### The DAE Platform

# Document Mining, Interpretation and Semantics Through the Prism of Performance Evaluation

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In collaboration with Lehigh University





# Speaker Background

Assistant Professor @ INPL – France

http://www.inpl-nancy.fr

•Research @ LORIA – INRIA

http://www.loria.fr

- Visiting Scientist @ Lehigh (2010-2011)
- •Contact Information:

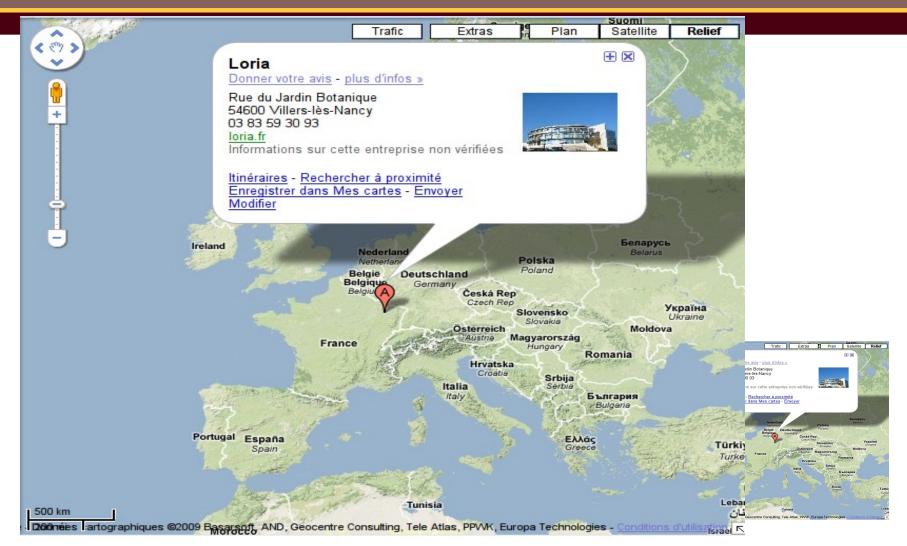
Bart.Lamiroy@loria.fr







# Speaker Background





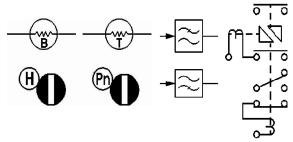


# Scientific Background

•1998: Image Indexing (Ph.D.)

•1998-2000: Stereo Tracking and Visual Servoing

•2000-2009: Graphical Document Analysis & "Image Semantics"



•2010-2011: The DAE project (http://dae.cse.lehigh.edu)











# Acknowledgements

Bryan Hodgson

Xingjian Zhang

Jeanne Steinberg

Yingjie Li

Yang Yu

Daniel Lopresti

Henry Baird

Weidong Chen

Dezhao Song

Qihan Long

Tao Sun

Sameer Antani

Mike Kot

Jeff Heflin

Hank Korth

**Brian Davison** 

Sai Lu Mon Aung

Austin Borden

Mike Caffrey







### **Outline of Talk**

#### General Context

Challenges of reproducible and traceable experimental research

#### The DAE Platform

A way of addressing the challenges, technical architecture and design choices

### First Experiments

Reports of actual uses of the platform

### Looking Beyond and New Challenges





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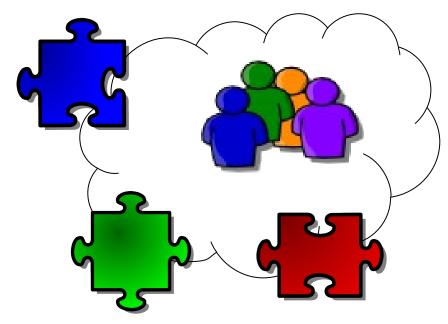
### Looking Beyond and New Challenges





### **General Context**

New idea – Claim – Algorithm



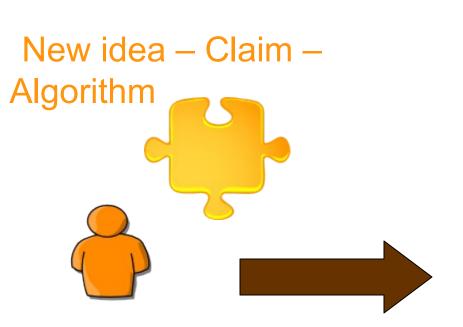
Existing algorithms – Stateof-the-Art – Other claims



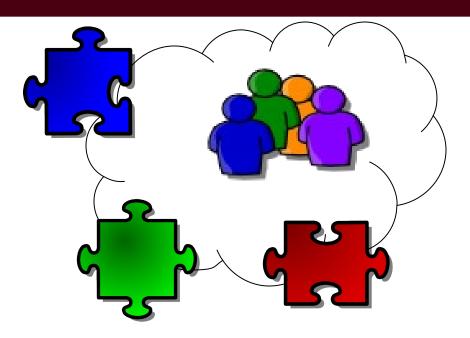




# Reproducible and Traceable Experimental Research



How do I perform?
How do I compare to the State-of-the-Art?

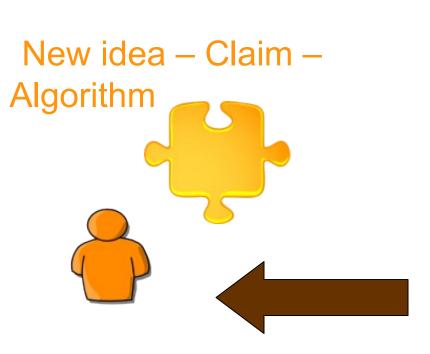


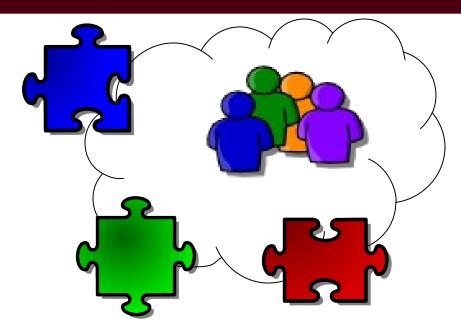
Existing algorithms – Stateof-the-Art – Other claims





# Reproducible and Traceable Experimental Research





Existing algorithms – Stateof-the-Art – Other claims

What is the legitimacy of this claim? Can it be verified/reproduced?





### A Generic Scenario

### Jane, a young starting researcher

Goal: solve a specific knowledge extraction problem (examples follow)

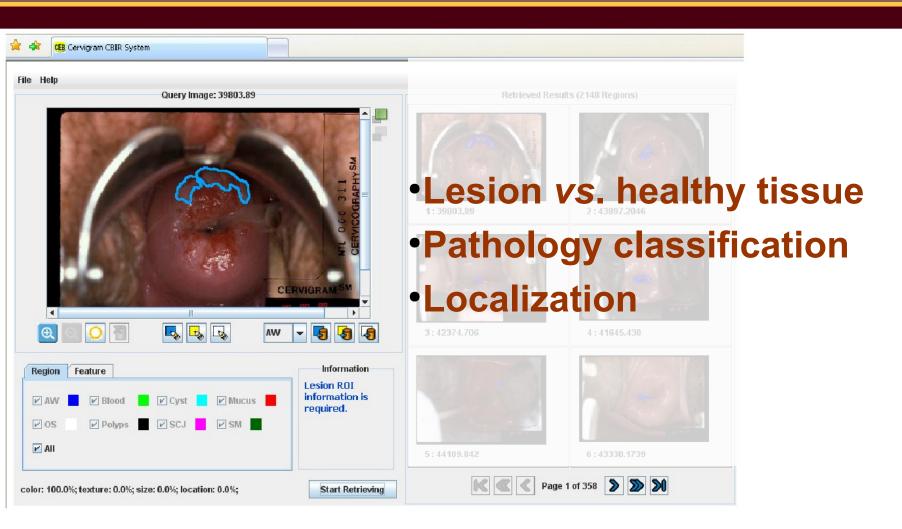
### Requirements:

- Find appropriate experimental data
- Prove genericity of approach (or establish boundaries)
- Compare with state of the art
- Report results





# Example 1







# Example 2

Graefe's Arch Clin Exp Ophthalmol (2002) 240:101-105

LABORATORY INVESTIGATION

DOI 10.1007/s00417-001-0389-z

Toshiya Sakurai Eiichi Yukawa Yoshiaki Hara Norio Mivata Manabu Mochizuki Indocyanine green angiography in patients with human T cell-lymphotropic virus type 1 uveitis

Received: 26 March 2001 Revised: 12 September 2001 Accepted: 12 September 2001 Published online: 29 January 2002 © Springer-Verlag 2002

This study was presented in part at the 4th International Symposium on ICG Angiography, Baden-Baden, Germany,

This study was approved by the ethics committee of Tane Memorial Eve Hospital

Tane Memorial Eye Hospital. 1-1-39 Sakaigawa, Nishi-ku, Osaka, 550-0024, Japan Tel.: +81-6-65815800 Fax: +81-6-65815063

E. Yukawa · Y. Hara Department of Ophthalmology, Nara Medical University, Nara, Japan

Miyata Eye Hospital, Miyakonojo, Miyazaki, Japan

M. Mochizuki Tokyo Medical and Dental University, Tokyo, Japan

Abstract Purpose: To determine the indocyanine green (ICG) angiographic features and to evaluate the choroidal involvement of human T-cell lymphotropic virus type 1 (HTLV-1)-associated uveitis. Methods: We performed ICG angiography using scanning laser ophthalmoscopy in 54 eyes of 27 patients (8 men and 19 women) diag-

nosed with HTLV-1 uveitis. The patient's mean age was 51.5 years with a range of 24-65 years. Results: The early phase of ICG angiography revealed ICG leakage from the choroidal vessels in the posterior pole, hyperfluorescent spots that which were not detected with fluorescein angiography, and small hypofluorescent lesions in the macula which most likely corresponded to microcirculatory disturbances in the choriocapillaris. Conclusions: We suggest that the ICG angiographic findings reflect choroidal lesions such as infiltration with leukocytes and edema. ICG angiography may provide useful information on choroidopathy in HTLV-1 uveitis.

#### Introduction

of the world. HTLV-1 carriers have been detected in the of Kyushu island, Japan. HTLV-1 is not easily transmissible under normal conditions, because cell-to-cell con- ogy [1, 5, 6, 7, 8, 9, 11, 14, 18, 20]. tact is generally required. The virus is transmitted by child via breast-feeding (HTLV-1 antigen-positive lymson to person via sexual contact (semen) or blood transfusion [16]. The virus is known to cause adult T-cell leualso related to many other diseases [15].

In 1989 [10], Oba et al. reported a relationship between HTLV-1 and uveitis. Mochizuki et al. performed Human T cell-lymphotropic virus type I (HTLV-1) is a epidemiologic, clinical, and virologic studies to investihuman retrovirus that is highly endemic in some regions gate the correlations between HTLV-1 and uveitis and defined HTLV-1-associated uveitis. The diagnosis of Caribbean islands, parts of central Africa, and the south HTLV-1 uveitis is based on seropositivity to HTLV-1 and the exclusion of other uveitic entities with defined etiol-

The diagnostic criteria are idiopathic uveitis, HTLV-1 three major routes: vertical transmission from mother to seropositivity, clinical features such as membranous vitreous opacities, and the presence of HTLV-1 infected phocytes in milk), and horizontal transmission from per- cells in the aqueous humor. Fluorescein angiography provides useful information on the retinal vessels, but the choroidal vessels and choroidal circulation cannot be kemia (ATL) and HTLV-1-associated myelopathy well visualized by this means. Recently, indocyanine (HAM). Recent studies have suggested that HTLV-1 is green (ICG) angiography has been used to observe choroidal lesions [2, 3, 4, 19]. ICG dye limits leakage from

- Author Title **Abstract localization**
- Language Identification
- Topic Identification
- Reference extraction







### **Current Consensus**

### Peer Reviewed Papers

- Re-implementation issues
- Claim verification

### Code Repositories

- Technology dependency
- IP and Copyright

#### Reference Data Collections

- Maintenance cost
- Non-evolution over time static
- Context restriction

#### Contests

Snapshots











### Problems with Consensus

#### Goal

 Develop robust algorithms that approach human levels of performance for specific tasks of interest.

#### Observations

- We want general algorithms, but often they are tested on small, overused datasets removed from real world.
- Published experimental results may be biased by algorithm developer's intimate knowledge of the data.
- Current practices fail to provide evidence of generality.
- What does "human levels of performance" mean? Even experts can disagree on all but the most trivial of cases.





### Problems with Consensus

#### Goal

Invent new methods that improve on known techniques.

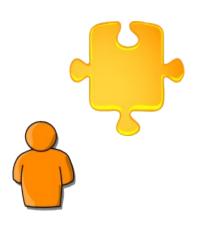
#### Observations

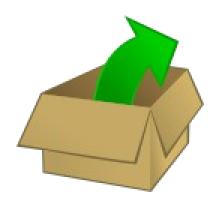
- How do we know when we have succeeded?
- Need to compare against previously published results creates reliance on standard datasets (a bad idea).
- Difficulties in re-implementing a published algorithm (incomplete descriptions, inherent conflict of interest).
- Competitions can be useful, but are snapshots.
- Many papers do not bother to make reproducible comparisons.

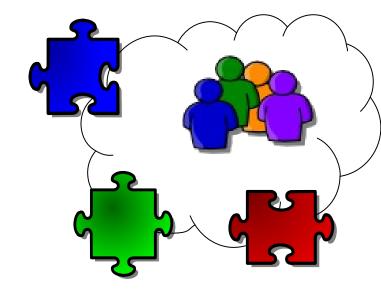




### **Needed Solution**







- Find appropriate testing data
- ► Find reference validation
- Certify experimental results
- Reproduce published results

- ► Provide reference data
- Provide access to algorithms





# Requirements for Solution

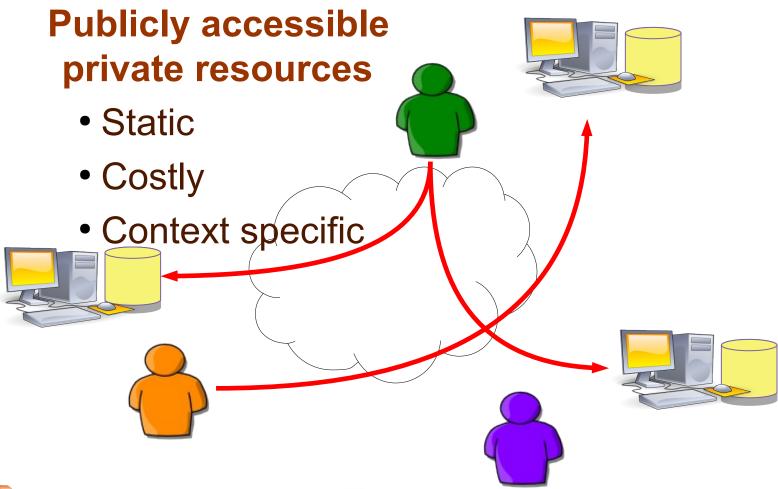
### Formalizing Experiments and their Environment

- Experimental DATA
  - Describe it
  - Make it available
  - Question dispute challenge extend
- Experiments
  - Describe them
  - Make them available
  - Reproduce them
  - Question adapt modify extend





### Before







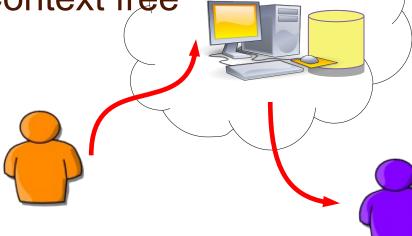
### DAE

# Publicly accessible shared resources



Shared cost









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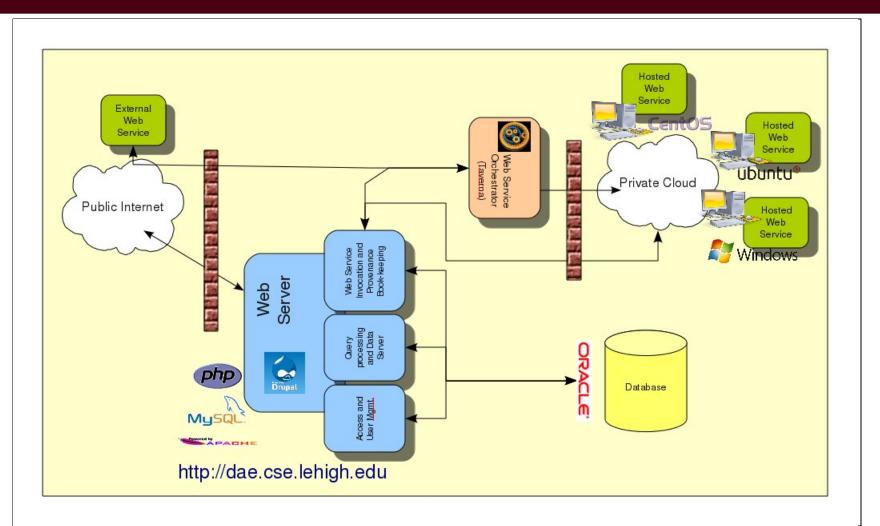
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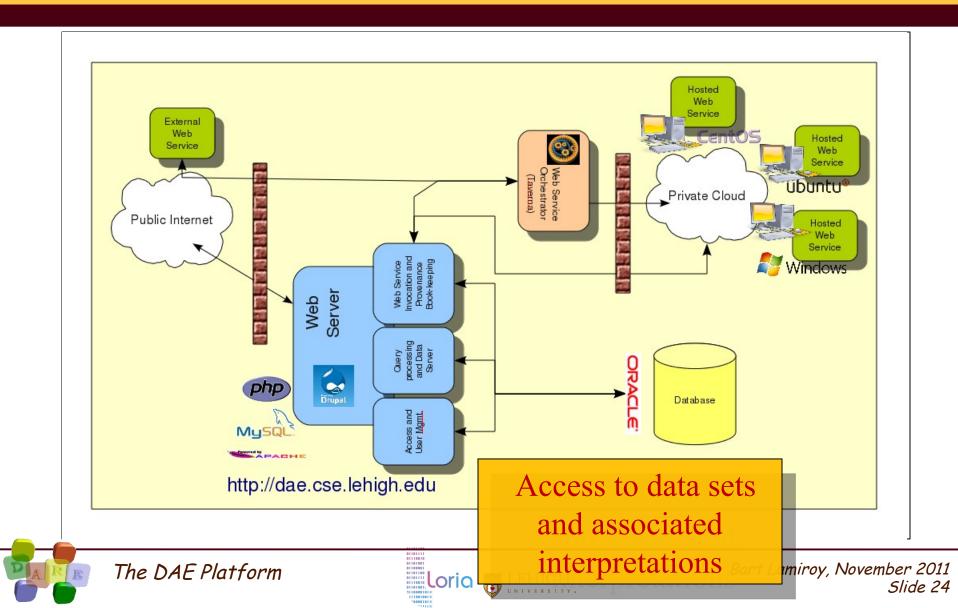
### DAE Platform - dae.cse.lehigh.edu



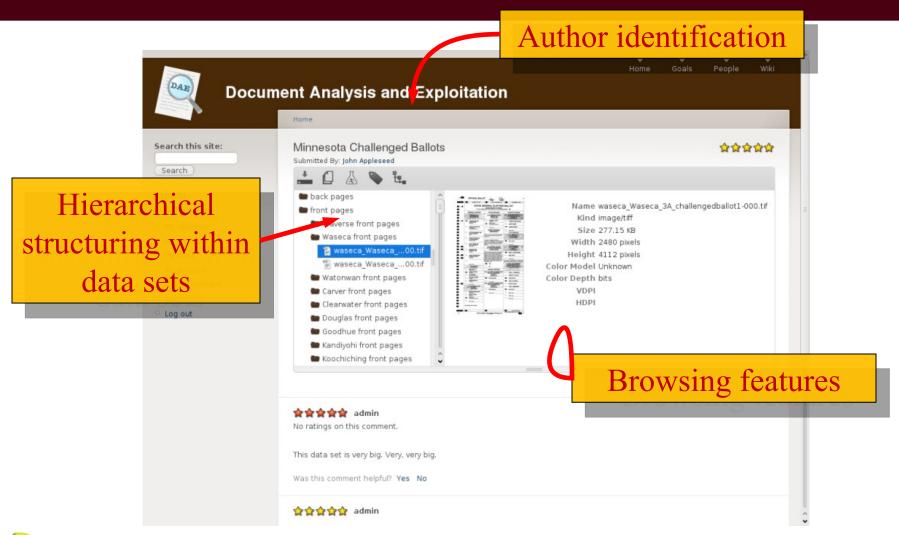




### DAE Platform - dae.cse.lehigh.edu

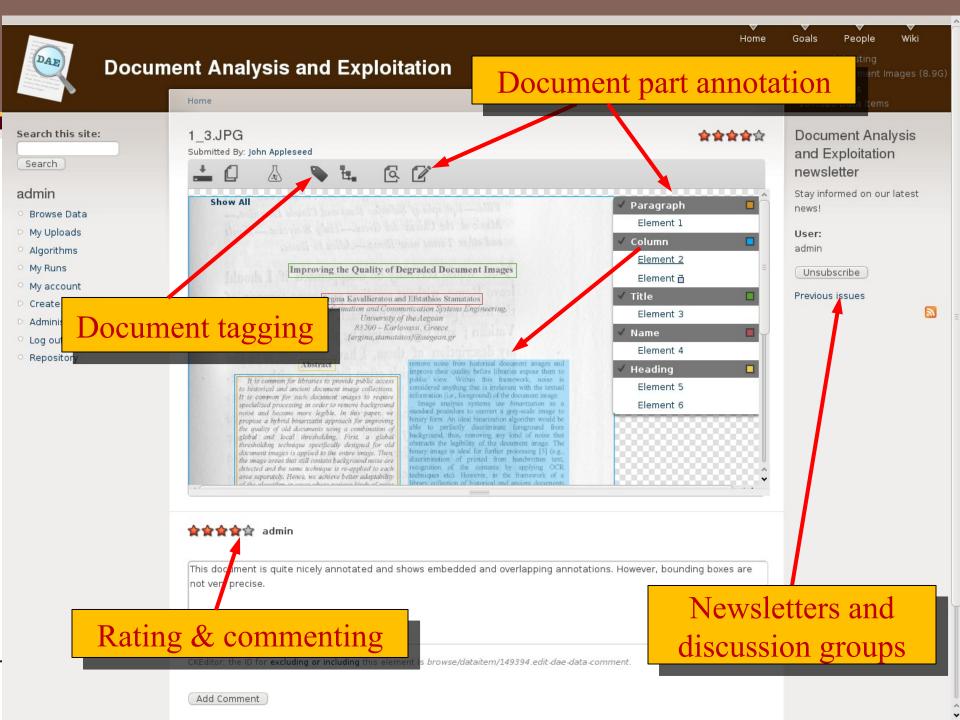


# **Browsing Data Sets**

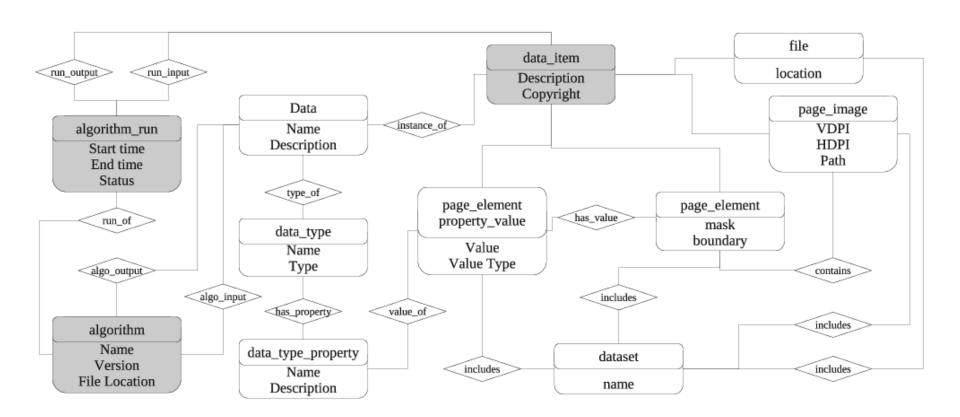








# **Underlying Data Model**

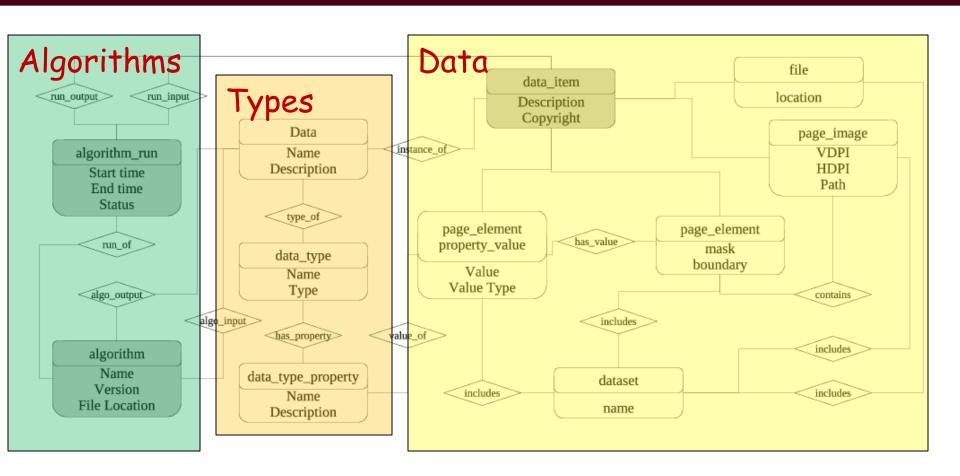








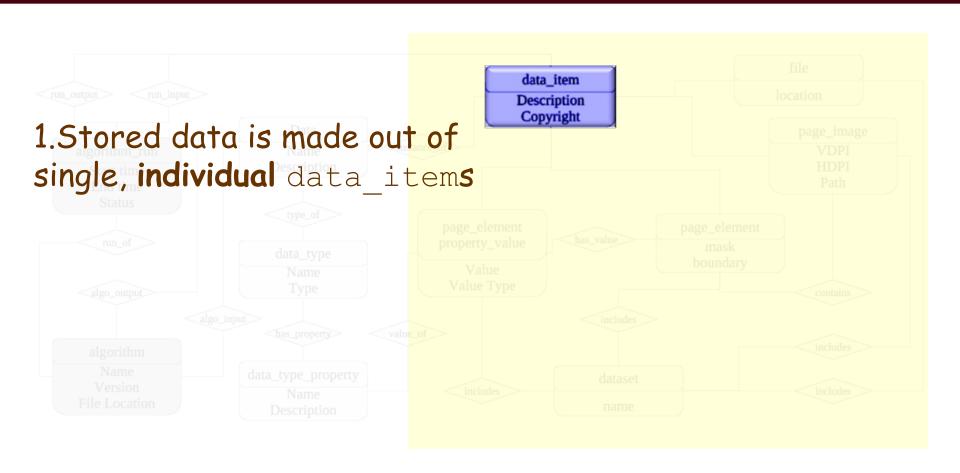
# **Underlying Data Model**







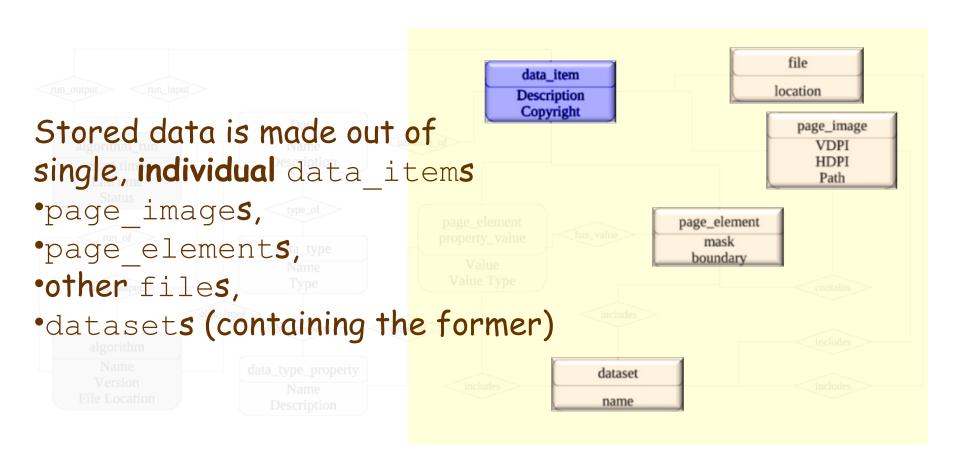
## Data Model: data\_items







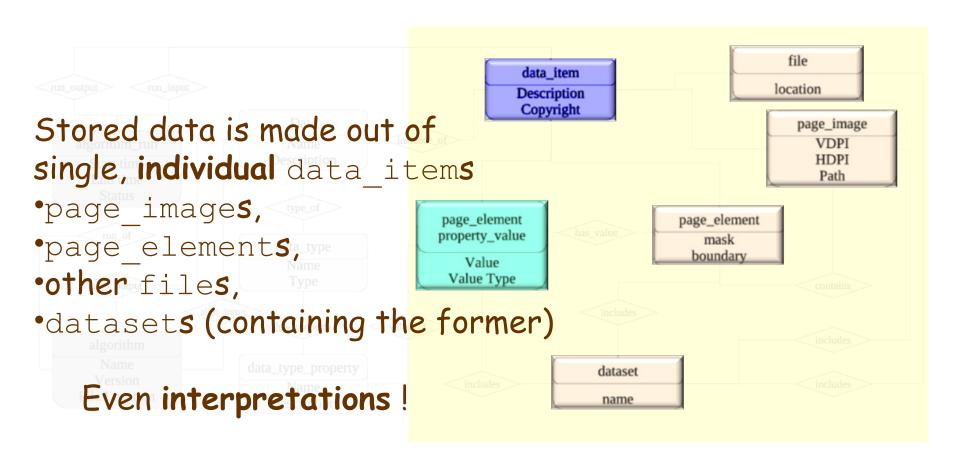
## Data Model: data\_items







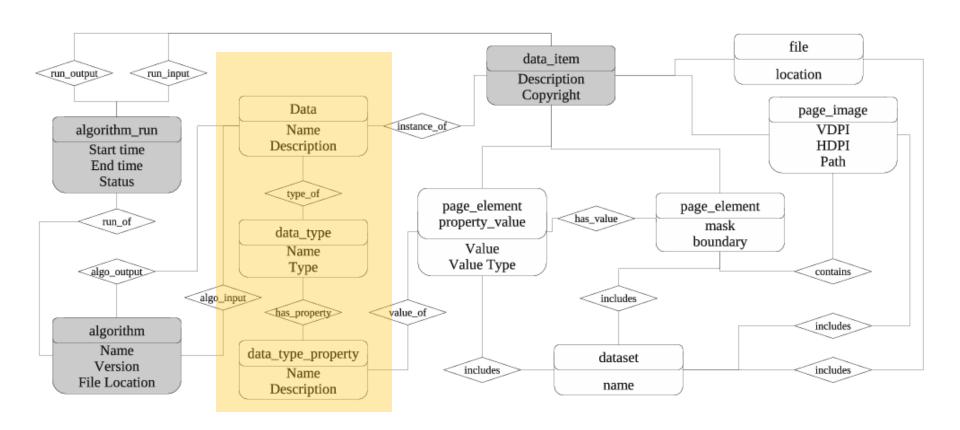
## Data Model: data\_items







# Data Model: datatypes



### 1. All data is typed





# Result: Fully Query-able Data

#### **Tobacco800 Dataset:**

Finding all name labels of identified signatures

```
select distinct to_char(pepv.value)

from datatype_property dp, value_of vo, page_element_property_value pepv
where

vo.data_type_PROPERTY_ID = dp.id

and dp.name = 'gedi_type_DLSignature'

and pepv.id = vo.page_element_property_value_id;
```

Finding all occurrences of a person's signature

```
select hv.page_element_id

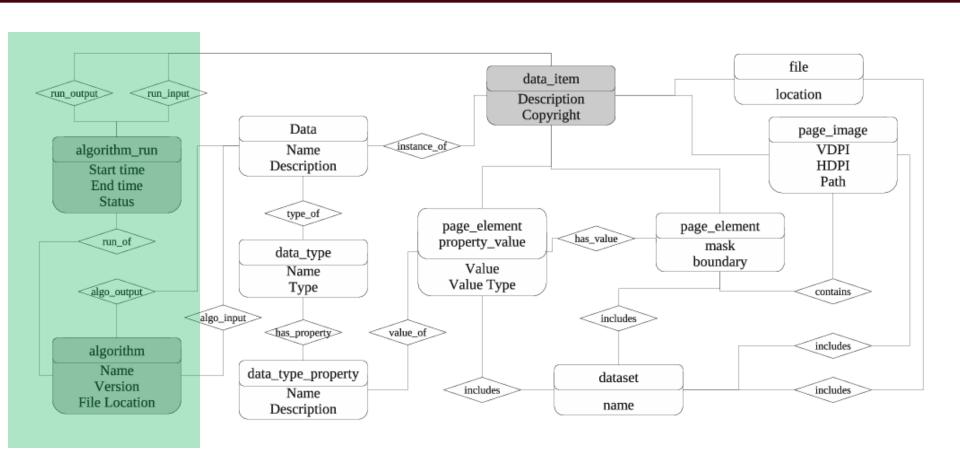
from datatype_property dp, value_of vo, page_element_property_value pepv, has_value hv
where

vo.data_type_PROPERTY_ID = dp.id
and dp.name = 'gedi_type_DLSignature'
and pepv.id = vo.page_element_property_value_id
and to_char(pepv.value) = 'John Doe'
and hv.page_element_property_value_id = pepv.id
```





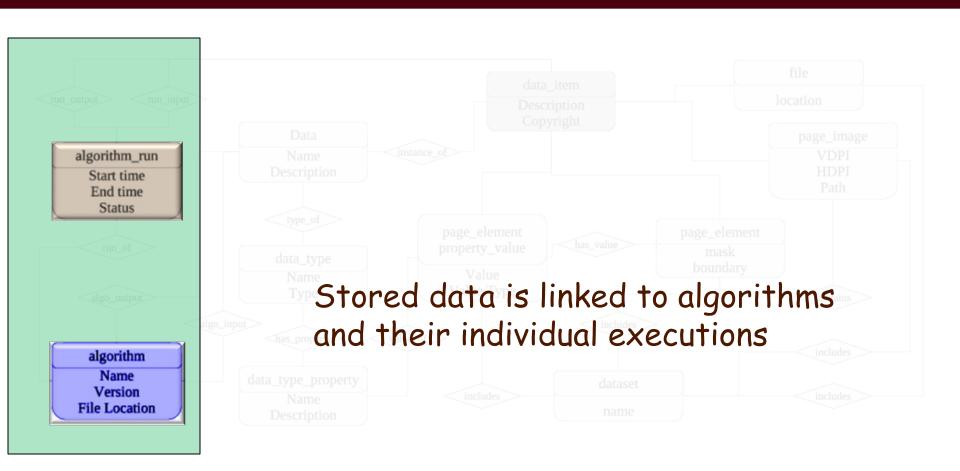
# Data Model: Algorithms







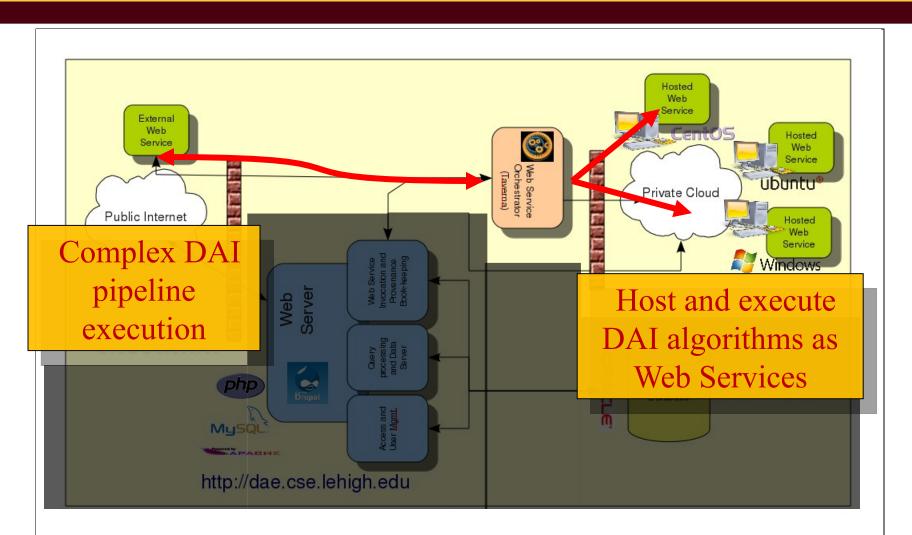
# Data Model: Algorithms







### DAE Platform - dae.cse.lehigh.edu







## Pitfalls

#### Coercion into

- programming languages
- data structures
- operating systems
- releasing control
- Obsoleteness
- Home-brew solutions
   (high maintenance costs)
- Steep learning curve





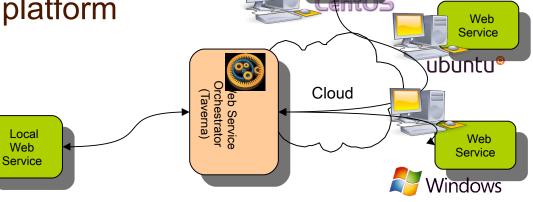
## Solution

### DAI Algorithms as Web Services

- WSDL open standard
- Not bound to any technology
- Formalized I/O

### Virtualized Environments

- Not bound to any platform
- Easily maintained
- Secure
- Cloud ready



Web





## Pitfalls

## Coercion into

- \*x programming languages
- \*X data structures
- \*X operating systems
- releasing control
- Obsoleteness
- Home-brew solutions

(high maintenance costs)

Steep learning curve ?

- No change in habits
- No change of existing code
- No need to release code
- High resilience and maintainability
- High level of interoperability





# WSDL & Learning Curve?

### Plug-and-Play wrappers

- No change in existing code
- For quick-win testing
   (15 lines of code in PhP)
- Taverna click-and-play

### More complex integration possible

- Can be progressive
- Supported by all major programming environments





## Benefits

- Pipelines = Experiments
  - Are formalized
  - Can be made available (published) for
    - Verification
    - Replay Re-use
    - Extension Modification
    - Debate
- Legacy code can be re-applied and maintained
- No requirement to integrate within the DAE platform





# Bonus for Integration within DAE

### • Full Provenance

- Find images on which algorithm A disagrees with algorithm B;
- Find images on which no algorithm has found the initially provided interpretation;
- Find interpretations with the strongest level of disagreement;
- Find all data seen by a user (or processed by an algorithm run by the same user);

### Etc.





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Challenges of reproducible and traceable experimental research

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A way of addressing the challenges, technical architecture and design choices

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## ICDAR 2011 Contest

#### Openain:

**Document Analysis** 

### • Challenge:

Extract Named Entities from scanned document

#### O Tools:

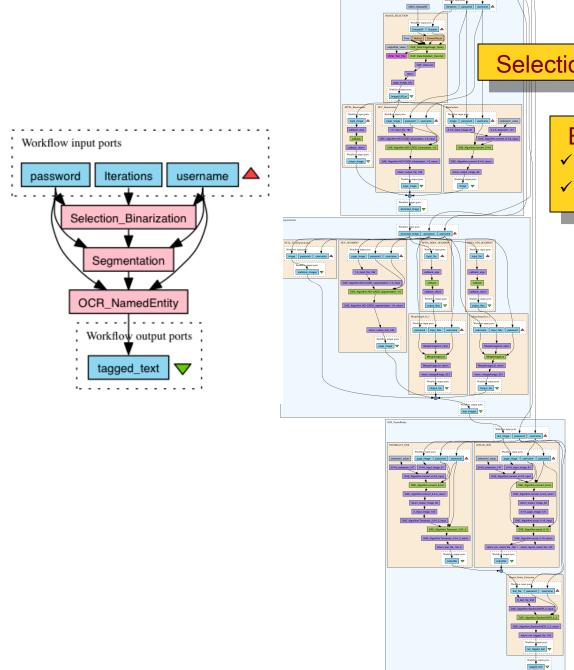
- DAE platform
- Binarization Segmentation OCR NLP
- Web Service wrapper code

#### O Task:

Contribute 1 component







Selection of Images on Image Server

#### Binarization Phase:

- ✓ 2 contributed algorithms
- ✓ 1 provided

#### Text Segmentation Phase:

- √ 3 contributed algorithms
- ✓ 1 provided

#### **OCR Phase:**

✓ 2 provided algorithms

Named Entity Detection

## Result

- 3 Binarization 4 Segmentation 2 OCR
  - = 24 pathways to compute same data.

#### Contributions were

- DAE hosted (Sun/Solaris, Linux, Windows)
- Remotely hosted
- Full Experimental Pipeline available and reproducible
- All Intermediate Data available and verifiable (including provenance: who, what, when)
- Full comparison possible by third party





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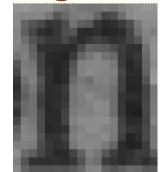




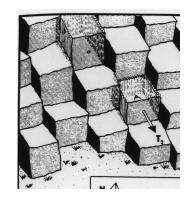
# Common Misconceptions

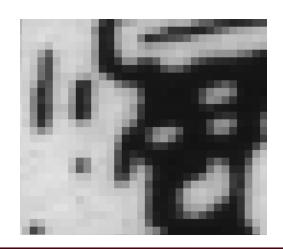
### Separating Signal form Background is Easy

Zamudio de Barboza, qua cto. El hombre debió se el hospital de Urgencia conductor del camión, Os salió ileso del percance niña y un joven



### **Shapes are Obvious**









# Common Misconceptions

## Experimental Validation Data is

- Unique
- Unambiguous
- Universal

## Experimental Validation Data

- Depends on context
- Can be open to controversy
- Can contain errors









# Truth vs. Interpretation

### Multiple Interpretation Contexts for Same Data

- Notion of "ground truth" is obsolete
- Data interpretation depends on
  - "intent" of the author (quasi-impossible to obtain)
  - "expression" of the author (usually undisputable)
  - "interpretation" of the user(s)
    - Human users
    - Algorithms



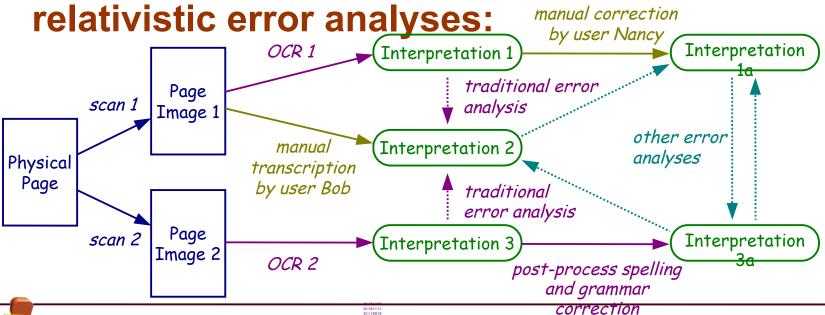


# Alternate Interpretations

### Entities on a page are located and interpreted:

- by humans interacting with the system
- by algorithms invoked by "others"

Possibility of alternate interpretations and





# Interpretations and Reputation

- •Which Interpretation is "the best"?
- •Which Interpretation is incompatible with my context?
  - Who has interacted with similar data?
  - Do I trust some more than others?
  - What is their interpretation context?
  - What is their reputation?





# Kinds of Reputation

#### Algorithm

How well does it address a stated task?

Implementation

Buggy or robust?

Dataset

Is it representative for a particular problem?

Interpretation

Is it trustworthy?

Publication

Is it highly cited?

Researcher

Amalgamation of contributions to above







## Further Work

### User Interaction

- Querying
- Semantics Discovery
- Multiple Interpretation Selection

### Promotion and Dissemination

- Multiple Platform Interaction
- Community Development





## Further Work

#### User Interaction

- Querying
- Semantics Discovery
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  - Multiple Platform Interaction
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# **Semantics Discovery**

#### Context

- The semantics of "Semantics" ill defined
- Interpretation and Context discovery

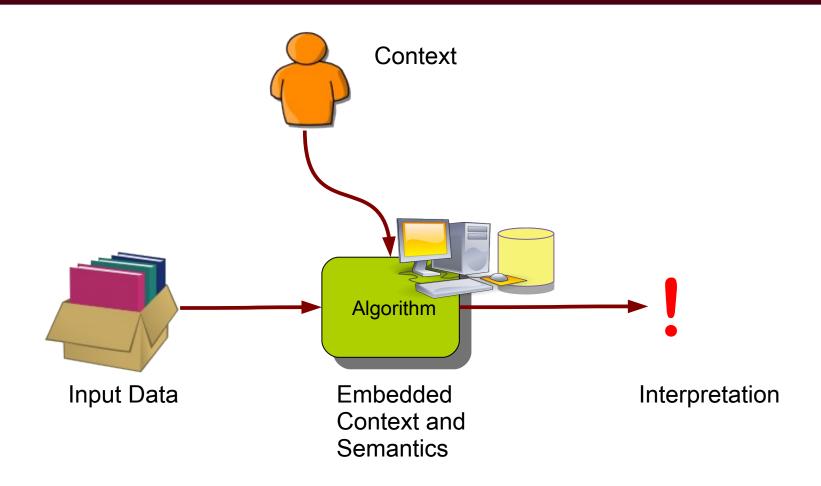
### Tools

- DAE Provenance data
- Data interaction ontology
- Researcher ontology





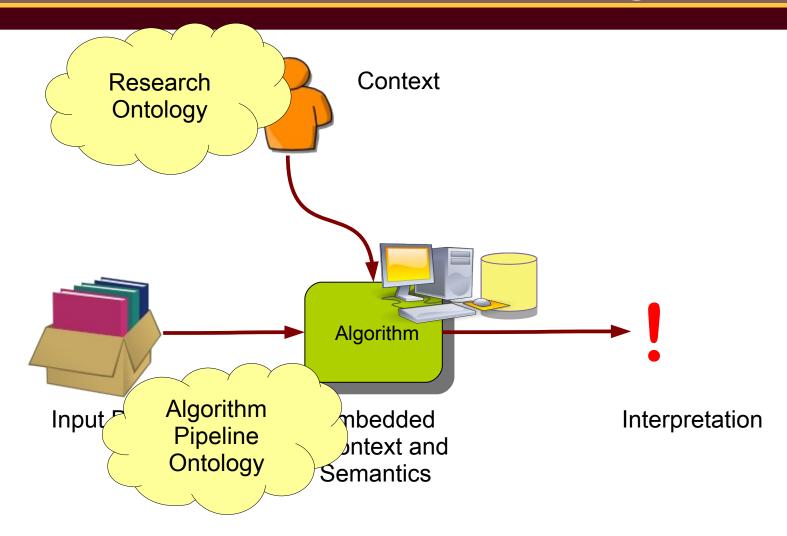
# Semantics Discovery







# Context Learning





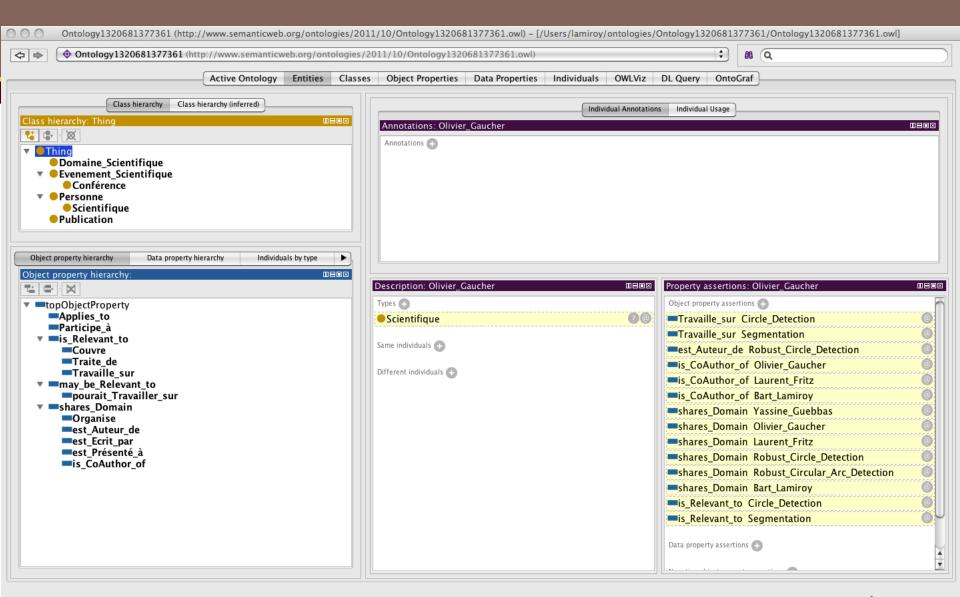


# Open Question

Res Appropriate Context **Ontologies DAE Provenance** Data Research Social **Network Information** Interpretation Context Document **Semantics? Learning** ?Semantics













## Further Work

#### User Interaction

- Querying
- Semantics Discovery
- Multiple Interpretation Selection

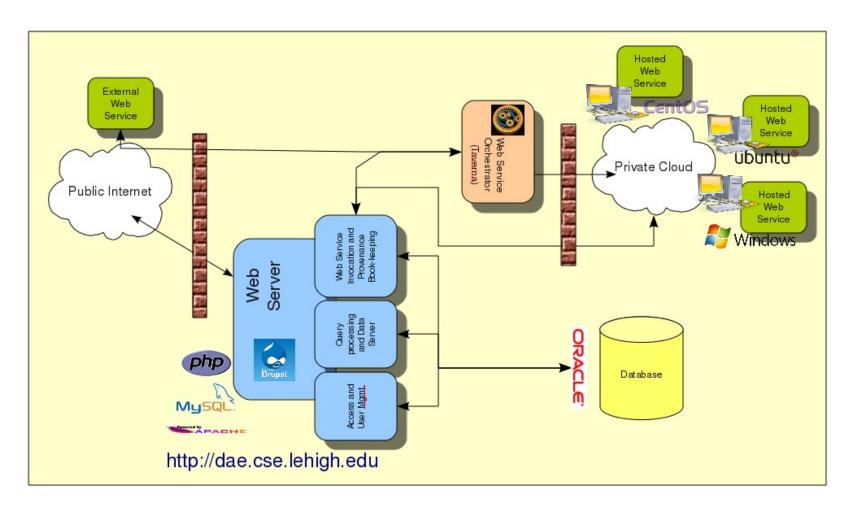
### Promotion and Dissemination

- Multiple Platform Interaction
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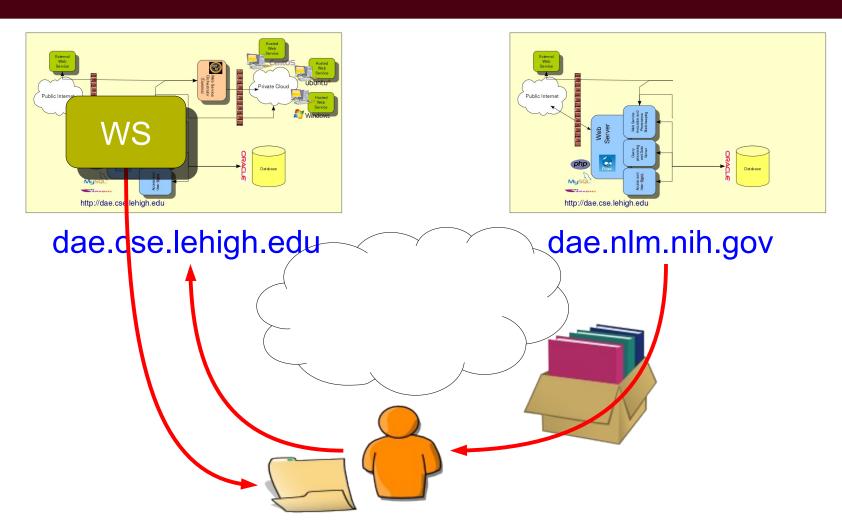
# DAE Architecture







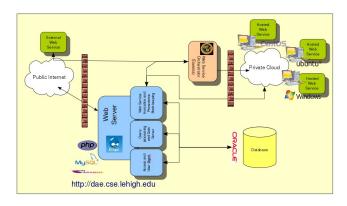
# Multiple DAE Platforms





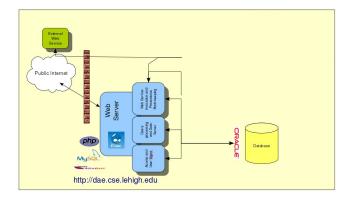


# Extending Provenance Registery



dae.cse.lehigh.edu

- Algorithm WS was run by UserX using DocY from dae-nlm.cse.lehigh.edu
- Register provenance
- Register dae-nlm



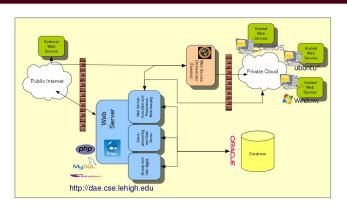
dae.nlm.nih.gov

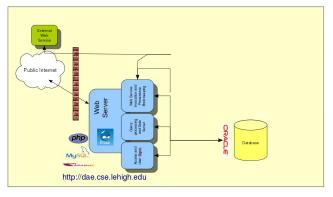
- UserX has accessed DocY from dae.cse.lehigh.edu
- Register dae

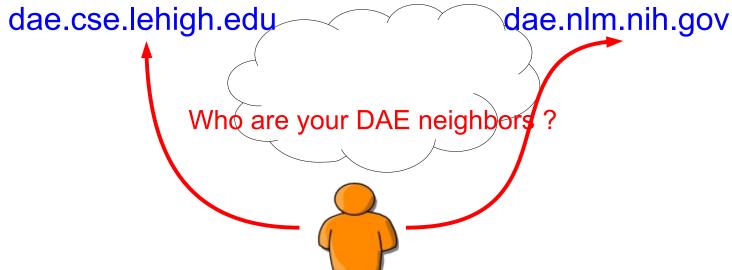




# Social Network Building



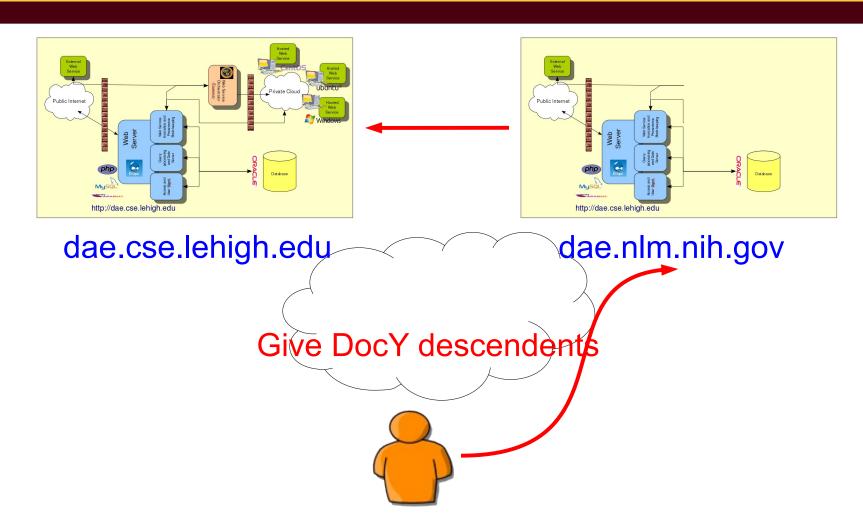








## Distributed Provenance







## Technical Issues

### Uniform User Identification

- Drupal accepts OpenID
- Some integration required for web-services
- Taverna?

### External data\_items

- OK in current model
- Handling resiliency (versioning) with caching
- Copyright issues of descendants?

### Constructing Distributed Provenance Queries

Avoiding loops?





## Conclusion

### **DAE Plaform is an Operational tool for:**

- large, open and re-usable collections
- handling multiple interpretations of data
- user defined and extended new data types
- certifiable interaction with data
- certifiable algorithm benchmarking





## Resources

- http://dae.cse.lehigh.edu
- http://sourceforge.net/projects/daeplatform/
- http://tinyurl.com/DAE-Wiki
- http://tinyurl.com/DAE-WebServices



