

An introduction to multi-precision linear algebra library MPLAPACK

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- What is and why MPLAPACK
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- Examples
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What is and why MPLAPACK



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- MPLAPACK is a rewrite of the widely used linear algebra packages **BLAS** and **LAPACK**, with **multiple precision** support, by Japanese scientist 中田真秀 (NAKATA Maho). It keeps the same interfaces of BLAS and LAPACK routines.
- LAPACK is for solving system of linear equations, the least squares problems, finding eigenvalues/eigenvectors and computing singular value decomposition (SVD), etc using the building blocks provided in BLAS for dense matrices.
- LAPACK and BLAS were written in Fortran (FORTRAN 77), containing subroutines separately supported for IEEE **single** (~7 digit accuracy) and **double precision** (~15 digit accuracy) real and complex matrices. Same routine, same interface e.g. `xGEMM`, `xGETRI`, with different letter `x` for different types.
- There are needs from physics, chemistry, engineering and mathematics where higher precision, e.g. accuracy of hundreds of digits, is wanted.
- LAPACK does not support arbitrary precision. If one wants to solve ill-conditioned problem $A^*x = b$ or compute the eigenvalues, or solving positive semi-definite programming where higher precision is desired, there is a gap between the needs for LAPACK and the needs for higher precision.
- MPLAPACK bridges the gap.

Example 1: Solving an ill-conditioned problem $A^*x = b$, where

$$A = \begin{bmatrix} 2 & 3.9998 \\ 1 & 2 \end{bmatrix}, \text{ let } x = \begin{bmatrix} 1 \\ 1 \end{bmatrix}, \quad b = Ax = \begin{bmatrix} 5.9998 \\ 3 \end{bmatrix}.$$

We construct the right hand side (RHS) b by multiplying matrix A by a unit vector.

We now make changes first in b and then in A and examine the solutions in MATLAB

- 1) Change b by less than 0.001, keep A the same

$$b' = \begin{bmatrix} 5.9998 \\ 3.0003 \end{bmatrix}, \quad \Rightarrow x' = \begin{bmatrix} -5.7589 \\ 4.3796 \end{bmatrix}.$$

- 2) Change the elements in A by less than 0.001, keep b the same

$$A' = \begin{bmatrix} 2.0000 & 3.9999 \\ 1.0005 & 2.0002 \end{bmatrix}, \quad \Rightarrow x' = \begin{bmatrix} -0.8036 \\ 1.9018 \end{bmatrix}.$$

Small changes in the input lead to large changes in the output.

Example 1: Solving an ill-conditioned problem $A^*x = b$ (cont'd)

$$A = \begin{bmatrix} 2 & 3.9998 \\ 1 & 2 \end{bmatrix}, \quad x = \begin{bmatrix} 1 \\ 1 \end{bmatrix}, \quad b = Ax = \begin{bmatrix} 5.9998 \\ 3 \end{bmatrix}.$$

Small changes in b and A lead to large changes in the solution.

- 1) Change b by less than 0.001, keep A the same

$$b' = \begin{bmatrix} 5.9998 \\ 3.0003 \end{bmatrix}, \quad \Rightarrow x' = \begin{bmatrix} -5.7589 \\ 4.3796 \end{bmatrix}.$$

- 2) Change the elements in A by less than 0.001, keep b the same

$$A' = \begin{bmatrix} 2.0000 & 3.9999 \\ 1.0005 & 2.0002 \end{bmatrix}, \quad \Rightarrow x' = \begin{bmatrix} -0.8036 \\ 1.9018 \end{bmatrix}.$$

Things like this may confuse us in real life. This matrix is ill-conditioned.



Example 1: Solving an ill-conditioned problem $A^*x = b$ (cont'd)

$$A = \begin{bmatrix} 2 & 3.9998 \\ 1 & 2 \end{bmatrix}, \quad b = \begin{bmatrix} 5.9998 \\ 3 \end{bmatrix}, \quad x = \begin{bmatrix} 1 \\ 1 \end{bmatrix}.$$

This raises the questions, especially when we face real problems:

- How do we know the solution we have obtained is correct?
- How reliable is the computed solution, if we have a small change in the input and we see a large change in the output?
- etc.

Example 2: Solving system of linear equations $H^*x = b$ with Hilbert matrix, with

$$h_{ij} = 1/(i + j - 1).$$

For example, for n=6*

$$H_6 = \begin{bmatrix} 1/1 & 1/2 & 1/3 & 1/4 & 1/5 & 1/6 \\ 1/2 & 1/3 & 1/4 & 1/5 & 1/6 & 1/7 \\ 1/3 & 1/4 & 1/5 & 1/6 & 1/7 & 1/8 \\ 1/4 & 1/5 & 1/6 & 1/7 & 1/8 & 1/9 \\ 1/5 & 1/6 & 1/7 & 1/8 & 1/9 & 1/10 \\ 1/6 & 1/7 & 1/8 & 1/9 & 1/10 & 1/11 \end{bmatrix}.$$

Example 2: Solving system of linear equations $H^*x = b$ with Hilbert matrix (cont'd). Hilbert matrix is ill-conditioned, solving the linear system with an ill-conditioned matrix becomes problematic when n is large.

$$H_6 = \begin{bmatrix} 1/1 & 1/2 & 1/3 & 1/4 & 1/5 & 1/6 \\ 1/2 & 1/3 & 1/4 & 1/5 & 1/6 & 1/7 \\ 1/3 & 1/4 & 1/5 & 1/6 & 1/7 & 1/8 \\ 1/4 & 1/5 & 1/6 & 1/7 & 1/8 & 1/9 \\ 1/5 & 1/6 & 1/7 & 1/8 & 1/9 & 1/10 \\ 1/6 & 1/7 & 1/8 & 1/9 & 1/10 & 1/11 \end{bmatrix}.$$

We construct the RHS b by multiplying H by a unit vector x , i.e. $b = H^*x$.

We then solve $H^*x = b$ and see if we get the solution the same, or "close" to the the unit vector.

Note: One may see in the literature the advice that Hilbert matrix is not a good test matrix because it is ill-conditioned. We use it intentionally for the purpose of testing the features of MPLAPACK.

What is and why MPLAPACK

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Example 2: Solving system of linear equations $H^*x = b$ with Hilbert matrix (cont'd). Hilbert matrix is ill-conditioned, solving the linear system via LU decomposition may become problematic when n becomes large.

Solving $H^*x = b$ of size 10	Solving $H^*x = b$ of size 11	Solving $H^*x = b$ of size 13	Solving $H^*x = b$ of size 50
Solution:	Solution:	Solution:	Solution:
[1]	[1]	[1]	[1]
1	1	0.9999999	1.000001
0.9999999	0.9999982	1.000013	0.999892
1.000003	1.000047	0.9994794	1.004564
0.9999758	0.9994746	1.0088	0.9173155
1.000116	1.003154	0.9199938	1.790654
0.999682	0.9888376	1.438536	-3.380056
1.000522	1.02444	-0.5437174	15.46936
0.9994947	0.9665197	4.60748	-26.46469
1.000266	1.027928	-4.656326	25.66145
0.9999415	0.9870307	6.882018	1.871251
	1.00257	-2.890713	-15.6143
		2.481753	15.50733
		0.7526833	-30.80703
			22.92981
			52.54445
			-54.27587
			...

When $n \geq 12$, one will get warning: the matrix is singular to machine precision.

Availability

From <https://github.com/nakatamaho/mplapack>

- Linux
- Mac OS
- Windows (mingw-64)

Two methods of having it on your computer:

- Using Docker, just follow the instruction; OR
- Building from tar-ball. Straightforward, requires **GMP** and **MPFR**, no hiccup, but may take more than an hour, keeping the fan humming.

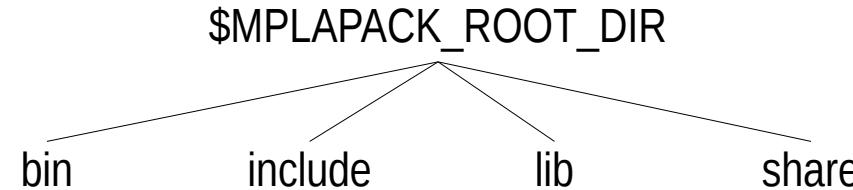


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The built package has a well organized structure



- It comes with a user manual and examples in share.
- The examples are easy to follow if one has used LAPACK or BLAS before.

Supported floating-point formats

- IEEE double precision (64-bit) ~15 digits.
- Extended (80-bit) double precision ~19 digits.
- Binary128 (128-bit) quadruple precision ~33 digits.
- Double-double ~32 digits and quad-double ~64 digits.
- GMP, but to be dropped in the future.
- MPFR for arbitrary precision.
- The precision can be set by environmental variables `MPLAPACK_GMP_PRECISION`, `MPLAPACK_MPFR_PRECISION` in binary digits (bits).



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Other related tools and packages

Related tools and packages

- Maple
- Mathematica
- GMP – GNU multiple precision arithmetic library.
- MPFR – GNU multiple precision floating-point reliable library
- MPFUN2020 <https://www.davidhbailey.com/dhbsoftware/>
- Arb – A C library for rigorous real and complex arithmetic with arbitrary precision
- Julia – with support through intrinsic data types by design.
- mpmath – A Python library for arbitrary-precision floating-point arithmetic, <http://mpmath.org/>.

Examples

Example 3: Solving system of linear equations $H^*x = b$ with Hilbert matrix, *revisited*. The RHS b is built by multiplying H by a unit vector x . The solution, however, deviates wildly when n gets bigger.

Solving $H * x = b$ of size 10

Solution:

```
[ 1 ]  
    1  
0.9999999  
1.000003  
0.9999758  
1.000116  
0.999682  
1.000522  
0.9994947  
1.000266  
0.9999415
```

$$\|dx\|_2/\|x\|_2 = 0.00026777989621392767$$

Solving $H * x = b$ of size 12

Solution:

```
[ 1 ]  
    1  
0.9999867  
1.000421  
0.9942459  
1.042286  
0.8138542  
1.519414  
0.05870534  
2.104559  
0.1904543  
1.336781  
0.9392925
```

$$\|dx\|_2/\|x\|_2 = 0.51517152763376584$$



Example 3: Solving system of linear equations $H^*x = b$ with Hilbert matrix, *revisited*. The RHS b is built by multiplying H by a unit vector x . The solution, however, deviates wildly when n gets bigger.

Solving $H * x = b$ of size 13

Solution:

```
[ 1 ]  
0.9999999  
1.000013  
0.9994794  
1.0088  
0.9199938  
1.438536  
-0.5437174  
4.60748  
-4.656326  
6.882018  
-2.890713  
2.481753  
0.7526833
```

$$\|dx\|_2 / \|x\|_2 = 2.7677028033584072$$

Solving $H * x = b$ of size 50

Solution:

```
[ 1 ]  
1.000001  
0.999892  
1.004564  
0.9173155  
1.790654  
-3.380056  
15.46936  
-26.46469  
25.66145  
1.871251  
-15.6143  
15.50733  
-30.80703  
22.92981  
52.54445  
-54.27587
```

$$\|dx\|_2 / \|x\|_2 = 27.025124940220273$$

Example 3: Solving system of linear equations $H^*x = b$ with Hilbert matrix (cont'd)

- The inverse of Hilbert matrix has a closed form containing integer values. But when $n \geq 12$, some of its values fall beyond what can be represented exactly in IEEE double precision.
- This means one can't get accurate answer by the multiplication of the inverse matrix and the right RHS.

The following is output of the 9th to 13th columns of the inverse of an Hilbert matrix of order 13 in julia*

1891439550//1	-1849407560//1	1160082924//1	-421848336//1	67603900//1
-285985659960//1	282454972800//1	-178652770296//1	654189419522//1	-10546208400//1
10724462248500//1	-10680328659000//1	6802547792040//1	-2505779115840//1	406029023400//1
-174769014420000//1	175266931840000//1	-112296027043200//1	41577371996160//1	-6767150390000//1
1542672646515000//1	-1556276461740000//1	1002242041360560//1	-372734643481200//1	60904353510000//1
-8251094840788800//1	8366542258314240//1	-5412107023347024//1	2020660279013376//1	-331319683094400//1
28450997395460640//1	-28977867717598800//1	18818568211899456//1	-7050482856248832//1	1159618890830400//1
-65321167489578000//1	66791723910912000//1	-43525940081944320//1	16357726068203520//1	-2697888848054400//1
100863567447142500//1	-103492384705710000//1	67651337791837800//1	-25495049614114080//1	4215451325085000//1
-103492384705710000//1	106518477825760000//1	-69822862214785680//1	26379357625420800//1	-4371579151940000//1
67651337791837800//1	-69822862214785680//1	45883595169716304//1	-17374404181470336//1	2885242240280400//1
-25495049614114080//1	26379357625420800//1	-17374404181470336//1	6592659294154752//1	-1096868950850400//1
4215451325085000//1	-4371579151940000//1	2885242240280400//1	-1096868950850400//1	182811491808400//1

* This matrix is generated in julia with SpecialMatrices package. The values are represented in Rational format.

Example 3: Solving $A^*x = b$ by calling LAPACK

```
Int main()
{
    ...
    // Create Hilbert matrix of order n
    for (int i = 0; i < n; i++)
    {
        for (int j = 0; j < n; j++)
            H[i + j * n] = one / ((i + 1) + (j + 1) - 1);
        x[i] = one; // True solution vector
    }

    // Create the RHS b = H*x
    dgemm("N", "N", n, ione, n, one, H, n, x, n, zero, b, n);

    // Solve linear equation A * x = b, store solution in b
    dgesv(n, ione, H, n, ipiv, b, n, info);

    // Display the solution stored in b
    cout << "\nSolution:\n";
    printmat_pretty(b.data(), n, 1, n, 1, 1);
}
```

Example 3: Solving $A^*x = b$ by calling MPLAPACK

```
#include <mpblas_mpfr.h>
#include <mplapack_mpfr.h>

int main()
{
    ...
    // Create Hilbert matrix of order n
    for (int i = 0; i < n; i++) {
        for (int j = 0; j < n; j++) {
            mtmp = (i + 1) + (j + 1) - 1;
            H[i + j * n] = one / mtmp;
        }
        x[i] = one;
    }

    // Create the RHS b = H*x
    Rgemm("N", "N", n, 1, n, one, H, n, x, n, zero, b, n);

    // Solve linear equation A * x = b, store solution in b
    Rgesv(n, (mplapackint)1, H, n, ipiv, b, n, info);

    // Display the solution stored in b
    printf("\nsol:\n"); printvec1(b.data(), n);
}
```

Example 3: Solving system of linear equations $H^*x = b$ with Hilbert matrix, revisited (cont'd).

The solution using MPLAPACK for size n=13 with precision MPLAPACK_MPFR_PRECISION=52 (~15 digits)

```
Solving H * x = b with size 13
```

```
sol:
```

```
+1.000000229180270316931000706972554326057434082031250000000000000e+00
+9.999652939882666125015475699910894036293029785156250000000000000e-01
+1.001303165205678524785071203950792551040649414062500000000000000e+00
+9.78740937561858093118871693150140345096588134765625000000000000000e-01
+1.187810149111231616814166045514866709709167480468750000000000000e+00
-5.194997667174320807315091030886833323165774345397949218750000000e-03
+4.467949041701382739688597212079912424087524414062500000000000000e+00
-6.965640166625449936077529855538159608840942382812500000000000000e+00
+1.33043039538297023227642057463526725769042968750000000000000000e+01
-1.162870483617172467916134337428957223892211914062500000000000000e+01
+9.25718176430800454568270652089267969131469726562500000000000000e+00
-2.112419206070680566256214660825207829475402832031250000000000000e+00
+1.514704831488293379493370593991130590438842773437500000000000000e+00
||dx||_2 / ||x||_2 = 5.9841935630246308
```

Example 3: Solving system of linear equations $H^*x = b$ with Hilbert matrix, revisited (cont'd).

The solution using MPLAPACK for size n=13 with precision set to 64 (~19 digits)

```
Solving H * x = b with size 13
```

```
sol:
```

```
+1.0000000000202592024102846157695978490664856508374214172363281250e+00
+9.999999713765458229279906032793689973914297297596931457519531250e-01
+1.000001011472766303082859717932251442107371985912322998046875000e+00
+9.999843658530147822273773106083183392911450937390327453613281250e-01
+1.0000131584531067633573458985907222995592746883630752563476562500e+00
+9.9993259840260664269651416646844666047400096431374549865722656250e-01
+1.0002234095538461025945431437378374539548531174659729003906250000e+00
+9.9950535075265004806427393546641724242363125085830688476562500000e-01
+1.0007386362758243786167181021973249244183534756302833557128906250e+00
+9.9926527577780360181609886627285277427290566265583038330078125000e-01
+1.0004666128738118594963465990410611539118690416216850280761718750e+00
+9.9982882411360100036266246315008743295038584619760513305664062500e-01
+1.0000275989100892654395935643663051450857892632484436035156250000e+00
||dx||_2 / ||x||_2 = 3.543685575947061e-4
```



Example 3: Solving system of linear equations $H^*x = b$ with Hilbert matrix, revisited (cont'd)

The solution using MPLAPACK for size n=13 with precision set to 100 (~30 digits)

Solving $H * x = b$ with size 13

sol:



Example 3: Solving system of linear equations $H^*x = b$ with Hilbert matrix, revisited (cont'd).

The solution using MPLAPACK for size n=50 with precision set to 64 (~19 digits)

```
Solving H * x = b with size 50
```

```
sol:
```

```
+9.999998323920491987261763289218663430801825597882270812988281250e-01
+1.0000040958644698844233131285186289005650905892252922058105468750e+00
+9.9975335853895302951586943063233547945856116712093353271484375000e-01
+1.0064168906572163326431854080134087325859582051634788513183593750e+00
+9.1017709361797309981995454664094324925827095285058021545410156250e-01
+1.7546457263592632714547386862946609653590712696313858032226562500e+00
-3.0254855778269326730886557008659565326524898409843444824218750000e+00
+1.4927053772695171149700876789268022548640146851539611816406250000e+01
-2.9906069249450969947320966291215427190763875842094421386718750000e+01
+4.2767396841196820132802169922570101334713399410247802734375000000e+01
-3.0697584028950302695998297863866355328354984521865844726562500000e+01
+2.211551147035071555706220447490295553261563181877136230468750000e+01
-4.1393720509945765184678379000615677796304225921630859375000000000e+01

' ' '
-6.2631474974109959945323122809668348054401576519012451171875000000e+01
+5.7912921674890292290244220296813182358164340257644653320312500000e+01
-1.3780150677677150288825969060013676426024176180362701416015625000e+01
||dx||_2 / ||x||_2 = 38.78403993623718
```

Example 3: Solving system of linear equations $H^*x = b$ with Hilbert matrix, revisited (cont'd).

The solution using MPLAPACK for size n=50 with precision set to 128 (~38 digits)

```
Solving H * x = b with size 50
```

```
sol:
```

```
+1.00000000000000038079415848112910489096635685977313350832873688e+00
+9.999999999999966750701030718839213402964044475822265094207334280e-01
+1.000000000007255172703651693486401563253154143934208671130269567e+00
+9.999999993038837272551621268809970416286823965442388084600946450e-01
+1.000000037115146126182340265678742155067632732281248938656716583e+00
+9.99998753888444599075067331553907452807942009956243269657735589e-01
+1.000028459629431368989744684343643758006993935196796059628634314e+00
+9.995345392071606549291704153566672717231530818676983009051771465e-01
+1.0005648957967554605483907518990365568580425813599501768911185623e+00
+9.9478467201851628393711200786220385336335657458783827004246119611e-01
+1.0372503209164496010899653761339063290379157475001644945029049109e+00
+7.9209278251439380606377111389207802732849034139528473607318024525e-01
+1.9097837552464119470205639731269624580606586446574511656723177134e+00

' ' '
+2.8721507790161390178564992472450103080885080664547374818894447845e+01
-6.4398157655023067179298982089858407629351504680616088425163990860e+00
+1.8268230699763150792418463987246657311249682987681419189201216862e+00
||dx||_2 / ||x||_2 = 27.676389927831689
```



Example 3: Solving system of linear equations $H^*x = b$ with Hilbert matrix, revisited (cont'd)

The solution using MPLAPACK for size n=50 with precision set to 256 (~77 digits)

Solving $H * x = b$ with size 50

sol:



Example 4: Consider a different problem, computing the eigenvalues of a Wilkinson matrix with

$$\text{diag } d_i = \begin{cases} \frac{n-1}{2} - i + 1 & \text{if } n \text{ is odd} \\ \frac{n}{2} - i + \frac{1}{2} & \text{if } n \text{ is even} \end{cases}, \quad 1 \text{ on upper and lower diagonals.}$$

For example, for $n=7$

$$W_7 = \begin{bmatrix} 3 & 1 & 0 & 0 & 0 & 0 & 0 \\ 1 & 2 & 1 & 0 & 0 & 0 & 0 \\ 0 & 1 & 1 & 1 & 0 & 0 & 0 \\ 0 & 0 & 1 & 0 & 1 & 0 & 0 \\ 0 & 0 & 0 & 1 & 1 & 1 & 0 \\ 0 & 0 & 0 & 0 & 1 & 2 & 1 \\ 0 & 0 & 0 & 0 & 0 & 1 & 3 \end{bmatrix}.$$



Example 4: Compute the eigenvalues of Wilkinson matrices using LAPACK routine xSTEV

```

-0.964132172690471
0.212662842513101
1.078164406734622
1.654730092254141
2.400771751580549
2.613723558709661
3.486274955522134
3.517639311060091
4.498968620932756
4.501192469653734
5.499953844648218
5.500050238625574
6.499998623277986
6.500001455954134
7.499999970478817
7.500000030730065
8.49999999521540
8.500000000493207
9.49999999993923
9.500000000006223
10.49999999999936
10.500000000000064
11.49999999999996
11.49999999999998
12.49999999999991
12.50000000000004
13.49999999999998
13.50000000000009
14.50000000000005
14.50000000000007
15.50000000000622
15.50000000000627
16.50000000054484
16.50000000054484
17.500000003808122
17.500000003808132
18.50000205070435
18.50000205070439
19.500008158672937
19.500008158672941
20.500225680185167
20.500225680185174
21.503952002665361
21.503952002665361
22.53894111930640
22.538941119306443
23.710678647330340
23.710678647330347
25.246194182903356
25.246194182903363

```

Using DSTEV (15 digits)

```

-9.6413217269047884692040697285411043762973215542899652064079418778e-01
+2.1266284251311072957020059527671627182428271130021357748773880303e-01
+1.07816440673461058675157671503333815109417574933494740212336183e+00
+1.65473009225413448557971351045780748191787144969566725194454193e+00
+2.4007717515805496534716909537495469146098359125574006611714139581e+00
+2.6137235587096797025166319830774682916874063209888454931352139e+00
+3.4862749555221273773161299389519470838933862097519522649236023426e+00
+3.5176393110600852069133567795676780908670222558498608123045414686e+00
+4.498968620932769352208490799602784307800740215554924361405874491e+00
+4.50119246965374796705999695642297689845978680978078045882284641266e+00
+5.4999538446482194288382926543188233478587356728439772268757224083e+00
+5.500050238625574936119251695257540242319736599712632596493e+00
+6.499998623277988915591971050985216826298867982964111433830112219e+00
+6.5000014559541367998971643177688920726270405126001605822239071131e+00
+7.499999970478817137165746958290769470187333153674477420281618834e+00
+7.500000030730059526457076085147310153672384203105139022273766994e+00
+8.499999995215409279787901564927029077201048323786380933597683907e+00
+8.5000000049320627776036358483980953301695241588459247141145169735e+00
+9.49999999999392072050662623541995438491980507933476474136114120e+00
+9.50000000062265744774814275714572980867619436225155368447304e+00
+1.04999999999993785719572442581814511172219330342159082647413015e+01
+1.050000000006353023028659717974101980852985604997229529544711e+01

```

$$\begin{aligned}
 & +2.1503952002665361328366097505690370667030819618048553820699453354e+01 \\
 & +2.1503952002665361328366150445249574060802011388204846298322081566e+01
 \end{aligned}$$

21.503952002665361
21.503952002665361



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Example 5: Compute SVD of matrices. Take again the Hilbert matrix for example

$$h_{ij} = 1/(i + j - 1).$$

For example, for n=6*

$$H_6 = \begin{bmatrix} 1/1 & 1/2 & 1/3 & 1/4 & 1/5 & 1/6 \\ 1/2 & 1/3 & 1/4 & 1/5 & 1/6 & 1/7 \\ 1/3 & 1/4 & 1/5 & 1/6 & 1/7 & 1/8 \\ 1/4 & 1/5 & 1/6 & 1/7 & 1/8 & 1/9 \\ 1/5 & 1/6 & 1/7 & 1/8 & 1/9 & 1/10 \\ 1/6 & 1/7 & 1/8 & 1/9 & 1/10 & 1/11 \end{bmatrix}.$$



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Example 5: Compute SVD of matrices (cont'd). For an m-by-n real matrix A, the SVD decomposition of A is

$$A = USV^T$$

where U is an m-by-m matrix, V an n-by-n matrix and S is an m-by-n matrix with only nonzero “singular values” on the diagonal

$$S = \begin{bmatrix} \sigma_1 & & \\ & \ddots & \\ & & \sigma_n \\ & & \vdots \\ 0 & \cdots & 0 \end{bmatrix} \quad \text{for } m \geq n, \quad S = \begin{bmatrix} \sigma_1 & & & 0 \\ & \ddots & & \\ & & \sigma_m & \cdots & 0 \end{bmatrix} \quad \text{for } m \leq n.$$

Among many things, one of the properties of singular values is it can tell us the rank and the condition number of a matrix, e.g. is it close to “singular”? that we've been looking at so far. We can compute the condition number in 2-norm by $\kappa_2 = \sigma_{\max}/\sigma_{\min}$ (see MATLAB).

Example 5: Compute SVD of matrices (cont'd) using LAPACK routine xGESVD, x being S for single and D for double. In MPLAPACK, the uniform interface is Rgesvd. The following is the code snippet

```
// Create/input the m x n matrix A
```

```
... ...
```

```
// Compute SVD of the matrix. Check the man page of DGESVD for details
```

```
Rgesvd("A", "A", m, n, A, m, S, U, m, VT, n, work, lwork, info);
```



except for the leading letter R, the interface is the same for both LAPACK and MPLAPACK

where

- First two letters indicating the options for computing U and VT respectively.
- A – m -by- n array, S – m -by- n matrix, U – m -by- m matrix, VT – n -by- n matrix.
- $work$ – work space of length $lwork$.



Example 5: Compute SVD of matrices (cont'd). Results of SVD of Hilbert matrix of size 13, with 15 and 154 digits (roughly)

S:

```
1.81383
0.3968331
0.04902942
0.004348755
0.0002951777
1.56237e-05
6.466419e-07
2.076321e-08
5.076552e-10
9.141281e-12
1.143512e-13
8.904175e-16
3.165503e-19
Cond(H(13)) : 5.7299894005078e+18
```

S:

```
+1.8138301187969768852554779111174893763742105990572791248194516291e+00
+3.9683307601762220579755614279608197831854306117823218168622528123e-01
+4.9029419419807657678273619984872715156853072246391090593441213011e-02
+4.3487550746417711561775810952824274325591344028204942187518701742e-03
+2.9517771353297399519509834580235921630962242284989409930465519070e-04
+1.5623703604059947550297296991582012752163573065275231760789850850e-05
+6.4664185629354604881522843894924507240718222321175303734376474269e-07
+2.0763214206157548764894005872912976923546949538644353117529335119e-08
+5.076551871014000227315722964240005183275518195695656294939343124e-10
+9.1412825065284215546059809754066319421083627349189596801271593715e-12
+1.1434960998078891102081478316358599670275540526922499695796038305e-13
+8.8782106985855237443916799922053800506780193406461192827889493846e-16
+3.2229010148608565932592373560345004561891011748649805132776692489e-18
Cond(H(13)) : +5.6279423737600765028088989122951008279945077348277108547214852203e+17
```

We see the condition number is huge.

Parallel processing

- Currently a few key routines in MPBLAS are optimized for acceleration via OpenMP
 - Raxpy – compute $y = a^*x + y$.
 - Rcopy – copy A to B.
 - Rdot – dot-product of two vectors.
 - Rgemm – matrix-matrix multiplication $C = \alpha A^*B + \beta C$
- Need to compile and link against the optimized MPBLAS library, e.g.
`-fopenmp -lmpblas_mpfr_opt -lmpfr -lmpc -lgmp`

Limitations

- Currently it only has the C/C++ (more C-style) interface. There is no Fortran interface (sorry!).
- Setting the precision (number of digits) is restricted by the underlying GMP/MPFR implementations.
- The data objects are opaque, handler functions for output are limited.
- Inter-language operations requires some coding work.



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- MPLAPACK is a multiple precision version of LAPACK for linear algebra operations.
- It is callable from C/C++, not readily from Fortran, but it has the potential being ported to other languages, thanks to its design philosophy – keep the interfaces identical to original BLAS and LAPACK.
- It is easy to install.
- It is quite straightforward to use, if one is familiar with LAPACK.
- The key building blocks in MPBLAS are optimized, capable of performing parallel processing with the support of OpenMP (on single computer).
- One needs some knowledge in numerical analysis as a guide line and needs to experiment with the accuracy by setting the number of digits.

- [1] **MPLAPACK**, <https://github.com/nakatamaho/mplapack>.
- [2] **LAPACK**, <https://netlib.org/lapack/>.
- [3] Fredrik Johansson et al, **mpmath**: A Python library for arbitrary-precision floating-point arithmetic, <http://mpmath.org/>.
- [4] Fredrik Johansson et al, **Arb**: A C library for arbitrary-precision interval arithmetic, <https://github.com/fredrik-johansson/arb/>.
- [5] David Bailey, **MPFUN2020**, David Bailey's website: High-Precision Software Directory, <https://www.davidhbailey.com/dhbsoftware/>.
- [6] **Eigen**: A C++ template library for linear algebra, <https://eigen.tuxfamily.org/>.
- [7] **Elemental**: A new framework for distributed memory dense linear algebra, <https://github.com/elemental/Elemental> (halted).
- [8] **Julia**, a language for scientific and high performance computing, <https://docs.julialang.org/en/v1/>.