



DSL

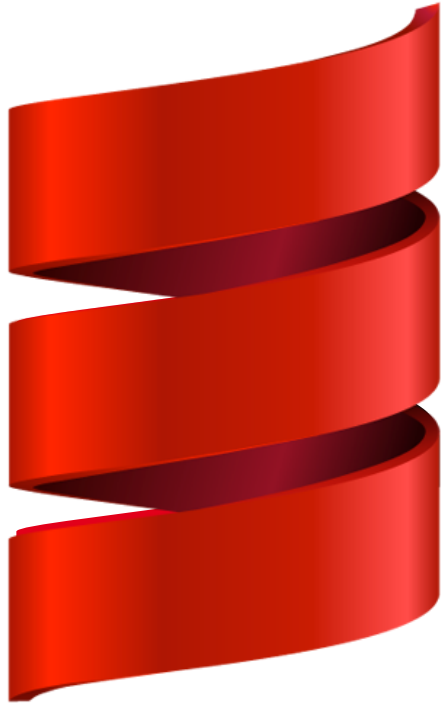
Embedding

in Scala

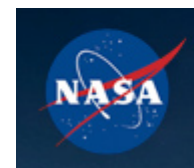
Tiark Rompf

PURDUE
UNIVERSITY

DSL Summer School, July 2015



Scala



Scala = scalable language

Small Scripts -> Large Systems

gradual, evolutionary

FP + OOP

```
class Matrix[T:Numeric](val rows: Int, val cols: Int) {  
  private val arr: Array[T] = new Array[T](rows * cols)  
  
  def apply(i: Int, j: Int): T = arr(i*cols + j)  
  
  def *(that: Matrix[T]) = { val res = new Matrix[T]; ...; res }  
  def +(that: Matrix[T]) = ...  
}  
  
// a,b,c,d: Matrix[Double]  
val x = a*b + a*c  
val y = a*c + a*d  
println(x+y)
```

Operator overloading, higher-order functions, by-name parameters, implicits, traits, ...

What about performance?

"The compiler / JVM will
make it run fast"

(wishful thinking)

10x – 1000x slowdown

(hard reality)

Yammer Moving from Scala to Java

Posted by [Alex Blewitt](#) on Nov 30, 2011

Sections [Development](#) Topics

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An [e-mail](#), sent from Yammer to Typesafe, ended up being that Yammer is moving it with complexity and performance.

Yammer PR Shelley Risk of Coda Hale, rather than original author has been Coda clarified that the message (CEO of [Typesafe](#)) following

“Via profiling and examining the bytecode we managed to get a 100x improvement by adopting some simple rules:

Don't ever use a for-loop

Don't ever use `scala.collection.mutable`

Don't ever use `scala.collection.immutable`

Always use `private[this]`

Avoid closures”

Update: Code has published [Yammer's official position](#) on the subject; which confirms the

Slow Program -> Fast Program?

Much harder!

"Any problem in computer science can be solved by another level of indirection"

-- David Wheeler

"Any problem in computer science can be solved by another level of indirection -- except problems caused by too many levels of indirection"

Abstraction Penalty

Type Classes

```
class Matrix[T:Numeric:Manifest](rows: Int, cols: Int) {  
  private val arr: Array[T] = new Array[T](rows * cols)  
  private val num = implicitly[Numeric[T]]; import num._
```

```
  def apply(i: Int, j: Int): T = {  
    arr(i*cols + j)
```

Indirection

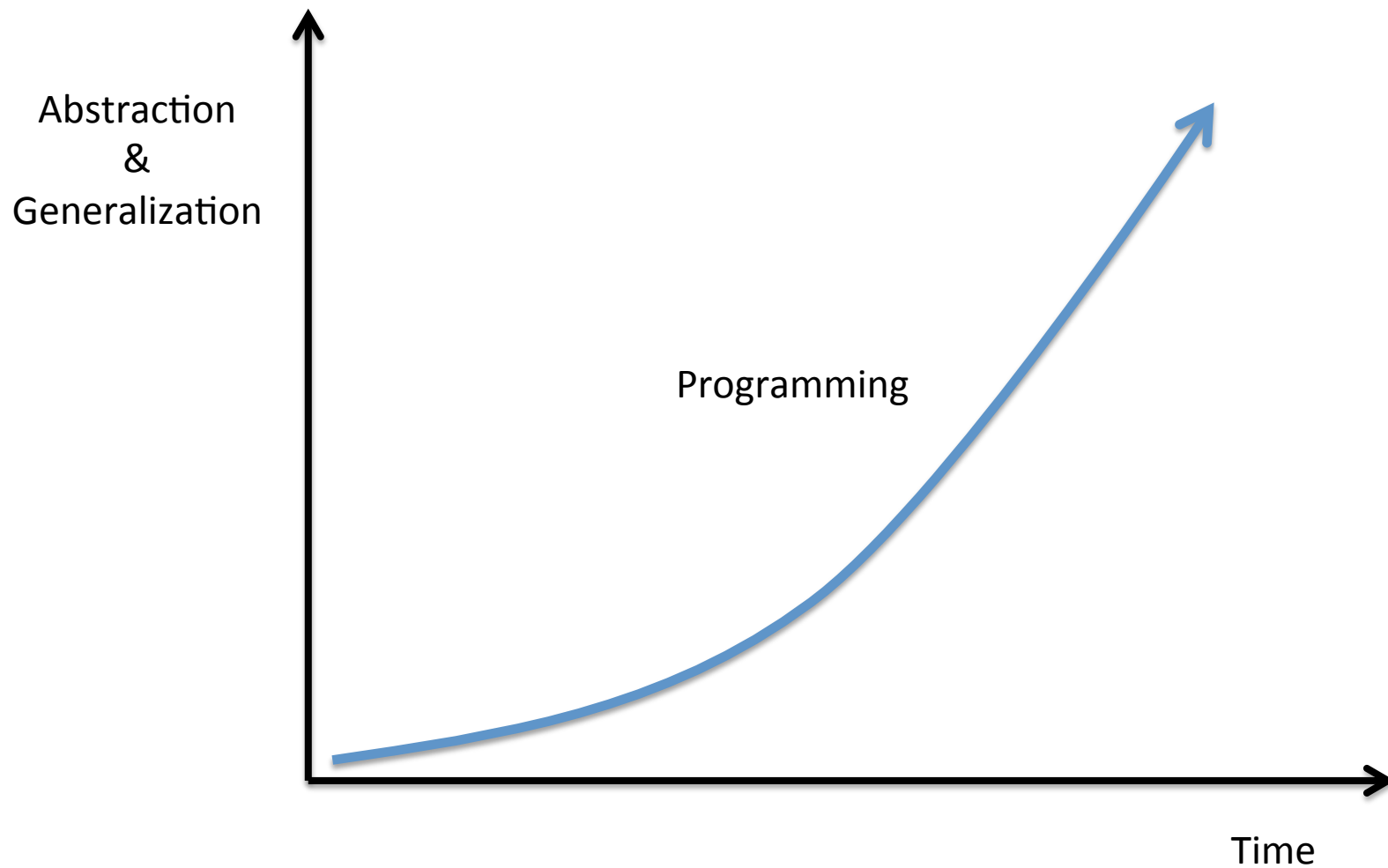
```
  def update(i: Int, j: Int, e: T): Unit =  
    arr(i*cols + j) = e
```

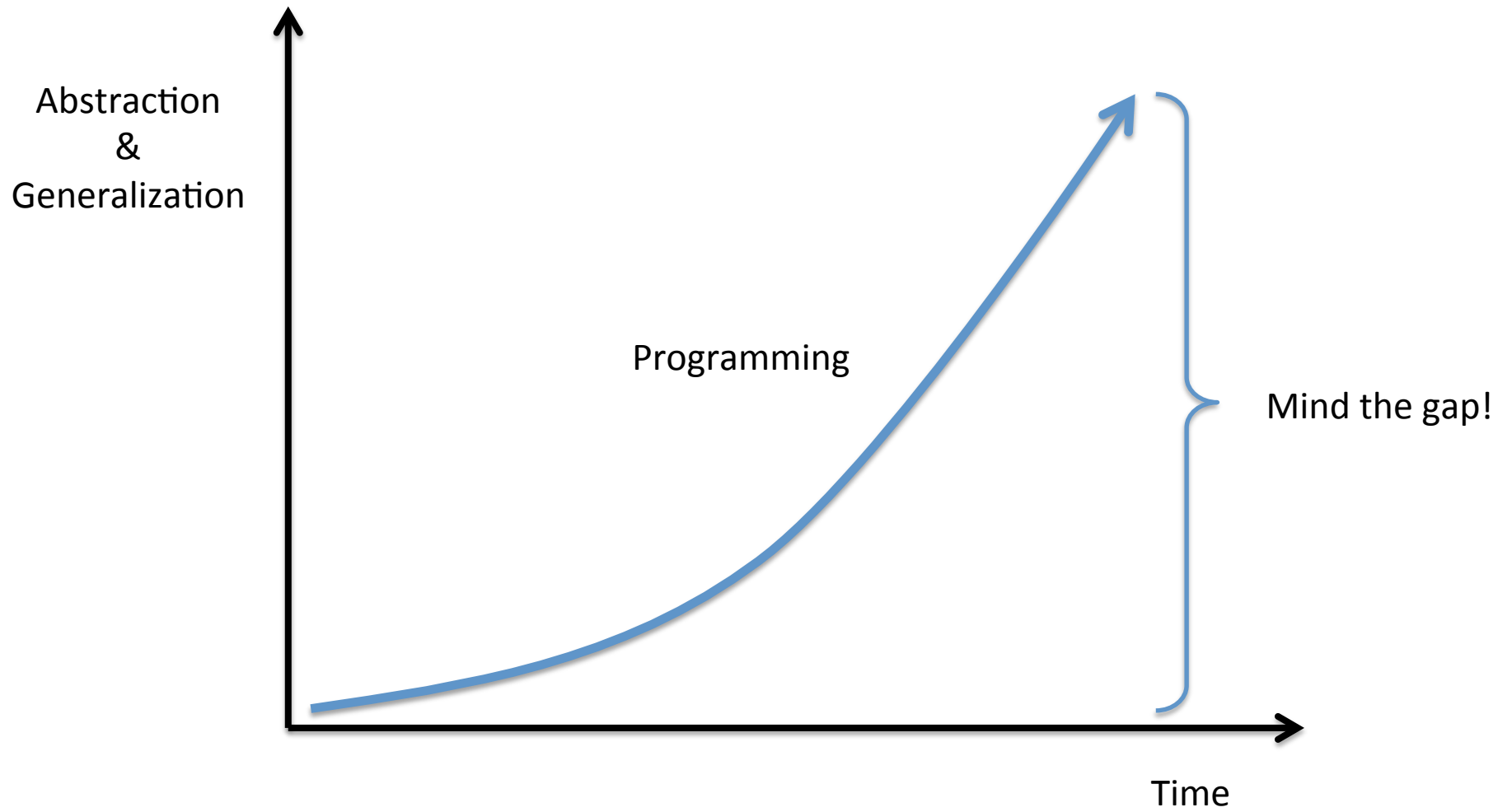
```
  def *(that: Matrix[T]) = {  
    val res = new Matrix[T](this.rows, that.cols)  
    for (i <- 0 until this.rows) {  
      for (j <- 0 until that.cols) {  
        for (k <- 0 until this.cols) {  
          res(i, j) += this(i, k) * that(k, j)
```

Closures
(and megamorphic
call sites)

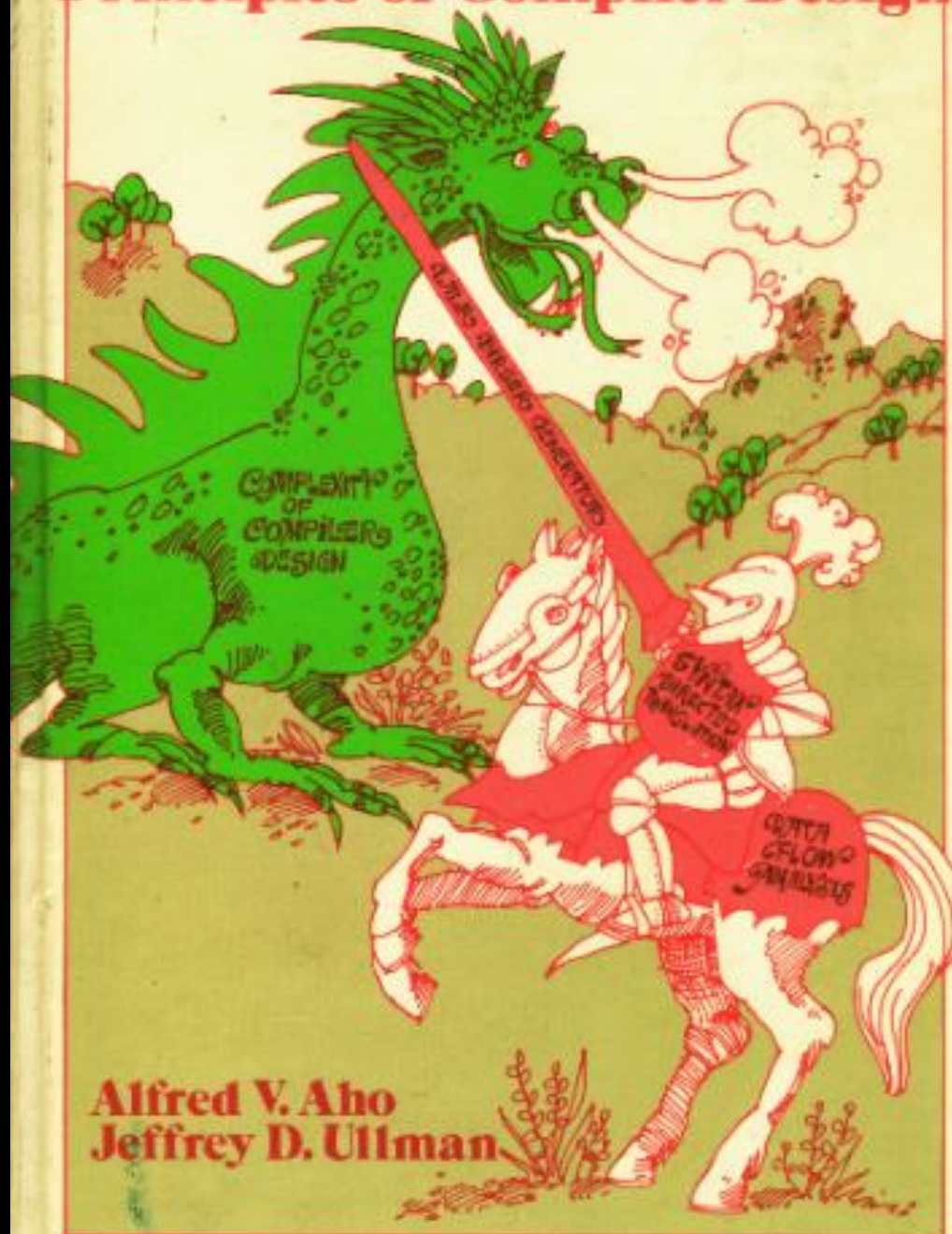
Method
Calls

```
        }  
      }  
    }  
    res  
  }
```

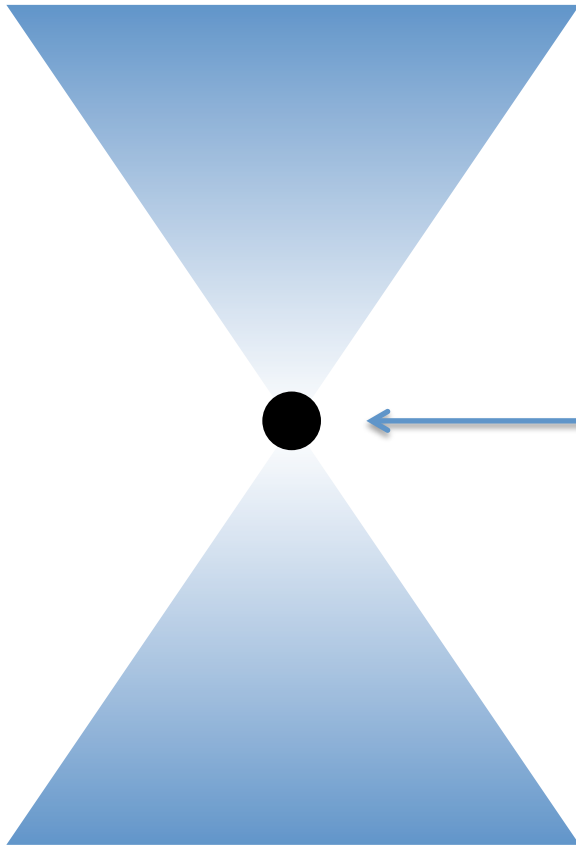




Principles of Compiler Design



Programmer

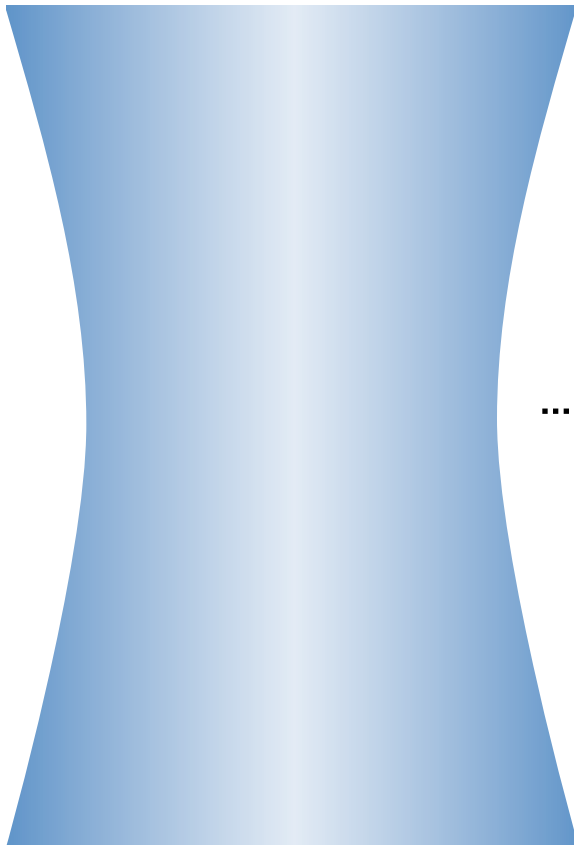


Hardware

general purpose compiler



Programmer



Hardware

..... Matrix, Graph, ...

..... Array, Struct, Loop, ...

..... SIMD, GPU, cluster, ...

- **horizontal and vertical** extensibility
- **generic optimizations** at each level (cse, dce, ...)

Secret Weapon: **LMS**

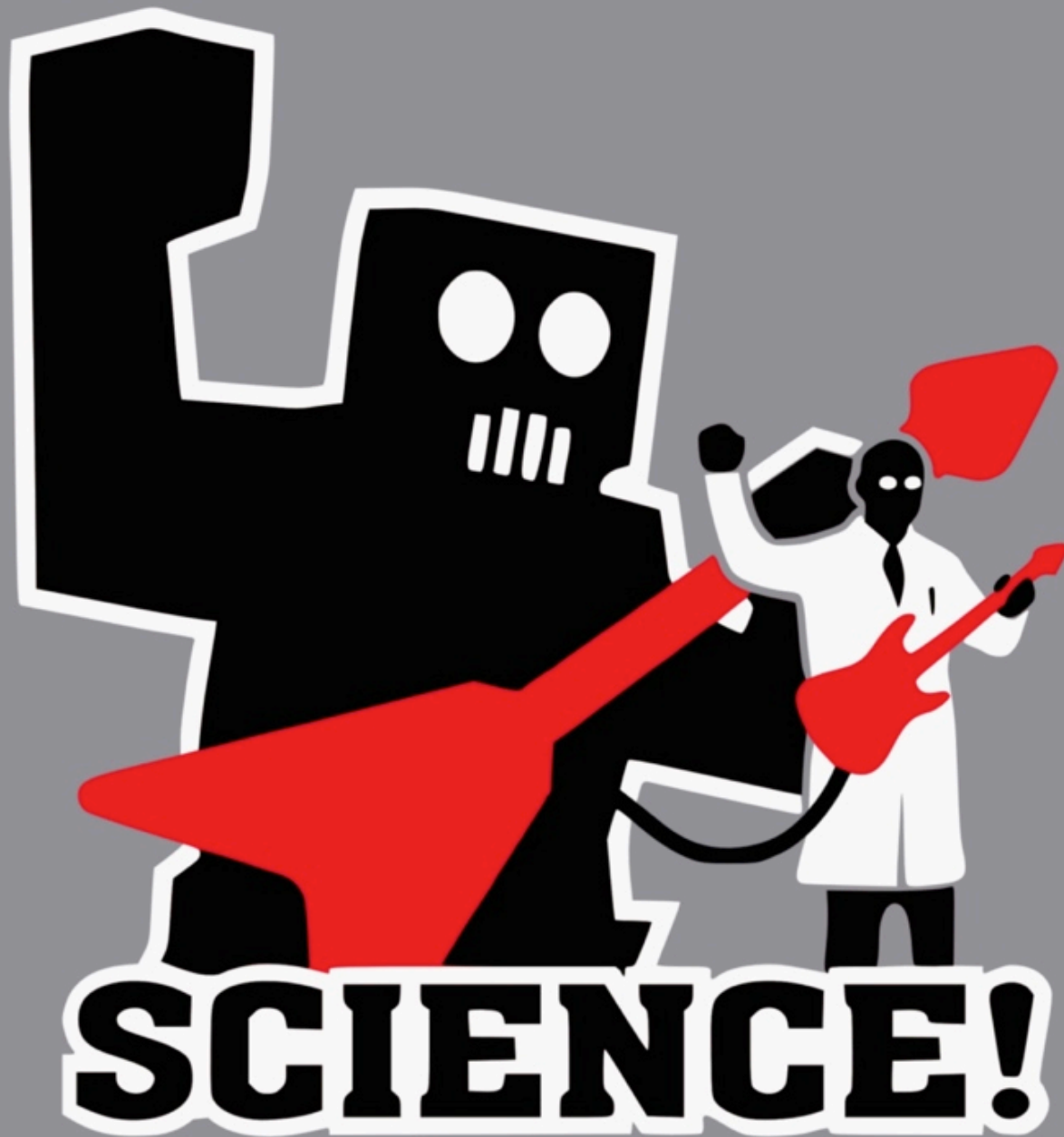
LMS = Lightweight Modular Staging

- An extensible compiler framework
- Implemented as a Scala library
- Execute 'now' vs 'compile and exec later'
- Specialize and compile program pieces at runtime

Staging = Multi-Stage Programming

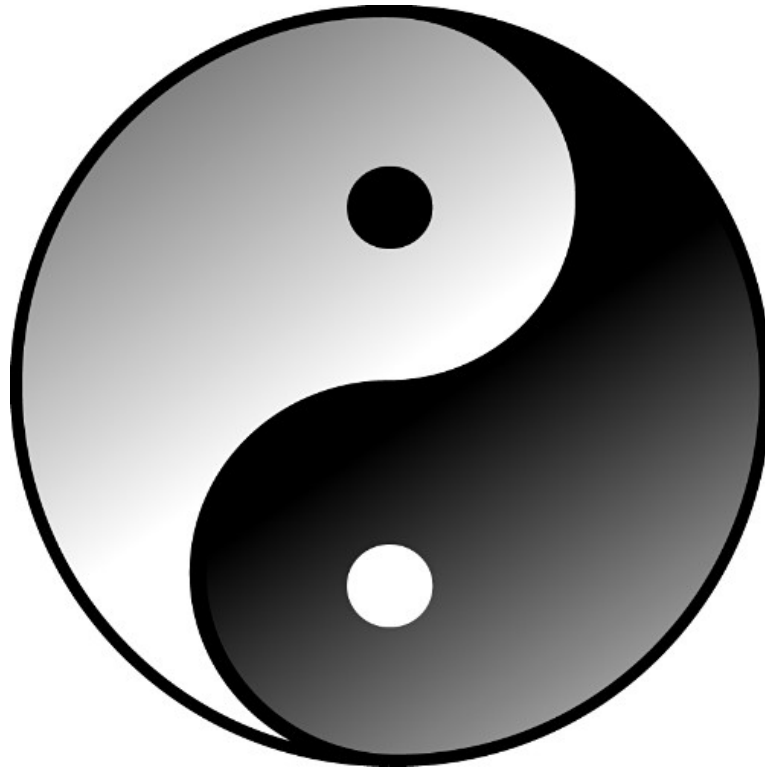
- Computations can generally be separated into stages (Jørring, Scherlis 1986), distinguished by:
 - frequency of execution
 - availability of data
- Multi-Stage Programming (Taha, Sheard 1997): make stages explicit in a program:
 - “delay” expressions to a generated stage
 - “run” delayed expressions
 - staged program fragments as first class values

Generative **METAPROGRAMMING**



(by Amorphia Apparel)

program generically ...



... and run specialized !

Projects / Collaborations

- **Delite (Stanford)**
Onward!'10, PACT'11, DSL'11, IEEE Micro 10/11, ECOOP'13, GPCE'13, TECS 4/14
 - DSLs and Big Data on heterogeneous devices
- **Spiral (ETH)** GPCE'13, ARRAY'14
 - Fast numeric libraries
- **LegoBase (EPFL DATA)** DCP'14, VLDB'14
 - Databases and query processing
- **Lancet (Oracle Labs)** PLDI'14
 - Integrate LMS with JVM / JIT compilation
- **Hardware (EPFL PAL)** FPT'13, FPL'14
 - Domain specific HW synthesis
- **Parser Combinators (EPFL LAMP)** OOPSLA'14
 - Protocols and dynamic programming
- **JavaScript (EPFL, INRIA Rennes)** ECOOP'12, GPCE'13
 - LMS for the web



LMS = Lightweight Modular Staging

- Int, String, T
 - "execute now"
- Rep[Int], Rep[String], Rep[T]
 - "generate code to execute later"
- if (c) a else b -> __ifThenElse(c,a,b)
 - "language virtualization"
- Extensible IR, transformers, loop fusion, ...
- "Batteries included"

Example: Matrix

```
class Matrix[T:Numeric:Manifest](val rows: Rep[Int], val cols: Rep[Int]) {  
  private val arr: Rep[Array[T]] = ArrayNew[T](rows * cols)  
  private val num = implicitly[Numeric[T]]; import num._  
  
  def apply(i: Rep[Int], j: Rep[Int]): A =  
    arr(i*cols + j)  
  
  def update(i: Rep[Int], j: Rep[Int], e: Rep[A]): Unit =  
    arr(i*cols + j) = e  
  
  def *(that: Matrix[T]) = {  
    val res = new Matrix[T](this.rows, that.cols)  
  
    for (i <- 0 until this.rows) {  
      for (j <- 0 until that.cols) {  
        for (k <- 0 until this.cols)  
          res(i, j) += this(i, k) * that(k, j)  
      }  
    }  
    res  
  }  
}
```

Matrices are “now”
objects, their data
arrays are “later”
objects

Generate Low-Level Code

```
var x27 = 500 * 500
var x28 = new Array[Double](x27)
var x29: Int = 0
while (x29 < 500) {
  var x30: Int = 0
  while (x30 < 500) {
    var x31: Int = 0
    while (x31 < 100) {
      ...
      x31 += 1
    }
    var x46 = ()
    x46
    x30 += 1
  }
  var x47 = ()
  x47
  x29 += 1
}
```

(still far from optimal:
should block loops for locality)

```
val m = randomMatrix(500, 100)
val n = randomMatrix(100, 500)
```

```
val p = m * n
```

```
--- generic took 2.691s
--- generic took 1.4s
--- generic took 1.464s
--- generic took 1.359s
--- generic took 1.244s
```

```
--- double took 1.062s
--- double took 1.228s
--- double took 1.076s
--- double took 1.03s
--- double took 1.076s
```

```
--- staged took 0.088s
--- staged took 0.058s
--- staged took 0.055s
--- staged took 0.054s
--- staged took 0.056s
```

20x!

User code

```
println(...)
val mystring = ... // Rep[String]
println(mystring.length)
```

DSL interface

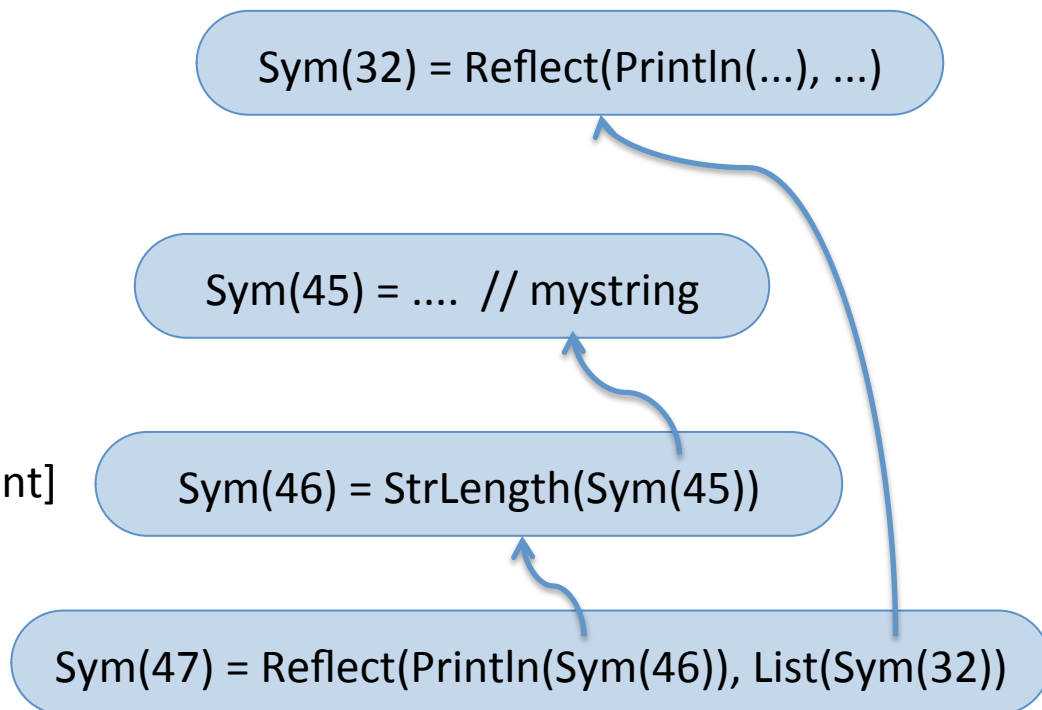
```
type Rep[T]
def infix_length(s: Rep[String]): Rep[Int]
def println(x: Rep[Any]): Rep[Unit]
```

DSL Implementation

```
type Rep[T] = Exp[T] // Sym[T] | Const[T]
```

```
case class StrLength(s: Exp[String]) extends Def[Int]
case class Println(s: Exp[Any]) extends Def[Unit]
```

```
def infix_length(s: Exp[String]): Exp[Int] = s match {
  case Const(s) => Const(s.length)
  case _ => reflectPure(StrLength(s))
}
def println(x: Exp[Any]): Exp[Unit] = reflectEffect(Println(x))
```



Demo Time

<https://scala-lms.github.io/tutorials/shonan.html>

<http://scala-lms.github.io/tutorials/linq.html>

A photograph of the Oracle campus, featuring several large, curved, blue glass buildings with white horizontal bands. The buildings are situated along a body of water, with a rocky shoreline in the foreground. The sky is clear blue. The word "ORACLE" is visible on one of the buildings.

LMS for Efficient Data Processing

**Yannis Klonatos, Christoph Koch (EPFL),
Hassan Chafi, Tiark Rompf (Oracle)**

Databases: State of the Art

- Popular generic DBMSs consist of > 1M lines of optimized C code
- Manual specialization for performance – e.g. PostgreSQL:
 - 20 implementations of memory page abstraction
 - 7 implementations of B-trees
- Difficult to adapt
 - e.g. disk based → in memory
- Still 10 – 100x slower than hand-written queries
(Stonebraker: time for a complete rewrite, Zukowski: Monetdb/x100)
- Commercial DBMS interpret query execution plans
 - some research on query compilation using LLVM (e.g. HyPer)

LegoBase

- New in-memory DB query engine, written in Scala
- Staged query interpreter
 - Compiles query execution plans (from Oracle DB) to C code
 - Supports all 22 TPCB queries
 - ~3000 lines of Scala code
- Use LMS for additional optimizations
 - Operator inlining
 - Optimizing data structures
 - Optimizing control flow (push vs pull)

It is indeed possible to write high performance systems in a high level language



2014

Very Large Data Base Endowment Inc.
(VLDB Endowment)

VLDB Best Paper Award
presented to

Tiark Rompf

for the paper entitled
**Building Efficient Query Engines in
a High-Level Language**

46th International Conference on Very Large Data Bases
September 1st-5th, Hangzhou, China

A SQL engine in 500 LOC

<https://scala-lms.github.io/tutorials/query.html>

Data is not only stored
but also transferred

Efficient, hand-optimized HTTP parser

```
switch (s) {  
  case s_req_spaces_before_url:  
    if (ch == '/' || ch == '*') {  
      return s_req_path;  
    }  
    if (IS_ALPHA(ch)) {  
      return s_req_schema;  
    }  
    break;  
  
  case s_req_schema:  
    if (IS_ALPHA(ch)) {  
      return s;  
    }  
  
    if (ch == ':') {  
      return s_req_schema_slash;  
    }  
    break;  
}
```

- Originally part of Nginx, later Node.js
- 2000+ lines of code
- Callbacks for header names/values triggered
- State-machine like code
- “Flat” code, loops/conditions

Staged Parser Combinators

```
def status: Parser[Int] =  
  ("HTTP/"~decimalNumber)~>wholeNumber<~(wildRegex~crLf) ^^ (_.toInt)  
  
def header: Parser[Option[(String,Any)]] =  
  (headerName<~":")~(wildRegex<~crLf) ^^ {  
    case hName~prop => collect(hName.toLowerCase, prop)  
  }  
  
def headers = rep(header)  
  
def response = status ~ headers  
  
...
```

- 200+ lines of code
- Fairly easy to change behaviour of a parser
 - Ex: ~ vs <~ vs ~>

Staged Parser Combinators

