



Semantic Challenges in Getting Work Done

Yolanda Gil

Information Sciences Institute and Department of Computer Science University of Southern California

> http://www.isi.edu/~gil @yolandagil gil@isi.edu

Keynote at the International Semantic Web Conference (ISWC)
October 21, 2014

Outline

- Managing work
 - Personal to do lists
- 2. Knowledge rich tasks in science
 - Semantic workflows
- 3. Collaborative tasks in science
 - Organic data science
- 4. Closing thoughts

Outline

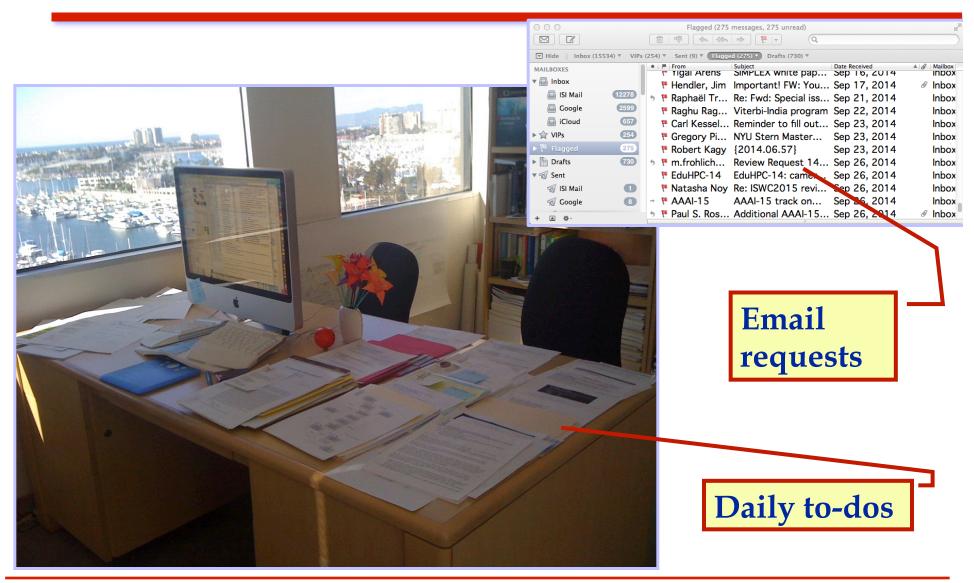
- Managing work
 - Personal to do lists
- 2. Knowledge rich tasks in science
 - Semantic workflows
- 3. Collaborative tasks in science
 - Organic data science
- 4. Closing thoughts

2 semantic challenges

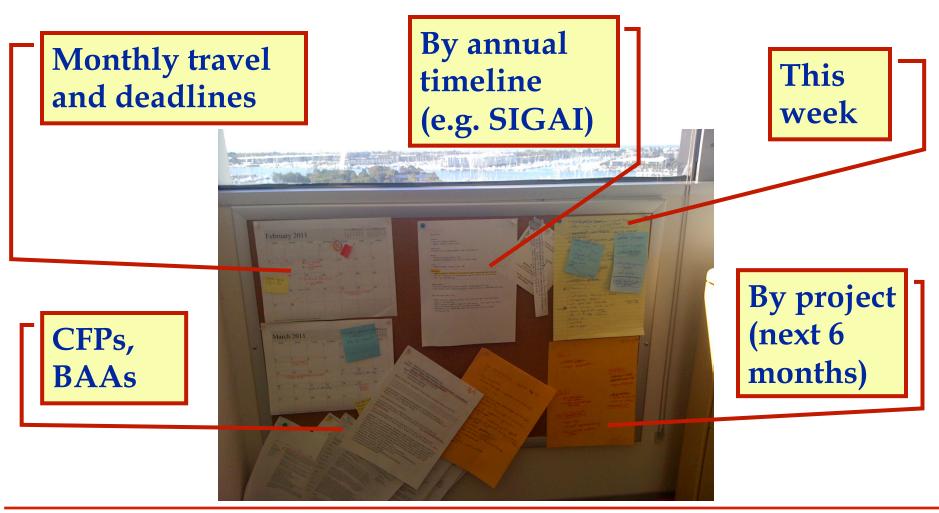
2 semantic challenges

2 semantic challenges

To Dos

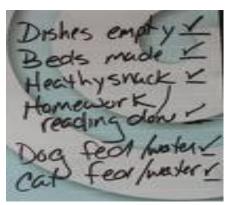


To Dos



To Do Lists

- To do lists are pervasive [Kirsh 01; Norman 91]
 - Used by more than 60% of people for personal information [Jones & Thomas 97]
 - Used more than calendars, contact lists, etc.
- Prior research focused on user studies
 - [Bellotti et al 04; Dey et al 00]









- Opportunity for assistance
 - Major potential impact on productivity

To Do List Management: Opportunities for Interpretation-based Assistance

To Do List Manager

Automate through agents

Anticipate missing entries & sub-tasks

Group and organize

Get advice from others

Get information from Web

What Are To Do Items Like: FB app

DIET!!!

Shoes? Debate... need sneakers

Renew registration for car

Be a true Christian

Hotel reservation Mafia blog update

Pay bills Think about more Facebook Money Making Ideas!

Apply for financial aid Ending of reproductive abilities

Print plane tickets Get off my lazy arse and start achieving some stuff

Renew BOFA card before you leave for summer

Mettre les images des captures sur Facebook

Spinatch and Bashamel Cupcakes

Skriva kod till Simons webbsida

Watch 'arry pottaaa!

Buy Air Blades Mk2 (Imperial)

Buy Ablative Shell (Imperial)

Ruff Racing Hyperblack 278 19'' 275/35 &

245/40 wrapped in NITTO 555R's

Order more AA Eneloop batteries

Return Fan to Westside via UPS

Return ugly jacket

- ~1500 items collected from ~325 people
- Many are not amenable to automation
- Many could be automated fully or in part

What Are To Do Items Like: Office

- Unusual structure
 - No verb: "quarterly report to Joe"
 - Abbreviations (also typos): "Sched wed 15 ISI"
 - Questions: "How to extract data for Steve"
- Many ways to refer to the same task
 - "Meet with John about paper", "Discuss paper with John", ...
- Incomplete task specifications
 - "Schedule meeting with John"
- Ambiguous references out of context
 - "Meet about paper"
 - "Meet with Raytheon folks"
- Personal items
 - "Walk the dog"

- Corpus of 2400 to-do entries from users of CALO office assistant
- 77% lack a verb
- 56% missing at least one argument
- 14% could be automated by agents

Opportunities

To Do List Manager

Automate through agents

Anticipate missing entries & sub-tasks

Group and organize

Get advice from others

Get information from Web

Agents

Colleagues

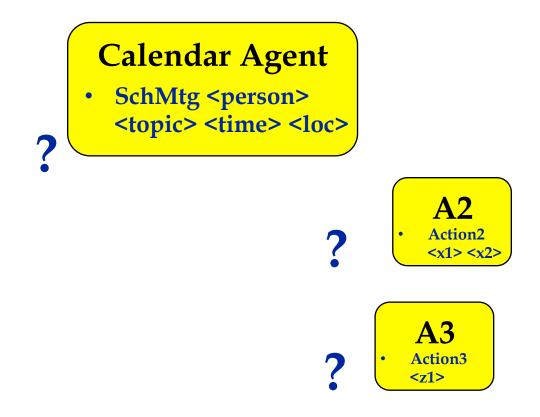
On-Line Resources

Agents: Beamer for CALO and Radar [Gil & Ratnakar AAAI 2008]

Match agent capabilities to user's to dos

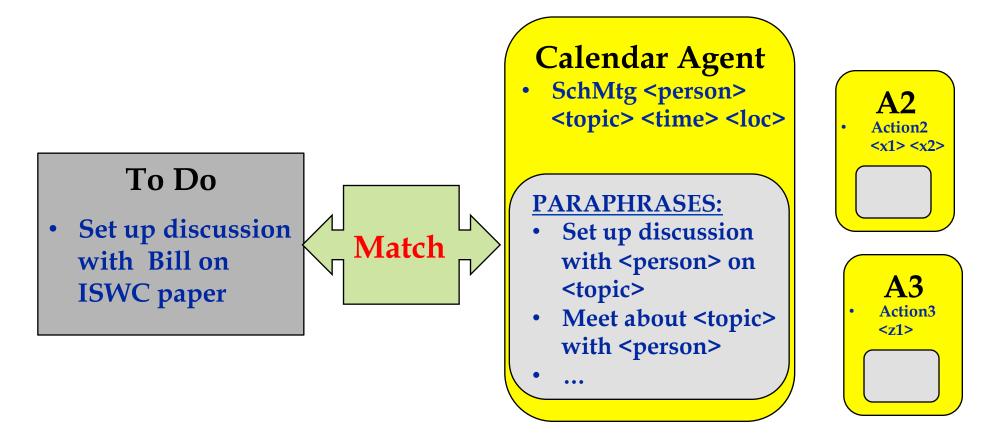
To Do

 Set up discussion with Bill on ISWC paper



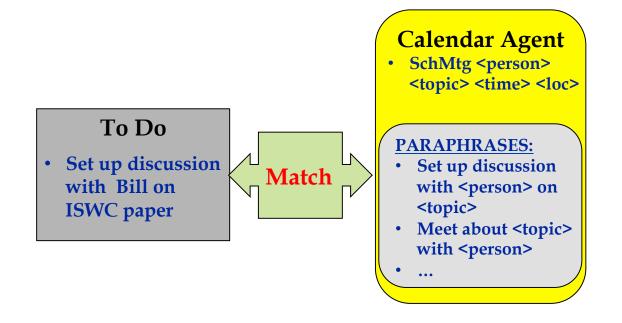
Agents: Beamer for CALO and Radar [Gil & Ratnakar AAAI 2008]

Use paraphrase patterns of agent capabilities to match them to user's to dos

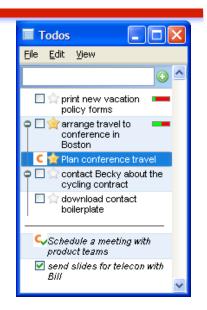


Agents: Beamer for CALO and Radar [Gil & Ratnakar AAAI 2008]

■ Use paraphrase patterns of agent capabilities to match them to user's to dos



- Evaluation with CALO office assistant corpus
 - 86.7% accuracy in detecting relevance to agents
 - only 0.2 to 0.4 edits needed to set up task parameters





The Need for Semantics

Knowledge

To Do List Manager

Automate through agents

Anticipate missing entries & sub-tasks

Group and organize

Get advice from others

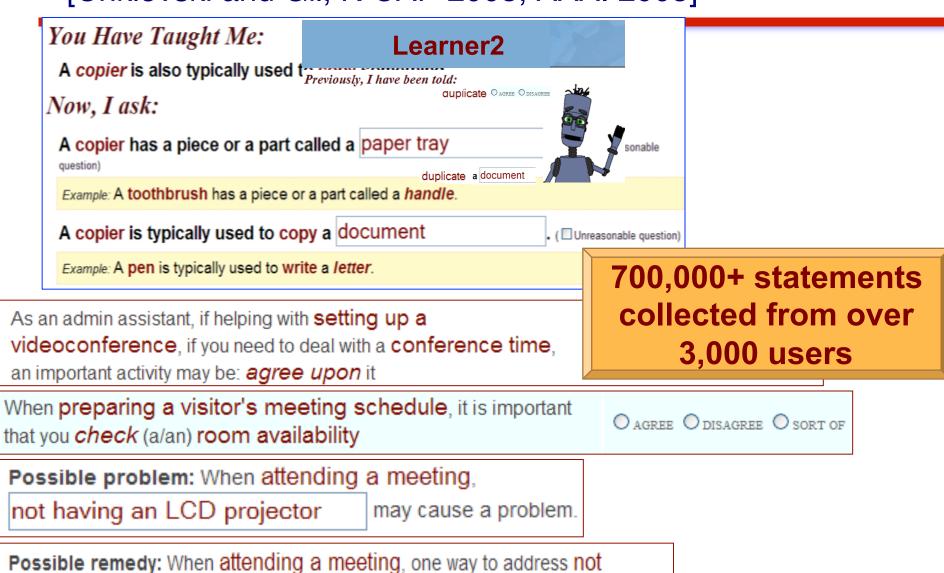
Get information from Web

Paraphrase Game [Chklovski 2005]

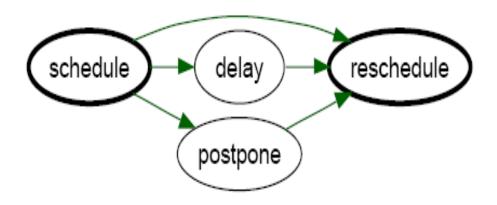


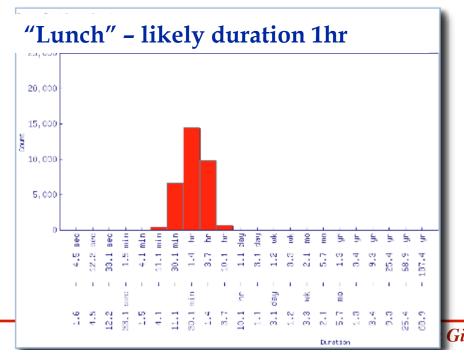
Common (Sense) Knowledge [Chklovski and Gil, K-CAP 2005, AAAI 2005]

having an LCD projector is to locate a portable projector

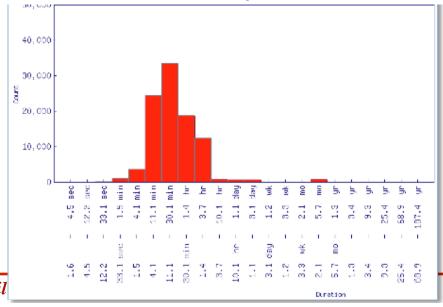


VerbOcean [Chkovski and Pantel, IJCNLP 2005]





"Presentation" - likely duration 10mins

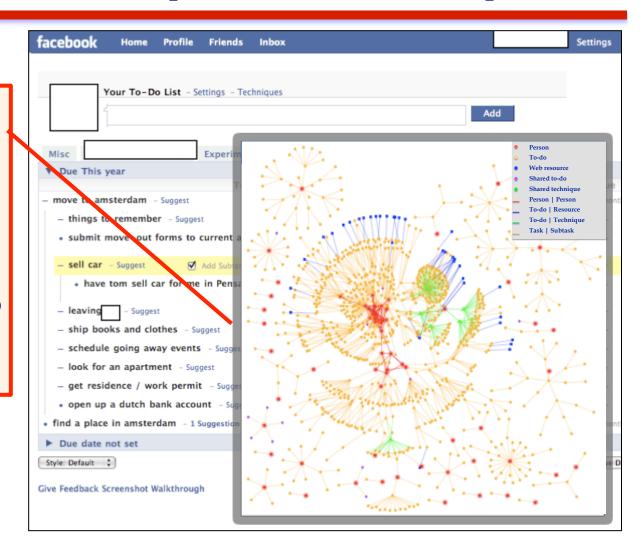


Managing To Dos through Colleagues: Social Task Networks [Groth et al 2010]

To Do app for FB

Social task networks

- People linked to their to-dos
- To-dos linked to their subtasks
- Tasks are linked to URIs which link to web resources
- Open TaskRepository usingLinked DataPrinciples



Managing To Dos through On-Line Resources [Vrandecic and Gil IUI 2011]



[debug] Displaying 6 How-Tos that we think may match. Tell us if they do! Search powered by Google.

pack for a camping trip

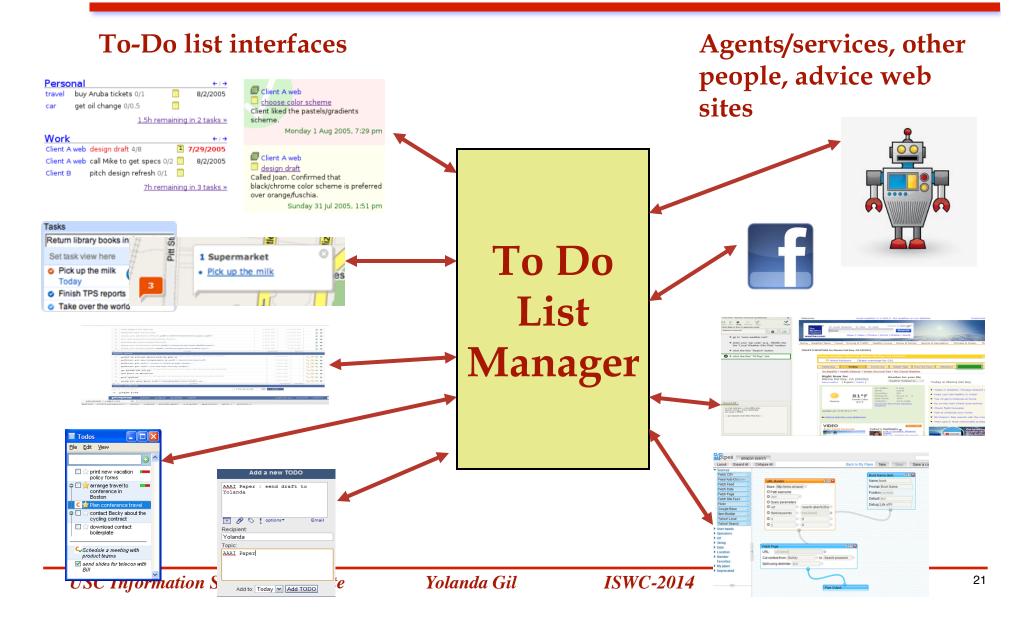
	Pack for a Camping Trip 13 steps	Source: wikiHow	Find out where you are going, how long, what you will be staying in, etc
•	Pack Light for a Camping	Trip Source: <u>wikiHow</u>	Three months before your camping trip, do a full inventory of your camping supplies
	Prepare for a Camping Tri	ip Source: <u>wikiHow</u>	Think about what you have to bring and what you want to bring
			Think about how much luggage is appropriate
	Pack for a Camping Trip As a Teen		
	Girl 10 steps	Source: wikiHow	Make a list of things to bring! (See Things Youll Need for a list of suggestions) Make sure you take enough things but dont over do it as you may bring too much and
	Look Good on a Camping Trip (Girls)		not find anything.
	7 steps	Source: <u>wikiHow</u>	
			Go over your list
	Pack for a Week Away Fro	om Home	

Submit

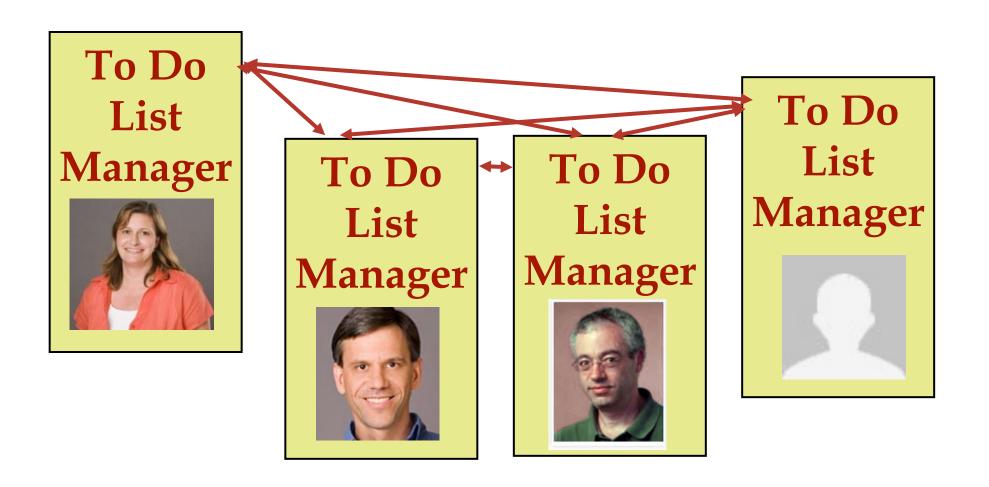
Some Readings

- Yolanda Gil, Varun Ratnakar, Timothy Chklovski, Paul T. Groth, Denny Vrandecic: "Capturing Common Knowledge about Tasks: Intelligent Assistance for To Do Lists." ACM Transactions on Interactive Intelligent Systems, 2(3). 2012.
- Hans Chalupsky, Yolanda Gil, Craig A. Knoblock, Kristina Lerman, Jean Oh, David V. Pynadath, Thomas A. Russ, Milind Tambe: "Electric Elves: Agent Technology for Supporting Human Organizations." AI Magazine 23(2): 11-24 (2002)

A Semantic Challenge: Managing Personal To Dos



A Semantic Challenge: Coordinating To Dos of Different People

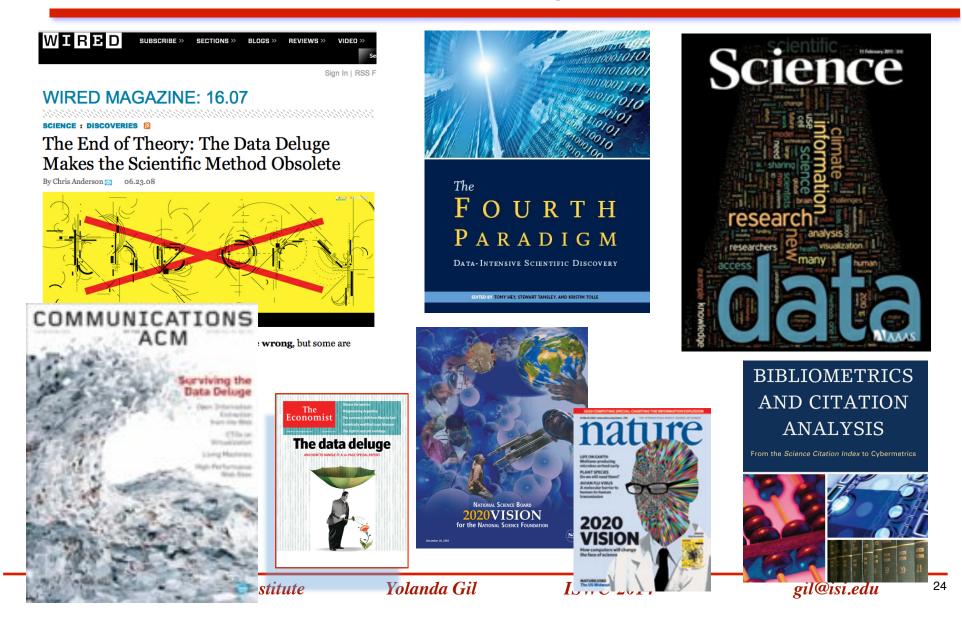


Semantic Challenges in Getting Work Done

- To dos
 - Managing personal to dos
 - Managing coordinated to dos
- Knowledge rich tasks in science

Open science

Data-Intensive Computing in Science



The Bottleneck is the Process, Not the Data!

Today: significant human bottleneck in the scientific process

What is the state of the art?

What is a good problem to work on?

What is a good experiment to design?

What data should be collected?

What is the best way to analyze the data?

What are the implications of the experiments?

What are appropriate revisions of current models?

- Need to help machines understand the scientific research process in order to assist scientists
 - Semantics can be a game changer

Text Extraction in Hanalyzer (L. Hunter, U. Colorado)



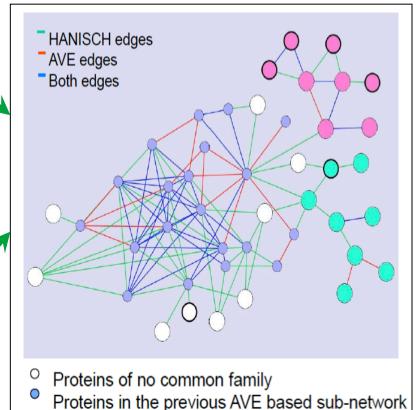
Text extraction from publications

The significance of the interaction between DAZAP1 and DAZL/DAZ remains to be defined. These proteins may act together to facilitate the expression of a set of genes in germ cells For example, DAZAP1 could be involved in the transport of the mRNAs of the target genes of DAZL. Alternatively, DAZL and DAZAP1 may act antagonistically to regulate the timing and the level of expression. Such an antagonistic interaction between two interacting RNA-binding proteins is exemplified by the neuron-specific nuclear RNA-binding protein, Nova-1. Nova-1 regulates the alternative splicing of the pre-mRNAs encoding neuronal inhibitory glycine eceptor α2 (GlyR α2) [23]. The ability of Nova-1 to activate exon selection in neurons is antagonized by a second RNA-binding protein, brPTB (brain-enriched polypyrimidine tract binding protein), which interacts with Nova-1 and inhibits its function [24]. DAZAP1 could function in a similar manner by binding to DAZL and inhibiting its function. Comparing the phenotypes of <code>DazII</code> and <code>DazapI</code> single and double knock-out mice may provide some clues to the significance of their interaction. Dazil knock-out mice have already been generated and studied [6]. The spermatogenic defect in the male becomes apparent only after day 7 post partum when the germ cells are committing to meiosis (H. Cooke, personal communication). The genomic structure of Dazap1, delineated here, should facilitate the generating of Dazap1 null mutation



Semantic integration of biomedical databases

Generation of interesting new hypotheses



oinferred synapse signaling proteins

Inferred myogenic proteins

Robot Scientist [King et al 2009]



Intelligent Science Assistants

What is the state of the art?

What is a good problem to work on?

What is a good experiment to design?

What data should be collected?

What is the best way to analyze the data?

Yolanda Gil

What are the implications of the experiments?

What are appropriate revisions of current models?

Timely Analysis of Environmental Data [Gil et al ISWC 2011]

With Tom Harmon (UC Merced), Craig Knoblock and Pedro Szekely (ISI)



California's Central Valley:

• Farming, pesticides, waste

S NOAA - National Oceanic and X

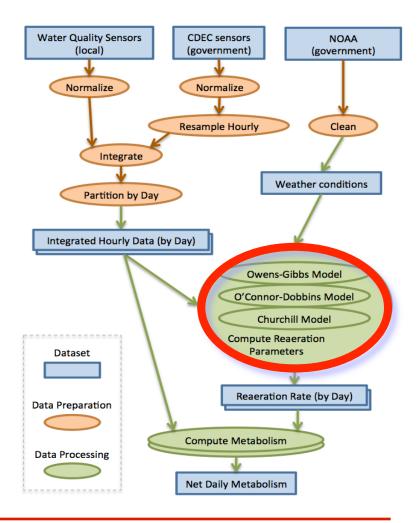
COUNTDOWN TO LAUNC

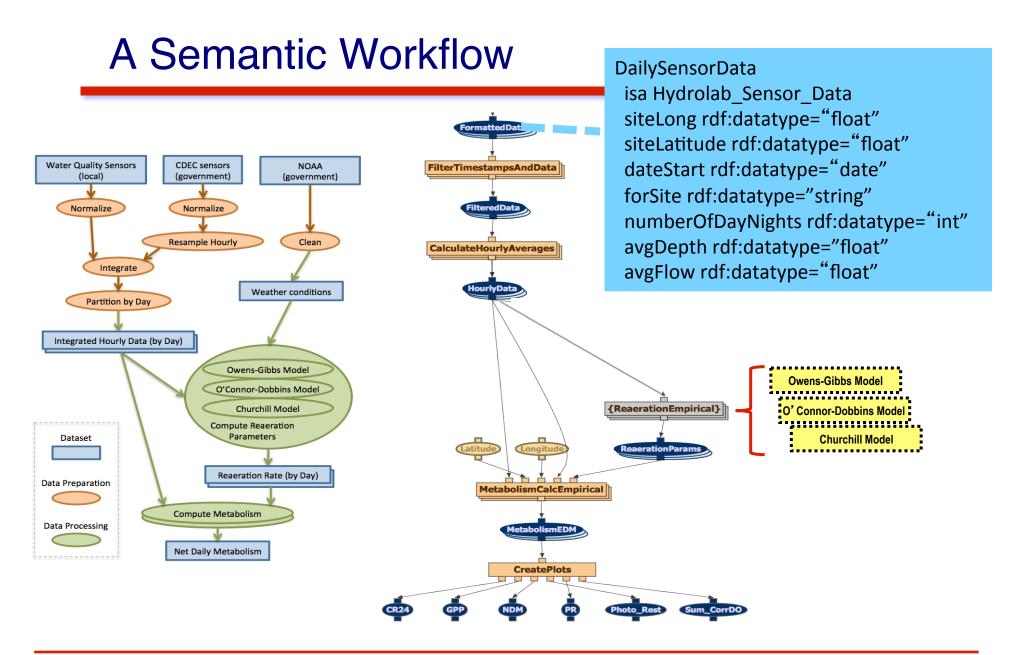
- Water releases
- Restoration efforts







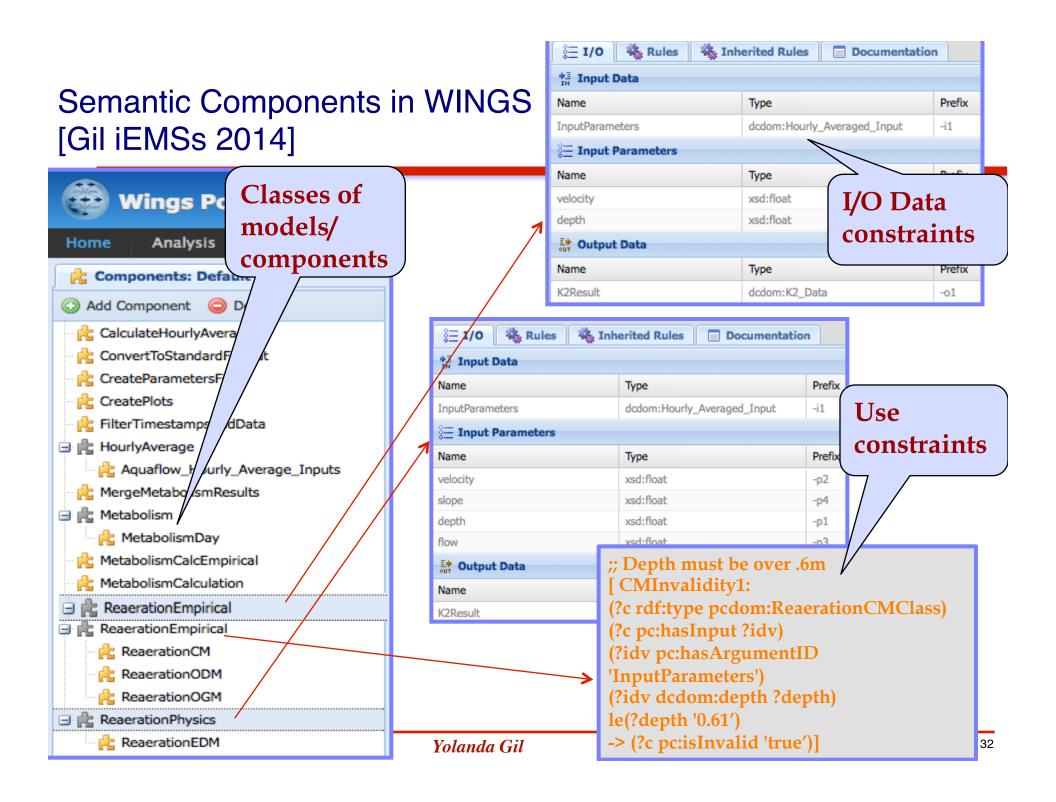




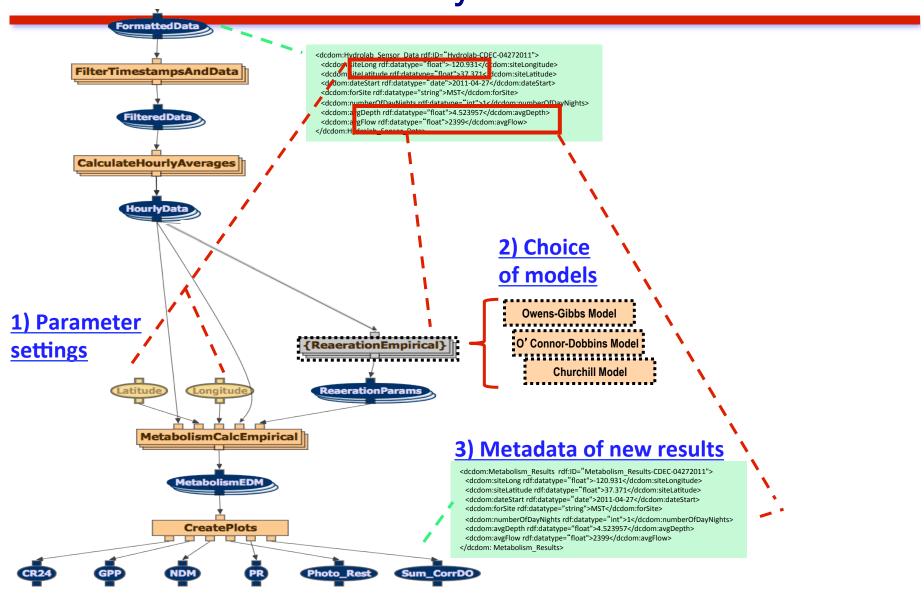
Semantic Workflows in Wings [Gil et al 10][Gil et al 09][Kim & Gil et al 08][Kim et al 06]

- Workflows are augmented with semantic constraints
 - Each workflow constituent has a variable associated with it
 - Workflow components, arguments, datasets
 - Constraints are used to restrict workflow variables
 - Can define abstract classes of components
 - Concrete components model exec. codes
- Workflow reasoners propagate and use semantic constraints
- Uses semantic web standards: OWL/ RDF, SPARQL
- Compilation of workflows to scalable execution infrastructure

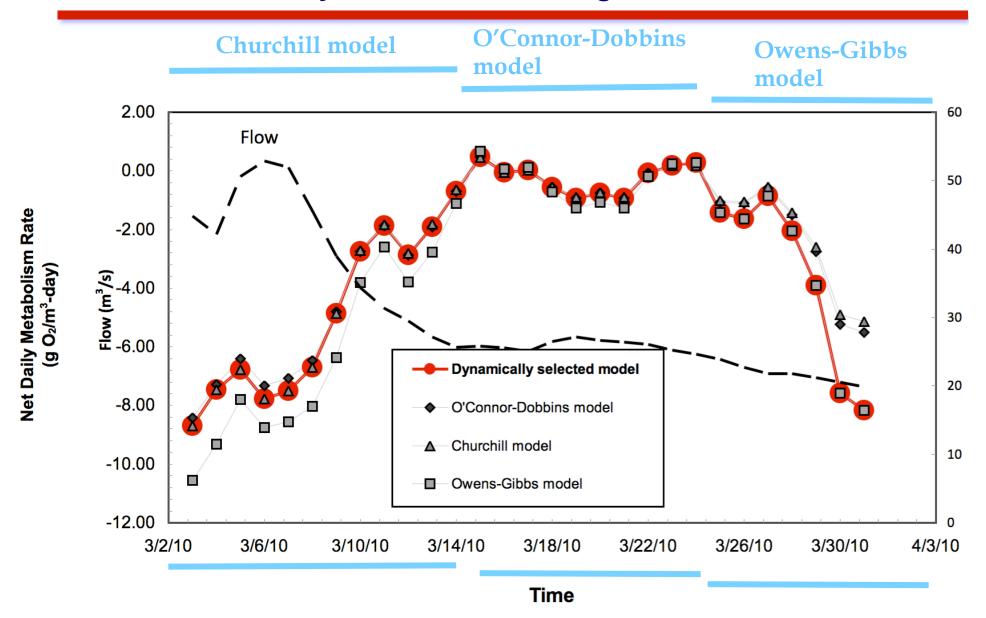




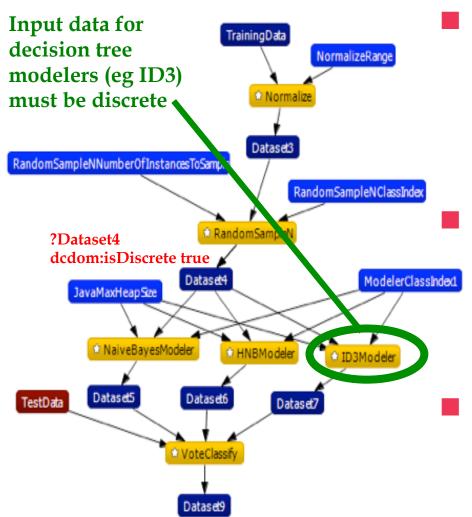
WINGS Specializes Workflow Based on Characteristics of Daily Data



WINGS Dynamically Selects Appropriate Model Based on Daily Sensor Readings



WINGS Workflow Reasoners



Key idea: Skeletal planning, where constraints for each component are propagated through a fixed workflow structure (the skeleton)

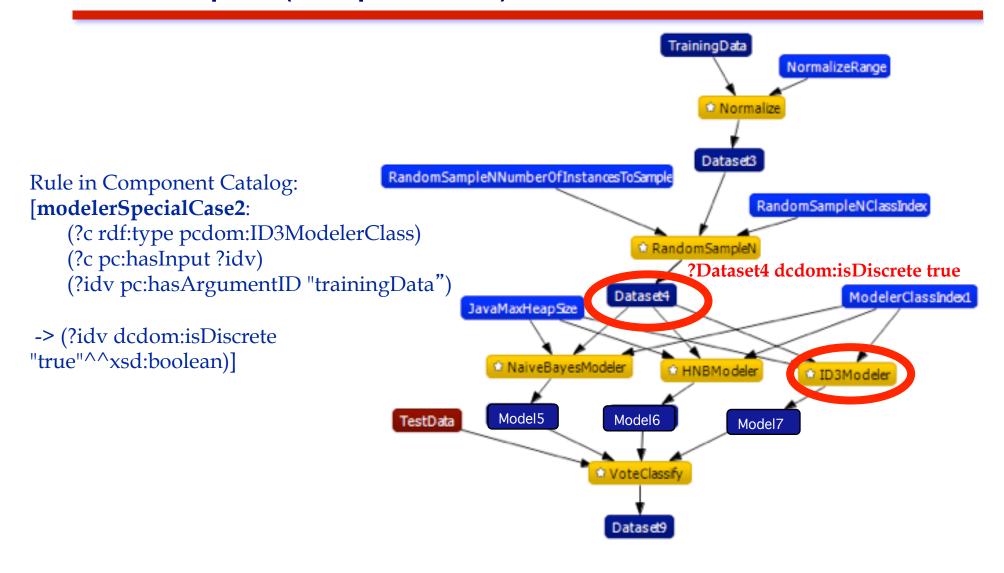
Phase 1: Goal Regression

- Starting from final products, traverse workflow backwards
- For each node, query for constraints on inputs

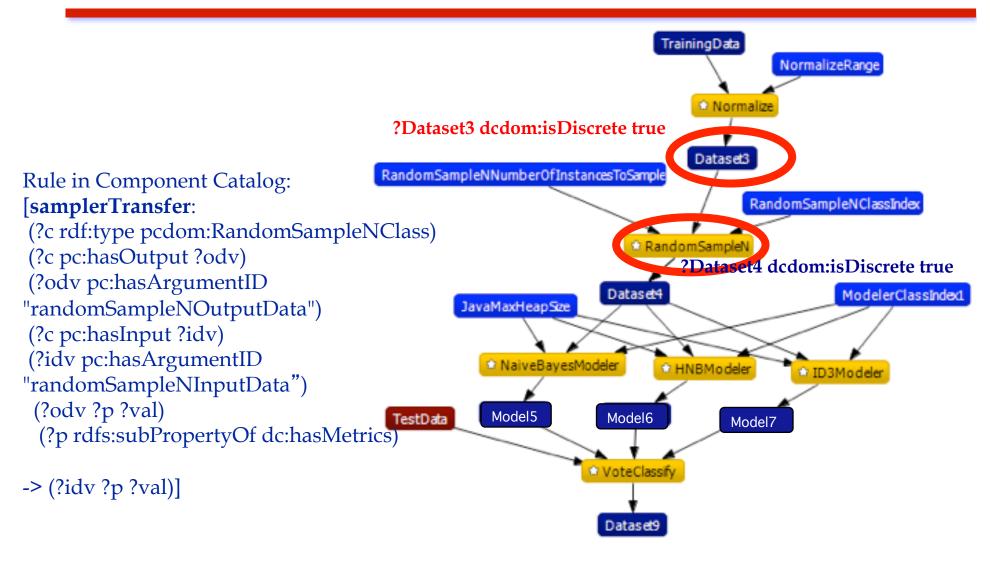
Phase 2: Forward Projection

- Starting from input datasets, traverse workflow forwards
- For each node, query for constraints

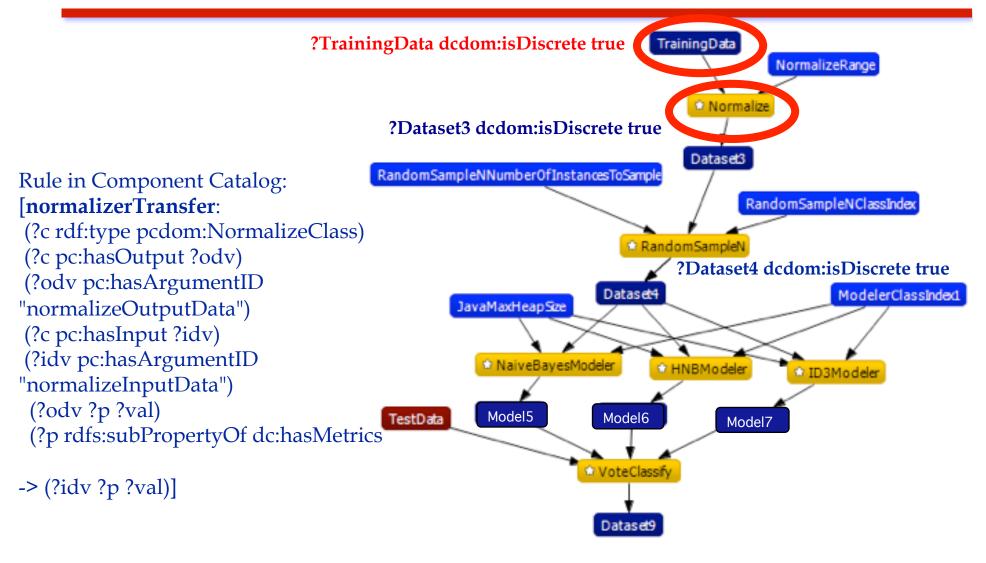
Example (Step 1 of 5)



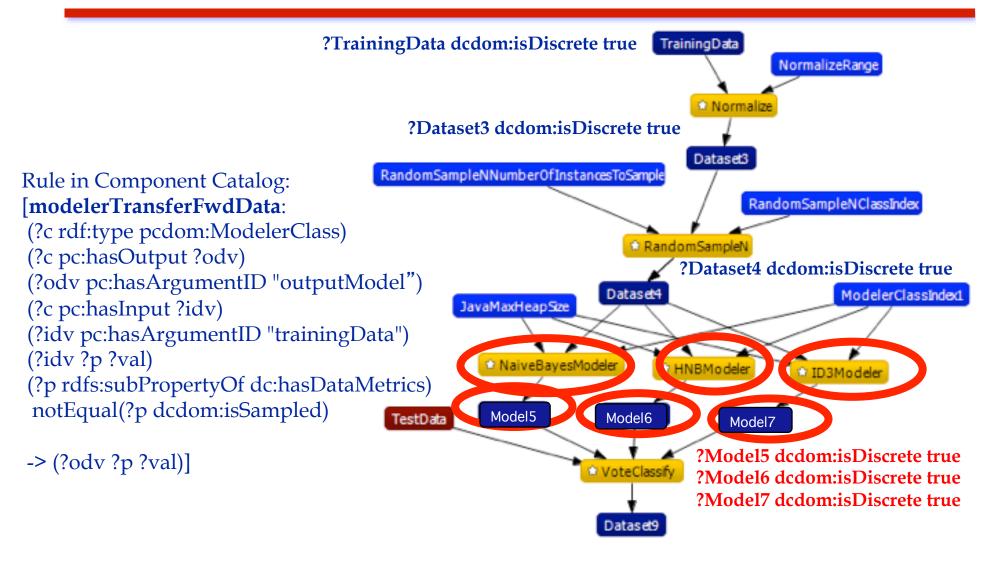
Example (Step 2 of 5)



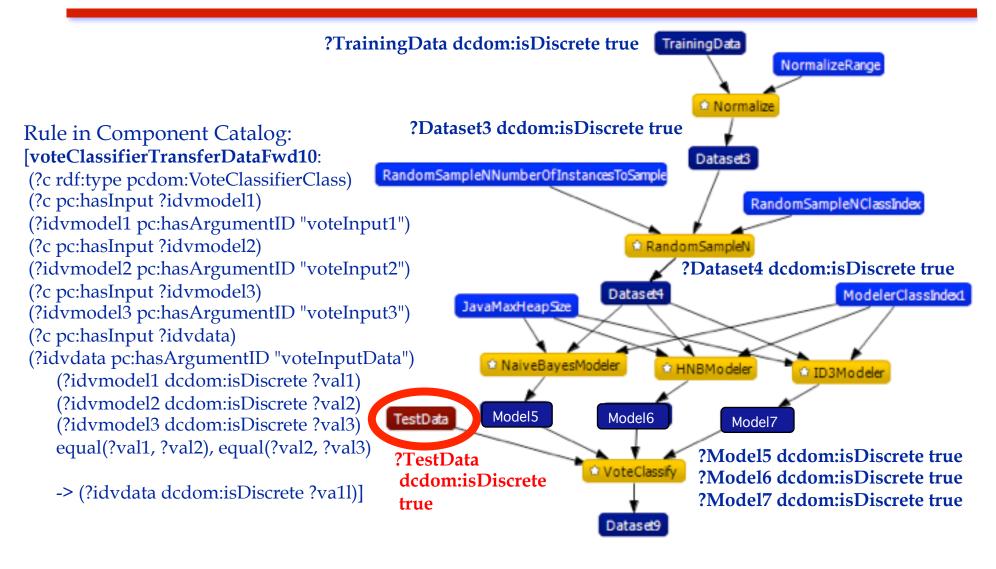
Example (Step 3 of 5)



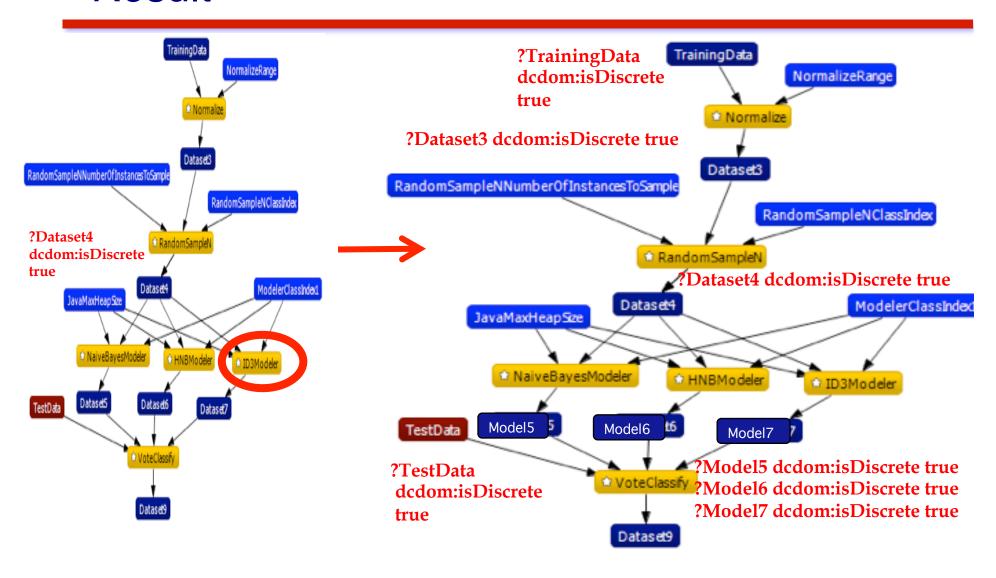
Example (Step 4 of 5)



Example (Step 5 of 5)

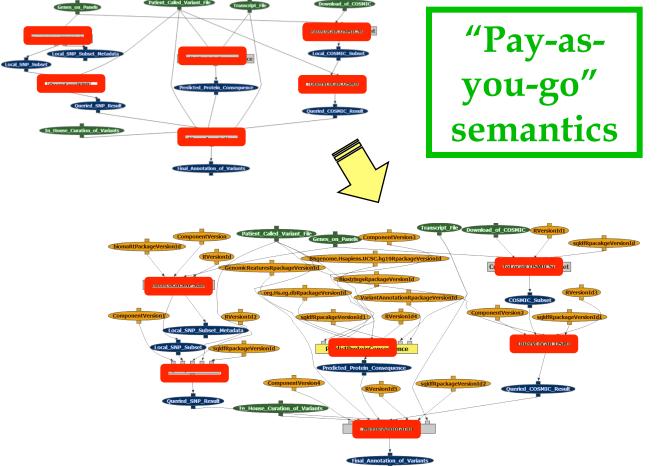


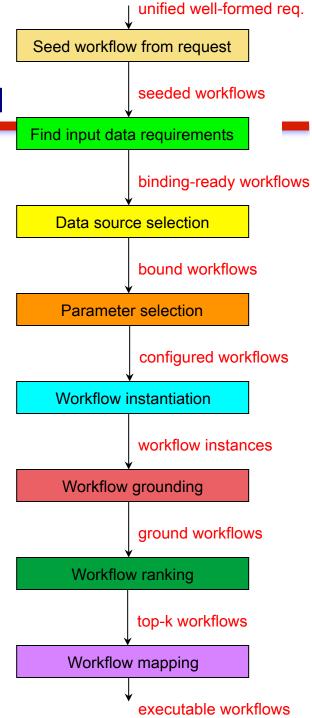
WINGS Workflow Reasoners: Result



WINGS Automatic Workflow Generation Algorithm [Gil et al JETAI 2011]

Work with P. Gonzalez (UCM) and Jihie Kim (ISI) Workflows with S. McWeeney & C. Zhang (OHSU)





Workflows

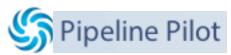
- Workflow systems
 - [Goble et al 2007]
 - [Ludaescher et al 2007]
 - [Freire et al 2008]
 - [Mattmann et al 2007]
 - [Mesirov et al 2009]
 - [Dinov et al 2009]
- Workflow representations
 - [Moreau et al 2010]
 - [IBM/MSR 2002]

























Semantic Process Models

Composition from first principles

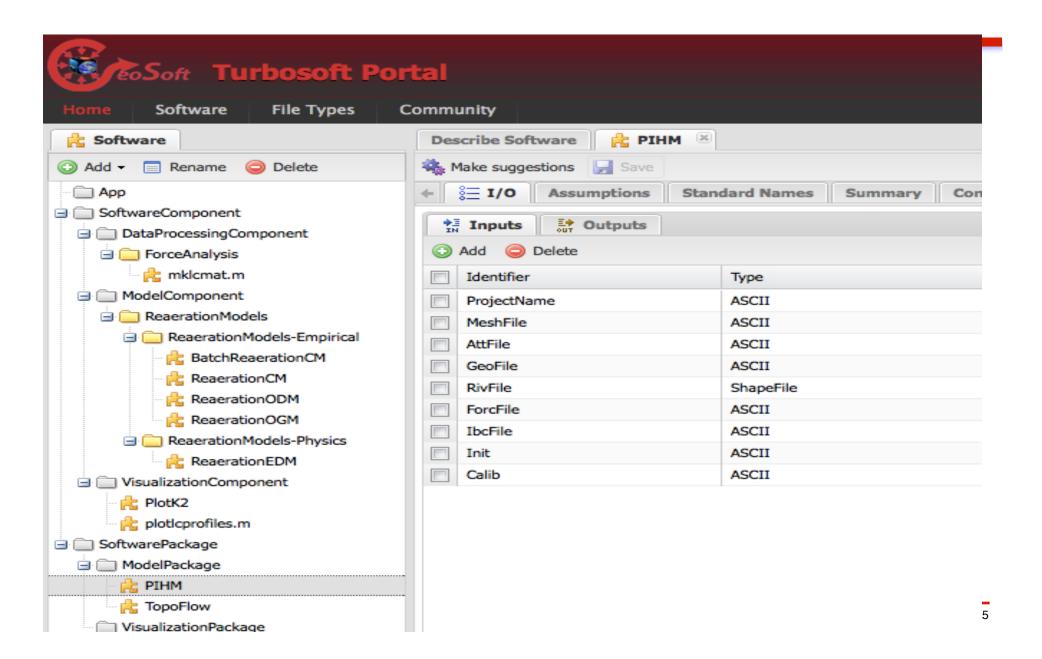
- [McIlraith & Son KR 2002] [Sohrabi et al ISWC 2006] [Sohrabi & McIlraith ISWC 2009] [Sohrabi & McIlraith ISWC 2010]
- [McDermott AIPS 2002]
- [Kuter et al ISWC 2004] [Sirin et al JWS 2005] [Kuter et al JWS 2005] [Lin et al ESWC 2008]
- [Lecue ISWC 2009]
- [Calvanese et al IEEE 2008]
- [Bertolli et al ICAPS 2009]
- [Li et al ISSC 2011]

Representations

- [Burstein et al ISWC 2002] [Martin et al ISWC 2007]
- [Domingue & Fensel IEEE IS 2008] [Dietze et al IJWSR 2011] [Dietze et al ESWC 2009]
- [Fensel et al 2011] [Vitvar et al ESWC 2008] [Roman et al AO 2005]

Semantic Descriptions of Software Components in Geosciences

Work with C. Duffy (PSU), S. Peckham (CU), C. Mattmann (JPL), J. Howison (UT)



CSDMS Standard Names [Peckham iEMSs 2014]

http://csdms.colorado.edu/



Objects

acetic_acid

air

air_ball

air_carbon_dioxide

air_helium_plume

air_to_fuel

air_visible_light

air_water

air_water_vapor

airfoil

airfoil_enclosing_curve

airplane

airplane_wing

alaskan_black_bear_brain_to_body

alaskan_black_bear_head

aluminum

atmosphere

atmosphere_aerosol

atmosphere_carbon_dioxide

atmosphere_cloud

atmosphere_clox_as_chlorine



Quantities

absorbed_shortwave_irradiation_flux actual_bond_angle

affinity

age

air_dried_pressure_head

albedo

altitude

amplitude

area

area_fraction

aspect_anale

aspect_ratio

avogadro_constant

backscattered_shortwave_irradiation_f

Yolanda Gil

Operators

anomaly

area_time_integral

azimuth_angle

curl

downward_component

downward_eastward_component

downward_northward_component

eastward_component

eastward_downward_component

gradient

increment

laplacian

lower_limit

magnitude

max

max_over_grid

max_over_grid_and_time

max_over_increment

max_over_time

mean_over_domain

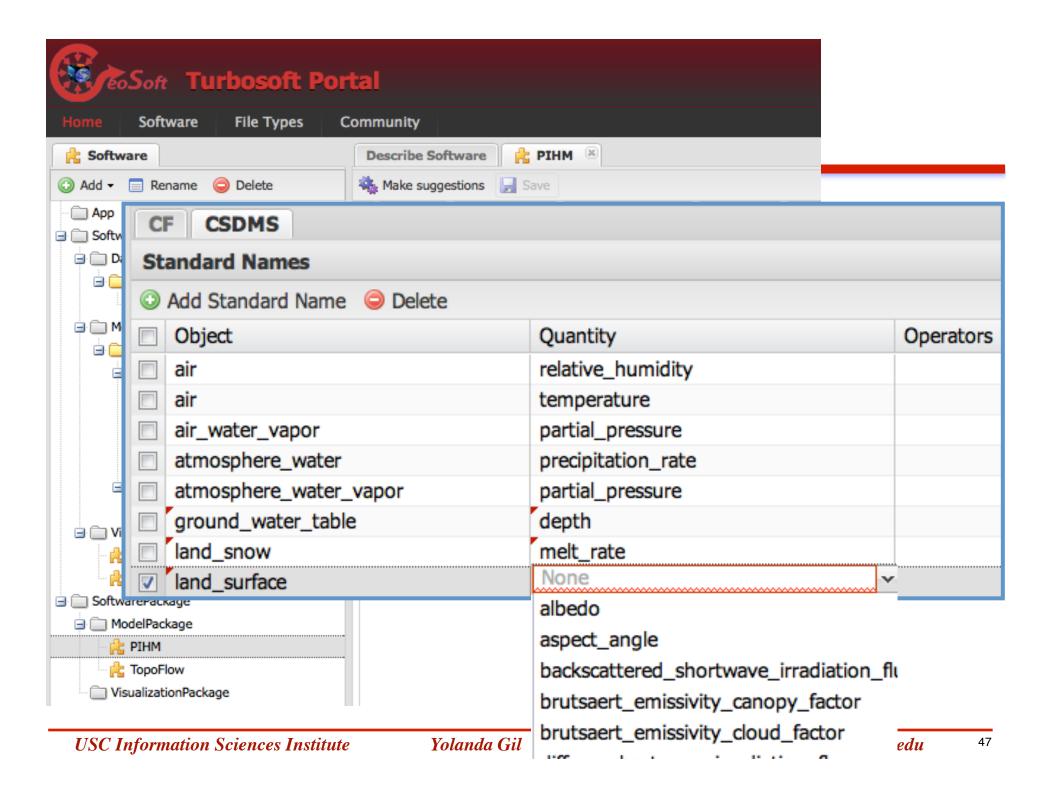
mean_over_increment

min

min_over_grid

min_over_increment

ISWC-2014 normal_component



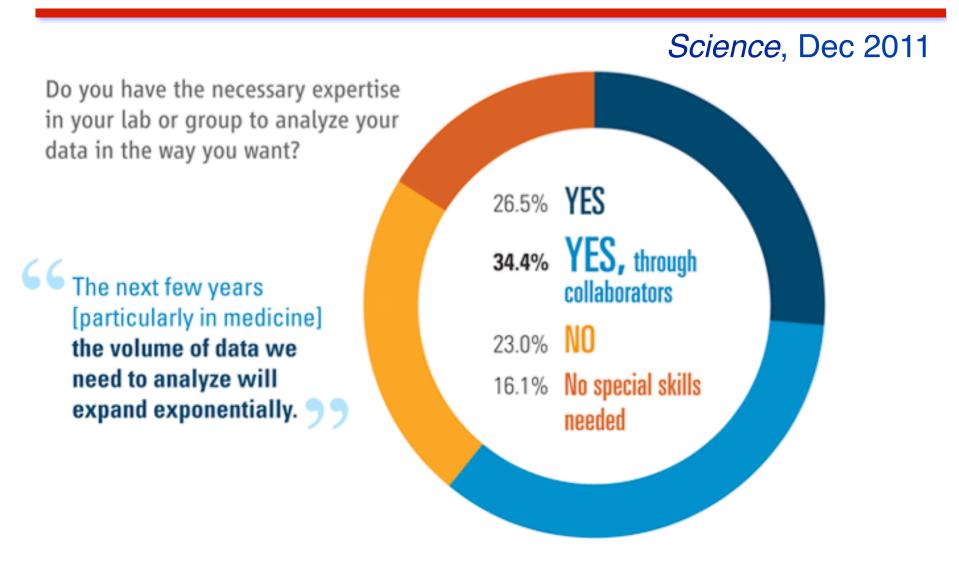
Benefits of Semantic Workflows:

1) Automatic Workflow Elaboration [Gil et al WORKS'13]

Workflowsdeveloped with Y. Liu (USC) and C. Mattmann (JPL) InputDataset Words LDA **Online LDA StopWords SmallWords FormatDataset Parallel** OutputData LDA **TrainTopics**

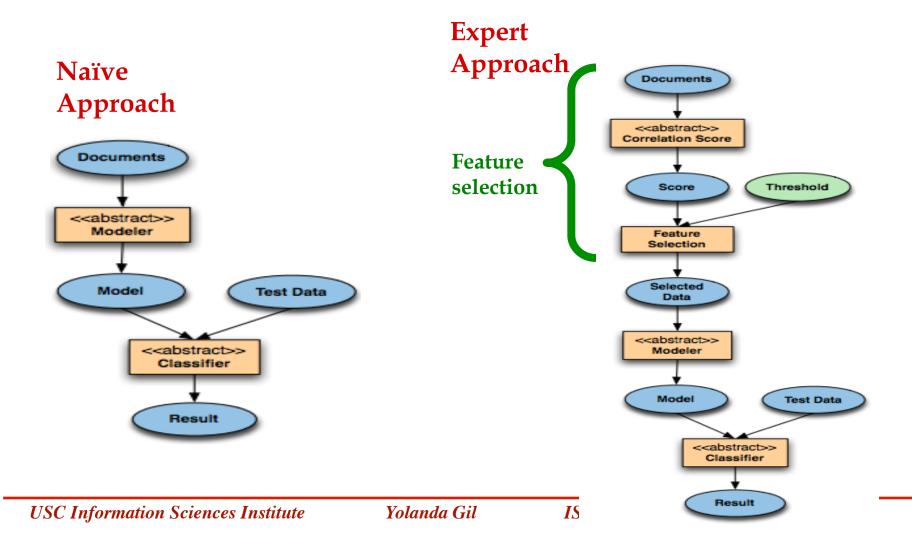
OutputTopics

Benefits of Semantic Workflows: 2) Access to Data Analytics Expertise



Capturing Expertise through Workflows [Hauder et al e-Science 2011]

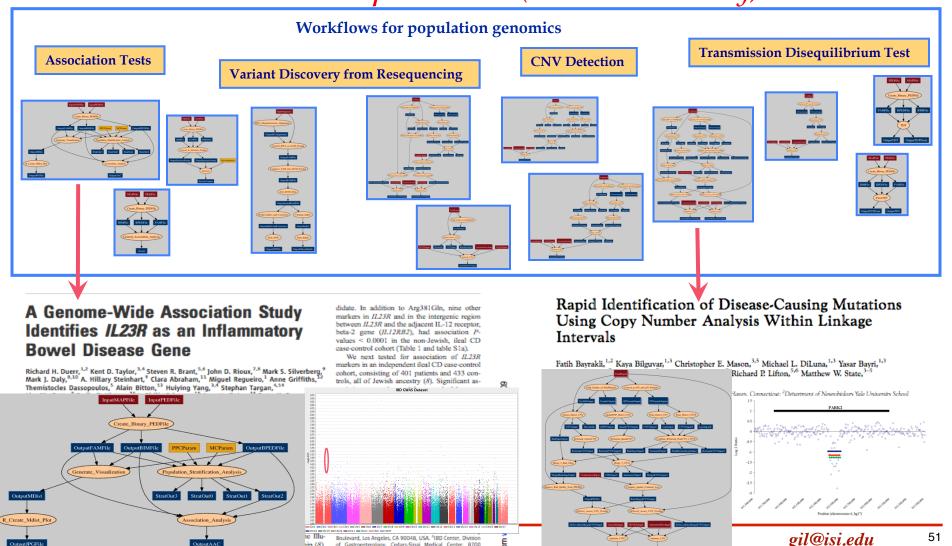
Workflows for text analytics, joint work with Yan Liu (USC) and Mattheus Hauder (TUM)



50

Capturing Expertise [Gil et al 2012]

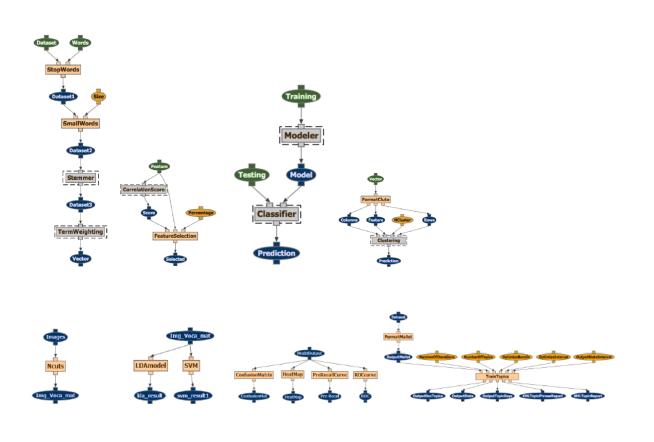
Work with Christopher Mason (Cornell University)

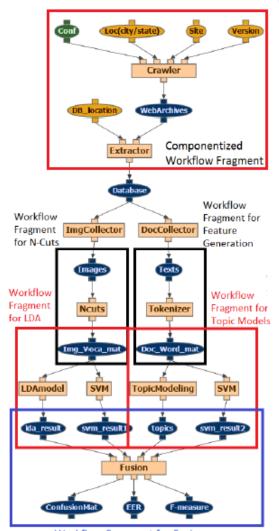


Beverly Boulevard, Los Angeles, CA 90048, USA, 5 Harvey M.

Benefits of Semantic Workflows: 3) Saving Time Through Reuse [Sethi et al MM'13]

Work with Ricky Sethi and Hyujoon Jo of USC



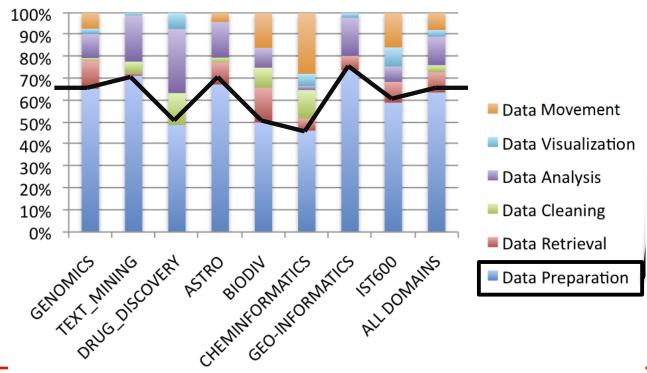


Workflow Fragment for Fusion

Saving Time through Reuse [Garijo et al FGCS'13]

Work with D. Garijo and O. Corcho (UPM), P. Alper, K. Belhajjame, and C. Goble (UM)

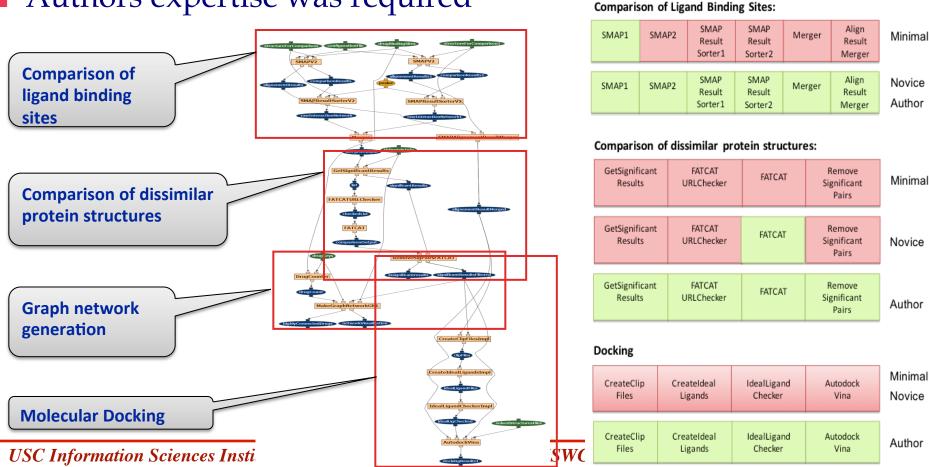
 "Scientists and engineers spend more than 60% of their time just preparing the data for model input or data-model comparison" (NASA A40)



Measuring Time Savings with "Reproducibility Maps" [Garijo et al PLOS CB12]

Work with D. Garijo of UPM and P. Bourne of UCSD

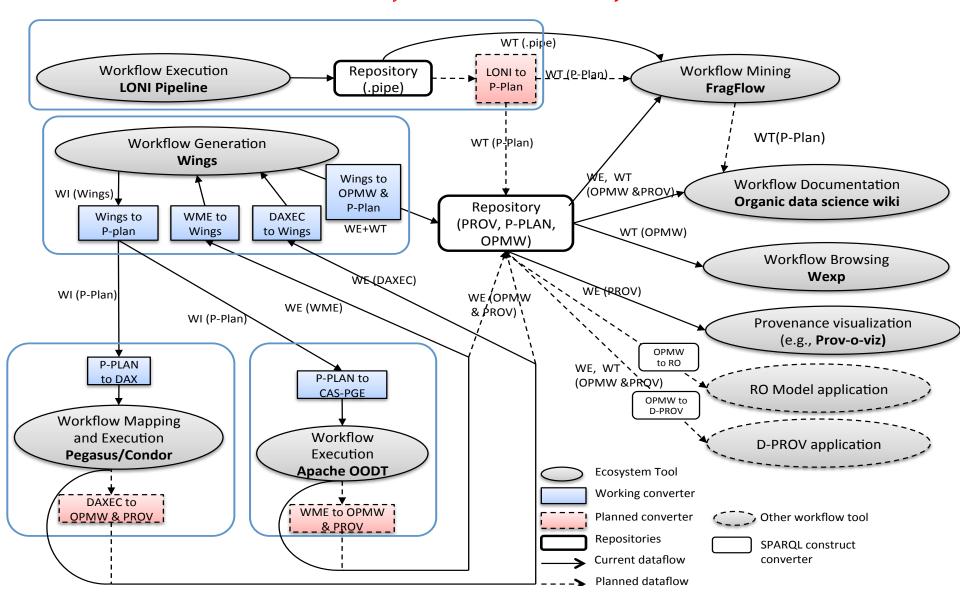
- 2 months of effort in reproducing published method (in PLoS' 10)
- Authors expertise was required



Benefits of Semantic Workflows:

4) Interoperability in a Workflow Ecosystem [Garijo et al 2014]

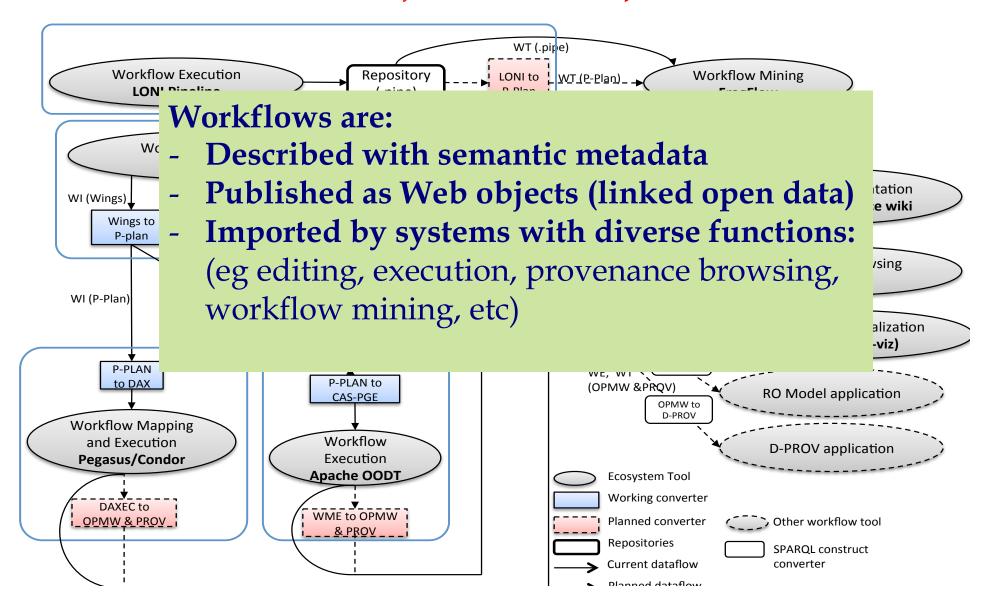
Work with D. Garijo and O. Corcho of UPM



Benefits of Semantic Workflows:

4) Interoperability in a Workflow Ecosystem [Garijo et al 2014]

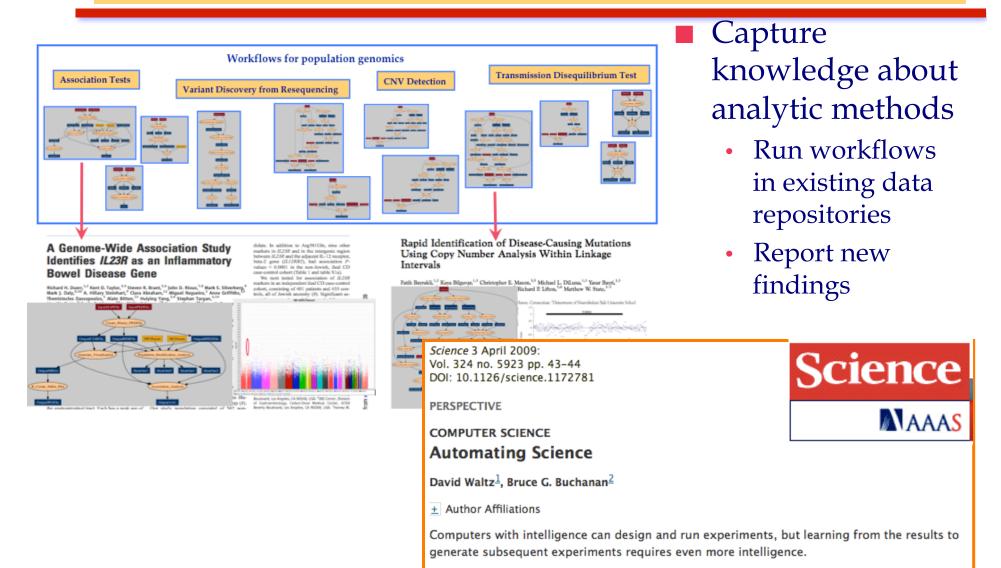
Work with D. Garijo and O. Corcho of UPM



Some Readings

- Yolanda Gil: "Intelligent Workflow Systems and Provenance-Aware Software." Proceedings of the Seventh International Congress on Environmental Modeling and Software (iEMSs), San Diego, CA, 2014.
- Yolanda Gil: "From Data to Knowledge to Discoveries: Artificial Intelligence and Scientific Workflows." Scientific Programming 17(3), 2009.
- Ewa Deelman, Chris Duffy, Yolanda Gil, Suresh Marru, Marlon Pierce, and Gerry Wiener: "EarthCube Report on a Workflows Roadmap for the Geosciences." National Science Foundation, Arlington, VA. 2012.

A Semantic Challenge: Automatic Paper Generator



A Semantic Challenge: A Web of Semantic Workflows/Processes

Assist people to:

- Share
- Copy
- Reuse
- Adapt
- Remix
- Update
- Certify
- Review
- ...

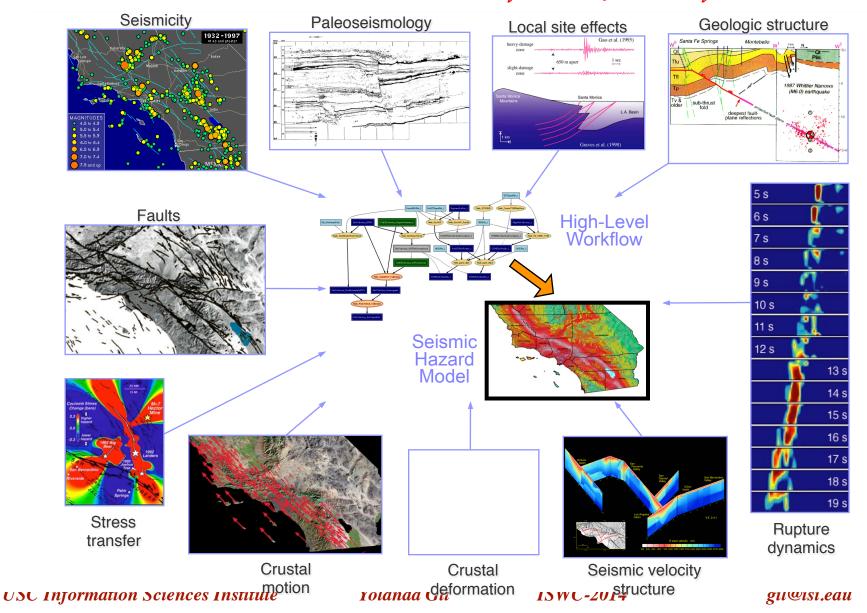
"Pay-as-you-go" semantics

Semantic Challenges in Getting Work Done

- To dos
 - Managing personal to dos
 - Managing coordinated to dos
- Knowledge rich tasks in science
 - Automatic paper generator
 - A Web of semantic workflows/processes
- Open science

Collaboration to Develop Workflows [Gil et al 2007]

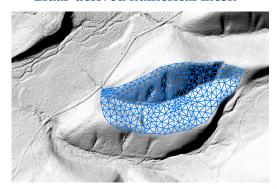
Slide from T. Jordan of USC and SCEC



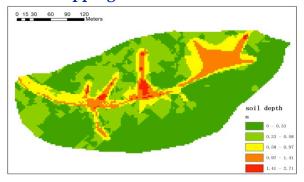
Understanding the "Age of Water"

Work with P. Hanson (UWisc), C. Duffy (PSU), and J. Read (USGS)

Lidar-derived numerical mesh



Mapping Bedrock GP Radar

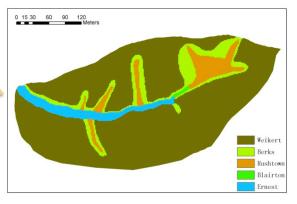


Linking Catchment Model-Data Assets Supported by NSF GEO-CZO

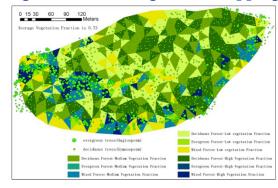


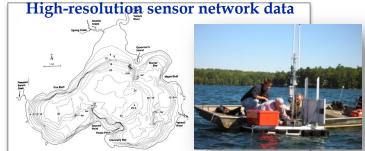
Linking Lake Model-Data
Assets Supported by
NSF BIO-GLEON, USGS CIDA

Research Soil Survey



High Resolution Vegetation Mapping





Read JS, et al. 2014. Ecological Modelling. 291C: 142-150. doi:10.1016/j.ecolmodel.2014.07.029

A New Kind of Collaborative Platform

Taxonomy of Science Communities [Bos et al 2007]

Shared Instruments	NEON
Community Data Systems	PDB
Open Community Contribution Systems	Zooniverse
Virtual Communities of Practice	GLEON
Virtual Learning Communities	VIVO
Distributed Research Centers	ENCODE
Community Infrastructure Projects	CSDMS

- Need a platform to support science collaborations that require:
 - Significant organization and coordination
 - Maintaining a community over the longer term
 - Growing the community based on unanticipated needs



Organic Data Science

Work with F. Michel and M. Hauder of TUM

- Organic data science is a novel approach to on-line scientific collaboration that supports:
 - Self-organization of communities by enabling any user to specify and decompose tasks
 - On-line community support by incorporating social sciences principles and best practices
 - An open science process by capturing new kinds of metadata about the collaboration that give necessary context to newcomers

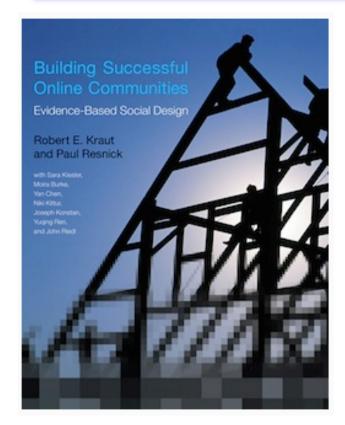
Self-Organization through Task Decomposition

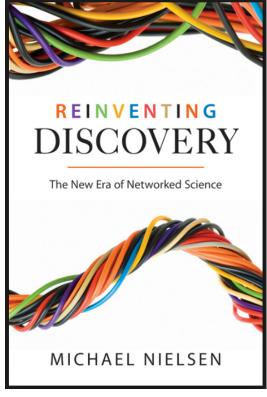
Implement the lake model for North Temperate Lakes

Document the GLM Lake Model Set up the GLM lake model for North Temperate Lakes Gather bathymetric and elevation data for North Temperate Lakes Obtain meteorological data for North Temperate Lakes Obtain initial conditions data for North Temperate lakes Obtain initial water level Obtain initial water temperature data Many tasks involved Obtain initial nutrient ar Necessary data resides in different repositories Obtain initial phytoplanl Different people understand different kinds of data Gather stream flow data for Where it is Collect daily volumetric How to use it If other data needed, unclear who has it Obtain water temperatu

Obtain or estimate nutrient and carbon concentration data for each major inflow

Social Principles for Online Communities















Social Principles: Some Examples

■ Starting communities, e.g.:

- Organize content, people, and activities into subspaces
- Inactive tasks should have "expected active times"

■ Encouraging contributions, e.g.:

- Simple tasks with challenging goals are easier to comply with
- Publicize that others have complied with requests

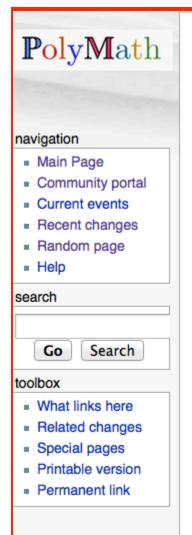
■ Encouraging commitment, e.g.:

Interdependent tasks increase commitments and reduce conflict

■ Dealing with newcomers, e.g.:

- Design common learning experiences for newcomers
- Provide sandboxes while they are learning

Opening Science: Polymath [Nielsen, Gowers 09]



Main Page

This is the wiki for *polymath* projects - massively collaborative online mathemat massively collaborative mathematics possible?

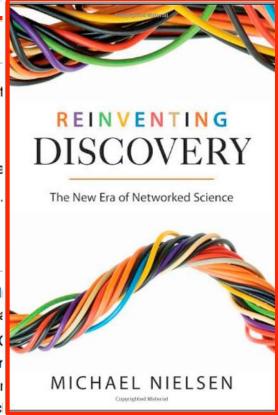
Many polymath projects will be proposed, planned, and run at This Blog ...

A Polymath logo is being trialled. If you have more suggestions, please add the

The wiki is currently locked down due to a major influx of spam (July 29, 2013). my best to reply quickly.

Existing polymath projects

- Polymath1: New proofs and bounds for the density Hales-Jewett theorem. I
- Polymath2: Must an "explicitly defined" Banach space contain c_0 or l_p ? Initial
- Mini-polymath1: Solving Problem 6 of the 2009 International Mathematical (
- Polymath3. The polynomial Hirsch conjecture. Proposed July 17, 2009; laur
- Polymath4: A deterministic way to find primes. Proposed July 27, 2009; laur
- Polymath5: The Erdős discrepancy problem. Proposed Jan 10, 2010; launc
- Mini-polymath2: Solving Problem 5 the 2010 International Mathematical Olympiad. Propose
- Polymath6: Improving the bounds for Roth's theorem. Proposed Feb 5, 2011.
- Mini-polymath3: Solving a problem from the 2011 International Ma
- Mini-polymath4: Solving a problem from the 2012 International Ma
- Polymath7: Establishing the Hot Spots conjecture for acute-angle
- Polymath8: Improving the bounds for small gaps between primes.



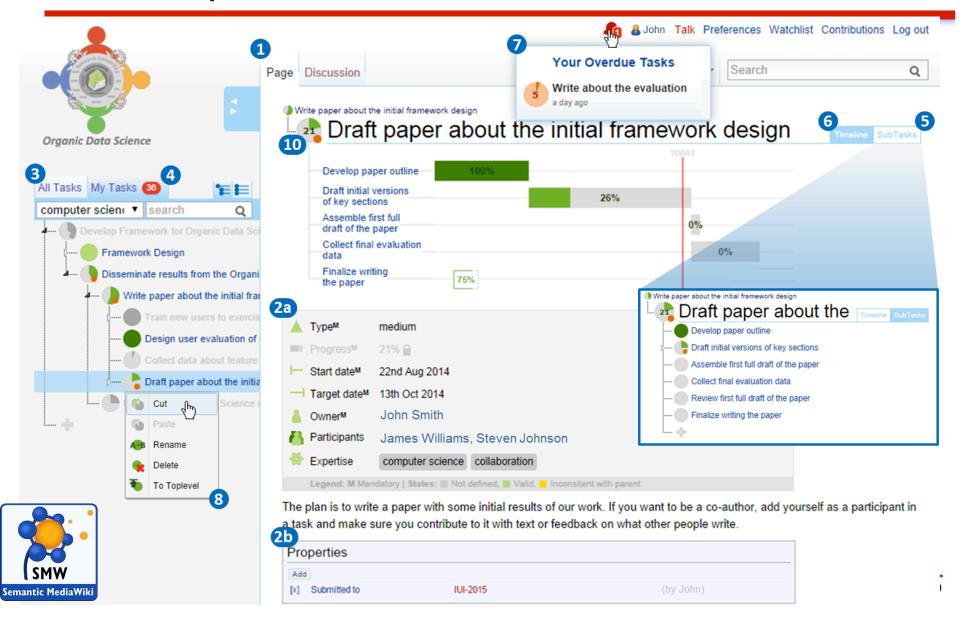
Organic Data Science



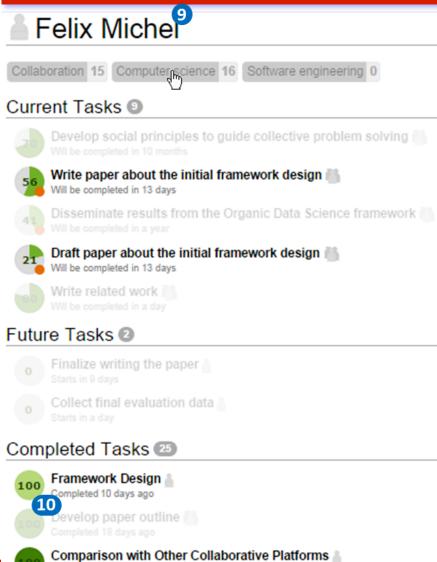
=> Task-oriented self-organizing on-line communities for open collaboration in science

- Organic data science is a novel approach to on-line scientific collaboration that supports:
 - Self-organization of communities by enabling any user to specify and decompose tasks
 - On-line community support by incorporating social sciences principles and best practices
 - An open science process by capturing new kinds of metadata about the collaboration that give necessary context to newcomers

Self-Organization through Dynamic Task Decomposition



Organic Data Science: Contributors



CDEC WEATHER 2010 03 02

Data

Data

- DOWNLOAD
- Data Types
 - Daily Sensor Data

Structured Properties				
Add				
[x]	Barpress	760	(By Admin)	
[x]	Depth	1.0214570760727	(By Admin)	
[x]	Flow	1550.6185302734	(By Admin)	
[x]	ForSite	SMN	(By Admin)	
[x]	HasSize	8316	(By Admin)	
[x]	SiteLatitude	37.347213745117	(By Admin)	
[x]	SiteLongitude	-120.97618103027	(By Admin)	
[x]	Slope	0.000099999997473788	(By Admin)	
[x]	Velocity	0.65311223268509	(By Admin)	

- . Used as Input in the following Workflows:
 - AF NTM Execution 2 March 2012 to 8 March 2012
 - AF EDM Execution 2 March 2012 to 8 March 2012
 - AF EM Execution 2 March 2012 to 8 March 2012

Models

AQUAFLOW NTM NIGHTTIMEMODEL

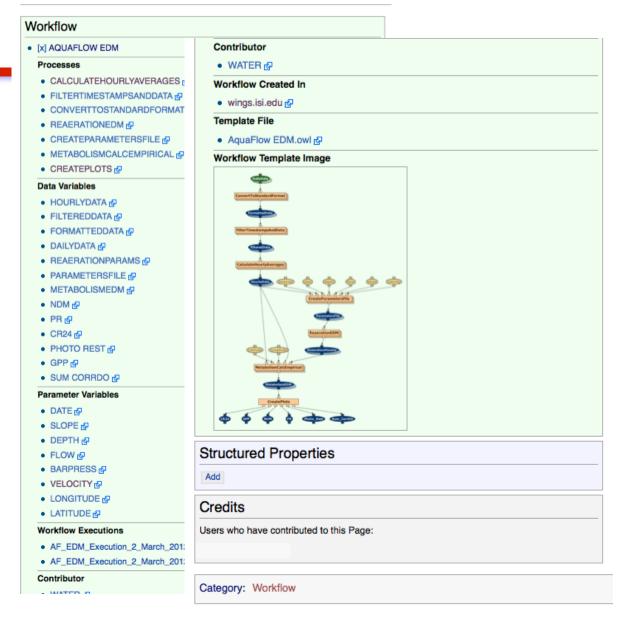
Component

- Input
 - PrevDayHourlyData
 - InputHourlyData
- Param
 - Longitude
 - Latitude
- Output
 - MeanNightDO
 - OutputNTMParams
- . Used in the following Templates:
 - AQUAFLOW NTM

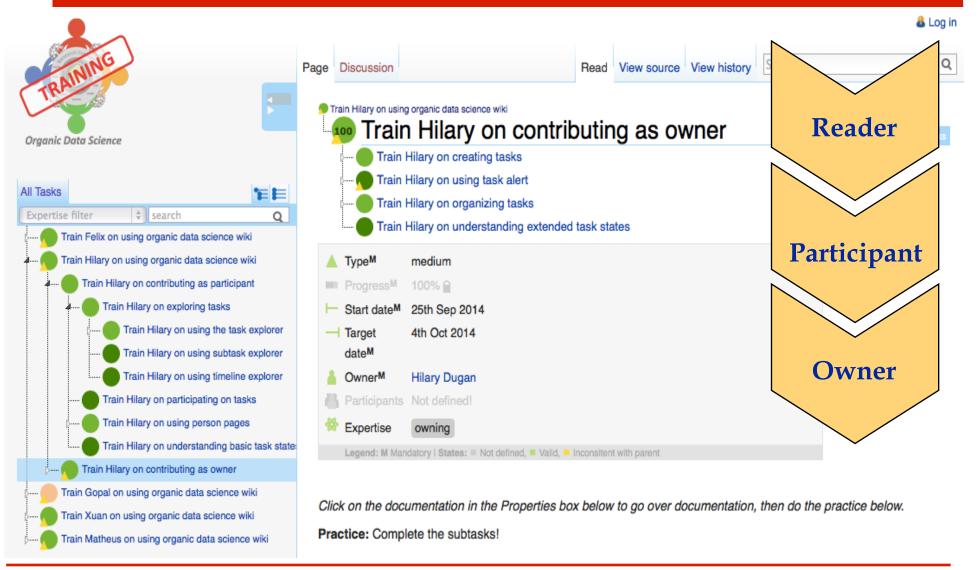
Structured Properties			
Add			
[x]	InputDataVariable	PrevDayHourlyData	(By Anonymous)
[x]	InputDataVariable	InputHourlyData	(By Anonymous)
[x]	InputParamVariable	Longitude	(By Anonymous)
[x]	InputParamVariable	Latitude	(By Anonymous)
[x]	OutputDataVariable	MeanNightDO	(By Anonymous)
[x]	OutputDataVariable	OutputNTMParams	(By Anonymous)

AQUAFLOW EDM

Workflows



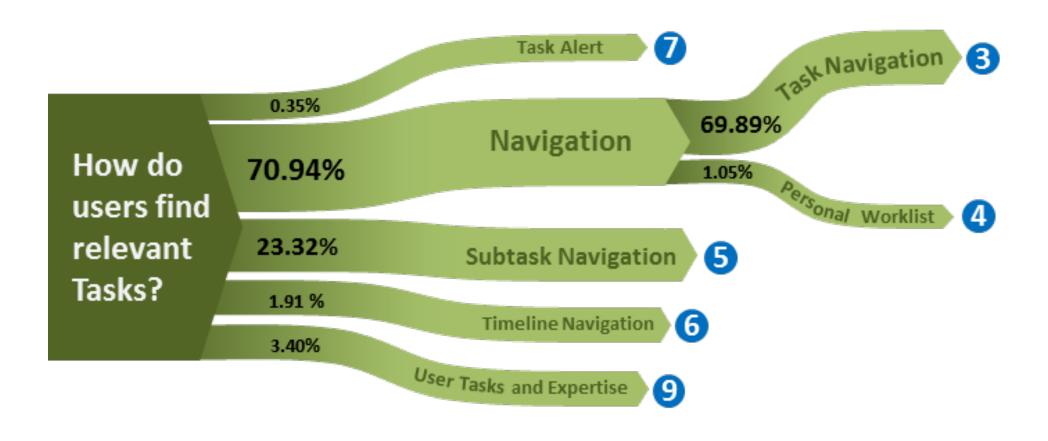
Training Newcomers



What Features Are Used to Manage Tasks?



How Do Users Find Relevant Tasks?



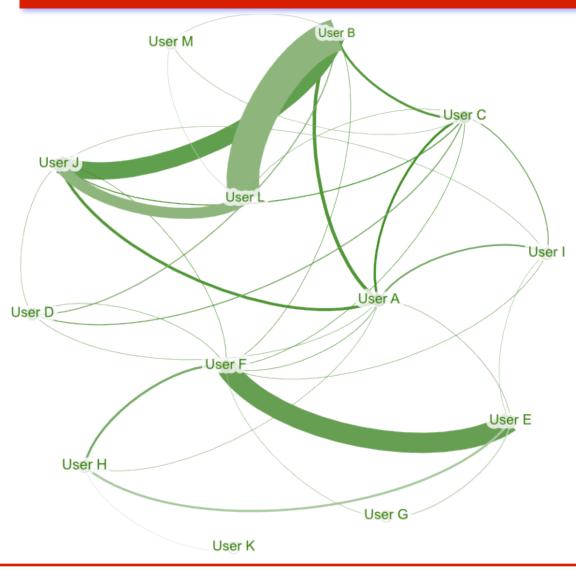
Are Users 100 2% Number of 25T 14T 10% 13% 14% 31T Collaborating? Different 16% 25T 3% Persons 8T 80 Tasks 14% 32T 70 **=** >4 32% 36T Percentage of 60 **4** 21% 471 50 **3** 88% 210T 81% 126T 40 30% 2 33T 30 **1** 48% 111T 20 24% 27T T = Number10 of Tasks

- A 52% of tasks are viewed by more than one person
- 1 72% of tasks have more than one person signed up
- C 19% of tasks have more than one person editing metadata

ß

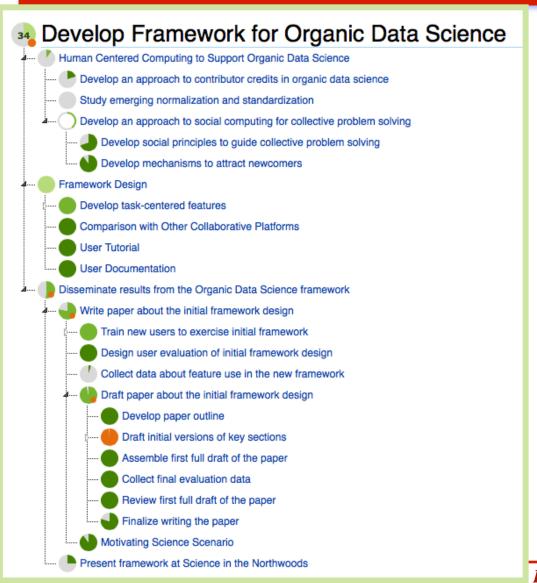
11% of tasks have more than one person editing content

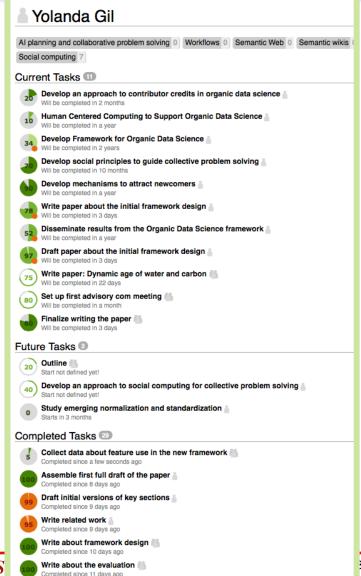
What Does the Social Network of Collaborators Look Like?



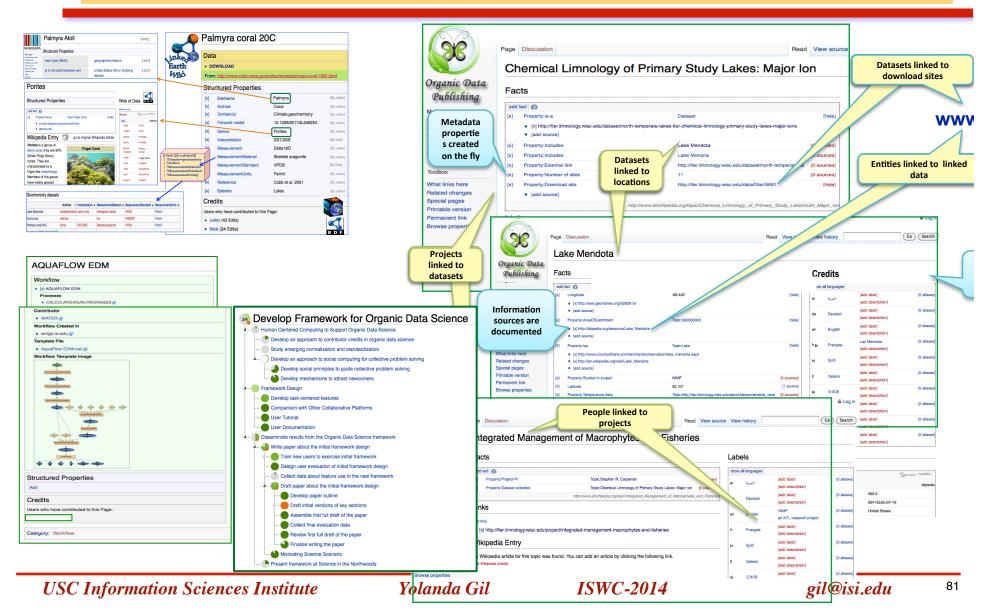
- Network of users (nodes) linked by shared tasks
- Links across all users
- Two distinct subgroups
 - 1. Water
 - 2. Software

A Semantic Challenge: Email-less Coordination for Projects





A Semantic Challenge: Open Science Processes



Semantic Challenges in Getting Work Done

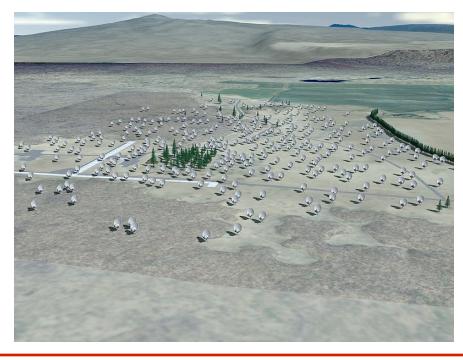
- To dos
 - Managing personal to dos
 - Managing coordinated to dos
- Knowledge rich tasks in science
 - Automatic paper generator
 - A Web of semantic workflows/processes
- Open science
 - Email-less coordination of projects
 - Open science processes

http://www.isi.edu/~gil
http://www.wings-workflows.org
http://www.organicdatascience.org
http://discoveryinformaticsinitiative.org

"We need bigger glasses and more hands in the water" – J. Tarter, SETI Institute











Discovery Informatics: Knowledge-Rich Science Infrastructure

SCIENCE sciencemag.org

10 OCTOBER 2014 • VOL 346 ISSUE 6206

ARTIFICIAL INTELLIGENCE

Amplify scientific discovery with artificial intelligence

Many human activities are a bottleneck in progress

By Yolanda Gil,¹ Mark Greaves,² James Hendler,^{3*} Haym Hirsh⁴

echnological innovations are penetrating all areas of science, making predominantly human activities a principal bottleneck in scientific progress while also making scientific advancement more subject to error and harder to reproduce. This is an area where a new generation of artificial intelligence (AI) systems can radically transform the prac-

increased the numbers of interested participants; Moore's law and steady exponential increases in computing power; and exponential increases in, and broad availability of, relevant data in volumes never previously seen. Those scientific efforts that have leveraged AI advances have largely harnessed sophisticated machine-learning techniques to create correlative predictions from large sets of "big data." Such work aligns well with the current needs of peta- and exascale science. However, AI has far broader capacity to ac-

information-finding beyond current search limitations.

We can project a not-so-distant future where "intelligent science assistant" programs identify and summarize relevant research described across the worldwide multilingual spectrum of blogs, preprint archives, and discussion forums; find or generate new hypotheses that might confirm or conflict with ongoing work; and even rerun old analyses when a new computational method becomes available. Aided by such a system, the scientist will focus on more of the creative aspects of research, with a larger fraction of the routine work left to the artificially intelligent assistant.

"AI-based systems that can represent hypotheses ... can reduce the error-prone human bottleneck in ... discovery."

Discovery Informatics Science Challenges for Intelligent Systems http://www.discoveryinformaticsinitiative.org

PSB Workshop (January 2015)

KDD Workshop (August 2014): http://ailab.ist.psu.edu/idkdd14/

AAAI Workshop (July 2014):

http://discoveryinformaticsinitiative/diw2014

AAAI Fall Symposium (Nov 2013):

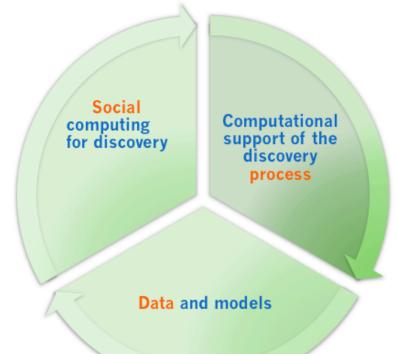
http://discoveryinformaticsinitiative/dis2013

AAAI Fall Symposium (Nov 2012):

http://discoveryinformaticsinitiative/dis2012

Microsoft eScience Summit (Aug 2012) Workshop on Web Observatories for Discovery Informatics

> PSB Workshop (Jan 2013): on Computational Challenges of Mass Phenotyping





NSF Workshop (Feb 2012): http://discoveryinformaticsinitiative/diw2012



A View from Biomedical Research: The NIH Big Data To Knowledge (BD2K) Initiative

PEBOURNE

Professional Developments Worth Sharing

HOME

ABOUT



Taking on the Role of Associate Director for Data Science at the NIH – My Original Vision Statement

by pebourne

On March 3, 2014 I will begin the job of Associate Director for Data Science (ADDS) at the National Institute of Health (NIH). I will report directly to NIH Director, Dr. Francis Collins. When I originally applied for the position in April 2013 I was asked to prepare a short vision statement. That statement

"Discovery informatics is in its infancy. Search engines are grappling with the need for deep search, but it is doubtful they will fulfill the needs of the biomedical research community when it comes to finding and analyzing the appropriate datasets. Let me cast the vision in a use case. As a research group winds down for the day algorithms take over, deciphering from the days on-line raw data, lab notes, grant drafts etc. underlying themes that are being explored by the laboratory (the lab's digital assets). Those themes are the seeds of deep search to discover what is relevant to the lab that has appeared since a search was last conducted in published papers, public data sets, blogs, open reviews etc. Next morning the results of the deep search are presented to each member as a personalized view for further post processing. We have a long way to go here, but programs that incite groups of computer, domain and social scientists to work on these needs will move us forward."

A View from Geoscieces: The NSF EarthCube Initiative

Outcomes

Transform practices within the geosciences community spanning over the next decade

Provide unprecedented new capabilities to researchers and educators

Vastly improve the productivity of community

Accelerate research on the Earth system

Provide a knowledge management framework for the geosciences













Data

Workflows Semantics Governance

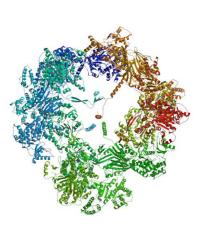
http://www.earthcube.org/

What Might the Future Look Like?

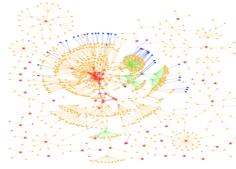
YOU: What are you working on?

OTHER PERSON: I am really busy, working on...









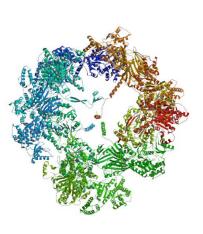


In the Future

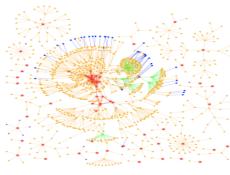
YOU: What are you working on?

OTHER PERSON: I am really busy, working on...











YOU: Yes, but aren't you glad that we can get our work done faster?

Thank you!













http://www.isi.edu/~gil

http://www.wings-workflows.org

http://www.organicdatascience.org

http://discoveryinformaticsinitiative.org

- Wings contributors: Varun Ratnakar, Ricky Sethi, Hyunjoon Jo, Jihie Kim, Yan Liu, Dave Kale (USC), Ralph Bergmann (U Trier), William Cheung (HKBU), Daniel Garijo (UPM), Pedro Gonzalez & Gonzalo Castro (UCM), Paul Groth (VUA)
- Wings collaborators: Chris Mattmann (JPL), Paul Ramirez (JPL), Dan Crichton (JPL), Rishi Verma (JPL), Ewa Deelman & Gaurang Mehta & Karan Vahi (USC), Sofus Macskassy (ISI), Natalia Villanueva & Ari Kassin (UTEP)
- Organic Data Science: Felix Michel and Matheus Hauder (TUM), Varun Ratnakar (ISI), Chris Duffy (PSU), Paul Hanson, Hilary Dugan, Craig Snortheim (U Wisconsin), Jordan Read (USGS)
- Biomedical workflows: Phil Bourne & Sarah Kinnings (UCSD), Chris Mason (Cornell), Joel Saltz & Tahsin Kurk (Emory U.), Jill Mesirov & Michael Reich (Broad), Randall Wetzel (CHLA), Shannon McWeeney & Christina Zhang (OHSU)
- Geosciences workflows: Chris Duffy (PSU), Paul Hanson (U Wisconsin), Tom Harmon & Sandra Villamizar (U Merced), Tom Jordan & Phil Maechlin (USC), Kim Olsen (SDSU)
- *And many others!*