

## Problem 1

Recursion in the forward direction is given by:

$$J_n(x) = \frac{2(n-1)}{x} J_{n-1}(x) - J_{n-2}(x)$$

Code 1: Forward code in C

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```
#include <stdio.h>
#include <stdlib.h>

#define SIZE 11

double * getBesselArray(int x, double * J){
    /* function to get J_i(x) for i = 2 to 10; returns filled array
    x = 1, 5, 50; J = pointer to first element of array */

    for (int i = 2; i < SIZE; i++){
        J[i] = (2 * (i-1) * J[i-1] / x) - J[i-2];
    }
    return J;
}

int main(){

    /* initialising first two elements with J_0 and J_1 */

    double * J1 = malloc(sizeof(double) * SIZE);
    J1[0] = 0.76519, J1[1] = 0.44005;

    double * J5 = malloc(sizeof(double) * SIZE);
    J5[0] = -0.17759, J5[1] = -0.32757;

    double * J50 = malloc(sizeof(double) * SIZE);
    J50[0] = 0.055812, J50[1] = -0.097511;

    J1 = getBesselArray(1, J1);
    J5 = getBesselArray(5, J5);
    J50 = getBesselArray(50, J50);

    printf("J_10(1) = %f\n", J1[10]);
    printf("J_10(5) = %f\n", J5[10]);
    printf("J_10(50) = %f\n", J50[10]);

    free(J1); free(J5); free(J50);

    return 0;
}
```

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The output of the following code is:

$$J_{10}(1) = 560.553310 \quad (1)$$

$$J_{10}(5) = 0.001585 \quad (2)$$

$$J_{10}(50) = -0.113847 \quad (3)$$

Values	x = 1	x = 5	x = 50
Absolute value (p)	$2.53 \times 10^{-10}$	$1.47 \times 10^{-3}$	-0.113847849
Calculated value ( $p^*$ )	560.5533	$1.59 \times 10^{-3}$	-0.113847000
Absolute error ( $ p - p^* $ )	560.5533	$1.17 \times 10^{-4}$	$8.49 \times 10^{-7}$
Relative error ( $\frac{ p-p^* }{ p }$ )	$2.13 \times 10^8$	0.079	$7.46 \times 10^{-6}$

We can see that the relative errors are unacceptably large for  $x = 1$ , just acceptable for  $x = 5$ , and quite good for  $x = 50$ .

## Problem 2

Recursion in the backward direction is given by:

$$J_n(x) = \frac{2(n+1)}{x} J_{n+1}(x) - J_{n+2}(x)$$

Code 2: Backward code in C

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```
double * getBesselArray(int x, double * J){
    /* function to get J_i(x) for i = 8 to 0 */

    for (int i = 8; i >= 0; i--){
        J[i] = (2 * (i+1) * J[i+1] / x) - J[i+2];
    }
    return J;
}
```

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The output is:

$$J_0(1) = 0.765190 \quad (4)$$

$$J_0(5) = -0.177594 \quad (5)$$

$$J_0(50) = 0.055807 \quad (6)$$

Values	x = 1	x = 5	x = 50
Absolute value (p)	$7.65197686 \times 10^{-1}$	$-1.7759677131 \times 10^{-1}$	$5.5812 \times 10^{-2}$
Calculated value ( $p^*$ )	$7.65190000 \times 10^{-1}$	$-1.7759400000 \times 10^{-1}$	$5.5807 \times 10^{-2}$
Absolute error ( $ p - p^* $ )	$7.686 \times 10^{-6}$	$2.77131 \times 10^{-6}$	$5 \times 10^{-6}$
Relative error ( $\frac{ p-p^* }{ p }$ )	$1.0044 \times 10^{-5}$	$1.56045 \times 10^{-5}$	$8.9586 \times 10^{-5}$

Clearly, the backward recursion has much less error than the forward one for all values of  $x$  (around  $\frac{1}{1000}^{th}$  to  $\frac{1}{100}^{th}$  percentage relative error).

### Problem 3

Can the error propagation be formally analyzed using the difference equation analysis we performed in class?

I think so. I'll try:

Say there is an error in the recording of the observations  $J_0(x)$  and  $J_1(x)$ ,  $= \Delta x_0$  and  $\Delta x_1$  respectively. Then, what we are actually calculating is:

$$\begin{aligned} J_n^*(x) &= \frac{2(n-1)}{x} [J_{n-1}(x) + \Delta x_1] - [J_{n-2}(x) + \Delta x_0] \\ \epsilon_n &= |J_n(x) - J_n^*(x)| \\ &= \frac{2(n-1)}{x} \Delta x_1 - \Delta x_0 \\ \epsilon_{n+1} &= \frac{2n}{x} \epsilon_n - \epsilon_{n-1} \\ \frac{\epsilon_{n+1}}{\epsilon_n} &= \frac{2n}{x} - \frac{\epsilon_{n-1}}{\epsilon_n} \end{aligned}$$

i.e., Current error rate  $= \frac{2n}{x}$  - Previous error rate. Let's call them  $R_{n+1}$  and  $R_n$  respectively,

$$\begin{aligned} R_{n+1} &= \frac{2n}{x} - R_n \\ \frac{R_{n+1}}{R_n} &= \frac{2n}{x} - 1 \\ \frac{R_{n+1}}{R_n} &\propto \frac{2n}{x} \end{aligned}$$

So, the rate of change of error rate is *directly* proportional to  $n$ , and *inversely* proportional to  $x$ .

Can the error behavior be understood by this analysis?

Yes. Since the rate of growth of error rate is *directly* proportional to  $n$ , we can write:  
(if we were working with continuous data, we would write in terms of derivatives, but since we have discrete data)

$$\begin{aligned}\frac{\Delta^2 \epsilon}{\Delta^2 n} &= k_1 \\ \frac{\Delta \epsilon}{\Delta n} &= k_1 n + k_2 \\ \epsilon &= k_1 n^2 + k_2 n + k_3\end{aligned}$$

So, the error grows quadratically with increasing  $n$ ; and that is why we see unconditional error growth in the case of forward recursion (since we are increasing  $n$ ), but not in the case of backward recursion (decreasing  $n$ ).

Also, we see most error when  $x = 1$ , less when  $x = 5$ , and even less so when  $x = 50$ ; since the rate of growth of error rate is *inversely* proportional to  $x$ .