Square root of a symmetric matrix

Tags: RKFIT, square root
Name: Mario Berljafa and Stefan Güttel
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1 Introduction

This is an example of RKFIT [1, 2, 3] being used for approximating $\sqrt{A}\boldsymbol{b}$, the action of the matrix square root onto a vector $\boldsymbol{b}$. RKFIT is described in [2, 3] and implemented in the RK Toolobox [1]. This code reproduces the example from [2, Section 5.2]. We also compare RKFIT to the vector fitting code VFIT3. Vector fitting is described in [4, 5].

We first define the matrix A, the vector $\boldsymbol{b}$, and the matrix F corresponding to $\sqrt{A}$. Our aim is then to find a rational function $r(z)$ of type (m, m) such that $\|F\boldsymbol{b} - r(A)\boldsymbol{b}\|_2/\|F\boldsymbol{b}\|_2 \to \min$.

```
N = 500;
A = gallery('tridiag',N);
b = eye(N,1);
f = @(x) sqrt(x); fm = @(X) sqrtm(full(A));
F = fm(A);
exact = F*b;
[U,D] = eig(full(A)); ee = diag(D).';
```

2 Running `rkfit`

In order to run RKFIT we only need to specify the initial poles of r. In this example we pretend to not know that a reasonable choice of initial poles is on

the negative real axis (the branch cut of $\sqrt{z}$) and choose 16 infinite poles ξ. As all quantities F, A, and $\boldsymbol{b}$, as well as the initial poles are real, it is advised running RKFIT with the `'real'` option. This will attempt to produce a rational approximant with perfectly complex conjugate (or even real) poles. We perform 10 RKFIT iterations.

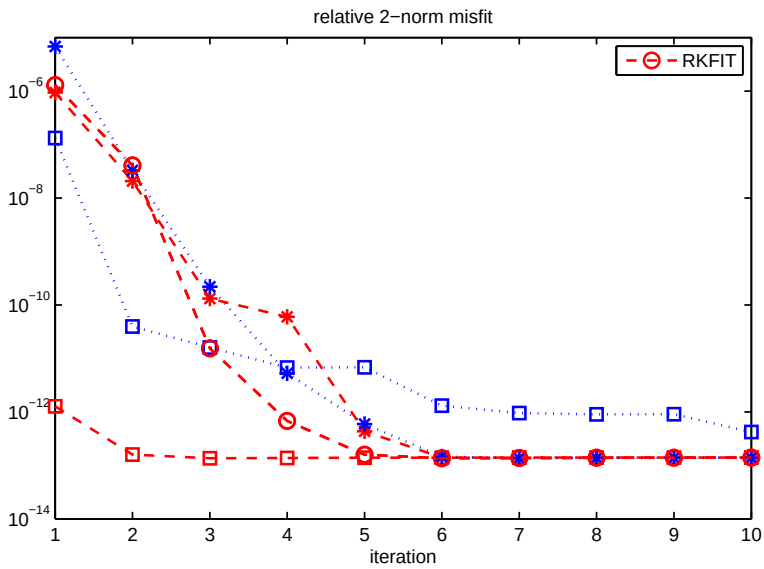
```
m = 16;
[xi, ratfun, misfit] = rkfit(F, A, b, repmat(inf, 1, m), ...
                                10, 'real');
```

It turns out that all computed poles are real negative:

```
xi

xi =
  Columns 1 through 6
  -1.5331e+02  -1.5895e+01  -4.9902e+00  -2.0811e+00
-9.6793e-01  -4.7037e-01
  Columns 7 through 12
  -2.3165e-01  -1.1374e-01  -5.5114e-02  -2.6156e-02
-1.2058e-02  -5.3360e-03
  Columns 13 through 16
  -2.2169e-03  -8.2578e-04  -2.4778e-04  -4.3227e-05
```

Here is a convergence plot of RKFIT, showing the relative misfit $\|F\boldsymbol{b} - r(A)\boldsymbol{b}\|_2/\|F\boldsymbol{b}\|_2$ at each iteration.

```
figure(1)
semilogy(misfit, 'ro--')
xlabel('iteration')
title('relative 2-norm misfit')
legend('RKFIT')
```



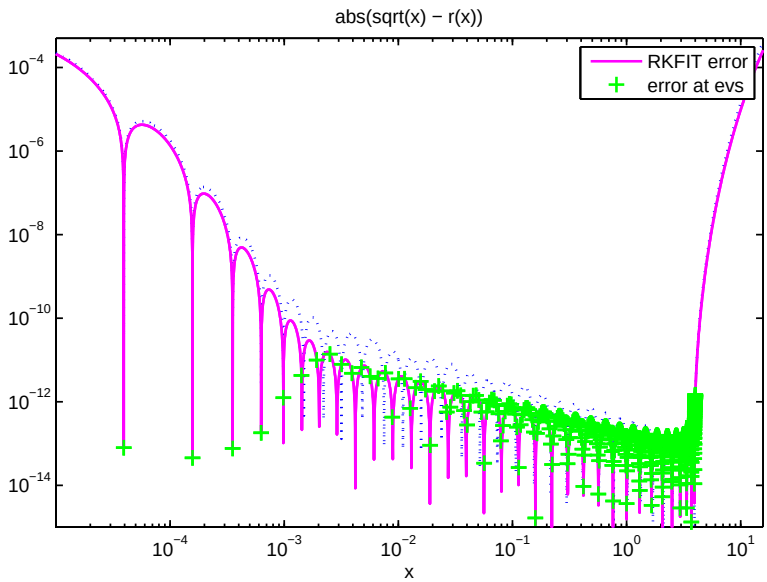
3 Evaluating the rational approximant

The second output `ratfun` is an object that can be used to evaluate the computed rational approximant. This evaluation is implemented in two ways. The first option is to evaluate a matrix function $r(A)\boldsymbol{b}$ by calling `ratfun(A,b)` with two input arguments. For example, here we are calculating the absolute misfit or r:

```
norm(F*b - ratfun(A, b))

ans =
   1.9712e-13
```

Alternatively, we can evaluate $r(z)$ pointwise by giving only one input argument. Let us plot the modulus of the scalar error function $\mathrm{err}(x) = f(x) - r(x)$ over the spectral interval of A, which is approximately $[0, 4]$.

```
figure(2)
xx = sort([logspace(-6,log10(15.728), 5000), ee ]);
errf = f(xx) - ratfun(xx);
loglog(xx,abs(errf), 'm-')
hold on
errf = f(ee) - ratfun(ee);
loglog(ee,abs(errf), 'g+')
xlabel('x')
title('abs(sqrt(x) - r(x))')
legend('RKFIT error','error at evs')
```



How often does this error function change sign on $[0, 4]$ (or rather on the discretisation set `xx`, which was chosen very fine with 5000 points)? Equivalently, how many interpolation points are there?

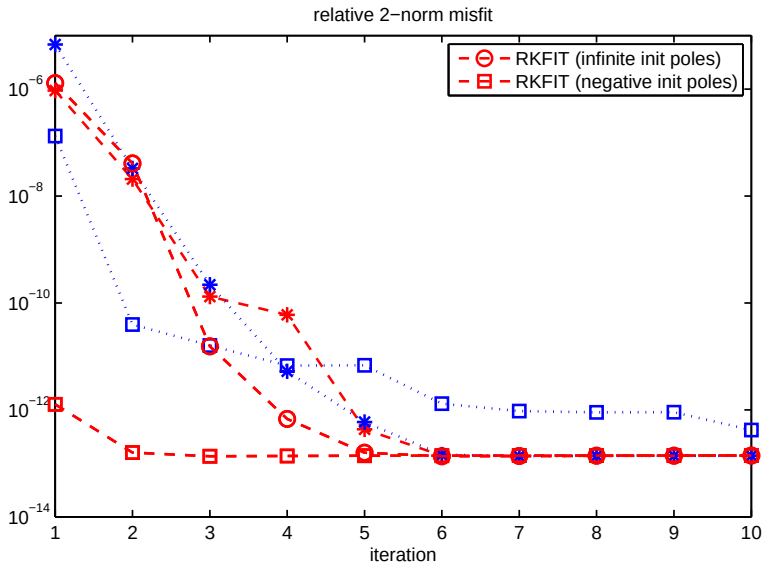
```
sc = length(find(diff(sign(errf))));
disp(['The error function has ' num2str(sc) ...
      ' sign changes on [0,4].'])
```

```
The error function has 32 sign changes on [0,4].
```

4 Some different choices for the initial poles

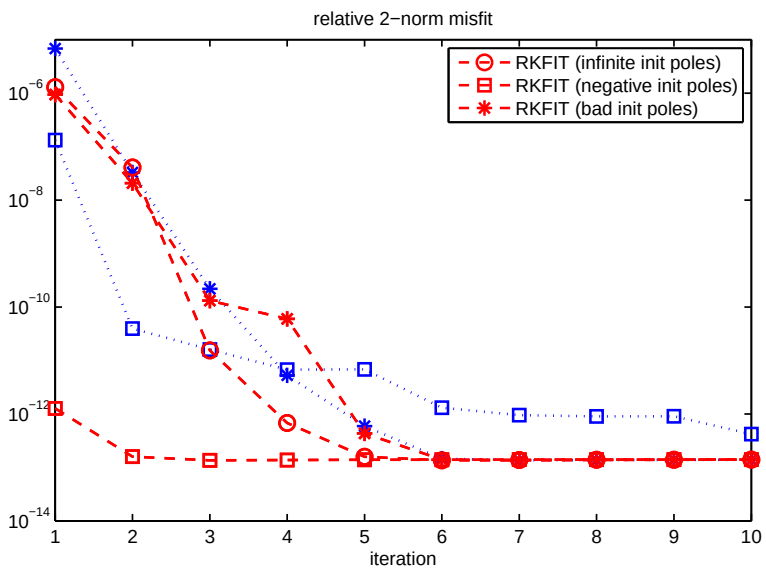
Choosing all initial poles equal to infinity seemed to work fine. Of course we can achieve convergence in fewer iterations by choosing negative initial poles, e.g., logarithmically spaced on the negative real axis. Let us rerun RKFIT with the new initial poles and compare the convergence.

```
[xi, ratfun, misfit] = rkfit(F, A, b, -logspace(-8, 8, m), ...
                             10, 'real');
figure(1), hold on
semilogy(misfit, 'rs--')
legend('RKFIT (infinite init poles)', ...
       'RKFIT (negative init poles)')
```



Now let us try a very bad choice for the initial poles, which is to place them right into the spectral interval $[0, 4]$ of A:

```
[xi, ratfun, misfit] = rkfit(F, A, b, linspace(0, 4, m), ...
                                 10, 'real');
semilogy(misfit, 'r*--')
legend('RKFIT (infinite init poles)', ...
       'RKFIT (negative init poles)', ...
       'RKFIT (bad init poles)')
```



In this example, RKFIT is not very affected by the choice of the initial poles and converges robustly.

5 Comparison with vector fitting

We now compare RKFIT with the vector fitting code described in [4, 5].

The following code makes sure the `vectfit3` implementation of VFIT is on MATLAB's search path, and if it is, defines some options.

```
if exist('vectfit3') ~= 2
  warning('Vector fitting VFIT3 not found in the Matlab path.');
  warning('VFIT can be downloaded from:')
  warning('http://www.sintef.no/Projectweb/VECTFIT/Downloads/VFUT3/')
  warning('Skipping comparison with VFIT3.');
  return
end

xi = -logspace(-8, 8, m);
opts.relax  = 1;   % Relaxed non-triviality constraint.
opts.stable = 0;   % Do not enforce stable poles.
opts.asymp  = 2;   % Options are 1, 2, or 3.
opts.skip_pole = 0;
opts.skip_res  = 0;
opts.spy1 = 0;    opts.spy2 = 0;
opts.logx = 0;    opts.logy = 0;
opts.errplot = 0; opts.phaseplot = 0;
```

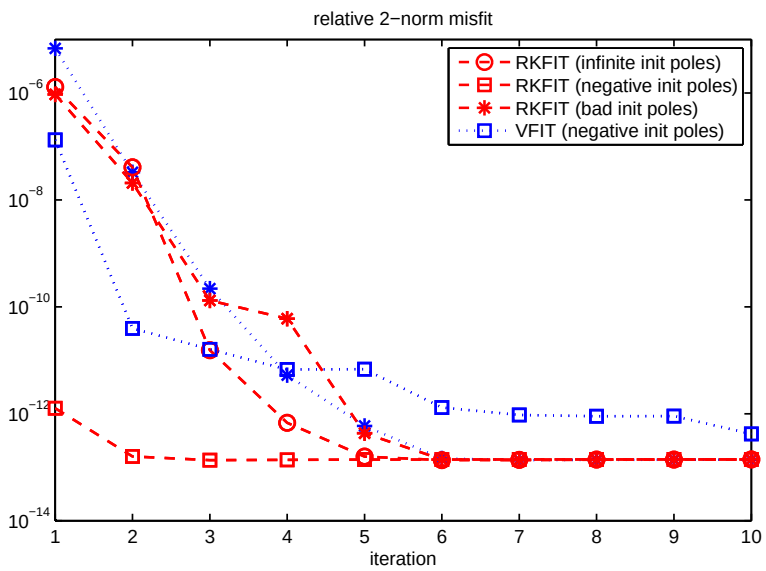
For VFIT3 we need interpolation nodes and weights, and in this example these need to be chosen as the eigenvalues of A and the components of $\boldsymbol{b}$ in the eigenvector basis of A, respectively.

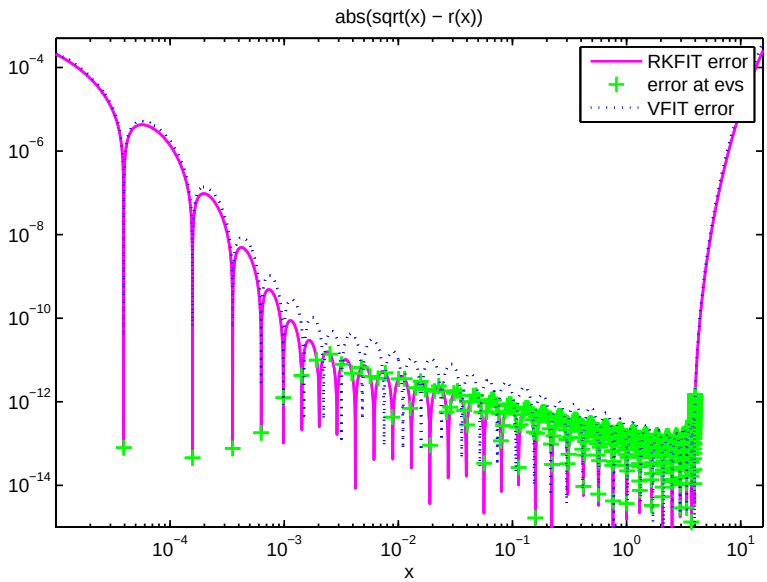
Note: *The below warnings outputted by VFIT are due to ill-conditioned linear algebra problems being solved, a problem that is circumvented with RKFIT by the use of discrete-orthogonal rational basis functions.*

```
nodes = diag(D); nodes = nodes(:).';
fvals = f(nodes);
weights = U'*b; weights = weights(:).';

for iter = 1:10
  [SER, xi, rmserr, fit] = ...
      vectfit3(fvals, nodes, xi, weights, opts);
  err_vfit(iter) = norm((fvals - fit).*weights)/norm(exact);
end

figure(1)
```

```
semilogy(err_vfit,'bs:')
legend('RKFIT (infinite init poles)','RKFIT (negative init poles)',...
       'RKFIT (bad init poles)','VFIT (negative init poles)')
axis([1, 10, 1e-14, 1e-5])

ffit = @(zz) arrayfun(@(z) ...
                      sum((SER.C).'./(z-diag(SER.A))) + ...
                      SER.D + z*SER.E,zz);
figure(2)
loglog(xx,abs(f(xx) - ffit(xx)), 'b:')
axis([1e-5, 15.7, 1e-15, 5e-4])
set(gca, 'XTick', 10.^(-5:1))
legend('RKFIT error', 'error at evs', 'VFIT error')
```

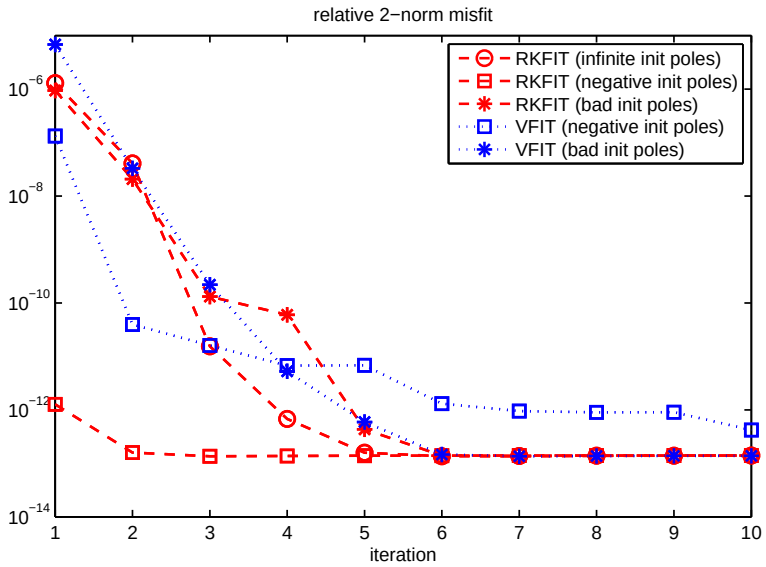
```
Warning: Matrix is close to singular or badly scaled. Results may be inaccurate.
RCOND =  1.702457e-16.
Warning: Rank deficient, rank = 15, tol =  7.021667e-15.
Warning: Rank deficient, rank = 15, tol =  7.021667e-15.
Warning: Rank deficient, rank = 16, tol =  7.021667e-15.
Warning: Rank deficient, rank = 16, tol =  7.021667e-15.
Warning: Rank deficient, rank = 16, tol =  7.021667e-15.
Warning: Rank deficient, rank = 16, tol =  7.021667e-15.
Warning: Matrix is close to singular or badly scaled. Results may be inaccurate.
RCOND =  2.037402e-16.
Warning: Rank deficient, rank = 16, tol =  7.021667e-15.
Warning: Matrix is close to singular or badly scaled. Results may be inaccurate.
RCOND =  2.012040e-16.
```





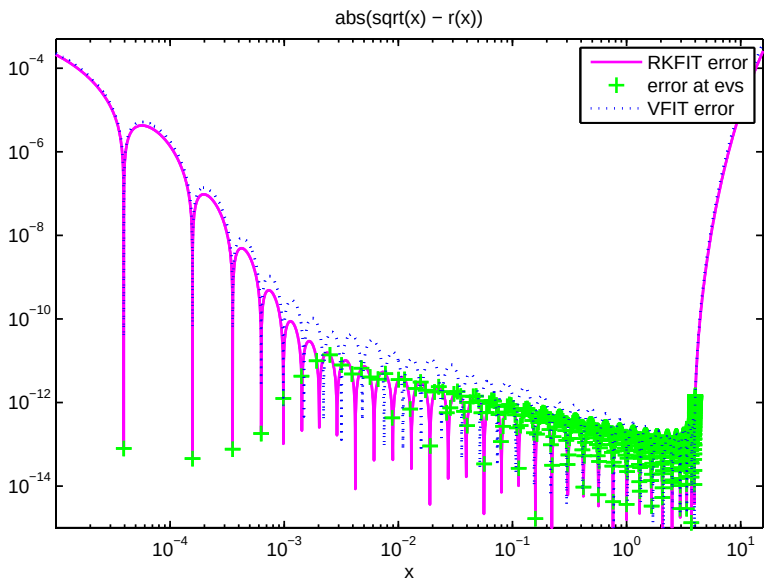
And the same with the "bad" initial poles.

```
xi = linspace(0, 4, m);
for iter = 1:10
  [SER, xi, rmserr, fit] = ...
      vectfit3(fvals, nodes, xi, weights, opts);
  err_vfit(iter) = norm((fvals - fit).*weights)/norm(exact);
end
figure(1)
semilogy(err_vfit, 'b*:')
legend('RKFIT (infinite init poles)', ...
       'RKFIT (negative init poles)', ...
       'RKFIT (bad init poles)',      ...
       'VFIT (negative init poles)',  ...
       'VFIT (bad init poles)')
```



This is the end of this example. The following is just to create a nice thumbnail.

```
figure(2), plot(NaN)
```



6 References

[1] M. Berljafa and S. Güttel. *A Rational Krylov Toolbox for MATLAB,* MIMS EPrint 2014.56 (http://eprints.ma.man.ac.uk/2199/), Manchester Institute for Mathematical Sciences, The University of Manchester, UK, 2014.

[2] M. Berljafa and S. Güttel. *Generalized rational Krylov decompositions with an application to rational approximation,* SIAM. J. Matrix Anal. Appl., 36(2):894–916, 2015.

[3] M. Berljafa and S. Güttel. *The RKFIT algorithm for nonlinear rational approximation,* MIMS EPrint 2015.38 (http://eprints.ma.man.ac.uk/2309/), Manchester Institute for Mathematical Sciences, The University of Manchester, UK, 2015.

[4] B. Gustavsen. *Improving the pole relocating properties of vector fitting,* IEEE Trans. Power Del., 21(3):1587–1592, 2006.

[5] B. Gustavsen and A. Semlyen. *Rational approximation of frequency domain responses by vector fitting,* IEEE Trans. Power Del., 14(3):1052–1061, 1999.