



An Extreme-Scale Computational Approach to Redistricting Optimization

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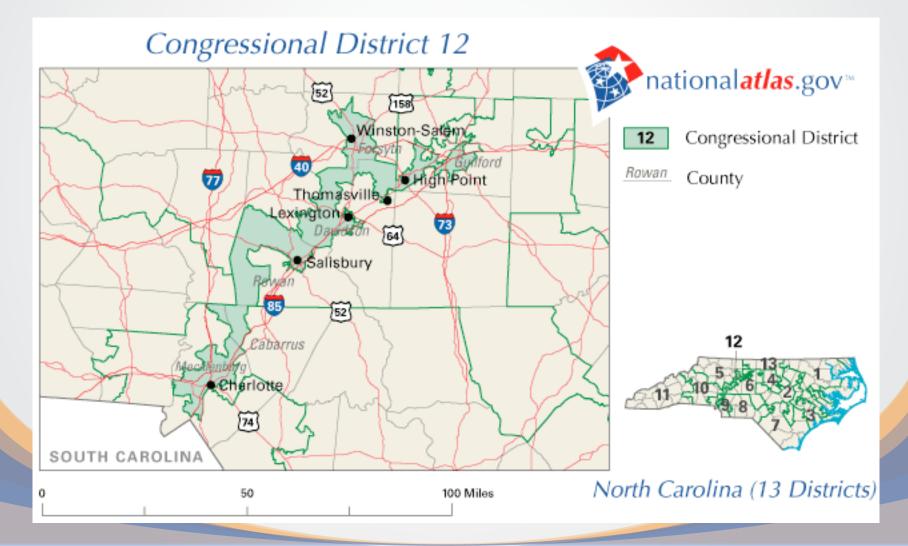
CyberGIS Center for Advanced Digital and Spatial Studies (CyberGIS Center) CyberInfrastructure and Geospatial Information Laboratory (CIGI Laboratory) Department of Geography and Geographic Information Science Department of Computer Science Department of Urban and Regional Planning Graduate School of Library and Information Science National Center for Supercomputing Applications (NCSA) University of Illinois at Urbana-Champaign

> NCSA Blue Waters Symposium for Petascale Science and Beyond May 12-15, 2014





Gerrymandering







Zoning Analytics

- Partitioning a group of indivisible geographic units into a smaller number of districts
 - Objectives and constraints
 - Contiguity, competitiveness, equal population, preservation of communities of interest and local political subdivisions, minority districts
 - Computational complexity
 - Number of possible solutions
 - Stirling number of the second kind: S(n, k)
 - Example: $S(55, 6) = 8.7 \times 10^{39}$
 - Computationally intractable
 - NP-hard



Cedar Falls Waterloo



Warren

Toledo

Findlay

Marion

33

Bucy

Hur

Morehead

Malaga

Complex Spatial Decision Making

Manchester

Dubuque



Apps - Visualization Community - Support - About - Feedback Home

Fort Dodge

Logout test's Profile

Renosti

Beloit

BioScope Wizard

Introduction

Overview

This website is to provide a prototype of CyberGIS-enabled web-based decision support platform for optimizing biomass feedstock provisions. The platform includes four major components: 1) BioScope optimization model; 2) a CyberGIS middleware that includes GISolve Open Service APIs and web server PHP functions; 3) a high performance cyberinfrastructure for optimization problem solving; and 4) an interactive web-based user interface.

The BioScope model optimizes a three-stage biomass provision system that includes farms, centralized storage and preprocessing (CSP) facilities, and biorefineries. The model can minimize biomass provision costs by optimizing the numbers, locations, and capacities of facilities and biomass transportation patterns.

How does it work

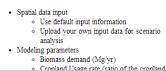
1. Job Submission

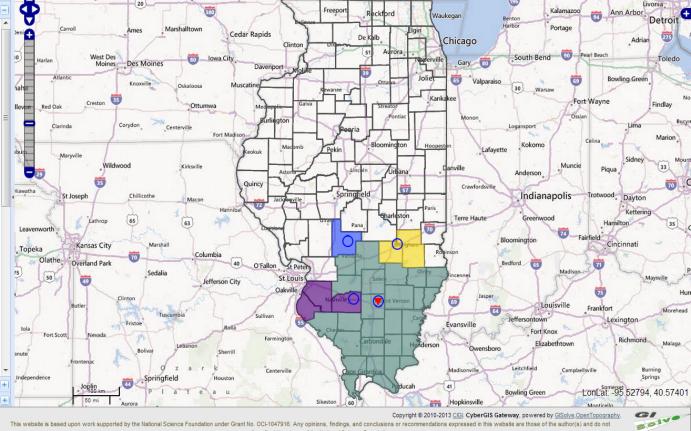
In the Job Panel tab, you can create your new job submission, check the submitted jobs and their associated status, and delete the submitted jobs. The panel will be connected to a database to view and manage your historical job submissions.

New job submission:

Job Panel

Layers





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Collaborative Work by Hao Hu, Tao Lin, Yan Liu, Luis F. Rodríguez, and Shaowen Wang





Exact Algorithms vs. Heuristics

- Exact algorithms
 - Guarantee to find an optimal solution
 - Methods
 - Branch & bound
 - Branch & cut
 - Etc.
 - Computationally intractable

Heuristics

- Algorithms that produce optimal or near-optimal solutions within a reasonable amount of time
- Population-based heuristics
 - Genetic algorithm
 - Swarm
 - Ant colony
 - Etc.





Genetic Algorithm (GA)

• Principles

- Evolutionary process
 - "survival of the fittest"
 - Iterative algorithm
- Solution population: a diverse set of initial solutions
- GA operators
 - Selection, crossover, mutation, replacement
- Stopping criteria
 - Solution quality
 - Time or the number of iterations

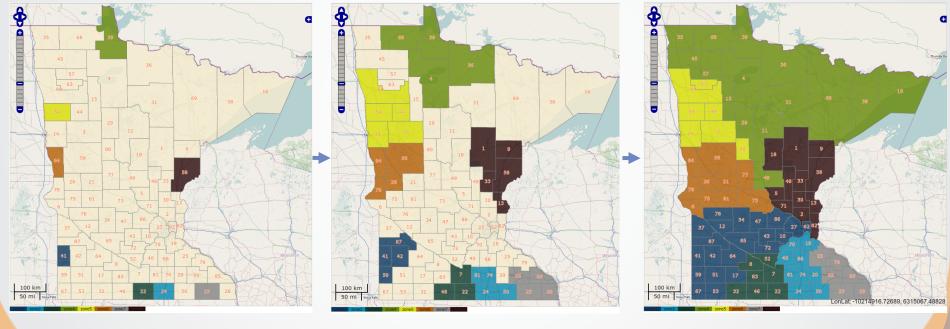
Spatial GA operators

- Solution generation
- Crossover
- Mutation





Spatial GA Operators – Feasible Solution Generation



Seeding

Expansion

Completion



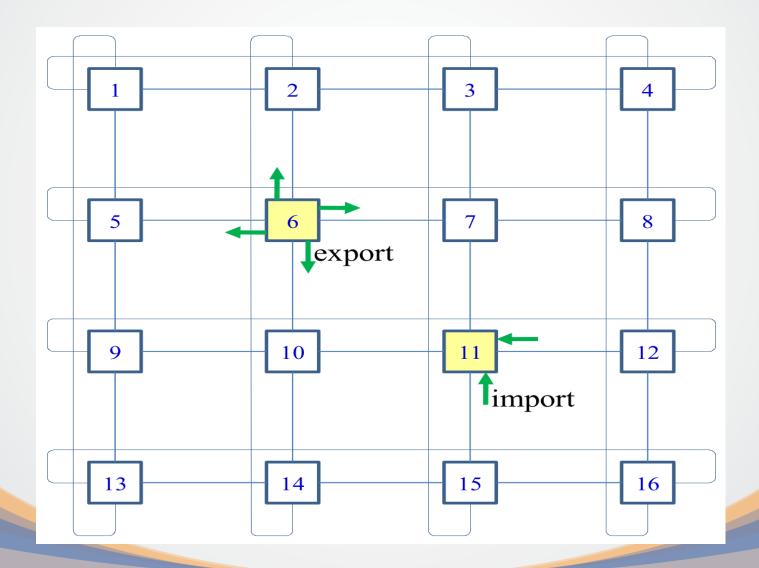


 Liu, Y.Y. and Wang, S. 2014. "A Scalable Parallel Genetic Algorithm for the Generalized Assignment Problem." *Parallel Computing*, <u>http://dx.doi.org/10.1016/j.parco.2014.04.008</u>





Parallel GA







Challenges for PGA

- Scalability to a large number of processor cores
 - Solution migration
 - Migration interval, rate
 - Traditional implementation: global barrier
 - Communication cost increases significantly when using a large number of cores

• Synchronous vs. Asynchronous

- Goals of asynchronous migration
 - Mimic the natural behavior of GA
 - Increase the overlapping of computing and communication
- Breaking the global barrier
 - Asynchronous migration
 - Buffer handling is inevitable

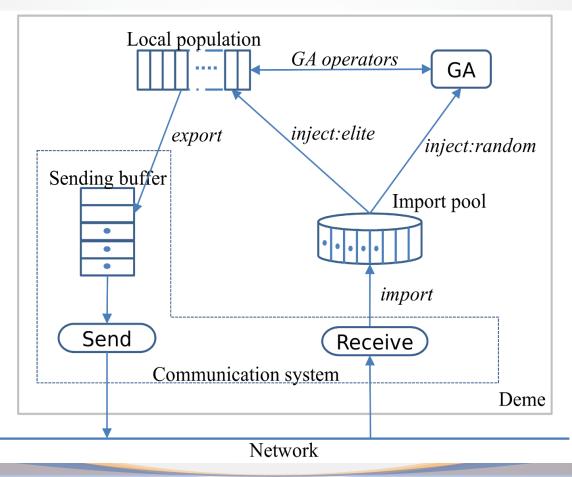




Asynchronous Migration

• Operators

o export, import, inject







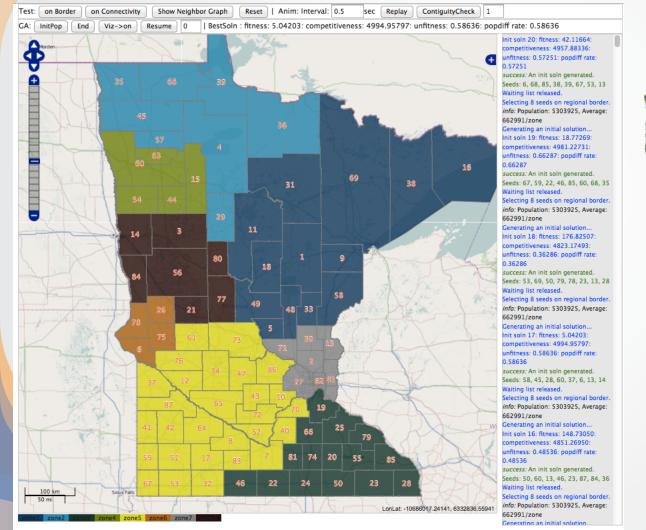
PGA Parameters

Parameters	Settings
Population size per deme	100
Initial population generation	Random with feasibility improve-
	ment or constraint-based im-
	provement
Selection	Binary tournament
Crossover	1-point. Probability: 0.8
Mutation	1-item mutation. Probability:
	0.2
Replacement	Replacing the unfittest or worst
Elitism	Yes
Stopping rules	No solution improvement,
	bounded solution quality
	reached, or fixed number of
	iterations
Connectivity d	4
Migration rate r	2
Export interval M_{expt}	100
Import interval M_{impt}	50
Probability of holding	1/20 (the probability to export
	when no better solution found
	during a previous export interval)
Sending buffer size $K_{sendbuf}$	20 solutions. Actual memory
	requirement is $(20 \times n \times 4 +$
	<i>buffer_overhead</i>) bytes
Import pool size K_{impt}	80 solutions. Actual memory re-
	quirement is $(80 \times n \times 4)$





Case Study







Computational Experiments

Problem size

- 4140 voting districts
- Number of districts: 8

• Data

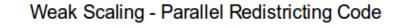
- Input: shape file; rook and queen neighborhood files
- Output: shape file; each core outputs one

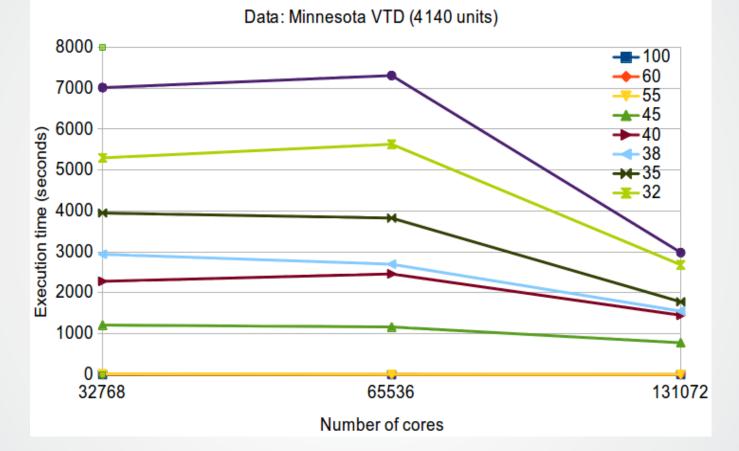
Number of cores

o 32768, 65536, 131072

CIGI







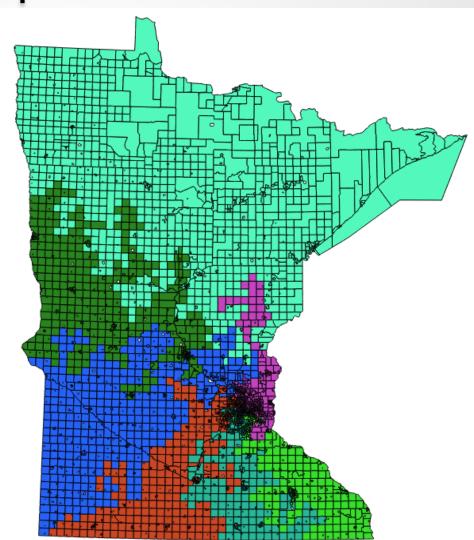
- Measurement: time taken to achieve different solution bounds (the smaller, the better)
- Using 131072 cores led to significant improvement





An Example Solution

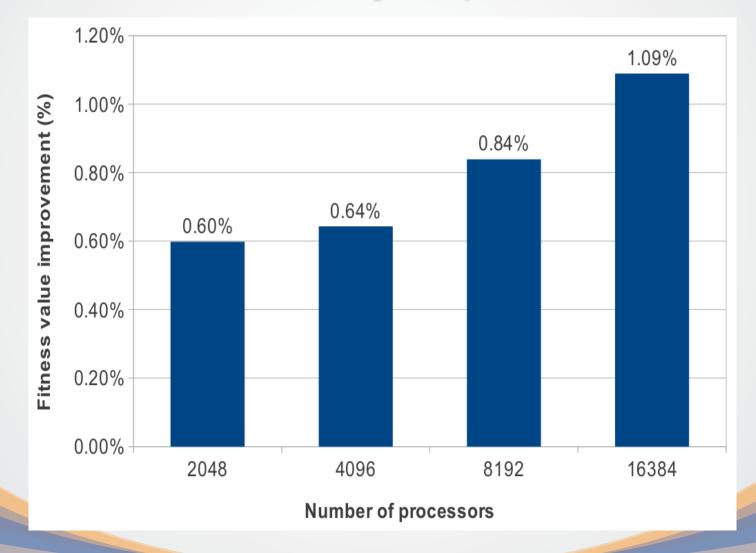
- Found by the run using 131072 cores
- Competitiveness as objective
 - R / (R + D)
 - o **0.499998**
- Constraint: population
 deviation







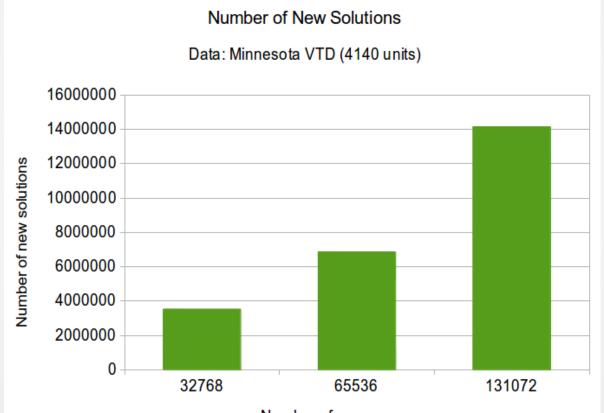
Solution Quality Improvement







Number of New Solutions

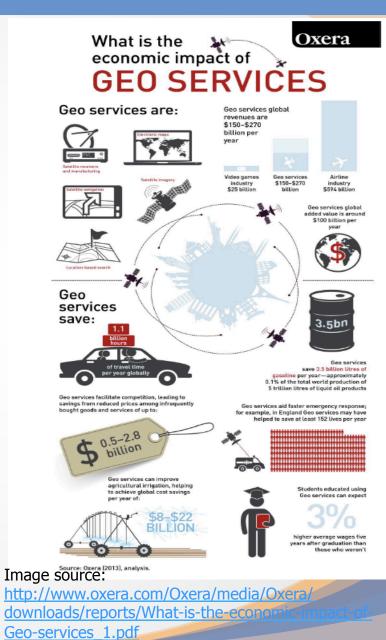


Number of cores



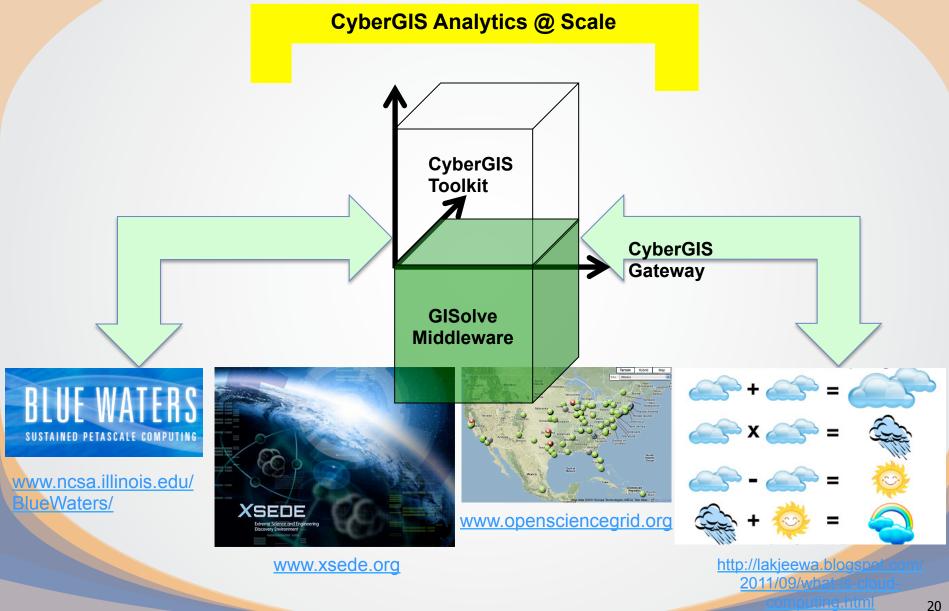


- What is GIS?
- Systems
- Science
- Services
- Society
- Synthesis
- Geo and spatial are special





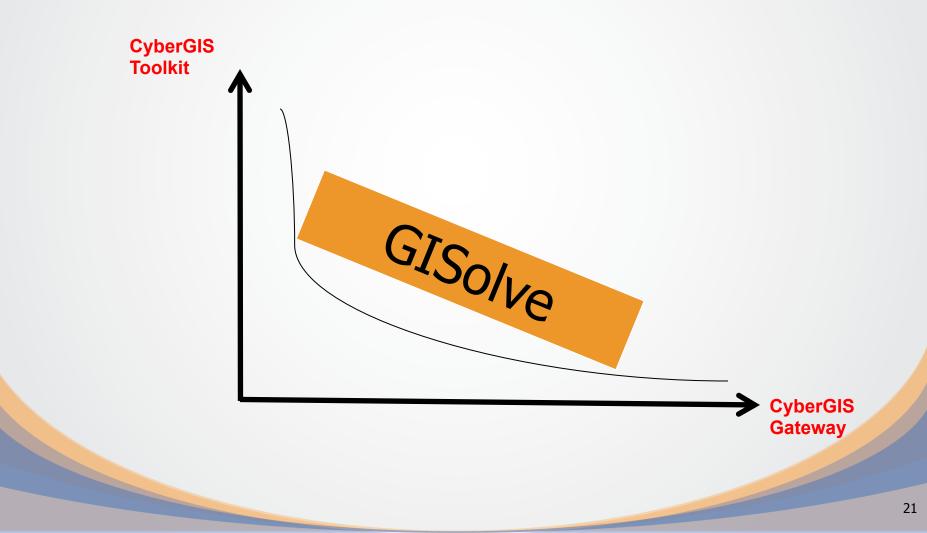








CyberGIS for What and Whom?







NSF SI2-SSI: CyberGIS Project \$4.43 million, Year: 2010-2015

Principal Investigator

– Shaowen Wang

Co-Principal Investigators

- Luc Anselin
- Budhendra Bhaduri
- Timothy Nyerges
- Nancy Wilkins-Diehr

Senior Personnel

- Michael Goodchild
- Sergio Rey
- Marc Snir
- E. Lynn Usery

Chair of the Science Advisory Committee

- Michael Goodchild

Project Manager

- Anand Padmanabhan

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- ASU: Wenwen Li and Rob Pahle
- ORNL: Ranga Raju Vatsavai
- SDSC: Choonhan Youn
- UIUC: Yan Liu and Anand Padmanabhan
- USGS: Michael Finn and David Mattli
- Graduate and undergraduate students

Industrial Partner: Esri

Steve Kopp and Dawn Wright



















CyberGIS Communities

Science and Technology Communities

- Advanced cyberinfrastructure
- Climate change impact assessment
- Emergency management
- Geographic information science
- Geography and spatial sciences
- Geosciences
- Social sciences
- Etc.

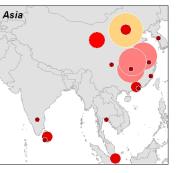
<u>User Communities</u>

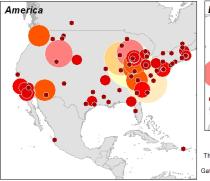
- Biologists
- Geographers
- Geoscientists
- Social scientists
- General public
- Broad GIS users
- Etc.

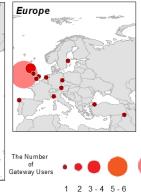


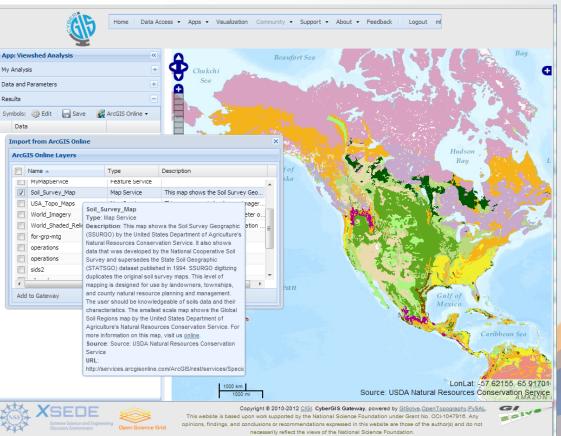
















CyberGIS Gateway and ArcGIS Online Integration

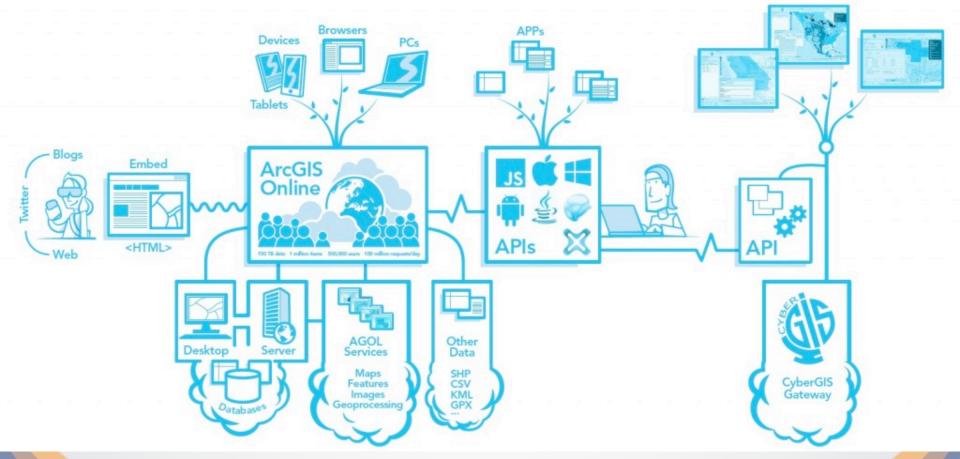
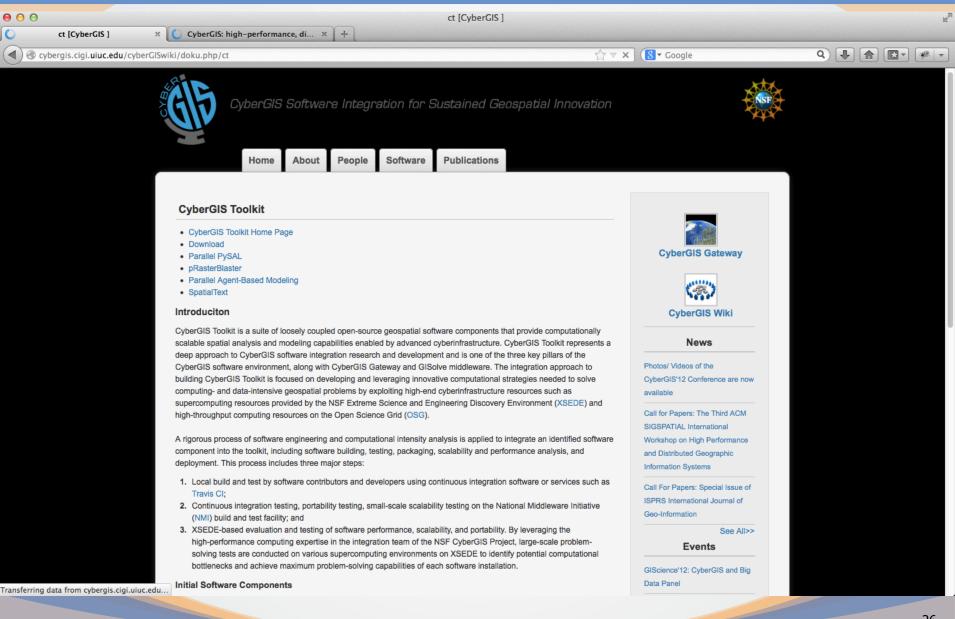


Image source: http://blogs.esri.com/esri/arcgis/2013/10/01/what-is-cybergis/



CyberInfrastructure and Geospatial Information Laboratory











- Curriculum and pedagogy
- Partnerships
- Open





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Thanks!

Comments / Questions?

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