Linear Optimization

Assignment 1

Sahil Goyal - 2020326

Anshak Goel – 2020283

Deeptorshi Mondal - 2020294

1.

Code for revised Simplex Method from scratch:

```
import numpy as np
def revised_simplex(c,A,b):
    m, n=A. shape
    basic_variables=np.arange(0,m)
    nonbasic_variables=np.arange(m,n)
    initial_soln=np.zeros(n)
    initial_soln[basic_variables]=np.linalg.solve(A[:,basic_variables],b)
    while True:
        A_b=A[:,basic_variables]
        yT=np.linalg.solve(A_b.T,c[basic_variables])
        temp=True
        k=0
        for i in nonbasic_variables:
            if np.dot(yT,A[:,i])>c[i]:
                temp=False
                k=i
                break
        if temp:
            break
        else:
            db=np.linalg.solve(A_b,-A[:,k])
            d=np.zeros(n)
            d[basic_variables]=db
            d[nonbasic_variables]=np.zeros(n-m)
            d[k]=1
            if np.all(d>=0):
                return "unbounded"
            else:
                lambd=np.min(-initial_soln[d<0]/d[d<0])</pre>
                j=np.argmin(-initial_soln[d<0]/d[d<0])</pre>
                initial_soln=initial_soln+lambd*d
                nonbasic_variables=np.append(nonbasic_variables,basic_variable
s[j])
```

```
nonbasic_variables=np.delete(nonbasic_variables,np.where(nonba
sic variables==k))
                nonbasic variables=np.sort(nonbasic variables)
                basic variables[j]=k
                basic variables=np.sort(basic variables)
    print(initial_soln)
    return initial_soln
if name == " main ":
    A=np.array([[6,8,-1,0],[7,12,0,-1]])
    b=np.array([100,120])
    c=np.array([12,20,0,0])
    x=revised_simplex(c,A,b)
    if type(x)==str:
        print("The problem is unbounded")
    else:
        print("The Optimal Solution is: ",np.dot(c,x))
```

The function revised_simplex takes 3 input parameters: the cost vector c, a matrix A which represents the constraints, and a vector b which represents the RHS of the equations. It gives as output the solution vector which can be multiplied with the cost vector in order to get the optimal value.

As an example, we have solved a question from lecture slides:

$$\min 12x_1 + 20x_2$$
s.t.
$$\begin{cases} 6x_1 + 8x_2 \ge 100 \\ 7x_1 + 12x_2 \ge 120 \\ x_1, x_2 \ge 0 \end{cases}$$

Ans: $x_1 = 15, x_2 = 5/4$, Optimal value=205.

```
A=np.array([[6,8,-1,0],[7,12,0,-1]])
b=np.array([100,120])
c=np.array([12,20,0,0])
x=revised_simplex(c,A,b)

v if type(x)==str:
    print("The problem is unbounded")

v else:
    print("The Optimal Solution is: ",np.dot(c,x))
```

Output:

```
[15. 1.25 0. 0. ]
The Optimal Solution is: 205.0000000000000000
```

7	For a general graph for this
	For a general graph for this problem ILP would be
	sext set at a skinning of the
	maximize Exe
	to day a
	ruch that
	Exe 51 for all ytV
	£xe ≤ 1 for all y t V xe € 40,13 for all et f
	1 Providing
	Here Vis set of ventices
	All edges are bringy with
	O representing that they are
	orepresenting that they are not the included and I representing that they are
	representing that they are
	included
	broph:
	G X
	(g) YY (b)
	Y A X
	3 (3)

```
c=np.array([-1,-1,-1,-1,0,0,0,0,0,0,0,0])
b=np.array([1,1,1,1,1,1,1,1])
A=np.array([
        [1,0,0,0,0,0,0,1,0,0,0],
        [1,1,0,1,0,0,0,0,0,1,0,0],
        [0,1,1,0,0,0,0,0,0,0,1,0],
        [0,0,1,1,0,0,0,0,0,0,0],
        [0,1,0,0,0,1,0,0,0,0,0],
        [0,0,1,0,0,0,1,0,0,0,0],
        [0,0,1,0,0,0,1,0,0,0,0]],
        [0,0,0,1,0,0,0,1,0,0,0,0]]
```

For the given adjacency matrix in network 1, we have solved the problem using the simplex code we have written. The output is:

```
[1. 0. 1. 0. 0. 1. 0. 1. 0. 0. 0. 0.]
The Optimal Solution is: 2.0
```

3.

xi + x; ≥ 1 for every crij € E x: E L 0, 1} All vertices (xi) were be with 0 representing that they are not included and prepresenting they are included (graph:

For the given adjacency matrix in network 1, we have solved the problem using the simplex code we have written. The output is:

```
[0.5 0.5 0.5 0.5 0.5 0.5 0.5 0. 0. 0. 0. ] The Optimal Solution is: 2.0
```

4.

We have written code for Interior Point Method from Scratch:

```
import numpy as np
import sympy
def ipopt(c,A,b):
    if np.linalg.matrix_rank(A)<np.min(A.shape):</pre>
        A=A[list(sympy.Matrix(A).T.rref()[1])]
    m, n=A. shape
    primal=np.ones(n,dtype=np.float64)
    dual=np.ones(m,dtype=np.float64)
    slack=np.ones(n,dtype=np.float64)
    flag=(abs(np.dot(primal,slack))>1e-6)
    while flag:
        sig=0.4
        mu=np.dot(primal,slack)/n
        b2=np.zeros(m+2*n,dtype=np.float64)
        b2[:m]=b-np.dot(A,primal)
        b2[m:m+n]=c-np.dot(A.T,dual)-slack
        b2[m+n:m+2*n]=sig*mu*np.ones(n,dtype=np.float64)-
np.dot(np.dot(np.diag(primal),np.diag(slack)),np.ones(n,dtype=np.float64))
        A2=np.zeros((m+2*n,m+2*n),dtype=np.float64)
        A2[:m,:n]=A
        A2[m:m+n,n:m+n]=A.T
        A2[m:m+n,n+m:m+2*n]=np.eye(n,dtype=np.float64)
        A2[m+n:m+2*n,:n]=np.diag(slack)
        A2[m+n:m+2*n,m+n:m+2*n]=np.diag(primal)
        delta=np.linalg.solve(A2,b2)
        dx,dl,ds=delta[:n],delta[n:m+n],delta[n+m:m+2*n]
        alpha max=1
        neg_dx=np.where(dx<0)[0]</pre>
        neg_ds=np.where(ds<0)[0]</pre>
        if len(neg_dx)>0:
            alpha_max=min(alpha_max,np.min(-primal[neg_dx]/dx[neg_dx]))
        if len(neg_ds)>0:
            alpha_max=min(alpha_max,np.min(-slack[neg_ds]/ds[neg_ds]))
        alpha=min(0.99*alpha_max,1)
        primal+=alpha*dx
        dual+=alpha*dl
        slack+=alpha*ds
        flag=(abs(np.dot(primal,slack))>1e-6)
    return primal
```

```
if __name__ == "__main__":
    A=np.array([[6,8,-1,0],[7,12,0,-1]],dtype=np.float64)
    b=np.array([100,120],dtype=np.float64)
    c=np.array([12,20,0,0],dtype=np.float64)
    x=ipopt(c,A,b)
    print(x)
    print("The Optimal Solution is: ",np.dot(c,x))
```

As an example, we have solved a question from lecture slides:

$$\min 12x_1 + 20x_2$$
s.t.
$$\begin{cases} 6x_1 + 8x_2 \ge 100 \\ 7x_1 + 12x_2 \ge 120 \\ x_1, x_2 \ge 0 \end{cases}$$

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```
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b=np.array([100,120],dtype=np.float64)
c=np.array([12,20,0,0],dtype=np.float64)
x=ipopt(c,A,b)
print(x)
print("The Optimal Solution is: ",np.dot(c,x))
```

Output:

```
[1.50000004e+01 1.24999976e+00 6.43515942e-07 1.07252627e-07]
The Optimal Solution is: 205.00000032175794
```

We have written the LPP for the given flow network

Il for network 2: max x, + xy + Xo ruch that: x, + x2 + x3 - x, 2 - x, 4 = 0 x5+x7+x8-x9-x15-X18=0 x + x 10 + x 11 - X8 - X16 - X19 = 0 X12 - X - - X22 20 X14 - X3 - X11 - X = 0 X, 5+ > 16+ ×17 - ×20 - ×22 - ×24 - ×2=0 x12+x19+x20+X21-X17-X29-X29=0 x22 + x3 - x26 - x 30 = 0 * ~ + x ~ 1 + 1 x 26 - X 21 - X 31 = 0 4, Ell, x 53, x, = 12, x 15, x 5 8, x, = 10, x 55, x854, Xq = b, x, = 3, x, <4, x, =18, R X17 = 11 , X19 = 6 , X19 = 3 , X16 = 17 , X17 = 4 , X19 = 6 , X19 = 13 , X19 = 13 , X19 = 15 , X19 = 15 , X19 = 15 , X19 = 15 xi 30 for all i 661,2, 313

We have solved the problem using the IPOPT code we have written. The output is as follows:

```
[1.10000000e+01 2.9999999e+00 5.77062147e-01 7.99999998e+00
4.00000000e+00 9.9999999e+00 4.9999999e+00 7.60620746e-01
5.16637612e+00 1.75998388e+00 6.07097689e-01 1.10945224e+01
8.00000000e+00 3.48253975e+00 2.26546903e+00 5.72429862e+00
2.99916057e+00 8.32877558e+00 1.04853832e+00 1.00083943e+00
4.93910506e-01 9.33453851e+00 4.63588794e+00 3.02491004e+00
4.37087364e+00 5.82785181e+00 2.29837991e+00 2.32729080e+00
 3.50202963e+00 8.14257464e+00 1.27297250e+01 8.97657061e-09
8.97657063e-09 1.14229379e+01 7.00000002e+00 4.00000000e+00
8.97657053e-09 8.97657054e-09 3.23937925e+00 8.33623878e-01
1.24001612e+00 3.39290231e+00 6.90547762e+00 8.00000000e+00
5.17460253e-01 7.34530971e-01 1.12757014e+01 1.00083943e+00
 2.67122442e+00 4.95146168e+00 2.99916057e+00 1.50608949e+00
3.66546149e+00 4.36411206e+00 9.75089958e-01 6.29126363e-01
1.17214819e+00 1.87016201e+01 6.72709199e-01 4.97970370e-01
8.57425361e-01 2.27027502e+00]
The Optimal Solution is: 28.999999964093718
```

We have written the Dual of the LPP for the given flow network

LPP to the dual LPP by iplying is with the period 04 x16417)

Yxy OE Xzy & 4 425 (0 = ×25 = 5) (0 c x 6 = 7) 427 (04 x17621) 428 (0 = x28 = 3) 428 (0 = x28 = 3) 429 (0 = x29 = 4) 430 (0 = x30 = 9) 06 X3, 515 432 (X, +x2+ x3 - 7, 12 - x14)=0 433 (xy-x2-x7)=0 Yzy (x,+x,+x,-x,-x,5-x,8)=0 439 (xy+x10+x11-x2-x16-x19)=0 (X12-X10-X22)-0 437 (X14-X3-X11-X27)=0 438 x19+ ×16+ ×17 - ×10-×23-×24-×29=0 4391 ×18 + ×19 + ×10 + ×11 - ×17 - ×19)=0 440 X22 + X22 - X26 - X30)=0 yyı (xzy + xzy+ xz6 - xz, -x3,)=0

real form of LPP min: 11 y, +3 y, 2 + 12 y, 4 15 y, 4 8 y + 5 y, 5 + 4 y, 6 + 6 y, 9 + 3 y, 6 + 4 y, 7 + 1 + 16 y 13 + 4 y, 9 + 3 y, 5 + 17 y, 16 + 4 y, 7 + 1 + 4 y, 0 + 2 y, 1 + 13 y, 2 + 4 y, 7 + 4 + 5 y, 5 + 7 y, 6 + 21 y, 7 + 3 y, 8 + 4 + 9 y, 0 + 15 y, 1 + 432 00 - 413 = 0 + 4347 411-437+43,20 416 - 44 5+ 42820

Y17 +418 - 429 20 Y18 - Y34 + 439 20 419-435 + 439 30 420 - 438 + 439 7 0 421 + 439 0 - 441 20 422-436+44030 423 - 438 + 440 70 924 - 428 + 441 = 0 425 - 439 + 44120 126 - ynd + yn, 70 427 - 437 20 728 - 43820 y L9 - 43970 420 - 44020 yz, - 94,20 y 20 for all i th yeer for all vetzz. We have solved the problem using the IPOPT code we have written. The output is as follows:

```
[9.99999986e-01 9.99999988e-01 1.06485393e-09 1.73790465e-09
 3.04133280e-09 9.99999977e-01 9.99999978e-01 3.75080501e-09
 1.45828538e-08 9.75275038e-09 3.58562266e-09 1.76110165e-09
 1.52066636e-09 2.35501611e-08 1.67185302e-08 1.07855052e-09
 1.21148800e-08 4.54881940e-09 2.45563813e-09 4.06077923e-09
 8.07556373e-09 3.32342894e-09 2.78563084e-09 1.24431587e-08
 1.93613008e-08 1.03953197e-08 6.50573432e-10 1.79372791e-08
 2.44104928e-08 1.41686711e-08 5.36544017e-09 2.19887428e+06
 2.19887428e+06 2.19887433e+06 2.19887333e+06 2.19885193e+06
 2.19885193e+06 2.19881280e+06 2.19881280e+06 2.19891275e+06
 2.19891275e+06 2.19888741e+06 2.19888741e+06 2.19876228e+06
 2.19876228e+06 2.19883363e+06 2.19883363e+06 2.19889273e+06
 2.19889273e+06 2.19890445e+06 2.19890445e+06 1.10593911e-09
 4.05510956e-09 2.11353054e-08 1.52066615e-09 3.04133280e-09
 1.21653300e-09 2.43306593e-09 1.60781953e-08 2.35608019e-09
 6.94087262e-09 2.00353560e-08 1.09674581e-09 1.52066636e-09
 3.49238902e-09 5.36478162e-09 2.12655640e-09 4.06072236e-09
 1.46119343e-09 1.16306882e-08 1.21143711e-08 2.46480073e-08
 1.30256307e-09 2.62589374e-09 4.02158708e-09 2.78288974e-09
 2.08677345e-09 5.28786986e-09 5.23887337e-09 3.47381983e-09
 1.49402972e-09 9.55443628e-10]
The Optimal Solution is: 29.000000705402222
```

5.

The following results are obtained when we solve the LPP's we made in the previous questions.

For Question 2:

Code:

```
reset;
var x{1..4} >= 0 binary;
maximize z: x[1]+x[2]+x[3]+x[4];
s.t. c1: x[1]<=1;
s.t. c2: x[1]+x[2]+x[4]<=1;
s.t. c3: x[3]+x[2]<=1;
s.t. c4: x[3]+x[4]<=1;
option solver cplex;
solve;
display x,z;</pre>
```

Output:

```
x [*] :=
2 0
   1
3
4 0
z = 2
For Question 3:
Code:
reset;
var x{1..4} >= 0 binary;
minimize z: x[1]+x[2]+x[3]+x[4];
s.t. c1: x[1]+x[2]>=1;
s.t. c2: x[2]+x[3]>=1;
s.t. c3: x[2]+x[4]>=1;
s.t. c4: x[3]+x[4]>=1;
option solver cplex;
solve;
display x, z;
Output:
 x [*] :=
 1
   1
 2
 3
   1
 4
   0
 z = 2
For Question 4:
Max-Flow:
Code:
reset;
var x\{1...31\} >= 0;
maximize z: x[1] + x[4] + x[6];
s.t. c1: x[1]+x[2]+x[3]-x[12]-x[14]=0;
s.t. c2: x[4]-x[2]-x[7]=0;
s.t. c3: x[6]+x[7]+x[8]-x[9]-x[15]-x[18]=0;
s.t. c4: x[9]+x[10]+x[11]-x[8]-x[16]-x[19]=0;
s.t. c5: x[12]-x[10]-x[22]=0;
s.t. c6: x[14]-x[3]-x[11]-x[27]=0;
s.t. c7: x[15]+x[16]+x[17]-x[20]-x[23]-x[24]-x[28]=0;
```

```
s.t. c8: x[18]+x[19]+x[20]+x[21]-x[17]-x[25]-x[29]=0;
s.t. c9: x[22]+x[23]-x[26]-x[30]=0;
s.t. c10: x[24]+x[25]+x[26]-x[21]-x[31]=0;
s.t. c11: x[1]<=11;
s.t. c12: x[2]<=3;
s.t. c13: x[3]<=12;
s.t. c14: x[4]<=15;
s.t. c15: x[5] \le 8;
s.t. c16: x[6]<=10;
s.t. c17: x[7] <=5;
s.t. c18: x[8]<=4;
s.t. c19: x[9]<=6;
s.t. c20: x[10]<=3;
s.t. c21: x[11]<=4;
s.t. c22: x[12]<=18;
s.t. c23: x[13]<=16;
s.t. c24: x[14]<=4;
s.t. c25: x[15]<=3;
s.t. c26: x[16]<=17;
s.t. c27: x[17]<=4;
s.t. c28: x[18]<=11;
s.t. c29: x[19]<=6;
s.t. c30: x[20]<=4;
s.t. c31: x[21]<=2;
s.t. c32: x[22]<=13;
s.t. c33: x[23]<=9;
s.t. c34: x[24] \le 4;
s.t. c35: x[25]<=5;
s.t. c36: x[26]<=7;
s.t. c37: x[27] <= 21;
s.t. c38: x[28] <= 3;
s.t. c39: x[29] \le 4;
s.t. c40: x[30] \le 9;
s.t. c41: x[31]<=15;
option solver cplex;
solve;
display x, z;
We formulated the LPP of flow network and have displayed the solution below.
x [*] :=
         5 0
                 9 4
                         13 0
                                 17 2
                                          21 0
                                                   25
                                                      5
 1 11
         6 10
                 10 1
                         14 4
                                 18 11
                                          22 9
                                                   26 0
 2 3
         7 5
                 11 0
                         15
                             0
                                 19 0
                                          23 0
 3 0
                                                   27
                                                      4
         8 0
                12 10
                         16 5
                                  20
                                     0
                                          24
                                              4
                                                      3
 4 8
                                                   28
z = 29
Min-Cut:
Code:
reset;
var y\{1...31\} >= 0;
var yu{1..10};
```

```
minimize z:
11*y[1] + 3*y[2] + 12*y[3] + 15*y[4] + 8*y[5] + 10*y[6] + 5*y[7] + 4*y[8] + 6*y[9] + 3*y[10] + 10*y[6] +
4*y[11]+18*y[12]+16*y[13]+4*y[14]+3*y[15]+17*y[16]+4*y[17]+11*y[18]+6*y[19]
+4*y[20]+2*y[21]+13*y[22]+9*y[23]+4*y[24]+5*y[25]+7*y[26]+21*y[27]+3*y[28]+
4*y[29]+9*y[30]+15*y[31];
s.t. c1: y[1]+yu[1]>=1;
s.t. c2: y[2]+yu[1]-yu[2]>=0;
s.t. c3: y[3]+yu[1]-yu[6]>=0;
s.t. c4: y[4]+yu[2]>=1;
s.t. c5: y[5] >= 0;
s.t. c6: y[6]+yu[3]>=1;
s.t. c7: y[7]-yu[2]+yu[3]>=0;
s.t. c8: y[8]+yu[3]-yu[4]>=0;
s.t. c9: y[9]-yu[3]+yu[4]>=0;
s.t. c10: y[10]+yu[4]-yu[5]>=0;
s.t. c11: y[11]+yu[4]-yu[6]>=0;
s.t. c12: y[12]-yu[1]+yu[5]>=0;
s.t. c13: y[13]>=0;
s.t. c14: y[14]-yu[1]+yu[6]>=0;
s.t. c15: y[15]-yu[3]+yu[7]>=0;
s.t. c16: y[16] - yu[4] + yu[7] >= 0;
s.t. c17: y[17]+yu[7]-yu[8]>=0;
s.t. c18: y[18] - yu[3] + yu[8] >= 0;
s.t. c19: y[19]-yu[4]+yu[8]>=0;
s.t. c20: y[20]-yu[7]+yu[8]>=0;
s.t. c21: y[21]+yu[8]-yu[10]>=0;
s.t. c22: y[22]-yu[5]+yu[9]>=0;
s.t. c23: y[23]-yu[7]+yu[9]>=0;
s.t. c24: y[24]-yu[7]+yu[10]>=0;
s.t. c25: y[25]-yu[8]+yu[10]>=0;
s.t. c26: y[26]-yu[9]+yu[10]>=0;
s.t. c27: y[27]-yu[6]>=0;
s.t. c28: y[28]-yu[7]>=0;
s.t. c29: y[29]-yu[8]>=0;
s.t. c30: y[30]-yu[9]>=0;
s.t. c31: y[31]-yu[10]>=0;
option solver cplex;
solve;
display y, yu, z;
```

We formulated the dual of the LPP of the network and displayed the output below.

```
:
1
2
3
          yu
                 :=
      У
      1
           0
      1
           1
      0
           0
4
      0
           0
5
      0
           0
6
      1
           0
7
      1
           0
8
      0
           0
9
      0
           0
10
      0
           0
11
      0
12
      0
13
      0
14
      0
15
      0
16
      0
17
      0
18
      0
19
      0
20
      0
21
      0
22
      0
23
      0
24
      0
25
      0
26
      0
27
      0
28
      0
29
      0
30
      0
31
 z = 29
```

As we can see, the optimal value in both the primal and dual form is the same. Therefore, we can see that for the given flow network min-cut = \max -flow = 29