- # Authored by Christopher Allred
- # A02233404
- # Numerical Methods
- # Spring 2020

#### PROBLEM 1.a

Linear algebraic equations can arise in the solution
of differential equations. For example, the following heat
equation describes the equilibrium temperature

$$T = T(x)(_{\circ}C)$$

at a point x (in meters m) along a long thin rod,

$$\frac{d^2T}{dx^2} = h'(T - T_a), \qquad (1)$$

$$\frac{d^2T}{dx^2} = h'(T - T_a),$$

$$d2T/dx2 = h'(T - Ta), (1)$$

where Ta = 10<sub>o</sub>C denotes the temperature of the surrounding air, and h' = 0.03 (m-2) is a heat transfer coefficient. Assume that the rod is 10 meters long (i.e.  $0 \le x \le 10$ ) and has boundary conditions imposed at its ends given by T(0) = 20<sub>o</sub>C and T(10) = 100<sub>o</sub>C.

a) Using standard ODE methods, which you do not need to repeat here, the general form of an analytic solution to (1) can be derived as

$$T(x) = A + Be^{\lambda x} + Ce^{-\lambda x},\tag{2}$$

$$T(x)=A+Be_{\lambda x}+Ce_{-\lambda x}$$
, (2)

where A, B, C, and  $\lambda$  are constants. Plug the solution of type (2) into both sides of equation (1). This should give you an equation that must be satisfied for all values of x, for  $0 \le x \le 10$ , for some fixed constants A, B, C, and  $\lambda$ . Analyze this conclusion to determine

### "NO COMPUTER CODE REQUIRED

$$T := A + B \cdot \exp(\operatorname{lambda} \cdot x) + C \cdot \exp(-\operatorname{lambda} \cdot x)$$

$$T := A + B e^{\lambda x} + C e^{-\lambda x}$$
(1)

$$\frac{\mathrm{d}^2}{\mathrm{d}x^2} T = h\_prime \cdot (T - Ta)$$

$$B \lambda^{2} e^{\lambda x} + C \lambda^{2} e^{-\lambda x} = h\_prime \left( A + B e^{\lambda x} + C e^{-\lambda x} - Ta \right)$$
 (2)

$$\lambda^{2}(B \cdot e^{\lambda x} + C \cdot e^{-\lambda x}) = h\_prime(T - Ta)$$

$$\lambda \left( B e^{\lambda x} + C e^{-\lambda x} \right)^2 = h\_prime \left( A + B e^{\lambda x} + C e^{-\lambda x} - Ta \right)$$
 (3)

$$T-A = B e^{\lambda x} + C e^{-\lambda x}$$

$$B e^{\lambda x} + C e^{-\lambda x} = B e^{\lambda x} + C e^{-\lambda x}$$
 (4)

$$\lambda^2 \cdot (T - A) = h\_prime (T - Ta)$$

$$\lambda^{2} \left( B e^{\lambda x} + C e^{-\lambda x} \right) = h\_prime \left( A + B e^{\lambda x} + C e^{-\lambda x} - Ta \right)$$
 (5)

#therfore

$$\lambda^2 = h_prime$$

$$\lambda^2 = h\_prime \tag{6}$$

$$A = Ta$$

$$A = Ta (7)$$

Found Lambda and A

(b) Next, impose the boundary conditions T (0) = 20 ₀C and T (10) = 100 ₀C to derive a system of 2 linear algebraic equations for B and C. Provide the system of two equations you have derived. ''' NO COMPUTER CODE REQUIRED

$$Ta := 10$$
 (8)

h prime := 0.3

$$h\_prime := 0.3$$
 (9)

Ta + B + C = 20

$$10 + B + C = 20 ag{10}$$

$$B + C = 20 - Ta \# Equation 1$$
  
 $B + C = 10$  (11)

$$Ta + B \cdot \exp(h\_prime \cdot 10) + C \cdot \exp(-h\_prime \cdot 10) = 100$$
  
  $10 + 20.08553692 B + 0.04978706837 C = 100$  (12)

$$\exp(h\_prime \cdot 10) + C \cdot \exp(-h\_prime \cdot 10) = 100 - Ta \# equation 2$$
  
 $20.08553692 + 0.04978706837 C = 90$  (13)

The system of two equations:

$$Mat1 := \begin{bmatrix} 1 & 1 & 10 \\ e^3 & e^{-3} & 90 \end{bmatrix}$$

PROBLEM 1.c

(c) Use one of the numerical algorithms you developed for homework 3 (Gauss elimination or LU decomposition) to solve the algebraic system you derived in question 2(b) above, and obtain an analytic solution to (1) of the form (2). By analytic solution we mean an explicit solution to equation (1) which is valid for each x in the interval [0, 10].

### Gauss elimination

# Authored by Christopher Allred

```
# A02233404
# Numerical Methods
# Spring 2020
import numpy as np
import math
TOLERANCE = 0.0001
def substitute(a, b):
   n = len(a)
   aCopy = a[:]
   x = [None]*(n)
   \# x[n] = b[n]/aCopy[n][n]
   for i in range(n-1,-1,-1):
        # sum = b[i]
       x[i] =aCopy[i][n]/aCopy[i][i]
        for j in range( i-1,-1,-1 ):
            \# sum = sum + aCopy[i][j]*x[j]
            aCopy[j][n] -= aCopy[j][i] *x[i]
   return x
def guss2(a, b):
   print(np.matrix(a))
   print("Solution:...")
   n = len(a)-1
   m = len(a[1])-1
    er = 0
   s = [None] * (n+1)
   for i in range(0,n+1):
        s[i] = abs(a[i][0])
        for j in range( 1,n+1):
            if abs(a[i][j]) > s[i]:
                s[i] = abs(a[i][j])
   #Elimination:
    for k in range(0,n+1):
        #Pivot:
```

```
p = k
    big = abs(a[k][k]/s[k])
    for ii in range(k+1,n+1):
        dummy = abs(a[ii][k]/s[ii])
        if dummy > big:
            big = dummy
            p = ii
    # Pivot
    if p != k:
        for jj in range( k,n+1):
            dummy = a[p][jj]
            a[p][jj] = a[k][jj]
            a[k][jj] = dummy
        dummy = b[p]
        b[p] = b[k]
        b[k] = dummy
        dummy = s[p]
        s[p] = s[k]
        s[k] = dummy
    if abs(a[k][k]/s[k]) < 0:
        er =- 1
        break #EXIT FOR
    for i in range(k+1,n+1):
        factor =- a[i][k]/a[k][k]
        for j in range (k,m+1):
            a[i][j] = a[i][j]+factor*a[k][j]
        b[i] = b[i] + factor*b[k]
if abs(a[n][n]/s[n]) < 0:
    er = -1
print(np.matrix(a))
print("Coefficients: " + str(np.matrix(b)))
# Elimination
if er != -1:
    #Substitute:
    vals = substitute(a,b)
```

```
print("Solution Values: "+ str(vals))
         return vals
if name == " main ":
    A1 = [[1, 1, 10],
         [math.exp(3), math.exp(-3), 90]]
    bConsts1 = [10,90]
    guss2(A1,bConsts1)
Terminal Output
                                                                             + \square
                                                     2: Python Debug Consc >
     PS C:\Users\Christopher\Documents\GitHub\NumMethods\HW\project1> ${env:PTVSD_LAUNCHER_P
     ORT}='54259'; & 'C:\Users\Christopher\AppData\Local\Programs\Python\Python37-32\python.
     exe' 'c:\Users\Christopher\.vscode\extensions\ms-python.python-2020.2.63990\pythonFiles
     \lib\python\new_ptvsd\wheels\ptvsd\launcher' 'c:\Users\Christopher\Documents\GitHub\Num
     Methods\HW\project1\NM_HW3.py'
     [[1.00000000e+00 1.00000000e+00 1.00000000e+01]
      [2.00855369e+01 4.97870684e-02 9.00000000e+01]]
     Solution:...
     [[ 1.
                                     10.
                        1.
                      -20.03574985 -110.85536923]]
     Coefficients: [[ 10.
                                   -110.85536923]]
     Solution Values: [4.467121518528577, 5.532878481471423]
     PS C:\Users\Christopher\Documents\GitHub\NumMethods\HW\project1>
   ⊗ 0 ♠ 0  Python: Current File (project1)
                                                Ln 160, Col 20 Spaces: 4 UTF-8 CRLF Python
                                      \frac{10(-9 + e^{-3})}{-e^{3} + e^{-3}}-\frac{10(e^{3} - 9)}{-e^{3} + e^{-3}}
```

Solution Values: [4.467121518528577, 5.532878481471423]

A = 4.467121518528577B = 5.532878481471423

#### PROBLEM 1.d

111

```
(d) Next we will discuss how to obtain a numerical solution to (1).
That is, we will seek to obtain an approximate solution to (1) which
describes the value of T at 9 intermediate points inside the interval
[0,10]. More precisely, the equation (1) can be transformed into a
linear algebraic system for the temperature at 9 interior points
```

```
T1 = T(1),

T2 = T(2),

T3 = T(3),

T4 = T(4),

T5 = T(5),

T6 = T(6),

T7 = T(7),
```

T8 = T(8),

using the following finite difference approximation for the second derivative at the ith interior point,

$$\frac{d^2T_i}{dx^2} = \frac{T_{i+1} - 2T_i + T_{i-1}}{(\Delta x)^2},\tag{3}$$

$$d2Ti/ dx2 = (Ti+1 -2Ti +Ti-1)/ (\Delta x)**2,$$
 (3)

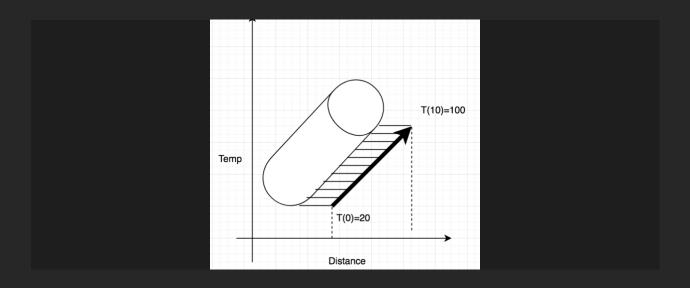
where

and  $\Delta x$  is the equal spacing between consecutive interior points (i.e. with 9 equally spaced interior points inside [0,10] it holds that  $\Delta x$  = 1). Use (3) to rewrite (1)

$$\frac{d^2T}{dx^2} = h'(T - T_a),\tag{1}$$

as a system of 9 linear algebraic equations for the unknowns T1, T2, T3, T4, T5, T6, T7, T8, and T9. Provide the system of 9 equations you have derived.

''' NO COMPUTER CODE REQUIRED



$$T := x \rightarrow A + B \cdot \exp(\operatorname{lambda} \cdot x) + C \cdot \exp(\operatorname{-lambda} \cdot x)$$

$$T := x \rightarrow A + B e^{\lambda x} + C e^{-\lambda x}$$

$$0 = \frac{3}{10} T - 3$$

$$0 = \frac{1}{10} T - 3$$

$$0 = \frac{3}{10} T - 3$$

$$0 = \frac{1}{10} T - 3$$

$$0 = \frac{1}{$$

for 
$$i$$
 from  $1$  by  $1$  while  $i \le 10$ do

$$Tmat[i] := \frac{T(i) - 2T(i-1) + T(i-2)}{(DeltaX^2)} - h\_prime \cdot (T(i-1) - Ta) = 0$$

end do:

$$Tmat[1] := 20 = T(0)$$

$$Tmat_1 := 20 = T(0)$$
 (9)

$$Tmat[11] := 100 = T(10)$$

$$Tmat_{11} := 100 = T(10)$$
 (10)

**Tmat** 

 $\_CMRTS := interface(rtablesize = \infty) : (11); interface(rtablesize = \_CMRTS) :$ 

$$20 = T(0)$$

$$T(2) - \frac{23}{10} T(1) + T(0) + 3 = 0$$

$$T(3) - \frac{23}{10} T(2) + T(1) + 3 = 0$$

$$T(4) - \frac{23}{10} T(3) + T(2) + 3 = 0$$

$$T(5) - \frac{23}{10} T(4) + T(3) + 3 = 0$$

$$T(6) - \frac{23}{10} T(5) + T(4) + 3 = 0$$

$$T(7) - \frac{23}{10} T(6) + T(5) + 3 = 0$$

$$T(8) - \frac{23}{10} T(7) + T(6) + 3 = 0$$

$$T(9) - \frac{23}{10} T(8) + T(7) + 3 = 0$$

$$T(10) - \frac{23}{10} T(9) + T(8) + 3 = 0$$

$$100 = T(10)$$

$$20 = T0$$

$$T2 - \frac{23}{10} TI + T0 + 3 = 0$$

$$T3 - \frac{23}{10} T2 + TI + 3 = 0$$

$$T4 - \frac{23}{10} T3 + T2 + 3 = 0$$

$$T5 - \frac{23}{10} T4 + T3 + 3 = 0$$

$$T6 - \frac{23}{10} T5 + T4 + 3 = 0$$

$$T7 - \frac{23}{10} T6 + T5 + 3 = 0$$

$$T8 - \frac{23}{10} T7 + T6 + 3 = 0$$

$$T9 - \frac{23}{10} T8 + T7 + 3 = 0$$

$$T10 - \frac{23}{10} T9 + T8 + 3 = 0$$

$$100 = T10$$

1	0	0	0	0	0	0	0	0	0	0	20	
1	$-\frac{23}{10}$	1	0	0	0	0	0	0	0	0	-3	
0	1	$-\frac{23}{10}$	1	0	0	0	0	0	0	0	-3	
0	0	1	$-\frac{23}{10}$	1	0	0	0	0	0	0	-3	
0	0	0	1	$-\frac{23}{10}$	1	0	0	0	0	0	-3	
0	0	0	0		$-\frac{23}{10}$		0	0	0	0	-3	
0	0	0	0	0	1	$-\frac{23}{10}$	1	0	0	0	-3	
0	0	0	0	0	0	1	$-\frac{23}{10}$	1	0	0	-3	
0	0	0	0	0	0	0	1	$-\frac{23}{10}$	1	0	-3	
0	0	0	0	0	0	0	0	1	$-\frac{23}{10}$	1	-3	
0	0	0	0	0	0	0	0	0	0	1	100	

...

(e) Use one of the numerical algorithms you developed for homework 3 (Gauss elimination or LU decomposition) to solve the system derived in question 1(d) above. Validate your numerical solution by comparison to the analytic solution that you obtained in 1(c) through depicting the two solutions on plots over the interval 0 ≤ x ≤ 10.

"

Solution Values:

[20.0,

16.27752228627173,

14.43830125842498,

13.930570608105715,

14.602011140218162,

16.654055014396057,

20.702315392892768,

27.961270389257304,

40.608606502399034.

62.43852456626046,

100.01

20.

 $16.27752228627172646416039605456400334424992124519769902458145032657775170877433608752527252618141803\\14.43830125842497086756891092549720769177481886395470775653733575112882893018097300130812681021726147\\13.93057060810570653124809907407957434683216214189812881545442190101855483064190181548341913731828336\\14.60201114021815415430171694488581330593915406241098851900783462121384718029540117430373720561479025\\16.65405501439604802364584989915779625682789220164714477826359772777329368403752088541517643559573422\\20.70231539289275630008373782317711808476499800137744447099844015266472829299089686215116859625539845\\27.96127038925729146654674709414957533813160320152097750503281462335558138984154189753251133579168222\\40.60860650239901407297378049336690519293768936212080379057703348105310890364464950217360747606547065\\62.43852456626044090129294804059430660562508233135687121329436238306656908854115195746678585915890028$ 

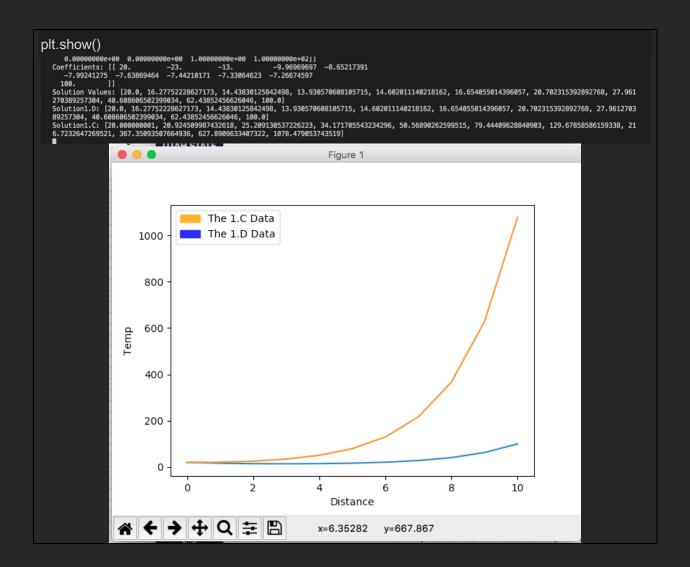
#### PROBLEM 1.e

(e) Use one of the numerical algorithms you developed for homework 3
(Gauss elimination or LU decomposition) to solve the system derived in question 1(d) above. Validate your numerical solution by comparison to the analytic solution that you obtained in 1(c) through depicting

the two solutions on plots over the interval  $0 \le x \le 10$ .

```
""
```

```
import NM_HW3 as hw
consts =[20,-3,-3,-3,-3,-3,-3,-3,-3,100]
print( np.matrix(consts))
solutions = hw.guss2(bigArray,consts)
import math
def Temp(x):
  A=10
  B=4.467121520
  C=5.532878481
  \lambda=math.sqrt(0.3)
  return A+B*math.exp(\lambda^*x) +C*math.exp(-\lambda^*x)
solutions2 = [None]*11
for i in range(0,11):
  solutions2[i] = Temp(i)
import matplotlib.pyplot as plt
import matplotlib.patches as mpatches
\# xVals = range(0,11)
print('Solution1.D:',solutions)
print('Solution1.C:',solutions2)
orange_patch = mpatches.Patch(color='orange', label='The 1.C Data')
blue_patch = mpatches.Patch(color='blue', label='The 1.D Data')
plt.legend(handles=[orange_patch,blue_patch])
plt.plot(solutions)
plt.plot(solutions2)
plt.ylabel('Temp')
plt.xlabel('Distance')
```



It was found that the analytical solution was more accurate in the larger temperature

# PROBLEM 1.f

(f) Write a function that takes as input the number of interior nodes n desired for your numerical solution (i.e. n = 9 in 1(d) above), and outputs the numerical solution to (1) in the form of the interior node values T1 = T( $\Delta x$ ), T2 = T( $2\Delta x$ ),..., Tn = T( $n\Delta x$ ).

```
def n_TempSolution(n,lengthOfRod):
  # deltaX = lengthOfRod/(n+1)
  # m= n+2
  \# solutionMat = [ [[0] * m for i in range(n+2)]]
  deltaX = lengthOfRod/(n+1)
  m= n+2
  solutionMat = [[0] * m for i in range(n+2)]
  consts = [None]*(n+2)
  # Rows
  for i in range(1,n+1):
     # Colloms
     for j in range(0,m):
       if i==j:
          solutionMat[i][j]= (-2/(deltaX**2) -.3)
       elif i+1 == j or i-1==j:
          solutionMat[i][j]=1/(deltaX**2)
       else:
          solutionMat[i][j]= 0
     solutionMat[i][m-1] = (-3)
     consts[i] = (-3)
  solutionMat[0][0] = 1
  solutionMat[0][m-1] = 20
  consts[0] = 20
  solutionMat[n+1][m-2] =1
  solutionMat[n+1][m-1] = 100
  consts[n+1] = 100
  print( np.matrix(solutionMat))
```

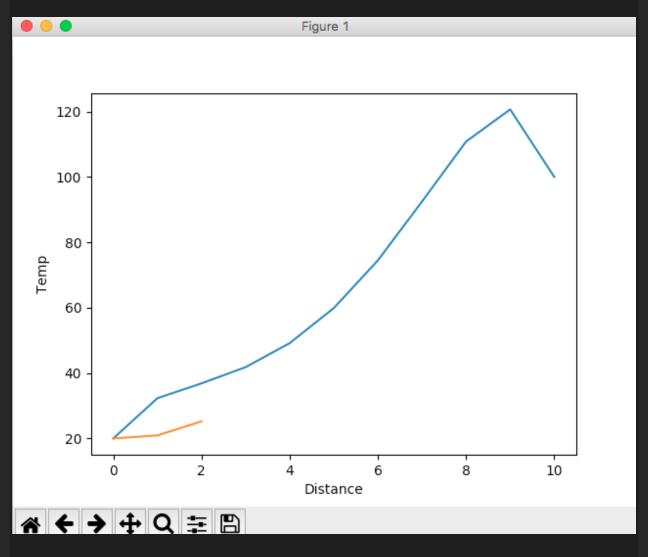
```
print( np.matrix(consts))
   solutions = hw.guss2(bigArray,consts)
   return solutions
lengthOfRod = 10
n_TempSolution(9,lengthOfRod)
                                                                                          1: Python Debug Console V + III iii ^
                                                       3. 100.]]
41.79543301545178, 49.14997991524187, 59.8738624702707, 74.46718676712038, 92.4547570
```

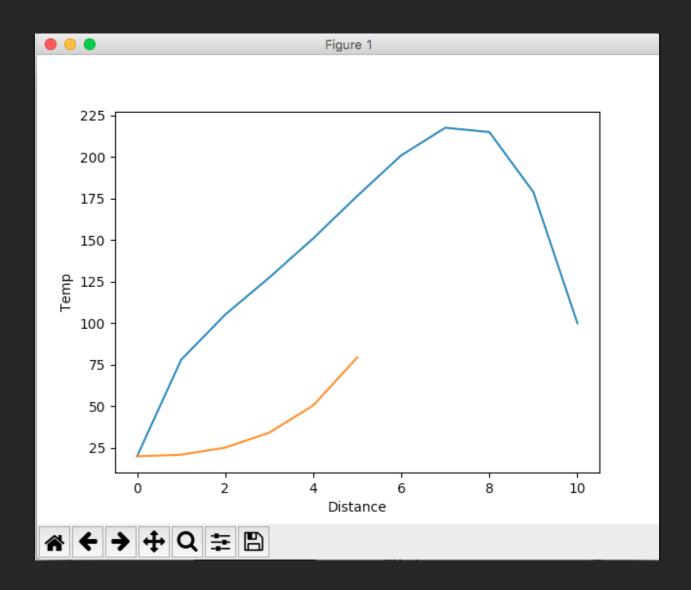
# PROBLEM 1.g

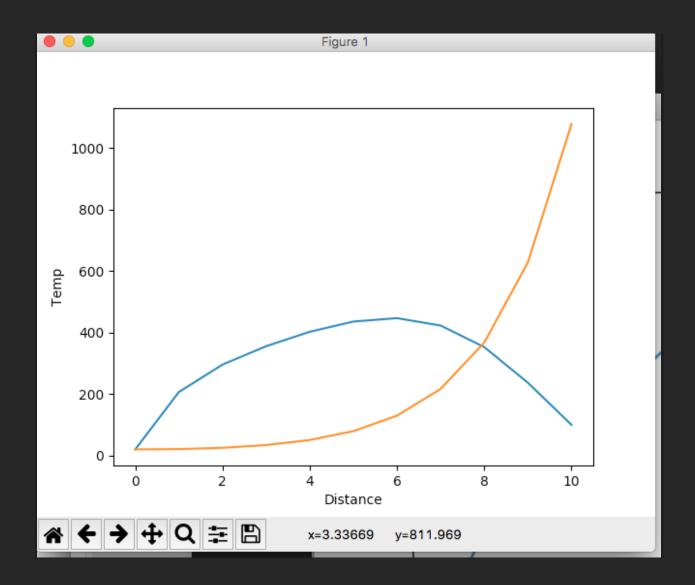
```
(g) Produce and submit 4 plots that compare your analytic solution
  to (1) derived in question 2(b) to the numerical solution generated
  in question 2(f) for n = 1, n = 4, n = 9, and n = 19, respectively.
""
lengthOfRod = 10
val = 1
solutions2 = [None]*(val+2)
for i in range(0,val+2):
  solutions2[i] = Temp(i)
```

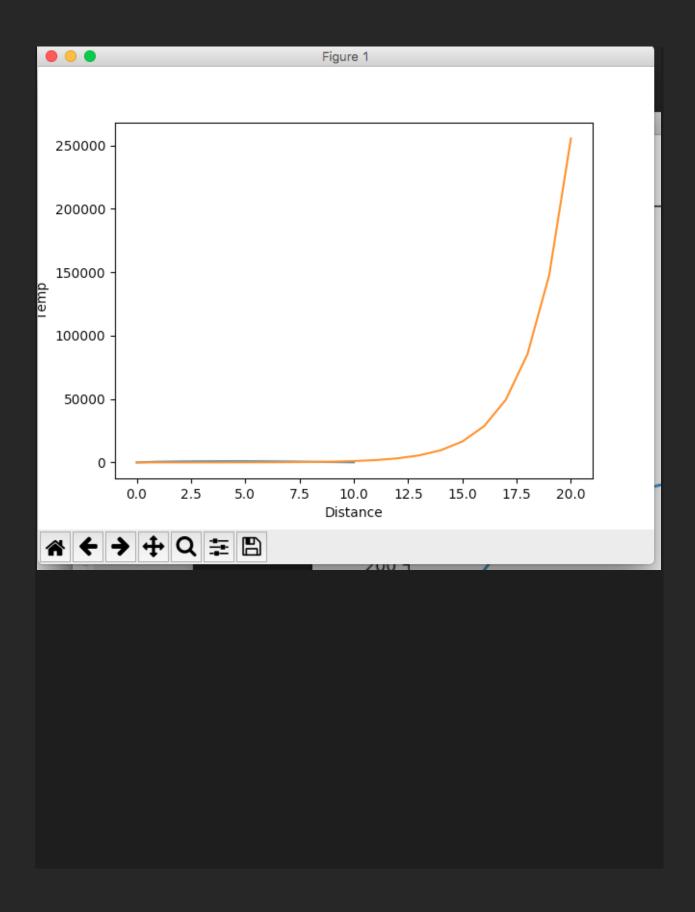
```
solutionG_1=n_TempSolution(1,lengthOfRod)
plt.plot(solutionG_1,)
plt.plot(solutions2)
plt.ylabel('Temp')
plt.xlabel('Distance')
plt.show()
val = 4
solutions2 = [None]*(val+2)
for i in range(0,val+2):
  solutions2[i] = Temp(i)
solutionG_4=n_TempSolution(4,lengthOfRod)
plt.plot(solutionG_4,)
plt.plot(solutions2)
plt.ylabel('Temp')
plt.xlabel('Distance')
plt.show()
val = 9
solutions2 = [None]*(val+2)
for i in range(0,val+2):
  solutions2[i] = Temp(i)
solutionG_9=n_TempSolution(9,lengthOfRod)
plt.plot(solutionG_9,)
plt.plot(solutions2)
plt.ylabel('Temp')
plt.xlabel('Distance')
plt.show()
val = 19
```

```
solutions2 = [None]*(val+2)
for i in range(0,val+2):
    solutions2[i] = Temp(i)
solutionG_19=n_TempSolution(19,lengthOfRod)
plt.plot(solutionG_19,)
plt.plot(solutions2)
plt.ylabel('Temp')
plt.xlabel('Distance')
plt.show()
```









### PROBLEM 2

,,,

2. Develop an algorithm that uses the golden section search to locate the minimum of a given function. Rather than using the iterative stopping criteria we have previously implemented, design the algorithm to begin by determining the number of iterations n required to achieve a desired absolute error |Ea| (not a percentage), where the value for |Ea| is input by the user. You may gain insight by comparing this approach to a discussion regarding the bisection method on page 132 of the textbook. Test your algorithm by applying it to find the minimum of f(x) = 2x+ (6/x) with initial guesses xl = 1 and xu = 5 and desired absolute error |Ea| = 0.00001.

,,,,

$$n = \frac{\log(\Delta x^0 / E_{a,d})}{\log 2} = \log_2\left(\frac{\Delta x^0}{E_{a,d}}\right)$$
(5.5)

```
PROBLEMS
                   TERMINAL
                                                 1: Python Debug Console
      New Lower Test Point = 1.7315417613997808
      f2 < f1
      New Upper Bound = 1.732661895728071
      New Lower Bound = 1.7315417613997808
      New Lower Test Point = 1.731969614641222
      New Upper Test Point = 1.7322340424866296
      New Upper Bound = 1.7322340424866296
      New Lower Bound = 1.7315417613997808
      New Upper Test Point = 1.731969614641222
      New Lower Test Point = 1.7318061892451884
       f2 < f1
      New Upper Bound = 1.7322340424866296
      New Lower Bound = 1.7318061892451884
      New Lower Test Point = 1.731969614641222
      New Upper Test Point = 1.732070617090596
       Final Lower Bound = 1.7318061892451884
       Final Upper Bound = 1.7322340424866296
       Final Local Minima = 1.732020115865909
       bash-3.2$
n 3.7.1 64-bit \otimes 0 \triangle 0 \triangleright Python: Current File (project1)
                                                             Spaces: 4
The Value for the local minima:
1.732020115865909
 Golden Search
 import math
 # goldenSearch
 def goldenSearch(function,xl, xu,tol):
   goldenRatio = 2/(math.sqrt(5) + 1)
   # Use the goldon ratio to set initial points
   x1 = xu - goldenRatio * (xu - xl)
   x2 = xl + goldenRatio * (xu - xl)
   f1 = function(x1)
```

```
f2 = function(x2)
iter = 0
deltaX = abs(xu - xl)
Ead = tol
numIteration = int(math.log2(deltaX/Ead))
# while (abs(xu - xl) > tol):
while( iter <= numIteration):
  iter = iter + 1
  if (f2 > f1):
     xu = x2
     print('New Upper Bound =', xu)
     print('New Lower Bound =', xl)
     # Set the new upper test point
     # Use result of the goldon ratio
     x2 = x1
     print('New Upper Test Point = ', x2)
     f2 = f1
     # Set the new lower test point
     x1 = xu - goldenRatio*(xu - xl)
     print('New Lower Test Point = ', x1)
     f1 = function(x1)
  else:
     print('f2 < f1')
     xI = x1
     print('New Upper Bound =', xu)
```

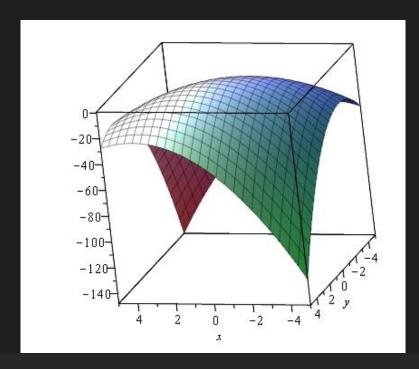
```
print('New Lower Bound =', xl)
     # Set the new lower test point
     x1 = x2
     print('New Lower Test Point = ', x1)
     f1 = f2
     # Set the new upper test point
     x2 = xI + goldenRatio*(xu - xI)
     print('New Upper Test Point = ', x2)
     f2 = function(x2)
print('Final Lower Bound =', xl)
print('Final Upper Bound =', xu)
finalPoint = (xl + xu)/2
print('Final Local Minima =', finalPoint )
return finalPoint
```

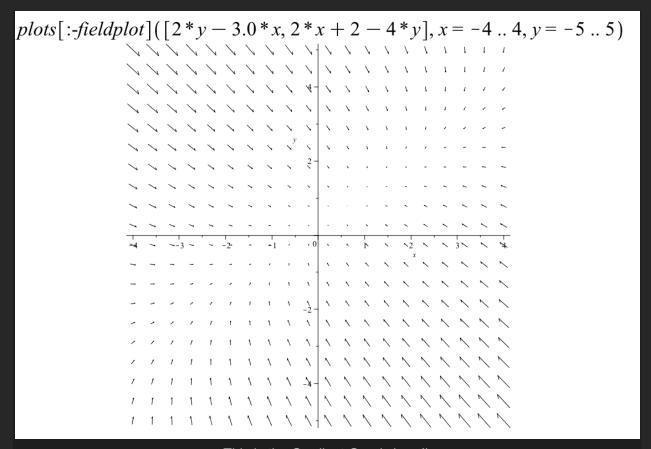
# PROBLEM 3

3. Given f(x,y)=2xy+2y-1.5x^2-2y^2,
(a) Start with an initial guess of (x0,y0) = (1,1) and determine
(by hand is fine) two iterations of the steepest ascent method to maximize f(x,y).

$$fun := 2 \cdot x \cdot y + 2 \cdot y - 1.5 \cdot x^2 - 2 \cdot y^2$$

$$fun := -2.0 \ h + 2 - 1.5 \ (-1.0 \ h + 1)^2$$
(1)





This is the Gradient Graph (woo!)

$$grad := Gradient(fun)$$
  
 $grad := (2y - 3.0x)\overline{e}_x + (2x + 2 - 4y)\overline{e}_y$ 

$$grado := \langle 2 yo - 3.0 xo, 2 xo + 2 - 4 yo \rangle$$

$$grado := \begin{bmatrix} 2 yo - 3.0 xo \\ 2 xo + 2 - 4 yo \end{bmatrix}$$
 (5)

 $point1 := \langle 1, 1 \rangle$ 

$$point1 := \begin{bmatrix} 1 \\ 1 \end{bmatrix}$$
 (6)

xo := 1

$$xo := 1$$
 (7)

yo := 1

$$yo := 1$$
 (8)

elivation := 
$$2 xo \cdot yo + 2 \cdot yo - 1.5 xo^2 - 2 \cdot yo^2$$
  
elivation :=  $0.5$  (9)

# Iteration 1

 $point2 := point1 + h \cdot grado$ 

$$point2 := \begin{bmatrix} -1.0 \ h + 1 \\ 1 \end{bmatrix}$$
 (10)

x := point2[1]

$$x := -1.0 h + 1 \tag{11}$$

y := point2[2]

$$y := 1 \tag{12}$$

fun

$$-2.0 h + 2 - 1.5 (-1.0 h + 1)^2$$
 (13)

isolate( (13), h)

$$h = -0.33333333333 \tag{14}$$

point2

xo := point2[1]

$$xo := 0.6666666667 \tag{17}$$

yo := point2[2]

$$yo := 1$$
 (18)

elivation := 
$$2 xo \cdot yo + 2 \cdot yo - 1.5 xo^2 - 2 \cdot yo^2$$
  
elivation :=  $0.666666666$  (19)

####2nd Iteration

h := 'h'

$$h := h \tag{20}$$

 $point3 := point2 + h \cdot grado$ 

$$point3 := \begin{bmatrix} 1. -1.0 \ h \\ -0.6666666667 \ h + 1 \end{bmatrix}$$
 (21)

x := point3[1]

$$x := 1. - 1.0 h \tag{22}$$

y := point3[2]

$$y := -0.666666667 h + 1 \tag{23}$$

fun

$$(1. -1.0 h) (-0.666666667 h + 1) - 1.3333333334 h + 2 - 1.5 (1.$$
 (24)  
-  $(1.0 h)^2 - 2 (-0.666666667 h + 1)^2$ 

isolate( **(24)**, h )

$$h = -0.3618162035 \tag{25}$$

(15)

$$h := 0.3618162035 \# take the oposit direction$$
  
 $h := 0.3618162035$  (26)

point3

$$xo := point3[1]$$

$$xo := 0.6381837965$$
 (28)

$$yo := point3[2]$$

$$yo := 0.7587891975$$
 (29)

$$yo := 0.7587891975$$
 (29)  
elivation :=  $2 xo \cdot yo + 2 \cdot yo - 1.5 xo^2 - 2 \cdot yo^2$   
elivation :=  $0.723632408$  (30)

(b) What point is the steepest ascent method converging towards? Justify your answer without computing any more iterations.

# " NO COMPUTER CODE REQUIRED

It looks like the function is converging towards the point (0.5, 0.75) which has an elevation of z=0.75, this is clear in the **small increase in the calculated h**. Also, the function has a **smaller difference in** the elevation thus pointing to the converging elevation of 0.75. If you need to more evidence the gradient graph also helps visualize the local maxima.