Astro 585: HW 4

Codename: The Maxwell-Jüttner Distribution

February 21, 2014

My git repository is here: https://github.com/hsgg/astro585, clone URL https://github.com/hsgg/astro585.git.

1 Profiling

```
#!/usr/bin/env julia
# Calculate a normalized Gaussian. Write it in this odd way, so we can see the
# profiler output better
function log_Gaussian(z, y, sigma)
   tmp1 = y - z
   tmp1 ./= sigma
   tmp1 = tmp1.^2
    tmp1 .*= -0.5
    tmp2 = sqrt(2 * pi)
    tmp2 .*= sigma
    tmp2 = log(tmp2)
    tmp1 -= tmp2
   return tmp1
end
function log_likelihood(log_pdf, y::Array, sigma::Array, z::Array)
   n = length(y)
    @assert n == length(sigma)
    Qassert n == length(z)
    sum = zero(y[1])
    for i in 1:n
        a = y[i]
        b = z[i]
        c = sigma[i]
        tmp = log_pdf(a, b, c)
        sum += tmp
    end
    return sum
end
```

```
# make data:
Nobs = 1000
srand(42)
z = zeros(Nobs)
sigma = 2. * ones(Nobs)
y = z + sigma .* randn(Nobs)
# profile:
Profile.clear()
t = @elapsed @profile (for i in 1:10^4;
        log_likelihood(log_Gaussian, y, sigma, z);
    end)
println("Time elapsed: ", t, " seconds")
Profile.print()
println()
# I expect the function calls to take the most time, and so that's where the
# profile is most often, 5800 times calling the function "log_Gaussian()", and
# 1715 times 'log()':
# julia> include("2_profile.jl")
# Time elapsed: 8.564663583 seconds
# 7690 boot.jl; include; line: 238
#
      7690 util.jl; anonymous; line: 42
#
             ...mp/hw4/2_profile.jl; log_likelihood; line: 21
        20
#
             ...mp/hw4/2_profile.jl; log_likelihood; line: 22
        15
#
        172 ...mp/hw4/2_profile.jl; log_likelihood; line: 23
             ...mp/hw4/2_profile.jl; log_likelihood; line: 24
#
#
             ...mp/hw4/2_profile.jl; log_likelihood; line: 25
#
        5800 ...mp/hw4/2_profile.jl; log_likelihood; line: 26
#
               ...mp/hw4/2_profile.jl; log_Gaussian; line: 11
#
           13
                ...p/hw4/2_profile.jl; log_Gaussian; line: 7
#
                ...p/hw4/2_profile.jl; log_Gaussian; line: 11
           1715 ...p/hw4/2_profile.jl; log_Gaussian; line: 12
#
                ...p/hw4/2_profile.jl; log_Gaussian; line: 14
#
#
                  inference.jl; typeinf_ext; line: 1092
#
                1 inference.jl; typeinf; line: 1196
                1 inference.jl; typeinf; line: 1382
#
#
                  1 inference.jl; inlining_pass; line: 1956
#
                    1 inference.jl; inlining_pass; line: 1972
#
                      1 inference.jl; inlining_pass; line: 2014
#
                        1 inference.jl; inlineable; line: 1832
#
                1 inference.jl; typeinf; line: 1385
#
                 1 inference.jl; tuple_elim_pass; line: 2244
```

```
1 inference.jl; find_sa_vars; line: 2181
#
#
        1589 ...mp/hw4/2_profile.jl; log_likelihood; line: 27
#
               float.jl; +; line: 132
          13
#
           8 inference.jl; typeinf_ext; line: 1092
              4 inference.jl; typeinf; line: 1259
#
                2 inference.jl; abstract_interpret; line: 958
#
#
                  2 inference.jl; abstract_eval; line: 814
#
                   2 inference.jl; abstract_eval_call; line: 789
                     2 inference.jl; abstract_call; line: 701
#
                       2 inference.jl; abstract_call_gf; line: 576
#
#
                          2 reflection.jl; _methods; line: 77
#
                              2 reflection.jl; _methods; line: 97
#
                                 2 reflection.jl; _methods; line: 97
                                    2 reflection.jl; _methods; line: 80
#
#
                2 inference.jl; abstract_interpret; line: 966
#
                 1 inference.jl; abstract_eval; line: 814
#
                  1 inference.jl; abstract_eval_call; line: 788
#
                  1 inference.jl; abstract_eval; line: 814
#
                   1 inference.jl; abstract_eval_call; line: 756
#
                    1 array.jl; getindex; line: 296
#
                     1 array.jl; copy!; line: 51
#
                      1 array.jl; unsafe_copy!; line: 37
#
              1 inference.jl; typeinf; line: 1379
#
                1 inference.jl; type_annotate; line: 1508
#
                  1 inference.jl; eval_annotate; line: 1471
#
                    1 inference.jl; eval_annotate; line: 1483
#
              3 inference.jl; typeinf; line: 1382
#
                3 inference.jl; inlining_pass; line: 1956
                  2 inference.jl; inlining_pass; line: 1972
#
#
                    1 inference.jl; inlining_pass; line: 1990
#
                      1 inference.jl; exprtype; line: 1662
#
                         1 inference.jl; abstract_eval; line: 798
#
                          1 inference.jl; abstract_eval_symbol; line: 910
#
                            1 inference.jl; abstract_eval_global; line: 898
#
                              1 inference.jl; abstract_eval_constant; line: 875
#
                    1 inference.jl; inlining_pass; line: 2014
#
                      1 inference.jl; inlineable; line: 1772
#
                         1 reflection.jl; _methods; line: 77
                             1 reflection.jl; _methods; line: 97
#
#
                                1 reflection.jl; _methods; line: 97
#
                                   1 reflection.jl; _methods; line: 80
#
                  1 inference.jl; inlining_pass; line: 2014
#
                    1 inference.jl; inlineable; line: 1910
#
                      1 inference.jl; sym_replace; line: 1560
                         1 inference.jl; sym_replace; line: 1560
#
```

```
1 inference.jl; sym_replace; line: 1560
# 8
       float.jl; +; line: 132
# Unfortunately, the the julia does not appear to inline the call to
# 'log_Gaussian()'. This would be nice. Indeed, inlineing it by hand results in
# a 5x speedup.
# Another way to speed it up would be to not loop in 'log_likelihood()', but
# use vectors in 'log_Gaussian()'. This might work, and also reduce the number
# of function calls to 'log()', but it could lead to many spurious memory
# allocations.
#
# Let's try that:
function log_likelihood_vec(log_pdf, y::Array, sigma::Array, z::Array)
    n = length(y)
    @assert n == length(sigma)
    @assert n == length(z)
    s = log_pdf(y, z, sigma)
    su = sum(s)
    return su
end
Profile.clear()
t = @elapsed @profile (for i in 1:10<sup>4</sup>;
        log_likelihood_vec(log_Gaussian, y, sigma, z);
    end)
println("Time elapsed: ", t, " seconds")
Profile.print()
# Nice, more thatn 2 times increase in speed:
# Time elapsed: 3.672488957 seconds
# 3205 boot.jl; include; line: 238
      3205 util.jl; anonymous; line: 72
#
        3143 ...mp/hw4/2_profile.jl; log_likelihood_vec; line: 65
#
               ...mp/hw4/2_profile.jl; log_Gaussian; line: 6
#
#
           73 array.jl; -; line: 135
           3 array.jl; -; line: 136
#
           8 array.jl; -; line: 926
#
#
           13 array.jl; -; line: 927
          1121 ...mp/hw4/2_profile.jl; log_Gaussian; line: 7
#
           901 broadcast.jl; ##broadcast_T_/#233; line: 187
#
#
                broadcast.jl; broadcast_shape; line: 27
                2 broadcast.jl; longer_size; line: 21
#
```

```
#
                  1 broadcast.jl; longer_size; line: 20
#
                  broadcast.jl; broadcast_shape; line: 29
#
               3 array.jl; fill!; line: 183
#
                  broadcast.jl; broadcast_shape; line: 42
#
           14 broadcast.jl; ##broadcast_T_/#233; line: 188
           204 broadcast.jl; ##broadcast_T_/#233; line: 189
#
#
             13 broadcast.jl; broadcast_args; line: 87
#
              1 array.jl; sizehint; line: 709
#
              5 broadcast.jl; calc_loop_strides; line: 63
#
              2 broadcast.jl; calc_loop_strides; line: 64
#
              1 broadcast.jl; calc_loop_strides; line: 66
#
               1 array.jl; sizehint; line: 709
#
              2 broadcast.jl; calc_loop_strides; line: 77
#
              2 broadcast.jl; calc_loop_strides; line: 83
#
                 broadcast.jl; broadcast_args; line: 88
               10 broadcast.jl; ##/_inner!#235; line: 129
#
#
                1 tuple.jl; ==; line: 81
#
                6 tuple.jl; ==; line: 84
#
                  1 promotion.jl; ==; line: 188
#
                   broadcast.jl; ##/_inner!#235; line: 131
#
               27 broadcast.jl; ##/_inner!#235; line: 136
#
               138 broadcast.jl; ##/_inner!#235; line: 166
#
               ...mp/hw4/2_profile.jl; log_Gaussian; line: 8
#
           572 array.jl; .^; line: 920
#
          100 ...mp/hw4/2_profile.jl; log_Gaussian; line: 9
#
           71 array.jl; .*; line: 135
#
             array.jl; .*; line: 136
#
           5 array.jl; .*; line: 944
           15 array.jl; .*; line: 945
#
#
               ...mp/hw4/2_profile.jl; log_Gaussian; line: 10
#
               ...mp/hw4/2_profile.jl; log_Gaussian; line: 11
#
           60 array.jl; .*; line: 135
#
           12 array.jl; .*; line: 136
#
             array.jl; .*; line: 937
#
           11 array.jl; .*; line: 938
#
           1 array.jl; .*; line: 941
#
          1058 ...mp/hw4/2_profile.jl; log_Gaussian; line: 12
#
           1058 operators.jl; log; line: 236
#
               ...mp/hw4/2_profile.jl; log_Gaussian; line: 13
#
           57 array.jl; -; line: 135
#
           3 array.jl; -; line: 136
#
           18 array.jl; -; line: 926
#
           10 array.jl; -; line: 927
#
             3 inference.jl; typeinf_ext; line: 1092
#
                1 inference.jl; typeinf; line: 1169
```

```
2 inference.jl; typeinf; line: 1382
#
                  2 inference.jl; inlining_pass; line: 1956
#
#
                    2 inference.jl; inlining_pass; line: 1972
#
                      2 inference.jl; inlining_pass; line: 2014
                        1 inference.jl; inlineable; line: 1808
#
                        1 inference.jl; inlineable; line: 1908
#
#
                          1 inference.jl; resolve_globals; line: 1624
#
                             1 inference.jl; resolve_globals; line: 1624
                              1 inference.jl; resolve_globals; line: 1624
#
#
             ...mp/hw4/2_profile.jl; log_likelihood_vec; line: 66
           43 abstractarray.jl; sum; line: 1487
#
#
            43 abstractarray.jl; sum_pairwise; line: 1481
             43 abstractarray.jl; sum_pairwise; line: 1481
#
              43 abstractarray.jl; sum_pairwise; line: 1481
#
#
               24 abstractarray.jl; sum_pairwise; line: 135
#
               17 abstractarray.jl; sum_pairwise; line: 1475
#
               2 abstractarray.jl; sum_pairwise; line: 1478
#
           3 inference.jl; typeinf_ext; line: 1092
#
              1 inference.jl; typeinf; line: 1259
                1 inference.jl; abstract_interpret; line: 966
#
#
                 1 inference.jl; abstract_eval; line: 814
#
                  1 inference.jl; abstract_eval_call; line: 788
#
              2 inference.jl; typeinf; line: 1382
#
                2 inference.jl; inlining_pass; line: 1956
#
                  1 inference.jl; inlining_pass; line: 1943
#
                   1 inference.jl; is_known_call; line: 2120
#
                  1 inference.jl; inlining_pass; line: 1994
# 1
       range.jl; colon; line: 36
       range.jl; colon; line: 38
# 1
    Loops, not Loops
#!/usr/bin/env julia
```

```
using Base.Test
using Devectorize
gaussian(x) = exp(-0.5 * x.^2) / sqrt(2 * pi)
# integrate_loop():
#
     - Calculates the integal from 'a' to 'b' of the function 'fn(x)', using no
#
       more than 'maxevals' function evaluations.
#
     - Returns a pair (I, E) similar to 'quadgk()', where I is the integral,
       and E an estimation of the error.
#
```

```
function integrate_loop(fn, a, b; maxevals=10^7)
    @assert maxevals >= 0
   N = maxevals
    sum = 0.0
   dx = (b-a)/(N+1)
    for i in 1:N
        sum += fn(a + i * dx)
    end
    I = sum * (b-a) / N
   # no idea if this is right, but it is in the right ballpark for the tests
    E = I / N
   return I, E
end
function integrate_vector(fn, a, b; maxevals=10^7)
   @assert maxevals >= 0
   N = maxevals
   x = linspace(a, b, N)
   y = gaussian(x)
   I = sum(y) * (b-a) / N
   E = I / N
   return I, E
end
function integrate_mapthenreduce(fn, a, b; maxevals=10^7)
    @assert maxevals >= 0
   N = maxevals
   x = linspace(a, b, N)
   y = map(gaussian, x)
   I = reduce(+, y) * (b-a) / N
   E = I / N
   return I, E
end
function integrate_mapreduce(fn, a, b; maxevals=10^7)
    @assert maxevals >= 0
   N = maxevals
   x = linspace(a, b, N)
    s = mapreduce(gaussian, +, x)
    I = s * (b-a) / N
```

```
E = I / N
    return I, E
end
function integrate_devectorize(fn, a, b; maxevals=10^7)
    # Hm, to install Devectorize package, julia wants to use the
    # --single-branch option to git clone, which git learned in version 1.7.10.
    # The Davey computers only have 1.7.9. We are almost there!
    # Doing this on my laptop now...
    @assert maxevals >= 0
   N = maxevals
   x = linspace(a, b, N)
    Odevec s = sum(exp(-0.5 .* x.^2) ./ sqrt(2 .* pi))
    I = s * (b-a) / N
    E = I / N
    return I, E
end
function test_integral(intfn; maxevals=10^7)
    a = 0.5
    numerically, ntol = intfn(gaussian, -a, a, maxevals=maxevals)
    analytically = erf(a/sqrt(2))
    @test_approx_eq_eps numerically analytically ntol
    if maxevals \geq 10^3
        @test_approx_eq_eps ntol 0.0 1e-4
    elseif maxevals >= 10^6
        @test_approx_eq_eps ntol 0.0 1e-8
    end
end
function test_integral_function(intfn)
    # test at least one odd case
    @test_throws intfn(-0.3, 0.4, maxevals=-1)
    # test the results
    test_integral(intfn, maxevals=0)
    test_integral(intfn, maxevals=1)
    test_integral(intfn, maxevals=2)
    test_integral(intfn, maxevals=3)
    test_integral(intfn, maxevals=10^4)
    test_integral(intfn, maxevals=10^6)
end
```

```
function time_integral_func(intfn, integrandfn;
       maxcalls=10^2, maxevals=10^4)
    a = -0.3
   b = 0.6
    time = @elapsed for i in 1:maxcalls
        numerically, tol = intfn(integrandfn, a, b, maxevals=maxevals)
    end
    return time
end
function run_timing_test(name, intfn, integrandfn)
    time1 = time_integral_func(intfn, integrandfn, maxcalls=10^2, maxevals=10^4)
    time2 = time_integral_func(intfn, integrandfn, maxcalls=10^4, maxevals=10^2)
    s = ("", 13 - length(name)) # 13 is the longest name
    println(name, ": ", s, time1, " sec each heavy,\t",
                time2, " sec called often")
end
println("Run tests...")
test_integral_function(integrate_loop)
test_integral_function(quadgk)
test_integral_function(integrate_vector)
test_integral_function(integrate_mapthenreduce)
test_integral_function(integrate_mapreduce)
test_integral_function(integrate_devectorize)
println("All tests passed.")
run_timing_test("loop", integrate_loop, gaussian)
run_timing_test("quadgk", quadgk, gaussian)
run_timing_test("vector", integrate_vector, gaussian)
run_timing_test("mapthenreduce", integrate_mapthenreduce, gaussian)
run_timing_test("mapreduce", integrate_mapreduce, gaussian)
run_timing_test("devectorize", integrate_devectorize, gaussian)
# Result before optimizing anything:
#
#
   Testing...
#
  All tests passed.
#
  loop:
                  0.484841521 sec each heavy, 0.603287741 sec called often
                  0.002404426 sec each heavy, 0.300752033 sec called often
  quadgk:
                  0.335771084 sec each heavy, 0.350315113 sec called often
  vector:
```

```
mapthenreduce: 0.904367259 sec each heavy, 0.917444621 sec called often
#
                   0.703033884 sec each heavy, 0.655952055 sec called often
#
   mapreduce:
#
    devectorize:
                   0.169991816 sec each heavy, 0.249352334 sec called often
# Most functions are a little slower when called often as opposed to having
# long loops in them. However, the mapreduce version is slightly faster. The
# comparison with 'quadgk()' is unfair, as it terminates whenever the desired
# tolerance is reached (and it's a different algorithm).
# Profiling devectorize version:
println()
Profile.clear()
@profile time_integral_func(integrate_devectorize, gaussian, maxcalls=10^1, maxevals=10^5)
Profile.print()
# 170 boot.jl; include; line: 238
#
      170 profile.jl; anonymous; line: 14
#
            170 ...4/3_integration.jl; time_integral_func; line: 110
#
                  80 ...3_integration.jl; integrate_devectorize; line: 71
#
                   80 array.jl; linspace; line: 238
                  90 ...3_integration.jl; integrate_devectorize; line: 72
#
#
# Lines 71 and 72 take most of the time. 71 just allocates memory: it is the
# call to 'linspace()'. Line 72 performs the loop. I don't see much potential
# for optimizing the loop, but getting rid of the allocation via 'linspace()'
# would make sense if the function gets called often.
# Profiling vector version:
println()
Profile.clear()
@profile time_integral_func(integrate_vector, gaussian, maxcalls=10^1, maxevals=10^5)
Profile.print()
#
# 306 boot.jl; include; line: 238
      306 profile.jl; anonymous; line: 14
#
            306 ...4/3_integration.jl; time_integral_func; line: 110
#
#
                  82 ...3_integration.jl; integrate_vector; line: 34
#
                   1 array.jl; linspace; line: 237
#
                   81 array.jl; linspace; line: 238
#
                  218 ...3_integration.jl; integrate_vector; line: 35
#
                   218 ...3_integration.jl; gaussian; line: 6
#
                    9 array.jl; .*; line: 135
#
                    1 array.jl; .*; line: 136
#
                    1 array.jl; .*; line: 938
#
                    41 array.jl; ./; line: 135
```

```
array.jl; ./; line: 136
#
#
                        array.jl; ./; line: 945
#
                    45 array.jl; .^; line: 920
#
                    117 operators.jl; exp; line: 236
#
                     ...3_integration.jl; integrate_vector; line: 36
                   6 abstractarray.jl; sum; line: 1487
#
#
                    6 abstractarray.jl; sum_pairwise; line: 1481
#
                     6 abstractarray.jl; sum_pairwise; line: 1481
#
                      6 abstractarray.jl; sum_pairwise; line: 1481
#
                       6 abstractarray.jl; sum_pairwise; line: 1481
#
                        6 abstractarray.jl; sum_pairwise; line: 1481
#
                         6 abstractarray.jl; sum_pairwise; line: 1481
#
                          6 abstractarray.jl; sum_pairwise; line: 1481
#
                           1 ...ractarray.jl; sum_pairwise; line: 1480
#
                           5 ...ractarray.jl; sum_pairwise; line: 1481
#
                            5 ...ractarray.jl; sum_pairwise; line: 1481
#
                             5 ...actarray.jl; sum_pairwise; line: 1481
#
                              4 ...actarray.jl; sum_pairwise; line: 135
#
                              1 ...actarray.jl; sum_pairwise; line: 1475
#
# Some of the time is spent in 'linspace()' again, but much more time is spent
# in 'gaussian()'. This is likely due to there being some temporary arrays
# copied for intermediate results.
```

3 Triad twists

#!/usr/bin/env julia

```
# I predict that the last twisted one would be fastest as it avoids branches
# ('abs()' better be implemented branch-less). Twist1 is likely the slowest, as
# the branch could go either way, with little predicting being possible. I
# don't expect too much variability depending on the vector size, except for
# twist2, as it needs to loop through them twice, and so would be much slower
# for large sizes.
#
# For 50% positive/negative: Little prediction possible, twist3 fastest.
# For all positive: Branch prediction should work well, twist2 slowest, not
# much difference otherwise.
# For all negative: The same as all positive.
# For 90% positive: Twist2 probably wouldn't be too bad, but not the fastest
# either.
#
# I predict twist3 to be the best in almost all situations, provided that
# 'abs()' does not incur a function call, and that it is implemented without
# branches.
```

```
function triad(b::Vector, c::Vector, d::Vector)
    assert(length(b) == length(c) == length(d))
    a = similar(b)
    for i in 1:length(a)
        a[i] = b[i] + c[i] * d[i]
    end
    return a
end
function triad_twist1(b::Vector, c::Vector, d::Vector)
    assert(length(b) == length(c) == length(d))
    a = similar(b)
    for i in 1:length(a)
        if c[i]<0.
            a[i] = b[i] - c[i] * d[i]
        else
            a[i] = b[i] + c[i] * d[i]
        end
    end
    return a
end
function triad_twist2(b::Vector, c::Vector, d::Vector)
    assert(length(b)==length(c)==length(d))
    a = similar(b)
    for i in 1:length(a)
        if c[i]<0.
            a[i] = b[i] - c[i] * d[i]
        end
    end
    for i in 1:length(a)
        if c[i]>0.
            a[i] = b[i] + c[i] * d[i]
        end
    end
    return a
end
function triad_twist3(b::Vector, c::Vector, d::Vector)
    assert(length(b)==length(c)==length(d))
    a = similar(b)
    for i in 1:length(a)
        cc = abs(c[i])
        a[i] = b[i] + cc * d[i]
```

```
end
   return a
end
##### benchmark stuff
function make_data(Nobs::Int)
    srand(4242424242)
   b = randn(Nobs)
   c = randn(Nobs)
   d = randn(Nobs)
   return b, c, d
end
function time_func(name, fn, b, c, d)
   t = @elapsed fn(b, c, d)
    s = (" ", 13 - length(name)) # 13 is the longest name
    println(name, ": ", s, t, " sec")
end
data = make_data(10^7)
time_func("triad", triad, data...)
time_func("triad_twist1", triad_twist1, data...)
time_func("triad_twist2", triad_twist2, data...)
time_func("triad_twist3", triad_twist3, data...)
# triad:
                0.28133741 sec
# triad_twist1: 0.413410574 sec
# triad_twist2: 0.635552881 sec
# triad_twist3: 0.283048662 sec
# ...as predicted.
##### profile stuff
function profile_func(name, fn, b, c, d)
   Profile.clear()
   @profile fn(b, c, d)
   println()
   println(name, ":")
   Profile.print()
end
```

```
profile_func("triad", triad, data...)
profile_func("triad_twist1", triad_twist1, data...)
profile_func("triad_twist2", triad_twist2, data...)
profile_func("triad_twist3", triad_twist3, data...)
# triad:
# 263 boot.jl; include; line: 238
           263 ...comp/hw4/3_triad.jl; profile_func; line: 14
#
                ...omp/hw4/3_triad.jl; triad; line: 24
#
             228 ...omp/hw4/3_triad.jl; triad; line: 25
#
# triad_twist1:
# 380 boot.jl; include; line: 238
#
           380 ...comp/hw4/3_triad.jl; profile_func; line: 14
#
             25 ...omp/hw4/3_triad.jl; triad_twist1; line: 33
                ...omp/hw4/3_triad.jl; triad_twist1; line: 34
#
#
             111 ...omp/hw4/3_triad.jl; triad_twist1; line: 35
#
             224 ...omp/hw4/3_triad.jl; triad_twist1; line: 37
#
# triad_twist2:
# 594 boot.jl; include; line: 238
           594 ...comp/hw4/3_triad.jl; profile_func; line: 14
#
                ...omp/hw4/3_triad.jl; triad_twist2; line: 46
#
                ...omp/hw4/3_triad.jl; triad_twist2; line: 47
#
             225 ...omp/hw4/3_triad.jl; triad_twist2; line: 48
#
                ...omp/hw4/3_triad.jl; triad_twist2; line: 51
#
                ...omp/hw4/3_triad.jl; triad_twist2; line: 52
#
             136 ...omp/hw4/3_triad.jl; triad_twist2; line: 53
#
                ...omp/hw4/3_triad.jl; triad_twist2; line: 56
#
# triad_twist3:
# 263 boot.jl; include; line: 238
#
           263 ...comp/hw4/3_triad.jl; profile_func; line: 14
#
                ...omp/hw4/3_triad.jl; triad_twist3; line: 62
#
             28 ...omp/hw4/3_triad.jl; triad_twist3; line: 63
#
             212 ...omp/hw4/3_triad.jl; triad_twist3; line: 64
# The computation line generally is hit the same number of times for each,
# except for the twist1 and twist2 implementations, where the sum of the two
# computational lines is much greater. This could be explained by the fact that
# branch mis-prediction by the CPU could have the processor spend time in that
# line even when it is not being executed. That's the best explanation I have
# right now.
```