Managing Hydrologic Data, Observations and Terrain Analysis

David Tarboton

Utah State University

http://www.engineering.usu.edu/dtarb/

Outline

The CUAHSI Hydrologic Information System



http://his.cuahsi.org/

Terrain Analysis Using Digital Elevation Models



http://hydrology.usu.edu/taudem

Hydrologic Data Challenges

- From dispersed federal agencies
- From investigators collected for different purposes
- Different formats
 - Points
 - Lines
 - Polygons
 - Fields

Data Heterogeneity

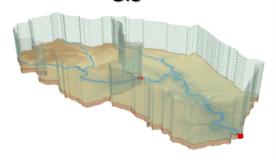
Water quality



Rainfall and Meteorology



GIS



Water quantity



Soil water



Groundwater



The way that data is organized can enhance or inhibit the analysis that can be done



I have your information right here ...

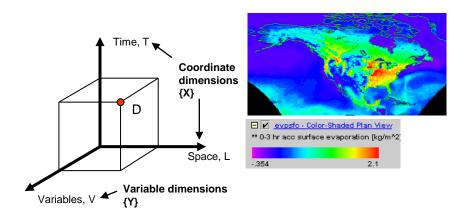
Hydrologic Science

It is as important to represent hydrologic environments precisely with data as it is to represent hydrologic processes with equations

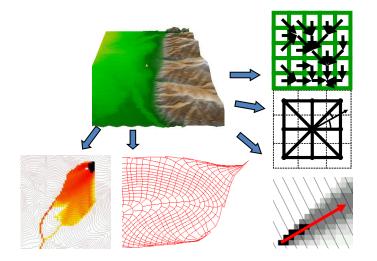
Physical laws and principles (Mass, momentum, energy, chemistry) Hydrologic Process Science (Equations, simulation models, prediction) Hydrologic conditions (Fluxes, flows, concentrations) **Hydrologic Information Science** (Observations, data models, visualization Hydrologic environment (Physical earth)

Data models capture the complexity of natural systems

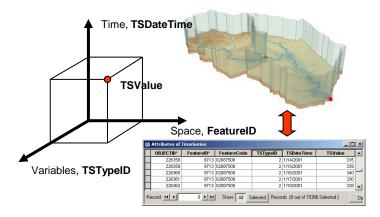
NetCDF (Unidata) - A model for Continuous Space-Time data



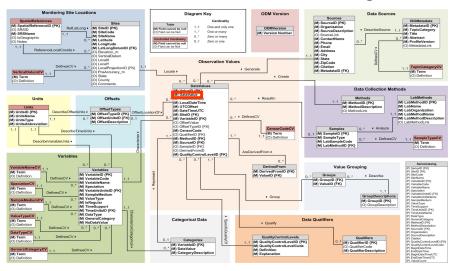
Terrain Flow Data Model used to enrich the information content of a digital elevation model



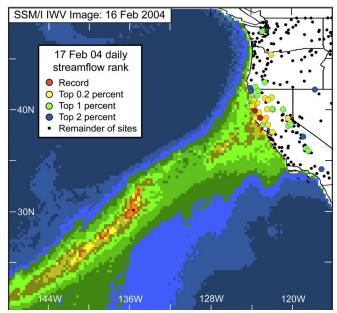
ArcHydro – A model for Discrete Space-Time Data

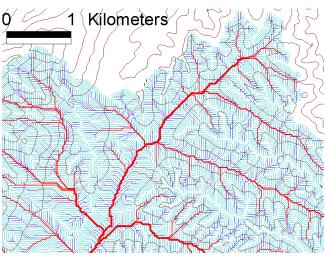


CUAHSI Observations Data Model: What are the basic attributes to be associated with each single data value and how can these best be organized?



Data intensive science synthesizes large quantities of information (Hey et al., 2009).

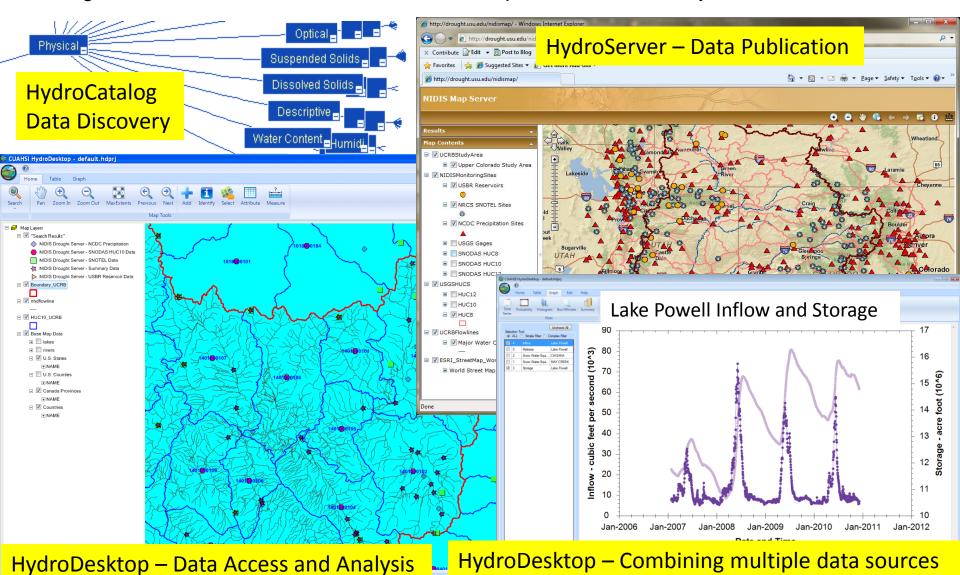




- exploiting advanced computational capability for the analysis and integration of large new datasets to elucidate complex and emergent behavior
- In hydrology, the image at left (Ralph et al., 2006) illustrates connection between extreme floods recorded in USGS stream gages and atmospheric water vapor from space based sensors
- Satellite remote sensing and massive datasets enhance understanding of multi-scale complexity in processes such as rainfall and river networks

CUAHSI HIS

The CUAHSI Hydrologic Information System (HIS) is an internet based system to support the sharing of hydrologic data. It is comprised of hydrologic databases and servers connected through web services as well as software for data publication, discovery and access.



The CUAHSI Community Hydrologic Information System

- University of Texas at Austin David Maidment, Tim Whiteaker,
 James Seppi, Fernando Salas, Jingqi Dong, Harish Sangireddy
- San Diego Supercomputer Center Ilya Zaslavsky, David Valentine, Tom Whitenack, Matt Rodriguez
- Utah State University Jeff Horsburgh, Kim Schreuders, Stephanie Reeder, Edward Wai Tsui, Ravichand Vegiraju, Ketan Patil
- University of South Carolina Jon Goodall, Anthony Castronova
- Idaho State University Dan Ames, Ted Dunsford, Jiří Kadlec, Yang Cao, Dinesh Grover
- Drexel University/CUNY Michael Piasecki
- WATERS Network Testbed Data Managers
- CUAHSI Program Office Rick Hooper, Yoori Choi, Conrad Matiuk
- ESRI Dean Djokic, Zichuan Ye

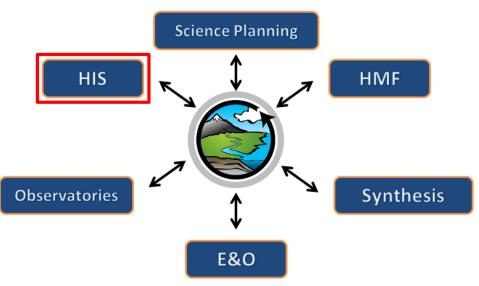




What is CUAHSI?

Consortium of Universities for the Advancement of Hydrologic Science, Inc.





- 110 US University members
- 6 affiliate members
- 12 International affiliate members (as of March 2009)

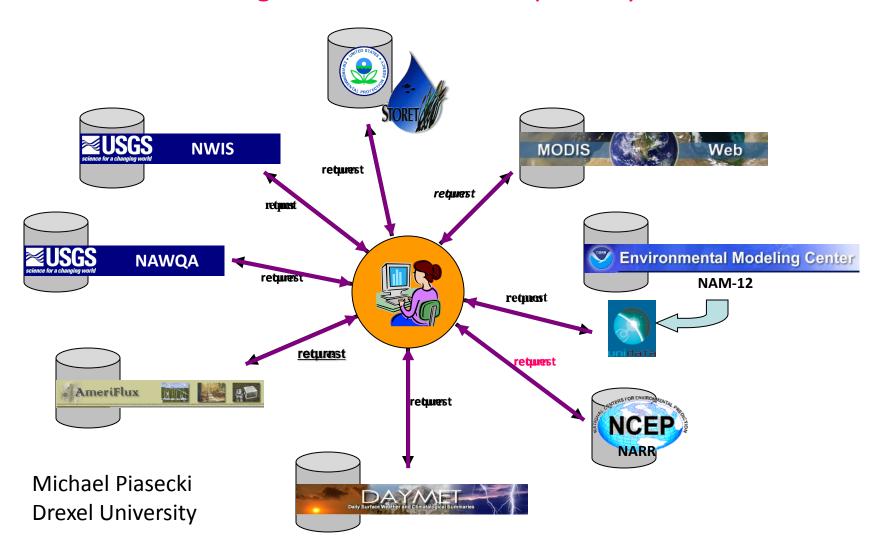
Infrastructure and services for the advancement of hydrologic science and education in the U.S.



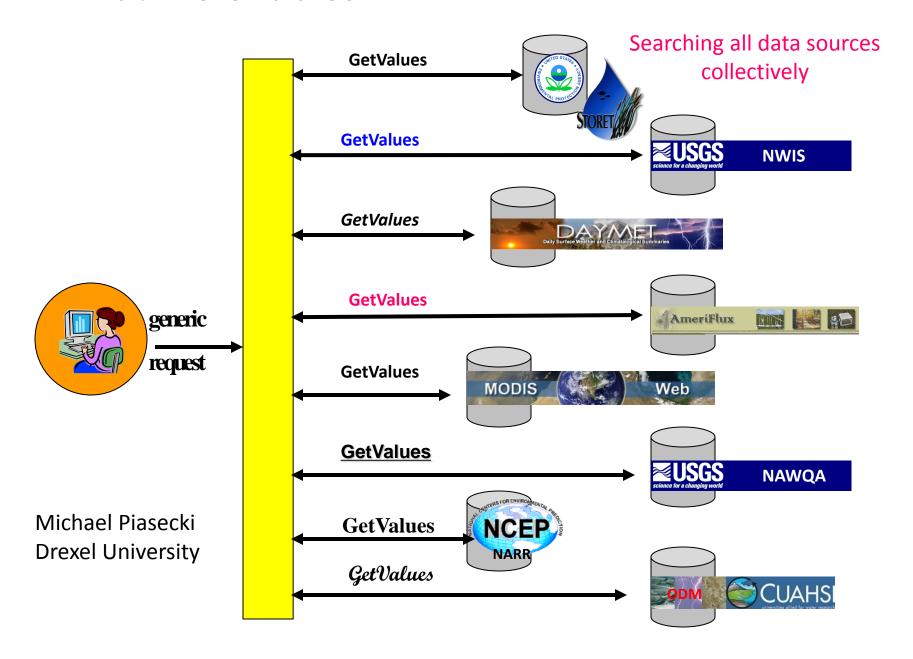


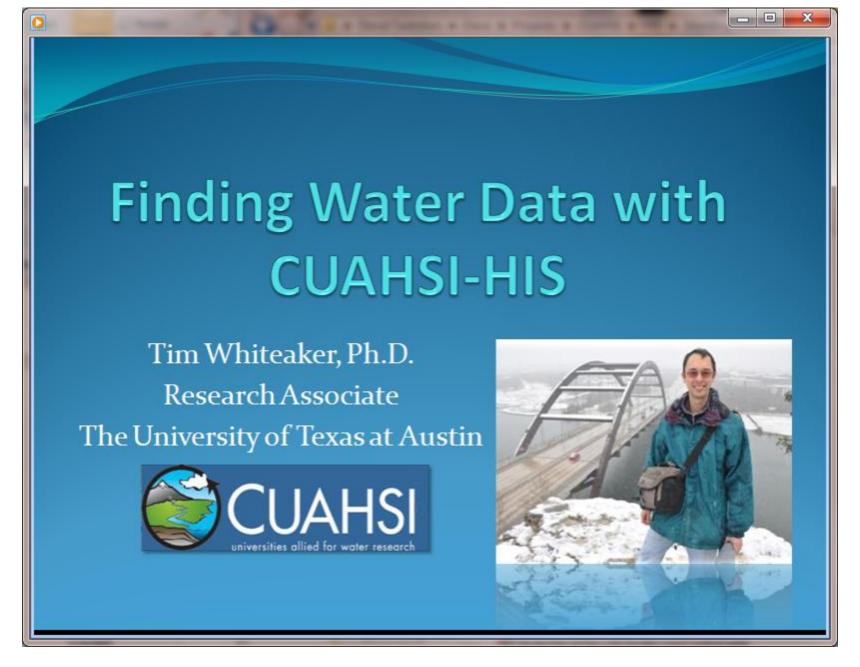
Data Searching – What we used to have to do

Searching each data source separately

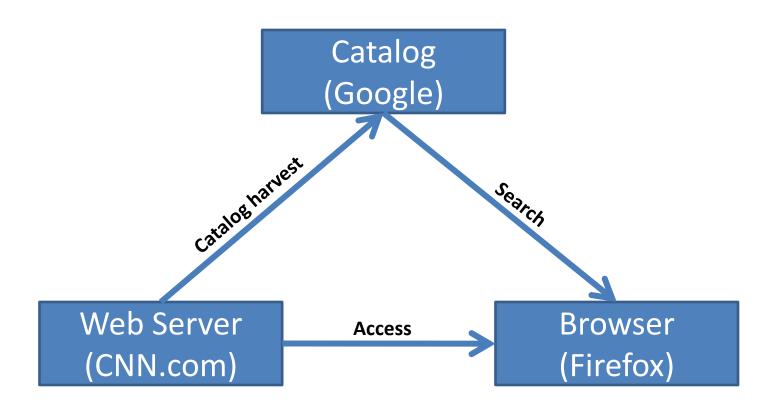


What HIS enables

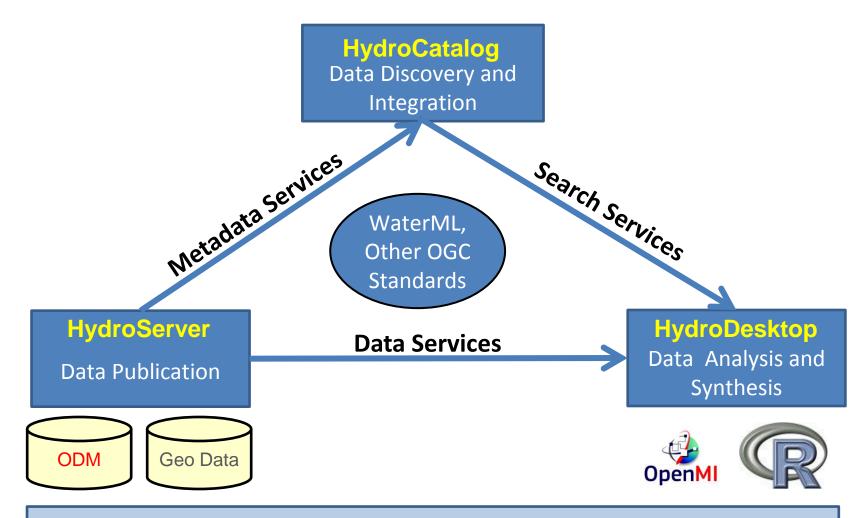




Web Paradigm



CUAHSI Hydrologic Information System Services-Oriented Architecture



Information Model and Community Support Infrastructure

What are the basic attributes to be associated with each single data value and how can these best be organized?

Value
DateTime
Variable
Location
Units

Interval

(support)

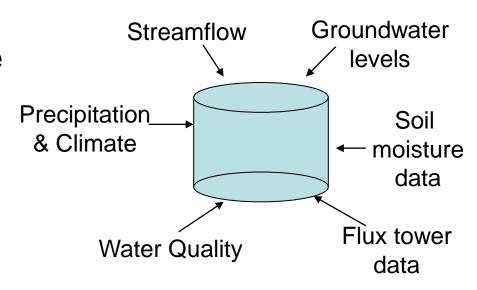
Accuracy

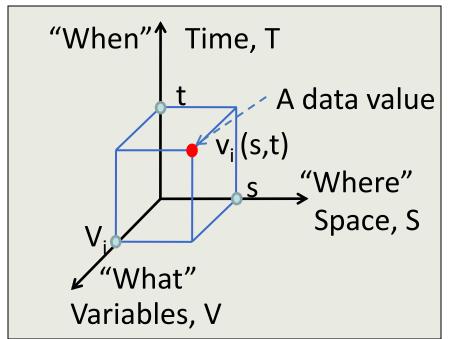
Offset
OffsetType/ Reference Point
Source/Organization
Censoring
Data Qualifying Comments

Method
Quality Control
Level
Sample Medium
Value Type
Data Type

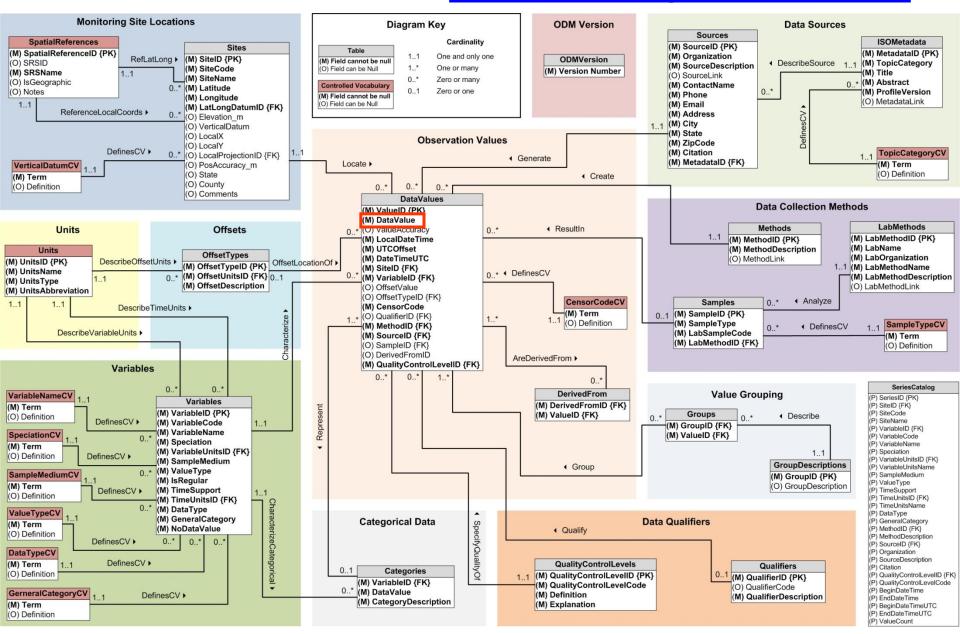
CUAHSI Observations Data Model

- A relational database at the single observation level (atomic model)
- Stores observation data made at points
- Metadata for unambiguous interpretation
- Traceable heritage from raw measurements to usable information
- Standard format for data sharing
- Cross dimension retrieval and analysis



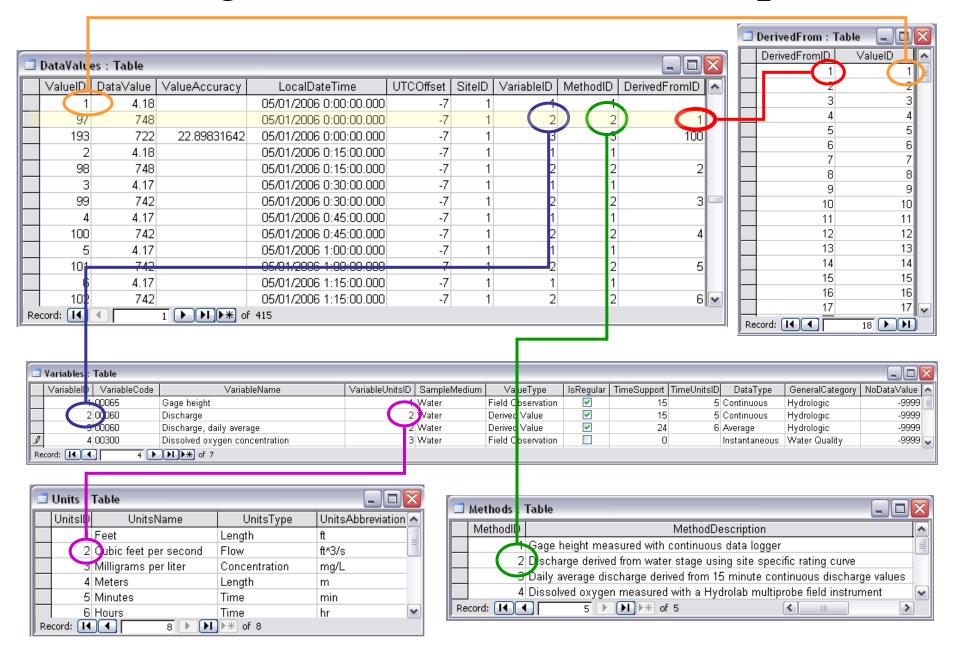


CUAHSI Observations Data Model http://his.cuahsi.org/odmdatabases.html



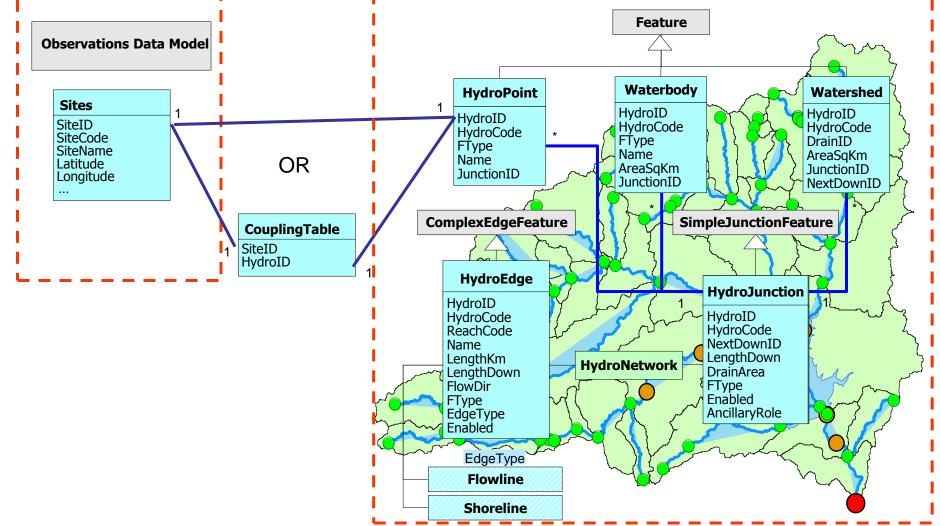
Horsburgh, J. S., D. G. Tarboton, D. R. Maidment and I. Zaslavsky, (2008), A Relational Model for Environmental and Water Resources Data, *Water Resour. Res.*, 44: W05406, doi:10.1029/2007WR006392.

Stage and Streamflow Example



Independent of, but can be coupled to Geographic Representation

ODM e.g. Arc Hydro



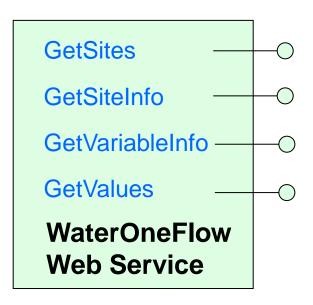
Importance of the Observations Data Model

- Provides a common persistence model for observations data
- Syntactic heterogeneity (File types and formats)
- Semantic heterogeneity
 - Language for observation attributes (structural)
 - Language to encode observation attribute values (contextual)
- Publishing and sharing research data
- Metadata to facilitate unambiguous interpretation
- Enhance analysis capability

WaterML and WaterOneFlow

WaterML is an XML language for communicating water data WaterOneFlow is a set of web services based on WaterML

Set of query functions



Returns data in WaterML

```
ctimeSeries>
 <sourceInfo xsi:type="SiteInfoType">
  <siteName>Colorado Rv at Austin, TX</siteName>
   <siteCode network="NWIS" siteID="4619631">0815800
- <geoLocation>
  - <geogLocation xsi:type="LatLonPointType" srs="EPSG</p>
      <latitude>30.24465429
      <longitude>-97.694448
    </geogLocation>
  </geoLocation>
 </sourceInfo>
 <variable>
  <variableCode vocabulary="NWIS" default="true" variable</p>
  <variableName>Discharge, cubic feet per second/vari
   <units unitsAbbreviation="cfs" unitsCode="35">cubic fee
 </variable>
 <values count="2545">
  <value dateTime="2006-12-31T00:00:00">129</value>
  <value dateTime="2006-12-31T00:15:00">129</value>
  <value dateTime="2006-12-31T00:30:00">129</value>
  <value dateTime="2006-12-31T00:45:00">129</value>
  <value dateTime="2006-12-31T01:00:00">124</value>
  <value dateTime="2006-12-31T01:15:00">129</value>
  <value dateTime="2006-12-31T01:30:00">124</value>
   <value dateTime="2006-12-31T01:45:00">124</value>
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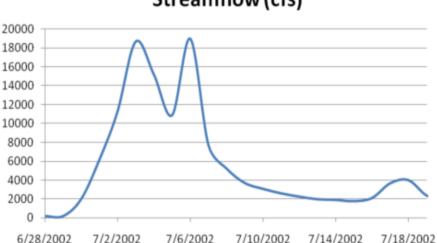
WaterML as a Web Language

USGS Streamflow data in WaterML language

<value qualifiers="A" dateTime="2002-07-18T00:00:00">4010/value>

Discharge of the San Marcos River at Luling, TX June 28 - July 18, 2002

```
<values count="21">
 <value qualifiers="A" dateTime="2002-06-28T00:00:00">203
 <value qualifiers="A" dateTime="2002-06-29T00:00">195</value>
                                                                                          Streamflow (cfs)
 <value qualifiers="A" dateTime="2002-06-30T00:00:00">2010</value>
 <value qualifiers="A" dateTime="2002-07-01T00:00:00">6170
                                                                     20000
 <value qualifiers="A" dateTime="2002-07-02T00:00:00">11300</value>
                                                                     18000
 <value qualifiers="A" dateTime="2002-07-03T00:00:00">18700
                                                                     16000
 <value qualifiers="A" dateTime="2002-07-04T00:00:00">15200
 <value qualifiers="A" dateTime="2002-07-05T00:00:00">10900 /value>
                                                                     14000
 <value qualifiers="A" dateTime="2002-07-06T00:00:00">19000
                                                                     12000
 <value qualifiers="A" dateTime="2002-07-07T00:00:00">7720</value>
                                                                     10000
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                                                                      8000
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                                                                      6000
 <value qualifiers="A" dateTime="2002-07-10T00:00:00">3090</value>
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                                                                      4000
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                                                                      2000
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 <value qualifiers="A" dateTime="2002-07-14T00:00:00">1920
                                                                                  7/2/2002
                                                                                            7/6/2002
                                                                       6/28/2002
 <value qualifiers="A" dateTime="2002-07-15T00:00:00">1780</value>
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 <value qualifiers="A" dateTime="2002-07-17T00:00:00">3680</value>
```

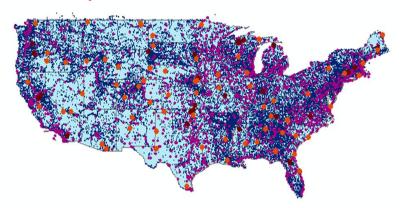


<qualifier qualifierCode="A" network="USGS" vocabulary="dv_rmk_cd">Approved for publication -- Processing and review completed.</qualifier: </values>

This is the WaterML GetValues response from NWIS Daily Values

Open Geospatial Consortium Web Service Standards

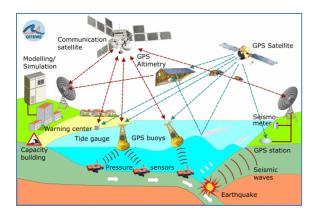
Map Services



- Web Map Service (WMS)
- Web Feature Service (WFS)
- Web Coverage Service (WCS)
- Catalog Services for the Web (CS/W)

These standards have been developed over the past 10 years by 400 companies and agencies working within the OGC

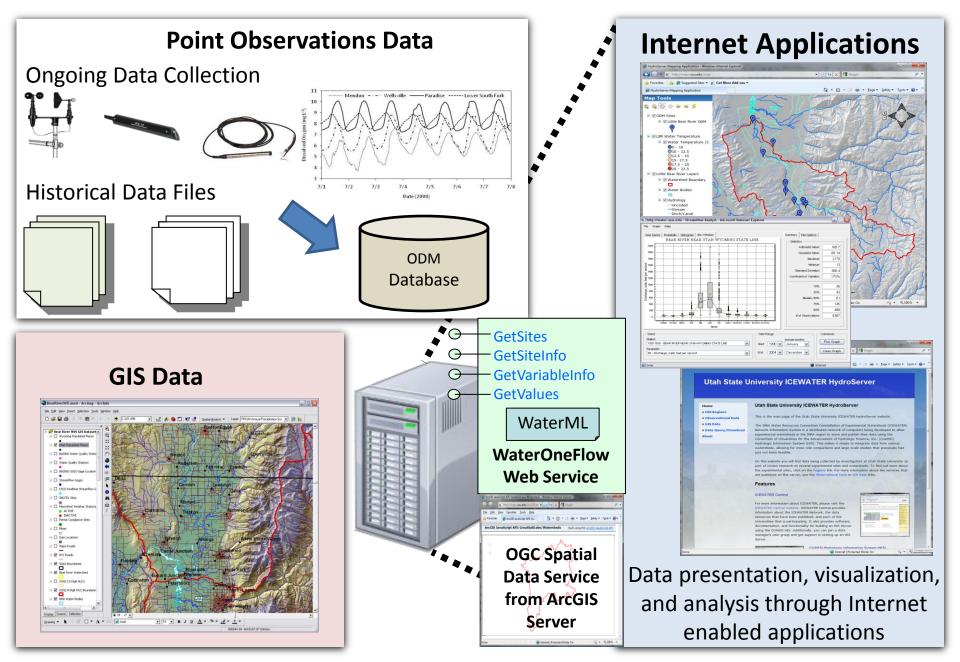
Observation Services



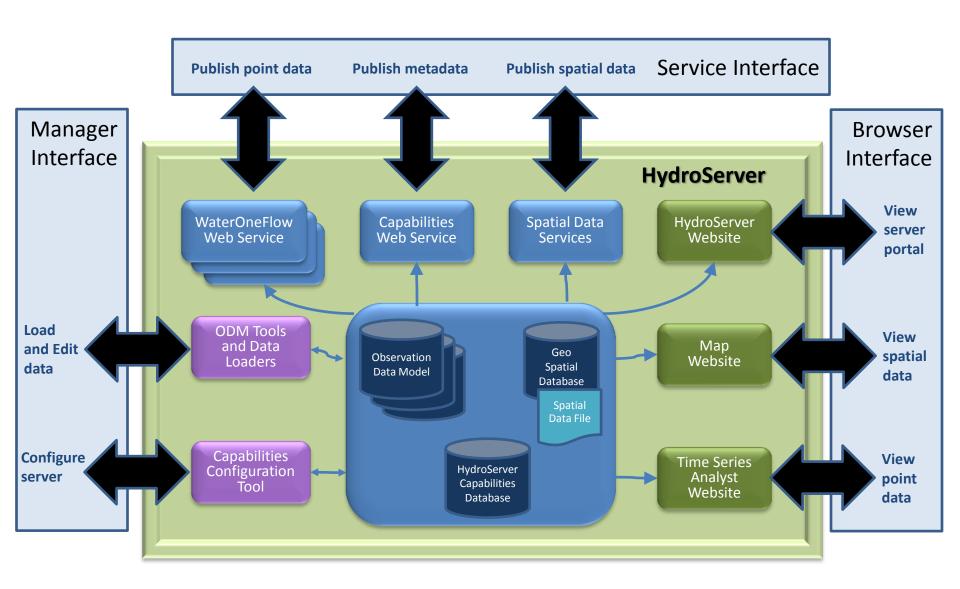
- Observations and Measurements Model
- Sensor Web Enablement (SWE)
- Sensor Observation Service (SOS)

OGC Hydrology Domain Working Group evolving WaterML into an International Standard http://www.opengeospatial.org/projects/groups/waterml2.0swg

HydroServer – Data Publication

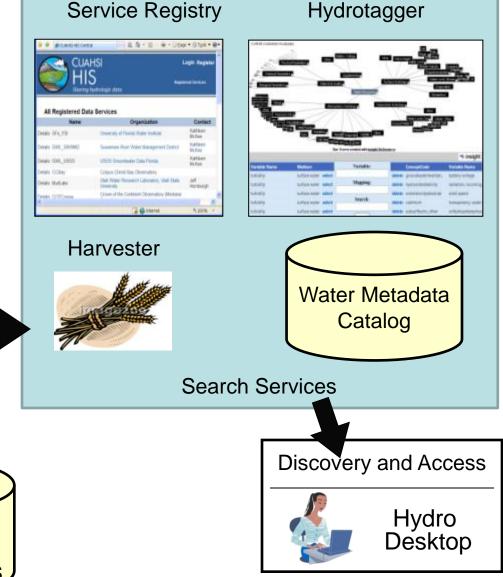


HydroServer



HydroCatalog

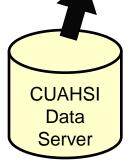
- Search over data services from multiple sources
- Supports concept based data discovery



WaterML

GetSites
GetSiteInfo
GetVariableInfo
GetValues

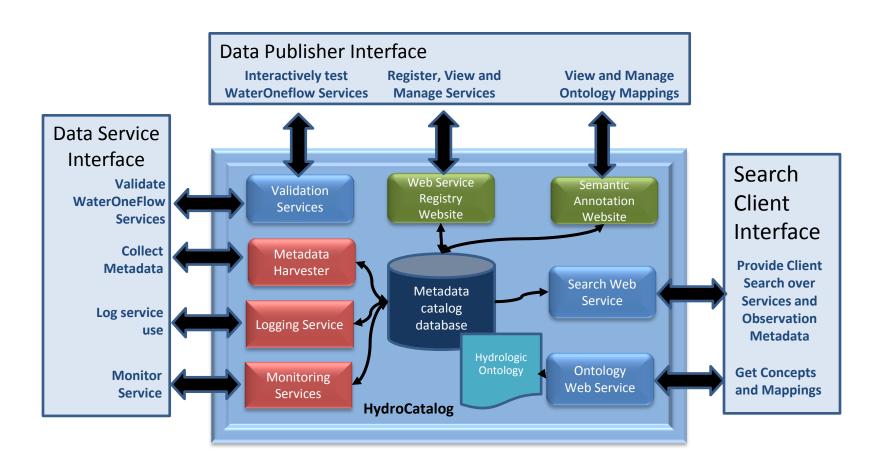
WaterOneFlow Web Service



3rd Party Server e.g. USGS

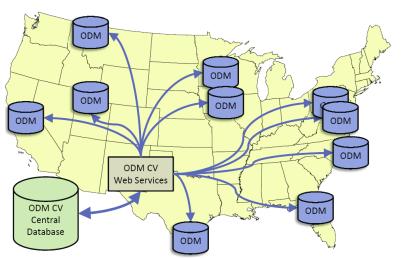
http://hiscentral.cuahsi.org

HydroCatalog

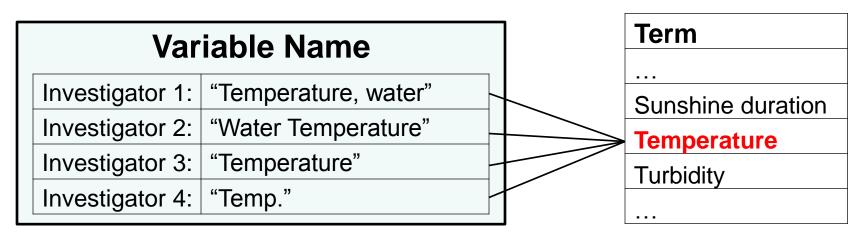


Overcoming Semantic Heterogeneity

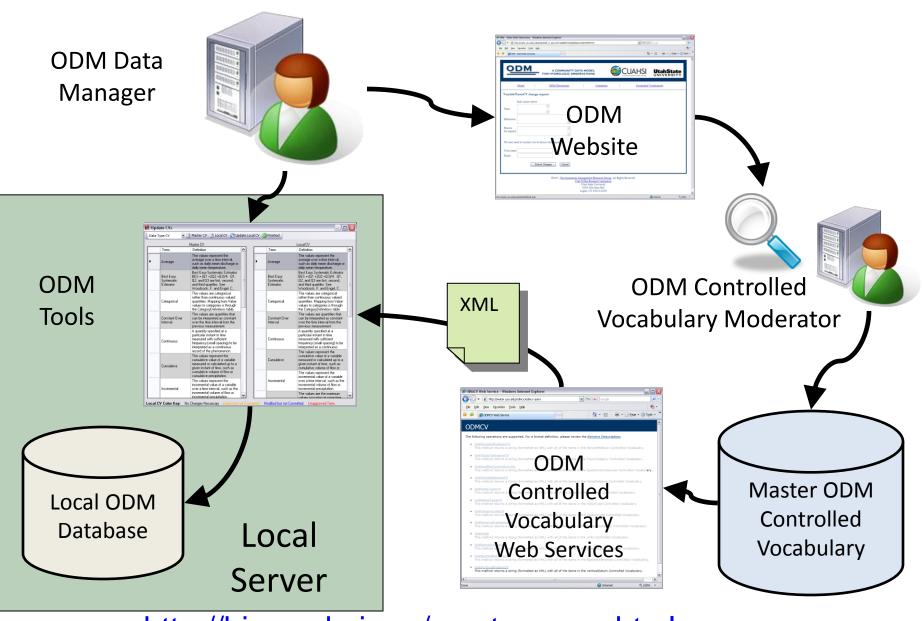
- ODM Controlled Vocabulary System
 - ODM CV central database
 - Online submission and editing of CV terms
 - Web services for broadcasting
 CVs



ODM VariableNameCV

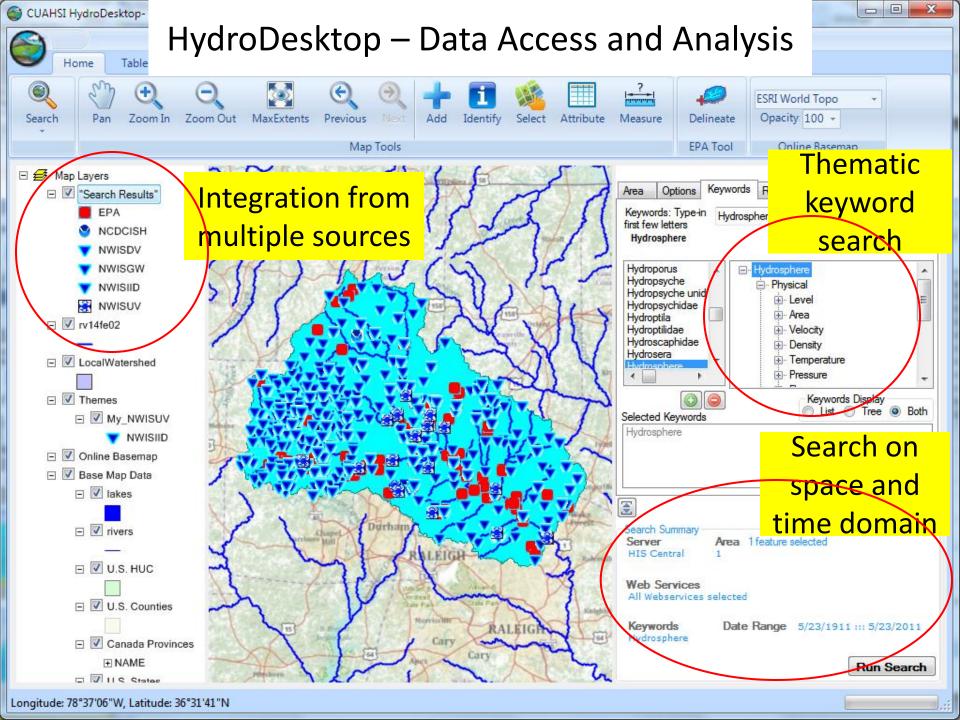


Dynamic controlled vocabulary moderation system

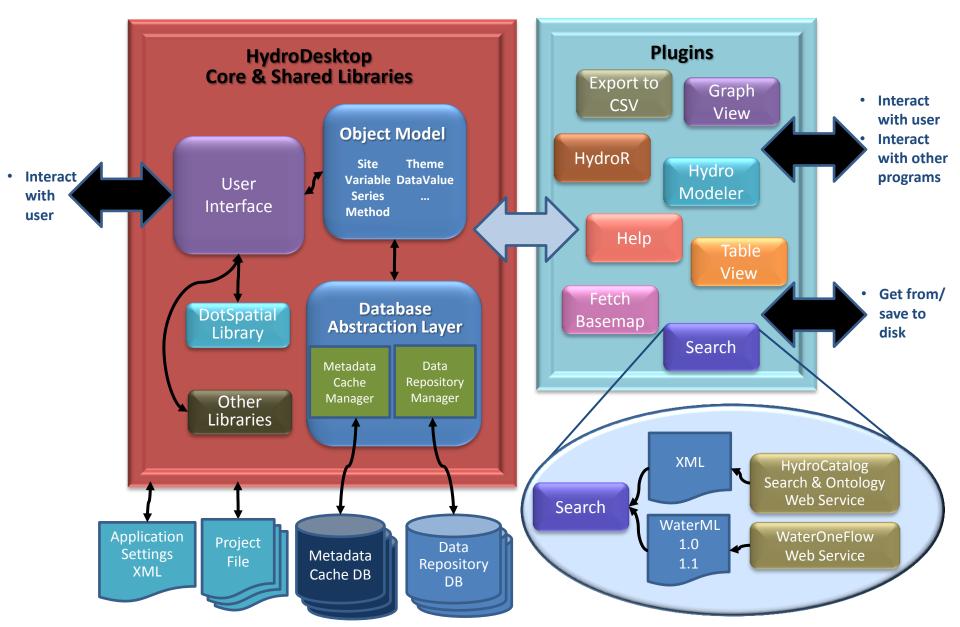


http://his.cuahsi.org/mastercvreg.html

From Jeff Horsburgh

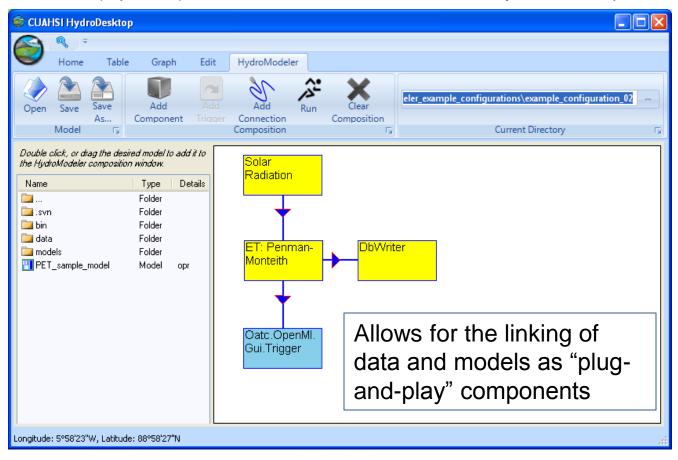


HydroDesktop



HydroModeler

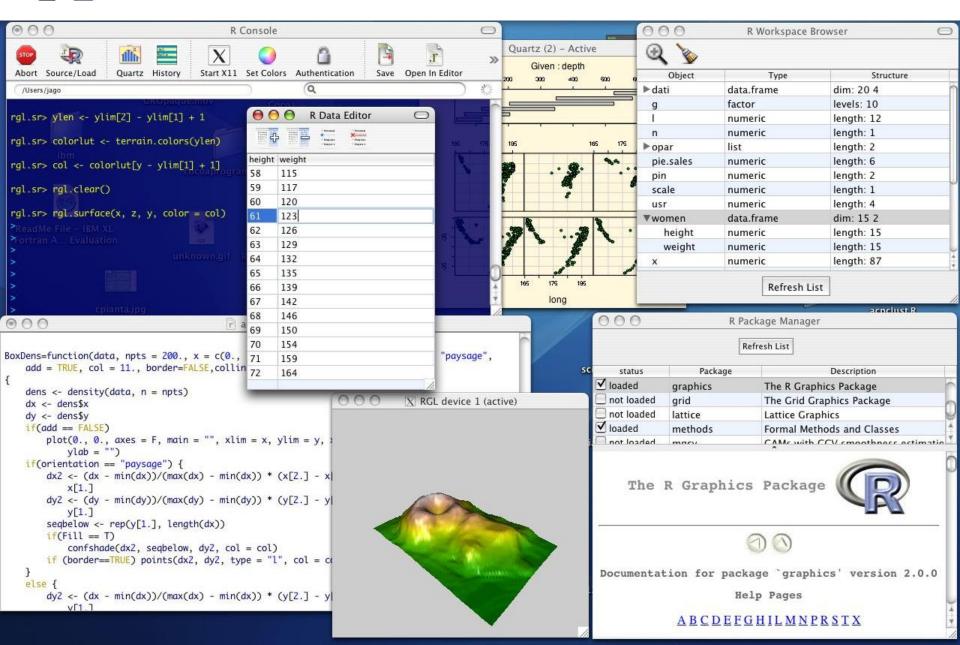
An integrated modeling environment based on the Open Modeling Interface (OpenMI) standard and embedded within HydroDesktop



In development at the University of South Carolina by Jon Goodall, Tony Castronova, Mehmet Ercan, Mostafa Elag, and Shirani Fuller



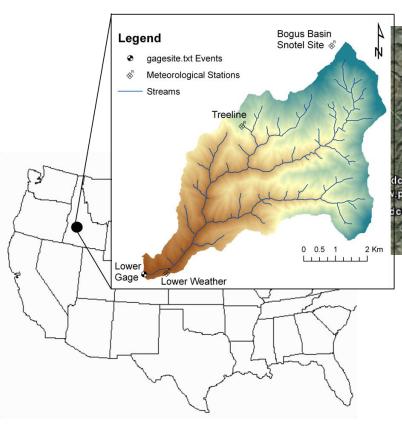
Integration with "R" Statistics Package



37 Water Data Services on HIS Central from 12 Universities

- University of Maryland,
 Baltimore County
- Montana State University
- University of Texas at Austin
- University of lowa
- Utah State University
- University of Florida
- University of New Mexico
- University of Idaho
- Boise State University
- University of Texas at Arlington
- University of California,
 San Diego
- Idaho State University

Dry Creek Experimental Watershed (DCEW) (28 km² semi-arid steep topography, Boise Front)



dcew.p51
dcew.p54
dcew.p56
ew.s1
68 Sites
24 Variables

dcew.p4

dcew.p49 dcew.s6

dcew.p52

Published by Jim McNamara, Boise State University

4,700,000+ values

Water Agencies and Industry

- USGS, NCDC, Corps of Engineers publishing data using HIS WaterML
- OGC Hydrology Domain Working Group evaluating WaterML as OGC standard
- ESRI using CUAHSI model in ArcGIS.com GIS data collaboration portal
- Kisters WISKI support for WaterML data publication
- Australian Water Resources
 Information System Water Accounting
 System has adopted aspects of HIS
- NWS West Gulf River Forecast Center Multi-sensor Precipitation Estimate published from ODM using WaterML

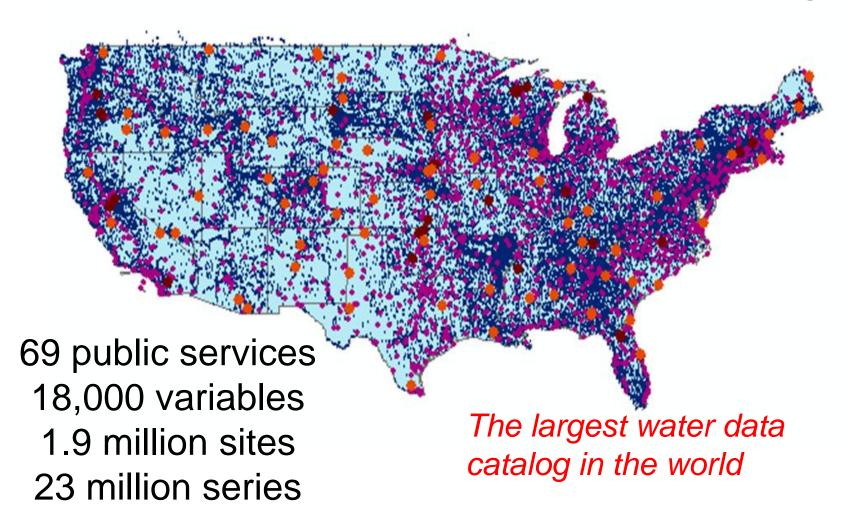








CUAHSI Water Data Services Catalog



5.1 billion data values (as of June 2011)

maintained at the San Diego Supercomputer Center 37

Open Development Model



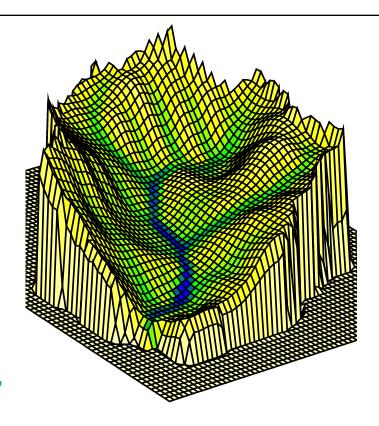
Summary

- Data Storage in an Observations Data Model (ODM) and publication through HydroServer
- Data Access through internet-based Water Data Services using a consistent data language, called WaterML from HydroDesktop
- Data Discovery through a National Water Metadata
 Catalog and thematic keyword search system at Central
 HydroCatalog (SDSC)
- Integrated Modeling and Analysis within HydroDesktop
 The combination of these capabilities creates a common window on water observations data for the United States unlike any that has existed before.

Terrain Analysis Using Digital Elevation Models (TauDEM)

David Tarboton¹, Dan Watson², Rob Wallace,³ Kim Schreuders¹, Jeremy Neff¹

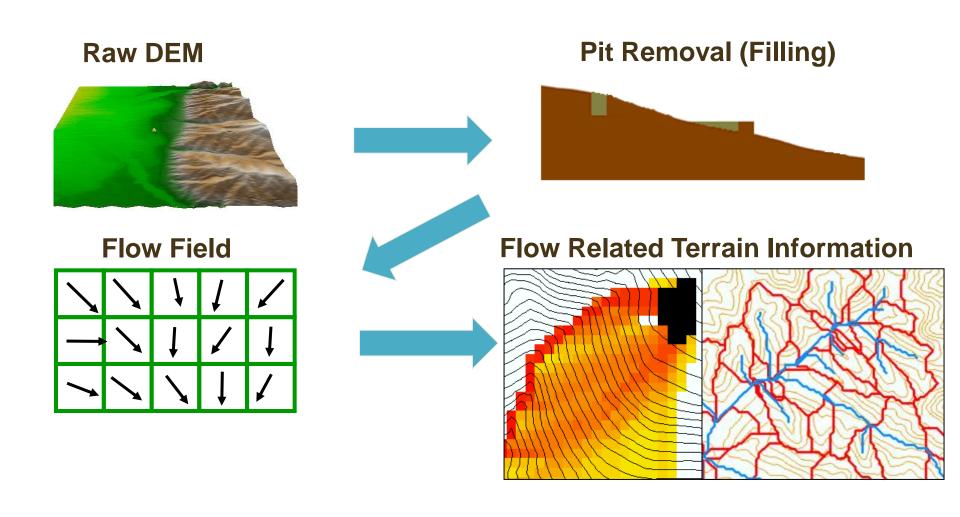
- ¹Utah Water Research Laboratory, Utah State University, Logan, Utah
- ²Computer Science, Utah State University, Logan, Utah
- ³US Army Engineer Research and Development Center, Information Technology Lab, Vicksburg, Mississippi







Deriving hydrologically useful information from Digital Elevation Models

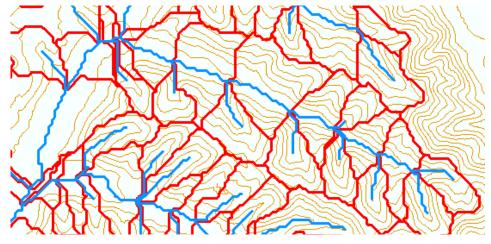


A parallel version of the TauDEM Software Tools

TauDEM Tools 🧞 Basic Grid Analysis D-Infinity Contributing Area D-Infinity Flow Directions D8 Contributing Area 🎤 D8 Flow Directions 🎤 Grid Network 🎤 Pit Remove 🖹 🥙 Specialized Grid Analysis D-Infinity Avalanche Runout D-Infinity Concentration Limited Accumulation D-Infinity Decaying Accumulation D-Infinity Distance Down D-Infinity Distance Up D-Infinity Reverse Accumulation D-Infinity Transport Limited Accumulation D-Infinity Upslope Dependence D8 Distance to Streams 🎤 Slope Average Down 済 Slope Over Area Ratio 🖹 🥙 Stream Network Analysis 🎤 D8 Extreme Upslope Value 🎤 Length Area Stream Source Move Outlets to Streams Peuker Douglas 🎤 Peuker Douglas Stream Definition Slope Area Combination 🎤 Slope Area Stream Definition Stream Definition By Threshold Stream Definition with Drop Analysis Stream Drop Analysis Stream Reach and Watershed

🎤 Watershed Grid to Shapefile

- Improved runtime efficiency
- Capability to run larger problems
- Platform independence of core functionality



Deployed as an ArcGIS Toolbox with tools that drive accompanying command line executables, available from http://hydrology.usu.edu/taudem/

The challenge of increasing Digital Elevation Model (DEM) resolution

1980's DMA 90 m 10² cells/km²

1990's USGS DEM 30 m 10³ cells/km²

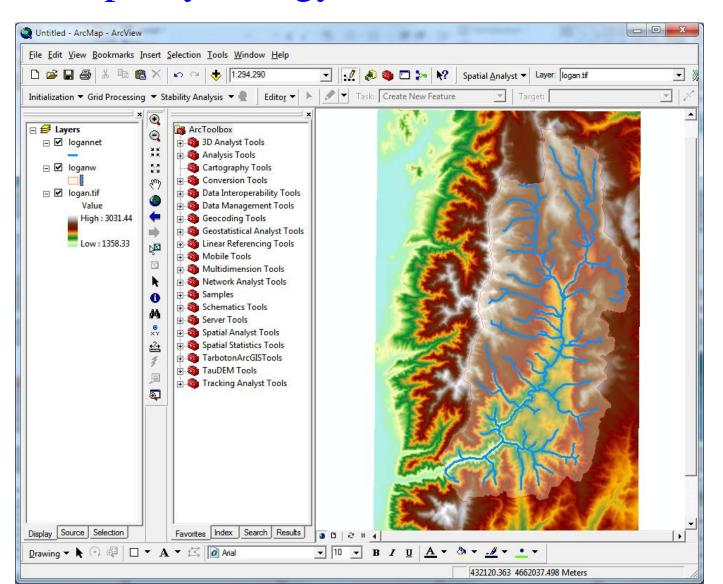
2000's NED 10-30 m 10⁴ cells/km²

2010's LIDAR ~1 m 10⁶ cells/km²

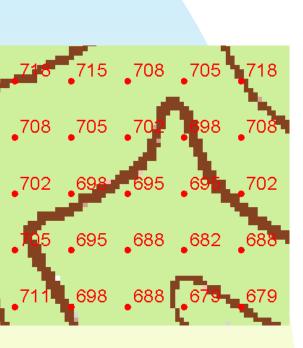


Website and Demo

http://hydrology.usu.edu/taudem

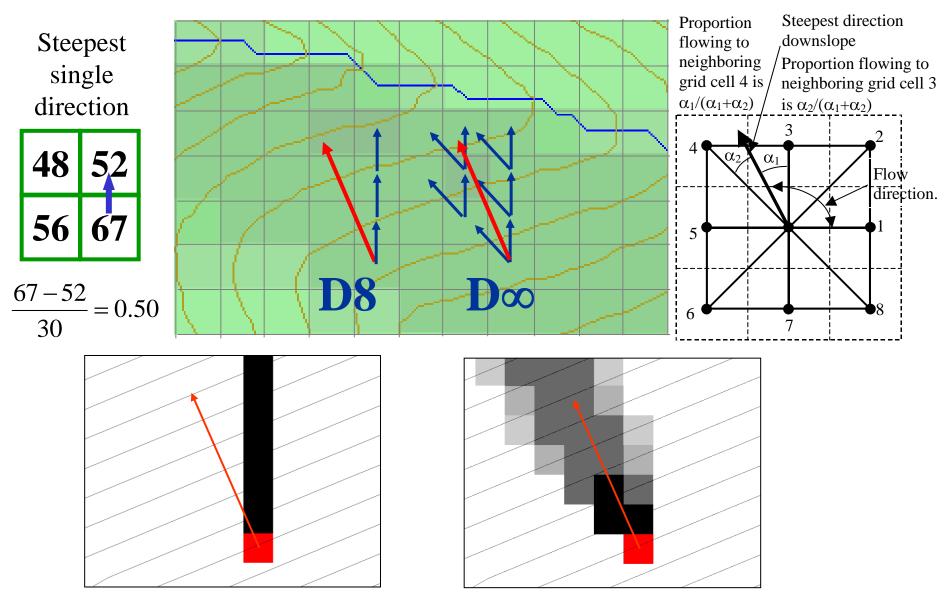


Grid Data Format Assumptions



- Input and output grids are uncompressed GeoTIFF
- Maximum size 4 GB
- GDAL Nodata tag preferred (if not present, a missing value is assumed)
- Grids are square $(\Delta x = \Delta y)$
- Grids have identical extents, cell size and spatial reference
- Spatial reference information is not used (no projection on the fly)

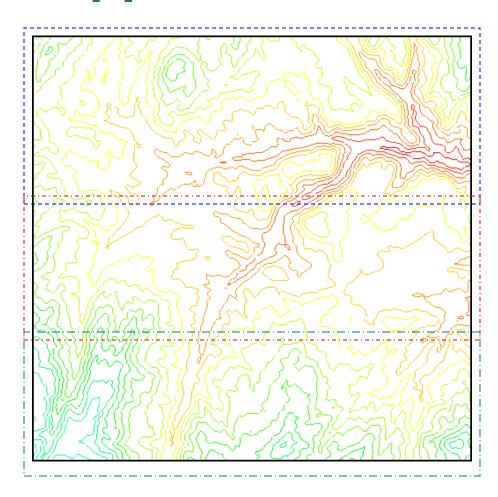
Representation of Flow Field



Tarboton, D. G., (1997), "A New Method for the Determination of Flow Directions and Contributing Areas in Grid Digital Elevation Models," Water Resources Research, 33(2): 309-319.)

Parallel Approach

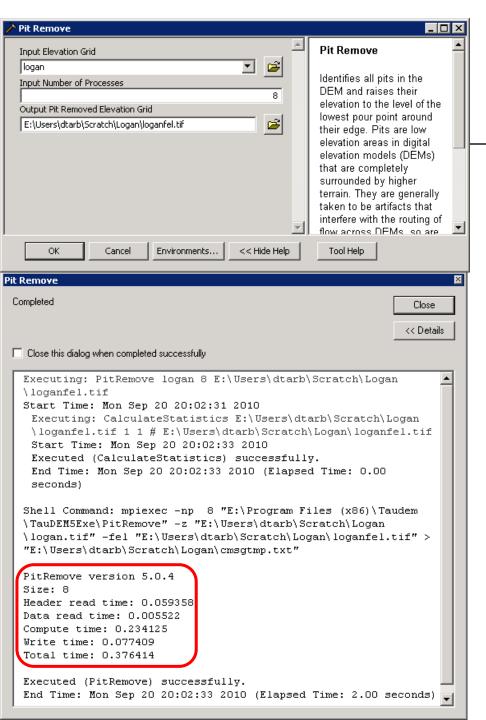
- MPI, distributed memory paradigm
- Row oriented slices
- Each process includes one buffer row on either side
- Each process does not change buffer row



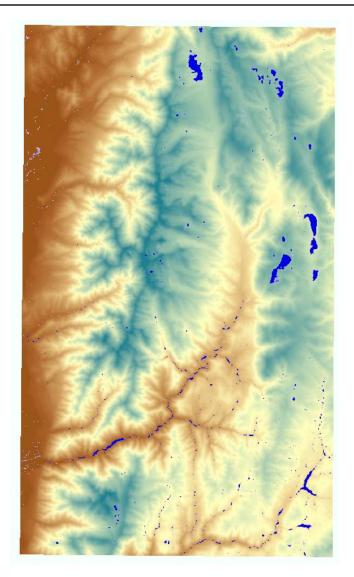
Illustrative Use Case: Delineation of channels and watersheds using a constant support area threshold

Steps

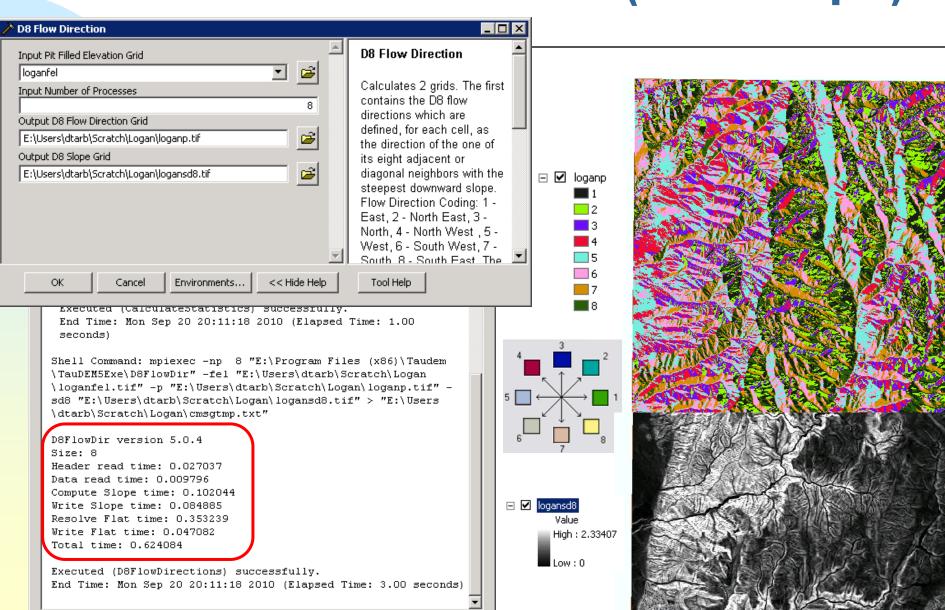
- Pit Remove
- D8 Flow Directions
- D8 Contributing Area
- Stream Definition by Threshold Threshold Stream Reach and Watershed Net.shp Stream **Output D8** Stream D8 Contributing Definition by Contributing Raster Threshold Watersheds Output D8 Stream Reach Stream D8 Flow Pit Remove Flow Input DEM Filled DEM and Watershed Order Direction Direction Coord.txt **Output D8** Slope Grid Tree.txt



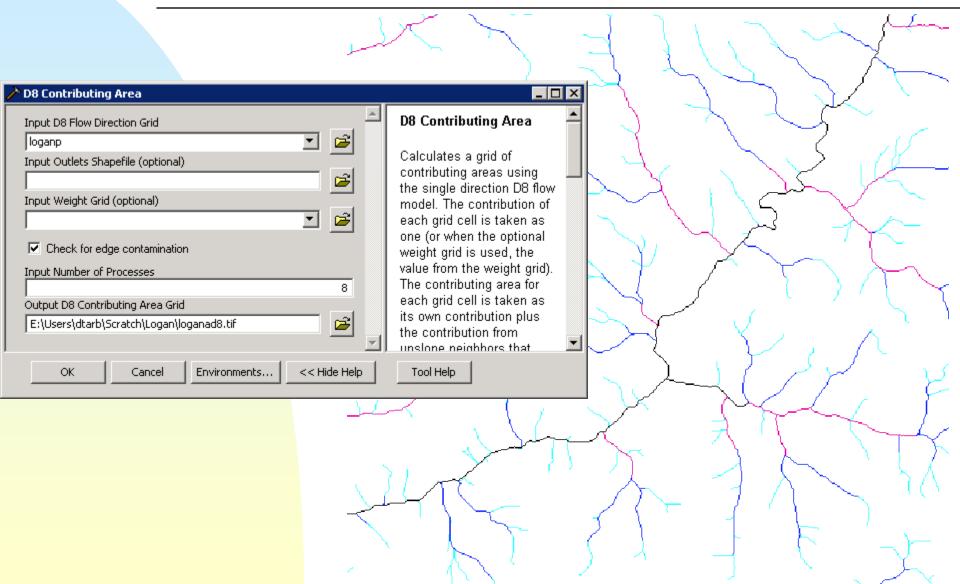
Pit Remove



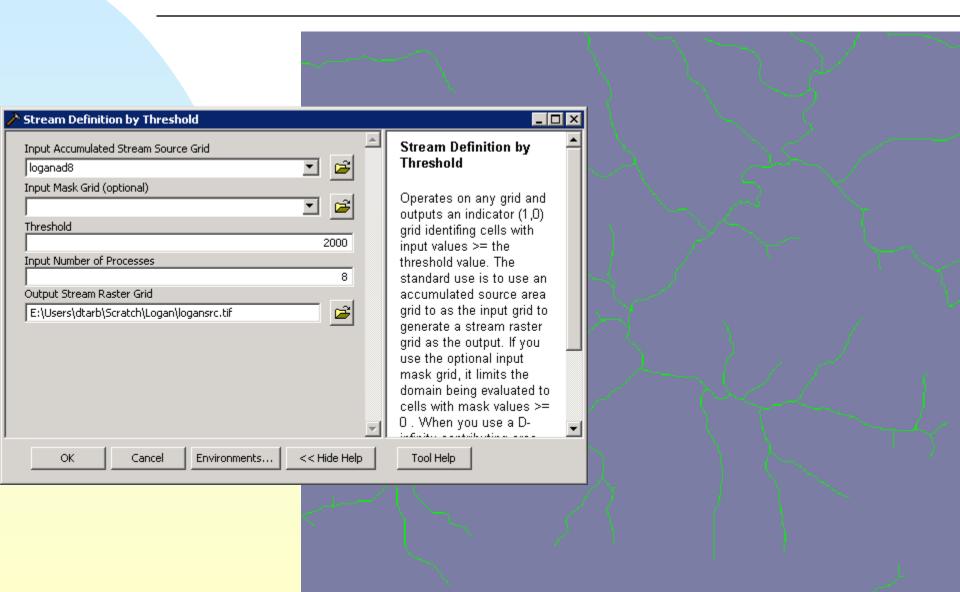
D8 Flow Direction (and Slope)



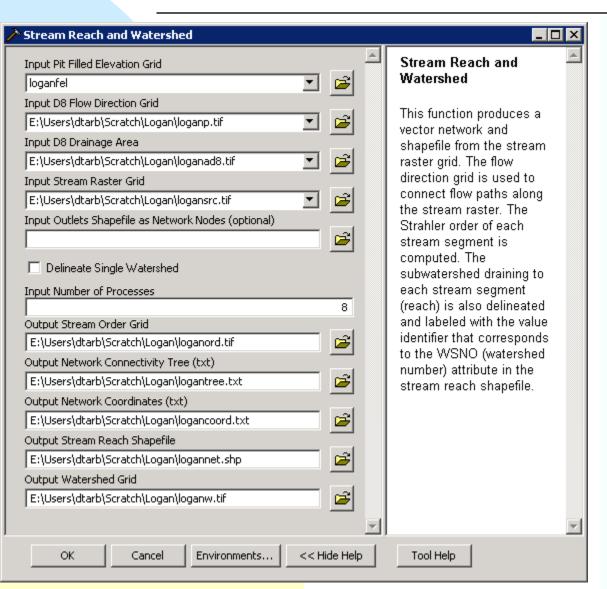
D8 Contributing Area

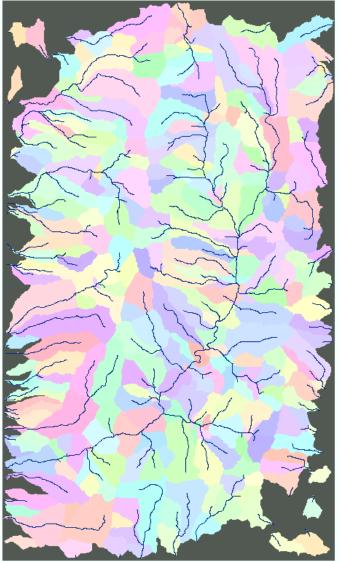


Stream Definition by Threshold

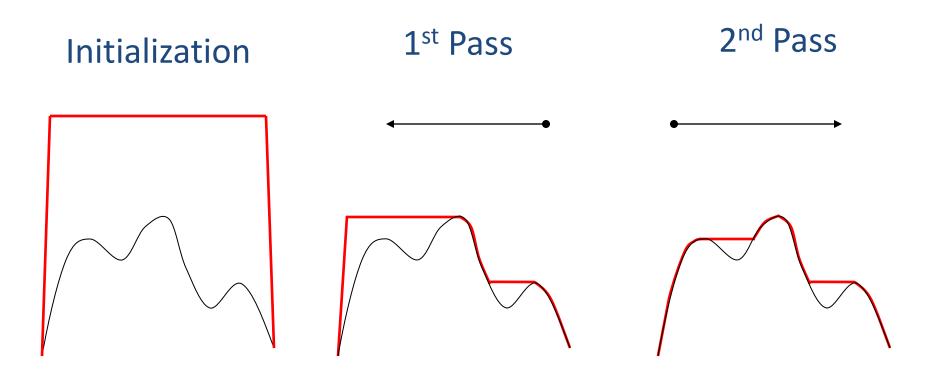


Stream Reach and Watershed



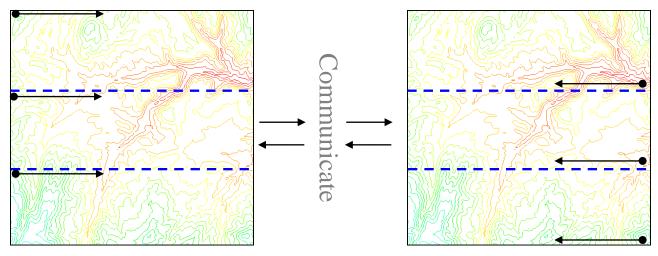


Some Algorithm Details Pit Removal: Planchon Fill Algorithm



Planchon, O., and F. Darboux (2001), A fast, simple and versatile algorithm to fill the depressions of digital elevation models, *Catena*(46), 159-176.

Parallel Scheme



```
Initialize(Z,F)
Do
      for all grid cells i
            if Z(i) > n
               F(i) \leftarrow Z(i)
            Else
               F(i) \leftarrow n
              i on stack for next pass
      endfor
      Send(topRow, rank-1)
      Send(bottomRow, rank+1)
      Recv(rowBelow, rank+1)
      Recv(rowAbove, rank-1)
Until F is not modified
```

Z denotes the original elevation.

F denotes the pit filled elevation.

n denotes lowest neighboring elevation i denotes the cell being evaluated

Iterate only over stack of changeable cells

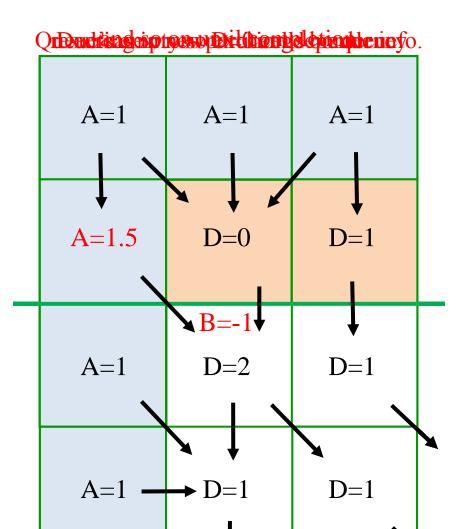
Parallelization of Contributing Area/Flow Algebra

1. Dependency grid

2. Flow algebra function

```
Executed by every process with D and Q initialized from FindDependencies. 

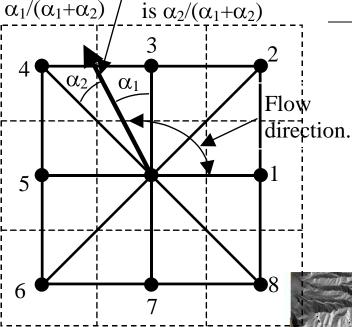
FlowAlgebra(P,Q,D,\theta,\gamma) while Q isn't empty get i from Q \underline{\theta}_i = FA(\underline{\gamma}_i, \underline{P}_{ki}, \underline{\theta}_k, \underline{\gamma}_k) for each downslope neighbor n of i if P_{in} > 0 D(n) = D(n) - 1 if D(n) = 0 add n to Q next n end while swap process buffers and repeat
```



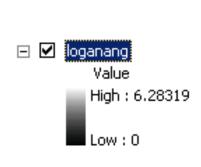
Proportion flowing to neighboring grid cell 4 is $\alpha_1/(\alpha_1+\alpha_2)$

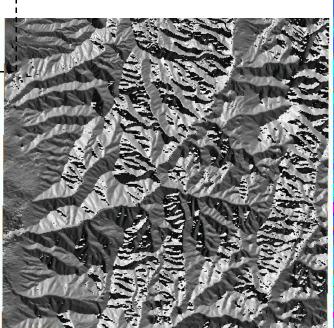
Steepest direction downslope Proportion flowing to neighboring grid cell 3

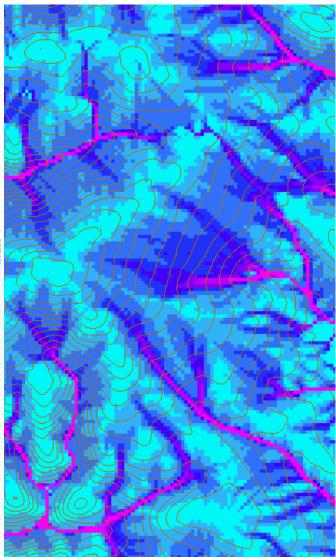
Illustration of some other functions



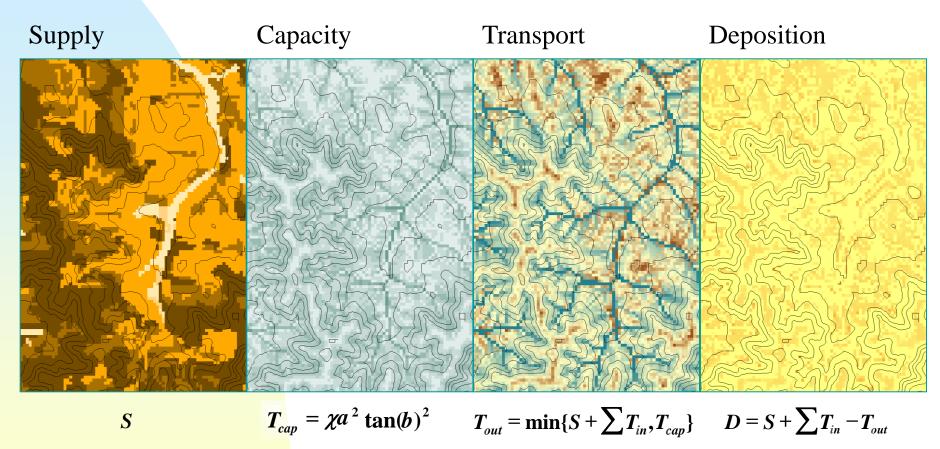
D-Infinity
Slope and
Contributing
Area







Transport limited accumulation



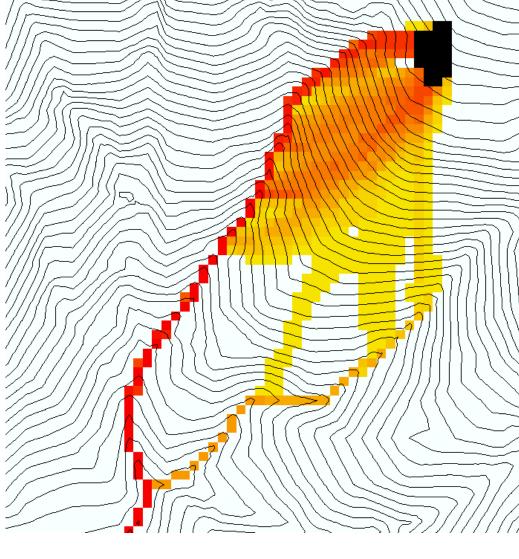
Useful for modeling erosion and sediment delivery, the spatial dependence of sediment delivery ratio and contaminant that adheres to sediment

Decaying Accumulation

A decayed accumulation operator DA[.] takes as input a mass loading field m(x) expressed at each grid location as m(i, j) that is assumed to move with the flow field but is subject to first order decay in moving from cell to cell. The output is the accumulated mass at each location DA(x). The accumulation of m at each grid cell can be numerically evaluated

$$DA[m(x)] = DA(i, j) = m(i, j)\Delta^{2} + \sum_{\substack{k \text{ contributing neighbors}}} p_{k}d(i_{k}, j_{k})DA(i_{k}, j_{k})$$

Here d(x) = d(i,j) is a decay multiplier giving the fractional (first order) reduction in mass in moving from grid cell x to the next downslope cell. If travel (or residence) times t(x) associated with flow between cells are available d(x) may be evaluated as $exp(-\lambda t(x))$ where λ is a first order decay parameter.

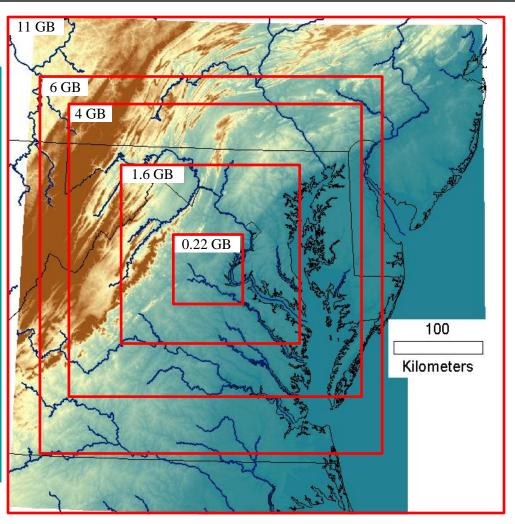


Useful for a tracking contaminant or compound subject to decay or attenuation

Capabilities Summary

Capability to run larger problems

Capability to run larger problems							
			Grid size				
		Processors used	Theoretcal limit	Largest run			
2008	TauDEM 4	1	0.22 GB	0.22 GB			
Sept 2009	Partial implement- ation	8	4 GB	1.6 GB			
June 2010	TauDEM 5	8	4 GB	4 GB			
Sept 2010	Multifile on 48 GB RAM PC	4	Hardware limits	6 GB			
Sept 2010	Multifile on cluster with 128 GB RAM	128	Hardware limits	11 GB			

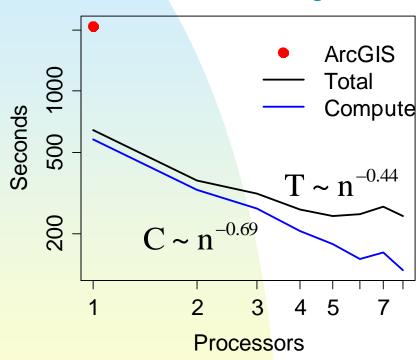


Single file size limit 4GB

At 10 m grid cell size

Improved runtime efficiency

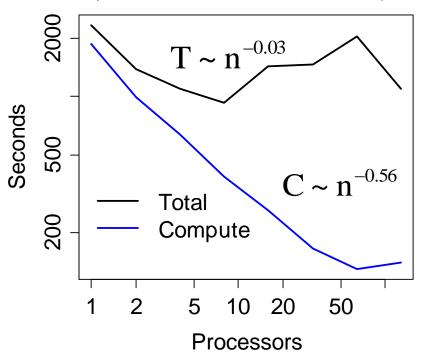
Parallel Pit Remove timing for NEDB test dataset (14849 x 27174 cells \approx 1.6 GB).



8 processor PC

Dual quad-core Xeon E5405 2.0GHz PC with 16GB

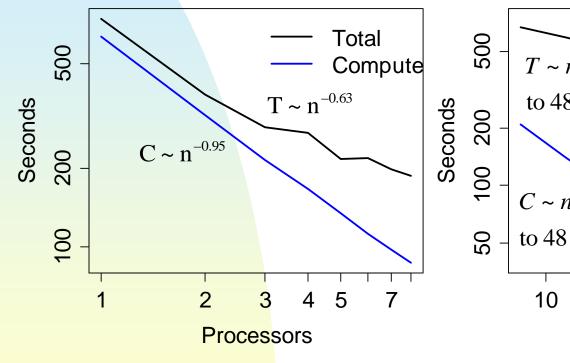
RAM



128 processor cluster
16 diskless Dell SC1435 compute nodes, each with 2.0GHz dual quad-core AMD Opteron 2350 processors with 8GB RAM

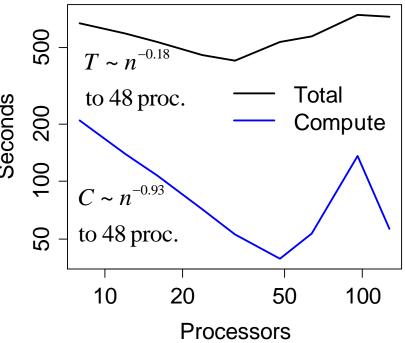
Improved runtime efficiency

Parallel D-Infinity Contributing Area Timing for Boise River dataset (24856 x 24000 cells ~ 2.4 GB)



8 processor PC

Dual quad-core Xeon E5405 2.0GHz PC with 16GB RAM



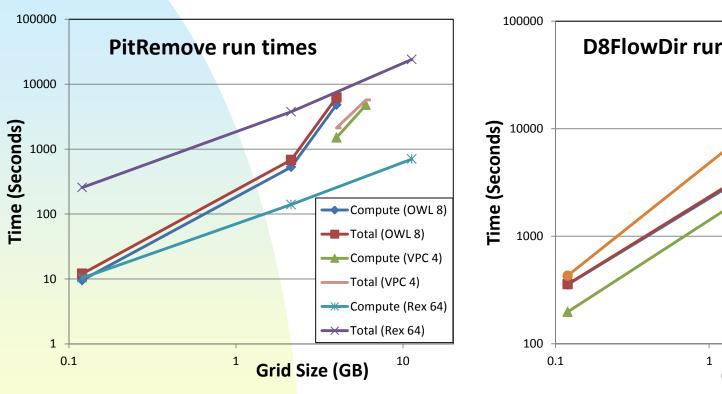
128 processor cluster
16 diskless Dell SC1435 compute nodes, each with 2.0GHz
dual quad-core AMD Opteron 2350 processors with 8GB RAM

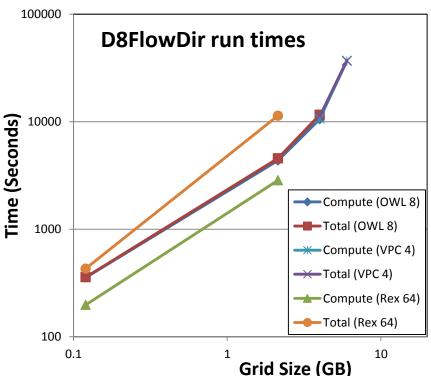
Scaling of run times to large grids

			Number of	PitRemove		D8FlowDir	
Dataset	Size	Hardware	Processors	(run time seconds)		(run time seconds)	
	(GB)			Compute	Total	Compute	Total
GSL100	0.12	Owl (PC)	8	10	12	356	358
GSL100	0.12	Rex (Cluster)	8	28	360	1075	1323
GSL100	0.12	Rex (Cluster)	64	10	256	198	430
GSL100	0.12	Mac	8	20	20	803	806
YellowStone	2.14	Owl (PC)	8	529	681	4363	4571
YellowStone	2.14	Rex (Cluster)	64	140	3759	2855	11385
Boise River	4	Owl (PC)	8	4818	6225	10558	11599
Boise River	4	Virtual (PC)	4	1502	2120	10658	11191
Bear/Jordan/Weber	6	Virtual (PC)	4	4780	5695	36569	37098
Chesapeake	11.3	Rex (Cluster)	64	702	24045		

- 1. Owl is an 8 core PC (Dual quad-core Xeon E5405 2.0GHz) with 16GB RAM
- 2. Rex is a 128 core cluster of 16 diskless Dell SC1435 compute nodes, each with 2.0GHz dual quad-core AMD Opteron 2350 processors with 8GB RAM
 - 3. Virtual is a virtual PC resourced with 48 GB RAM and 4 Intel Xeon E5450 3 GHz processors
 - 4. Mac is an 8 core (Dual quad-core Intel Xeon E5620 2.26 GHz) with 16GB RAM

Scaling of run times to large grids





- Owl is an 8 core PC (Dual quad-core Xeon E5405 2.0GHz) with 16GB RAM
- Rex is a 128 core cluster of 16 diskless Dell SC1435 compute nodes, each with 2.0GHz dual quad-core AMD Opteron 2350 processors with 8GB RAM
 - Virtual is a virtual PC resourced with 48 GB RAM and 4 Intel Xeon E5450 3 GHz processors

Summary and Conclusions

- Parallelization speeds up processing and partitioned processing reduces size limitations
- Parallel logic developed for general recursive flow accumulation methodology (flow algebra)
- Documented ArcGIS Toolbox Graphical User Interface
- 32 and 64 bit versions (but 32 bit version limited by inherent 32 bit operating system memory limitations)
- PC, Mac and Linux/Unix capability
- Capability to process large grids efficiently increased from 0.22 GB upper limit pre-project to where < 4GB grids can be processed in the ArcGIS Toolbox version on a PC within a day and up to 11 GB has been processed on a distributed cluster (a 50 fold size increase)

Limitations and Dependencies

- Uses MPICH2 library from Argonne National Laboratory http://www.mcs.anl.gov/research/projects/mpich2/
- TIFF (GeoTIFF) 4 GB file size (for single file version)
- Capability to use multiple files to cover domain not yet on web site
- Processor memory
- Toolbox user interface only for ArcGIS 9.3.1

Are there any questions?

