### Knowledge and Cache Conscious Data Mining: Algorithms and Systems Support

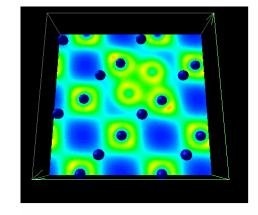
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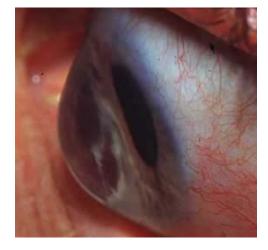
Joint work with A. Ghoting, G. Buehrer, M. Goyder, S. Tatikonda, T. Kurc and J. Saltz

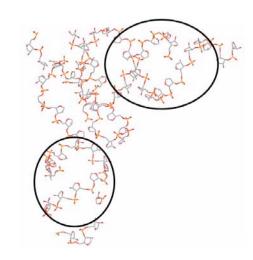
NSF NGS CNS0406386

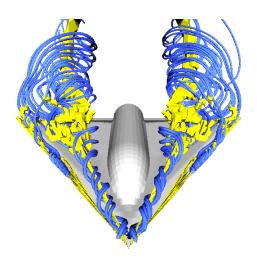
#### Motivation

- Advances in technology → huge data collections
  - Sensor networks
  - Massive legacy data in business or financial settings
  - Large scale simulations
  - Homeland security applications
  - Biomedical imaging
  - Bioinformatics

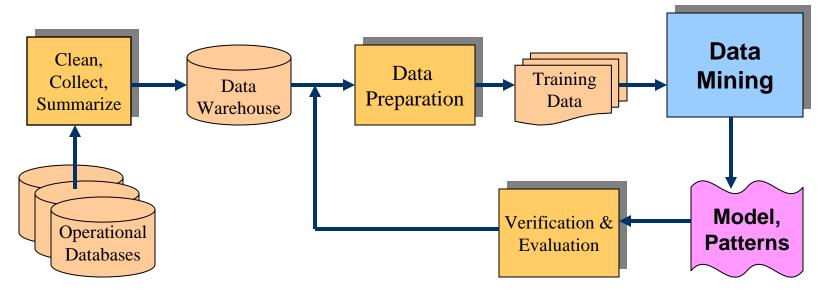








### **Knowledge Discovery Process**



- Knowledge discovery and data mining
  - Goal: extracting useful and actionable information (models, rules, patterns) from such massive data stores.
  - Multi-billion dollar industry
- Time consuming process **Compute and Data Intensive**
- Human-in-the-loop (verification) Interactive
- Impedence Mismatch!

## Next Generation Data Analysis

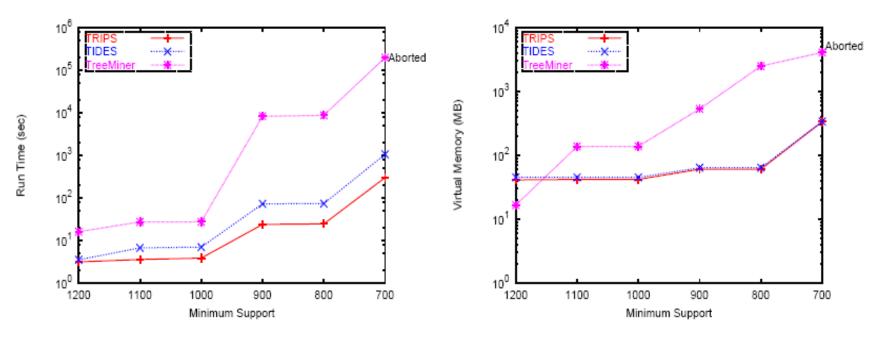
- Potential Solution: Leverage commodity high performance computing solutions to resolve this impedance mismatch.
  - Services oriented architecture
    - Scheduling Services
    - I/O and Data Services
    - Knowledge and Data Caching Services
  - Algorithms that can leverage such services
- Challenges
  - Highly irregular very data and application dependent
  - Often rely on large meta-data housed in dynamic data structures
    - Used to prune search space (pointer-based)
    - May be out-of-core!
  - Data is also often dynamic (time varying)

### Key Idea: Predicting and Exploiting Re-Use at Multiple Levels

- Cache Conscious Data Mining
  - At the algorithmic level
  - Improve spatial and temporal locality through careful understanding of (repetitive) access patterns
    - Leverage memory placement and data structure partitioning
  - Leverage architectural features (e.g. SMT) effectively to hide latency
    - Co-schedule threads that work on same data (different tasks)
- Knowledge Conscious Data Mining
  - At the methodological level
  - Leverage the iterative and interactive nature of process
  - Store and re-use previously computed knowledge to drive future requests
  - Effective in collaborative data analysis tasks but also across iterations of same algorithm

# **Cache Conscious Tree Mining**

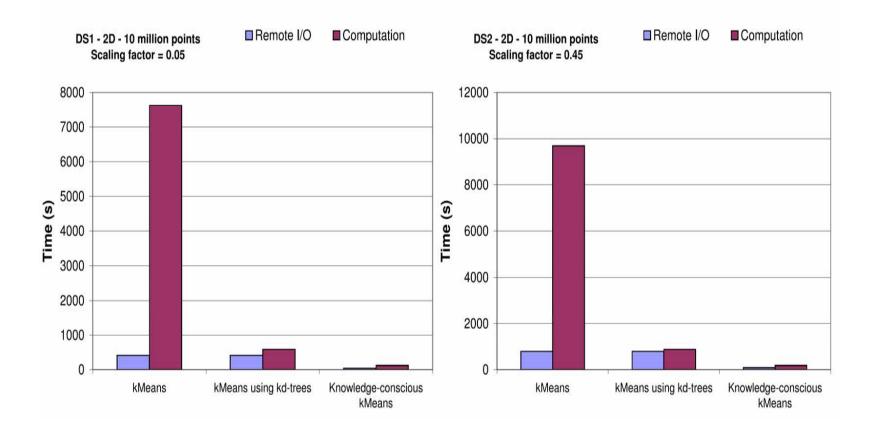
- Applications: bioinformatics, linguistics, program analysis, bug detection, web mining etc.
- Essentially converted pointer-based trees to sequences (housed in arrays) and operated on sequence space (bijection)
- Up to 355 speedup, using 40% less memory over state-of-art



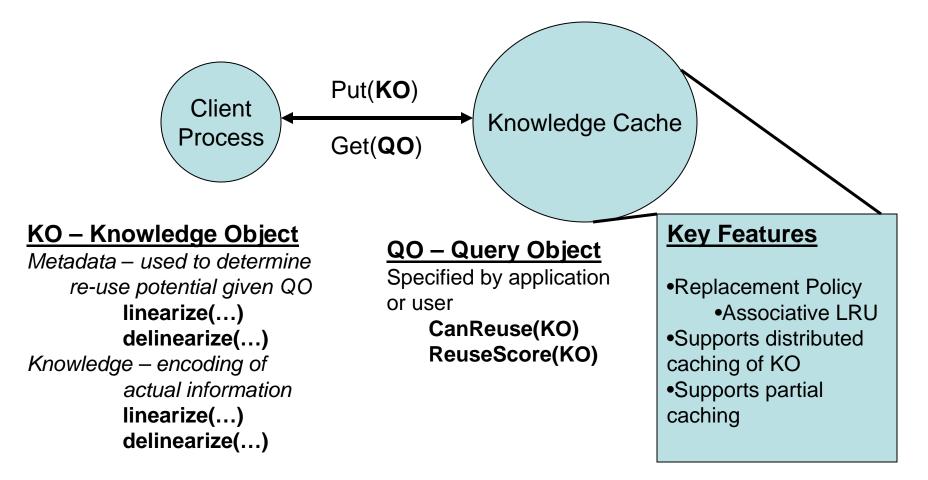
Itanium, 1.3 Ghz, 4GB Memory, CSLOGS – weblog dataset

## **Knowledge Conscious Clustering**

- Fundamental approach with a host of applications
- Single client system. Benefits include
  - Re-use knowledge across iterations of algorithm
  - Remote (Client-side) caching of KO
  - Up to10 fold improvement across the board



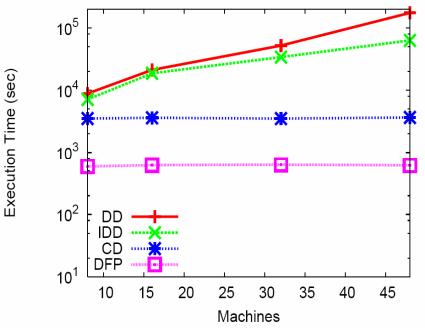
#### Knowledge Caching System Overview



Additionally we also support Data Objects (DO) – data subsets

## Summary and Current Status

- Designed Cache Conscious Solutions
  - Frequent Pattern Mining (VLDB Journal 06, KDD 06)
    - Tera-scale mining (PPOPP 2007)
  - Tree Mining (CIKM 2006)
    - Parallelization in progress
- Designed Knowledge
  Conscious Solutions
  - Clustering (PKDD 2006)
  - Frequent Pattern Mining & Classification (in progress)
- Systems Support
  - Design in place, implementation being debugged (in progress)



Weak Scalability on Frequent Pattern Mining

- Stripped down linearize/delinearize
  - 10 fold reduction in communication
- Efficient even when meta-data is out-of-core
- Order of magnitude over state-of-art

## Acknowledgements

- Other grant acknowledgements
  - NSF CAREER IIS-0347662
  - NSF RI CNS-0403342
  - IBM PhD Fellowship (A. Ghoting)
- For more information
  - http://www.cse.ohio-state.edu/~srini
  - http://dmrl.cse.ohio-state.edu