

MultiPrecisionArrays.jl: A Julia package for iterative refinement

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Summary

[MultiPrecisionArrays.jl](#) Kelley (2023b), Kelley (2023a) provides data structures and solvers for several variations of iterative refinement (IR). IR can speed up an LU matrix factorization by factoring a low precision copy and using the low precision factorization in a residual correction loop. The additional storage cost is the low precision copy, so IR is at time vs storage tradeoff. IR is an old algorithm and a good account of the classical theory is in Higham (1996).

Statement of need

The package [IterativeRefinement.jl](#) is an implementation of the IR method from J.Dongarra et al. (1983). It has not been updated in four years.

The unregistered package [ltref.jl](#) implements IR and the GMRES-IR method from Amestoy et al. (2023) and was used to obtain the numerical results in that paper. It does not provide the data structures for preallocation that we do and does not seem to have been updated lately.

Algorithm

This package will make solving dense systems of linear equations faster by using the LU factorization and IR. It is limited to LU for now. A very generic description of this for solving a linear system $Ax = b$ is

IR(A, b)

- $x = 0$
- $r = b$
- Factor $A = LU$ in a lower precision
- While $\|r\|$ is too large
 - $d = (LU)^{-1}r$
 - $x = x + d$
 - $r = b - Ax$
- end
- end

In Julia, a code to do this would solve the linear system $Ax = b$ in double precision by using a factorization in a lower precision, say single, within a residual correction iteration. This means that one would need to allocate storage for a copy of A in the lower precision and factor that

34 copy. Then one has to determine what the line $d = (LU)^{-1}r$ means. Do you cast r into the
 35 lower precision before the solve or not? **MultiPrecisionArrays.jl** provides data structures and
 36 solvers to manage this. The **MPLArray** structure lets you preallocate A , the low precision copy,
 37 and the residual r . The factorizations factor the low-precision copy and the solvers use that
 38 factorization and the original high-precision matrix to run the while loop in the algorithm.

39 IR is a perfect example of a storage/time tradeoff. To solve a linear system $Ax = b$ in R^N
 40 with IR, one incurs the storage penalty of making a low precision copy of A and reaps the
 41 benefit of only having to factor the low precision copy.

42 Installation

43 The standard way to install a package is to type `import Pkg; Pkg.add("MultiPrecisionArrays")`
 44 at the Julia prompt. One can run the unit tests with `Pkg.test("MultiPrecisionArrays")`.
 45 After installation, type using `MultiPrecisionArrays` when you want to use the functions in
 46 the package.

47 Example

48 Here is a simple example to show how iterative refinement works. We will follow that with
 49 some benchmarking on the cost of factorizations. The functions we use are **MPLArray** to create
 50 the structure and **mplu!** to factor the low precision copy. In this example high precision is
 51 `Float64` and low precision is `Float32`. The matrix is the sum of the identity and a constant
 52 multiple of the trapezoid rule discretization of the Greens operator for $-d^2/dx^2$ on $[0, 1]$

$$Gu(x) = \int_0^1 g(x, y)u(y) dy$$

53 where

$$g(x, y) = \begin{cases} y(1-x); & x > y \\ x(1-y); & x \leq y \end{cases}$$

54 The code for this is in the `/src/Examples` directory. The file is **Gmat.jl**. You need to do

```
55 using MultiPrecisionArrays
56 using MultiPrecisionArrays.Examples
57 to get to it.
```

58 The example below compares the cost of a double precision factorization to a **MPLArray**
 59 factorization. The **MPLArray** structure has a high precision (TH) and a low precision (TL) matrix.
 60 The structure we will start with is

```
61 struct MPLArray{TH<:AbstractFloat, TL<:AbstractFloat}
62     AH::Array{TH, 2}
63     AL::Array{TL, 2}
64     residual::Vector{TH}
65     onthefly::Bool
66 end
```

67 The structure also stores the residual. The `onthefly` Boolean tells the solver how to do the
 68 interprecision transfers. The easy way to get started is to use the `mplu` command directly on
 69 the matrix. That will build the **MPLArray**, follow that with the factorization of `AL`, and put in
 70 all in a structure that you can use with `\`.

71 Now we will see how the results look. In this example we compare the result with iterative
72 refinement with $A \backslash b$, which is LAPACK's LU. As you can see the results are equally good.
73 Note that the factorization object MPF is the output of `mplu`. This is analogous to $AF = \text{lu}(A)$ in
74 LAPACK.

```
75 julia> using MultiPrecisionArrays
76
77 julia> using MultiPrecisionArrays.Examples
78
79 julia> using BenchmarkTools
80
81 julia> N=4096;
82
83 julia> G=Gmat(N);
84
85 julia> A = I - G;
86
87 julia> MPF=mplu(A); AF=lu(A);
88
89 julia> z=MPF\b; w=AF\b;
90
91 julia> ze=norm(z-x,Inf); zr=norm(b-A*z,Inf)/norm(b,Inf);
92
93 julia> we=norm(w-x,Inf); wr=norm(b-A*w,Inf)/norm(b,Inf);
94
95 julia> println("Errors: $ze, $we. Residuals: $zr, $wr")
96 Errors: 8.88178e-16, 7.41629e-14. Residuals: 1.33243e-15, 7.40609e-14
```

97 So the results are equally good.
98 The compute time for `mplu` should be roughly half that of `lu`.

```
99 julia> @belapsed mplu($A)
100 8.55328e-02
101
102 julia> @belapsed lu($A)
103 1.49645e-01
```

104 It is no surprise that the factorization in single precision took roughly half as long as the one
105 in double. In the double-single precision case, iterative refinement is a great example of a
106 time/storage tradeoff. You have to store a low precision copy of A , so the storage burden
107 increases by 50% and the factorization time is cut in half.

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