## **Dive Into Catalyst**

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"In almost every computation a great variety of arrangements for the succession of the processes is possible, and various considerations must influence the selection amongst them for the purposes of a Calculating Engine. One essential object is to choose that arrangement which shall tend to reduce to a minimum the time necessary for completing the calculation."



Ada Lovelace, 1843



## What Is Catalyst?

Catalyst is a functional, extensible query optimizer used by Spark SQL.

- Leverages advanced FP language (Scala) features
- Contains a library for representing <u>trees</u> and applying <u>rules</u> on them



#### **Trees**

Tree is the main data structure used in Catalyst

- A tree is composed of node objects
- A node has a node type and zero or more children
- Node types are defined in Scala as subclasses of the TreeNode class



#### **Trees**

- Literal(value: Int)
- Attribute(name: String)
- Add(left: TreeNode, right: TreeNode)

```
Add(
Attribute(x),
Add(Literal(1), Literal(2)))
```

```
Add

Attribute(x)

Add

Literal(1)

Literal(2)
```



#### Rules are functions that transform trees

- Typically functional
- Leverage pattern matching
- Used together with
  - TreeNode.transform (synonym of transformDown)
  - TreeNode.transformDown (pre-order traversal)
  - TreeNode.transformUp (post-order traversal)



## Examples

- Simple constant folding

```
tree.transform {
  case Add(Literal(c1), Literal(c2)) => Literal(c1 + c2)
  case Add(left, Literal(0)) => left
  case Add(Literal(0), right) => right
}
```



## Examples

- Simple constant folding

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tree.transform {
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```

Pattern



## Examples

- Simple constant folding

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## Examples

- Simple constant folding

```
tree.transform {
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}
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Rule



Brainsuck is an optimizing compiler of the Brainfuck programming language



Brainsuck is an optimizing compiler of the Brainfuck programming language

 Yeah, it's not a typo, just wanna avoid saying dirty words all the time during the talk :^)



Brainsuck is an optimizing compiler of the Brainfuck programming language

Inspired by <u>brainfuck optimization strategies</u>
 by Mats Linander



Brainsuck is an optimizing compiler of the Brainfuck programming language

Goals

Illustrate the power and conciseness of the Catalyst optimizer



Brainsuck is an optimizing compiler of the Brainfuck programming language

- Goals

As short as possible so that I can squeeze it into a single talk (292 loc)



Brainsuck is an optimizing compiler of the Brainfuck programming language

Non-goal

Build a high performance practical compiler



Brainsuck is an optimizing compiler of the Brainfuck programming language

- Contains
  - Parser, IR, and interpreter of the language
  - A simplified version of the Catalyst library, consists of the TreeNode class and the rule execution engine
  - A set of optimization rules



```
object MergeMoves extends Rule[Instruction] {
  override def apply(tree: Instruction) = tree.transformUp {
    case Move(n, Move(m, next)) => if (n + m == 0) next else Move(n + m, next)
object MergeAdds extends Rule[Instruction] {
  override def apply(tree: Instruction) = tree.transformUp {
    case Add(n, Add (m, next)) \Rightarrow if (n + m == 0) next else Add(n + m, next)
object Clears extends Rule[Instruction] {
  override def apply(tree: Instruction) = tree.transformUp {
    case Loop(Add( , Halt), next) => Clear(next)
object Scans extends Rule[Instruction] {
  override def apply(tree: Instruction) = tree.transformUp {
    case Loop(Move(n, Halt), next) => Scan(n, next)
object MultisAndCopies extends Rule[Instruction] {
  override def apply(tree: Instruction) = tree.transform {
    case Loop(Add(-1, MoveAddPairs(seq, offset, Move(n, Halt))), next) if n == -offset =>
      seq.foldRight(Clear(next): Instruction) {
        case ((distance, 1), code) => Copy(distance, code)
        case ((distance, increment), code) => Multi(distance, increment, code)
object MoveAddPairs {
  type ResultType = (List[(Int, Int)], Int, Instruction)
  def unapply(tree: Instruction): Option[ResultType] = {
    def loop(tree: Instruction, offset: Int): Option[ResultType] = tree match {
      case Move(n, Add(m, inner)) =>
        loop(inner, offset + n).map { case (seq, finalOffset, next) =>
          ((offset + n, m) :: seq, finalOffset, next)
      case inner => Some((Nil, offset, inner))
    loop(tree, 0)
```

Optimization rules in Brainsuck

Optimization rules in bfoptimization by Mats Linander



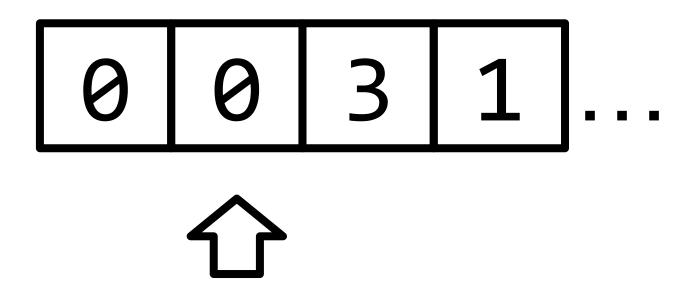
```
databricks
```

```
d opt_copylcop(ir):
return opt_multiloop(ir, True)
 The copy loop is a common construct in brainfact, where scretch
like [-Markersen] is used to duplicate the correct cell to two
other cells. This exempte can be replaced with three committee operations; (opyr1) Copyr1) Clear(0).
The multiplication loop is similar but edds a multiplicative factor to the value being copied. E.g. [-secondaries] can be poplaced with Mul(2, 5) Mul(3, 1) Clear(6).
 ir = ir(i)
     * find the next 'leaf loop', i.e. the next loop that doesn't * hold any other loope, break if no such loop exists.

while i* leafing:
    if issuetance(ir[i], Open);
      # find the nest block of 'effsetable' instructions
%LOCKOFS = (Add, Sub, Left, Right, Clear, Is, Out)
while i < len(ir) and ir[i]._class_ set is %LOCKOFS
      j = 1
while j < len(ir) and ir(j)._class_ is BLOCKOPS;</pre>
                        nally reposition the pointer to wherever it ended up
```

#### **Machine Model**

A dead simple simulation of Turing machine





## **Instruction Set**

Instruction	Meaning	
(Initial)	<pre>char array[infinitely large size] = {0}; char *ptr = array;</pre>	
>	++ptr; // Move right	
<	ptr; // Move left	
+	++*ptr; // Increment	
-	*ptr; // Decrement	
•	<pre>putchar(*ptr); // Put a char</pre>	
,	*ptr = getchar(); // Read a char	
]	while (*ptr) { // Loop until *ptr is 0	
]	} // End of the loop	



## Sample Program



## Sample Program



## **Brainsuck Instruction Set**

Instruction	Meaning
Move(n)	ptr += n;
Add(n)	*ptr += n;
Loop(body)	while (*ptr) { body; }
Out	<pre>putchar(*ptr);</pre>
In	*ptr = getchar();
Scan(n)	while (*ptr) { ptr +=n; }
Clear	*ptr = 0
Copy(offset)	*(ptr + offset) = *ptr;
Multi(offset, n)	*(ptr + offset) += (*ptr) * n;
Halt	abort();



## **Brainsuck Instruction Set**

Instruction	Meaning
Move(n)	ptr += n;
Add(n)	*ptr += n;
Loop(body)	while (*ptr) { body; }
Out	<pre>putchar(*ptr);</pre>
In	*ptr = getchar();
Scan(n)	while (*ptr) { ptr +=n; }
Clear	*ptr = 0
<pre>Copy(offset)</pre>	<pre>*ptr = 0  *(ptr + offset) = *ptr;  *(ptr + offset) += (*ptr) *  abort();</pre>
<pre>Multi(offset, n)</pre>	*(ptr + offset) += (*ptr) *
Halt	abort();



```
trait TreeNode[BaseType <: TreeNode[BaseType]] {
    self: BaseType =>

    def children: Seq[BaseType]

    protected def makeCopy(args: Seq[BaseType]): BaseType

    def transform(rule: PartialFunction[BaseType, BaseType]): BaseType = ...

    def transformDown(rule: PartialFunction[BaseType, BaseType]): BaseType = ...

    def transformUp(rule: PartialFunction[BaseType, BaseType]): BaseType = ...

    ...
}
```



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    protected def makeCopy(args: Seq[BaseType]): BaseType

    def transform(rule: PartialFunction[BaseType, BaseType]): BaseType = ...

    def transformDown(rule: PartialFunction[BaseType, BaseType]): BaseType = ...

    def transformUp(rule: PartialFunction[BaseType, BaseType]): BaseType = ...

    ...
}
```



## Helper classes for leaf nodes and unary nodes

```
trait LeafNode[BaseType <: TreeNode[BaseType]] extends TreeNode[BaseType] {
    self: BaseType =>
    override def children = Seq.empty[BaseType]
    override def makeCopy(args: Seq[BaseType]) = this
}

trait UnaryNode[BaseType <: TreeNode[BaseType]] extends TreeNode[BaseType] {
    self: BaseType =>
    def child: BaseType
    override def children = Seq(child)
}
```



#### Represented as tree nodes

```
sealed trait Instruction extends TreeNode[Instruction] {
   def next: Instruction
   def run(machine: Machine): Unit
}

sealed trait LeafInstruction extends Instruction with LeafNode[Instruction] {
   self: Instruction =>
    def next: Instruction = this
}

sealed trait UnaryInstruction extends Instruction with UnaryNode[Instruction] {
   self: Instruction =>
    def next: Instruction = child
}
```



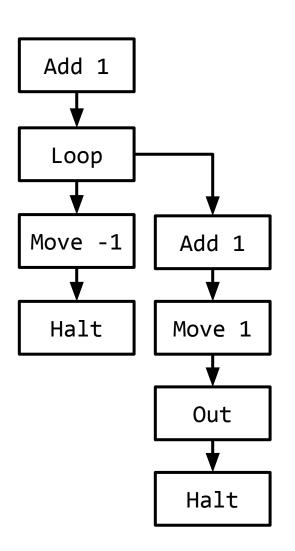
```
case object Halt extends LeafInstruction {
  override def run(machine: Machine) = ()
}

case class Add(n: Int, child: Instruction) extends UnaryInstruction {
  override protected def makeCopy(args: Seq[Instruction]) =
      copy(child = args.head)
  override def run(machine: Machine) = machine.value += n
}

case class Move(n: Int, child: Instruction) extends UnaryInstruction {
  override protected def makeCopy(args: Seq[Instruction]) =
      copy(child = args.head)
  override def run(machine: Machine) = machine.pointer += n
}
```

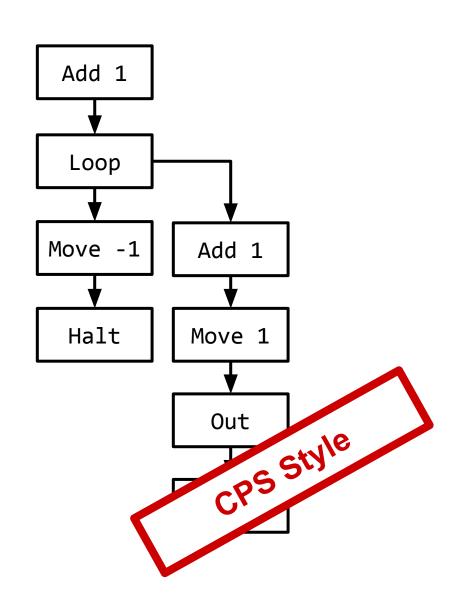


```
// +[<]+>.
Add(1,
  Loop(
    Move(-1,
      Halt),
    Add(1,
      Move(1,
        Out(
           Halt)))))
```





```
// +[<]+>.
Add(1,
  Loop(
    Move(-1,
      Halt),
    Add(1,
      Move(1,
        Out(
           Halt)))))
```





#### Rule Executor

## Rules are functions organized as Batches

```
trait Rule[BaseType <: TreeNode[BaseType]] {
  def apply(tree: BaseType): BaseType
}

case class Batch[BaseType <: TreeNode[BaseType]](
  name: String,
  rules: Seq[Rule[BaseType]],
  strategy: Strategy)</pre>
```



#### Rule Executor

Each Batch has an execution strategy

```
sealed trait Strategy {
  def maxIterations: Int
}
```

A Batch of rules can be repeatedly applied to a tree according to some Strategy



#### Rule Executor

## Each Batch has an execution strategy

```
sealed trait Strategy {
  def maxIterations: Int
}
```

## Apply once and only once

```
case object Once extends Strategy {
  val maxIterations = 1
}
```



#### Rule Executor

#### Each Batch has an execution strategy

```
sealed trait Strategy {
  def maxIterations: Int
}
```

#### Apply repeatedly until the tree doesn't change

```
final case class FixedPoint(maxIterations: Int) extends Strategy

case object FixedPoint {
  val Unlimited = FixedPoint(-1)
}
```



# Finally... Optimizations!



#### **Optimization 1: Merge Moves**

#### Patterns:

- >>>>
- <<<<
- >><<<>>>

#### Basic idea:

- Merge adjacent moves to save instructions



## **Optimization 1: Merge Moves**

#### Examples:

- Move(1, Move(1, next) ⇒
  Move(2, next)
- Move(1, Move(-1, next) ⇒
  next



## **Optimization 1: Merge Moves**

```
object MergeMoves extends Rule[Instruction] {
  override def apply(tree: Instruction) = tree.transformUp {
    case Move(n, Move(m, next)) =>
      if (n + m == 0) next else Move(n + m, next)
  }
}
```

#### Strategy:

- FixedPoint

We'd like to merge all adjacent moves until none can be found



## **Optimization 2: Merge Adds**

#### **Patterns**

```
- ++++
- ----
- ++---++
```

#### Basic idea:

Merge adjacent adds to save instructions



## **Optimization 2: Merge Adds**

#### **Examples:**

- Add(1, Add(1, next) ⇒
  Add(2, next)
- Add(1, Add(-1, next) ⇒
  next



#### **Optimization 2: Merge Adds**

```
object MergeAdds extends Rule[Instruction] {
  override def apply(tree: Instruction) = tree.transformUp {
    case Add(n, Add(m, next)) =>
      if (n + m == 0) next else Add(n + m, next)
  }
}
```

#### Strategy:

- FixedPoint

We'd like to merge all adjacent adds until none can be merged



## **Optimization 3: Clears**

#### **Patterns**

- [-]
- [--]
- [+]

#### Basic idea:

- These loops actually zero the current cell. Find and transform them to Clears.



#### **Optimization 3: Clears**

#### Examples:

- Loop(Add(-1, Halt), next) ⇒
  Clear(next)
- Loop(Add(2, Halt), next) ⇒
  Clear(next)



## **Optimization 3: Clears**

```
object Clears extends Rule[Instruction] {
  override def apply(tree: Instruction) = tree.transform {
    case Loop(Add(n, Halt), next) =>
      Clear(next)
  }
}
```

#### Strategy:

- Once

After merging all adds, we can find all "clear" loops within a single run



## **Optimization 4: Scans**

#### **Patterns**

- [<]
- **-** [<<]
- **-** [>]

#### Basic idea:

 These loops move the pointer to the most recent zero cell in one direction. Find and transform them to Scans.



#### **Optimization 4: Scans**

#### **Examples:**

- Loop(Move(-1, Halt), next) ⇒
  Scan(-1, next)
- Loop(Move(2, Halt), next) ⇒
  Scan(2, next)



## **Optimization 4: Scans**

```
object Scans extends Rule[Instruction] {
  override def apply(tree: Instruction) = tree.transform {
    case Loop(Scan(n, Halt), next) =>
        Scan(n, next)
  }
}
```

#### Strategy:

- Once

After merging all adds, we can find all "scan" loops within a single run



#### Patterns:

```
- [->>++<<]

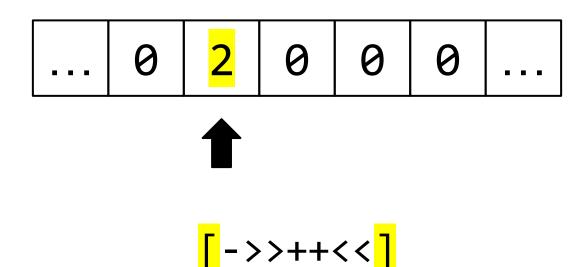
*(ptr + 2) += (*ptr) * 2;

- [->>++<<<--->]

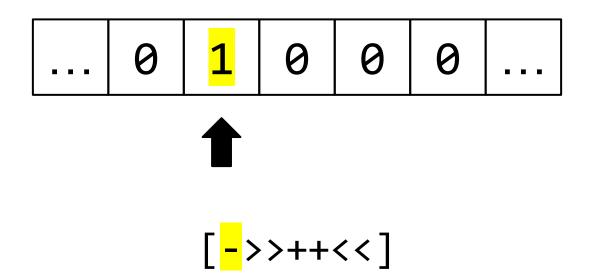
*(ptr + 2) += (*ptr) * 2;

*(ptr - 1) += (*ptr) * (-3);
```

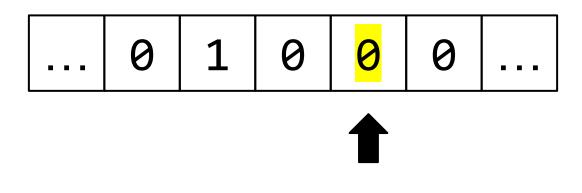




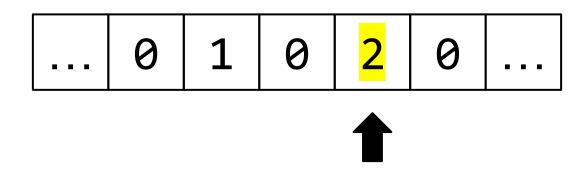




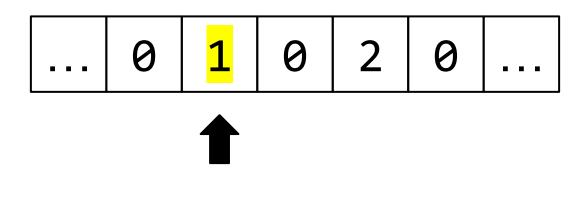




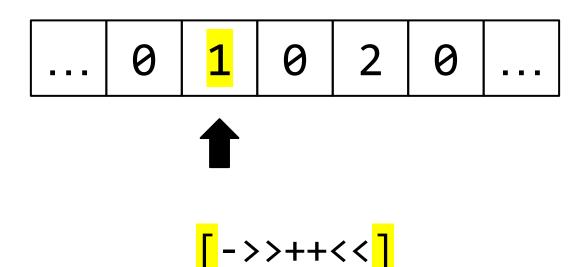




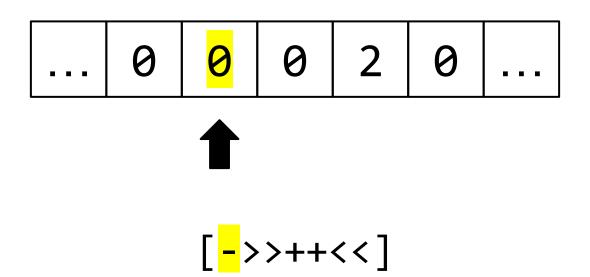




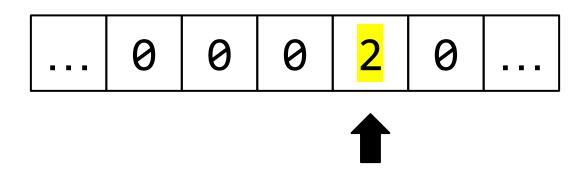




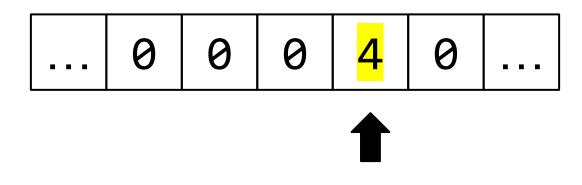




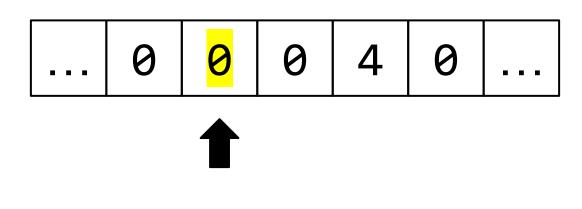




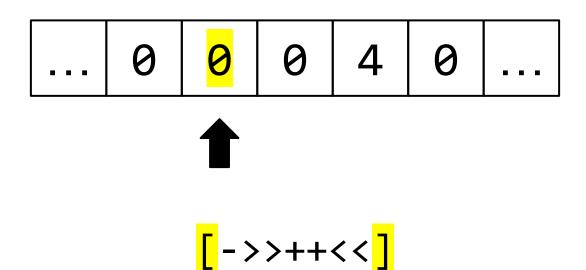














#### **Examples:**

```
- Loop(
   Add(-1,
        Move(2,
        Add(2,
        Move(-2, Halt)))), next) ⇒
   Multi(2, 2,
   Clear(next))
```



#### **Examples:**

```
Loop(
   Add(-1,
    Move(2, Add(2,
     Move(-3, Add(-3,
      Move(1, Halt))))), next) \Rightarrow
  Multi(2, 2,
   Multi(-1, -3,
    Clear(next)))
```



This pattern is relatively harder to recognize, since it has variable length:

```
tree.transform {
  case Loop(Add(-1, ???))) => ???
}
```



What we want to extract from the pattern?

- (offset, step) pairs
   For generating corresponding Multi instructions sequences
- finalOffset
   To check whether the pointer is reset to the original cell at the end of the loop body



#### **Examples:**

```
- [->>++<<]
- Pairs: (2,2) :: Nil, finalOffset: 2
- [->>++<---<]
- Pairs: (2,2)::(1,-3)::Nil, finalOffset: 1
- [-<<<+++>-->>]
- Pairs: (-3,3)::(-2,-2)::Nil, finalOffset: -2
```



#### **Extractor Objects**

In Scala, patterns can be defined as <u>extractor</u> <u>objects</u> with a method named <u>unapply</u>. This methods can be used to recognize arbitrarily complex patterns.



An extractor object which recognizes move-add pairs and extracts offsets and steps:



An extractor object which recognizes move-add pairs and extracts offsets and steps:

```
object MoveAddPairs {
  type ResultType = (List[(Int, Int)], Int, Instruction)

def unapply(tree: Instruction): Option[ResultType] = {
  def loop(tree: Instruction, offset: Int): Option[ResultType] = tree match {
    case Move(n, Add(m, inner)) =>
        loop(inner, offset + n).map { case (seq, finalOffset, next) =>
        ((offset + n, m) :: seq, finalOffset, next)
    }
    case inner => Some((Nil, offset, inner))
  }
  loop(tree, 0)
}
```



```
object MultisAndCopies extends Rule[Instruction] {
  override def apply(tree: Instruction): Instruction = tree.transform {
    case Loop(Add(-1, MoveAddPairs(seq, offset, Move(n, Halt))), next)
        if n == -offset =>
        seq.foldRight(Clear(next): Instruction) {
        case ((distance, increment), code) => Multi(distance, increment, code)
    }
}
```

#### Strategy:

- Once

After merging moves and adds, we can find all multiplication loops within a single run



## **Optimization 6: Copies**

#### Patterns:

- [->>+<<]
- [->>+<<<+>]

#### Basic idea:

- These are actually special forms of multiplication loops with 1 as multiplicand
- Replace Multi with a cheaper Copy



#### **Optimization 6: Copies**

#### **Examples:**

```
Loop(
   Add(-1,
    Move(2, Add(1,
     Move(-3, Add(1,
      Move(1, Halt))))), next) \Rightarrow
  Copy(2,
   Copy(-1),
    Clear(next)))
```



## **Optimization 6: Copies**

```
object MultisAndCopies extends Rule[Instruction] {
  override def apply(tree: Instruction): Instruction = tree.transform {
    case Loop(Add(-1, MoveAddPairs(seq, offset, Move(n, Halt))), next)
        if n == -offset =>
        seq.foldRight(Clear(next): Instruction) {
        case ((distance, 1), code) => Copy(distance, code)
        case ((distance, increment), code) => Multi(distance, increment, code)
    }
}
```

#### Strategy:

- Once

After merging moves and adds, we can find all copy loops within a single run



## Composing the Optimizer

```
val optimizer = new Optimizer {
  override def batches = Seq(
    Batch(
      "Contraction",
      MergeMoves :: MergeAdds :: Nil,
      FixedPoint.Unlimited),

    Batch(
      "LoopSimplification",
      Clears :: Scans :: MultisAndCopies :: Nil,
      Once)
  ).take(optimizationLevel)
}
```

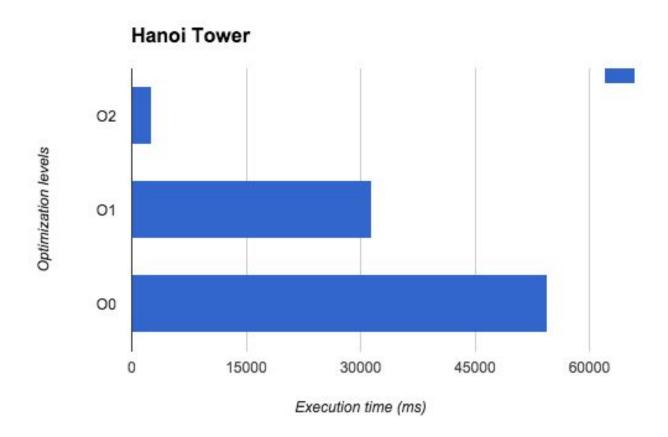


#### **Demo: Hanoi Tower**

Towers of Hanoi in Brainf\*ck
Written by Clifford Wolf <a href="http://www.clifford.at/bfcpu/">http://www.clifford.at/bfcpu/</a>

xXXXXXx xXXXXXXXXXXXXXXXXXXXXXXXXX



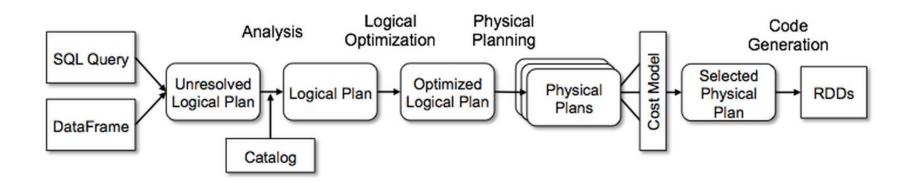




The following structures are all represented as TreeNode objects in Spark SQL

- Expressions
- Unresolved logical plans
- Resolved logical plans
- Optimized logical plans
- Physical plans



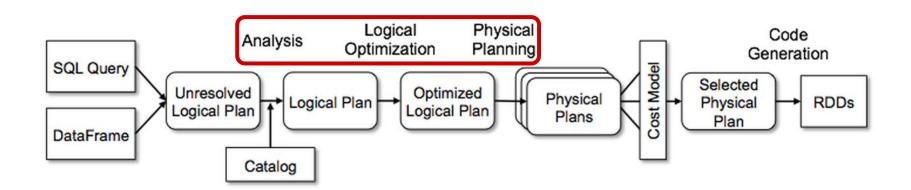




The actual Catalyst library in Spark SQL is more complex than the one shown in this talk. It's used for the following purposes:

- Analysis (logical ⇒ logical)
- Logical optimization (logical ⇒ logical)
- Physical planning (logical ⇒ physical)







#### References

- Spark SQL and Catalyst
  - Deep Dive into Spark SQL's Catalyst Optimizer (Databricks blog)
  - Spark SQL: Relational Data Processing in Spark (Accepted by SIGMOD15)
- Brainfuck language and optimization
  - Brainfuck optimization strategies
  - Optimizing brainfuck compiler
  - Wikipedia: Brainfuck
- Brainsuck code repository
  - https://github.com/liancheng/brainsuck



# **Thanks**

Q&A

