

# Scientific Computing - Homework 1

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## 1 Page 45, 1.7 (Scientific Computing: An Introductory Survey)

(a) The accurate value of  $\sec(1)$  is approximately 3.42551.

For  $h = 10^{-k}, k = 0, \dots, 16$ , the absolute error of the finite difference method first decreases as  $k$  increases. After the absolute error achieves its minimum when  $k = 8$ , it increases as  $k$  increases.

This agrees with  $\sqrt{\epsilon_{mach}} = 1.49012 \times 10^{-8}$ .

(b) For central difference method, the curve of error is similar but achieves its minimum when  $k = 7$ , a little earlier than the finite difference method.

The following figure shows the variation of  $\log(\text{absolute error})$  and  $\log(h)$ .

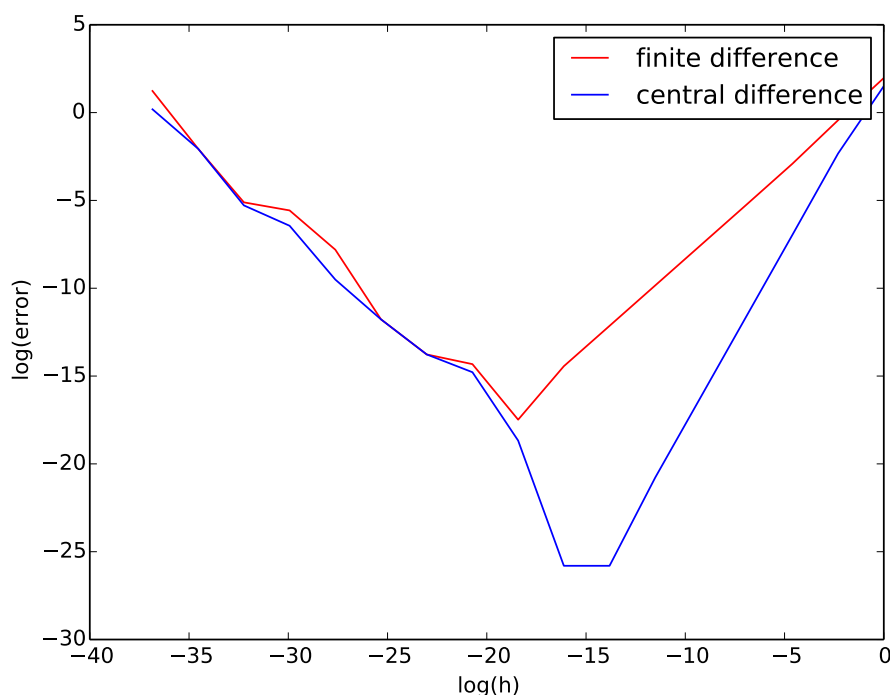


Figure 1: Finite Difference Error and Central Difference Error

## 2 Page 46, 1.9 (Scientific Computing: An Introductory Survey)

(b) Since the Taylor expansion has the formulation  $e^x = \sum_{i=0}^{+\infty} \frac{x^i}{i!}$ ,

I set up the stopping criterion as  $\left| \frac{x^i}{i!} \right| < \epsilon$ , where  $\epsilon = 10^{-9}$ .

(c) For  $x = 1, 5, 10, 15, 20$ , the Taylor expansion results are similar to the built-in function  $\exp(x)$ .

Both the absolute error and the relative error are satisfyingly small if I set  $\epsilon$  small enough.

For  $x = -1, -5, -10, -15, -20$ , the relative error is quite bigger than above.

When  $x = -20$ , the relative error is always bigger than 1 even when  $\epsilon < 10^{-20}$ . In this case summing in the natural order is no longer useful.

(d) For  $x < 0$ , I modified the program as  $e^{-x} = \frac{1}{e^x}$  and find out when  $x = -1, -5, -10, -15, -20$ , the absolute error and relative error are both satisfyingly small (smaller than  $10^{-10}$ ).

(e) I think not. However I rearrange the terms, the results will be still closed to 0 when  $x \ll 0$ .

In this case rearranging the terms can not alleviate the rounding error.

### 3 Matlab test

For  $n = 500, 1000, 2000, 4000, 8000$ , generate an  $n \times n$  random matrix,  $B$ , and an  $n \times 1$  vector,  $b$ . Find the symmetric matrix  $A = B' * B$ .

- 1) Using function 'eig' to test the time cost for eigen decomposition;
- 2) Using  $x = A \backslash b$  and  $x = A^{-1} * b$  to test the time cost in finding the solution of  $Ax = b$ ;
- 3) Plot all the time costs as a function of  $n$  and the power law for the time costs.

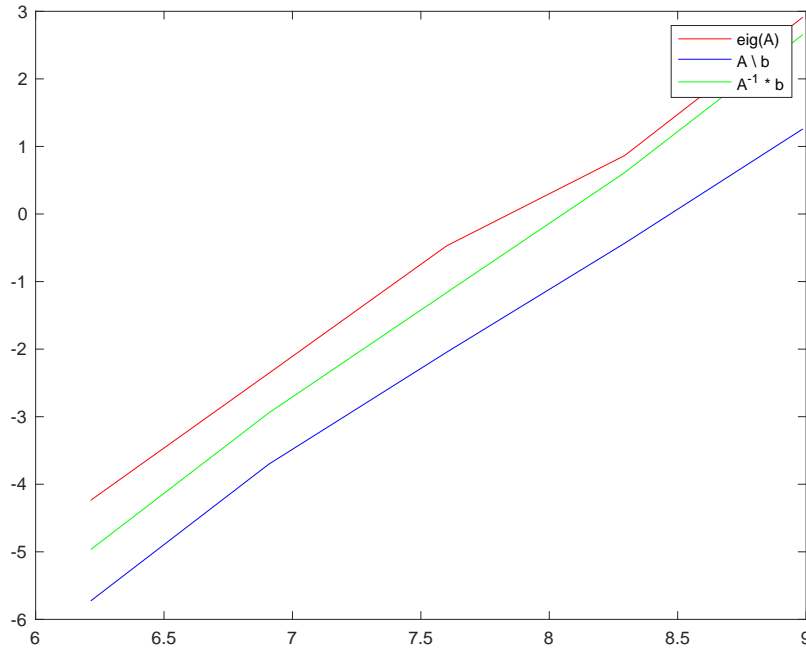


Figure 2: Matlab test

To solve the equation  $Ax = b$ , computing  $x = A^{-1} * b$  is much slower than computing  $x = A \backslash b$ .

In the above figure, the slopes of the three lines are all approximately 3, which means the time cost of  $\text{eig}(A)$ ,  $A \backslash b$  and  $A^{-1} * b$  are all proportional to  $n^3$ .