

two_language_problem

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Introduction to Julia: Solving Two-Language Problem

1 1. Two-Language Problem

- Dynamic languages (e.g., Python, R, Matlab) tend to be slow
- Lower-level, compiled languages (e.g., C, C++, Fortran) are very fast, but are more time-consuming to write and debug

1.1 The Standard Approach - Write core algorithms in lower-level language - Then wrap that in higher-level language using some interface language/package (e.g., Cython, Rcpp)

1.1 1.2 What's the Problem?

1. Higher barrier to entry for package authors
2. Package authors must know (at least) two languages
3. Creates a sharp divide between package authors and package users

Meet Julia!

2 2. Julia Language

1. Dynamic technical computing language
2. Multi-paradigm (functional, imperative, "OO"-ish)
3. Just-in-time compiled using LLVM
4. Under active development (current stable version: 1.5.3)
5. Fast!
6. Fun!

2.1 2.1 History of Julia

- Started in 2009 as PhD thesis of Jeff Bezanson at MIT
- Jeff has been collaborating with Stefan Karpinski and Viral Shah
- First official release in 2012
- Influenced by: C, Python, R, Ruby, Lisp
- Specifically designed to be as fast as C and as expressive as Python or Ruby

2.2 2.2 Julia Basics

- Similar to Python and R in that:
 - Very flexible
 - Has elements of imperative, functional, and object-oriented programming
 - Functions are first-class citizens
 - * (e.g., pass as arguments, return from other functions)

2.2.1 2.2.1 Julia Basics (cont.)

- Differs from Python and R in that:
 - Compiled, not interpreted
 - Not object-oriented in the classical sense
 - No classes with private data and private methods
 - No concept of inheritance
 - Designed with parallelism in mind
 - Multiple dispatch
 - Excellent for generic programming
 - Metaprogramming

2.3 2.3 Base Language Data Types

1. Primitives

- Any numeric type you can imagine: Int64, Float64, BigInt, Complex, Irrational, Rational
- Many abstract types: Any, Real, Number, Integer
- String and Chars

2. Container Types

- Array (vectors, matrices, N-dimensional arrays)
- Set
- Dict
- Tuple and NamedTuple

3. Composite Types

- struct

In [3]: *# Simple function that computes Fibonacci numbers*

```
function fib(n)
    nums = ones{Int, n}
    for i = 3:n
        nums[i] = nums[i - 1] + nums[i - 2]
    end
    return nums[n]
end
```

```
Out[3]: fib (generic function with 1 method)
```

```
In [4]: fib(50)
```

```
Out[4]: 12586269025
```

```
#  
Julia is Fast!
```

```
In [5]: function qsort(a, lo, hi)  
        i = lo  
        j = hi  
        while i < hi  
            pivot = a[div(lo + hi, 2)]  
            while i <= j  
                while a[i] < pivot  
                    i += 1  
                end  
                while a[j] > pivot  
                    j -= 1  
                end  
                if i <= j  
                    a[i], a[j] = a[j], a[i]  
                    i += 1  
                    j -= 1  
                end  
            end  
            if lo < j  
                qsort(a, lo, j)  
            end  
            lo = i  
            j = hi  
        end  
        return a  
    end
```

```
Out[5]: qsort (generic function with 1 method)
```

```
###  
R quicksort example
```

```
In [7]: # quick sort in Julia  
        using BenchmarkTools  
  
        n = 100_000  
        x = randn(n)  
  
        @time x2 = qsort(x, 1, n);  
  
0.010187 seconds
```

2.4 Understanding Julia's Speed

1. Type system makes inferring concrete types easy
2. Julia aggressively specializes run-time types
3. LLVM generates fast native code
4. Very smart people working hard!

Parallelism in Julia

2.5 Parallel Programming in Julia

1. Can use multi-processing or multi-threading
2. Excellent support for SIMD operations
3. SharedArray data type in base language
4. DistributedArray data structure allows for Hadoop-like distributed computing

Suppose we want to constrain range of values a vector's elements can take to be between -0.5 and 0.5

```
In [8]: function squeeze_range!(x)
        n = length(x)
        for i = 1:n
            if !(-0.5 < x[i] < 0.5)
                x[i] = 0.5 * sign(x[i])
            end
        end
    end
```

```
Out[8]: squeeze_range! (generic function with 1 method)
```

```
In [9]: n = 100_000_000
        v = randn(n)
```

```
@time squeeze_range!(v)
```

```
0.599651 seconds (20.43 k allocations: 1.113 MiB)
```

2.5.1 Using Threads for Parallelism

```
In [10]: # NOTE: JULIA_NUM_THREADS env variable is set in .bash_profile
        using Base.Threads
```

```
function thr_squeeze_range!(x)
    n = length(x)
    @threads for i = 1:n
        if !(-0.5 < x[i] < 0.5)
            x[i] = 0.5 * sign(x[i])
        end
    end
end
```

```
        end
    end
```

```
Out[10]: thr_squeeze_range! (generic function with 1 method)
```

```
In [12]: n = 100_000_000
        v = randn(n)
```

```
@time thr_squeeze_range!(v)
```

```
0.123301 seconds (36 allocations: 4.281 KiB)
```

```
#
Integration with other Languages
```

2.6 Julia and other Languages

Julia has zero-cost interfaces with many technical computing languages:

1. PyCall.jl
 2. RCall.jl
 3. ccall()
 4. Cxx.jl
 5. Matlab.jl
- Introduction to Julia

2.7 Arithmetic

```
In [13]: 3 + 7      # addition
```

```
Out[13]: 10
```

```
In [14]: 10 - 3     # subtraction
```

```
Out[14]: 7
```

```
In [15]: 20 * 5     # multiplication
```

```
Out[15]: 100
```

```
In [16]: 100 / 10   # division
```

```
Out[16]: 10.0
```

```
In [17]: 10 ^ 2     # exponentiation
```

```
Out[17]: 100
```

2.8 Logical Operators

```
In [ ]: false && true # logical AND
```

```
In [19]: false || true # logical OR
```

```
Out[19]: true
```

2.9 Comparisons

```
In [20]: 1 == 1.0 # Equality
```

```
Out[20]: true
```

```
In [21]: 3 <
```

```
Out[21]: true
```

```
In [22]: 1 <= 1
```

```
Out[22]: true
```

2.10 Assignment

Assignment in Julia is done with the single `=`. All it does is associates a name (on the left) to a value (on the right).

```
In [ ]: x = 1
```

```
In [23]: x = "hello!"
```

```
Out[23]: "hello!"
```

2.11 Arrays

Julia has highly efficient multidimensional arrays, both constructed and indexed with square brackets.

```
[item1, item2, ...]
```

```
In [24]: # create array (vector) with []  
squares = [1, 4, 9, 15, 25]
```

```
Out[24]: 5-element Array{Int64,1}:
```

```
 1  
 4  
 9  
15  
25
```

```
In [25]: squares[1] # indexing is similar to R
```

```
Out[25]: 1
```

```
In [26]: squares[1:3] # slicing is same as R
```

```
Out[26]: 3-element Array{Int64,1}:
```

```
 1  
 4  
 9
```

2.12 Loops

The syntax for a for loop is

```
for *var* in *loop iterable*  
    *loop body*  
end
```

```
In [27]: for x in 1:3  
        println("yo")  
        end
```

```
yo  
yo  
yo
```

```
In [28]: x = 3  
        while x < 6  
            println(x)  
            x += 1  
        end
```

```
3  
4  
5
```

3 if/else

```
In [29]: if 4 > 2  
        println("potato!!")  
        end
```

```
potato!!
```

```
In [30]: if 3 < 0.4  
        println("no")  
        else  
            println("soup!!")  
        end
```

```
soup!!
```

4 Functions in Julia

```
In [31]: function add_one(a)
           # function body
           b = a + 1
           return b
       end
```

```
Out[31]: add_one (generic function with 1 method)
```

```
In [32]: add_one(42)
```

```
Out[32]: 43
```