# Building Mashups by Example

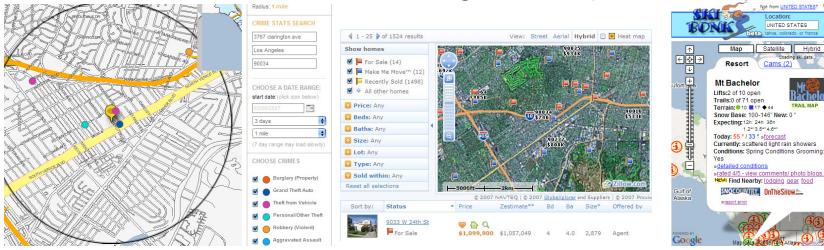
Rattapoom Tuchinda

**Doctoral Defense** 

July 22, 2008

# What's a Mashup?

A website or application that combines content from more than one source into an integrated experience [wikipedia]



a) LA crime map -Crime Report from different counties -Map b) zillow.com-Real Estate Listing-Property Tax

c) Ski bonk-Weather-Snow Report-Snow Resorts

### Combined Data gives new insight / provides new services

NEW! Add SkiBonk to about ite mobiles

Only Operating Resorts

Name

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Popular Se

Ski Resorts' showr

oo many resorts to display. Zoom in to see m

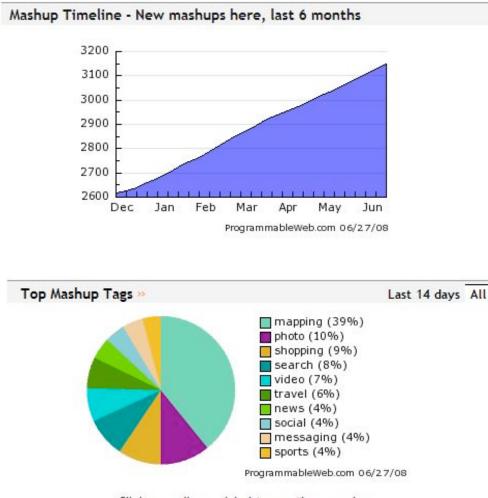
Runs Lifts Conditions

Heavenly (4/5)

auze d'Oulx (3.5/5

and Jun 03-04

# **Statistics and Trends**



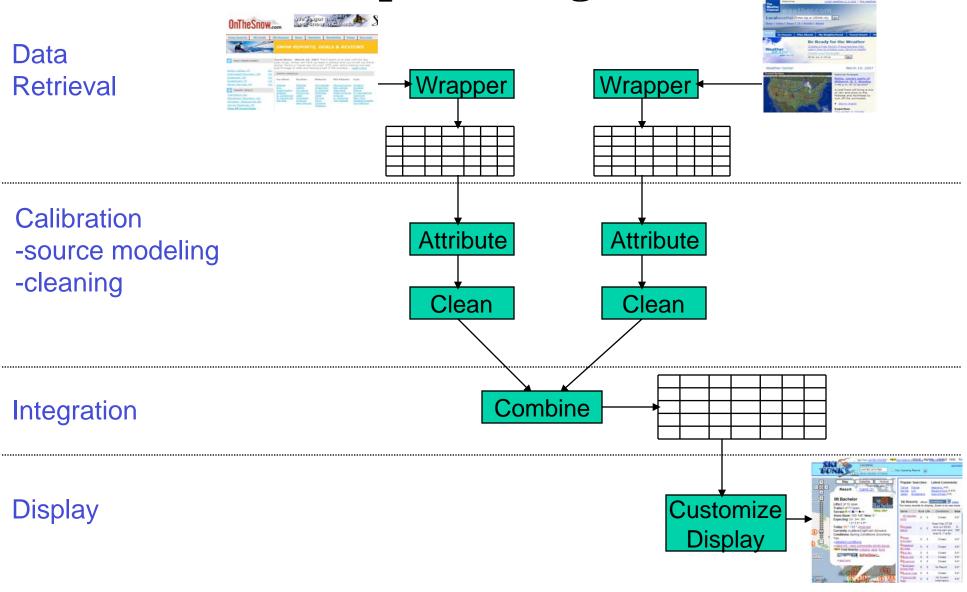
Click on a slice or label to see those mashups

### **Survey of top 50 Mashups**

• Divide into five categories based on programming structures

• Focus of this thesis is on the first four categories which account for 47% of the most popular Mashups

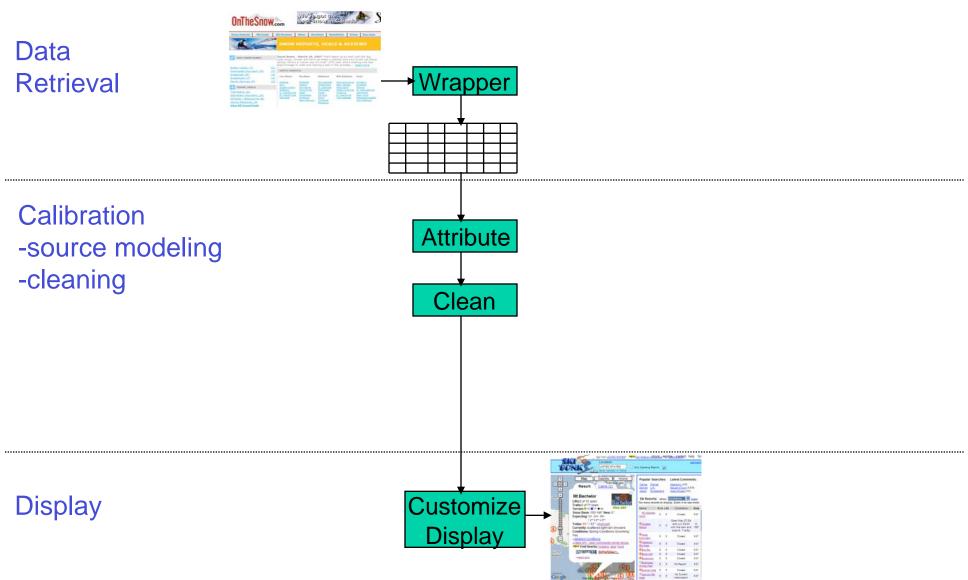
# **Mashup Building Issues**



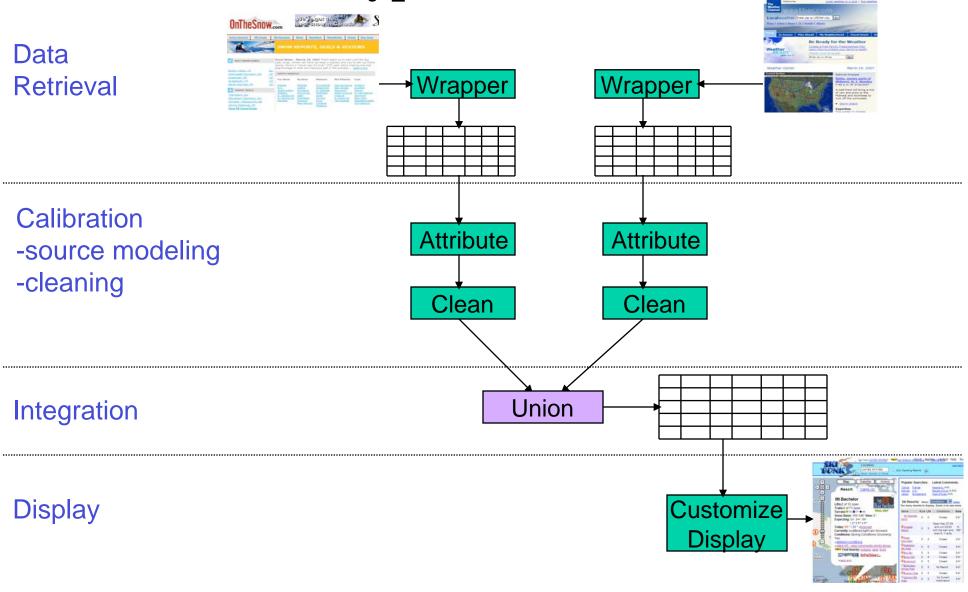
Introduction • Approach • Evaluation • Related Work • Conclusion

Δ

# **Type 1: One Simple Source**



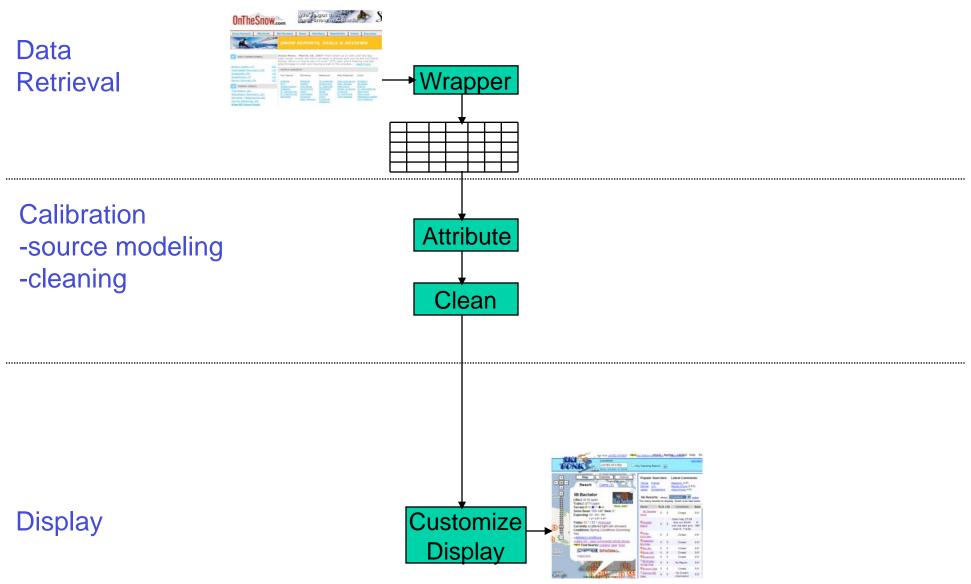
# **Type 2: Union**



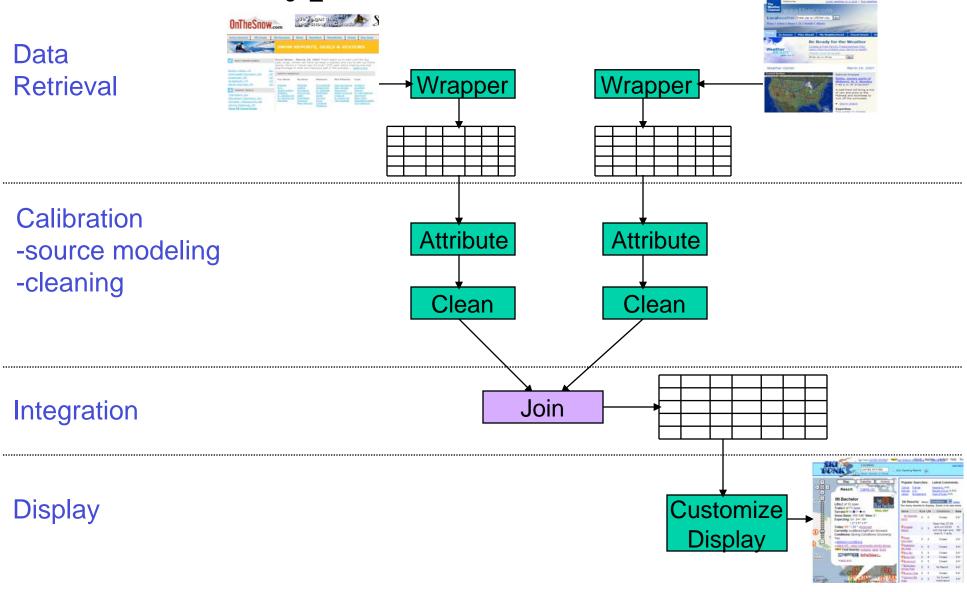
Introduction • Approach • Evaluation • Related Work • Conclusion

6

# **Type 3: One Source with Form**

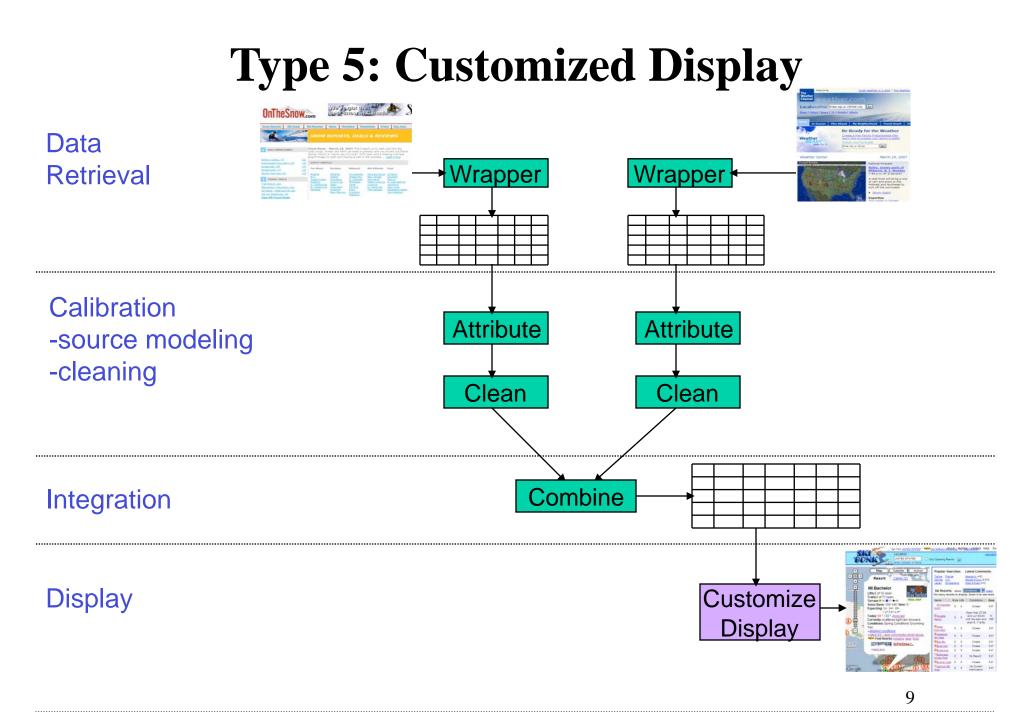


# **Type 4: Database Join**



Introduction • Approach • Evaluation • Related Work • Conclusion

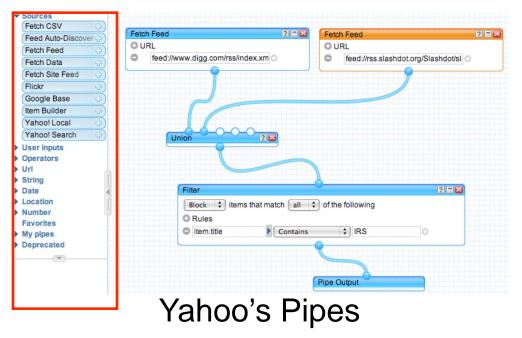
8



# **Existing Approaches**

Goal: Create Mashups without Programming

• Doesn't translate to not having to understand programming.



Widget Paradigm

- Widgets (i.e., 43 for Pipes, 300+ for MS) represents an operation on the data.
- Locating and learning to customize widget can be time consuming
- Most tools focus on particular issues and ignore others.

Can we come up with a framework that addresses all of the issues while still making the Mashup building process easy?

## **Thesis Statement**

Web users can build Mashups effectively using an integrated framework that lets them solve the problems of data extraction, source modeling, data cleaning, and data integration by specifying examples instead of programming operations.

# Contributions

- A programming by demonstration approach that uses a single table for building a Mashup
- An integrated approach that links data extraction, source modeling, data cleaning, and data integration together.
- A query formulation technique that allows users to specify examples to build complicated queries.

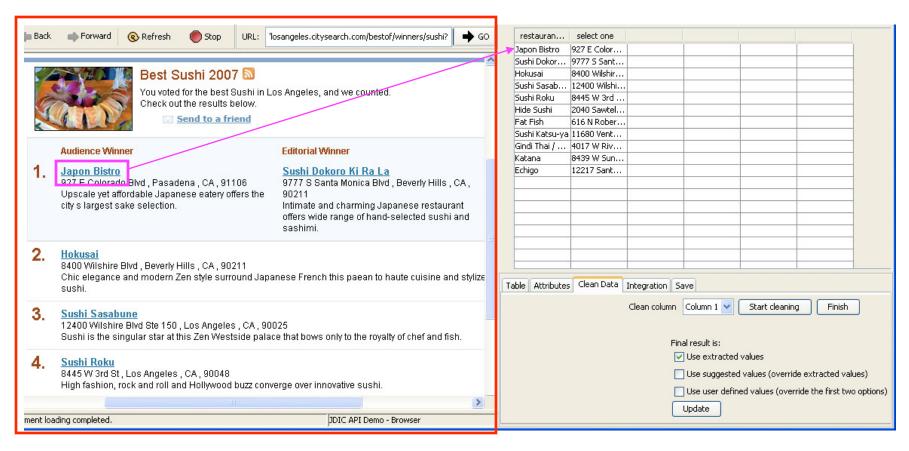
# **Key Ideas**

- Focus on data, not operations
  - Users are more familiar with data.
- Leverage existing database
  - Help source modeling, cleaning, and data integration.
- Consolidate as opposed to Divide-And-Conquer
  - Solving a problem in one issue can help solve another issue.
  - Interacting within a single spreadsheet platform

### **Embedded Browser**

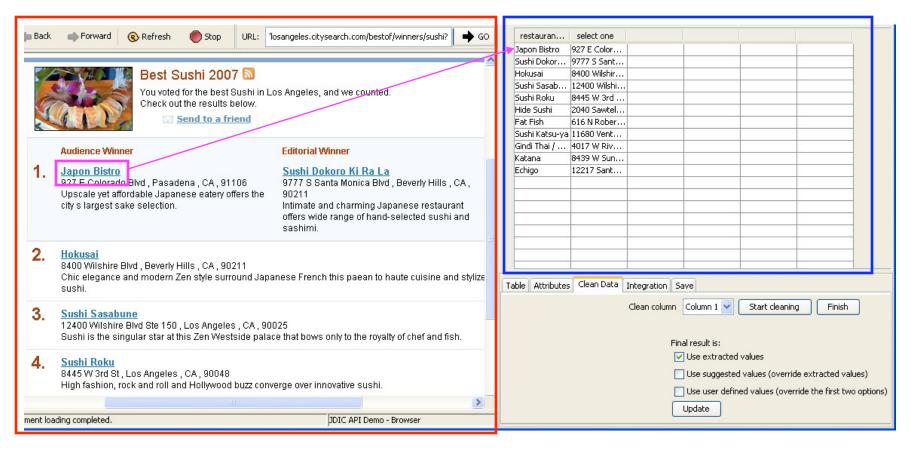
	8445 W 3rd St , Los Angeles , CA , 90048 High fashion, rock and roll and Hollywood buzz conv	erge over innovative sushi.			•	e extracted values) ide the first two opt
3. 4.	Sushi Sasabune 12400 Wilshire Blvd Ste 150 , Los Angeles , CA , 900 Sushi is the singular star at this Zen Westside palac Sushi Roku				al result is: ] Use extracted	
2.	<mark>Hokusai</mark> 8400 Wilshire Blvd , Beverly Hills , CA , 90211 Chic elegance and modern Zen style surround Japa sushi.	nese French this paean to haute cuisine and stylize	Table Attribute	s Clean Data		
1.	Japon Bistro 927 E Colorado Blvd , Pasadena , CA , 91106 Upscale yet affordable Japanese eatery offers the city s largest sake selection.	Sushi Dokoro Ki Ra La 9777 S Santa Monica Blvd , Beverly Hills , CA , 90211 Intimate and charming Japanese restaurant offers wide range of hand-selected sushi and sashimi.	Echigo	12217 Sant		
	Audience Winner	Editorial Winner	Katana	4017 W Riv 8439 W Sun		
G.	Check out the results below.			8445 W 3rd 2040 Sawtel 616 N Rober 11680 Vent		
Ź	Best Sushi 2007 D		Hokusai	8400 Wilshir 12400 Wilshi		
			Japon Bistro Susbi Dokor	927 E Color 9777 S Sant		

### **Embedded Browser**



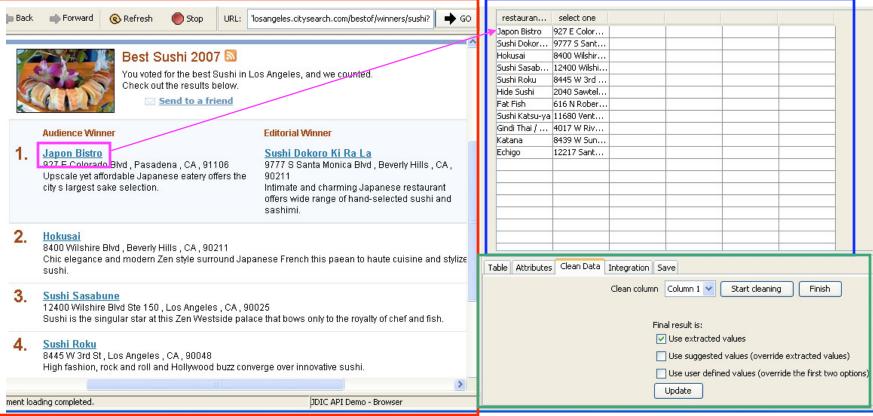
### **Embedded Browser**

### Table

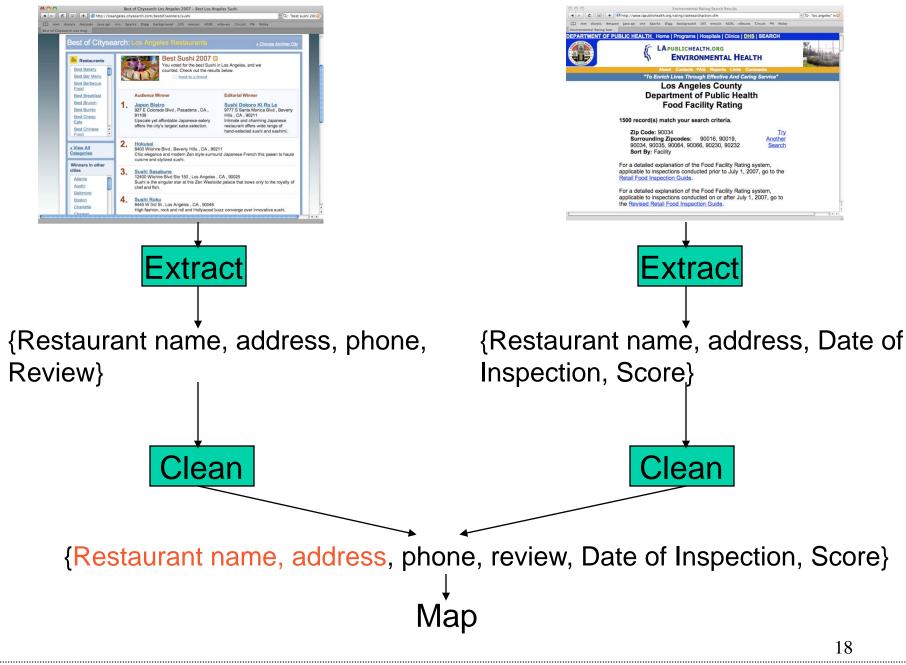


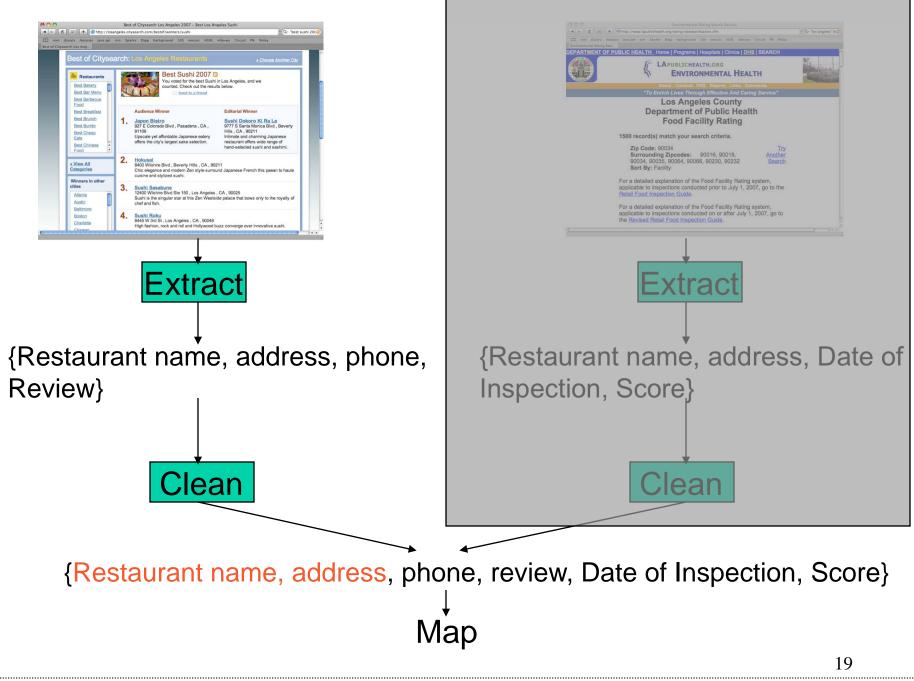
### **Embedded Browser**

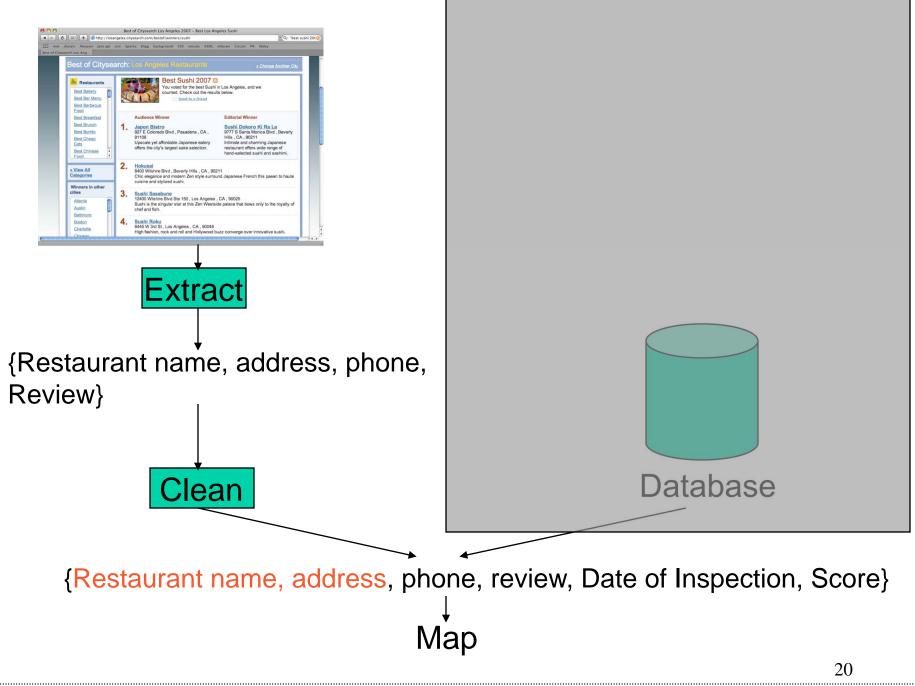
### Table

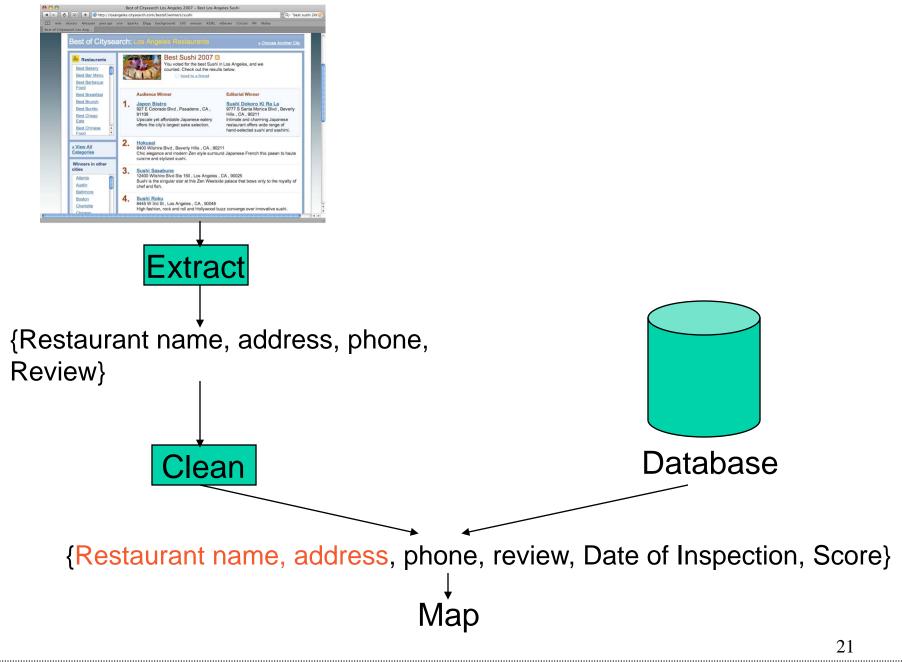


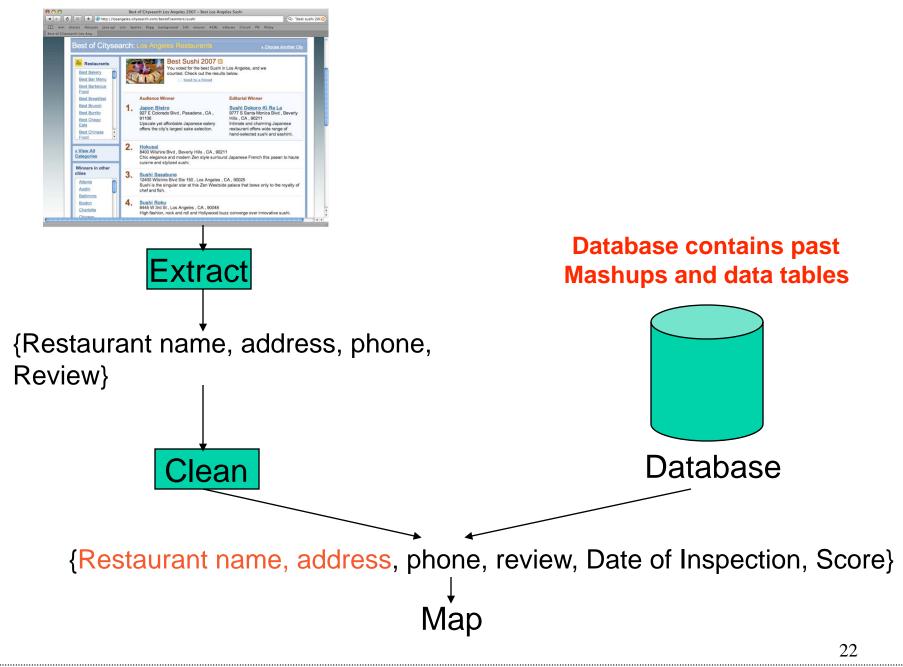
### **Interaction Modes**











## **Data Retrieval: Extraction**

### Japon Bistro

927 E Colorado B vd , Pasadena , CA , 91106 Upscale yet affordable Japanese eatery offers the city's largest sake selection.

### Sushi Dokoro Ki Ra La

9777 S Santa Monica Blvd , Beverly Hills , CA 90211 Intimate and charming Japanese restaurant offers wide range of hand-selected sushi and sashimi.

#### 2. Hokusai

8400 Wilshire Blvd , Beverly Hills , CA , 90211

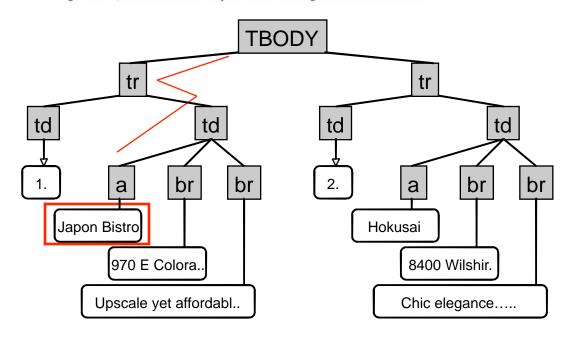
Chic elegance and modern Zen style surround Japanese French this paean to haute cuisine and stylized sushi.

### 3.

Sushi Sasabune 12400 Wilshire Blvd Ste 150 , Los Angeles , CA , 90025 Sushi is the singular star at this Zen Westside palace that bows only to the royalty of chef and fish.

#### Sushi Roku 4.

8445 W 3rd St, Los Angeles, CA, 90048 High fashion, rock and roll and Hollywood buzz converge over innovative sushi.



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Japon Bistro		
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## **Data Retrieval: Extraction**

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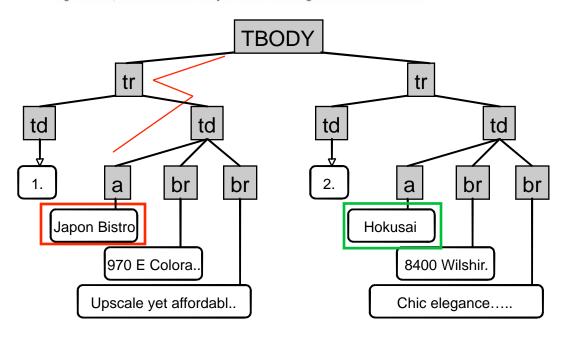
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select one	
Japon Bistro	
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Sushi Roku	
Hide Sushi	
Fat Fish	
Sushi Katsu-ya	
Gindi Thai /	
Katana	
Echigo	

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## **Data Retrieval: Navigation**

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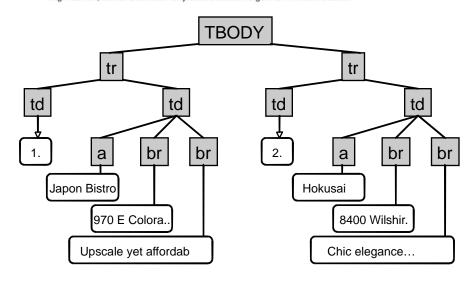
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### **Japon Bistro**

927 E Colorado Blvd Pasadena, CA 91106 Phone: (626) 744-1751





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Lin	to this page
Har	re Updated Info?
Sug	gest a correction

select one	address	select one	select one
Japon Bistro	927 E Color	Upscale yet	31 Reviews
Sushi Dokor	9777 S Sant	Intimate an	
Hokusai	8400 Wilshir	Chic eleganc	
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Sushi Roku	8445 W 3rd	High fashion	
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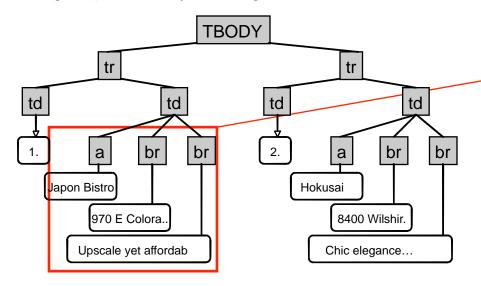
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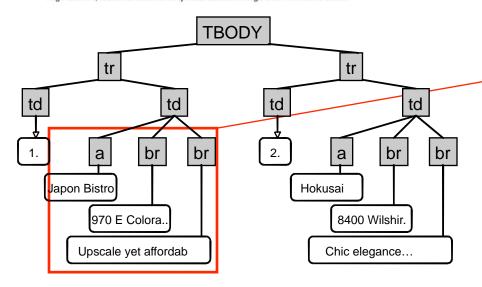
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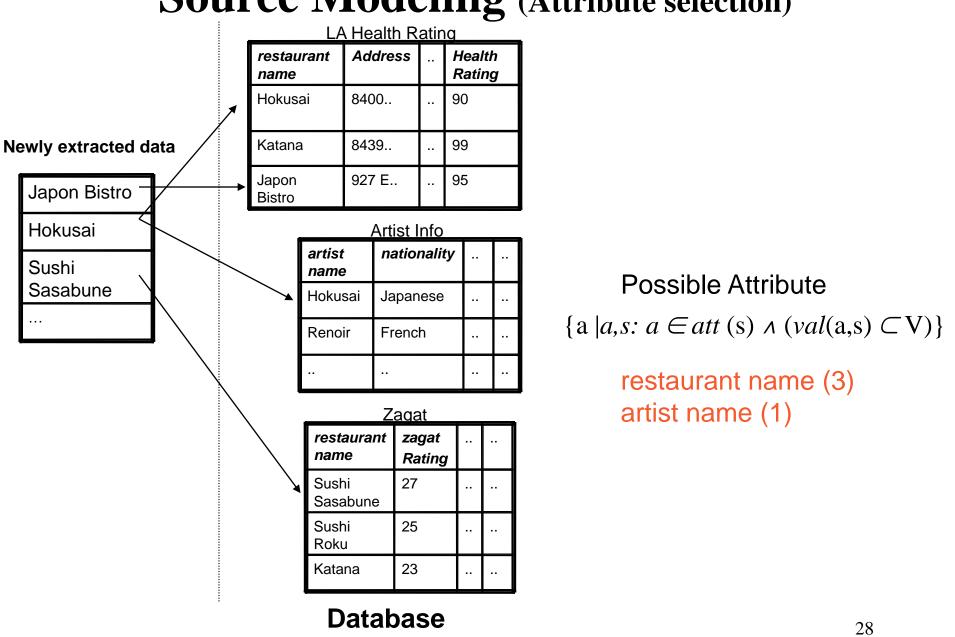




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select one	address	select one	select one
Japon Bistro	927 E Color	Upscale yet	31 Reviews
Sushi Dokor	9777 S Sant	Intimate an	3 Reviews
Hokusai	8400 Wilshir	Chic eleganc	30 Reviews
Sushi Sasab	12400 Wilshi	Authentic Ja	66 Reviews
Sushi Roku	8445 W 3rd	High fashion	62 Reviews
Hide Sushi	2040 Sawtel	No fuss, jus	25 Reviews
Fat Fish	616 N Rober	Inventive ro	38 Reviews
Sushi Katsu-ya	11680 Vent	The MOCA o	49 Reviews
Gindi Thai /	4017 W Riv	Burbank res	29 Reviews
Katana	8439 W Sun	Rustic Japa	96 Reviews
Echigo	11217 Sant	Stellar sushi	49 Reviews



Source Modeling (Attribute selection)

## **Data Cleaning:** using existing values

Data	repository	
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Newly ext	racted	data
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Japon Bistro Hokusai Sushi Sasabune Sushi Roka

Restaurant name

restaurant name	Address	 Health Rating
Hokusai	8400	 90
Katana	8439	 99
Japon Bistro	927 E	 95

I A Hoalth Pating

Zagat

restaurant name	zagat Rating	 
Sushi Sasabune	27	 
Sushi Roku	25	 
Katana	23	 

## **Data Cleaning:** using existing values

**Data repository** 

### LA Health Rating restaurant Address Health Rating name Newly extracted data 8400.. Hokusai 90 .. Japon Bistro Katana 8439... 99 .. Hokusai 927 E., Japon 95 ... **Bistro** Sushi Sasabune Zagat Sushi restaurant zagat Roka name Rating Sushi 27 • • . . Sasabune Restaurant name 25 Sushi . . • • Roku 23 Katana . . • •

## **Data Cleaning:** using predefined rules

	description	number of r	suggest	user defined	final
	Upscale yet	31 Reviews		31	1
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ai.	Chic eleganc	30 Reviews			
	Authentic Ja	66 Reviews			
ă,	High fashion	62 Reviews	M		
	No fuss, jus	25 Reviews			
	Inventive ro	38 Reviews			
	The MOCA o	49 Reviews			
	Burbank res	29 Reviews			
ai.	Rustic Japa	96 Reviews			
	Stellar sushi	49 Reviews			

31 Reviews  $\rightarrow$  31 Subset Rule:  $(s_1s_2..s_k) \rightarrow (d_1d_2...d_t) \land$   $(k \leq t) \land$   $s_i \in \{d_1, d_2, ..., d_t\} \land$  $d_i \neq d_j$ 

Predefined Rules

## **Data Cleaning:** using predefined rules

	description	number of r	suggest	user defined	final
17	Upscale yet	31 Reviews		31	
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10	, Authentic Ja	66 Reviews			
<i>(</i> )	High Fashion	62 Reviews			
	. No fuss, jus	25 Reviews			
	, Inventive ro	38 Reviews			
	The MOCA o	49 Reviews			
	Burbank res	29 Reviews			
6	. Rustic Japa	96 Reviews			
11	Stellar sushi	49 Reviews			
Subset	ews $\rightarrow 3^{\circ}$ Rule: () $\rightarrow (d_1 d_2)$	-			
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``	$d_2, \dots, d_t \} \land$	<b>`</b>		Pre	edefined

Introduction • Approach • Evaluation • Related Work • Conclusion

# Data Integration [tuchinda 2007]

restauran	address	description	number o	
Japon Bistro	927 E Color	Upscale yet	31	
Sushi Dokor	9777 S Sant	Intimate an	3	
Hokusai	8400 Wilshir	Chic eleganc	30	
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Sushi Roku	8445 W 3rd	High fashion	62	
Hide Sushi	2040 Sawtel	No fuss, jus	25	
Fat Fish	616 N Rober	Inventive ro	38	
Sushi Katsu-ya	11680 Vent	The MOCA o	49	
Gindi Thai /	4017 W Riv	Burbank res	29	
Katana	8439 W Sun	Rustic Japa	96	
Echigo	12217 Sant	Stellar sushi	49	

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# Data Integration [tuchinda 2007]

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# Data Integration [tuchinda 2007]

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# **Data Integration (cont.)**

restaurant	address	description	number of r
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Sushi Dokor	9777 S Sant	Intimate an	3
Hokusai	8400 Wilshir	Chic eleganc	30
Sushi Sasab	12400 Wilshi	Authentic Ja	66
Sushi Roku	8445 W 3rd	High fashion	62
Hide Sushi	2040 Sawtel	No fuss, jus	25
Fat Fish	616 N Rober	Inventive ro	38
Sushi Katsu-ya	11680 Vent	The MOCA o	49
Gindi Thai /	4017 W Riv	Burbank res	29
Katana	8439 W Sun	Rustic Japa	96
Echigo	11217 Sant	Stellar sushi	49

Data repository

### LA Health Rating

restaurant name	Address	 Health Rating
Hokusai	8400	 90
Katana	8439	 99
Japon Bistro	927 E	 95

Z	Zagat	
restaurant name	zagat Rating	 
Sushi Sasabune	27	 
Sushi Roku	25	 
Katana	23	 

## **Data Integration (cont.)**

restaurant	address	description	number of r	-							
Japon Bistro	927 E Color	Upscale yet	31		Data repository						
Sushi Dokor	9777 S Sant	Intimate an	3				Data rept	<b>3</b> 10	Ory		
Hokusai	8400 Wilshir	Chic eleganc	30			1.0		- 41.00			
Sushi Sasab	12400 Wilshi	Authentic Ja	66			LA	Health Ra	atin	q		
Sushi Roku	8445 W 3rd	High fashion	62			restaurant	Address		Health		
Hide Sushi	2040 Sawtel	No fuss, jus	25			name			Rating		
Fat Fish	616 N Rober	Inventive ro	38				8400		00		
Sushi Katsu-ya	11680 Vent	The MOCA o	49			Hokusai			90		
Gindi Thai /	4017 W Riv	Burbank res	29								
Katana	8439 W Sun	Rustic Japa	96			Katana	8439		99		
Echigo	11217 Sant	Stellar sushi	49			Rataria	0400		00		
						Japon Bistro	927 E		95		

Zagat							
restaurant name	zagat Rating						
Sushi Sasabune	27						
Sushi Roku	25						
Katana	23						

## **Data Integration (cont.)**

restaurant	address	description	number of r
Japon Bistro	927 E Color	Upscale yet	31
Sushi Dokor	9777 S Sant	Intimate an	3
Hokusai	3400 Wilshir	Chic eleganc	30
Sushi Sasab	12400 Wilshi	Authentic Ja	66
Sushi Roku	3445 W 3rd	High fashion	62
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Sushi Katsu-ya	11680 Vent	The MOCA o	49
Gindi Thai /	1917 W Riv	Burbank res	29
Katana	3439 W. Sun	Rustic Japa	96
Echigo	11217 Sant.	Stellar sushi	49

Data repository

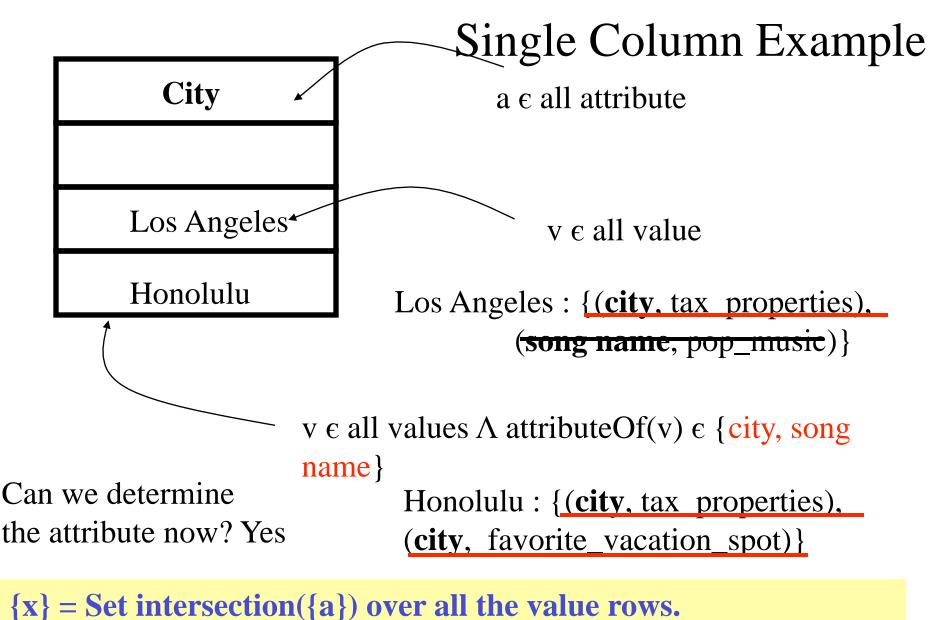
#### LA Health Rating

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2	Zagat							
restaurant name	zagat Rating							
Sushi Sasabune	27							
Sushi Roku	25							
Katana	23							

## **Data Integration (cont.)**

				<u> </u>	_					
restaurant	address	description	number of r							
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Sushi Roku	3445 W 3rd	High fashion	62			restaurant	Addre	SS		Health
Hide Sushi	2040 Sawtel	No fuss, jus	25			name				Rating
Fat Fish	516 N Rober	Inventive ro	38		Ī	I I a luva a i	0.400			
Sushi Katsu-ya	11680 Vent	The MOCA o	49		1	Hokusai	8400			90
Gindi Thai /	1017 W Riv	Burbank res	29		1					
Katana	3439 W Sun	Rustic Japa	96		1	Katana	8439			99
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$\{v\} = \operatorname{val}(a)$	a,s) wher	e <i>a {x}</i>		ue 10 w 5.		Sushi Roku	25			
s is any so	ource wh	ere <i>att(s)</i>	${x} \neq {}$			Katana	23			



 $\{v\} = \operatorname{val}(a,s)$  where  $a \in \{x\} \land s$  is any source where  $att(s) \cap \{x\} \neq \{\}$ 

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Introduction • Approach • Evaluation • Related Work • Conclusion

#### **Map Generation**

🖨 Back	rorward 📦	🛞 Refresh 🛛 🥚	Stop	URL:	'losangeles.citysearch.com/best	:of/winners/sushi?init_search=1	•	50
restaurant descriptio			💽 addi	ress: [			•	~
	3.1	Panorama City Van Nuys Sepulveda Dam Recreation Area Incino	re au B du Ja ha Sa MTNS Recreat	ddres everly escrip odern apanes aute cu umber ealth_	Map Sate ant_name:Hokusai s:8400 Wilshire Blvd, Hills, CA, 90211 tion:Chic elegance and Zen style surround the French this paean to this and stylized sushi. tr_of_reviews:30 pating:90 Ner Hollywood Ner Hollywood Ner Hollywood	ellite Hybrid Terrace La Crescen La C Controse Flin Montrose Glendale James La C S Controse Flin Montrose Glendale C C C C C C C C C C C C C C C C C C C		
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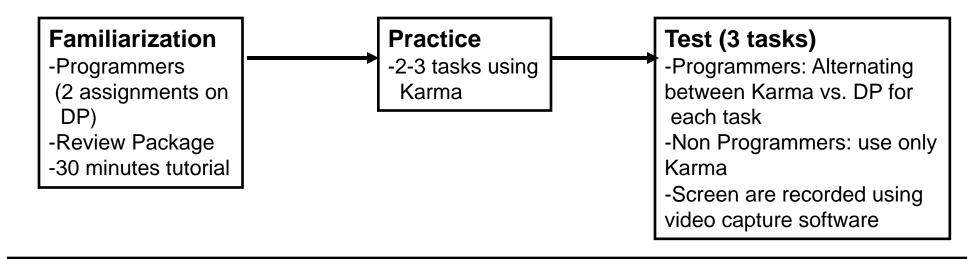
Introduction • Approach • Evaluation • Related Work • Conclusion

# **Evaluation**

- Baseline: A combination of Dapper/Pipes
- Claims:
  - Users with no programming experiences can build all four Mashup types.
    - 2. Karma takes less time to complete each subtask.
    - 3. Overall, the user takes less time to build the same Mashup in Karma compared to Dapper/Pipes
- Users:
  - Programmers (20)
  - Non-programmers (3)

If subjects (programmers) who are familiar with workflow and widgets spend more time on Dapper/Pipes in general, then the non-programmer subjects would spend more time on Dapper/Pipes as well if they were to learn how to use those systems.

## **Evaluation: Setup**



Task2		Dapper/Pipes						Karma				
Subject	Е	Μ	С	Ι	Total	Е	Μ	С	Ι	Total		
No.1	4:38	0:22	2:45	1:15	9:00	1:26	0:43	0:43	0:00	2:52		
No.2	1:35	0:12	3:30	0:12	5:29	0:50	0:57	0:57	0:00	2:44		
No.3	*5:00	0:25	*5:00	*5:00	15:25	2:52	1:00	<b>3:</b> 00	0:00	5:52		
No.4	4:49	0:17	3:29	0:38	9:14	1:26	0:48	1:03	0:00	3:18		
No.5	*5:00	0:29	1:44	1:16	8:29	1:43	0:45	1:20	0:00	3:48		
No.6	*5:00	0:20	*5:00	*5:00	15:20	2:07	0:30	0:50	0:00	3:27		

Using 5 minutes cut off time

Introduction • Approach • Evaluation • Related Work • Conclusion

#### **Evaluation:** Tasks

Task No.	Mashup Type	Data Extraction	Source Modeling	Data Cleaning	Data Integration
1	1 (1 source)	Moderate	Simple	Difficult	N/A
2	2,3 (union+form)	Difficult	Simple	Simple	Union (simple)
3	4 (join 2 sources)	Simple	Simple	N/A	Join (difficult)

#### Claim 1

Users with no programming experiences can build all four Mashup types.

#### **Evaluation:** Tasks

Task No.	Mashup Type	Data Extraction	Source Modeling	Data Cleaning	Data Integration
1	1 (1 source)	Moderate	Simple	Difficult	N/A
2	2,3 (union+form)	Difficult	Simple	Simple	Union (simple)
3	4 (join 2 sources)	Simple	Simple	N/A	Join (difficult)

#### Claim 2

When the Mashup subtask is difficult, Karma takes less time to complete that subtask.

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#### **Evaluation:** Tasks

Task No.	Mashup Type	Data Extraction	Source Modeling	Data Cleaning	Data Integration
1	1 (1 source)	Moderate	Simple	Difficult	N/A
2	2,3 (union+form)	Difficult	Simple	Simple	Union (simple)
3	4 (join 2 sources)	Simple	Simple	N/A	Join (difficult)

#### Claim 3

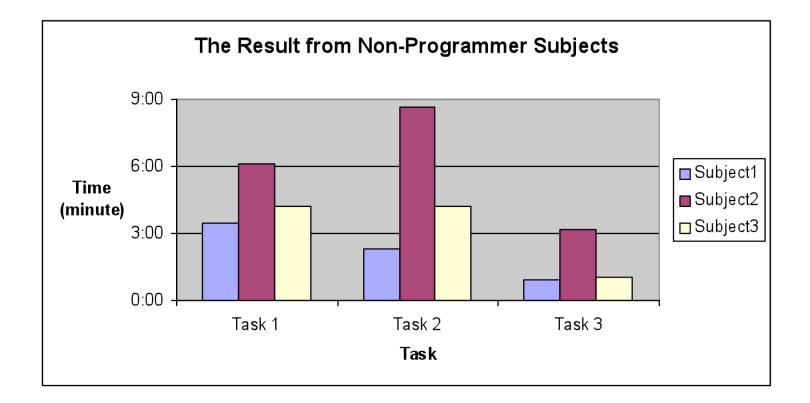
Overall, the user takes less time to build the same Mashup in Karma compared to Dapper/Pipes

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Introduction • Approach • Evaluation • Related Work • Conclusion

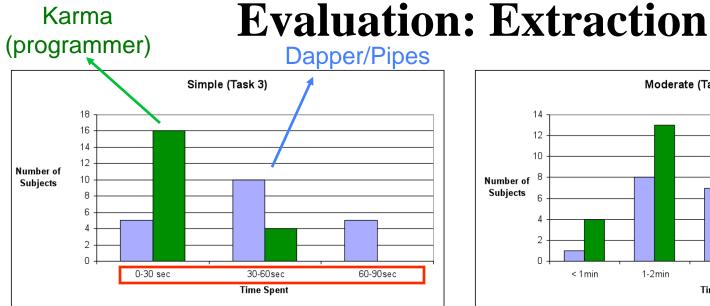
Claim 1: Users with no programming experiences can build all four Mashup types

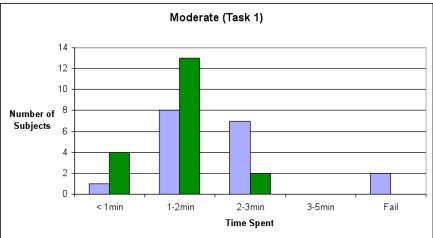
#### **Evaluation: Non-Programmers**

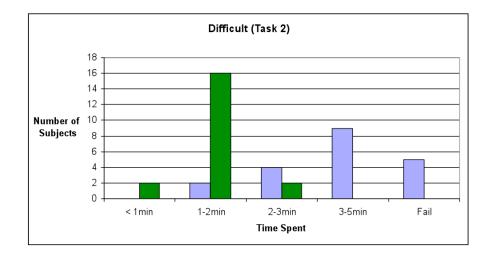


Introduction • Approach • Evaluation • Related Work • Conclusion

# Claim 2: Karma takes less time to complete each subtask





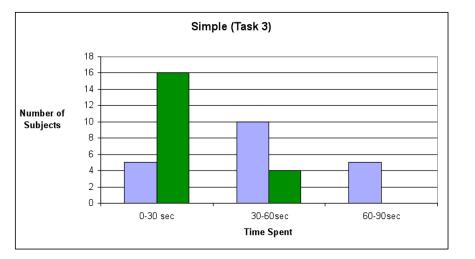


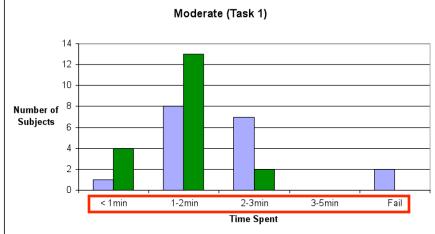
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Dapper/Pipes 📕 Karma Introduction • Approach • Evaluation • Related Work • Conclusion

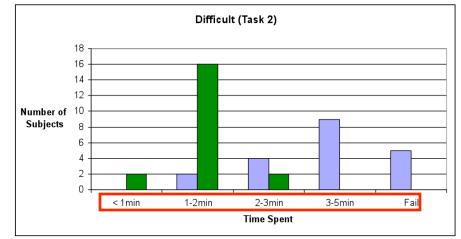
#### **Evaluation: Extraction**





- As the extraction task gets more difficult, Dapper/Pipes takes
  - longer

- more subjects failing to complete the task (11% for moderate and 25% for difficult)

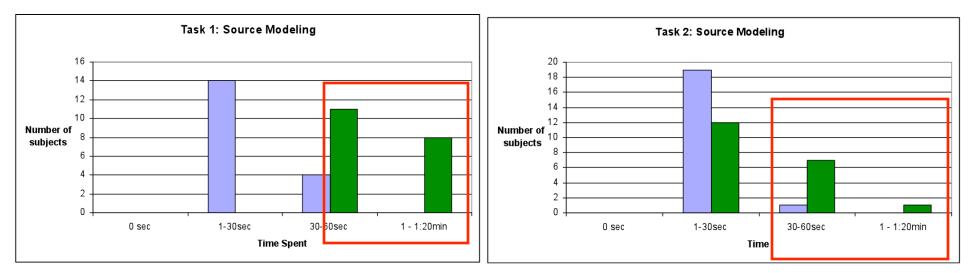


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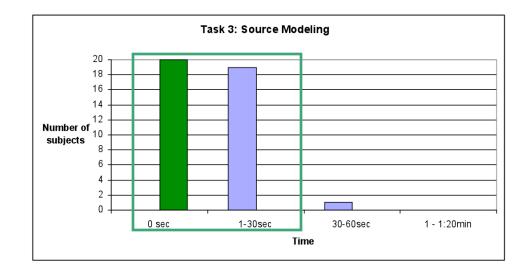
 Dapper/Pipes
 Karma

 Introduction • Approach • Evaluation • Related Work • Conclusion

# **Evaluation: Source Modeling**



- Karma performed worse in task 1 and tasks 2
  - only 30 sec difference
  - -subjects take times selecting attributes
  - the saving will be realized in the data integration step.
- Karma performed better in task 3 because of union

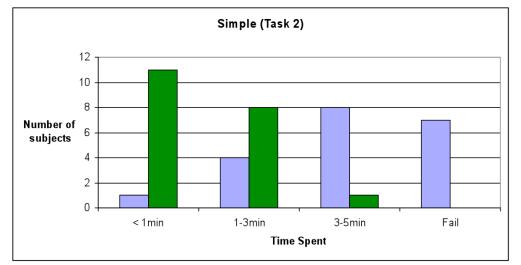


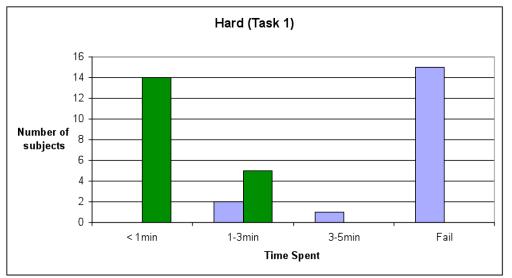
52

 Dapper/Pipes
 Karma

 Introduction • Approach • Evaluation • Related Work • Conclusion

#### **Evaluation: Data Cleaning**





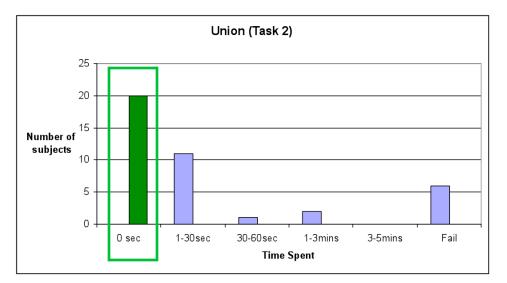
- Karma performed better in both tasks
- When the cleaning task gets harder, more subjects are failing in Dapper/ Pipes (35% for simple and 83% in hard)

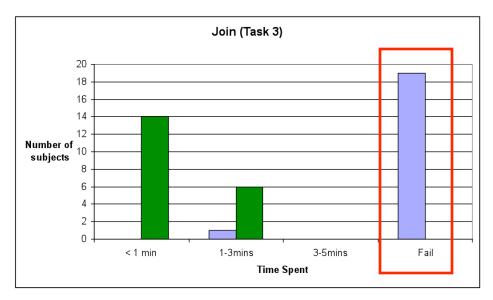
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Introduction • Approach • Evaluation • Related Work • Conclusion

Dapper/Pipes Karma

#### **Evaluation: Data Integration**





- Because of the table structure, subjects can specify union indirectly by dropping data into the right cell
- The time spent in source modeling step allows Karma to suggest the linking source
- Dapper/Pipes: 30% fail in the union case and 95% fail in the join case

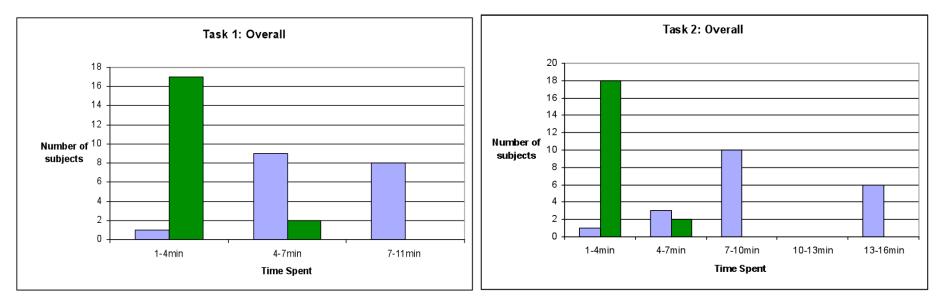
54

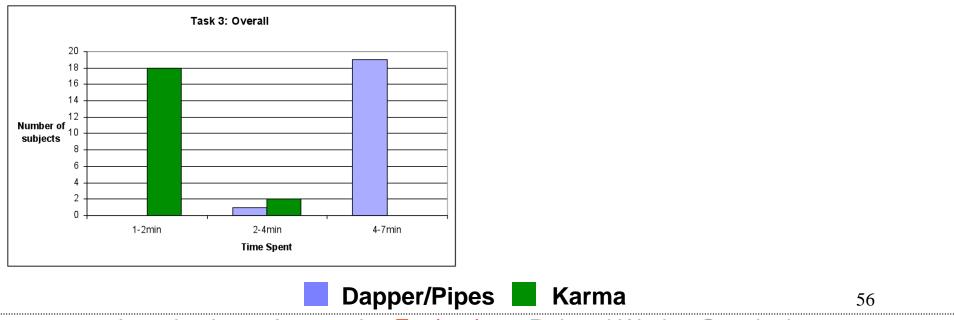
Dapper/Pipes 📕 Karma

Introduction • Approach • Evaluation • Related Work • Conclusion

Claim 3: Overall, the user takes less time to build the same Mashup in Karma compared to Dapper/Pipes

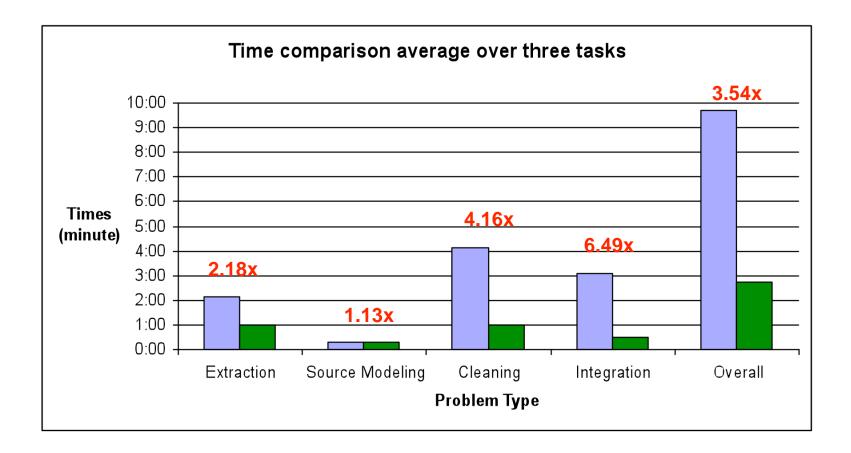
#### **Evaluation: Overall**





Introduction • Approach • Evaluation • Related Work • Conclusion

#### **Evaluation:** Average



Dapper/PipesKarmaIntroduction • Approach • Evaluation • Related Work • Conclusion

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System	Data	Source	Data	Data	Mashup Type
	Retrieval	Modeling	Cleaning	Integration	Supported
MIT's Simile	DOM	Manual	N/A	N/A	1
MIT's Pot Luck	RDF	Manual	PBD	Manual	1,3,4
Dapper	DOM	Manual	Manual	Join only	1,2,4
Yahoo's Pipes	Widgets	Manual	Widgets	Union only	1,2,3
MS's Popfly	Widgets	Manual	Widgets	Widgets	1,2,4
CMU's Marmite	Widgets	Manual	Widgets	Widgets	1,2,4
Intel's Mashmaker	Dapper	Manual	Widgets	Expert	1,2,3,4
Google MyMap	Widgets	Manual	N/A	Union only	1,2
Agent Wizard	Q/A	Q/A	Q/A	Q/A	1,3,4
Cards	DOM	Manual	N/A	Manual	1,2,4
Karma	DOM	Database	PBD	PBD	1,2,3,4

System	Data	Source	Data	Data	Mashup Type
	Retrieval	Modeling	Cleaning	Integration	Supported
MIT's Simile	DOM	Manual	N/A	N/A	1
MIT's Pot Luck	RDF	Manual	PBD	Manual	1,3,4
Dapper	DOM	Manual	Manual	Join only	1,2,4
Yahoo's Pipes	Widgets	Manual	Widgets	Union only	1,2,3
MS's Popfly	Widgets	Manual	Widgets	Widgets	1,2,4
CMU's Marmite	Widgets	Manual	Widgets	Widgets	1,2,4
Intel's Mashmaker		Require	an expert		1,2,3,4
Google MyMap	Widgets	Manual	N/A	Union only	1,2
Agent Wizard	Q/A	Q/A	Q/A	Q/A	1,3,4
Cards	DOM	Manual	N/A	Manual	1,2,4
Karma	DOM	Database	PBD	PBD	1,2,3,4

System	Data	Source	Data	Data	Mashup Type
	Retrieval	Modeling	Cleaning	Integration	Supported
MIT's Simile	Early wo	ork. Focus	on DOM,	too basic	1
MIT's Pot Luck	RDF	Manual	PBD	Manual	1,3,4
Dapper	DOM	Manual	Manual	Join only	1,2,4
Yahoo's Pipes	Widgets	Manual	Widgets	Union only	1,2,3
MS's Popfly	Widgets	Manual	Widgets	Widgets	1,2,4
CMU's Marmite	Widgets	Manual	Widgets	Widgets	1,2,4
Intel's Mashmaker		Require	an expert		1,2,3,4
Google MyMap	Widgets	Manual	N/A	Union only	1,2
Agent Wizard	Q/A	Q/A	Q/A	Q/A	1,3,4
Cards	DOM	Manual	N/A	Manual	1,2,4
Karma	DOM	Database	PBD	PBD	1,2,3,4

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System	Data	Source	Data	Data	Mashup Type	
	Retrieval	Modeling	Cleaning	Integration	Supported	
MIT's Simile	Early wo	ork. Focus	on DOM,	too basic	1	
MIT's Pot Luck		/ Manually			1,3,4	
Dapper	DOM	Manual	Manual	Join only	1,2,4	
Yahoo's Pipes	Widgets	Manual	Widgets	Union only	1,2,3	
MS's Popfly	Widgets	Manual	Widgets	Widgets	1,2,4	
CMU's Marmite	Widgets	Manual	Widgets	Widgets	1,2,4	
Intel's Mashmaker		Require	an expert		1,2,3,4	
Google MyMap	Widgets	Manual	N/A	Union only	1,2	
Agent Wizard	Q/A	Q/A	Q/A	Q/A	1,3,4	
Cards	DOM	Manual	N/A	Manual	1,2,4	
Karma	DOM	Database	PBD	PBD	1,2,3,4	

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System	Data	Source	Data	Data	Mashup Type
	Retrieval	Modeling	Cleaning	Integration	Supported
MIT's Simile	Early wo	ork. Focus	on DOM,	too basic	1
MIT's Pot Luck		/ Manually			1,3,4
Dapper	Mainly	focus on	extraction	/ linear	1,2,4
Yahoo's Pipes	Widgets	Manual	Widgets	Union only	1,2,3
MS's Popfly	Widgets	Manual	Widgets	Widgets	1,2,4
CMU's Marmite	Widgets	Manual	Widgets	Widgets	1,2,4
Intel's Mashmaker		Require	an expert		1,2,3,4
Google MyMap	Widgets	Manual	N/A	Union only	1,2
Agent Wizard	Q/A	Q/A	Q/A	Q/A	1,3,4
Cards	DOM	Manual	N/A	Manual	1,2,4
Karma	DOM	Database	PBD	PBD	1,2,3,4

System	Data	Source	Data	Data	Mashup Type
	Retrieval	Modeling	Cleaning	Integration	Supported
MIT's Simile	Early wo	ork. Focus	on DOM,	too basic	1
MIT's Pot Luck	RDF	/ Manually	<u>specify</u> d	ata int	1,3,4
Dapper	Mainly	focus on	extraction	/ linear	1,2,4
Yahoo's Pipes		Wid	gets		1,2,3
MS's Popfly	Fa	ancier UI/ r	nore wido	lets	1,2,4
CMU's Marmite			U U	n workflow	1,2,4
Intel's Mashmaker			an expert		1,2,3,4
Google MyMap	Widgets	Manual	N/A	Union only	1,2
Agent Wizard	Q/A	Q/A	Q/A	Q/A	1,3,4
Cards	DOM	Manual	1,2,4		
Karma	DOM	Database	PBD	PBD	1,2,3,4

System	Data Source		Data	Data	Mashup Type
	Retrieval	Modeling	Cleaning	Integration	Supported
MIT's Simile	Early wo	ork. Focus	on DOM,	too basic	1
MIT's Pot Luck	RDF	/ Manually	v specify d	ata int	1,3,4
Dapper	Mainly	focus on	extraction	/ linear [	1,2,4
Yahoo's Pipes		Wid	gets		1,2,3
MS's Popfly	Fa Fa	ancier UI/ r	nore wido	lets	1,2,4
CMU's Marmite				n workflow	1,2,4
Intel's Mashmaker			an expert		1,2,3,4
Google MyMap		Create poi	<u>nts on Ma</u>	p	1,2
Agent Wizard	Q/A	Q/A	Q/A	Q/A	1,3,4
Cards	DOM Manual N/A Man				1,2,4
Karma	DOM	Database	PBD	PBD	1,2,3,4

System	Data	Source	Data	Data	Mashup Type	
	Retrieval	Modeling	Cleaning	Integration	Supported	
MIT's Simile	Early wo	ork. Focus	on DOM,	too basic	1	
MIT's Pot Luck	RDF	/ Manually	specify d	ata int	1,3,4	
Dapper	Mainly	focus on	extraction	/ linear [	1,2,4	
Yahoo's Pipes		1,2,3				
MS's Popfly	Fa Fa	ancier UI/ 1	nore wido	lets	1,2,4	
CMU's Marmite				n workflow	1,2,4	
Intel's Mashmaker			an expert		1,2,3,4	
Google MyMap		<u>Create poi</u>	nts on Ma	p	1,2	
Agent Wizard	Q/A a	<u>pproach /  </u>	<u>inear / sc</u>	alability	1,3,4	
Cards	DOM	DOM Manual N/A Manual				
Karma	DOM	Database	PBD	PBD	1,2,3,4	

System	Data	Source	Data	Data	Mashup Type			
	Retrieval	Modeling	Cleaning	Integration	Supported			
MIT's Simile	Early wo	ork. Focus	on DOM,	too basic	1			
MIT's Pot Luck		/ Manually			1,3,4			
Dapper	Mainly	focus on	extraction	/ linear	1,2,4			
Yahoo's Pipes		Wid	gets		1,2,3			
MS's Popfly	l Fa	ancier UI/ 1	nore wido	lets	1,2,4			
CMU's Marmite				n workflow	1,2,4			
Intel's Mashmaker			an expert		1,2,3,4			
Google MyMap		<u>Create poi</u>	nts on Ma	p	1,2			
Agent Wizard	Q/A a	pproach /	<u>linear / sca</u>	alability	1,3,4			
Cards	Tuple = c	Tuple = card. Drawing links for relations						
Karma	DOM	Database	PBD	PBD	1,2,3,4			

# **Related Work: Data Extraction**

- Automatic extraction: table and lists only
  - RoadRunner (exploit HTML structure)
  - Adel (grammer induction to detect rows)
  - VisualWeb (OCR technique to detect tables)
- Semi-Automatic: require more label examples
  - WIEN (inductive less expressive than stalker)
  - Stalker (Cotesting)
  - SoftMealy (finite state transducer)
  - WHISK (rigid format, exact delimiter)
- DOM: rely on well-formed HTML and less labeling
  - Simile
  - Dapper
  - Interactive Wrapper Generation (ML + prediction on DOM) [Irmak+ 2006]
  - PLOW (add natural language)
  - Cards
  - Karma

[Crescenzi et al., 2001]

[Lerman+ 2001]

[Gatterbauer+ 2007]

[Kushmerick 1997]

[Muslea+ 1999]

[Soderland 1998]

[Huynh+2005]

[Allen+ 2007]

[Dontcheva+ 2007]

[Tuchinda+ 2008]

[Hsu 1998]

# **Related Work: Source Modeling**

- 1:1 mapping, N:M mapping
  - Schema-level match
  - TranScm [Milo+ 98] [Palopoli+99] • DIKE • Artemis [Castano+01] [Clifton+97] • Delta - +Instance-based matcher [Li 00] • SemInt • LSD [Doan 01] • ILA [Etzioni 95] • iMapp [Dhamanka 04] [Ling 01] • Clio (interactive) • Inducing Source Description [Carman 07] Karma leverages existing techniques to narrow candidate matches
    - String Similarities

[Cohen+ 2003]

# **Related Work: Data Cleaning**

- Commercial Tools: Focus on writing transformation
  - ACR/Data, Migration Architect [Chaudhuri+ 1997]
- Discrepancy Detection: Use as a stepping stone for record linkage and cleaning system
  - Levenshtein distance [Needleman+ 70] [Baeza-Yates+99]
  - Vector based
  - EM
  - SVM
- Record linkage & cleaning systems: Focus on ranking ٠
  - Fuzzy Match
  - Apollo
  - Phoebus
  - Potter's wheel
- Karma
  - Gained reference sources through source modeling process
  - Provided predefined transformations

[Ristad+ 98]

[Bilenko+ 03]

[Winkler 06]

[Chaudhuri+ 03]

[Michelson+07]

[Raman+ 01]

[Michalowski+ 05]

# **Related Work: Data Integration**

- Universal Relation: Make it easier to formulate the query but users still need to formulate the query [Ullman 1980, 1988]
- Query by example: Need to know which data sources to use and the query may not return results
  - QBE
- Retrieval by formulation: Need to understand domain model to formulate partial description
  - Helgon
  - RABBIT

- Gql

- Graphical Query Language: Users still need to navigate through sources (graphs)
  - [Benzi 1998, Haw 1994, Papantonakis 1988]
- Question-Answering Technique: Understanding about database operations required.
  - Agent Wizard [Tuchinda+ 2004]
- Interactive Schema/data integration: Understanding about source schema required
  - Clio
- Karma is based on Programming by Demonstration [Cyper 2001; Lau2001]

[Ling 01]

[Zloof 1975]

[Fischer 1989]

[Williams 1982]

# Conclusion

- Mashup is a fast growing area
  - Need an efficient way to for casual web users to build it.
- Contributions
  - A PBD approach that uses a single table for building a Mashup
  - An integrated approach that links four Mashup buildling issues.
  - A query formulation technique that allows users to specify examples to build complicated queries.
- Evaluated the validity of the Karma approach
  - Subjects were able to complete Mashup building tasks in Karma
  - The overall improvement is at least 3.5

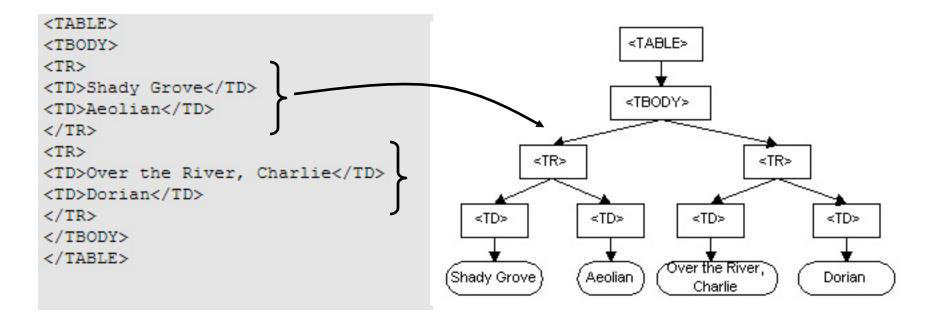
#### **Future Work**

- Customizing display by examples
- Implementing feedback and quality
- Adding planning components to handle dynamic data.

# Thank You!

# Backup Slides

# **Document Object Model (DOM)**



# **Vertical Expansion**

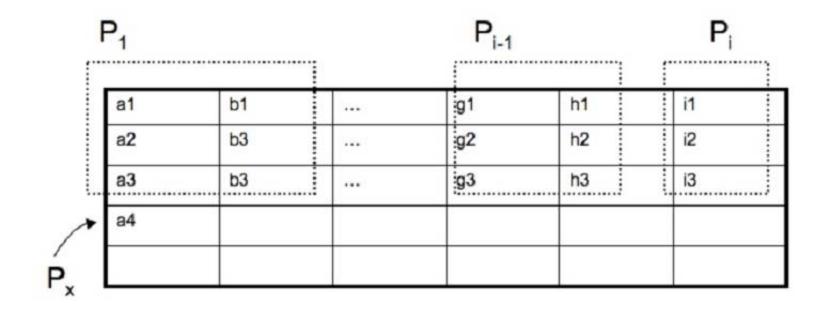


Table 8.4: Normalized data for task 1. E stands for data extraction, M stands for source modeling, and C stands for data cleaning. The asterisk indicates time substitution when failures happen. The data is reported in minutes.

Task1		Dappe	er/Pipes			Ka	rma	
Subject	E	Μ	$\mathbf{C}$	Total	Е	Μ	С	Total
No.1	*5:00	0:20	*5:00	10:20	2:19	1:08	1:00	4:27
No.2	1:43	0:30	*5:00	7:13	1:00	0:40	0:29	2:09
No.3	n/a	n/a	n/a	n/a	n/a	n/a	n/a	n/a
No.4	0:52	0:48	*5:00	6:40	1:12	1:00	0:50	3:02
No.5	<b>5:</b> 00	0:35	*5:00	10:35	1:15	1:18	1:20	3:53
No.6	2:30	0:15	*5:00	7:45	1:00	0:54	0:28	2:22
No.7	1:20	0:22	*5:00	6:42	0:51	0:51	0:46	2:28
No.8	1:40	0:14	*5:00	6:54	1:04	0:41	0:33	2:19
No.9	1:26	0:16	*5:00	6:42	1:14	1:00	1:10	3:24
No.10	1:39	0:10	*5:00	6:49	0:53	0:42	0:50	2:26
No.11	2:00	0:19	*5:00	7:19	1:04	1:00	0:53	2:57
No.12	2:00	0:49	2:00	4:49	1:07	1:00	0:40	2:47
No.13	2:00	0:05	*5:00	7:05	0:58	0:50	0:56	1:44
No.14	2:46	0:15	*5:00	8:01	1:12	0:45	0:48	2:45
No.15	2:27	0:14	3:11	5:52	1:10	0:49	1:20	3:19
No.16	1:16	0:12	*5:00	6:28	0:58	0:42	0:25	2:05
No.17	n/a	n/a	n/a	n/a	2:00	1:00	0:50	3:50
No.18	2:30	0:14	*5:00	7:44	1:06	1:10	1:46	4:02
No.19	1:38	0:47	1:20	3:45	1:20	0:49	0:35	2:44
No.20	1:30	0:16	*5:00	6:46	1:04	0:44	0:35	2:23

Table 8.5: Normalized data for task 2. E stands for data extraction, M stands for source modeling, C stands for data cleaning, and I stands for data integration. The asterisk indicates time substitution when failures happen. The data is reported in minutes.

Task2		Da	apper/P	ipes				Karm	a	
Subject	E	Μ	$\mathbf{C}$	Ι	Total	Е	Μ	С	Ι	Total
No.1	4:38	0:22	2:45	1:15	9:00	1:26	0:43	0:43	0:00	2:52
No.2	1:35	0:12	3:30	0:12	<b>5</b> :29	0:50	0:57	0:57	0:00	2:44
No.3	*5:00	0:25	*5:00	*5:00	15:25	2:52	1:00	3:00	0:00	5:52
No.4	4:49	0:17	3:29	0:38	9:14	1:26	0:48	1:03	0:00	3:18
No.5	*5:00	0:29	1:44	1:16	8:29	1:43	0:45	1:20	0:00	3:48
No.6	*5:00	0:20	*5:00	*5:00	15:20	2:07	0:30	0:50	0:00	3:27
No.7	2:17	0:15	4:46	0:18	7:36	1:13	0:25	0:52	0:00	2:31
No.8	3:23	0:21	*5:00	*5:00	13:44	1:10	0:21	0:24	0:00	1:55
No.9	4:11	0:21	*5:00	*5:00	14:32	1:22	0:47	2:11	0:00	4:20
No.10	2:16	0:07	3:14	0:20	5:50	1:04	0:20	1:06	0:00	2:30
No.11	3:04	0:17	*5:00	*5:00	13:21	1:06	0:34	0:53	0:00	2:33
No.12	2:00	0:27	*5:00	0:20	7:47	1:23	0:30	0:37	0:00	2:30
No.13	*5:00	0:07	1:43	0:10	7:00	1:42	0:32	0:41	0:00	2:55
No.14	3:03	0:23	4:42	0:10	8:21	1:40	0:31	0:56	0:00	3:07
No.15	2:06	0:12	3:13	0:22	5:53	1:30	0:24	2:05	0:00	3:59
No.16	3:58	0:11	3:29	0:27	8:05	0:51	0:17	1:00	0:00	2:08
No.17	4:15	0:28	3:39	0:30	8:52	1:04	0:28	1:18	0:00	2:50
No.18	*5:00	0:23	*5:00	*5:00	15:23	1:17	0:30	1:10	0:00	2:57
No.19	4:01	0:14	2:42	0:21	7:16	1:39	0:21	0:50	0:00	2:50
No.20	1:36	0:43	0:36	0:22	3:17	1:07	0:28	0:40	0:00	2:15

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Table 8.6: Normalized data for task 3. E stands for data extraction, M stands for source modeling, and I stands for data integration. The asterisk indicates time substitution when failures happen. The data is reported in minutes.

Task3		Dapp	er/Pipes	3	Karma			
Subject	E	Μ	Ι	Total	E	Μ	Ι	Total
No.1	1:30	0:26	*5:00	6:56	0:14	0:00	2:16	2:30
No.2	0:30	0:10	*5:00	5:40	0:25	0:00	0:26	0:54
No.3	1:00	0:15	*5:00	6:15	0:15	0:00	0:44	0:59
No.4	0:40	0:16	*5:00	5:56	0:20	0:00	1:06	1:26
No.5	0:40	0:14	*5:00	5:54	0:20	0:00	0:37	0:57
No.6	0:30	0:10	*5:00	5:40	0:20	0:00	0:31	0:51
No.7	0:27	0:10	*5:00	5:37	0:14	0:00	0:50	1:04
No.8	0:29	0:20	*5:00	5:49	0:30	0:00	0:51	1:21
No.9	0:40	0:23	*5:00	6:03	0:13	0:00	0:44	0:57
No.10	0:30	0:10	*5:00	5:40	0:20	0:00	0:35	0:55
No.11	0:51	0:20	*5:00	6:11	0:16	0:00	1:05	1:21
No.12	1:05	0:18	*5:00	6:23	0:30	0:00	0:46	1:16
No.13	0:31	0:14	*5:00	5:45	0:16	0:00	0:57	1:13
No.14	0:36	0:14	*5:00	5:50	0:14	0:00	2:00	2:14
No.15	0:26	0:21	*5:00	5:47	0:30	0:00	0:45	1:15
No.16	0:27	0:13	*5:00	5:40	0:15	0:00	0:56	1:11
No.17	0:33	0:38	1:56	3:07	0:30	0:00	0:46	1:16
No.18	1:03	0:07	*5:00	6:10	0:20	0:00	1:10	1:30
No.19	0:33	0:17	*5:00	<b>5:5</b> 0	0:25	0:00	1:20	1:45
No.20	0:18	0:13	*5:00	5:31	0:12	0:00	0:44	0:56