



Planning for Information Gathering

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These slides are based in part on slides from José Luis Ambite and Rao Kambhampati, which are in turn based in part on slides from Alon Halevy.

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Planning on the Web

- Part I: Planning for Information Gathering
- Part II: Plan Execution for Information Gathering

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Outline

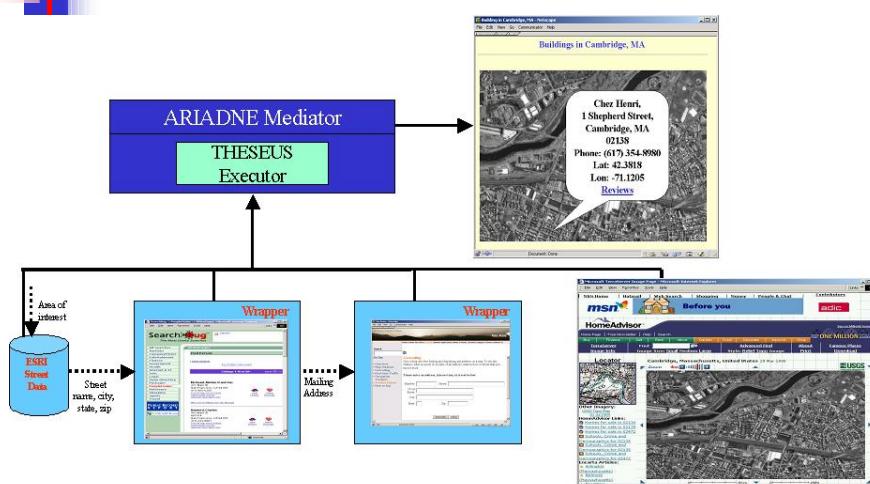
- Information Gathering
- Planning for Information Gathering
 - View Integration
 - Query Reformulation
 - Source Capabilities
- Optimizing Information Gathering Plans
 - Removing Redundant Sources
 - Optimizing Sources and Queries
- Interleaving Planning and Sensing
 - Sensing to Handle Incomplete Information
 - Sensing to Optimize Plans
- Contingent Planning for Information Gathering
- Planning to Compose Web Sources
- Discussion

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Information Gathering Example



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Wrappers for Accessing Online Information Sources

- Wrappers provide uniform querying and data extraction

The screenshot shows a web page titled "Yellow Pages" with a navigation bar including "Home", "Yellow Pages", and "Results". Below this is a "RESULTS" section for "Restaurants (1 - 1 of 1)". The result for "Casablanca Restaurant" is listed with address "220 Lincoln Boulevard, Venice, CA 90291" and phone number "(310) 392-5751". A note says it appears in the "Restaurants" category. At the bottom is a "Jump to Top" link.

NAME Casablanca Restaurant
STREET 220 Lincoln Boulevard
CITY Venice
PHONE (310) 392-5751

- State of the art in wrapper induction
 - Data extraction is based on Web page layout (Muslea *et al.* 1999, Kushmerick *et al.* 1997)
 - User labels examples of data on pages
 - Induction algorithm learns extraction rules for data

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Planning for Information Gathering

- Database query access planning
 - Specialized planner optimized for task
 - Sources are fixed
 - Mappings predefined in global schema
 - Complete plan is generated and then executed
 - Assumes closed-world and complete information
- Distributed, heterogeneous environments:
 - Sources and mappings are not fixed
 - Sources are autonomous
 - Overlapping and redundant sources
 - Sources may be incomplete
 - Sources may be unavailable
 - Additional information may be required to access a source
 - Access to sources may be costly

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Database Query Access Plans

Declarative SQL query

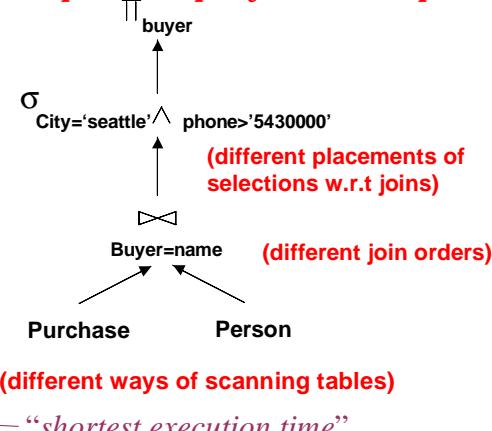
```
SELECT S.buyer  
FROM Purchase P, Person Q  
WHERE P.buyer=Q.name AND  
    Q.city='seattle' AND  
    Q.phone > '5430000'
```

Inputs:

- the query
- statistics about the data (indexes, cardinalities, selectivity factors)
- available memory

Ideally: Want to find best plan. *Practically:* Avoid worst plans!

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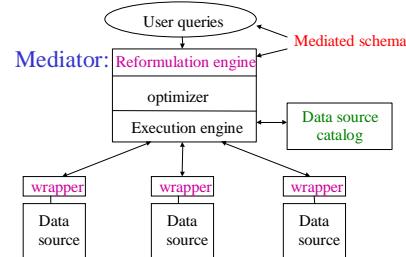
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Virtual Integration Architecture

- Leave the data in the sources
- When a query comes in:
 - Determine the relevant sources to the query
 - Break down the query into sub-queries for the sources
 - Get the answers from the sources, and combine them appropriately
- Data is fresh. Approach scalable
- Issues:
 - Relating Sources & Mediator
 - Reformulating the query
 - Efficient planning & execution



Garlic [IBM], Hermes[UMD]; Tsimmis, InfoMaster[Stanford]; DISCO[INRIA]; Information Manifold [AT&T]; SIMS/Ariadne[USC]; Emerac/Havasu[ASU]

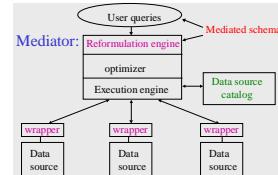
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Desiderata for Relating Source-Mediator Schemas

- **Expressive power:** distinguish between sources with closely related data. Hence, be able to prune access to irrelevant sources.
- **Easy addition:** make it easy to add new data sources.
- **Reformulation:** be able to reformulate a user query into a query on the sources efficiently and effectively.
- **Nonlossy:** be able to handle all queries that can be answered by directly accessing the sources



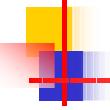
Reformulation

- Given:
 - A query Q posed over the mediated schema
 - Descriptions of the data sources
- Find:
 - A query Q' over the data source relations, such that:
 - Q' provides only *correct answers* to Q , and
 - Q' provides *all possible answers* to Q given the sources.

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Source Descriptions

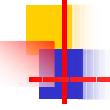
Elements of source descriptions:

- Contents: source contains movies, directors, cast.
- Constraints: only movies produced after 1965.
- Completeness: contains *all* American movies.
- Capabilities:
 - Negative: source requires movie title or director as input
 - Positive: source can perform selections, joins, ...

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Approaches to Specification of Source Descriptions

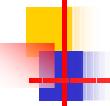
- **Global-as-View (GAV):**
Mediator relation defined as a view over source relations
Ex: TSIMMIS (Stanford), HERMES (Maryland)
- **Local-as-View (LAV):**
Source relation defined as view over mediator relations
Ex: Information Manifold (AT&T), Tukwila(UW), InfoMaster (Stanford), Ariadne (USC)

View ~ named query ~ logical formula

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Views and Conjunctive Queries

```
CREATE VIEW Big-LA-buyers AS  
    SELECT buyer, seller, price  
    FROM Person, Purchase  
    WHERE Person.city = "Los Angeles" AND  
          Person.name = Purchase.buyer AND  
          Purchase.price > 10000
```

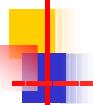
big-LA-buyers(Buyer, Seller, Price) :-
 person(Buyer, "Los Angeles"),
 purchase(Buyer, Seller, Product, Price),
 Price > 10000.

Datalog rule ~ view definition
Rule body ~ select-from-where construct of SQL

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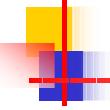
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Query Reformulation

Problem: rewrite the user query expressed in the mediated schema into a query expressed in the source schemas

Given a query Q in terms of the mediated-schema relations, and descriptions of the information sources,

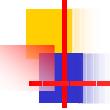
Find a query Q' that uses only the source relations, such that

- $Q' \models Q$ (i.e., answers are correct; i.e., $Q' \subseteq Q$) and
- Q' provides all possible answers to Q given the sources

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Global-as-View (GAV)

Each mediator relation is defined as a view over source relations.

```
MovieActor(title,actor) ←  
    DB1(id,title,actor,year)  
MovieActor(title,actor) ←  
    DB2(title,director,actor,year)  
MovieReview(title, review) ←  
    DB1(id,title,actor,year) ^ DB3(id,review)
```

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Query Reformulation in GAV

Query reformulation = rule unfolding+simplification

Query: *Find reviews for ‘DeNiro’ movies*

$q(\text{title}, \text{review}) :- \text{MovieActor}(\text{title}, \text{'DeNiro'}),$
 $\quad \text{MovieReview}(\text{title}, \text{review})$

1. $q'(\text{title}, \text{review}) :- \text{DB1}(\text{id}, \text{title}, \text{'DeNiro'}, \text{year}),$ Redundant
 ~~$\quad \text{DB1}(\text{id}, \text{title}, \text{actor}, \text{year}), \text{DB3}(\text{id}, \text{review})$~~

2. $q'(\text{title}, \text{review}) :-$ Redundant
wrt 1
 ~~$\quad \text{DB2}(\text{title}, \text{director}, \text{'DeNiro'}, \text{year}),$~~
 ~~$\quad \text{DB1}(\text{id}, \text{title}, \text{actor}, \text{year}), \text{DB3}(\text{id}, \text{review})$~~

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Local-as-View (LAV)

- Each source relation is defined as a view over mediator relations

$V1(\text{title}, \text{year}, \text{director}) \xrightarrow{\subseteq} \text{Movie}(\text{title}, \text{year}, \text{director}, \text{genre})$
 $\wedge \text{American}(\text{director}) \wedge \text{year} \geq 1960 \wedge \text{genre} = \text{'Comedy'}$

$V2(\text{title}, \text{review}) \rightarrow \text{Movie}(\text{title}, \text{year}, \text{director}, \text{genre}) \wedge$
 $\text{year} \geq 1990 \wedge \text{MovieReview}(\text{title}, \text{review})$

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Query Reformulation in LAV

Query: *Reviews for comedies produced after 1950*

$q(\text{title}, \text{review}) :- \text{Movie}(\text{title}, \text{year}, \text{director}, \text{'Comedy'}), \text{year} \geq 1950, \text{MovieReview}(\text{title}, \text{review})$

Reformulated query:

$q'(\text{title}, \text{review}) :- V1(\text{title}, \text{year}, \text{director}), V2(\text{title}, \text{review})$

$$q' \subseteq q$$

$V1(\text{title}, \text{year}, \text{director}) \rightarrow \text{Movie}(\text{title}, \text{year}, \text{director}, \text{genre}) \wedge \text{American}(\text{director}) \wedge \text{year} \geq 1960 \wedge \text{genre} = \text{'Comedy'}$

$V2(\text{title}, \text{review}) \rightarrow \text{Movie}(\text{title}, \text{year}, \text{director}, \text{genre}) \wedge \text{year} \geq 1990 \wedge \text{MovieReview}(\text{title}, \text{review})$

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Inverse-Rules Algorithm

[Duschka+1997]

Idea: Construct an equivalent logic program which evaluation yields the answer to the query

- The antecedent of the query and views is in term of mediator predicates
- Would like to have source predicates in antecedent so that program can be evaluated

⇒ Invert the rules

(simply by using standard logical manipulations)

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The Inverse-Rules Algorithm: Example

$\forall \text{dept, course} \rightarrow \text{Enrolled}(\text{student}, \text{dept}) \wedge \text{Registered}(\text{student}, \text{course})$

$$a \rightarrow b \equiv \neg a \vee b$$

$$\begin{aligned} & \forall D, C [v1(D, C) \rightarrow \exists S [e(S, D) \wedge r(S, C)]] \\ & \equiv \neg v1(D, C) \vee [e(f(D, C), D) \wedge r(f(D, C), C)] \\ & \equiv [\neg v1(D, C) \vee e(f(D, C), D)] \wedge [\neg v1(D, C) \vee r(f(D, C), C)] \\ & \equiv [v1(D, C) \rightarrow e(f(D, C), D)] \wedge [v1(D, C) \rightarrow r(f(D, C), C)] \\ & \equiv \\ & \quad e(f(D, C), D) \leftarrow v1(D, C) \\ & \quad r(f(D, C), C) \leftarrow v1(D, C) \end{aligned}$$

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The Inverse-Rules Algorithm: Example

$q(D) \leftarrow \text{Enrolled}(S, D) \wedge \text{Registered}(S, "DB")$
 $v1(D, C) \rightarrow \text{Enrolled}(S, D) \wedge \text{Registered}(S, C)$

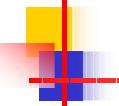
$q(D) \leftarrow \text{Enrolled}(S, D) \wedge \text{Registered}(S, "DB")$
 $\text{Enrolled}(f(D, C), D) \leftarrow v1(D, C)$
 $\text{Registered}(f(D, C), C) \leftarrow v1(D, C)$
 $q(D) \leftarrow v1(D, "DB")$

$\text{Ext}(v1) = \{("CS", "DB"), ("EE", "DB"), ("CS", "AI")\}$
 $\text{Ext}(q) = \{("CS"), ("EE")\}$

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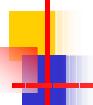
GAV VS. LAV

- Not modular
 - Addition of new sources changes the mediated schema
 - Can be awkward to write mediated schema without loss of information
 - Query reformulation easy
 - reduces to view unfolding (polynomial)
 - Can build hierarchies of mediated schemas
 - Best when
 - Few, stable, data sources
 - well-known to the mediator (e.g. corporate integration)
 - Garlic, TSIMMIS, HERMES
- Modular--adding new sources is easy
 - Very flexible--power of the entire query language available to describe sources
 - Reformulation is hard
 - Involves answering queries only using views (can be intractable)
 - Best when
 - Many, relatively unknown data sources
 - possibility of addition/deletion of sources
 - Information Manifold, InfoMaster, Emerac

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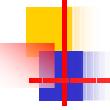
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Modeling Source Capabilities

Negative capabilities:

- A web-site may require certain inputs (in an HTML form) to answer a query
- Need to consider only valid query execution plans

Positive capabilities:

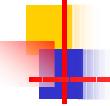
- A source may be database (understands SQL)
- Need to decide the placement of operations according to capabilities

Problem: how to describe and exploit source capabilities

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Negative Capabilities: Binding Patterns

Sources:

AAAIdb^f(X) → AAAIPapers(X)

CitationDB^{bf}(X,Y) → Cites(X,Y)

AwardDB^b(X) → AwardPaper(X)

Query: find all the award winning papers:

$q(X) \leftarrow \text{AwardPaper}(X)$

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Recursive Rewritings

$q(X) \leftarrow \text{AwardPaper}(X)$

- Problem: *Unbounded* union of conjunctive queries

$q_1(X) \leftarrow \text{AAAIdb}(X), \text{AwardDB}(X)$

$q_1(X) \leftarrow \text{AAAIdb}(X_1), \text{CitationDB}(X_1, X), \text{AwardDB}(X)$

...

$q_1(X) \leftarrow \text{AAAIdb}(X_1), \text{CitationDB}(X_1, X_2), \dots, \text{CitationDB}(X_n, X), \text{AwardDB}(X)$

- Solution: Recursive Rewriting

$\text{papers}(X) \leftarrow \text{AAAIdb}(X)$

$\text{papers}(X) \leftarrow \text{papers}(Y), \text{CitationDB}(Y, X)$

$q'(X) \leftarrow \text{papers}(X), \text{AwardDB}(X)$

$\text{AAAIdb}^{\textbf{f}}(X) \rightarrow \text{AAAIpapers}(X)$

$\text{CitationDB}^{\textbf{bf}}(X, Y) \rightarrow \text{Cites}(X, Y)$

$\text{AwardDB}^{\textbf{b}}(X) \rightarrow \text{AwardPaper}(X)$

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Inverse-Rules Algorithm Binding Patterns

Sources:

$\text{AAAIdb}^{\textbf{f}}(X) \rightarrow \text{AAAIpapers}(X)$

$\text{CitationDB}^{\textbf{bf}}(X, Y) \rightarrow \text{Cites}(X, Y)$

$\text{AwardDB}^{\textbf{b}}(X) \rightarrow \text{AwardPaper}(X)$

Query: find all the award winning papers:

$q(X) \leftarrow \text{AwardPaper}(X)$

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Inverse-Rules Algorithm Inverse + Domain Rules (1)

Inverted Rules:

$\text{AAAPapers}(X) \leftarrow \text{AAIdb}(X)$
 $\text{Cites}(X, Y) \leftarrow \text{dom}(X) \wedge \text{CitationDB}(X, Y)$
 $\text{AwardPaper}(X) \leftarrow \text{dom}(X) \wedge \text{AwardDB}(X)$

Domain Rules:

$\text{dom}(Y) \leftarrow \text{dom}(X) \wedge \text{CitationDB}(X, Y)$
 $\text{dom}(X) \leftarrow \text{AAIdb}(X)$

Query:

$q(X) \leftarrow \text{AwardPaper}(X)$

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Inverse-Rules Algorithm Inverse + Domain Rules (2)

Simplifying the program:

$q(X) \leftarrow \text{paper}(X) \wedge \text{AwardDB}(X)$
 $\text{paper}(Y) \leftarrow \text{paper}(X) \wedge \text{CitationDB}(X, Y)$
 $\text{paper}(X) \leftarrow \text{AAIdb}(X)$

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Managing Source Overlap

- Often, sources on the Internet have overlapping contents
 - The overlap is *not* centrally managed (unlike DDBMS—data replication etc.)
- Reasoning about overlap is important for plan optimality
 - We cannot possibly call all potentially relevant sources!
- Qns: How do we characterize and exploit source overlap?

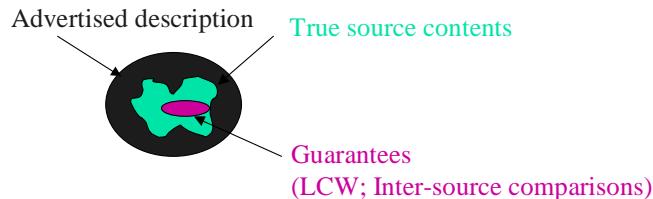
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Local Completeness Information

- If sources are incomplete, we may need to look at all of them
- Often, sources are *locally complete*
- Movie(title, director, year) complete for years after 1960, or for American directors
- **Question:** given a set of local completeness statements, is a query Q' a complete answer to Q?



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Using LCW rules to minimize plans

Basic Idea:

- If reformulation of Q leads to a union of conjunctive plans
 - $P_1 \vee P_2 \vee \dots \vee P_k$
 - Then, if P_1 is “complete” for Q (under the given LCW information), then we can minimize the reformulation by pruning $P_2 \dots P_k$
 - $[P_1 \wedge LCW]$ contains $P_1 \vee P_2 \vee \dots \vee P_k$
- [Duschka, AAAI-97]
- For Recursive Plans (obtained when the sources have access restrictions)
 - We are allowed to remove a rule r from a plan P , if the “complete” version of r is already contained in $P-r$

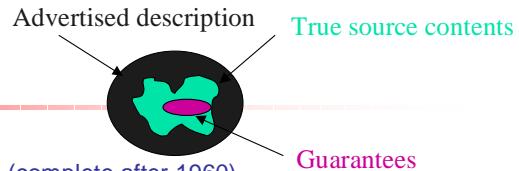
Emerac [Lambrecht & Kambhampati, 99]

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Example



- S1: Movie(title, director, year) (complete after 1960)
 $S1(T,D,Y) \rightarrow M(T,D,Y)$
- S2: Show(title, theater, city, hour)(complete for Seattle)
 $S2(T,Th,C,H) \rightarrow Sh(T,Th,C,H)$
LCW: $S2(T,Th,C,H) \leftarrow Sh(T,Th,C,H) \& C = \text{Seattle}$
- S3: Show(title, theater, city, hour)
 $S3(T,Th,C,H) \rightarrow Sh(T,Th,C,H)$
- Query: Find movies and directors playing in Seattle
 $Q(T,D) \leftarrow M(T,D,Y) \& Sh(T,Th,C,H) \& C = \text{"Seattle"}$
- Plan: Combine S1 with S2 or S3
 $Q(T,D) \leftarrow S1(T,D,Y) \& S2(T,Th,C,H) \& C = \text{"Seattle"}$
 $Q(T,D) \leftarrow S1(T,D,Y) \& S3(T,Th,C,H) \& C = \text{"Seattle"}$
- Optimized Plan: Use LCW to prune S3
 $Q(T,D) \leftarrow S1(T,D,Y) \& S2(T,Th,C,H) \& C = \text{"Seattle"}$

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Planning by Rewriting

[Ambite & Knoblock, 1998]

- Efficiently generate an initial solution plan (possibly of low quality)
- Iteratively rewrite the current plan
 - using a set of declarative plan rewriting rules
 - improving plan quality
 - until an acceptable solution or resource limit reached



Efficient High-Quality Planning

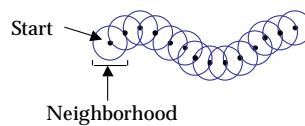
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Planning by Rewriting as Local Search

- PbR: efficient high-quality planning using local search
- Main issues:
 - Selection of initial feasible point: Initial plan generation
 - Generation of a local neighborhood: Set of plans obtained from application of the plan rewriting rules
 - Cost function to minimize: Measure of plan quality
 - Selection of next point: Next plan to consider -- determines how the global space is explored



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Planning by Rewriting for Query Planning in Mediators

- Initial plan generation: random parse of the query
- Plan rewriting rules: based on properties of:
 - relational algebra,
 - distributed environment,
 - integration axioms
- Plan quality: query execution time (size estimation)
- Search Strategies: gradient descent+restart, simulated annealing, variable-depth rewriting, ...

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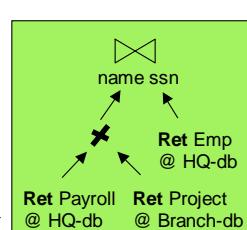
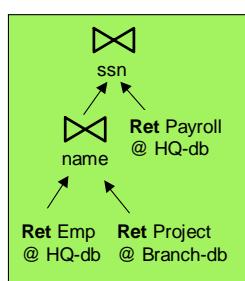
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Query Planning in PbR

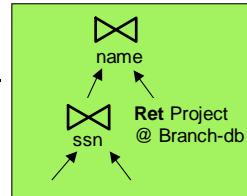
$a(name \text{ sal } proj) :- Emp(name ssn) \wedge Payroll(ssn sal) \wedge Projects(name proj)$

HQ-db
Emp(name ssn)
Payroll(ssn sal)

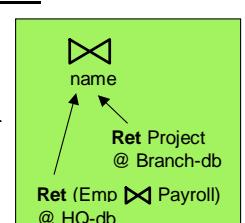
Branch-db
Project(name proj)



Join Swap



Remote Join Eval



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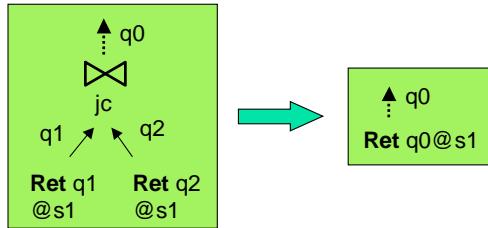
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Rewriting Rules: Distributed Environment remote-join-eval

```
(define-rule remote-join-eval
  :if (:operators ((?n1 (retrieve ?source ?query1))
                  (?n2 (retrieve ?source ?query2))
                  (?n3 (join ?join-conds ?query0 ?query1 ?query2)))
    :constraints (capability ?source join))
  :replace (:operators (?n1 ?n2 ?n3))
  :with (:operators ((?n4 (retrieve ?source ?query0)))))

  
```



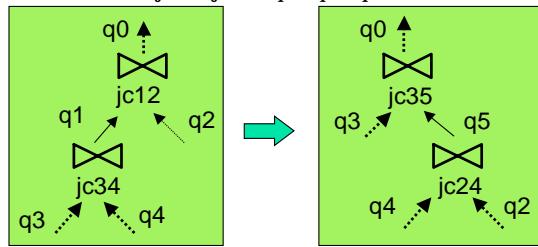
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Rewriting Rules: Relational Algebra join-associativity

```
(define-rule :name join-associativity
  :if (:operators ((?n1 (join ?jc34 ?q1 ?q3 ?q4)
                  (?n2 (join ?jc12 ?q0 ?q1 ?q2)))
    :constraints (join-swappable ?jc34 ?q1 ?q3 ?q4 ?jc12 ?q0 ?q2 ;;in
                           ?jc24 ?jc35 ?q5) ;; out
  :replace (:operators (?n1 ?n2))
  :with (:operators ((?n3 (join ?jc24 ?q5 ?q4 ?q2))
                  (?n4 (join ?jc35 ?q0 ?q3 ?q5))))
```



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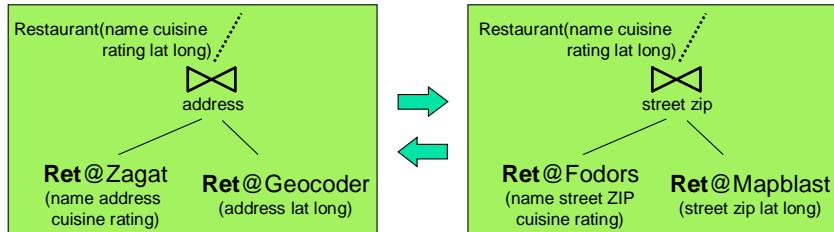
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Rewriting Rules: Integration Axioms

- Rules computed from integration axioms relevant to query:

$\text{Restaurant}(\text{name cuisine rating lat long}) =$

- $\text{Zagat}(\text{name address cuisine rating}) \wedge \text{Geocoder}(\text{address lat long})$
- $\text{Fodors}(\text{name street zip cuisine rating}) \wedge \text{Mapblast}(\text{street zip lat long})$



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PbR in Query Planning: Summary

- Operators: output, retrieve, assign, select, join, union
- Plan rewriting rules:
 - Relational algebra: join-swap, selection-push-in, selection-push-out, assignment-push-in, assignment-push-out, selection-assignment-swap, push-join-thru-union, and push-union-thru-join.
 - Distributed environment: source-swap, remote-join-eval, remote-selection-eval, and remote-assignment-eval.
 - Integration axioms: computed automatically from the relevant integration axioms for classes in the query
- Search: gradient descent + random restart
 - first-improvement
 - steepest descent

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Planning for the Internet Softbot

[Golden et al., 1996, Golden, 1998]

- XII and Puccini planners for the Internet Softbot
- Plans both gathering and manipulation actions
 - e.g., ls -a, chmod +r *
- Used to model Internet resources such as netfind
- Each resource modeled as an operator

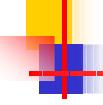
```
Name: (netfind ?person)
Preconds:
  (current.shell csh)
  (isa netfind.server ?server)
  (firstname ?person ?firstname)
  (lastname ?person ?lastname)
  (or
    (person.city ?person ?keyword)
    (person.institution ?person ?keyword))
Postconds:
  (userid ?person !userid)
  (person.machine ?person !machine)
```

Netfind Operator from XII

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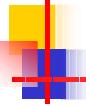
Observational Effects and Knowledge Preconditions

- **Observational Effects**
 - Effect that changes the state of the world
`chmod +r foo.tex -- cause(readable(foo.tex))`
 - Effect that changes the agent's model of the world
`wc -- observe(word.count(file, !word))`
- **Knowledge Preconditions**
 - Information goal -- `find-out(length(paper.tex, l))`
 - Goals of achievement -- `satisfy(readable(f) False)`
- **Verification Links**
 - Alternative to knowledge preconditions
 - Assume secondary condition is true and then use an observational effect to determine whether it is true after execution

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Sensing for Locally Complete Information

- **Reasons about incomplete information**
 - Uses LCW to reason about what it knows and what it doesn't know
 - e.g., `ls -a *` gives it locally complete information about the current directory
- **Interleaves sensing actions to gather LCW information**
 - LCW statements are a way of satisfying universally quantified goals
- **Provides fine-grained reasoning**
 - e.g., can request all recent techreports by X not already stored locally

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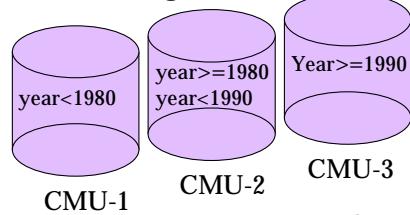
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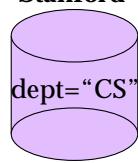
Sensing to Determine Relevant Sources [Ashish, Knoblock, & Levy, 1997]

Technical Report Repositories

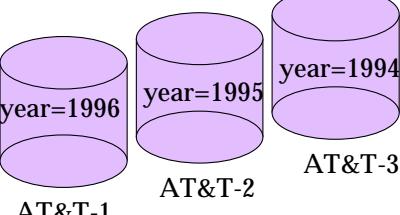
Carnegie Mellon



Stanford



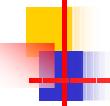
AT&T Labs



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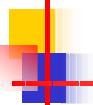
Building a Discrimination Matrix

- Discrimination matrix specifies the relevant sources for each region of each attribute
- Approach:
 - Analyze source descriptions to build a discrimination matrix
 - Matrix partitions sources along some attribute
 - Discrimination matrix used to estimate the cost of querying with and without sensing
- Useful discriminations provided when:
 - Sources can be partitioned by some attribute
 - Exists another source that provides that attribute
- Example: Information about the year of a tech report reduces the relevant sources from 7 to 3

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Discrimination Matrix

<u>Region</u>	<u>Relevant Sources</u>
< 1980	CMU-1, Stanford
[1980,1990)	CMU-2, Stanford
[1990,1994)	CMU-3, Stanford
[1994,1994]	CMU-3, Stanford, AT&T-1
[1995,1995]	CMU-3, Stanford, AT&T-2
[1996,1996]	CMU-3, Stanford, AT&T-3
> 1996	CMU-3, Stanford

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Planning with Discriminating Queries

- Consider inserting a discriminating query for any subquery that:
 - Requires accessing multiple sources
 - There exists a discriminating attribute in the matrix
- Compare the cost of no discrimination to the combined cost of discriminating and querying
- Since we cannot know the results of the discrimination, use the average estimated cost
- Potentially relevant sources: $S = S_1, \dots, S_6$
- Discriminating queries: R_1, R_2
- Possible plans: $S, R_1 S', R_2 S'', R_1 R_2 S'''$
 - $R_1: \{\{S_1, S_2\}, \{S_3, S_4, S_5\}, \{S_6\}\}$
 - $R_2: \{\{S_1\}, \{S_2, S_3\}, \{S_4, S_5, S_6\}\}$
 - $R_1 R_2: \{\{S_1\}, \{S_2\}, \{S_3\}, \{S_4, S_5\}, \{S_6\}\}$

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Plan without Sensing

Retrieve GNP
where Org=NATO
from Afghanistan Page
of the World Factbook

⋮
267 countries
⋮

⋮
Retrieve GNP
where Org=NATO
from Zimbabwe Page
of the World Factbook

Union GNP
of Countries
where Org=NATO

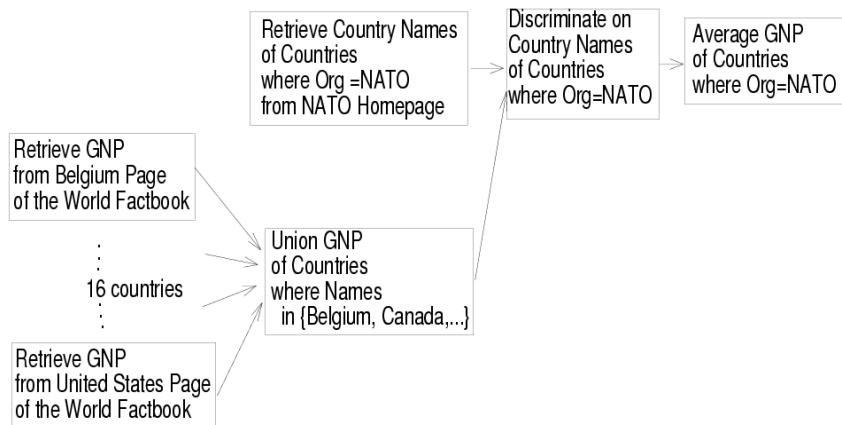
Average GNP
of Countries
where Org=NATO

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Plan with Sensing



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Contingent Planning for Information Gathering [Friedman & Weld '97]

- Use subsumption relationships to make a plan more resource conscious
 - Determined based on LCW statements
- Execution policies:
 - Brute force – ignore subsumption and execute everything greedily
 - Aggressive – execute multiple alternatives and cancel others once a subsumed source is successful
 - Frugal – execute the most general source first and only execute others if it fails

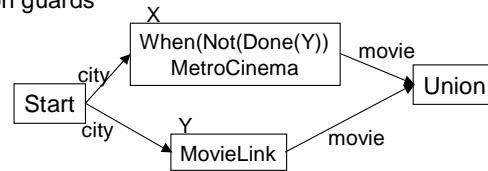
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Augmenting the Plans

- Contingent plans
 - Operator can fire when its guard is true
 - Status variable for each operator
 - Sleeping, running, failed, and done
 - Approach:
 - Nodes initialized to running
 - Running nodes fired when input is available
 - Update status based on guards
- Guards
 - Aggressive policy:
 - Failed(Y)
 - Frugal policy:
 - Failed(Y)



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Composing Web Services

- Information sources only have inputs and outputs
 - Possibly with some additional constraints on those
- Services have:
 - Inputs and outputs
 - Preconditions and effects
- Could be cast as a traditional planning problem with preconditions and effects
- Example:
 - To purchase a book on Amazon has a precondition of having the money and has the effects of having the book and less money
- Services can be composed into compound services [McIlraith & Fadel, 2002]

Craig Knoblock Stored and reused similar to Macros [Eikes, 1972]

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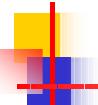
Discussion

- Is this planning?
 - Not in the sense of composing sequences of actions with interacting effects
 - Certainly in the broader sense of formulating a scheme or program for the accomplishment or attainment of some goal
- Good ideas can be shared across fields
 - Planning by rewriting based on traditional approaches to query planning
- Lots of interesting problems with real world applications
 - Optimizing the plans (e.g., interleaving sensing actions)
 - Interleaving source selection and plan optimization
 - Efficient execution of the plans (next class)

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Bibliography

- Planning for Information Gathering

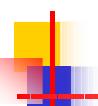
- View Integration

- Levy, Alon Y. (2000). Logic-based Techniques in Data Integration. *Logic Based Artificial Intelligence*, Edited by Jack Minker, Kluwer Publishers.
 - Halevy, Alon Y. (2001). Answering Queries Using Views: A Survey. *VLDB Journal*.
 - Duschka, Oliver M. (1997). Query Planning and Optimization in Information Integration. Ph.D. Thesis, Stanford University, Computer Science Technical Report STAN-CS-TR-97-1598.
 - Duschka, Oliver M. and Alon Y. Levy (1997). Recursive Plans for Information Gathering. *Proceedings of IJCAI-97*

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Bibliography

- Planning for Information Gathering

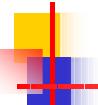
- Traditional Planning Approaches

- Lambrecht, Eric and Subbarao Kambhampati (1997). Planning for Information Gathering: A Tutorial Survey. ASU CSE Technical Report 96-017.
 - Knoblock, Craig A. (1995). Planning, Executing, Sensing, and Replanning for Information Gathering. In *Proceedings of IJCAI-95*.
 - Knoblock, Craig A. (1996) Building a planner for information gathering: A report from the trenches. In *Proceedings of AIPS-96*.
 - Kwok, Chung T. and Daniel S. Weld (1996). Planning to Gather Information. In *Proceedings of AAAI-96*.

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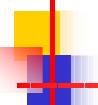
Bibliography

- Optimizing Information Gathering Plans
 - Removing Redundant Sources
 - Duschka, Oliver M. (1997). Query Optimization Using Local Completeness. In *Proceedings of AAAI-97*.
 - Lambrecht, Eric and Subbarao Kambhampati, and Senthil Gnanaprakasam. (1999) Optimizing Recursive Information Gathering Plans. In *Proceedings of IJCAI-99*.
 - Optimizing Sources and Queries
 - Ambite, Jose Luis and Craig A. Knoblock (2000). Flexible and Scalable Cost-based Query Planning in Mediators: A Transformational Approach. *Artificial Intelligence Journal*, 118(1-2):115—161.
 - Jose Luis Ambite and Craig A. Knoblock (1997). Planning by Rewriting: Efficient Generating High-Quality Plans. In *Proceedings of AAAI-1997*.

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Bibliography

- Interleaving Planning and Sensing
 - Sensing to Handle Incomplete Information
 - Golden, Keith, Oren Etzioni, and Daniel S. Weld (1996). Planning with Execution and Incomplete Information. University of Washington, Department of Computer Science, Technical Report UW-CSE-96-01-09.
 - Golden, Keith (1998) Leap Before you Look: Information Gathering in the PUCCINI Planner. In *Proceedings of AIPS-98*.
 - Sensing to Optimize Plans
 - Ashish, Naveen, Craig A. Knoblock, and Alon Y. Levy (1997). Information Gathering Plans with Sensing Actions, *Recent Advances in AI Planning: 4th European Conference on Planning, ECP'97*. Springer-Verlag, New York, 1997.

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Bibliography

- Contingent Planning for Information Gathering
 - Friedman, Marc and Daniel S. Weld. (1997). Efficiently Executing Information-Gathering Plans. In *Proceedings of IJCAI-97*, Nagoya, Japan, August 1997.
- Planning to Compose Web Sources
 - McIlraith, Sheila and Ronald Fadel (2002). Planning with Complex Actions. In *Proceedings of AIPS-2002 Workshop: Is There Life Beyond Operator Sequencing? - Exploring Real-World Planning*.