

presented at the Texas Instruments Workshop on Query Optimizers, 1993

# Prairie: An Algebraic Framework for Rule Specification in Query Optimizers

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September 13, 1993

# Overview of talk

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## What is the problem?

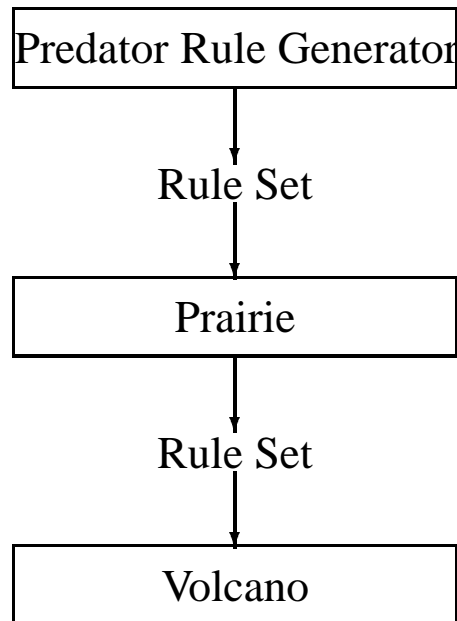
- Rule-based query optimization.
- Develop a simple, well-defined framework called Prairie to specify rules in a rule-based query optimizer.
  - Abstraction of rule specification to insulate user from rule engine.
  - Support automatic generation of rule sets.
- Test framework by using it to write query optimizers.
- Develop a pre-processor to translate Prairie rule specification to Volcano.
- Test performance of Prairie vs. Volcano.

## Why is it important?

- A simple framework makes it easy to read, write and understand rules with fewer errors.
- Need to distill the essence of what needs to be specified by the user.
  - Too many details to specify.
  - Software Engineering approach.
  - Research based on Volcano.

# Motivation

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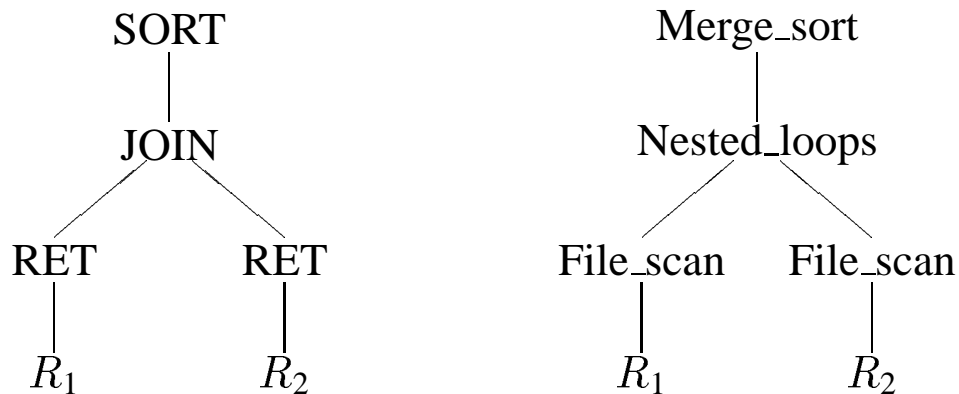


- Abstract specification of rules – only essential things to be specified.
- Minimize complexity (i.e., dependence on implementation details) of rule set specification.
- Admits different types of rule set optimizations:
  - Collapsing of rule sets.
  - Generate implementation details.
- Allows automatic generation of rule sets to be easier.

# Prairie: Terminology

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- **Stored files and streams.** Stored files are relations. Streams are produced by computations on stored files or other streams.
- **Operators.** Abstract computations on streams or files. Describes what the *semantics* are, not *how* they are implemented.
- **Algorithms.** Concrete implementations of operators.
- **Operator Tree.** Rooted tree with operators or algorithms as internal nodes, stored files as leaves.
- **Access Plan.** Operator tree with algorithms as internal nodes.
- **Property.** Information used by optimizer to choose between different operator trees/access plans.



# Prairie: Terminology

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- **Descriptors.** List of  $\langle property, value \rangle$  pairs that encode operator tree node information.

*Example:*

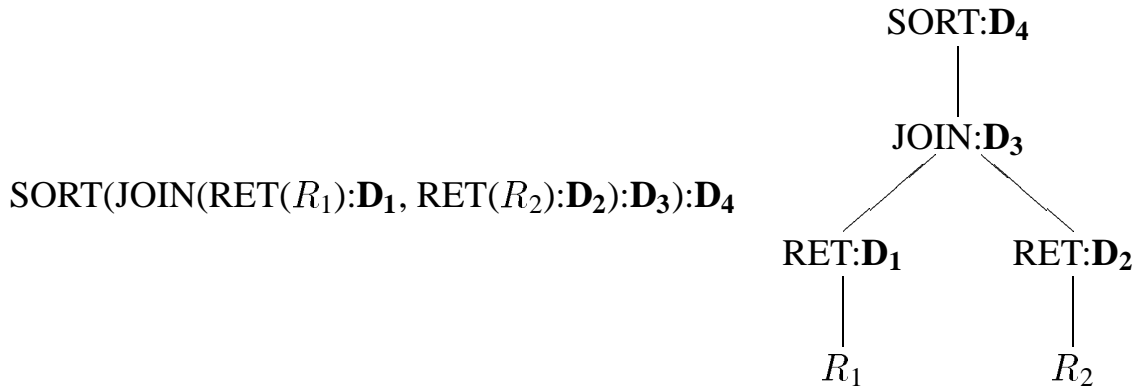
Property	Meaning
join_predicate	join predicate for JOIN operator
tuple_order	tuple order of stream
num_records	number of tuples in stream
tuple_size	size of individual tuple in stream
attribute_list	list of attributes
cost	estimated cost of algorithm

- Lot of effort in coding properties. We ultimately may want to provide with Prairie a library of pre-written implementations of certain properties.

# Prairie: Terminology

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- **Expression.** Operator tree with descriptor information attached.



Descriptor members are accessed by a structure member relationship. Thus,

$D_2.\text{num\_records} = \#$  of records returned by  $\text{RET}(R_2)$ .

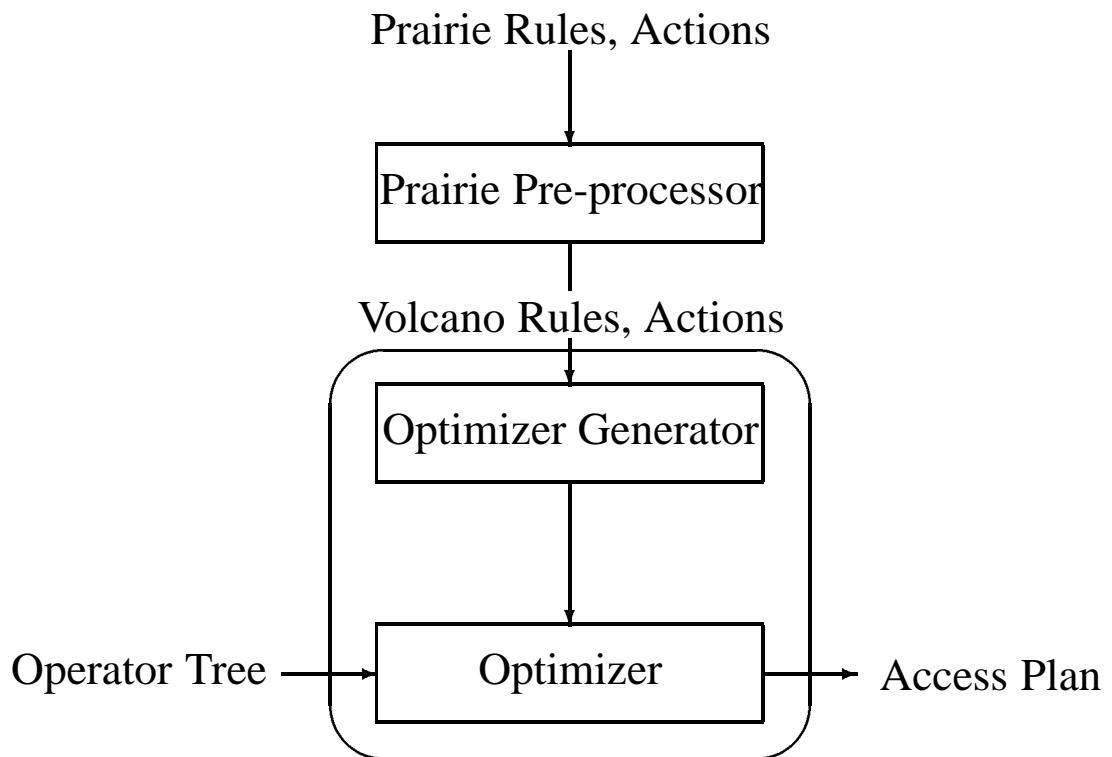
$D_4.\text{num\_records} = \#$  of records returned by the entire expression.

- **Rules.** (Largely influenced by Volcano)

Rules transform one operator tree into another. There are two types of rewrite rules: T-rules (“transformation rules”) and I-rules (“implementation rules”). Each type has a test and actions associated with it.

# Prairie: Model

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- Actions written in C.
- Top-down optimization
- Goal is to generate code with comparable efficiency to a hand-written version, yet input is much simpler.

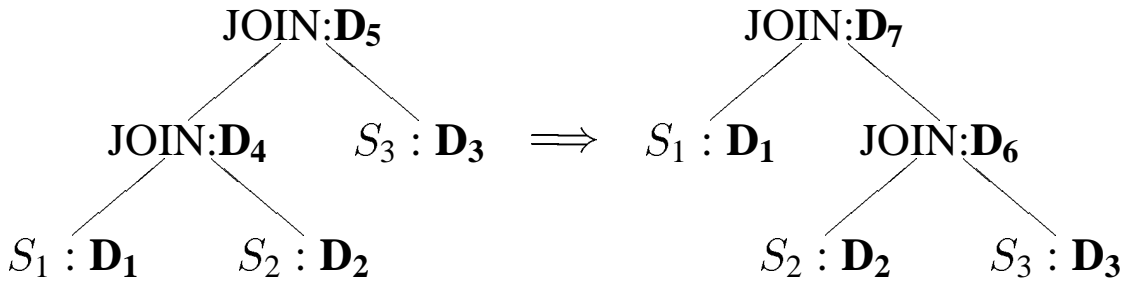
# Prairie: T-Rules

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**General Form:** (influenced by Volcano)

$$\begin{array}{c}
 E(x_1, \dots, x_n) : \mathbf{D}_1 \implies E'(x_1, \dots, x_n) : \mathbf{D}_2 \\
 \{\{ \text{pre-test statements} \}\} \\
 \text{test} \\
 \{\{ \text{post-test statements} \}\}
 \end{array}$$

**Example: JOIN Associativity**



$$\text{JOIN}(\text{JOIN}(S_1, S_2) : \mathbf{D}_4, S_3) : \mathbf{D}_5 \implies \text{JOIN}(S_1, \text{JOIN}(S_2, S_3) : \mathbf{D}_6) : \mathbf{D}_7$$

{  
}

! attr\_in\_attr\_list ( $\mathbf{D}_5$ .join\_predicate.operand1,  $\mathbf{D}_2$ .attributes)

{

$\mathbf{D}_7 = \mathbf{D}_5$ ;

$\mathbf{D}_7$ .join\_predicate =  $\mathbf{D}_4$ .join\_predicate;

$\mathbf{D}_6$ .attributes = union( $\mathbf{D}_2$ .attributes,  $\mathbf{D}_3$ .attributes);

$\mathbf{D}_6$ .join\_predicate =  $\mathbf{D}_5$ .join\_predicate;

$\mathbf{D}_6$ .tuple\_size = estimate\_tuple\_size( $\mathbf{D}_2$ .tuple\_size,  $\mathbf{D}_3$ .tuple\_size);

$\mathbf{D}_6$ .num\_records = estimate\_num\_records( $\mathbf{D}_2$ .num\_records,  $\mathbf{D}_3$ .num\_records);

}



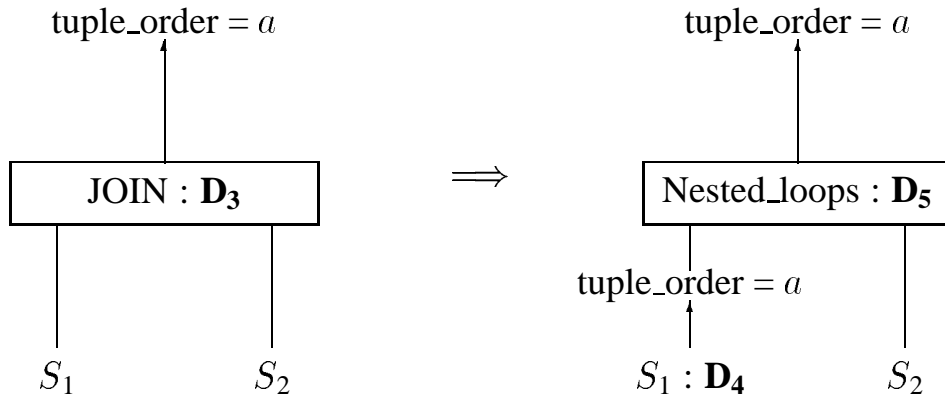
# Prairie: I-Rules

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**General Form:** (influenced by Volcano)

$$\begin{array}{c}
 E(x_1, \dots, x_n) : \mathbf{D}_1 \implies A(x_1, \dots, x_n) : \mathbf{D}_2 \\
 \text{test} \\
 \{\{ \text{pre-opt statements} \}\} \\
 \{\{ \text{post-opt statements} \}\}
 \end{array}$$

**Example: Nested Loops**



$$\begin{array}{c}
 \text{JOIN}(S_1, S_2) : \mathbf{D}_3 \implies \text{Nested\_Loops}(S_1 : \mathbf{D}_4, S_2) : \mathbf{D}_5 \\
 \text{TRUE}
 \end{array}$$

{{

$$\mathbf{D}_5 = \mathbf{D}_3;$$

$$\mathbf{D}_4 = \mathbf{D}_1;$$

$$\mathbf{D}_4.\text{tuple\_order} = \mathbf{D}_3.\text{tuple\_order};$$

}}

{{

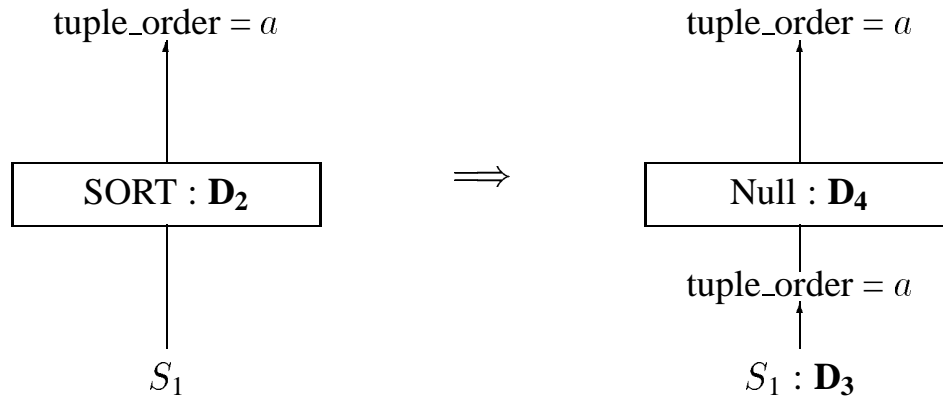
$$\mathbf{D}_5.\text{cost} = \mathbf{D}_4.\text{cost} + (\mathbf{D}_4.\text{num\_records}) * \mathbf{D}_2.\text{cost};$$

}}

# Prairie: Null Algorithm

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The Null algorithm serves to pass constraints of a node down to its input. Since operators are *explicit* in Prairie, we need a way to remove operators as necessary from operator trees.



## Example: Null SORT

$$\text{SORT}(S_1) : \mathbf{D}_2 \implies \text{Null}(S_1 : \mathbf{D}_3) : \mathbf{D}_4$$

TRUE

{{

$$\mathbf{D}_4 = \mathbf{D}_2;$$

$$\mathbf{D}_3 = \mathbf{D}_1;$$

$$\mathbf{D}_3.\text{tuple\_order} = \mathbf{D}_2.\text{tuple\_order};$$

}}

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$$\mathbf{D}_4.\text{cost} = \mathbf{D}_3.\text{cost};$$

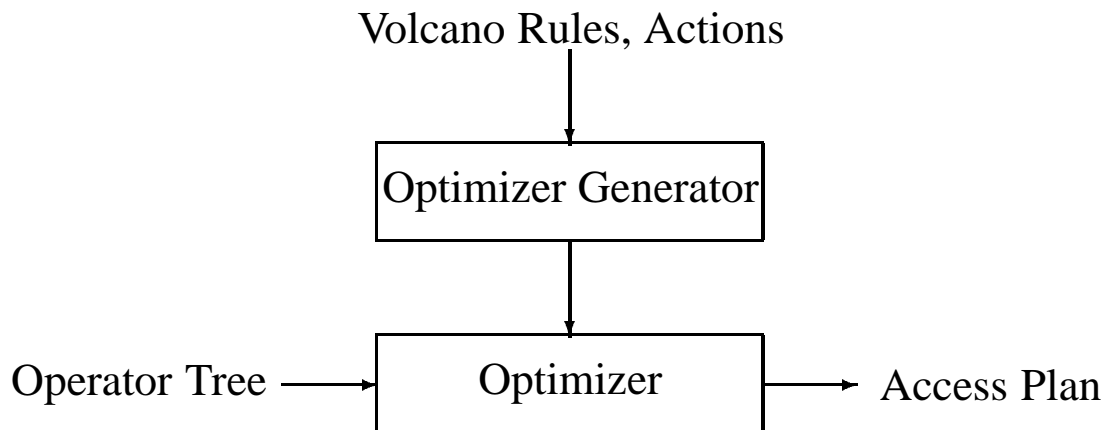
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# Volcano: Model

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## Volcano Model (Graefe, 1990)

- Top-down query optimization.



- Extensibility — can add new operators, algorithms.
- Two types of rules:
  - Transformation rules (non-cost-based rewrites).
  - Implementation rules (cost-based rewrites).
- Efficient storage of equivalence classes of operator trees.
- Branch-and-bound control of search space.
- Constraint-based generation of “interesting expressions” (ala System R).

# Volcano: Model

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- “Hidden” algorithms, rules: enforcers.

RET  
|  
 $R_1$

Merge\_sort  
|  
File\_scan  
|  
 $R_1$

## Why do we care?

- Hard to visualize where enforcers will appear in operator tree since enforcers don't appear in rules.
- Hard to see how operator trees are rewritten. Also, greater potential for errors in specifying rewrite rules.

*Conclusion:* Make all operators and algorithms *explicit*.

# Volcano: Model

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- Operator Tree node information represented using five different structures:
  - Operator/Algorithm arguments
  - Logical property
  - System property
  - Physical property
  - Cost
- Adding new operators/algorithms may require repartition of properties.

*Example.* Adding a relational project operator changes attribute list from a logical to physical property.

# Volcano: Model

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- Requires writing of functions to map properties between operator trees.
  - derive\_log\_prop.
  - derive\_sys\_prop.
  - derive\_phy\_prop.
  - do\_any\_good.
  - get\_input\_pv.
  - cost.

*Conclusion:* Use *one* structure to encode node information.

- Volcano implementation rules hard to read, write, understand because of scattered functions.
  - easy to make mistakes.

## The Problem

- Translating Prairie input to Volcano.
  - Automatic mapping of descriptor to different vectors in Volcano.
  - Automatically generate mapping functions in Volcano.

# Preliminary Experimental Results

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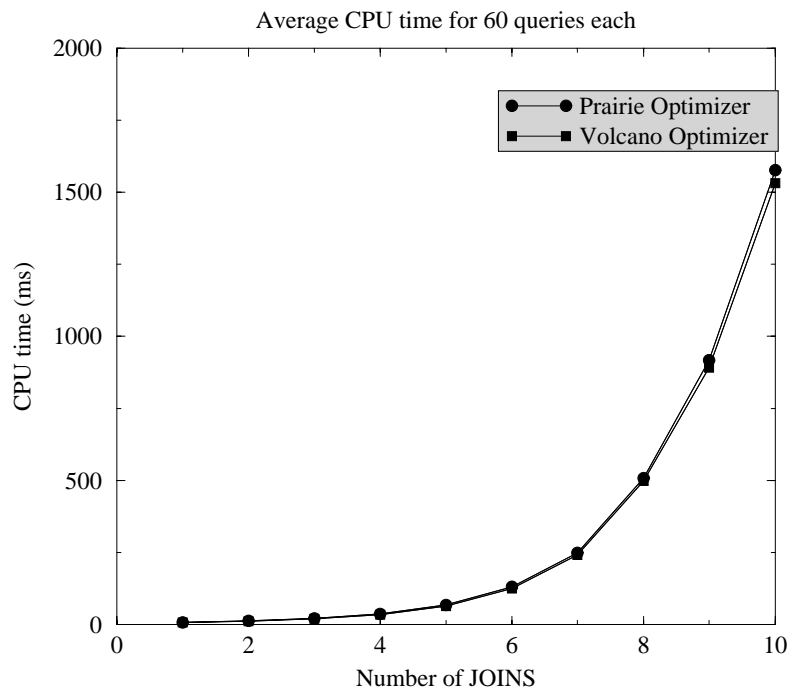
Specification of a simple centralized optimizer:

- JOIN, RET, SORT operators.
- One JOIN algorithm, Jmerge; one RET algorithm, File\_scan.
- One SORT algorithm, Merge\_sort.

## Lines of code

Prairie	Volcano from P2V	Hand-written Volcano
700	1300	1400

## n-way JOIN queries



**Other experiments:** Multiple JOIN algorithms (Jmerge, Jhash), collapsing of multiple rules.

# Future Work

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- Encode TI's OODB optimizer to see if Prairie scales.
- Investigate extending Prairie to build layered query optimizers.
  - Each layer consists of a set of rules.
  - Layers can be composed in arbitrary ways.
  - Rules in different layers can be composed for efficiency.
- May investigate extending Prairie to specify bottom-up query optimizers.