

# MAT 423: Solution homework 2 (09/05)

**Ex 1.** [2pts] See scripts at the end.

**Ex 2.** [4pts] + [extra 2pts] Let  $f(x) = xe^x - 1$ .

- a) We use bisection method with  $a = 0$  and  $b = 1$ . In order to have precision  $\varepsilon = 10^{-16}$ , a sufficient condition on the number of iteration  $n$  is given by:

$$\frac{1}{2^{n+1}} \cdot |b - a| < \varepsilon \quad \Rightarrow \quad -(n + 1) \ln 2 < -16 \ln 10 \quad \Rightarrow \quad n > 53.$$

1pt

After  $n = 53$  iterations, we obtain:

$$x_{ref} = 0.567143290409784.$$

1pt

- b) Using the reference solution  $x_{ref}$ , we estimate the error  $e_n = |x_n - x_{ref}|$  for the three methods at after each iterations. We use  $x_0 = 1$  as initial point for the Newton method,  $(x_0, x_1) = (0, .1)$  for the secant method. The errors are plotted in figure 1 with log scale in the y-axis.

1pt

1pt

- c\*) [extra +2pts] To further estimate the decay of each method, we need to estimate the parameters  $C$  and  $\alpha$  such that:  $\frac{e_{n+1}}{e_n} \approx C$  in other words  $e_{n+1} \approx Ce_n^\alpha$ . To make this relationship linear, we take the logarithm:

$$\log e_{n+1} \approx \log C + \alpha \log e_n.$$

Therefore, denoting  $y = \log e_{n+1}$  and  $x = \log e_n$ , we have:

$$y \approx \beta + \alpha x,$$

with  $\beta = \log C$ . We plot the data points  $(x_i, y_i)_i$  for the three methods and estimate the linear-relationship in figure

+1pt

We obtain the following constant:

+1pt

- bisection method:  $\alpha = .991 \approx 1$  and  $C = .438 \approx \frac{1}{2}$ .
- secant method:  $\alpha = 1.62 \approx \frac{1+\sqrt{5}}{2}$  and  $C = .926$ .
- Newton method:  $\alpha = 1.94 \approx 2$  and  $C = .713$ .

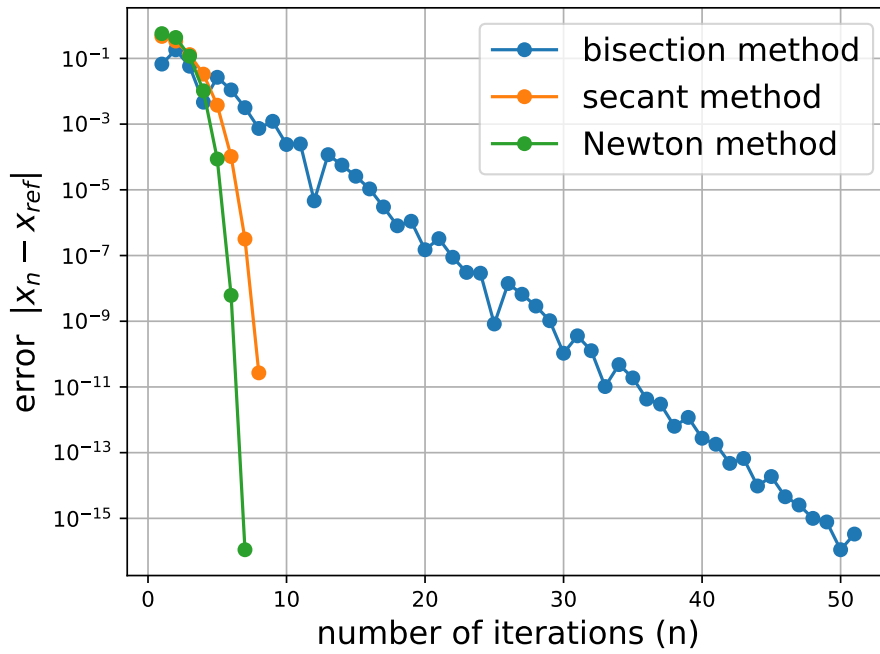


Figure 1: Evolution of the error  $e_n = |x_n - x_{ref}|$  for the three methods. In log-scale, the convergence of the bisection is linear whereas secant and Newton methods are faster.

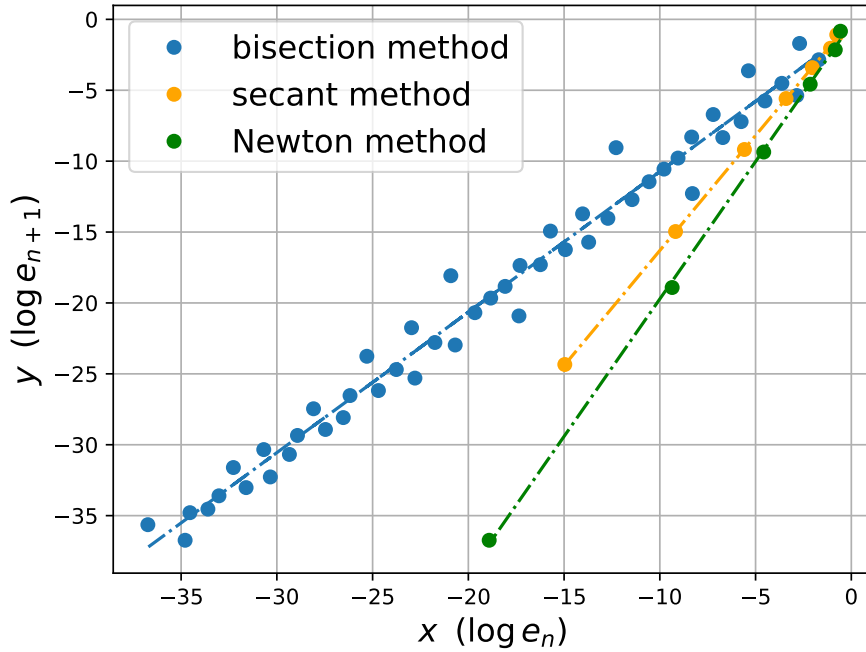


Figure 2: Estimation of the relationship  $e_{n+1} \approx C \cdot e_n^\alpha$  using log-log plot.

**Ex 3. 4pts**

Let  $\varphi(x) = \sqrt{x+1}$ .

a) We show that  $\varphi$  is a **contraction** on  $I = [0, \infty)$ .

◦ For any  $x \geq 0$ ,  $\varphi(x) \geq 0$ . Thus,  $\varphi(I) \subset I$ . .5pt

◦  $|\varphi'(x)| = \left| \frac{1}{2(x+1)} \right| \leq \frac{1}{2}$  on  $I$ . .5+.5pt

Thus, for any  $x, y \in I$ , we have:  $|\varphi(x) - \varphi(y)| \leq \frac{1}{2}|x - y|$ .

Therefore,  $\varphi$  is a contraction on  $I$ . We deduce that there **exists a unique fixed point**  $x_*$  of  $\varphi$  on  $I$ . .5pt

b) Let  $J = [1, 2]$ . We have  $\varphi(J) \subset J$ , thus we can narrow our interval ( $x_* \in J$ ).

Consider the sequence  $\{x_n\}_n$  defined recursively by:  $x_{n+1} = \varphi(x_n)$  with  $x_0 = 1$ . We have:

$$|x_n - x_*| \leq \frac{k^n}{1 - k} \cdot |b - a| = 1 \cdot \left(\frac{1}{2}\right)^{n-1}. \quad \text{span style="float: right; border: 1px solid black; padding: 2px;">1pt}$$

Thus, in order to have the error  $e_n = |x_n - x_*|$  less than  $10^{-4}$ , a sufficient condition is:

$$\frac{1}{2^{n-1}} \leq 10^{-4} \Rightarrow n \geq \frac{4 \ln 10}{\ln 2} + 1 \approx 14.3. \quad \text{span style="float: right; border: 1px solid black; padding: 2px;">1pt}$$

After **15 iterations**, the error is less than  $10^{-4}$ .

**Ex 4.** Let  $\varphi(x) = \sqrt{x^2 + 1}$  and  $I = [0, \infty)$ .

a) By the **mean value theorem**:

$$|\varphi(x) - \varphi(y)| = |\varphi'(c)(x - y)|,$$

with  $c \in [x, y]$ . Since  $|\varphi'(s)| = \left| \frac{s}{\sqrt{s^2+1}} \right| < 1$ , we deduce that:  $|\varphi(x) - \varphi(y)| < |x - y|$ .

b) We notice that  $x_n = \sqrt{n}$ . Thus, the sequence does not converge.

c)  $\varphi(x) = x$  implies:  $\sqrt{x^2 + 1} = x$  which is not possible. Thus,  $\varphi$  does not have a fixed point.

This does not contradict the *fixed-point theorem* since  $\varphi$  is **not** a contraction: **it does not exist**  $k < 1$  such that  $|\varphi(x) - \varphi(y)| \leq k|x - y|$ .

```

%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%
%%      Newton/secant methods (MATLAB/OCTAVE)      %%
%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%
function saveX=Newton_method(f,df,x0,N)
%% Newton Method in 1D:
%%    $x_{n+1} = x_n - f(x_n)/f'(x_n)$ 
%%
%% A) initialisation
saveX = zeros(1,N+1);
x = x0;
saveX(1) = x0;
%%-----%%
%% B)      Loop      %%
%%-----%%
for i=1:N
    x = x - f(x)/df(x);      % update
    saveX(i+1) = x;          % save
end
end

function saveX=secant_method(f,x0,x1,N)
%% secant Method in 1D:
%%    $x_{n+1} = x_n - f(x_n)*(x_n-x_{n-1})/(f(x_n)-f(x_{n-1}))$ 
%%
%% A) initialisation
saveX = zeros(1,N+1);
x_nm = x0;
x_n = x1;
saveX(1) = x1;              % does not keep x0
%%-----%%
%% B)      Loop      %%
%%-----%%
for i=1:N
    x_np = x_n - f(x_n)*(x_n-x_nm)/(f(x_n)-f(x_nm));
    saveX(i+1) = x_np;      % save
    x_nm = x_n;
    x_n = x_np;
end
end
end

```

```

#####
##          Newton/secant methods (Python)          ##
#####
import numpy as np

def Newton_method(f,df,x0,N):
    ''' Newton Method in 1D:
     $x_{n+1} = x_n - f(x_n)/f'(x_n)$ 
    '''
    ## A) initialisation
    saveX = np.zeros(N+1)
    x = x0
    saveX[0] = x0
    ##-----##
    ## B)          Loop          ##
    ##-----##
    for i in range(N):
        x = x - f(x)/df(x)      # update
        saveX[i+1] = x          # save
    ## C) output
    return saveX

def secant_method(f,x0,x1,N):
    ''' secant Method in 1D:
     $x_{n+1} = x_n - f(x_n)*(x_n-x_{n-1})/(f(x_n)-f(x_{n-1}))$ 
    '''
    ## A) initialisation
    saveX = np.zeros(N+1)
    x_nm,x_n = x0,x1
    saveX[0] = x1                # does not keep x0
    ##-----##
    ## B)          Loop          ##
    ##-----##
    for i in range(N):
        x_np = x_n - f(x_n)*(x_n-x_nm)/(f(x_n)-f(x_nm))
        saveX[i+1] = x_np        # save
        x_nm = x_n
        x_n = x_np
    ## C) output
    return saveX

```

```

#####
##          Newton/secant methods (JULIA)          ##
#####
function Newton_method(f,df,x0,N)
    #= Newton Method in 1D:
     $x_{n+1} = x_n - f(x_n)/f'(x_n)$ 
    =#
    ## A) initialisation
    saveX = zeros(N+1)
    x = x0
    saveX[1] = x0
    ##-----##
    ## B)          Loop          ##
    ##-----##
    for i=1:N
        x = x - f(x)/df(x)      # update
        saveX[i+1] = x          # save
    end
    ## C) output
    return saveX
end

function secant_method(f,x0,x1,N)
    #= secant Method in 1D:
     $x_{n+1} = x_n - f(x_n)*(x_n-x_{n-1})/(f(x_n)-f(x_{n-1}))$ 
    =#
    ## A) initialisation
    saveX = zeros(N+1)
    x_nm,x_n = x0,x1
    saveX[1] = x1              # does not keep x0
    ##-----##
    ## B)          Loop          ##
    ##-----##
    for i=1:N
        x_np = x_n - f(x_n)*(x_n-x_nm)/(f(x_n)-f(x_nm)) # update (could be speed up...)
        saveX[i+1] = x_np      # save
        x_nm = x_n
        x_n = x_np
    end
    ## C) output
    return saveX
end

```