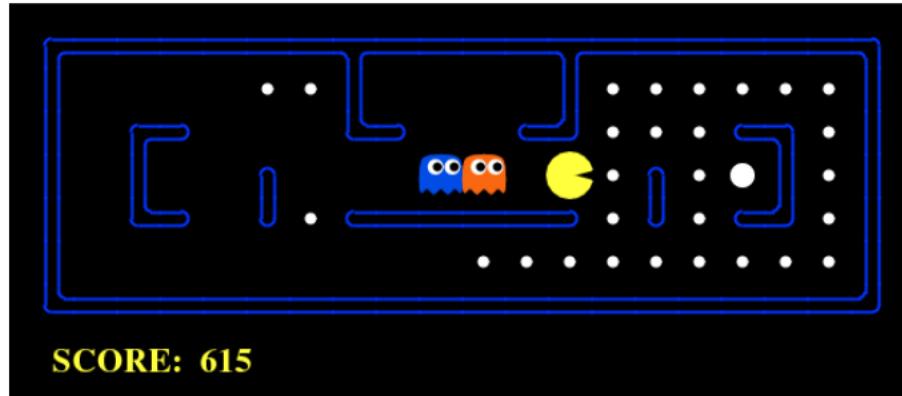
[Ghostbusters Project](#)

[Classification Project](#)

[edX Pilot](#)

[Contact](#)

## The Pac-Man Projects



### Overview

The Pac-Man projects were developed for UC Berkeley's introductory artificial intelligence course, CS 188. They apply an array of AI techniques to playing Pac-Man. However, these projects don't focus on building AI for video games. Instead, they teach foundational AI concepts, such as informed state-space search, probabilistic inference, and reinforcement learning. These concepts underly real-world application areas such as natural language processing, computer vision, and robotics.

We designed these projects with three goals in mind. The projects allow students to visualize the results of the techniques they implement. They also contain code examples and clear directions, but do not force students to wade through undue amounts of scaffolding. Finally, Pac-Man provides a challenging problem environment that demands creative solutions; real-world AI problems are challenging, and Pac-Man is too.

In our course, these projects have boosted enrollment, teaching reviews, and student

Start State

1	2	3
4		6
7	5	8



1	2	3
4	5	6
7		8



1	2	3
4	5	6
7	8	

Goal State

# Thesis: Tile-Sliding Puzzle



# Artificial Intelligence

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## SteppingStone: An Empirical and Analytical Evaluation\*

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### Abstract

Decomposing a difficult problem into simpler subproblems is a classic problem solving technique. Unfortunately, the most difficult subproblems can be as difficult, if not more difficult, than the original problem. This is not an obstacle to problem solving if the difficult subproblems recur in other problems. If the difficult subproblems recur often, then its solution need only be learned once and reused. SteppingStone is a learning problem solver that decomposes a problem into simple and difficult-but-recurring subproblems. It solves the

SteppingStone operates on problems defined with a state space representation consisting of a set of goals, a set of operators, and an initial state. The goal orderer takes as input a set of goals. It orders these goals so that the constrained search method will likely solve them. It does this by ordering them so as to reduce the likelihood of subgoal interactions using a domain independent heuristic we call *openness* [Ruby and Kibler, 1989]. It produces an ordered set of subgoals as output.

The constrained search component takes as input an

# My Idea...

# Memories are constructed.. Not stored complete!

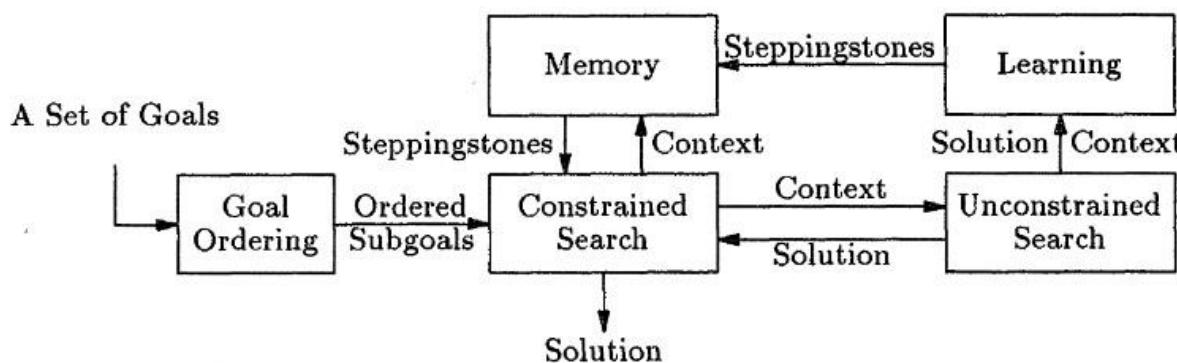


Figure 1: Overview of SteppingStone

the original impasse state.

When memory fails to return any useful stepping-stones the constrained search component calls the unconstrained search component. The unconstrained search component takes as input a context, just as the memory component did. Unconstrained search relaxes the protection on the solved subgoals in its search for a solution. If it resolves the impasse, it returns the sequence of moves found to the constrained search component. The unconstrained search component also sends its impasse solution, along with the context, to the learner.

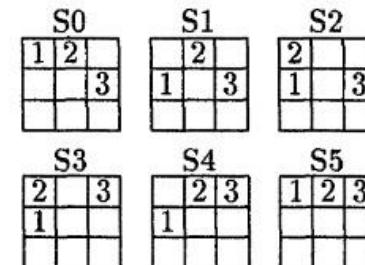
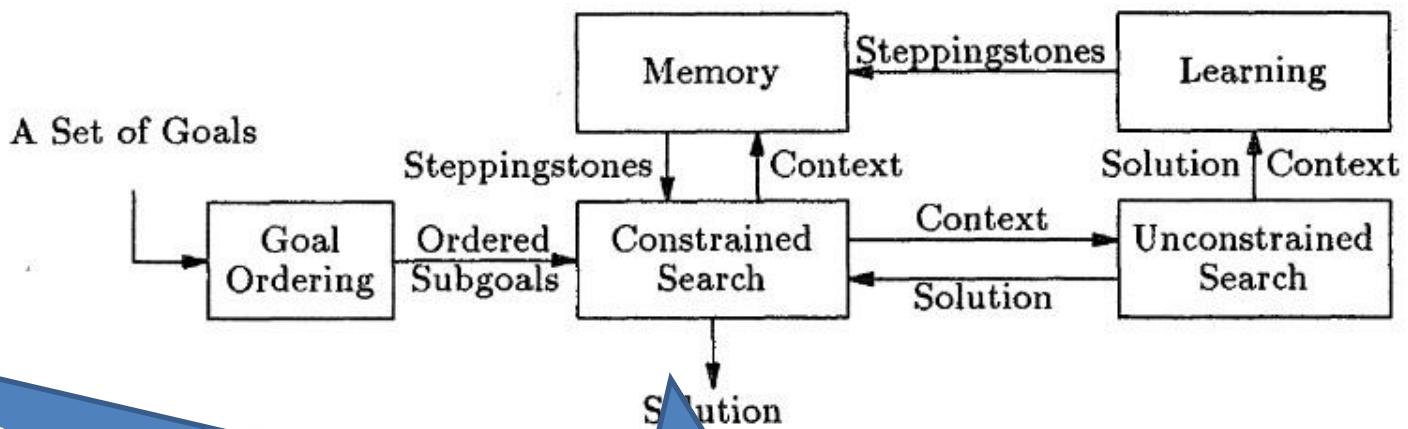


Figure 2: Steppingstones from Memory

# Thesis: Tile Sliding Domain



1: Overview of SteppingStone

When stones the constrained search component takes as input a context from the memory component did. Unconstrained search removes the protection on the solved subgoals in its search for a solution. If it resolves the impasse, it returns the sequence of moves found to the constrained search component. The unconstrained search component also sends its impasse solution, along with the context, to the learner.

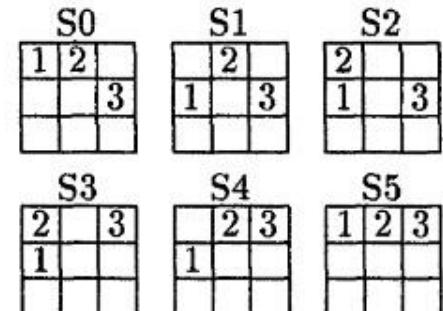
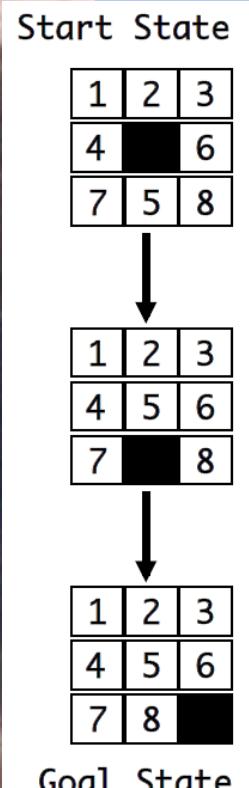


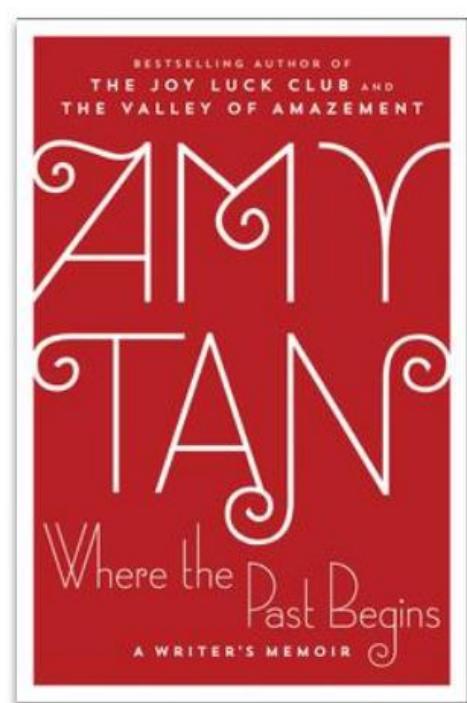
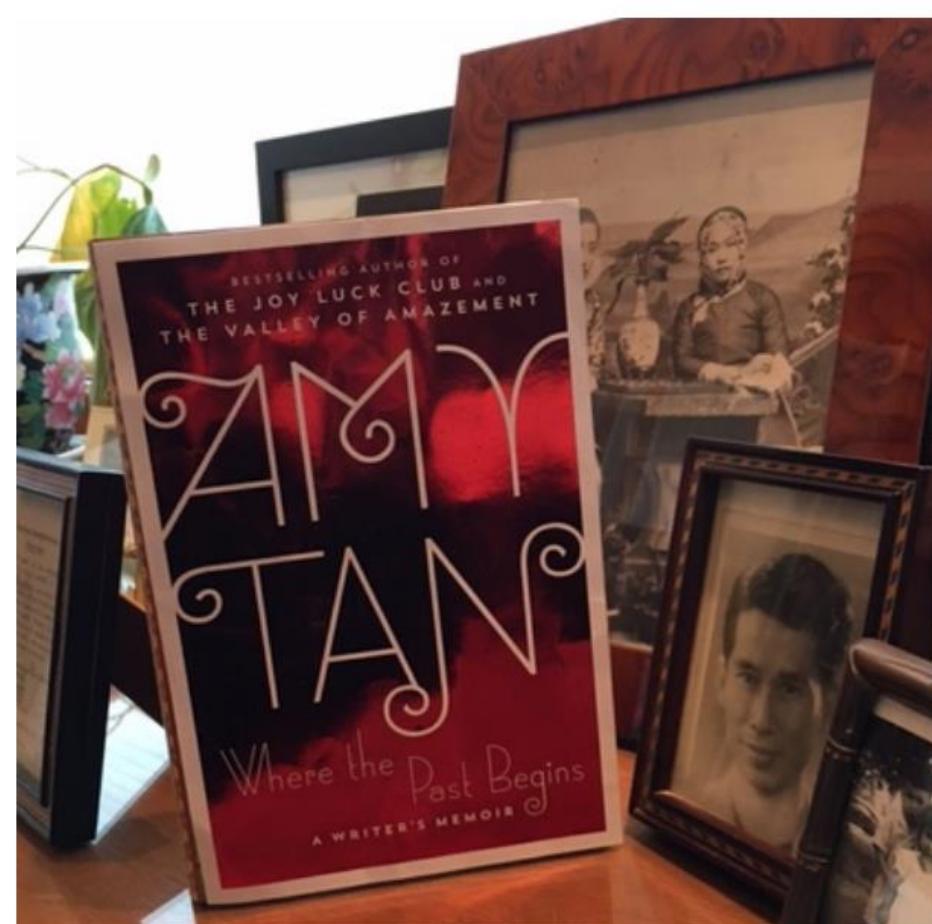
Figure 2: Steppingstones from Memory

# My Memories..

- Home Hedges Maze.
- Also – First time w/ Sliding Tile Puzzle



# Emotional Memory: Process of a Writer



*coming*  
*Oct 17, 2017*  
[Pre-order](#)

# Malleable Memory (Gaps)

## Learning & Memory w/ Elizabeth Loftus



# CORRUPTED MEMORY

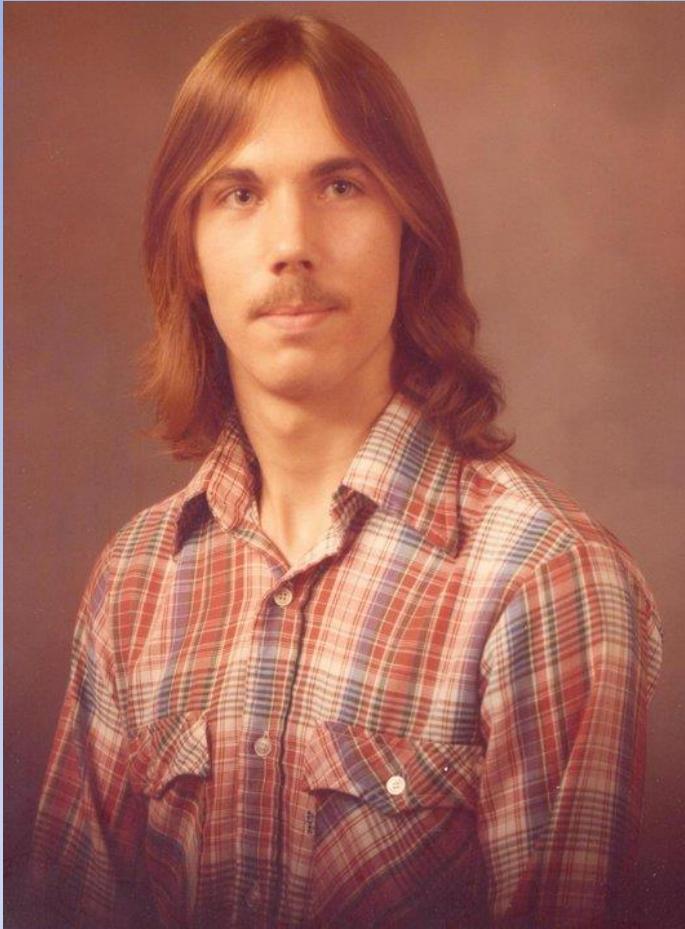
*Elizabeth Loftus has spent decades exposing flaws in eyewitness testimony. Her ideas are gaining fresh traction in the US legal system.*

BY MOHEB COSTANDI



Elizabeth Loftus is a cognitive psychologist at the University of California Irvine.

# eXciting Mazes Memories!



- ME:
  - Do you remember the FUN maze?
- NEIGHBOR:
  - Do YOU remember this other HORRIBLE thing??
- ME:
  - Hmm .. I guess not.
- Language influencing memory ??

# Memories & Learning

JOURNAL OF VERBAL LEARNING AND VERBAL BEHAVIOR 13, 585-589 (1974)

## **Reconstruction of Automobile Destruction: An Example of the Interaction Between Language and Memory'**

ELIZABETH F. LOFTUS AND JOHN C. PALMER

*University of Washington*

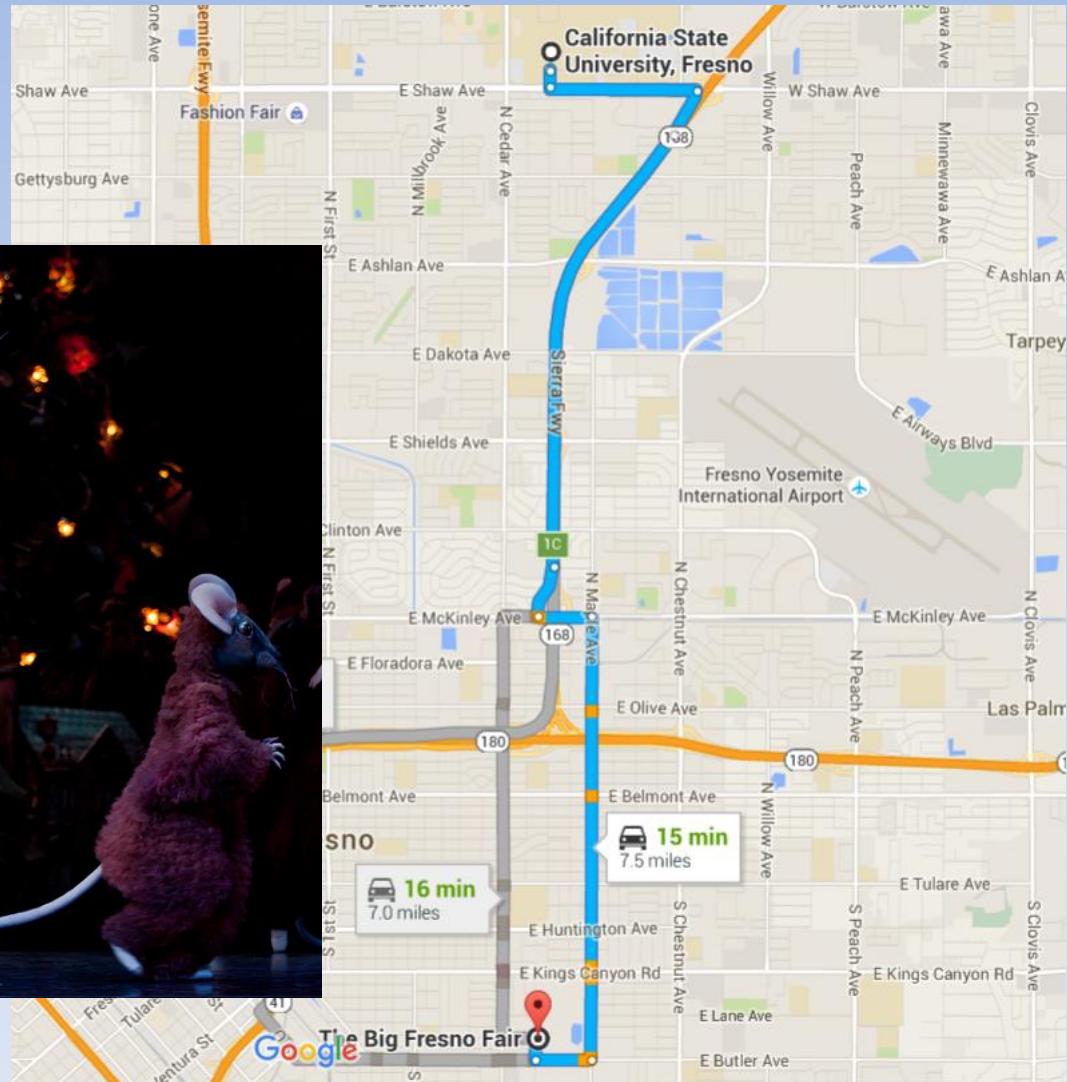
Two experiments are reported in which subjects viewed films of automobile accidents and then answered questions about events occurring in the films. The question, "About how fast were the cars going when they smashed into each other?" elicited higher estimates of speed than questions which used the verbs *collided*, *bumped*, *contacted*, or *hit* in place of *smashed*. On a retest one week later, those subjects who received the verb *smashed* were more likely to say "yes" to the question, "Did you see any broken glass?", even though broken glass was not present in the film. These results are consistent with the view that the questions asked subsequent to an event can cause a reconstruction in one's memory of that event.

# Computer Science / Memories

- Puzzles
- Abstractions
- Memories

# Current Interest: Abstraction

- Hello, World!



# Abstraction: Computational Thinking

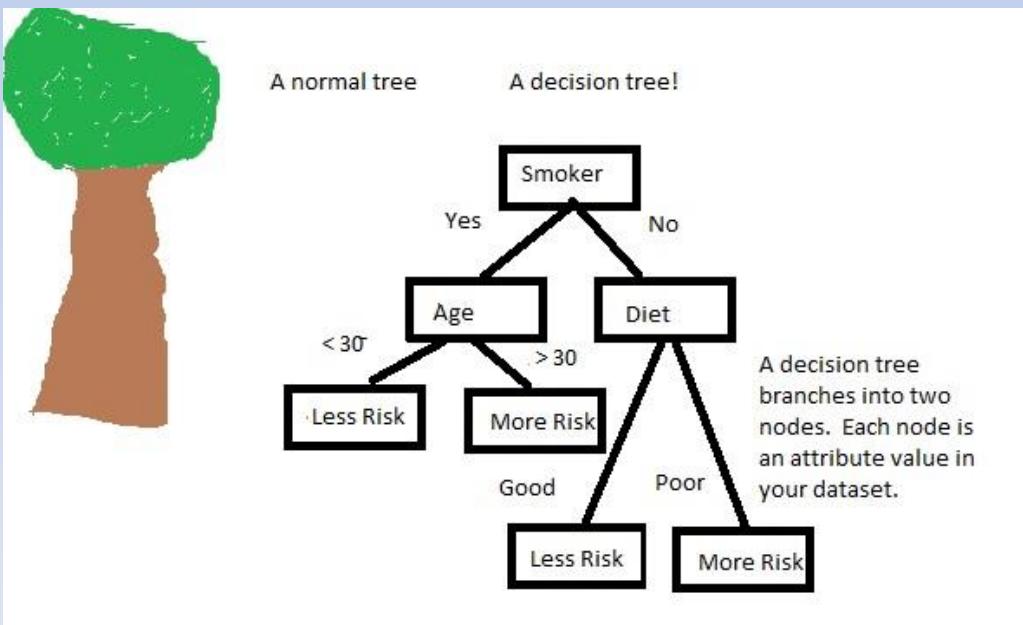
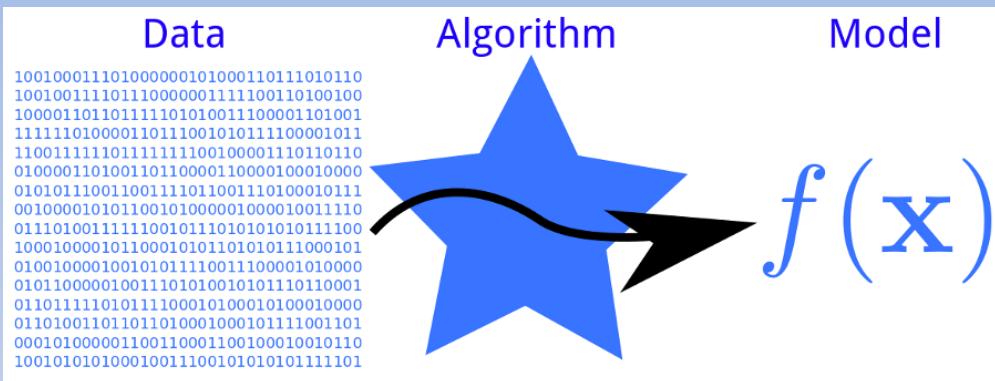
- Abstraction
- Automation
- Algorithms/Analysis

# Intelligence (Problem Solving) Requires..

## ..Learning from Experiences

- Intelligence requires learning from experiences.

# Machine Learning - AI



THE MASTER  
ALGORITHM

HOW THE QUEST FOR  
THE ULTIMATE  
LEARNING MACHINE WILL  
REMAKE OUR WORLD

PEDRO DOMINGOS

## ■ UPCOMING AAAI EVENTS

### July 2014

The Twenty-Eighth AAAI Conference will be held in Québec City, Québec, Canada July 27–31.

The Twenty-Sixth IAAI Conference on Innovative Applications of Artificial Intelligence will be held in Québec City, Québec, Canada July 29–31.

### October 2014

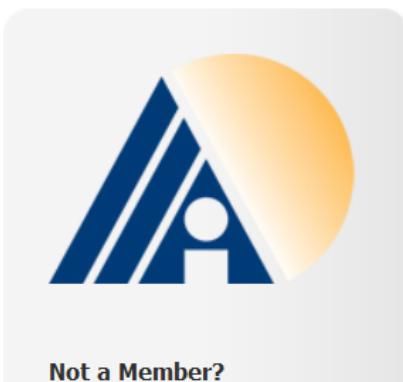
The Tenth AAAI Conference on Artificial Intelligence and Interactive Digital Entertainment will be held October 3–7 in Raleigh, North Carolina USA



AAAI Conference on Artificial Intelligence and Interactive Digital Entertainment

## ■ WELCOME TO THE ASSOCIATION FOR THE ADVANCEMENT OF ARTIFICIAL INTELLIGENCE!

Founded in 1979, the Association for the Advancement of Artificial Intelligence (AAAI) (formerly the American Association for Artificial Intelligence) is a nonprofit scientific society devoted to advancing the scientific understanding of the mechanisms underlying thought and intelligent behavior and their embodiment in machines. AAAI aims to promote research in, and responsible use of, artificial intelligence. AAAI also aims to increase public understanding of artificial intelligence, improve the teaching and training of AI practitioners, and provide guidance for research planners and funders concerning the importance and potential of current AI developments and future directions. [More...](#)



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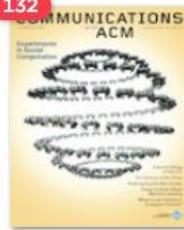
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- Shelf 2
- Shelf 3
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**Bookcase 1**

Shelf 1



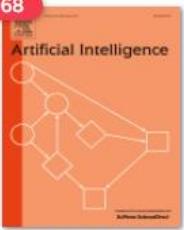
132 COMMUNICATIONS ACM

Communication of the ACM



16 Journal of the ACM

Journal of the ACM



68 Artificial Intelligence

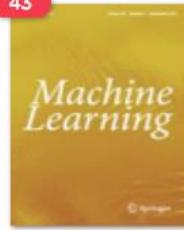
Artificial Intelligence



35 ACM Transactions on Knowledge Discovery from Data

ACM Transactions on Knowledge Discovery from Data

Shelf 2



43 Machine Learning

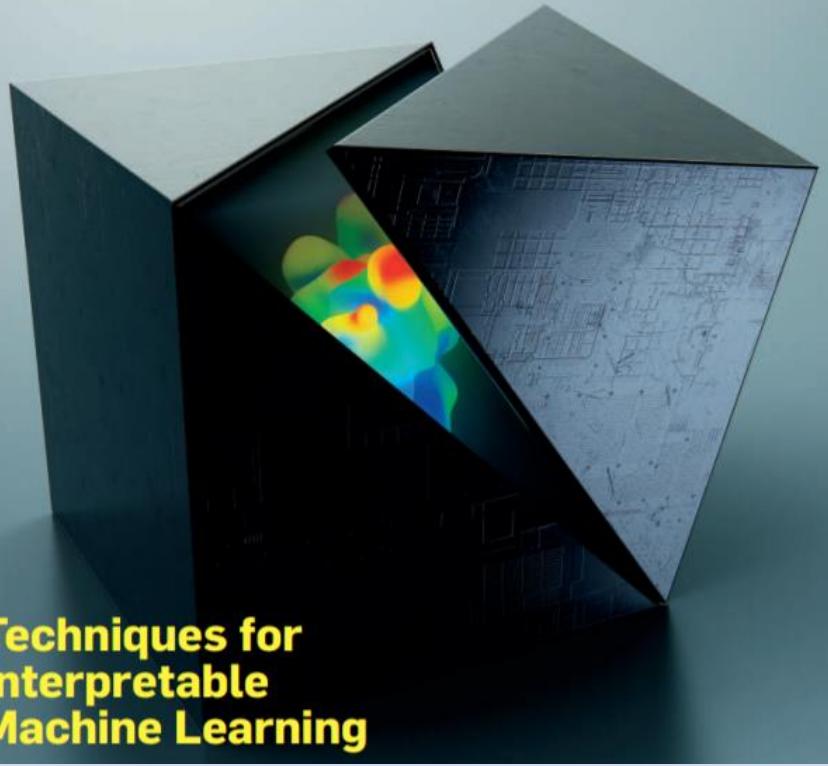
Machine Learning

66

# COMMUNICATIONS OF THE ACM

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01/2020 VOL.63 NO.01



## Techniques for Interpretable Machine Learning

V viewpoints

DOI:10.1145/3326600

James W. Davis and Jeff Hachtel

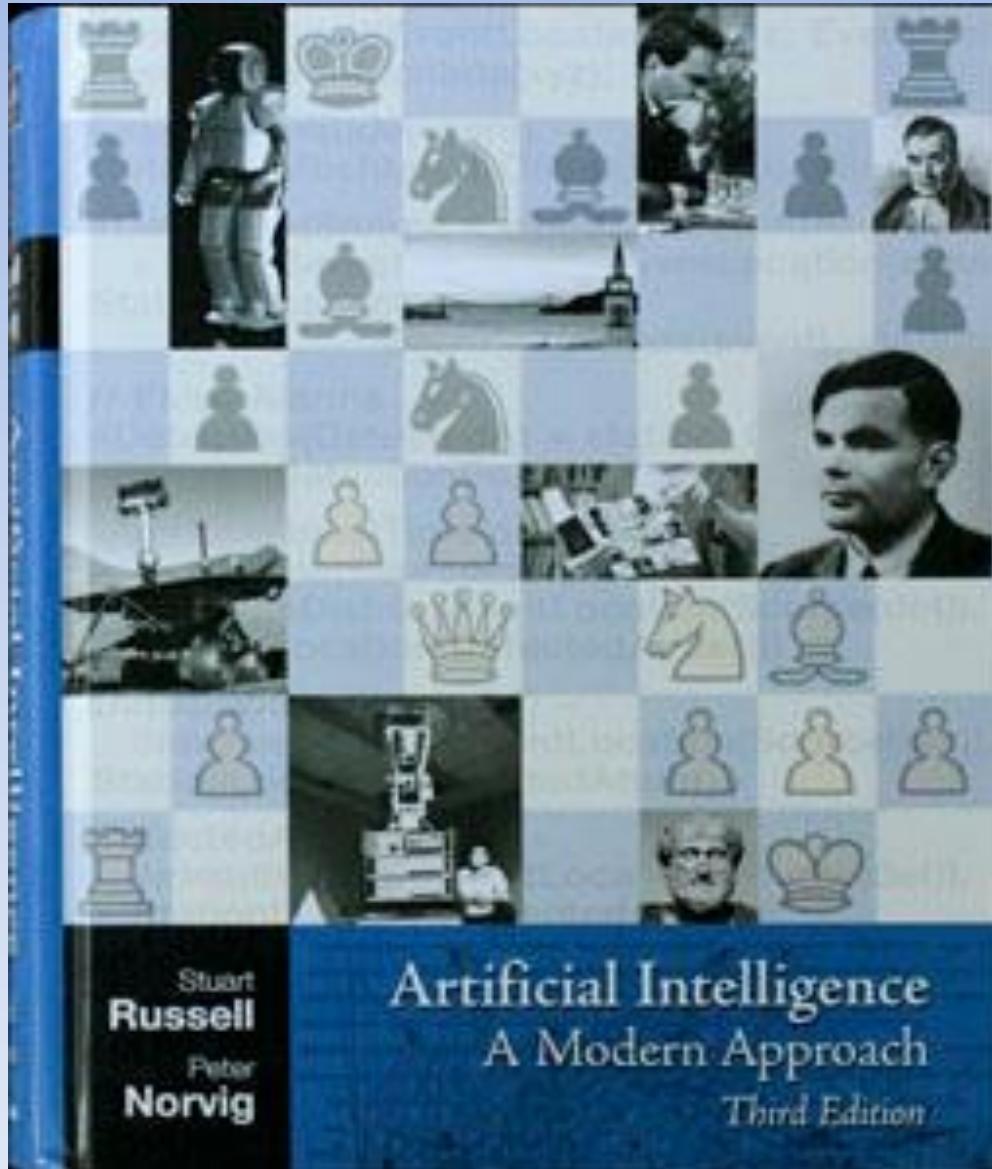
## Viewpoint A\* Search: What's in a Name?

*A search for algorithmic answers returns unique results.*

**O**RIGINALLY PUBLISHED IN 1968 by Hart, Nilsson, and Raphael,<sup>1</sup> the well-known A\* search algorithm is a foundational pathfinding



# Tentative Course Schedule



- Chapters 1, 2
- Chapter 18
- Chapters 3, 5, 6
- Chapter 7
- Chapter 13, 14?

# Chapter 1

# What is AI?

- What is Intelligence?



# **Chapter 1**

# **What is AI?**

- Introduce Science of Artificial Intelligence
- History of AI

# Chapter 2: Intelligent Agents

- Agents operate in world
- Is Agent Intelligent?

# Chapter 2:

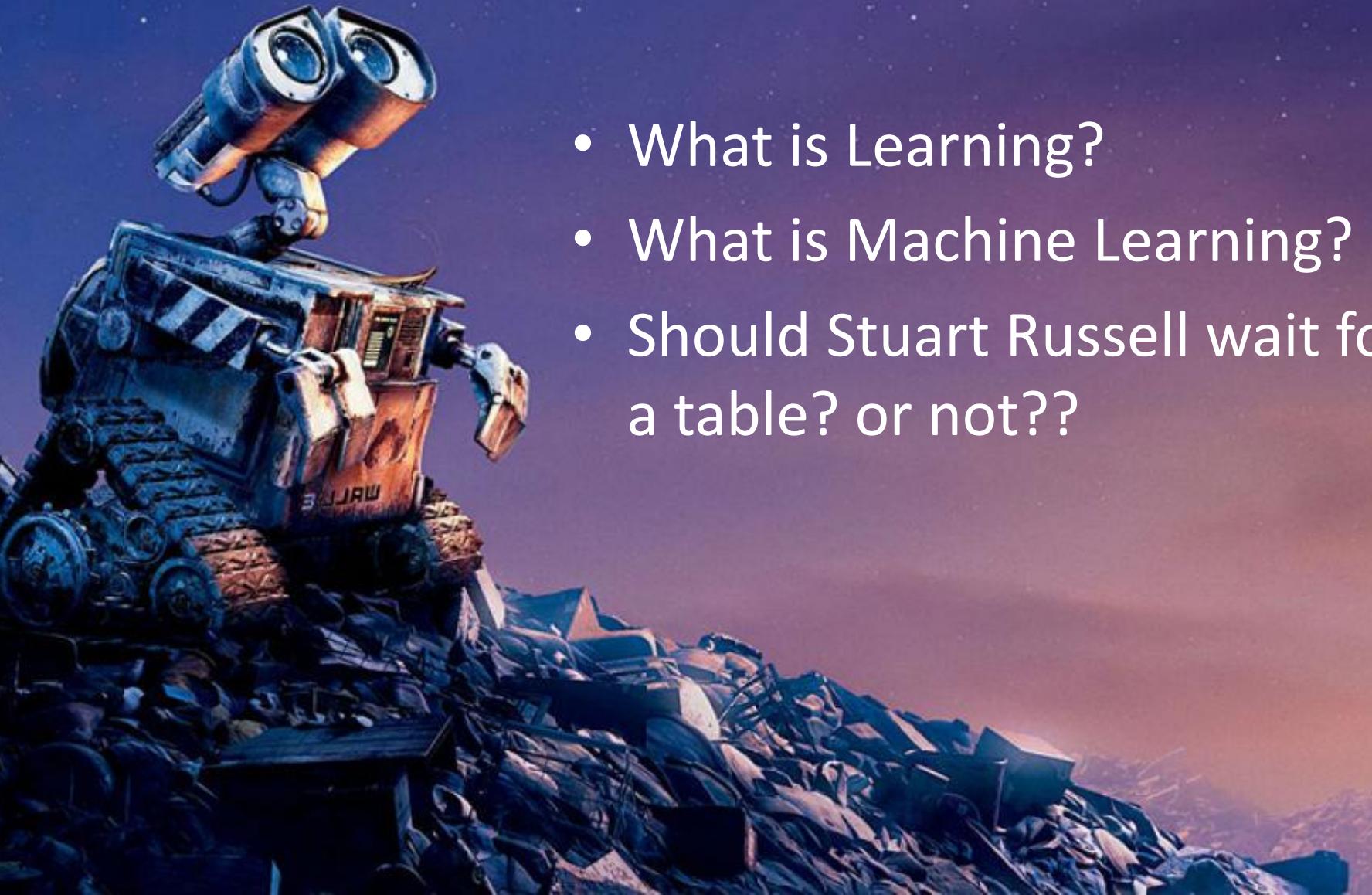
## Problem Characteristics

- Examine problems
  - Problems people solve
  - Problems not currently solved by computers
- Develop characteristics for describing problems.

# Chapter 18

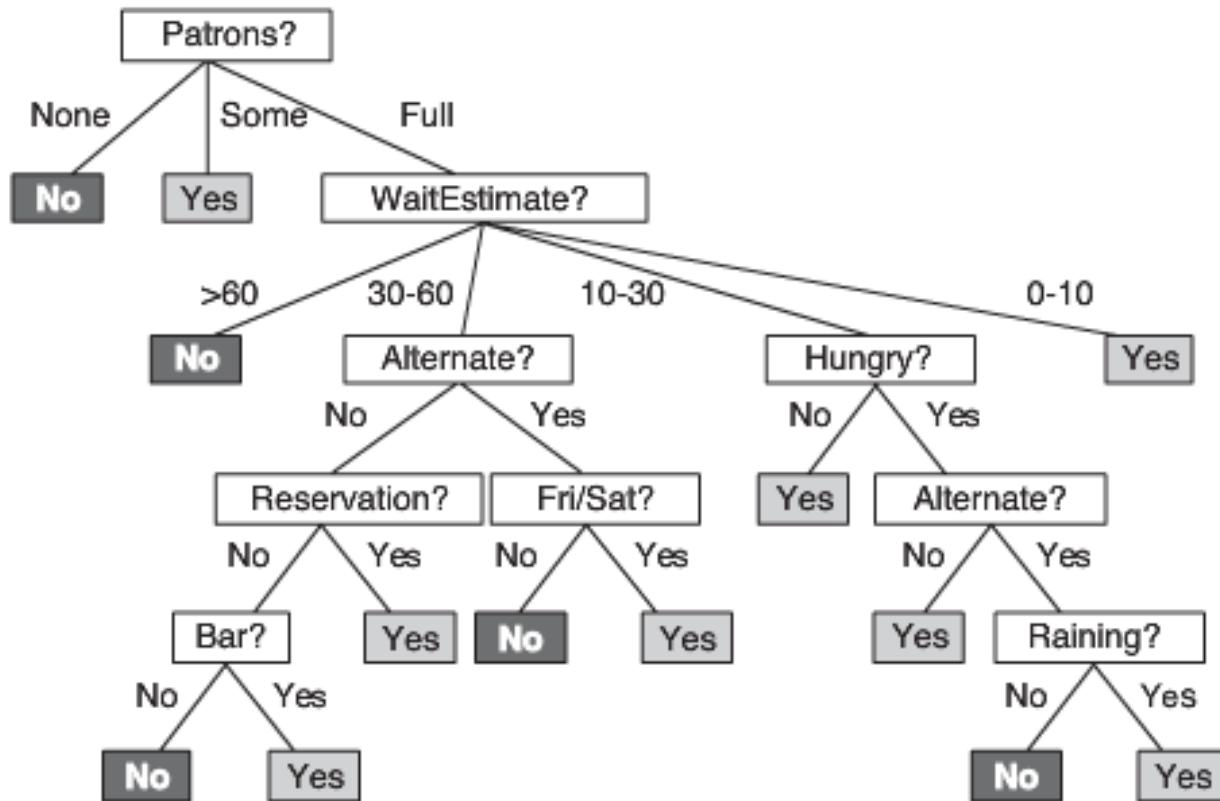
# Machine Learning

- What is Learning?
- What is Machine Learning?
- Should Stuart Russell wait for a table? or not??



# Chapter 18:

## Decision Trees



**Figure 18.2** A decision tree for deciding whether to wait for a table.

# Linear Models

- Gradient Descent
  - Learning Rate
- Logistic Regression
- Regularization
  - Regularization Parameter

# Neural Nets

- Neural Models
- Multi-Layer Models
- Training Multi-Layer Models w/  
Backpropagation
- Deep Learning

# Midterm

- Chapters 1, 2, 18

# Search

# Chapter 3:

## State-Space Search

- Uniformed Search
- Heuristics
- Informed Search
  - A\*

# Chapter 5 Adversarial Search

- Multiplayer Games
- Minimax
- Alpha Beta Cutoff



# Chapter 6: CSP

- Constraint Satisfaction Problems
- AC-3 Algorithm
- Cutsets
- Etc...



# Knowledge Representation

# Chapter 7: Logical Agents

- Propositional Logic

246

Chapter 7. Logical Agents

$P$	$Q$	$\neg P$	$P \wedge Q$	$P \vee Q$	$P \Rightarrow Q$	$P \Leftrightarrow Q$
<i>false</i>	<i>false</i>	<i>true</i>	<i>false</i>	<i>false</i>	<i>true</i>	<i>true</i>
<i>false</i>	<i>true</i>	<i>true</i>	<i>false</i>	<i>true</i>	<i>true</i>	<i>false</i>
<i>true</i>	<i>false</i>	<i>false</i>	<i>false</i>	<i>true</i>	<i>false</i>	<i>false</i>
<i>true</i>	<i>true</i>	<i>false</i>	<i>true</i>	<i>true</i>	<i>true</i>	<i>true</i>

Figure 7.8 Truth tables for the five logical connectives. To use the table to compute, for

249  $P \wedge Q$  is false, first look on the left for the row  
 $\neg P$ ). Then look in that row under the  $P \vee Q$  column

Section 7.5. Propositional Theorem Proving

$$\begin{aligned}(\alpha \wedge \beta) &\equiv (\beta \wedge \alpha) \text{ commutativity of } \wedge \\(\alpha \vee \beta) &\equiv (\beta \vee \alpha) \text{ commutativity of } \vee \\((\alpha \wedge \beta) \wedge \gamma) &\equiv (\alpha \wedge (\beta \wedge \gamma)) \text{ associativity of } \wedge \\((\alpha \vee \beta) \vee \gamma) &\equiv (\alpha \vee (\beta \vee \gamma)) \text{ associativity of } \vee \\\neg(\neg\alpha) &\equiv \alpha \text{ double-negation elimination} \\(\alpha \Rightarrow \beta) &\equiv (\neg\beta \Rightarrow \neg\alpha) \text{ contraposition} \\(\alpha \Rightarrow \beta) &\equiv (\neg\alpha \vee \beta) \text{ implication elimination} \\(\alpha \Leftrightarrow \beta) &\equiv ((\alpha \Rightarrow \beta) \wedge (\beta \Rightarrow \alpha)) \text{ biconditional elimination} \\\neg(\alpha \wedge \beta) &\equiv (\neg\alpha \vee \neg\beta) \text{ De Morgan} \\-\neg(\alpha \vee \beta) &\equiv (\neg\alpha \wedge \neg\beta) \text{ De Morgan} \\(\alpha \wedge (\beta \vee \gamma)) &\equiv ((\alpha \wedge \beta) \vee (\alpha \wedge \gamma)) \text{ distributivity of } \wedge \text{ over } \vee \\(\alpha \vee (\beta \wedge \gamma)) &\equiv ((\alpha \vee \beta) \wedge (\alpha \vee \gamma)) \text{ distributivity of } \vee \text{ over } \wedge\end{aligned}$$

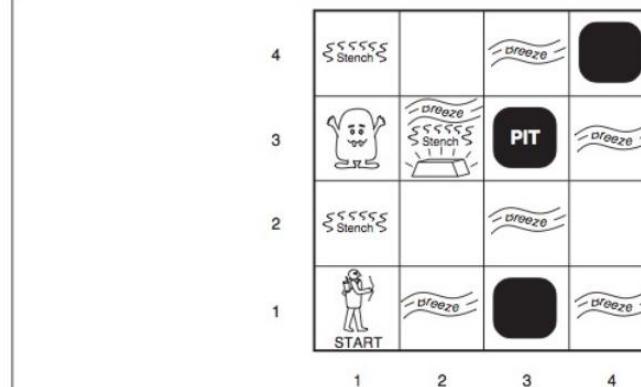
Figure 7.11 Standard logical equivalences. The symbols  $\alpha$ ,  $\beta$ , and  $\gamma$  stand for arbitrary sentences of propositional logic.

# Chapter 7: Logical Agents w/ Wumpus World

- Wumpus World

238

Chapter 7. Logical Agents



**Figure 7.2** A typical wumpus world. The agent is in the bottom left corner, facing right.

$P_{x,y}$  is true if there is a pit in  $[x, y]$ .

$W_{x,y}$  is true if there is a wumpus in  $[x, y]$ , dead or alive.

$B_{x,y}$  is true if the agent perceives a breeze in  $[x, y]$ .

$S_{x,y}$  is true if the agent perceives a stench in  $[x, y]$ .

# PROBLEMS!!!

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Chapter 7. Logical Agents

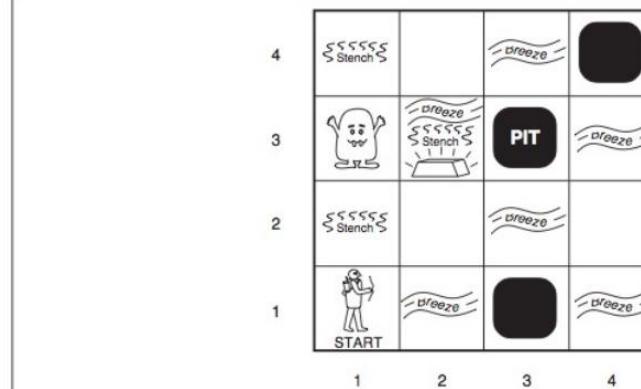


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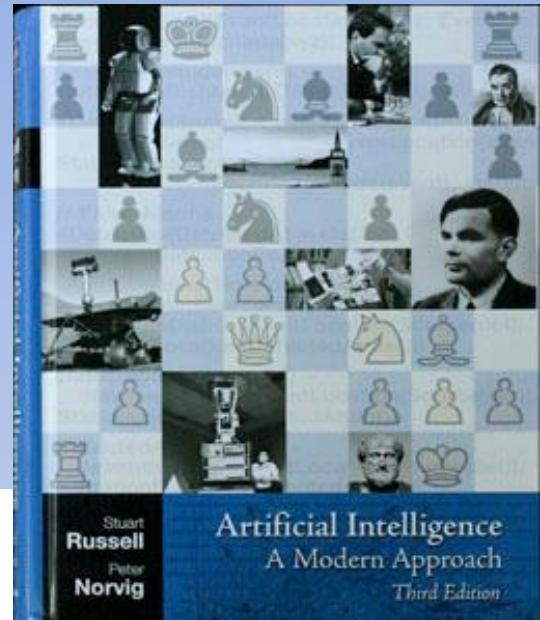
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# Uncertainty

uncertainty

# Dealing w/ Uncertainty



## IV Uncertain knowledge and reasoning

### 13 Quantifying Uncertainty

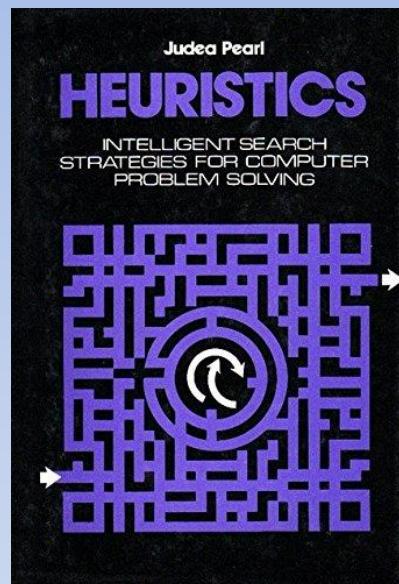
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### 14 Probabilistic Reasoning

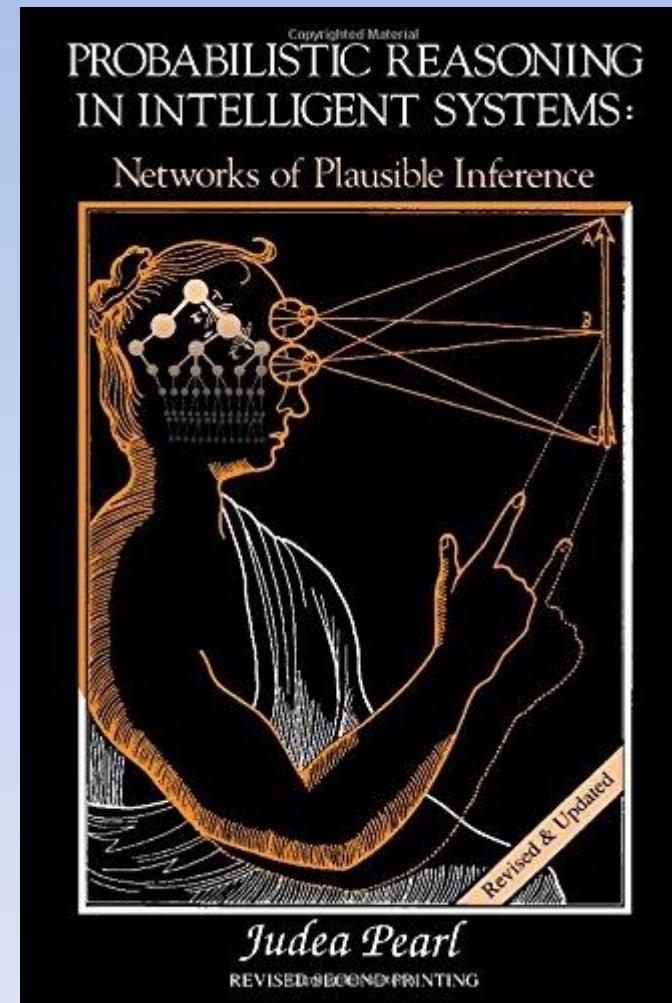
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# Judea Pearl

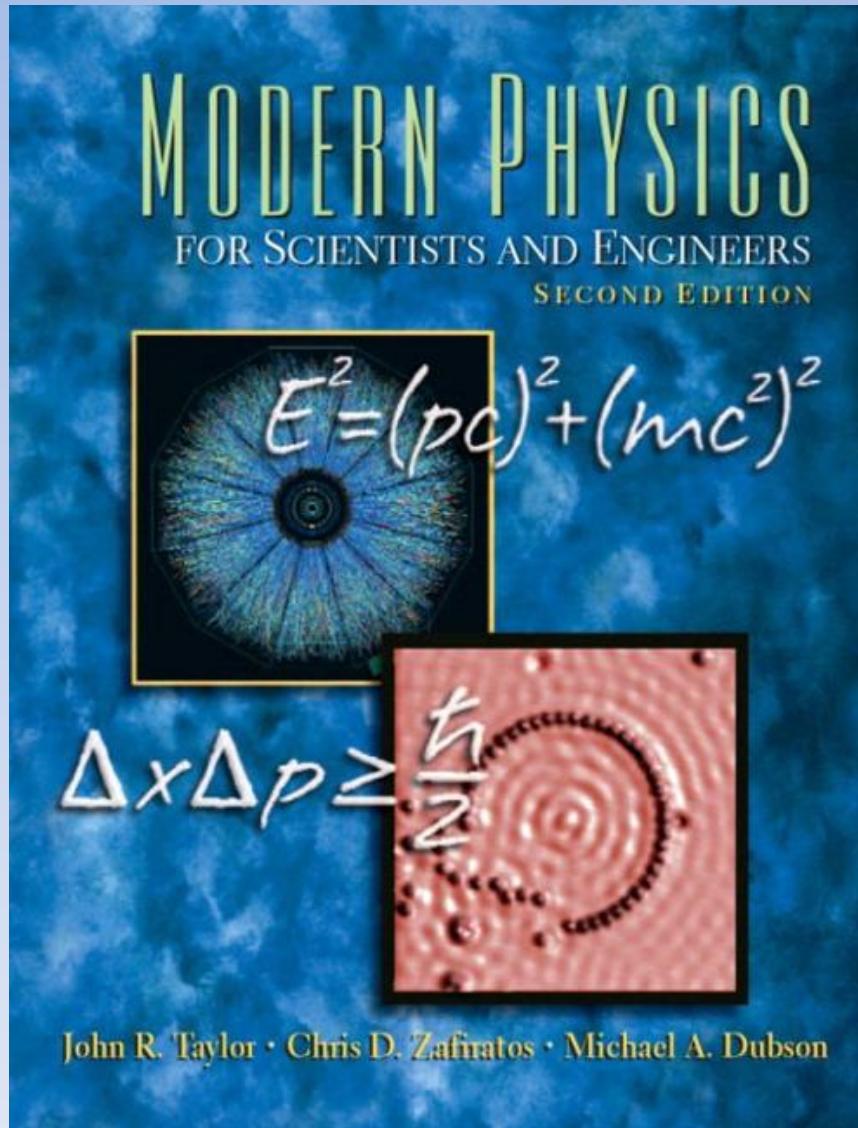
## Heuristics -> Probabilistic Reasoning



- April 1984
- September 15, 1988



# Uncertainty in Science



# Classical Mechanics



# Probability & Physics



- Albert Einstein
- Quantum mechanics is certainly imposing. But an inner voice tells me that it is not yet the real thing. The theory says a lot, but does not really bring us any closer to the secret of the "old one." **I, at any rate, am convinced that *He* does not throw dice.**
  - Letter to [Max Born](#) (4 December 1926); *The Born-Einstein Letters* (translated by Irene Born) (Walker and Company, New York, 1971)  
[ISBN 0-8027-0326-7](#).
- In a 1943 conversation with William Hermanns recorded in Hermanns' book *Einstein and the Poet*, Einstein said: "**As I have said so many times, God doesn't play dice with the world.**" ([p. 58](#)).

# Uncertainty in Science

## Schrödinger's Equation

$$i\hbar \frac{\partial}{\partial t} \psi(\mathbf{r}, t) = -\frac{\hbar^2}{2m} \nabla^2 \psi(\mathbf{r}, t) + V(\mathbf{r}, t) \psi(\mathbf{r}, t)$$

$i$  is the imaginary number,  $\sqrt{-1}$ .

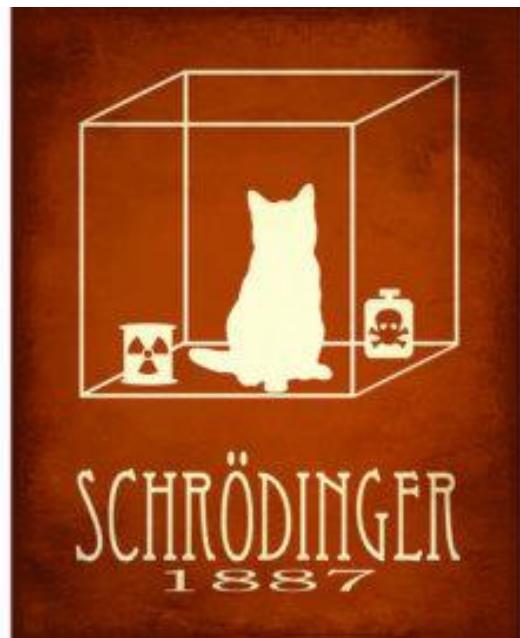
$\hbar$  is Planck's constant divided by  $2\pi$ : 1

$\psi(\mathbf{r}, t)$  is the wave function, defined over

$m$  is the mass of the particle.

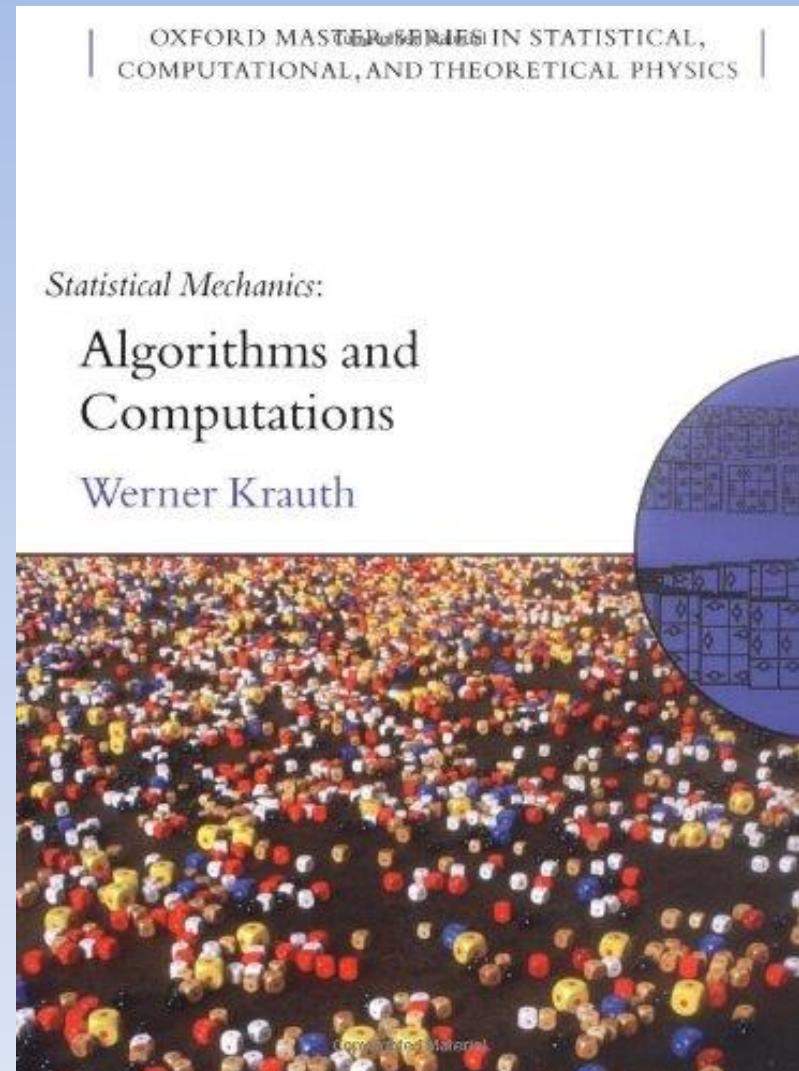
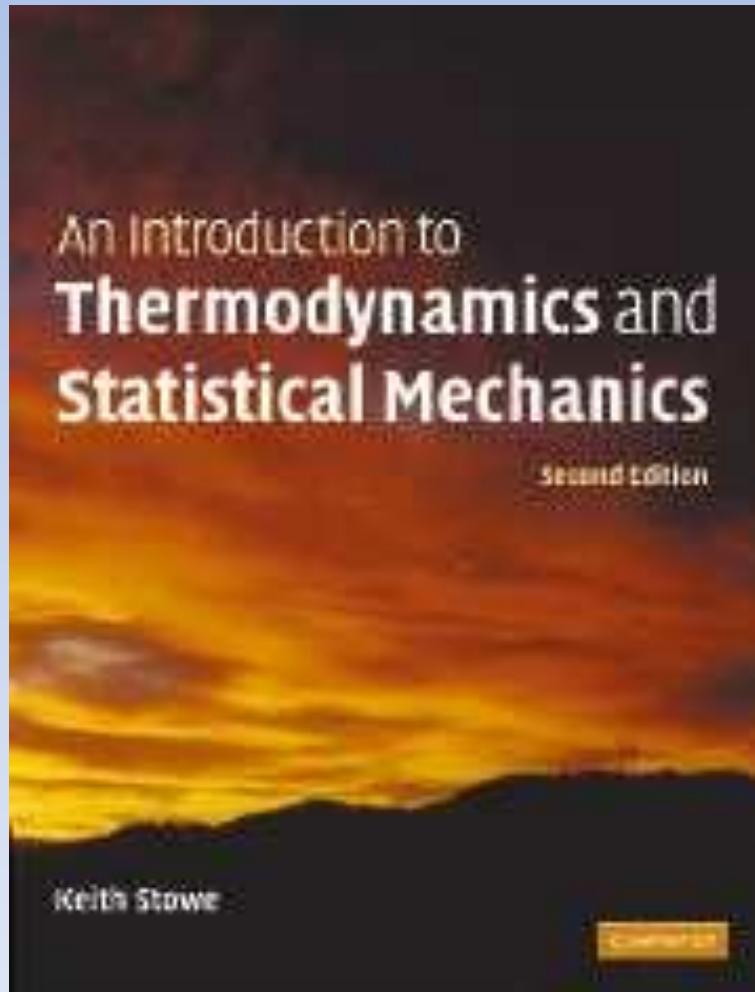
$\nabla^2$  is the Laplacian operator,  $\frac{\partial^2}{\partial x^2} + \frac{\partial^2}{\partial y^2} + \frac{\partial^2}{\partial z^2}$

$V(\mathbf{r}, t)$  is the potential energy influencing



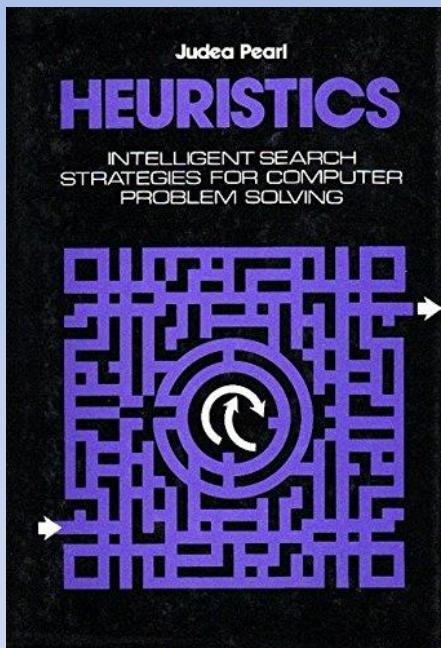
# Uncertainty in Science

## Classical Mechanics versus Statistical Mechanics

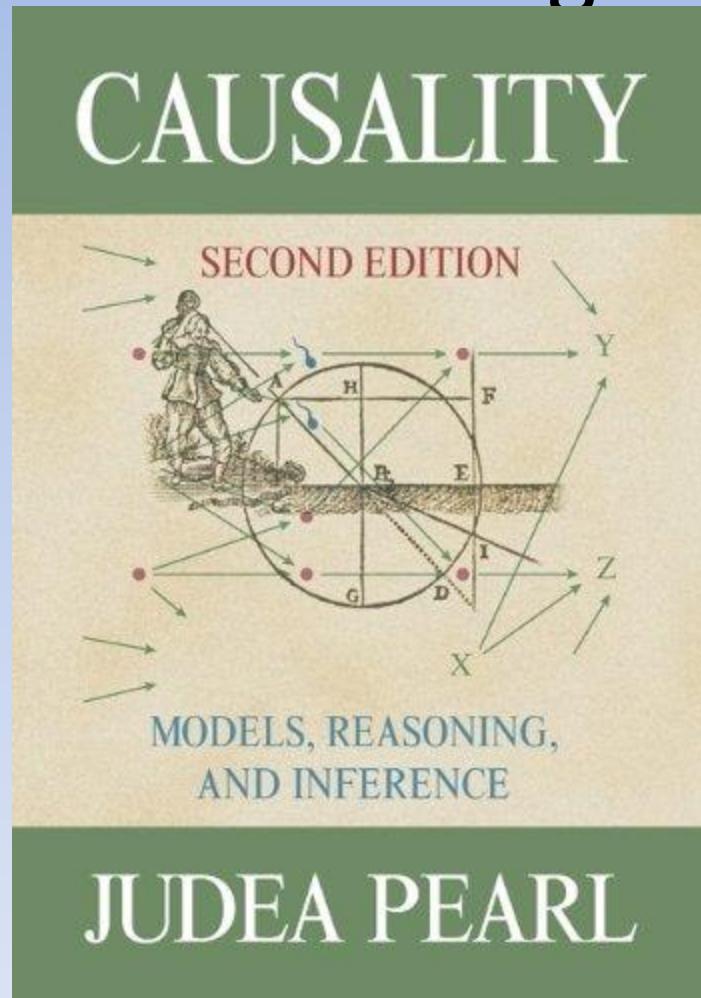
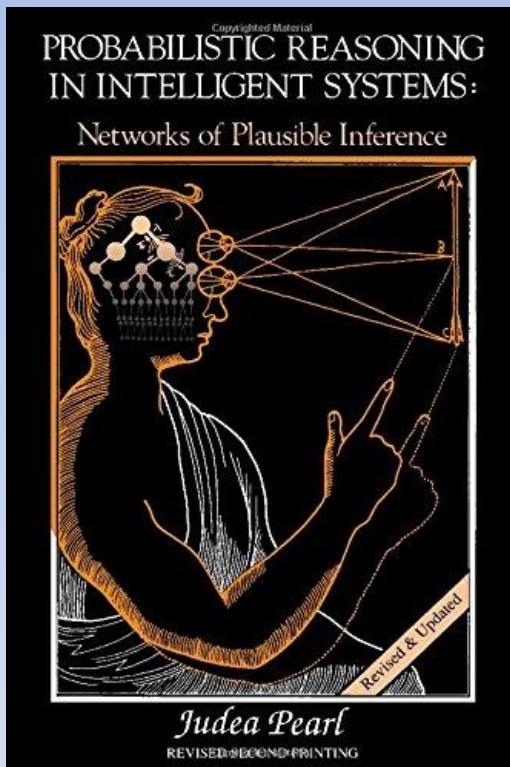


# Judea Pearl

## Heuristics -> Probabilistic Reasoning



- April 1984
- September 15, 1988



September 14<sup>th</sup>, 2009

# History of Probability (AI Text)

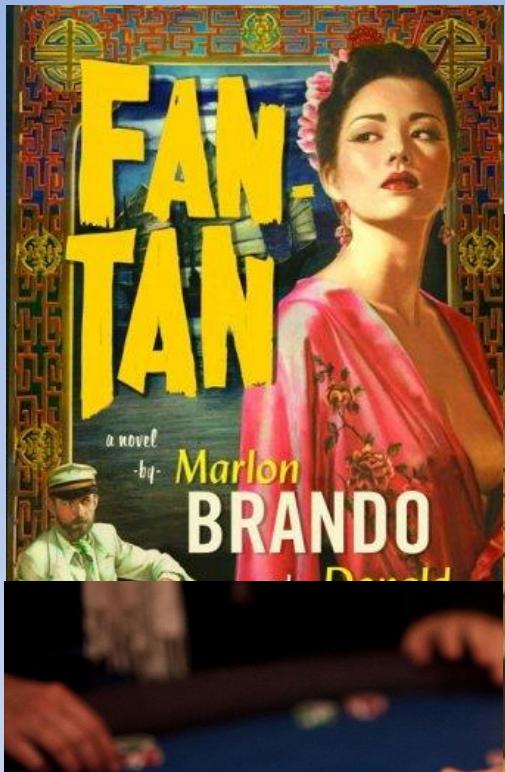
- In about 850 A.D. the Indian mathematician Mahaviracarya described how to arrange a set of bets that can't lose (what we now call a Dutch book).

## Mahāvīra (mathematician)

From Wikipedia, the free encyclopedia

Mahāvīra (or Mahaviracharya, "Mahavira the Teacher") was a 9th-century Jain mathematician from Mysore, India.<sup>[1][2][3]</sup> He was the author of *Ganitasārasaṅgraha* (or *Ganita Sara Samgraha*, c. 850), which revised the *Brāhmaś�uṭasiddhānta*.<sup>[1]</sup> He was patronised by the Rashtrakuta king Amoghavarsha.<sup>[4]</sup> He separated astrology from mathematics. It is the earliest Indian text entirely devoted to mathematics.<sup>[5]</sup> He expounded on the same subjects on which Aryabhata and Brahmagupta contended, but he expressed them more clearly. His work is a highly syncopated approach to algebra and the emphasis in much of his text is on developing the techniques necessary to solve algebraic problems.<sup>[6]</sup> He is highly respected among Indian mathematicians, because of his establishment of terminology for concepts such as equilateral, and isosceles triangle; rhombus; circle and semicircle.<sup>[7]</sup> Mahāvīra's eminence spread in all South India and his books proved inspirational to other mathematicians in Southern India.<sup>[8]</sup> It was translated into Telugu language by Pavuluri Mallana as *Saar Sangraha Ganitam*.<sup>[9]</sup>

# Uncertainty w/ Gambling Games



# Gambling Concepts

- Games of Chance
  - Poker
  - Roulette
  - Fan Tan
- Odds
  - Sports Book
  - Horse Racing

# 1565 Girolamo Cardano

- In Europe, the first significant systematic analyses were produced by Girolamo Cardano around 1565, although publication was posthumous (1663).
- Gambling Motivated



# Meaning and Probability Theory

- 20 to 1 in the 5<sup>th</sup>
  - What does 20 to 1 Mean?
  - Where does 20 come from?

# Joints & Marginals

		Intelligence		
		low	high	
Grade	A	0.07	0.18	0.25
	B	0.28	0.09	
	C	0.35	0.03	
			0.3	1.0

- $P(\text{Intelligence}=\text{high}) = ?$

# More Problems...

		Intelligence		
		low	high	
Grade	A	0.07	0.18	0.25
	B	0.28	0.09	
	C	0.35	0.03	
			0.3	1.0

- $P(\text{Intelligence}=\text{high}) = ?$

**INTRACTABILITY**

**YTBICATBATNI**

# Computational Thinking!!!

## Algorithms & Data Structures

- Representation
  - Bayesian Networks
- Algorithms
  - Exact Inference
  - Approximate Inference

# PLAN

## The Bayesian Network (Chapter 14)

- Exploit Islands of Tractability in High Dimensional Space Probability Distributions
  - Worst Case is Intractable
  - Real World Frequently NOT Worst Case
  - Efficiently Exploit properties in Real World Probability Distributions to induce Tractability
- Primary Tool:

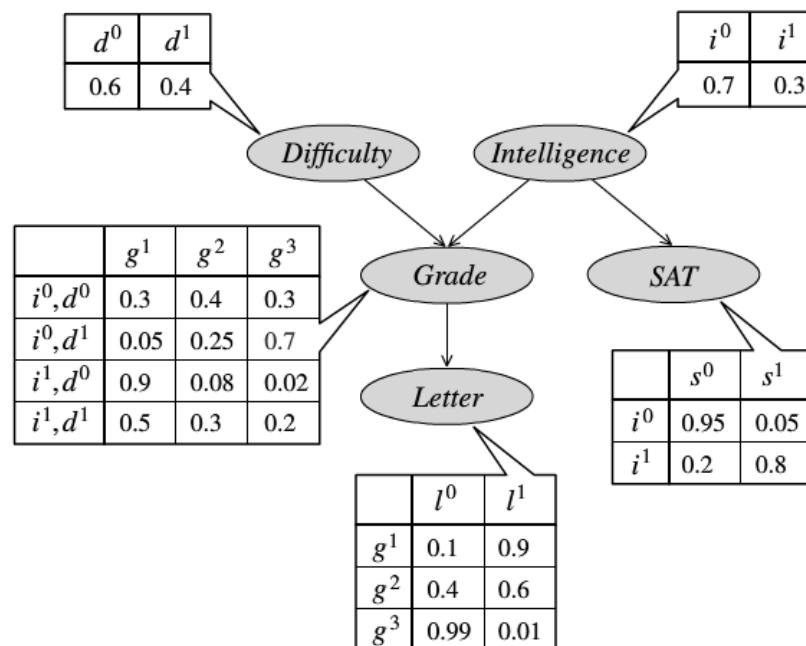
**INDEPENDENCE!**

# Bayesian Networks : (CPD's)

- Each variable is associated with a conditional probability distribution (CPD) that specifies a distribution CPD over the values of X given each possible joint assignment of values to its parents in the model.
- For a node with no parents, the CPD is conditioned on the empty set of variables.

3.2. Bayesian Networks

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# Final

- Chapters 1, 2
- Chapters 3, 5, 6
- Chapter 7
- Chapters 13, 14
- Chapter 18

# Grading

Period	Presentations/Examinations/Assignments	Points
Various	Attendance	100
	In-Class	100
	Quizzes	100
	Qwiklabs	200
	Midterm	200
	Final	300