Julia Programming Language Benchmark Using a Flight Simulation

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- Julia is a relatively new computer language that aims to reduce the challenge for mathmodelers to develop fast computer tools and simulations. It potentially combines the ease-of-coding feature of scripting languages (like Python) with the performance of compiled languages (like C++).
- A key question for Julia application to the simulation domain is, "Can Julia, with its obvious coding simplicity, provide runtime speeds comparable to conventional compiled languages for flight simulation?"
- A unique combination of existing elements can be employed to address the previous question
 - Extensively documented object-oriented simulation architecture
 - Industry standard rocket flight simulation
 - Separate versions (C++, Java, and Python) already benchmarked

Julia Language Overview

Key Features:

- "Pythonic" syntax
- Dynamically-typed
- Built-in REPL (read-evaluate-print-loop)
- JIT (just-in-time) compilation
- Built-in library support for multi-threading, multi-core, and distributed processing
- Direct calling of C and Fortran code without "glue" code
- Automatic garbage collection (i.e. memory leak control)
- Easy extension to multi-processing

Practical Considerations:

- Designed for technical computing (like MATLAB & Fortran)
- Free and open-source
- High-level and easy-to-learn (like MATLAB & Python)
- High-performance (like C, C++, and Fortran)



Mini-Rocket Description



DESCRIPTION:

- A missile trajectory generator
- Desktop PC tool
- Object-oriented for maximum versatility
- Easy to use
- Modern coding structure
- · Useful as component in larger system models
- Developed and used for over 30 years for missiles ranging from tactical to strategic

Very easy to configure five degree-of-freedom missile flyout model that accurately generates trajectories in threedimensional space, including maneuver characteristics without 6DOF overhead

FEATURES:

- Very fast running with unique *osculating-plane* formulation without the overhead of rigid body equations-of-motion (runtime speed of a 3-DOF with additional two degrees-of-freedom)
- Motion in three-dimensional space
- Two independent channels (pitch and yaw) for steering and guidance
- 1-d, 2-d, or 3-d table lookups to model aerodynamics and propulsion characteristics
- Capability to model angle-of-attack variations in lift and drag
- Constraints on lateral acceleration based on angle-of-attack and closed-loop airframe response time
- Detailed models for control and guidance subsystems not required

PAST APPLICATIONS:

- Generate trajectory given missile aerodynamic, mass, and propulsion parameters
- Generate family of trajectories (off-line) for higher-level system simulations or system time-line studies
- In-line missile (offensive or defensive) model for engagement simulations
- In-line missile dynamics model for HWIL
- Tool for quickly examining different guidance laws
- Tool for mapping battlespace engagement constraints
- Trajectory reconstruction/flight characteristics estimation

EXTENSIVE LEGACY OF APPLICATION (partial list)

Strategic

- Endo-interceptors (HEDI, ARROW, THAAD)
- Exo-interceptors (ERIS, GBI, E2I)
- Anti-Satellite boosters (KE-ASAT)
- Strategic Target Vehicles (STARS, ARES)
- Numerous booster survey studies

Tactical

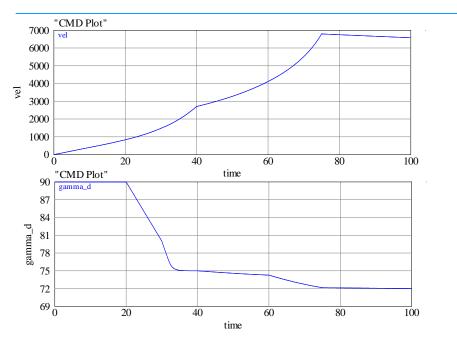
- NLOS-PAM
- EAPS
- Future Missile
- HAPS
- Modular Missile Technology
- IDEEAS constructive simulation (embedded interceptor)

Mini-Rocket Case for Benchmark

Benchmark Rocket

- 3-stage hypothetical rocket (to avoid SBU concerns)
- vertical launch with pitch-over
- rotating earth
- · pre-programmed maneuver in pitch and yaw channels
- lift-off weight = 2500 kg
- flight time = 100 sec
- stage splits = 40, 75 at 2708, 6790 m/sec
- final velocity = 6574 m/sec at t = 100 sec
- Benchmark case replicated for all three languages: C++, Python, Java

Benchmark Trajectory



 a case was chosen to illustrate 5DOF maneuver model (i.e. 6DOF minus roll)

30000

40000

50000

• pitch-over (red)

0

200000

150000

50000

0

Ζ

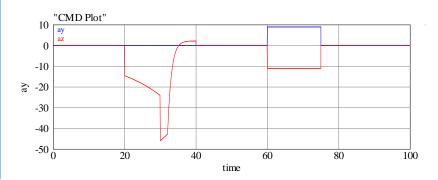
out-of-plane yaw (blue)

10000

20000

Y

 start pitch over at 20 sec to 75 deg, then execute dual pitch/yaw ~1g maneuvers for 15 sec



10000

20000

∠ 30000 x 40000

50000

Julia Mini-Rocket Coding Highlights – Object Oriented

DIFFERENTIAL EQUATION (DE) ENGINE (the "main" program) **OBJECT-ORIENTED** all simulation entities are a hierarchy of objects #### SIMULATION clock mr = build rocket("rocket.dat", "**** rocket ****") rocket integrators = [mr.myclock, mr.mass, mr.vmx, mr.vmy, mr.vmz, rocket stages mr.pmx, mr.pmy, mr.pmz] · atmosphere model etc. down-the-hierarchy for ii in 1:1 # MC LOOP **Differential Equation Engine orchestrates execution** clock = Clock1(0.0, 0.01)init(mr) println("Launch!!!") # CREATE ATMOSPHERE OBJECT while clock.t <= 100.0 + 1e-6 mr.atm = Atm62()#### UPDATE update(mr, clock) # CREATE STAGE OBJECTS #### REPORT stage1 = stage read(fname, "**** stage 1 ****") report(mr, clock) stage2 = stage read(fname, "**** stage 2 ****") #### PROPAGATE stage3 = stage read(fname, "**** stage 3 ****") [propagate(state, clock) for state in integrators] mr.stages = [stage1, stage2, stage3] update(clock) end end

TABLE OBJECT CREATION AND USAGE

```
# CREATE TABLE OBJECT WITHIN STAGE OBJECT
txv_table = table1_read( fname, "txv_table", line0)
ca_off_table = table2_read( fname, "ca_off", line0)
ca_on_table = table2_read( fname, "ca_on", line0)
cna_table = table1_read( fname, "ca_on", line0)
guide_mode_table = table1_read( fname, "guide_mode_table", line0)
ay_table = table1_read( fname, "ay_table", line0)
az_table = table1_read( fname, "ay_table", line0)
gammad_table = table1_read( fname, "gammad_table", line0)
gamma_table = table1_read( fname, "gammad_table", line0)
...
# ACCESS TABLE OBJECT TO INTERPOLATE
txv = interp( txv table, clock.t)
```

MINI-ROCKET JULIA BUILD PROCESS

- Simulation components built incrementally starting with DE engine and support elements (tables, atm. model, stages, integrator (clock), ...)
- Extensive experimentation with coding techniques and structures to leverage Julia features with emphasis on code readability and then timing
- Benchmark timing studies conducted with DE engine and table elements to understand optimum Julia coding practice for speed
 - Multiple, independent 2nd order transfer functions ran in parallel to prototype different DE engine representations and numerical integrators
 - Large-scale table lookup benchmarks conducted to optimize speed (without sacrificing readability)
 - After much experimentation, Julia was ~1/2 as fast as equivalent C++ code for DE and table look-up benchmarks

Lines-of-Code Comparison

	C++	Java	Python	Julia ⁴
DE Engine ¹	360	310	227	98 ²
rocket model	830	753	596	613
vector utilities	533	524	351	0 ³
table utilities	650	384	252	165
misc. utilities	104	153	64	82
TOTAL	2477	2124	1490	958

NOTES:

1. All models coded to same O-O simulation kernel architecture

<u>Reference</u>

Sells, H.R., A Code Architecture to Streamline the

Missile Simulation Life Cycle

AIAA Modeling and Simulation Technologies Conference, 12 January 2017

- 2. Julia DE engine does not encompass full functionality for OSK train-of-objects built into other DEs. This lack of functionality was not relevant for this study.
- 3. Julia built-in linear algebra functionality used
- 4. Julia characters-of-code would have been GREATLY reduced without specifying types (at cost of great speed penalty and code readability)

	1 run		10 runs	5	100 rur	15
C++ (CMD) ² Borland C++ 5.5	0.125 1.0	00	0.126	1.01	0.127	1.02
Java 12.0.1	0.169 1.3	35	0.139	1.11	0.099	0.79
Python 2.7.16	3.765 <mark>30</mark>	0.12	3.752	30.02	3.762	30.10
Julia 1.1.1	1.470 <mark>11</mark>	1.76	0.378	3.02	0.269	2.15

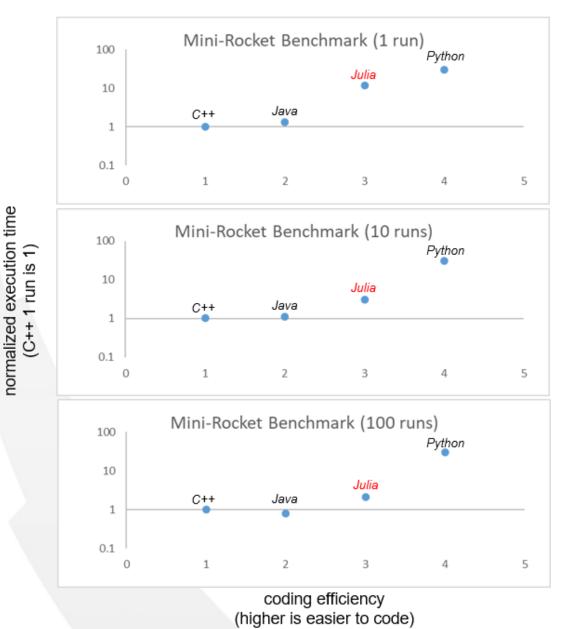
colored numbers normalized to C++ single run

- initial data file read not included in timing loop
- small amount of screen output redirected to buffered file (SSD)
- no explicit compiler optimization flags for C++ and Java
- all cases run on Windows 10, Dell Precision 7820, Wintel High Performance Dual Socket Engineering Workstation, Intel Xeon Gold 6130 CPU @ 2.10GHz (2 processors, 32 cores each), normal single-thread processing used for these runs

References:

- 1. "Mini-Rocket User Guide", U.S. Army Technical Report AMR-SS-07-27, online link https://apps.dtic.mil/docs/citations/ADA472173
- 2. "C++ Model Developer", U.S. Army Technical Report, U.S. Army Technical Report AMR-SG-05-12, online link https://apps.dtic.mil/docs/citations/ADA433836

Benchmark Plots – Mini-Rocket



- Nature of Just-In-Time (JIT) is evident for Java and Julia
- JIT compile on-the-fly is more efficient for large # runs (JIT compilation consumes less amount of total run time)
- Surprise #1: Java as efficient (or better than C++) – previous experience was Java ~ 4x slower*
- Surprise #2: Python much better than previous experience ~ 100x slower*

*my personal experience as well as generally-accepted independent benchmarks

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- Julia has very easy-to-extend functionality for utilizing multiple cores.
- The Julia Distributed package adds functionality to extend the single execution thread used-to-now to multiple cores.
- All cases run on Dell Precision 7820, Wintel High Performance Dual Socket Engineering Workstation, Intel Xeon Gold 6130 CPU @ 2.10GHz (2 processors, 32 cores each)

configuration	# concurrent runs	time (sec)	time/run
1 run/core	65 <mark>1</mark>	3	0.046
10 runs/core	650	7.627	0.0117
100 runs/core ³	6500	53	0.00815
1000 runs/core ³	65000	525	0.00808

NOTES:

- 1. 65 runs executed concurrently including the master process and its 64 worker processes.
- 2. All output turned off, except final report
- 3. More runs per core appears to reduce overhead of distributing runs

Wrap-Up

- Challenges still exist in providing the tools and environment for quickly and efficiently constructing dynamical system simulations that address every step in the missile & rocket simulation life cycle
- The potential contribution of Julia is to give "non-expert" coders (scripters) the ability to build high performance simulations
- Julia was well suited to coding the object-oriented structure in MR with an exceptional economy-of-code
- Julia execution speed was much faster than Python but still slower than C++ and Java

Path Forward

- The economy of Julia to express complex programming constructs makes it attractive as a simulation experiment "testbed" for prototyping any future simulation applications
- Although only touched upon in these results, parallel computing capability and its application to time-domain dynamic system simulations is especially compelling for further flight simulation research