Web and Grid Services from Pitt/CMU

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Serial and Parallel Applications

- ITR collaboration: University of Pittsburgh and Carnegie Mellon University.
- Astronomers, Computer Scientists and Statisticians
- Developing fast, usually tree-based, algorithms to reduce large volumes of data.
- Sample projects:
 - Source detection and cross match
 - Parallel Correlation Functions
 - Parallelized serial applications (through GridShell on the Teragrid)
 - Object classification (Morphologies and general classification)
 - Anomaly finding
 - Density estimation
 - Intersections in parameter space

WSExtractor: Source Detection

Wrapping existing services as webservices Accessible through client and webservices Prototype Sextractor (and interface to other services) Communication through SOAP messages and attachments Full configurable through webservice interface Outputs VOTables Dynamically cross matches with openskyquery.net Returns cross matched VOTables Plots through Aladin and VOPlot Wsext





Welcome to the homepage of WSextractor

There are just six steps to getting your source catalog back. If you are interested in testing out this service, here is a test file that works.

Step 1: Specify the file you want to upload

/home/simon/images/756/ Browse...

Step 2: Select the catalog you would like to crossmatch with.

ROSAT	٠
GALEX	1
DLS	
RC3	
SDSS	
SDSSDR2	Ŧ

Step 3: Submit your file for processing

Submit

Contact US-VO help desk with any problems



SExtractor Output Fields

Step 4: Select the output fields you would like in your catalog.

FLUXERR_AUTO MAG_AUTO MAGERR_AUTO FLUX_BEST FLUXERR_BEST MAG_BEST MAGERR_BEST KRON_RADIUS

Output Fields from SDSSDR2

Step 5: Choose the columns you would like included in your CrossMatched catalog



Step 6: Go do it.



Contact US-VO help desk with any problems



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Aladin sky atlas

<u>CDS</u> · <u>Simbad</u> · <u>VizieR</u> · <u>Aladin</u> · <u>Catalogues</u> · <u>Nomenclature</u> · <u>Biblio</u> · <u>Tutorial</u> · <u>Developer's corner</u>





Aladin sky atlas

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Issues

VOTable and Java
DataHandler type

 Working with C and .NET

Aladin as applet on server

 Eliminates multiple transfers of images

Concurrent Access – sessions with WS
Value validation

Grid Services

Harnessing parallel grid resources in the NVO data mining framework

N-pt correlation functions for SDSS data:

- 2-pt: hours
- 3-pt: weeks
- 4-pt: 100 years!

 With the NVO, computational requirements will be much more extreme.

There will be many more problems for which throughput can be substantially enhanced by parallel computers.

The challenge of Parallelism

Parallel programs are hard to write!

- Parallel codes (especially massively parallel ones) are used by a limited number of "boutique enterprises" with the resources to develop them.
- Scientific programming is a battle between run time and development time:
 - Development time must be less than run time!
 - Large development time means they must be reused again and again and again to make the development worth it (e.g. N-body hydrodynamic code).
- Even the "boutiques" find it extraodinarily difficult to conduct new queries in parallel.
 - For example, all of the U.Washington "N-Body Shop's" data analysis code is *still* serial.

A Scalable Parallel Analysis Framework

- Canned routines (e.g. "density calculator", "cluster finder", "2-pt correlation function") restrict inquery space.
- Bring the power of a distributed TeraScale computing grid to NVO users. Provide seamless scalability from single processor machines to TeraFlop platforms while not restricting inquery space.
- **Toolkit**:
 - Highly general
 - Highly flexible
 - Minimizes development time
 - Efficiently and scalably parallel
 - Optimized for common architectures (MPI, SHMEM, POSIX, etc)

Methodology

 Identify Critical Abstraction Sets and Critical Methods Sets from work done on serial algorithm research by existing ITR.

 Efficiently parallelize these abstraction and methods sets

Distribute in the form of a parallel toolkit.
Developing a fast serial algorithm is completely different than implementing that algorithm in parallel.

Parallel Npt

Developing an efficient & scalable parallel 2-, 3-, 4-point correlation function code.

- Development on Terascale Computing System
- Based on the parallel gravity code PKDGRAV:
 - Highly portable (MPI, POSIX Threads, SHMEM, & others)
 - Highly scalable





92% linear speedup on 512 PEs!

By Amdahl's Law, this means 0.017% of the code is actually serial.



Number of Processors

T3D Science Rate (for θ =0.7 Hexadecapole)

Parallel Npt performance

2-pt Correlation Function (2°)



So far, only 74% speedup on 128 PEs

Parallel Npt Performance

2-pt Correlation Function (2°)



Issues and Future directions

Problem is communication latencies

- N-body inter-node communication small (npt large)
- Minimize off processor communication
- Extend to npt (in progress)
- Adaptive learning for allocating resources and work load
- Interface with webservices
- Not another IRAF