

```
In [1]: import numpy as np
from scipy.optimize import fsolve
from IPython.display import display, Markdown
from sympy import *
from sympy import latex as ltx

# Derivatives
x = symbols(r'x', real=True)
N1, N2 = symbols(r'N_1 N_2', real=True)
d1, d2, d3 = symbols(r'd_1 d_2 d_3', real=True)
N1 = x * d1 / ( 10 - d1) - 0.5 * d2**2
N2 = d2 - d1

display ( Markdown( f"$\frac{{\partial N_1}}{{\partial d_1}} = {\tex(diff(N1,d1))}$" ) )
display ( Markdown( f"$\frac{{\partial N_1}}{{\partial d_2}} = {\tex(diff(N1,d2))}$" ) )
display ( Markdown( f"$\frac{{\partial N_2}}{{\partial d_1}} = {\tex(diff(N2,d1))}$" ) )
display ( Markdown( f"$\frac{{\partial N_2}}{{\partial d_2}} = {\tex(diff(N2,d2))}$" ) )
```

$$\frac{\partial N_1}{\partial d_1} = \frac{d_1 x}{(10-d_1)^2} + \frac{x}{10-d_1}$$

$$\frac{\partial N_1}{\partial d_2} = -1.0 d_2$$

$$\frac{\partial N_2}{\partial d_1} = -1$$

$$\frac{\partial N_2}{\partial d_2} = 1$$

```
In [2]: import numpy as np
from scipy.optimize import fsolve

# HELPER
def cN1(x,d_) :
    if not len(d_) : return []
    return x * d_[0] / (10. - d_[0]) - 0.5 * d_[1]**2
def cN2(x,d_) :
    if not len(d_) : return []
    return d_[1] - d_[0]
def cN(x,d_) :
    if not len(d_) : return []
    return [ cN1(x,d_), cN2(x,d_) ]
def cG( dd_, R_) : return dd_ @ R_

#
# MAIN ENGINE
#
def run_sim( x, line_search=False, modif_nr=False, line_search_maxit=1000, line_search_fast_s=False, BFGS=False ) :
    global G0,d, delta_d, K_inv, v_, alpha_, w_, delta_R_, R_, R
    # Load steps
    dF_ = np.array( [ 0.25, 0 ] )
    F_ = np.zeros(2)

    s=1 # Line search
    RET_F, RET_D, RET_I = [], [], []

    # Load steps
    d_ = np.array( [ 0. , 0. ] )
    N_ = cN(x,d_)

    for n in range(40) :
        F_ += dF_

        #print(f"Running LOAD STEP {n} (F={F_})")
        CONVERGED = False

        # INITIALIZE THE RESIDUAL
        N_ = cN(x,d_)
        R_ = F_ - N_
        r0 = np.linalg.norm(R_)
        r = r0
        # NEWTON ITERATIONS
        for i in range(15) :
            #print(f" Netwon step {i} (r:{r:.4f})")
            # UPDATE THE TANGENT MATRIX
            if not modif_nr or not i :
                K = np.array( [ [ d_[0] * x / (10 - d_[0])**2 + x / (10 - d_[0]), -d_[1] ] ,
                                [ -1, 1 ] ] )

            # UPDATE BFGS PREDICTOR AFTER FIRST TIMESTEP
            if i and BFGS :
                Rbfgs_ = R_ + ( v_ @ R_ ) * w_
                dd0_ = np.linalg.inv( K ) @ Rbfgs_
                dd_ = dd0_ + ( w_ @ v_ ) * dd0_
            else :
                # UPDATE THE SEARCH DIRECTION
```

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dd_ = np.linalg.inv(K) @ R_

# UPDATE LINE SEARCH G0 (before d_ update)
g0 = cG( dd_, R_ )

# SAVE & UPDATE d_
prev_d_ = d_.copy()
d_ += dd_
# TRIAL RESIDUAL
tR_ = F_ - cN(x,d_)

# UPDATE LINE SEARCH G1 (after d_ update)
g1 = cG( d_, tR_ )

# UPDATE LINE SEARCH
if line_search :
    #print("          Line search started.")

    def foo(_s):
        gs = cG( dd_ ,          F_ - cN(x, prev_d_ + _s*dd_) )
        [N1,N2] = cN(x, prev_d_ + _s*dd_)
        if line_search_fast_s :
            if abs(gs) < abs(g0/2) :
                #print("          G(s) < G(0)/2. Enough.")
                return 0
        return gs

    s = fsolve(foo,0, maxfev=line_search_maxit)[0]

    test_f = abs(foo(s))
    if abs(test_f)>abs(g0/2)+1e-4 :
        print(f"[FAILED] Cannot find acceptable s in line search - test_f[{{test_f:.4e}} > abs(g0/2)[{{abs(g0/2):.4e}}]]")
        return RET_D, RET_F, RET_I

# UPDATE THE DOFs BASED ON THE LINE SEARCH
d_ = prev_d_ + dd_ * s

# SAVE & UPDATE R_
prev_R_ = R_.copy()
R_ = F_ - cN(x,d_)

# UPDATE GS
gs = cG( dd_, R_ )

r = np.linalg.norm(R_)
if r > 1e50 : break # Crash!
if r/r0 < 1e-4 :
    #print(f"          Converged (r/r0={{r/r0:.4e}})")
    CONVERGED = True
    break

# UPDATE BFGS VECTORS
if BFGS :
    v_ = dd_ / (gs - g0)
    alpha = np.sqrt( np.max( [ 0, - s * (gs - g0) / g0 ] ) )
    w_ = (alpha - 1) * R_ - prev_R_
    #print("          Update BFGS vectors.")

if not CONVERGED :
    print(f"[FAILED] The solver did not converge after {i} iterations. Giving up.")
    return RET_D, RET_F, RET_I

RET_D.append(d_.copy())
RET_I.append(i+1)
RET_F.append(F_[0])

return RET_D, RET_F, RET_I

```

```
#full_run( line_search=0, modif_nr=1, line_search_maxit=10000, line_search_fast_s=False, BFGS=1 )
```

1) Exact $N_1(d_1, d_2 = d_1)$ vs. d_1 for $x=15$ and $x=25$

```

In [3]: import matplotlib.pyplot as plt
import numpy as np

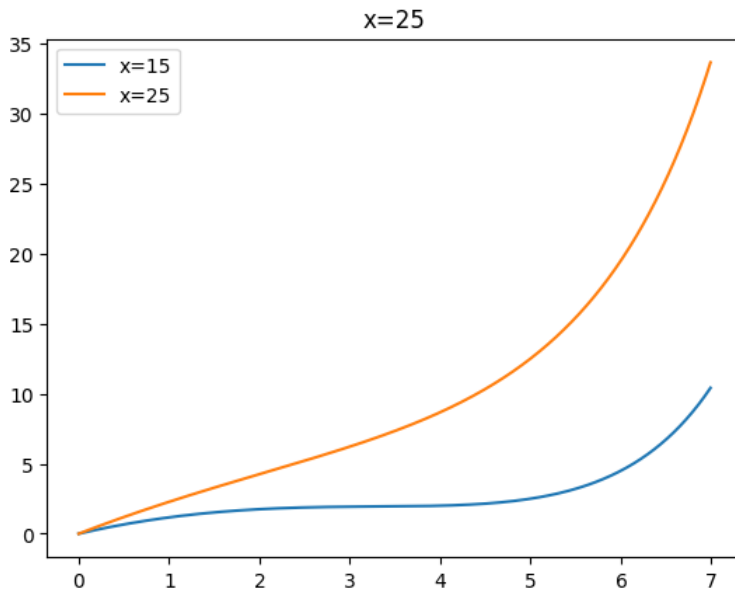
d0_ = np.arange(0,7,.01)
d0 = np.array( [ i for i in d0_ if i != 10] )

d0 = [ d0, d0 ]
N15exact = cN1(15,d0)
plt.plot( d0[0], N15exact, label="x=15" )
plt.title("x=15")
N25exact = cN1(25,d0)
plt.plot( d0[0], N25exact, label="x=25" )

```

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plt.title("x=25")
plt.legend()
```

Out[3]: <matplotlib.legend.Legend at 0x1f62d45ee10>



```
In [4]: import matplotlib.pyplot as plt
def full_run(line_search, modif_nr=False, line_search_maxit=1000, line_search_fast_s=False, BFGS=False) :
    x=15
    D15,F15,I15 = run_sim( x=x, line_search=line_search, modif_nr=modif_nr, line_search_maxit=line_search_maxit, line_search_fast_s=line_search_fast_s, BFGS=BFGS )
    d15 = np.array( [ [i,i] for i,j in D15 ] ).T
    N15 = cN1(x,d15)

    x=25
    D25,F25,I25 = run_sim( x=x, line_search=line_search, modif_nr=modif_nr, line_search_maxit=line_search_maxit, line_search_fast_s=line_search_fast_s, BFGS=BFGS )
    d25 = np.array( [ [i,i] for i,j in D25 ] ).T
    N25 = cN1(x,d25)

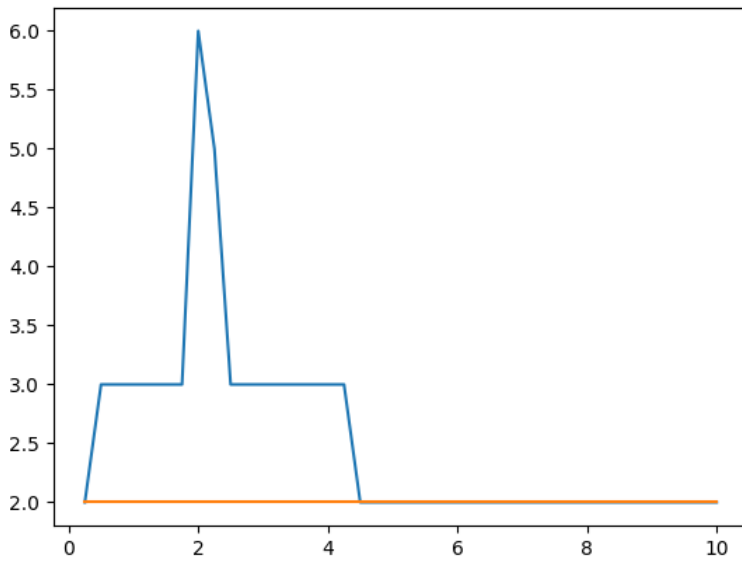
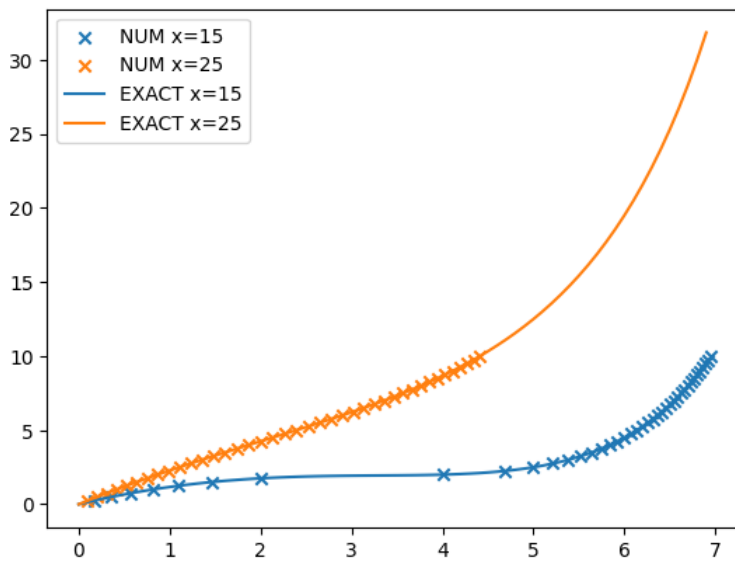
    # PLOT NUMERICAL
    if len(d15) : plt.scatter( d15[0], N15, label="NUM x=15", marker='x' )
    if len(d25) : plt.scatter( d25[0], N25, label="NUM x=25", marker='x' )
    # PLOT EXACT
    d0_ = np.arange(0,7,.1)
    d0 = np.array( [ i for i in d0_ if i != 10 ] )
    d0 = [ d0, d0 ]
    N15exact = cN1(15,d0)
    plt.plot( d0[0], N15exact, label="EXACT x=15" )
    N25exact = cN1(25,d0)
    plt.plot( d0[0], N25exact, label="EXACT x=25" )
    plt.legend()

    plt.figure()
    plt.plot(F15,I15,label="Iterations x=15")
    plt.plot(F25,I25,label="Iterations x=25")
```

Newton raphson - consistent tangent

The solver converged in both cases (x=15 and x=25)

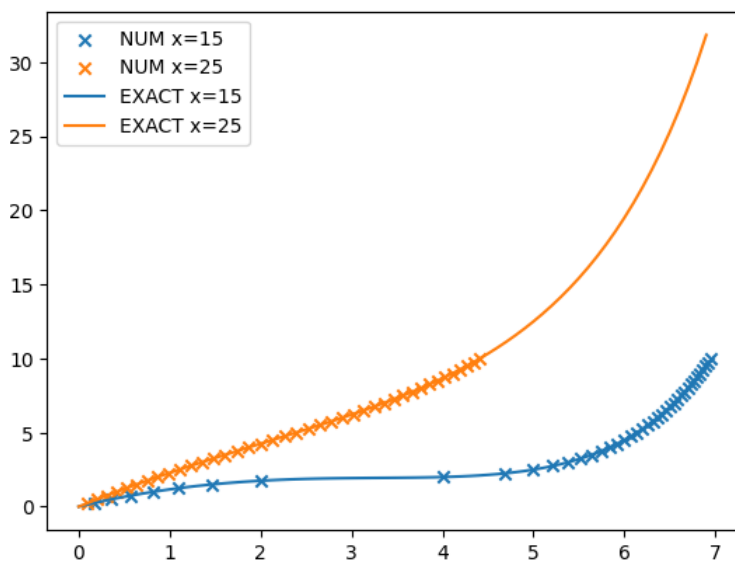
```
In [5]: full_run( line_search=0, modif_nr=False, line_search_maxit=10000, line_search_fast_s=False )
```

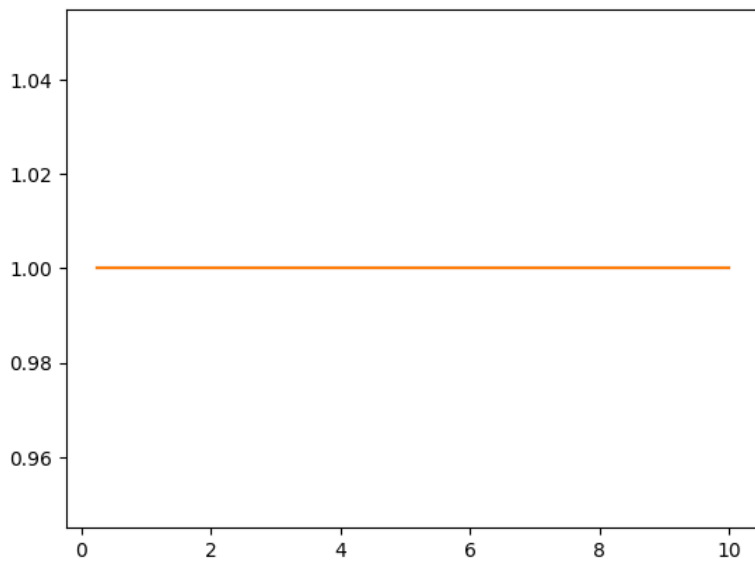


Newton raphson - consistent tangent + Line Search

The solver converged in both cases ($x=15$ and $x=25$), in one iteration

```
In [6]: full_run( line_search=True, modif_nr=False, line_search_maxit=10000, line_search_fast_s=False )
```



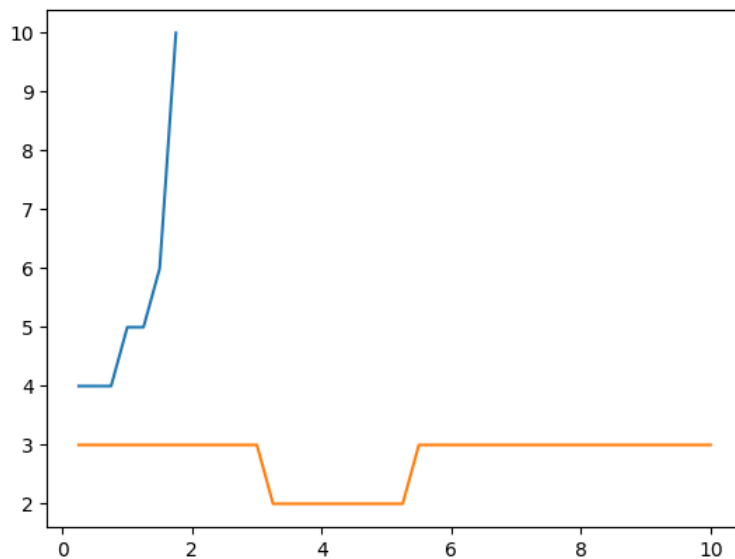
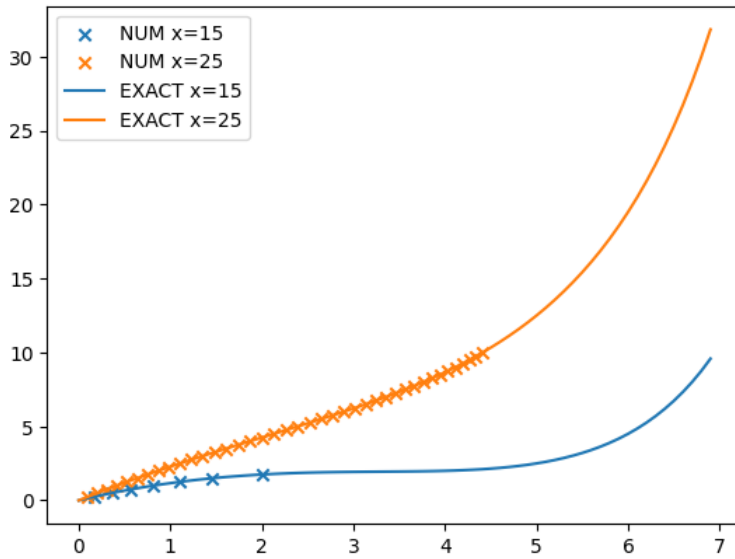


Modified Newton Raphson

The solver converged for x=25 but diverged after some loads for x=25.

```
In [7]: full_run( line_search=False, modif_nr=True, line_search_maxit=10000, line_search_fast_s=False )
```

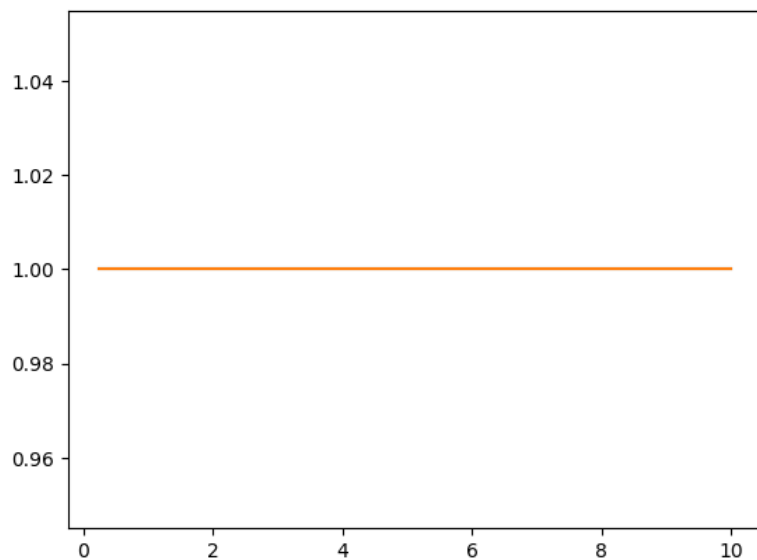
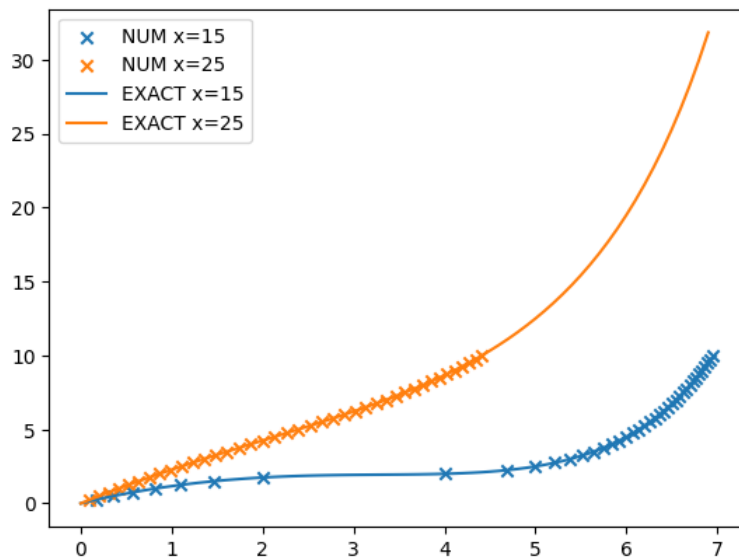
[FAILED] The solver did not converge after 14 iterations. Giving up.



Modified Newton Raphson + Line Search, full resolution

The solver converged in both cases ($x=15$ and $x=25$) in one iteration.

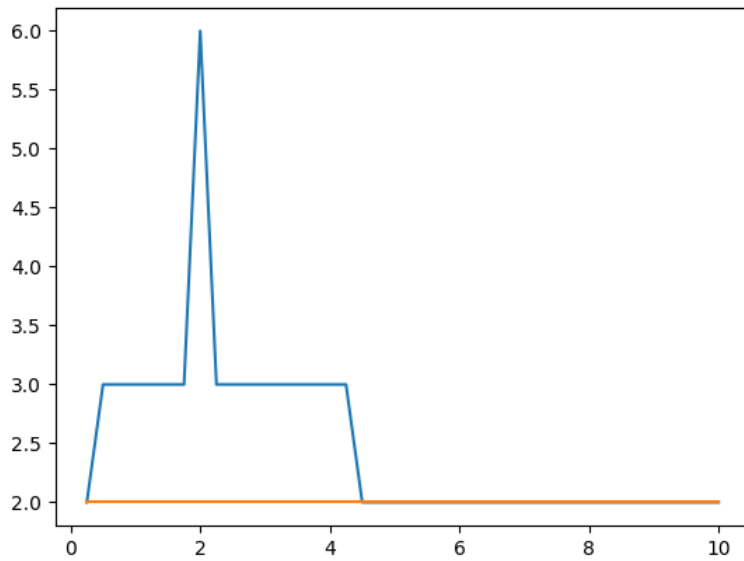
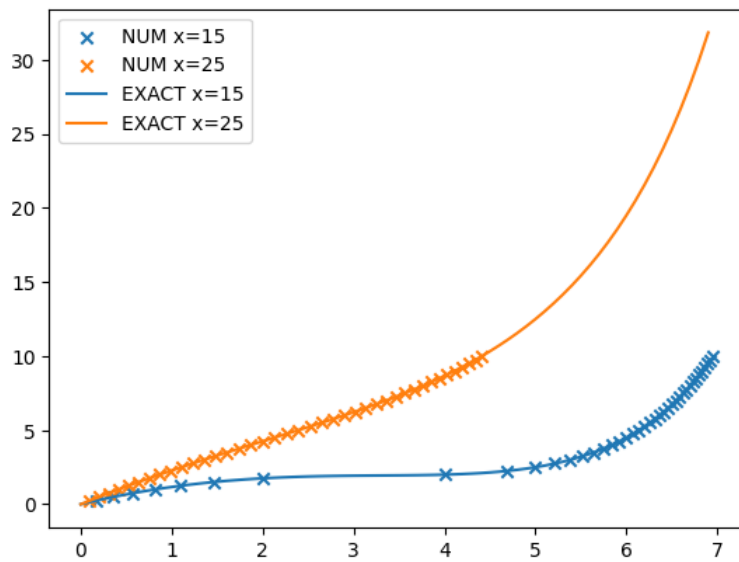
```
In [8]: full_run( line_search=True, modif_nr=True, line_search_maxit=1000, line_search_fast_s=False )
```



Modified Newton Raphson + Line Search, stop when $G(s) < 1/2 G(0)$

The solver converged in both cases ($x=15$ and $x=25$) in more than one iteration.

```
In [9]: full_run( line_search=True, modif_nr=True, line_search_maxit=1000, line_search_fast_s=True )
```

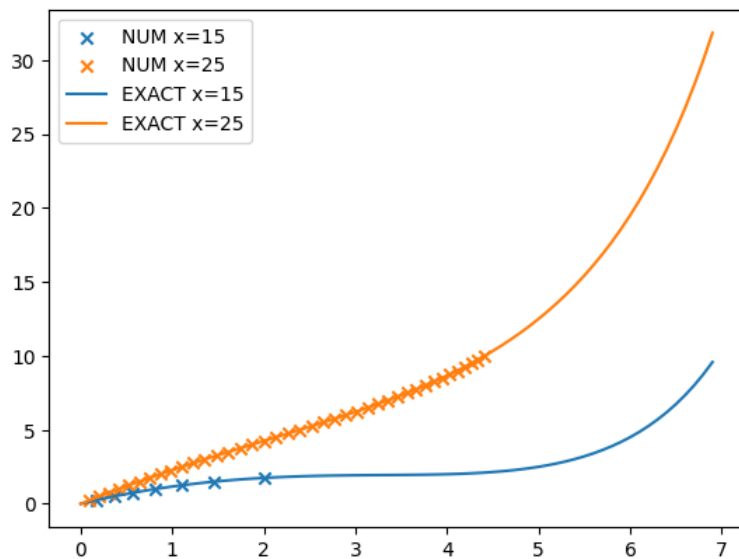


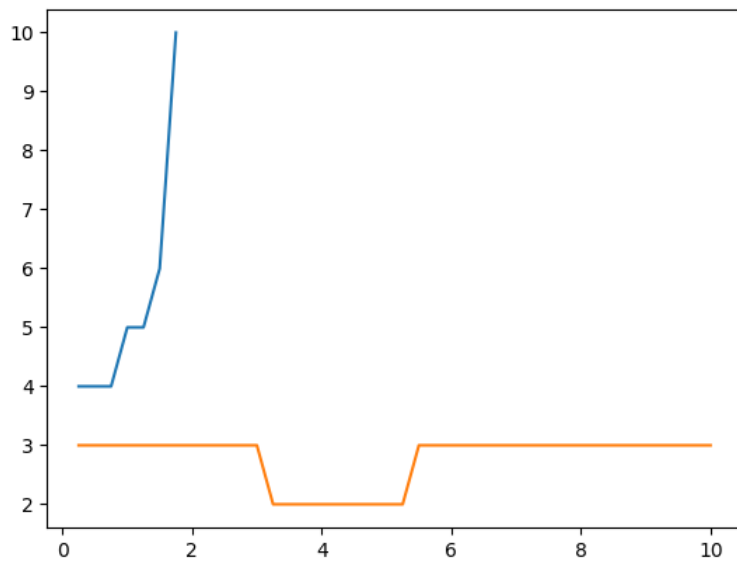
Modified Newton Raphson + Line Search, maximum of 5 iterations

The solver did not converge for x=15.

```
In [10]: full_run( line_search=False, modif_nr=True, line_search_maxit=5, line_search_fast_s=False, BFGS=False )
```

[FAILED] The solver did not converge after 14 iterations. Giving up.





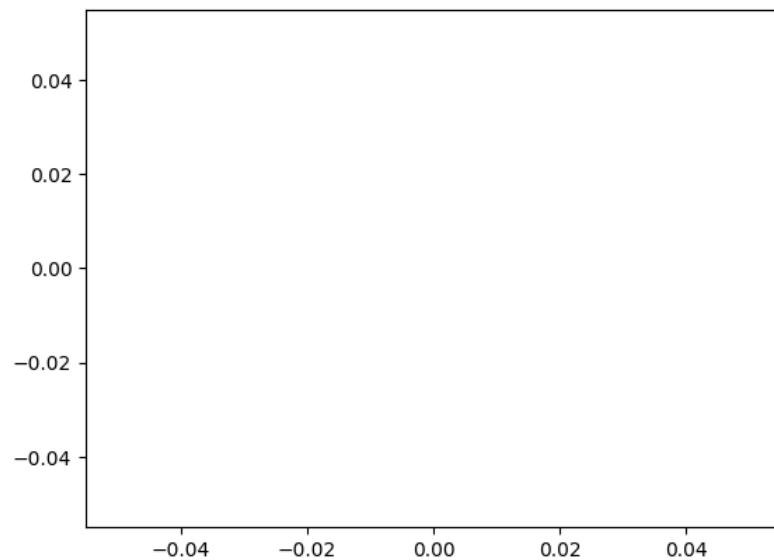
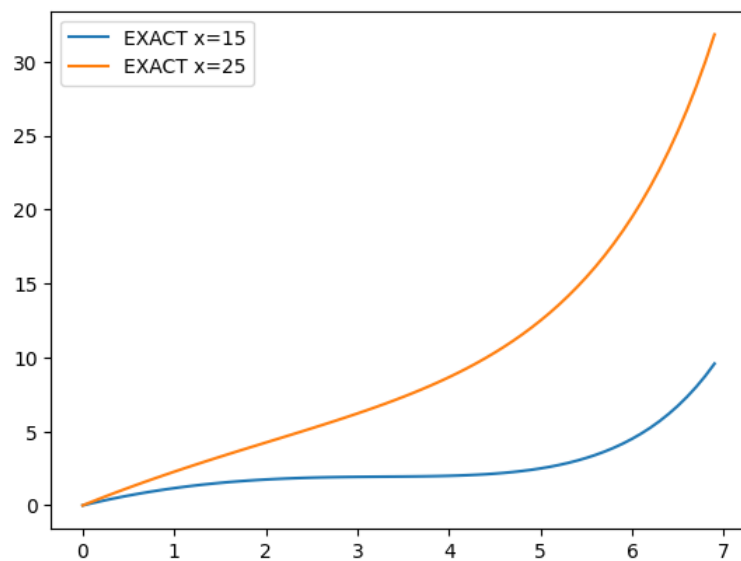
Modified Newton Raphson + BFGS

The solver did not converge.

```
In [11]: full_run( line_search=False, modif_nr=True, line_search_maxit=10000, line_search_fast_s=False, BFGS=True )
```

[FAILED] The solver did not converge after 14 iterations. Giving up.

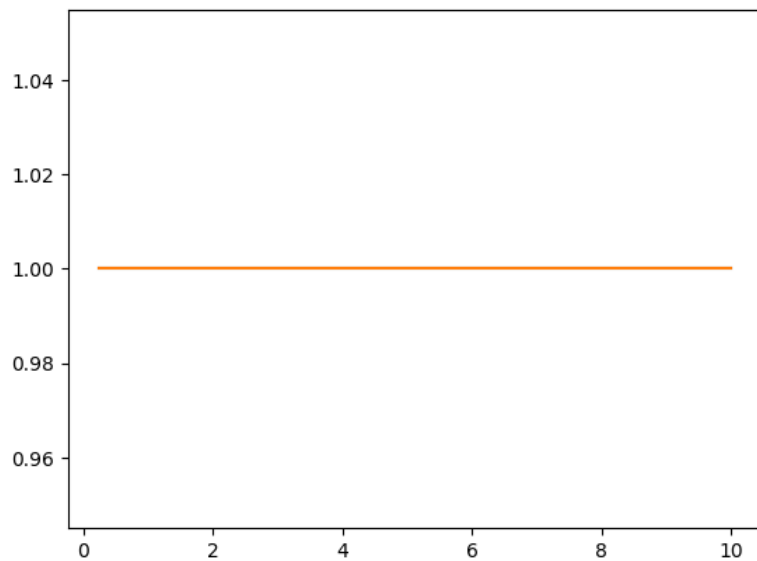
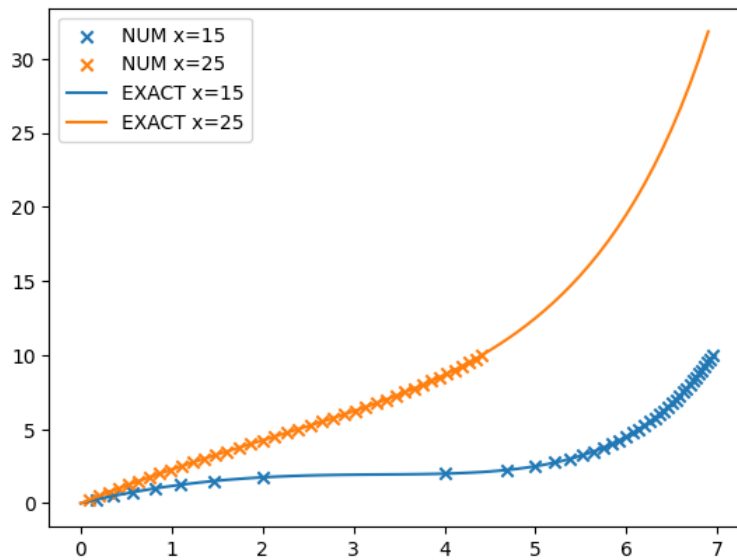
[FAILED] The solver did not converge after 13 iterations. Giving up.



Modified Newton Raphson + BFGS + line search

The solver converged in one iteration (before using the BFGS vectors).

```
In [12]: full_run( line_search=True, modif_nr=True, line_search_maxit=10000, line_search_fast_s=False, BFGS=True )
```



Comments

1. The consistent tangent strategy works in both cases.
2. The modified Newton fails when the tangent changes significantly.
3. The modified Newton works in both cases when associated with line search.
4. The line search works in both cases, and saves time (less Newton cycles).
5. Even limiting the resolution of the search parameter (max 5 iterations in the s solver), the line search is effective
6. BFGS worked only with line search enabled. This suggests this problem is not suitable to assess BFGS capabilities.

In []: