Ask and You Shall Receive: Empirically Evaluating Declarative Approaches to Finding Data in Unstructured Heaps

#### William F. Jones and Gregory M. Kapfhammer

Allegheny College http://www.cs.allegheny.edu/~gkapfham/

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#### Overview of the Presentation



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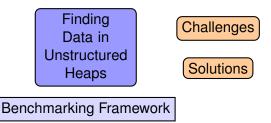
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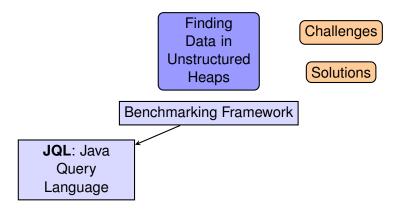
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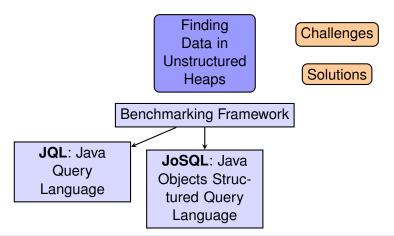
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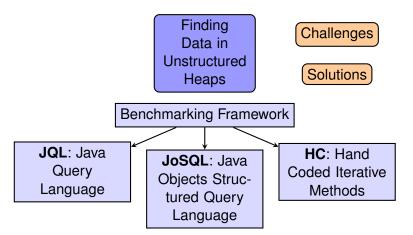
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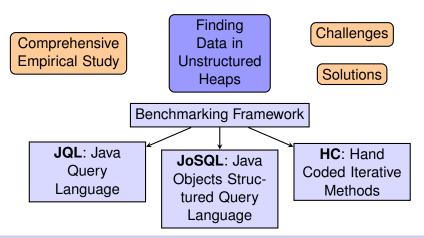
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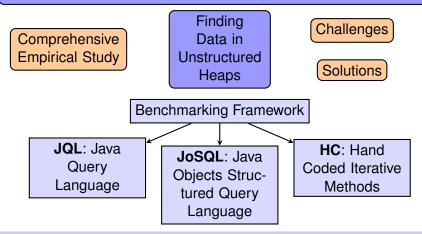
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## Overview of the Presentation

Experiments Reveal Trade-offs in Performance and Overall Viability



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#### Correctly and Efficiently Finding Objects in the Heap

The unstructured heap in a Java virtual machine stores objects that are connected in complex and unpredictable ways (Xu and Rountev, ICSE 2008)

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## Correctly and Efficiently Finding Objects in the Heap

The unstructured heap in a Java virtual machine stores objects that are connected in complex and unpredictable ways (Xu and Rountev, ICSE 2008)

When is an Object Allocated to the Heap?

LinkedList list = new LinkedList()

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### Correctly and Efficiently Finding Objects in the Heap

The unstructured heap in a Java virtual machine stores objects that are connected in complex and unpredictable ways (Xu and Rountev, ICSE 2008)

When is an Object Allocated to the Heap?

LinkedList list = new LinkedList()

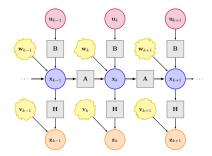
#### Let's Allocate Some Objects to the Heap!

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## Correctly and Efficiently Finding Objects in the Heap

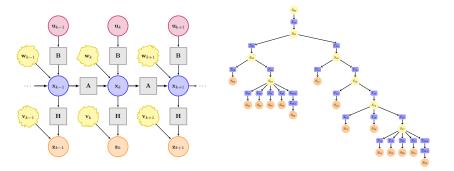


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## Correctly and Efficiently Finding Objects in the Heap



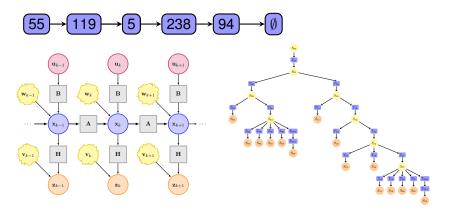
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# Correctly and Efficiently Finding Objects in the Heap



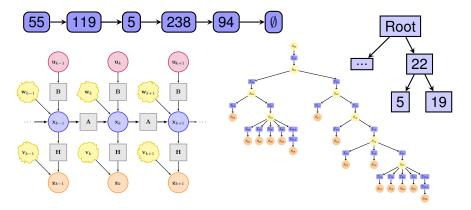
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# Correctly and Efficiently Finding Objects in the Heap



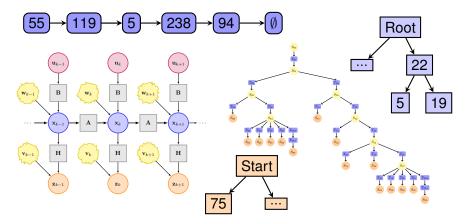
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# Correctly and Efficiently Finding Objects in the Heap



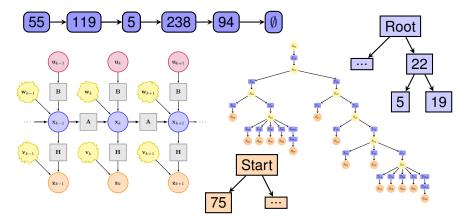
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# Correctly and Efficiently Finding Objects in the Heap



#### LinkedList Node(s) with Values Greater Than Those in the Trees

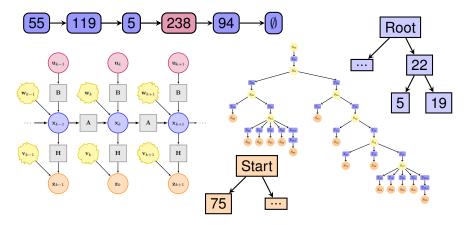
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# Correctly and Efficiently Finding Objects in the Heap



#### LinkedList Node(s) with Values Greater Than Those in the Trees

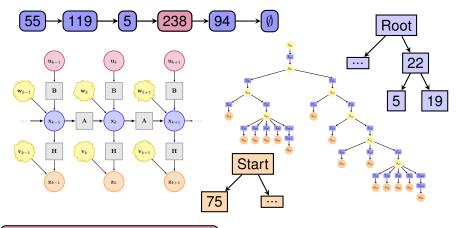
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#### Correctly and Efficiently Finding Objects in the Heap



#### (How Do We Find These Nodes?)

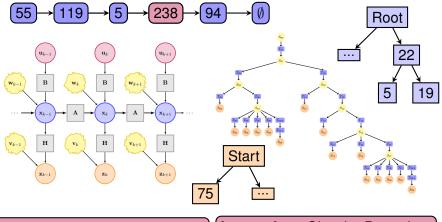
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# Correctly and Efficiently Finding Objects in the Heap



(How Do We Find These Nodes?)

Imperative - Give the Procedure

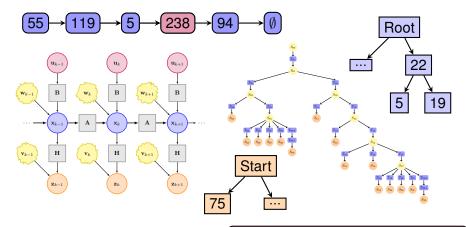
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# Correctly and Efficiently Finding Objects in the Heap



How Do We Find These Nodes?

Declarative - Give the Specification

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## **Object Query Languages and Bicycles**



#### Efficiency - Bicycle: Low wind resistance and time to destination

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## **Object Query Languages and Bicycles**



#### Efficiency - Query: Minimal space overhead and a low response time

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## **Object Query Languages and Bicycles**



#### Effectiveness - Bicycle: Transports all item(s) with no break downs

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## **Object Query Languages and Bicycles**



#### Effectiveness - Query: Always returns the correct result(s) to a query

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## **Object Query Languages and Bicycles**



#### Cost - Bicycle: Frame material(s) and components cause price to vary

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## **Object Query Languages and Bicycles**



#### Cost - Query: Must consider installation and development challenges

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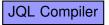
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# JQL: Java Query Language



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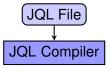
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# JQL: Java Query Language



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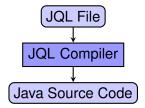
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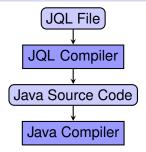
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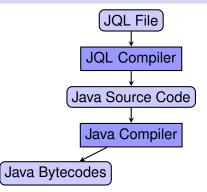
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## JQL: Java Query Language



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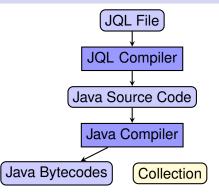
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## JQL: Java Query Language



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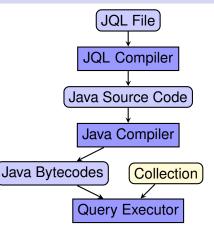
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## JQL: Java Query Language



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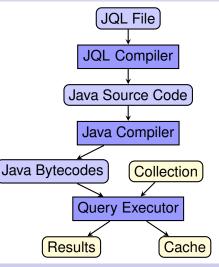
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### JQL: Java Query Language



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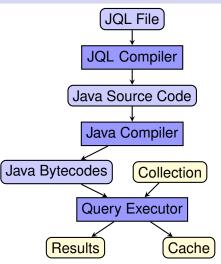
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### JQL: Java Query Language

#### **Features**

- Pre-compilation
- AOP with AspectJ
- Method Queries
- Caching
- Optimizations



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### JQL: Java Query Language

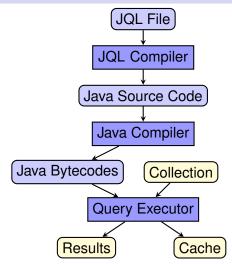
#### Features

- Pre-compilation
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#### References

• (Willis et al. ECOOP 2006)

 (Willis et al. OOPSLA 2008)



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# JoSQL: Java Objects Structured Query Language

### Parse SQL

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## JoSQL: Java Objects Structured Query Language

SQL String Query Object Parse SQL

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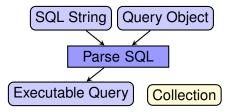
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## JoSQL: Java Objects Structured Query Language



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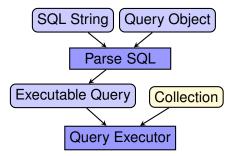
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## JoSQL: Java Objects Structured Query Language



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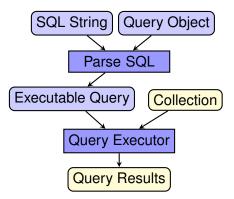
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## JoSQL: Java Objects Structured Query Language



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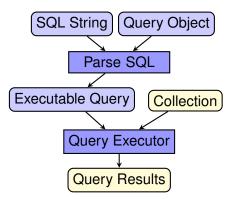
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# JoSQL: Java Objects Structured Query Language

#### Features

- SQL Statements
- String Parsing
- Java Reflection
- Query Facilities



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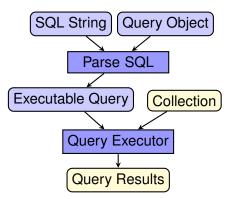
# JoSQL: Java Objects Structured Query Language

#### Features

- SQL Statements
- String Parsing
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- Query Facilities

#### Reference

http://josql.sf.net



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### Comparison of Data Finding Methods

As the number of collections and objects increases, imperative programming may lead to applications that are complicated, error-prone, and hard to maintain (Xu and Rountev, ICSE 2008)

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### Comparison of Data Finding Methods

As the number of collections and objects increases, imperative programming may lead to applications that are complicated, error-prone, and hard to maintain (Xu and Rountev, ICSE 2008)

JQL: Compile-Time

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### Comparison of Data Finding Methods

As the number of collections and objects increases, imperative programming may lead to applications that are complicated, error-prone, and hard to maintain (Xu and Rountev, ICSE 2008)

JQL: Compile-Time

JoSQL: Run-Time

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## Comparison of Data Finding Methods

As the number of collections and objects increases, imperative programming may lead to applications that are complicated, error-prone, and hard to maintain (Xu and Rountev, ICSE 2008)

JQL: Compile-Time

JoSQL: Run-Time

Performance Trade-Offs?

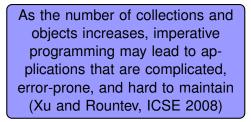
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### Comparison of Data Finding Methods



JQL: Compile-Time

JoSQL: Run-Time

Performance Trade-Offs?

Effectiveness Concerns?

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# Comparison of Data Finding Methods

Benchmarking Framework Helps to Answer These Questions

As the number of collections and objects increases, imperative programming may lead to applications that are complicated, error-prone, and hard to maintain (Xu and Rountev, ICSE 2008)

JQL: Compile-Time

JoSQL: Run-Time

Performance Trade-Offs?

Effectiveness Concerns?

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# Benchmarking Framework to Evaluate Query Methods

Random Collection Generator

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# Benchmarking Framework to Evaluate Query Methods

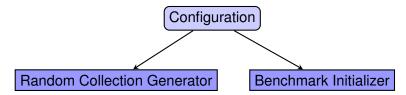
Random Collection Generator

Benchmark Initializer

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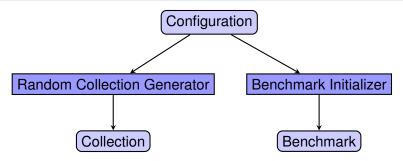
# Benchmarking Framework to Evaluate Query Methods



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# Benchmarking Framework to Evaluate Query Methods

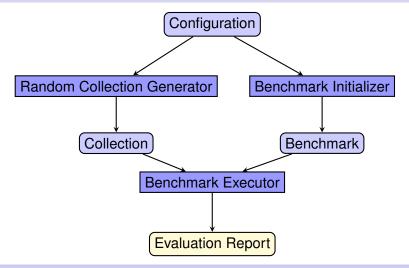


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# Benchmarking Framework to Evaluate Query Methods



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## Configuration of the Benchmarking Framework

Possible Configurations









### Explored a wide variety of benchmark configurations

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# Configuration of the Benchmarking Framework

Possible Configurations









### What operations do we run to evaluate the query methods?

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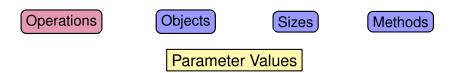
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## Configuration of the Benchmarking Framework

Possible Configurations



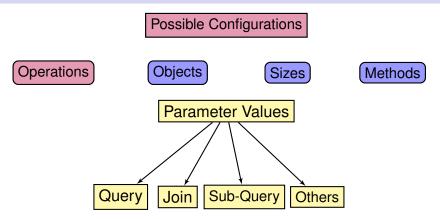
### What operations do we run to evaluate the query methods?

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# Configuration of the Benchmarking Framework



### What operations do we run to evaluate the query methods?

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# Configuration of the Benchmarking Framework

Possible Configurations









### What objects will we allocate to the JVM's heap?

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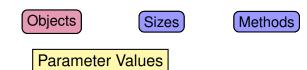
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## Configuration of the Benchmarking Framework

Possible Configurations





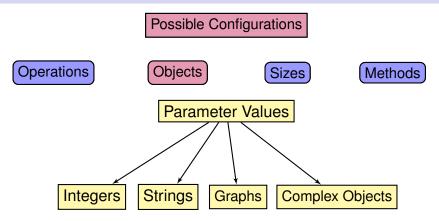
#### What objects will we allocate to the JVM's heap?

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# Configuration of the Benchmarking Framework



#### What objects will we allocate to the JVM's heap?

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# Configuration of the Benchmarking Framework

Possible Configurations









### How big should we make the objects and the collections?

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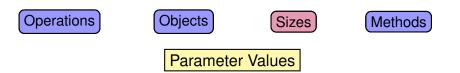
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## Configuration of the Benchmarking Framework

Possible Configurations



### How big should we make the objects and the collections?

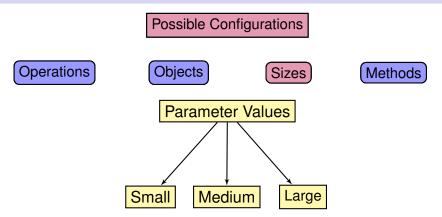
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## Configuration of the Benchmarking Framework



### How big should we make the objects and the collections?

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# Configuration of the Benchmarking Framework

Possible Configurations









### Which methods should be part of the framework?

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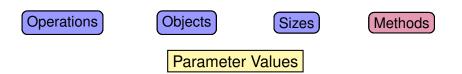
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## Configuration of the Benchmarking Framework

Possible Configurations



### Which methods should be part of the framework?

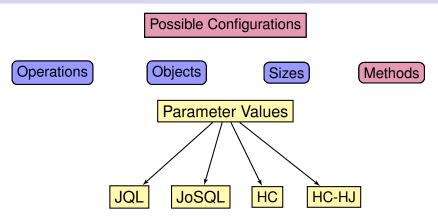
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## Configuration of the Benchmarking Framework



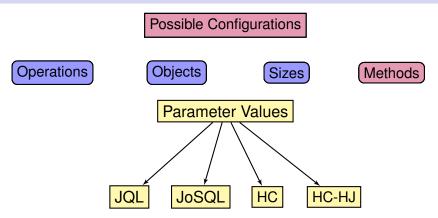
#### Which methods should be part of the framework?

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# Configuration of the Benchmarking Framework



### See the paper for further operator and configuration details

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### Analysis Techniques: Regression Tree Models

Method: HC-HJ, JQL

#### Tree Models: Recursive partitioning creates hierarchical view of data

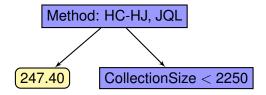
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# Analysis Techniques: Regression Tree Models



#### Tree Models: Recursive partitioning creates hierarchical view of data

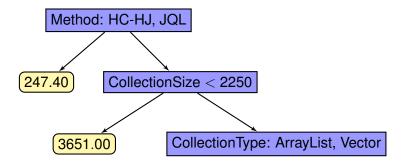
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# Analysis Techniques: Regression Tree Models



#### Tree Models: Recursive partitioning creates hierarchical view of data

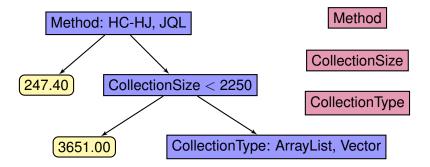
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# Analysis Techniques: Regression Tree Models



#### Explanatory Variable: Configuration of the benchmarking framework

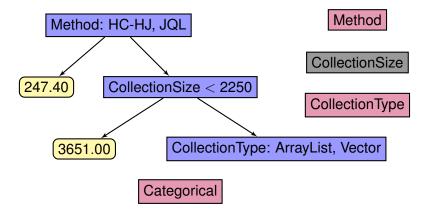
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# Analysis Techniques: Regression Tree Models



#### Non-parametric techniques that handles different variable types

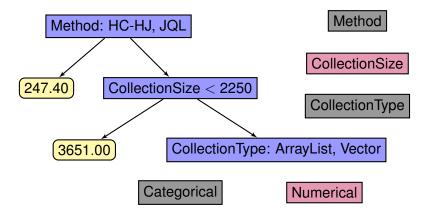
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# Analysis Techniques: Regression Tree Models



#### Non-parametric techniques that handles different variable types

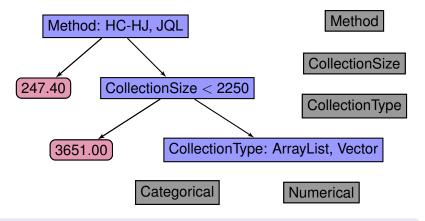
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# Analysis Techniques: Regression Tree Models



#### Response Variable: Response time of the benchmark

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## Join Benchmark with Integers and Strings

Method: HC-HJ, JQL

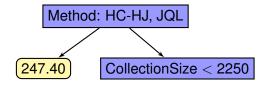
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## Join Benchmark with Integers and Strings



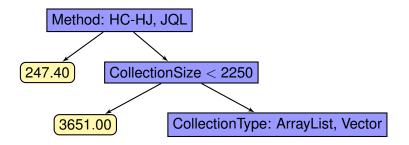
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## Join Benchmark with Integers and Strings



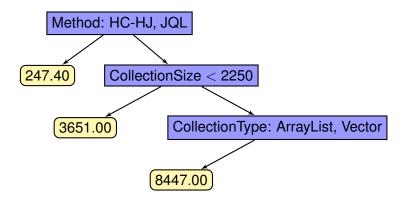
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## Join Benchmark with Integers and Strings



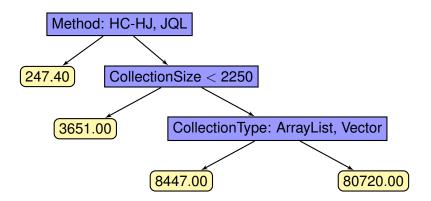
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## Join Benchmark with Integers and Strings



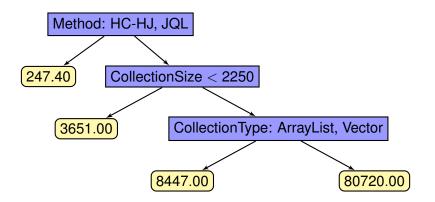
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# Join Benchmark with Integers and Strings



#### Reflection's Impact: HC-HJ and JQL exhibit lower values than JoSQL

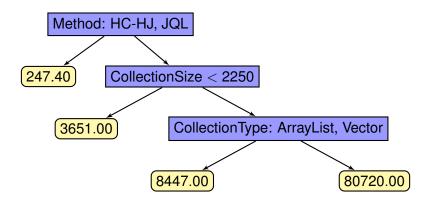
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Conclusion

# Join Benchmark with Integers and Strings



#### Reflection's Impact: LinkedList further degrades JoSQL's performance

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Conclusion

## Impact of Object Size on Joining

Small Obje	mall Objects Collection Size				
	Method	Small	Medium	Large	
	JQL	57.2	390.2	981.8	
	HC-HJ	69.3	378.1	923.5	
	JoSQL	997.3	3620.2	8823.1	

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Large Objects		<b>Collection Size</b>			
	Method	Small	Medium	Large	
	JQL	35.4	80.8	255.4	
	HC-HJ	11.4	63.3	217.8	
	JoSQL	930.3	3107.3	8165.9	

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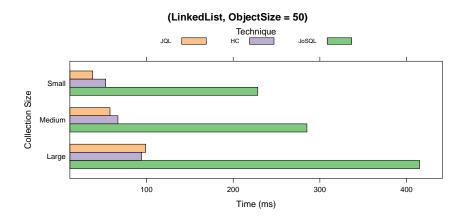
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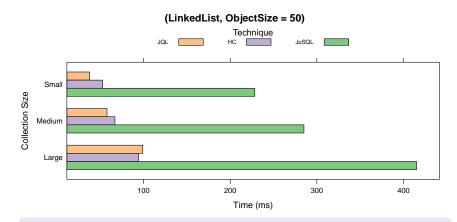
# SubQuery Benchmark with Graphs Containing Strings



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# SubQuery Benchmark with Graphs Containing Strings

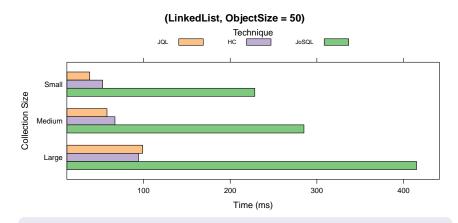


### JQL is Faster Than HC When the Collection Size is Small and Medium

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# SubQuery Benchmark with Graphs Containing Strings

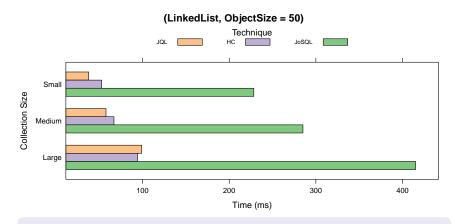


#### HC is Faster Than JQL When the Collection Size is Large

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# SubQuery Benchmark with Graphs Containing Strings

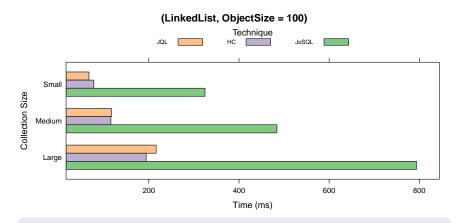


#### Why? JQL Must Track All of the Objects in the Heap

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# SubQuery Benchmark with Graphs Containing Strings



#### Trend is Even More Pronounced as the Object Size Increases

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# Conclusions and Future Work

### **Concluding Remarks**

- Comprehensive empirical study of query methods
- Interesting trends concerning JQL, JoSQL, HC, and HC-HJ
- Refer to the paper for many more insights

#### **Future Work**

- Integrate new benchmarks and object types
- Consider different sizes of objects and collections
- Incorporate different data finding methods
- Leverage additional statistical analysis techniques

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Ask and You Shall Receive: Empirically Evaluating Declarative Approaches to Finding Data in Unstructured Heaps

> Thank you for your attention! Questions?



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"Ask, and you will receive. Search, and you will find. Knock, and the door will be opened for you." Matthew 7:7 (GWT) http://bible.cc/matthew/7-7.htm