Building Software Agents for Planning Monitoring, and Optimizing Travel

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Acknowledgements

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Introduction

Wealth of travel-related data available online Web provides unprecedented access to information to end users Abundance of computing power available We can exploit these three factors to: Support better planning of travel Provide real-time monitoring of travel plans Exploit data mining techniques to minimize problems and cost



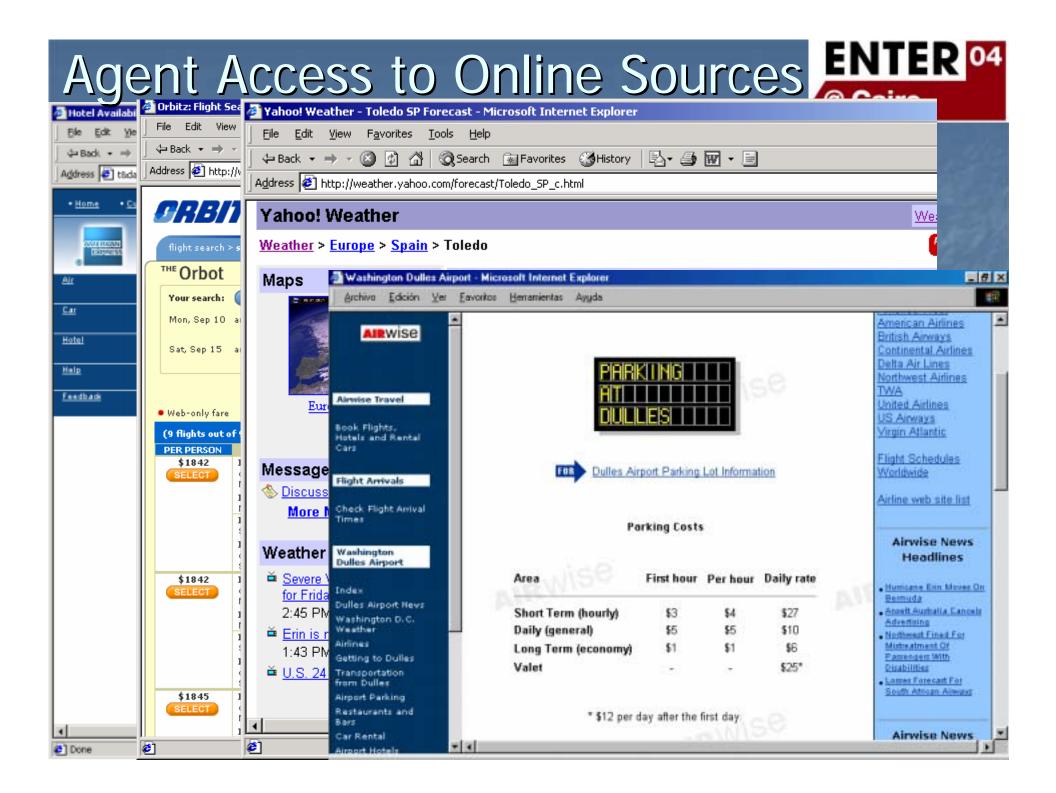
Outline

Agent Access to Online Sources
Interactive Planning of a Trip
Building Agents for Monitoring Travel
Mining Online Sources to Optimize Travel
Conclusions



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Problem: Information Not in a Usable Format

- Web pages are intended for human consumption
- Web services and XML are designed to solve this problem, but not available for most data
 - Need to turn these online sources into 'agentenabled' sources
 - Support database like querying by a software agent
 - Return information in a structured format, such as XML



Wrappers for Live Access Cairo to Online Sources

Wrapper

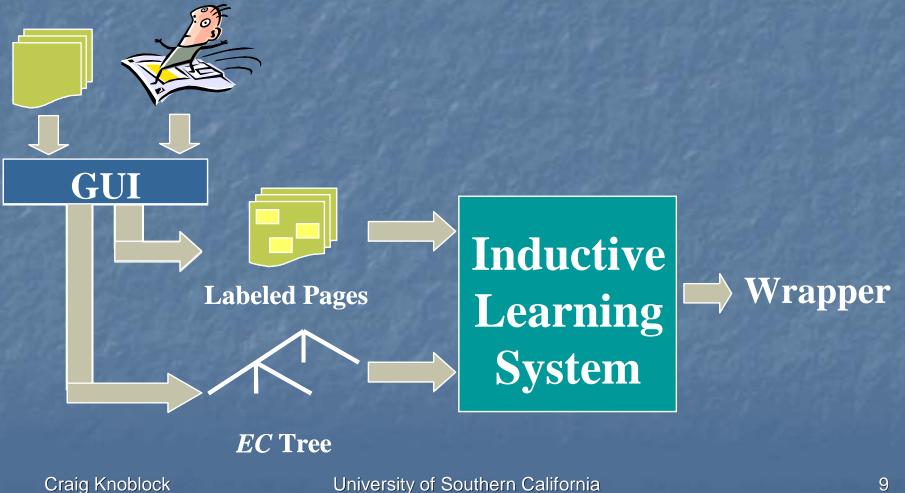
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Learning a Wrapper





Status

Almost any source on the Web can be turned into an agent-enabled source Time to construct a wrapper ranges from a few minutes to a few hours Tools are easy to learn Makes it possible to exploit the huge amount of information available online Wrapper learning technology has been licensed to Fetch Technologies, which has a commercial product available



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Interactive Trip Planning

- Current systems provide support to select flights, hotels and cars
 - Integrates the planning at the level of dates and locations
- There are many more factors involved in planning a trip
 - Which airports to fly into and out of
 - Whether to drive or take a taxi to the airport
 - How to get form the airport to the destination
 - Proximity of hotel to meeting
 - Etc...
- Ideally a system will
 - Provide all of the data required to make these decisions
 - Provide a way to consider the tradeoffs of the various choices



Heracles Constraint-based Planning

Framework for building integrated applications
 Extract and integrate data for a given task
 Live access to online sources using the wrappers
 Constraint-based decides what sources to query and how to integrate the results
 Tight integration of user choices

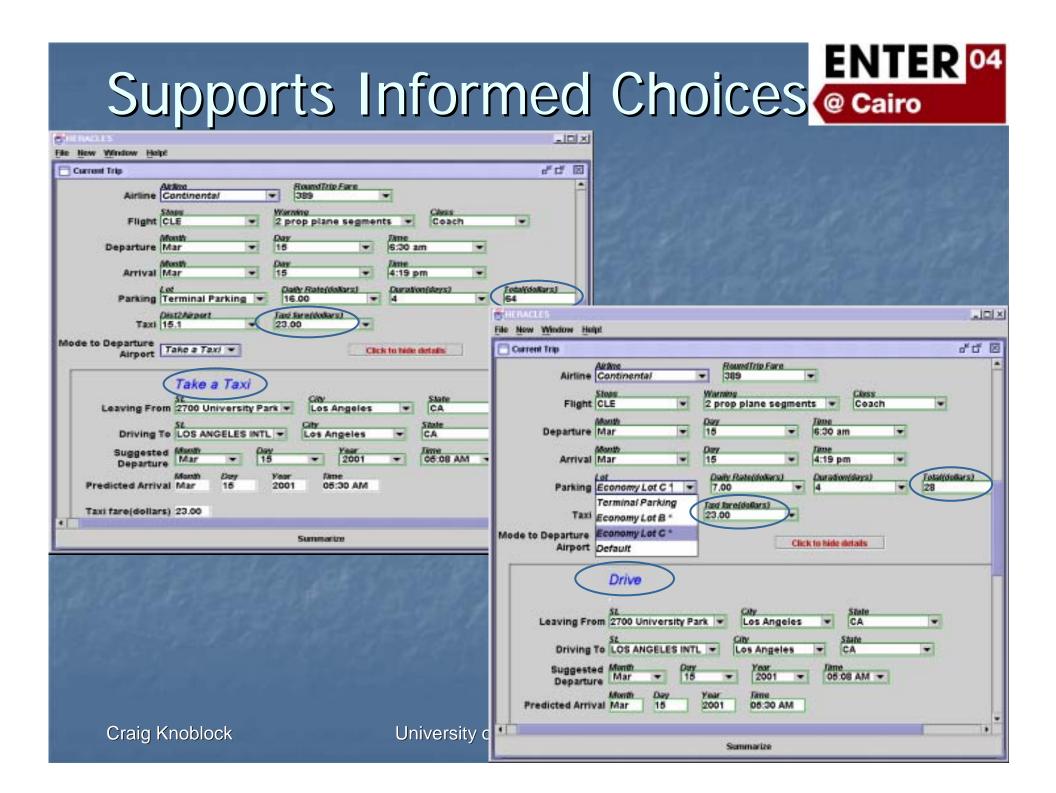
Travel Planner

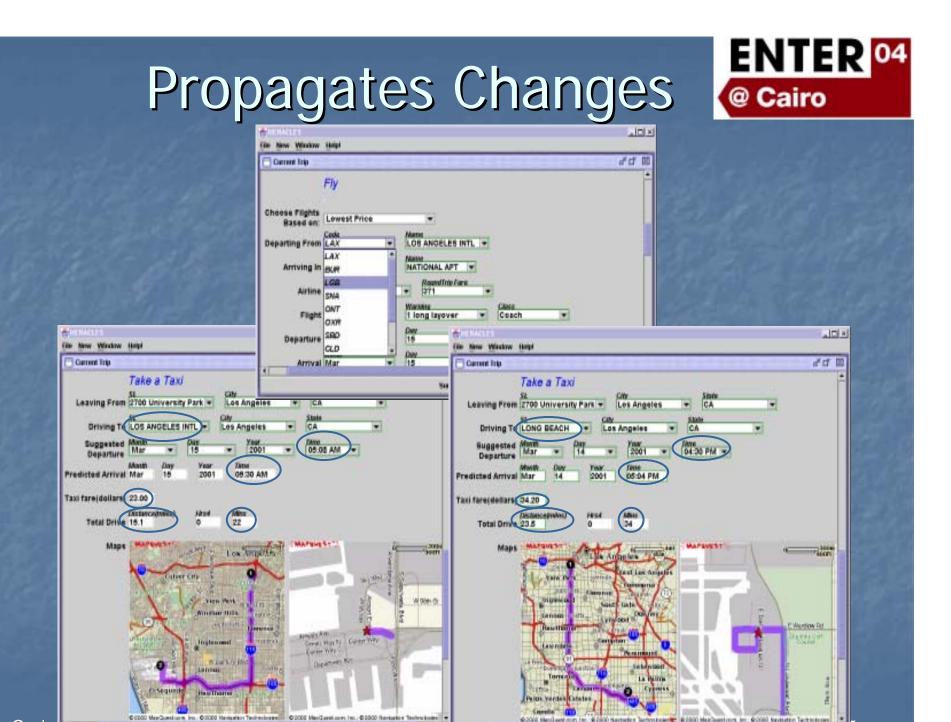


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Dynamically Updates Slots as Cairo

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Sommarize:

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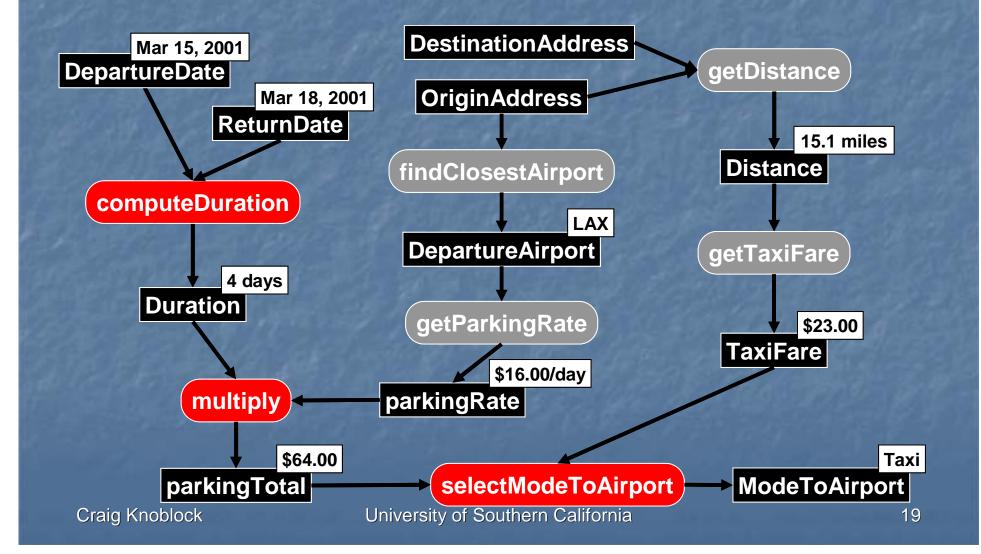
User Can Specify

High-Level Preferences

File New Window Help!	The New Window (help!
Current Trip af Cf 🖂	Current Trip of CF 🗵
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Constraint Network: Drive or Taxi?





Summary

 Integration of wide range of data from many different sources
 Tight integration of data using constraints to capture the dependencies
 Supports better decision making

 Easy to consider costs of specific choices
 Easy to compare tradeoffs



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Agents for Monitoring Travel

Many opportunities and possible problems can arise during travel

Current environment:

- Wide access to data
- Abundance of computer resources
- Availability of cell phones and portable computers
- Makes it possible to monitor all aspects of a trip

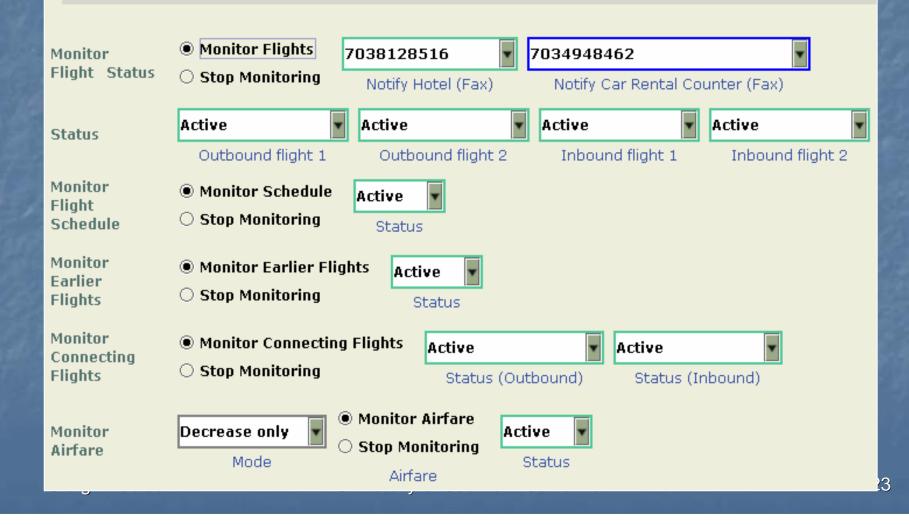
Create personal assistants that monitor your travel plan to

- exploit opportunities
- avoid problems

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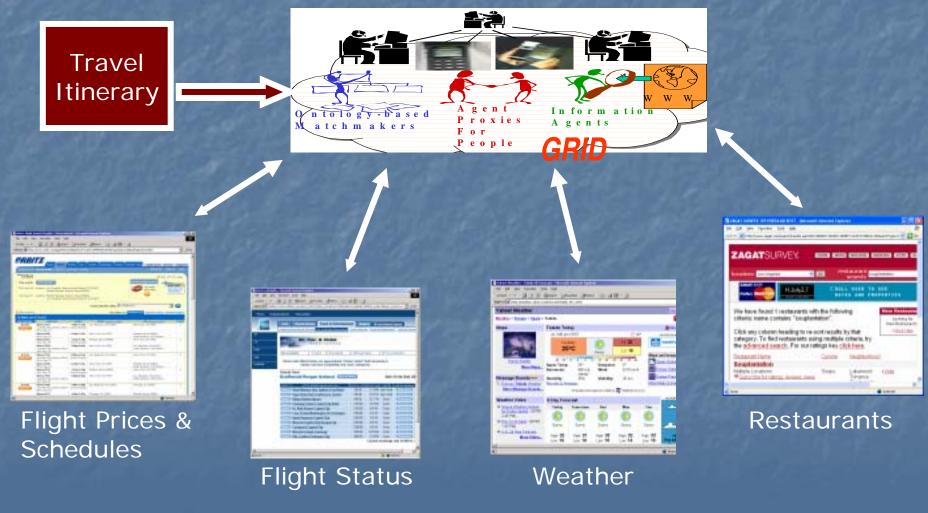
Automatically Configuring Agents

Monitoring Tasks





Agents Deployed to Monitor Travel Itinerary



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Actual Messages Sent @Cairo

Flight-Status Agent:

Flight delayed message:

Your United Airlines flight 190 has been delayed. It was originally scheduled to depart at 11:45 AM and is now scheduled to depart at 12:30 PM. The new arrival time is 7:59 PM.

Flight cancelled message:

Your Delta Air Lines flight 200 has been cancelled.

Fax to hotel message:

Attention: Registration Desk

I am sending this message on behalf of David Pynadath, who has a reservation at your hotel. David Pynadath is on United Airlines 190, which is now scheduled to arrive at IAD at 7:59 PM. Since the flight will be arriving late, I would like to request that you indicate this in the reservation so that the room is not given away.

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Actual Messages Sent (cont.)

- Airfare Agent: Airfare dropped message
 The airfare for your American Airlines itinerary
 (IAD LAX) dropped to \$281.
- Earlier-Flight Agent: Earlier flights message
 The status of your currently scheduled flight is:
 # 190 LAX (11:45 AM) IAD (7:29 PM) 45 minutes Late
 If you would like to return earlier, the following
 United Airlines flights will arrive earlier than your
 scheduled flights:

946 LAX (8:31 AM) - IAD (3:35 PM) 11 minutes Late

388 LAX (9:25 AM) - DEN (12:25 PM) 10 minutes Late # 1534 DEN (1:20 PM) - IAD (6:06 PM) On Time Craig Knoblock
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Challenges in Building Cairo Monitoring Agents

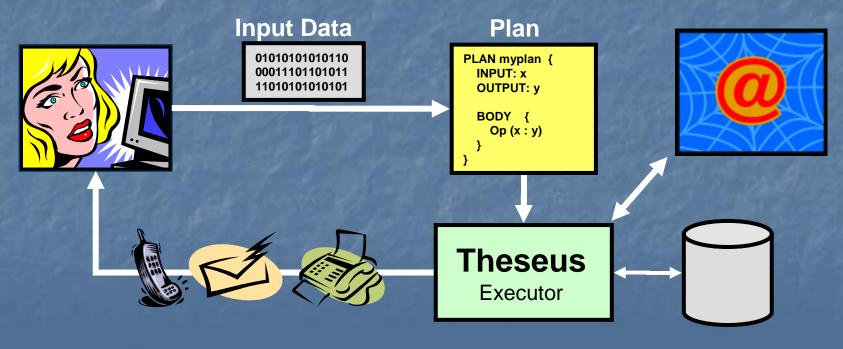
Problem

- Information gathering may involve accessing and integrating data from many sources
 Total time to execute these plans may be large
- Why?
 - Slow remote sources
 - Unpredictable network latencies
 - Binding patterns
 - Source cannot be queried until a previous query has been answered
 - Result: execution is often I/O-bound



Theseus Agent Execution System

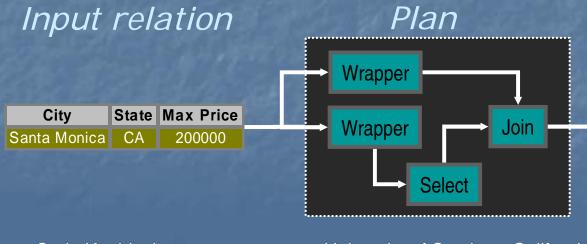
- Plan language and execution system for Web-based information integration
 - Expressive enough for monitoring a variety of sources
 - Efficient enough for real-time monitoring



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Streaming Dataflow Plans consist of a network of operators Examples: Wrapper, Select, etc. Operators produce and consume data Operators "fire" upon any input data



Output relation

Address 100 Main St., Santa Monica, 90292 520 4th St. Santa Monica, 90292 2 Ocean Blvd, Venice, 90292

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Current Work



- Challenge: How to build monitoring agents without the need to program them?
- We are developing an agent wizard that leads the user through a series of questions and then builds the required agent

Agent Wizard				
Please select the Wrappe which you want to use	er/Plan			
House Search Wrapper House Search Wrapper Geocoder Wrapper	Enter the Inputs for Geocode			
ITA Wrapper	1. ADDRESS Other Plan 2. CITYSTATE Other Plan 3. COUNTRY Constant Input			
	Back	Specify the Conditions you want to che	L 🗆 🗙	
		'Geocoder : Latitude' < 'Geocoder : Longitude' >	Plan Input : min_lat' AND Plan Input : max_lat' AND "Plan Input : min_lon' AND New Plan Input 👻 max_lon	- Agent
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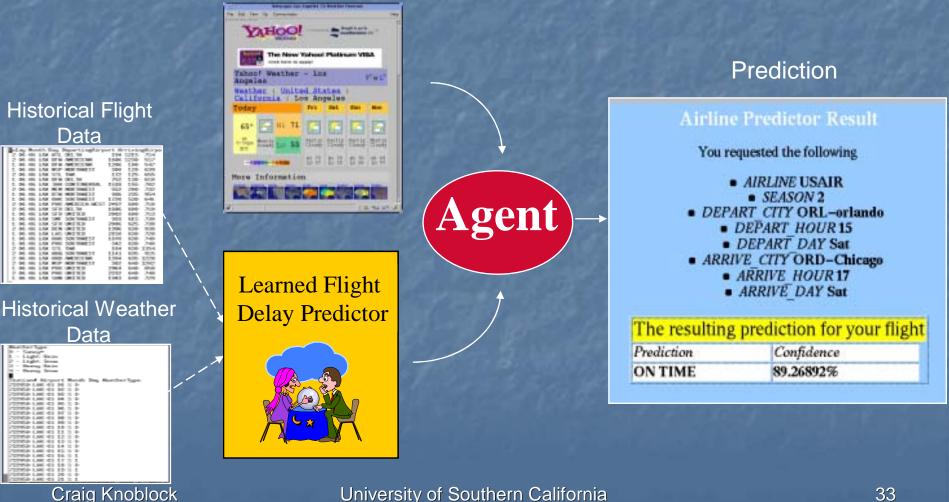


Mining Online Sources to **Optimize Travel**

- Wealth of online data provides many opportunities for data mining
- Two examples:
 - Predicting flight delays from historical flight delays and weather forecasts
 - Predicting airline prices to minimize cost



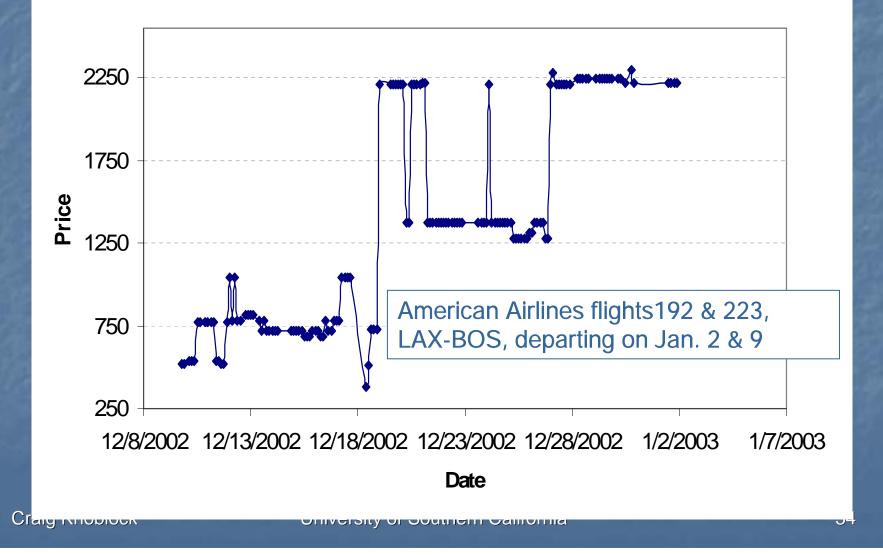
Predicting Weather-related Flight Delays



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Predicting Airline Prices





Hamlet: To Buy or Not to Buy

Collected airline flight data over several months
Developed a learning algorithm to predict whether to buy immediately or wait to buy a ticket
Exploits the fact that airline pricing is done with a relatively static, but unknown algorithm
Pricing can be learned by considering the pricing on the same flight on previous days



Data Set

Extracted data from online sources using wrappers Collected over 12,000 price observations: Lowest available fare for a one-week roundtrip LAX-BOS and SEA-IAD 6 airlines including American, United, etc. 21 days before each flight, every 3 hours



Learning Algorithm

Stacking with three base learners:
Rule learning (Ripper) (e.g., R=wait)
Time series
Q-learning (e.g., Q=buy)
Ripper used as the meta-level learner.
Output: classifies each decision point as 'buy' or 'wait'.



Experimental Results

Real price data; Simulated passengers
Learner run once per day on "past data"
Execution: label each purchase point until buy (or sell out)
Compute savings (or loss)



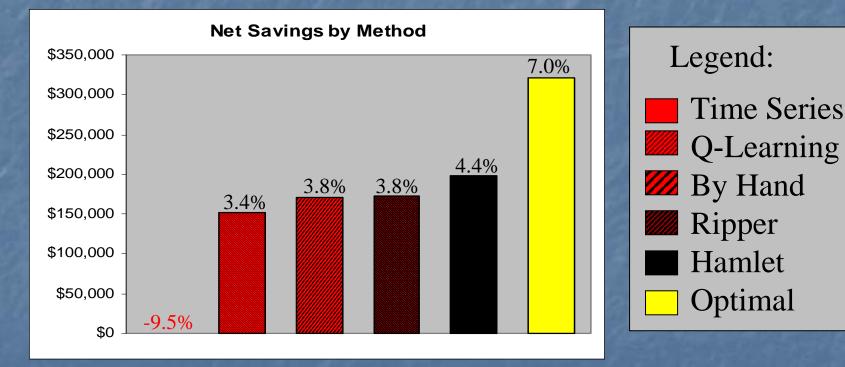
Savings by Method

Savings over "buy now".
Penalty for sell out = upgrade cost.
Total ticket cost is \$4,579,600.

Method	Savings	Losses	Upgrade Cost	% Upgrades	Net Savings	% Savings	% of Optimal
Optimal	\$320,572	\$0	\$0	0%	\$320,572	7.0%	100.0%
By hand	\$228,318	\$35,329	\$22,472	0.36%	\$170,517	3.8%	53.2%
Ripper	\$211,031	\$4,689	\$33,340	0.45%	\$173,002	3.8%	54.0%
Time Series	\$269,879	\$6,138	\$693,105	33.00%	-\$429,364	-9.5%	-134.0%
Q-learning	\$228,663	\$46,873	\$29,444	0.49%	\$152,364	3.4%	47.5%
Hamlet	\$244,868	\$8,051	\$38,743	0.42%	\$198,074	4.4%	61.8%



Savings by Method



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Upgrade Penalty

Method	Upgrade Cost	% Upgrades	
Optimal	\$0	0%	
By hand	\$22,472	0.36%	
Ripper	\$33,340	0.45%	
Time Series	\$693,105	33.00%	
Q-learning	\$29,444	0.49%	
Hamlet	\$38,743	0.42%	

Savings on "Feasible" Flights



Comparison of Net Savings (as a percent of total ticket price) on Feasible Flights

Method	Net Savings
Optimal	30.6%
By hand	21.8%
Ripper	20.1%
Time Series	25.8%
Q-learning	21.8%
Hamlet	23.8%

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Conclusions

The Web provides unprecedented access to data

- Build wrappers to turn these sources into agent-enabled sources
- Combine these sources to build an integrated travel planning system
- Automatically generate a set of agents to monitor all aspects of a travel plan
- Mine the data sources to advise a traveler about prices, chances of delays, etc.
- There are many more uses of this widely available data...



More Information

Email: knoblock@isi.edu

Papers available from my homepage: http://www.isi.edu/~knoblock







Ripper

Features include price, airline, route, hoursbefore-takeoff, etc.
Learned 20-30 rules...

IF hours-before-takeoff ≥ 252 AND price ≥ 2223 AND route = LAX-BOS THEN *wait*

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Simple Time Series

$$p_{t+1} = \frac{\sum_{i=1}^{k} \alpha(i)}{\sum_{i=1}^{k} \alpha(i)}$$



Q-learning

Natural fit to problem

 $Q(a,s) = R(a,s) + \gamma \cdot \max_{a'} (Q(a',s'))$

Q(b,s) = -price(s) $Q(w,s) = \begin{cases} -300000 & \text{if flight sells out after } s. \\ \max(Q(b,s'), Q(w,s')) & \text{otherwise.} \end{cases}$