# Julia A Fast Dynamic Language for Technical Computing

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## Why Julia?

Dynamic languages are extremely popular for numerical work:

- Matlab, R, NumPy/SciPy, Mathematica, etc.
- very simple to learn and easy to do research in

However, all have a "split language" approach:

- high-level dynamic language for scripting low-level operations
- C/C++/Fortran for implementing fast low-level operations

Libraries in C — no productivity boost for library writers

Forces vectorization — sometimes a scalar loop is just better

### **Three Key Features**

Sophisticated *dynamic* type system exposed in the language

- dependent parametric types; abstract types, a.k.a. traits
- polymorphism of all kinds ad hoc, parametric & duck

Multiple dispatch as the core unifying paradigm

- when well-implemented, fast & ubiquitous, it is qualitatively different
- many features can be seen as special cases of multiple dispatch

One language for a broad spectrum of programming levels

- a\*b can compile down to a single machine instruction
- a\*b can start a computation on a cluster of 1000s of machines

### Low-Level Code

```
function qsort!(a,lo,hi)
    i, j = lo, hi
    while i < hi
        pivot = a[(lo+hi)>>>1]
        while i <= j
            while a[i] < pivot; i = i+1; end
            while a[j] > pivot; j = j-1; end
            if i <= j
                a[i], a[j] = a[j], a[i]
                i, j = i+1, j-1
            end
        end
        if lo < j; qsort!(a,lo,j); end</pre>
        lo, j = i, hi
    end
    return a
end
```

### **Medium-Level Code**

```
function randmatstat(t,n)
    v = zeros(t)
    w = zeros(t)
    for i = 1:t
        a = randn(n, n)
        b = randn(n, n)
        c = randn(n, n)
        d = randn(n, n)
        P = [a b c d]
        Q = [a b; c d]
        v[i] = trace((P'*P)^4)
        w[i] = trace((Q'*Q)^4)
    end
    std(v)/mean(v), std(w)/mean(w)
end
```

### **High-Level Code**

```
function copy to(dst::DArray, src::DArray)
    @sync begin
        for p in dst.pmap
        end
    end
    return dst
end
function copy to(dest::AbstractArray, src)
    i = 1
    for x in src
        dest[i] = x
        i += 1
    end
    return dest
end
```

### @spawnat p copy\_to(localize(dst), localize(src,dst))

## Calling C/Fortran

getpid() = ccall(:getpid, Uint32, ())

system(cmd::String) = ccall(:system, Int32, (Ptr{Uint8},), cmd)

libfdm = dlopen("libfdm") besselj0(x::Float64) = ccall(dlsym(libfdm, :j0), Float64, (Float64,), x)

function fill!{T<:Union(Int8,Uint8)}(a::Array{T}, x::Integer)</pre> ccall(:memset, Void, (Ptr{T},Int32,Int), a, x, length(a)) return a end

### **The Numeric Promotion Dilemma**

- Most languages allow you to mix numeric types
  - not having this gets annoying very quickly Ada, OCaml (?), assembly

Traditional solution is to build promotion rules into the language

- otherwise too slow
- but doesn't work for user-defined types

Ideally want something generic, extensible & fast

1 + 2.5 0.5 + 3im ...

### **Promotion via Multiple Dispatch**

**Built-in definitions:** 

function promote{T,S}(x::T, y::S)
 P = promote\_type(T,S)
 convert(P,x), convert(P,y)
end

+(x::Number, y::Number) = +(promote(x,y)...)

When adding a new type:

## Type System — What's Normal

Nominative type hierarchy

Bits types, composite types, abstract types

Tuple types (argument lists)

Union types

Types have parameters (invariant) Rational {Int32}(1,2) Rational(1,2)

### Type System — What's Unusual

Parametric methods and "type patterns"

 $r{T<:Integer}(x::T, y::T) = Rational{T}(x,y)$ r(1,2)

Singleton kinds

sizeof(::Type{Int16}) = 2 sizeof(Int16)  $\Rightarrow$  2

Example — matrices that can be passed to LAPACK:

typealias StridedMatrix{T,A<:Array}</pre> Union(Matrix{T},SubArray{T,2,A})

## **Dynamic Type Inference**

Tags, not types

Tries to "guess" the tags

Entirely run-time semantics

Can improve the algorithm without updating the spec

## **Dynamic Type Inference**

- Abstract interpretation of lowered form:
  - assignments, calls, conditional branches, exception handlers
- Apply type transfer functions to handle calls
- Small set of primitives with simple, known t-functions
- The t-function for generic functions is

$$T(f, t_{\operatorname{arg}}) = \bigsqcup_{(s,g) \in f}$$

 $T(g, t_{\operatorname{arg}} \sqcap s)$ 

### **Key Optimizations**

- Aggressive method specialization
- Lots of inlining
- Elimination of apply()
  - apply(f, (a, b))  $\Rightarrow$  f(a,b) apply(f, t::(T,S))  $\Rightarrow$  f(t[1], t[2])

Multiple value cons-elimination

 $(a, b) = (f(), g()) \Rightarrow t1 = f(); t2 = g()$ a = t1; b = t2

### Performance



### Disadvantages

Method ambiguities

can print a very specific warning (using type intersection)

Generated code, compiler data structures and type information take up memory

realistically, can't run Julia in < 200Mb today</p>

About 144 bytes/LOC in the library

Building from scratch is slow

~15 sec system image build time to prime the cache (but done off-line)

Modularity is a bit tricky with multiple dispatch

Type info only flows "forward" — no return type overloading

### **People Like It!**

"Frustrated matlab and R user wanting a language that doesn't sacrifice performance."

"Where has Julia been this past two years!? I had searched for it high and low, day and night, to the point of nearly driving myself insane."

"I'm having a lot of \*fun\* (productive fun!) using Julia and hope to be able to contribute."

"...everything I wished I'd had in MATLAB and for data analysis for years now..."

"I'm really excited that you're building a language that looks very much like what I've wanted for over ten years now."

### **Project Statistics**

Hundreds of popular numerical functions

Getting traction as an open-source project:

- ► 350,000+ page views
- 100,000+ visitors
- ► 3,000+ downloads
- 1,000+ GitHub followers
- ► 40+ contributors
- ▶ 4+ Stefans

http://julialang.org/