

# Teme Tehnici de Optimizare - Miruna-Andreea Zavelca

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## 1 TEMA 1

De dat exemple și de reprezentat grafic cele 4 cazuri de compatibilitate în care se poate afla un program liniar

```
[1]: import numpy as np
import matplotlib.pyplot as plt
```

```
[2]: def adaugaLaGrafic(x, y1, y2):
    plt.plot(x, y1)
    plt.fill_between(x, y1, y2, alpha=0.2)
```

### 1.1 Program liniar compatibil cu optim finit cu solutie optima unica:

Consider urmatorul sistem:

$$\begin{aligned} \max \quad & x + y \\ \text{s.t.} \quad & x - y \leq 2 \\ & 2x + y \leq 1 \end{aligned}$$

Calculez coordonatele punctului de optim:

$$\begin{aligned} y &= x - 2 \\ y &= 1 - 2 * x \\ \Rightarrow x - 2 &= 1 - 2 * x \\ \Rightarrow 3 * x &= 3 \\ \Rightarrow x &= 1 \\ \Rightarrow y &= -1 \end{aligned}$$

```
[3]: def programFinitSolutieUnica():
    plt.figure()
    plt.xlim(-5, 5)
    plt.ylim(-5, 5)
    x = np.linspace(-5, 5, 100)

    # x - y <= 2
    y = x - 2
    adaugaLaGrafic(x, y, y - 20)

    # 2x + y <= 1
```

```

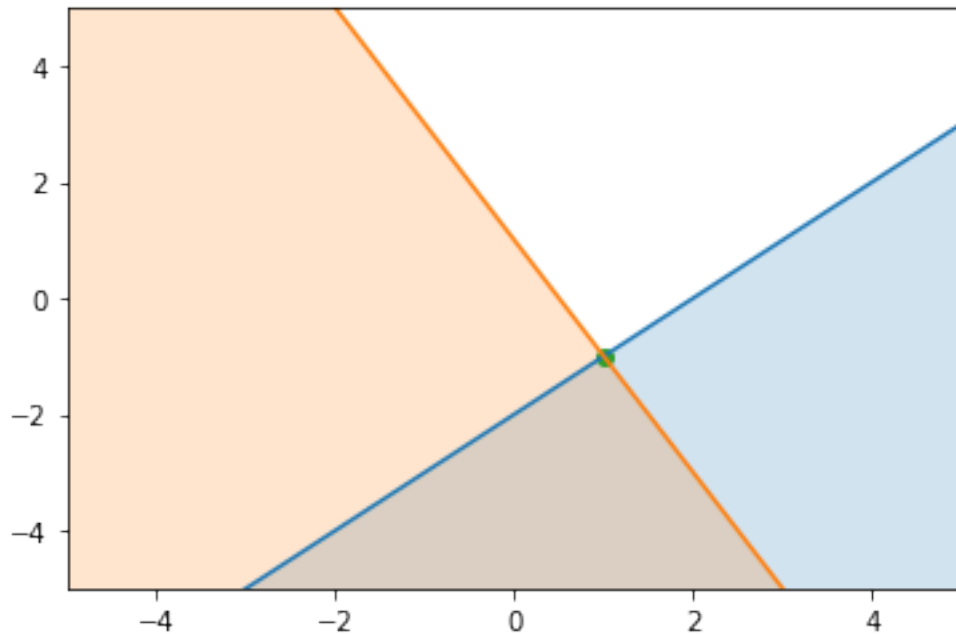
y = 1 - 2 * x
adaugaLaGrafic(x, y, y - 20)

# punctul de optim
plt.scatter(1, -1)

plt.show()

programFinitSolutieUnica()

```



## 1.2 Program linear compatibil cu optim finit cu o infinitate de solutii optime:

Consider urmatorul sistem:

$$\begin{aligned} \max \quad & x + y \\ \text{s.t.} \quad & y \leq -1 \end{aligned}$$

```

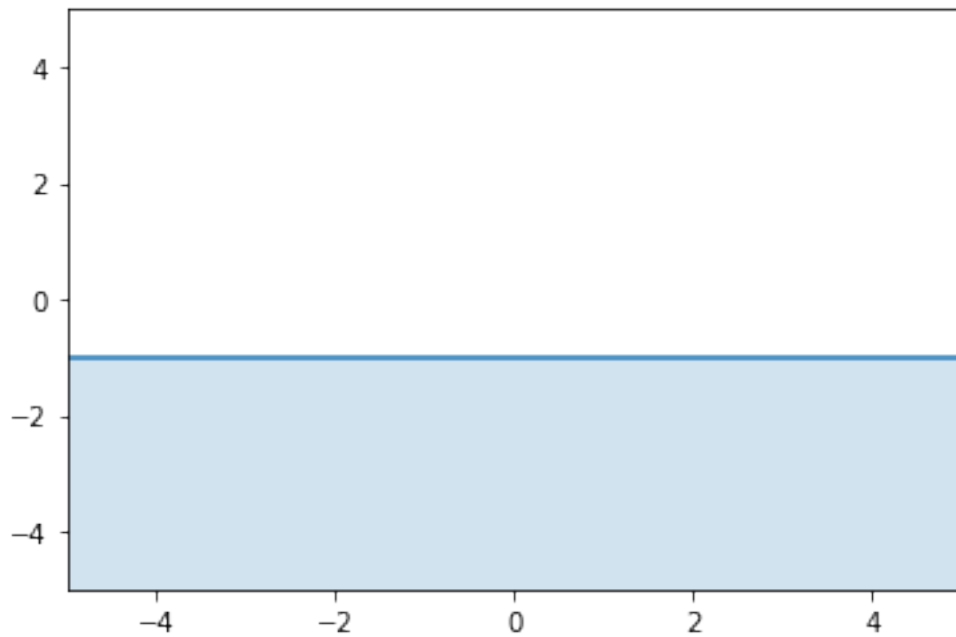
[4]: def programFinitInfinitateSolutii():
    plt.figure()
    plt.xlim(-5, 5)
    plt.ylim(-5, 5)
    x = np.linspace(-5, 5, 100)

    # y ≤ -1
    y = -x * 0 - 1
    adaugaLaGrafic(x, y, y - 10)

```

```
plt.show()

programFinitInfinitateSolutii()
```



### 1.3 Program liniar compatibil cu optim infinit:

Consider urmatorul sistem:

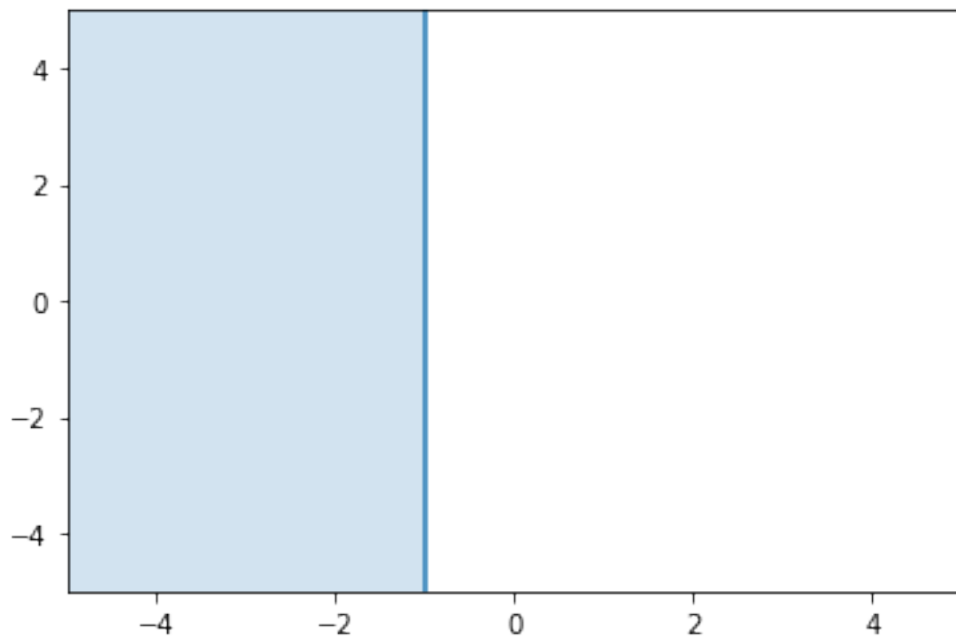
$$\begin{aligned} \max \quad & x + y \\ \text{s.t.} \quad & x \leq -1 \end{aligned}$$

```
[5]: def programOptimInfinit():
    plt.figure()
    plt.xlim(-5, 5)
    plt.ylim(-5, 5)
    x = np.linspace(-5, 5, 100)

    #  $x \leq -1$ 
    plt.plot(0 * x - 1, x)
    plt.fill_between(x, -1, -10, alpha=0.2)

    plt.show()

programOptimInfinit()
```



#### 1.4 Program liniar incompatibil:

Consider urmatorul sistem:

$$\begin{aligned} \max & x + y \\ -4x + 11y & \geq 3 \\ 2x + y & \geq 7 \\ y & \leq -1 \end{aligned}$$

```
[6]: def programIncompatibil():
    plt.figure()
    plt.xlim(-5, 5)
    plt.ylim(-5, 5)
    x = np.linspace(-5, 5, 100)

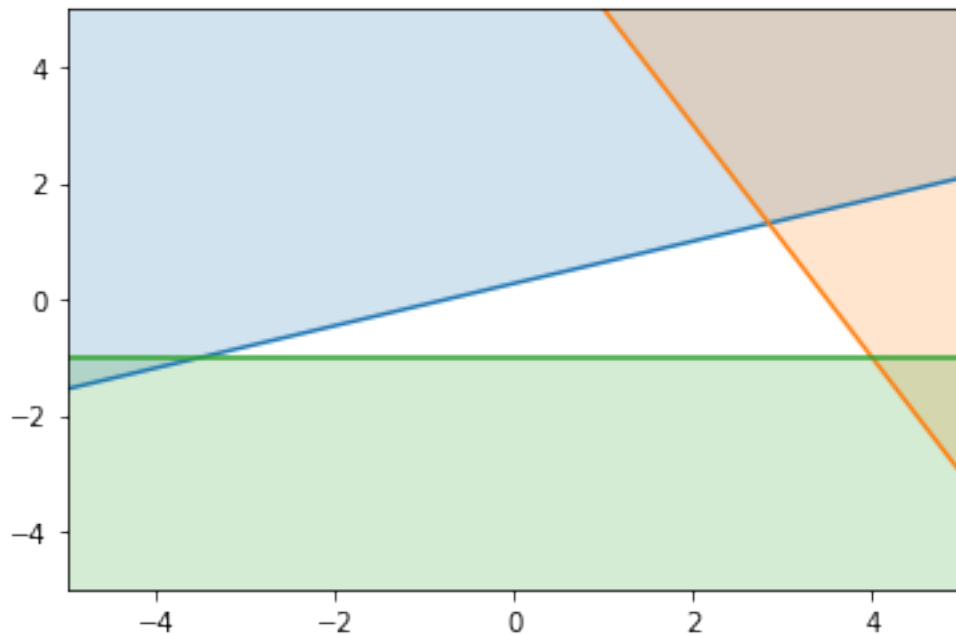
    # -4x + 11y >= 3
    y = (3 + 4 * x) / 11
    adaugaLaGrafic(x, y, y + 10)

    # 2x + y >= 7
    y = 7 - 2 * x
    adaugaLaGrafic(x, y, y + 10)

    # y <= -1
    y = -x * 0 - 1
    adaugaLaGrafic(x, y, y - 10)
```

```
plt.show()

programIncompatibil()
```



## 2 TEMA 2

1. De calculat inversa unei matrici
2. De aplicat lema substituției

Calculeaza  $x$  astfel incat  $Ax = B$  folosind metoda substitutiei descendente:

```
[7]: def substitutieDescendenta(a, b):
    assert a.shape[0] == a.shape[1], 'Matricea introdusa nu este patratica!'
    assert a.shape[0] == b.shape[0], 'Matricea introdusa si vectorul b nu se_
    ↪potrivesc!'

    n = b.shape[0] - 1
    x = np.zeros(shape=n+1)
    x[n] = b[n] / a[n, n]

    for k in range(n - 1, -1, -1):
        s = np.dot(a[k, k + 1:], x[k + 1:])
        x[k] = (b[k] - s) / a[k, k]

    return x
```

Calculeaza x astfel incat  $Ax = B$  folosind metoda Gauss cu pivotare totala:

```
[8]: def gaussPivotareTotala(a, b):
    assert a.shape[0] == a.shape[1], 'Matricea introdusa nu este patratica!'
    assert a.shape[0] == b.shape[0], 'Matricea introdusa si vectorul b nu se
    ↪potrivesc!'
    assert np.linalg.det(a) != 0, 'Sistem incompatibil sau sistem compatibil
    ↪nedeterminat!'

    a_ext = np.concatenate((a, b[:, None]), axis=1)
    n = a.shape[0] - 1
    index = np.array(range(n + 1))

    for k in range(n):
        poz = np.argmax(np.abs(a_ext[k:, k:n]))
        p = k + int(poz / (n - k))
        m = k + poz - ((p - k) * (n - k))

        if a_ext[p, m] == 0:
            raise AssertionError('Sistem incompatibil sau sistem compatibil
            ↪nedeterminat.')

        if p != k:
            a_ext[[p, k], :] = a_ext[[k, p], :]

        if m != k:
            a_ext[:, [m, k]] = a_ext[:, [k, m]]
            index[[m, k]] = index[[k, m]]

        for j in range(k + 1, n + 1):
            mjk = a_ext[j, k] / a_ext[k, k]
            a_ext[j, :] -= mjk * a_ext[k, :]

    if a_ext[n, n] == 0:
        raise AssertionError('Sistem incompatibil sau sistem compatibil
        ↪nedeterminat.')

    y = substitutieDescendenta(a_ext[:, :-1], a_ext[:, -1])

    x = list(range(n + 1))
    for i in range(n + 1):
        x[index[i]] = y[i]

    return np.array(x)
```

## 2.1 Calculeaza inversa matricei A folosind metoda Gauss pentru determinarea inversei:

```
[9]: def calculeazaInversa(a):  
    assert a.shape[0] == a.shape[1], 'Matricea introdusa nu este patratica!'  
    assert np.linalg.det(a) != 0, 'Matricea introdusa nu este inversabila!'  
  
    inversa = np.zeros_like(a)  
    n = a.shape[0]  
    ident = np.eye(n)  
  
    for i in range(n):  
        inversa[:, i] = gaussPivotareTotala(a, ident[:, i])  
  
    return inversa
```

```
[10]: a = np.array([  
        [2, 0, 0],  
        [0, 1, 0],  
        [0, 0, 4]  
    ], dtype=np.float)  
print('Matricea initiala:\n', a, '\n')  
  
b = calculeazaInversa(a)  
print('Inversa matricei:\n', b, '\n')
```

Matricea initiala:

```
[[2. 0. 0.]  
 [0. 1. 0.]  
 [0. 0. 4.]
```

Inversa matricei:

```
[[0.5  0.   0.  ]  
 [0.   1.   0.  ]  
 [0.   0.   0.25]]
```

## 2.2 Aplica lema substitutiei pentru matricea B si vectorul C:

```
[11]: def lemaSubstitutiei(B, C, k=1):  
    assert k in range(B.shape[0]), 'k nu apartine intervalului [1, m] !'  
    assert C.shape[0] == B.shape[0], 'C nu apartine R^m !'  
  
    B_inv = calculeazaInversa(B)  
  
    B_tilde = B.copy()  
    B_tilde[:, k] = C
```

```

y = B_inv @ C
if y[k] == 0:
    raise AssertionError('B nu e nesingulara!')

eta = -y / y[k]
eta[k] = 1 / y[k]

E = np.eye(B.shape[0])
E[:, k] = eta

B_tilde_inv = E @ B_inv

return B_tilde, B_tilde_inv

```

```

[12]: a = np.array([
        [2, 0, 0],
        [0, 1, 0],
        [0, 0, 4]
    ], dtype=np.float)
print('Matricea initiala:\n', a, '\n')

c, d = lemaSubstitutiei(a, np.array([1, 5, 2]))
print('B tilde:\n', c, '\n')
print('Inversa lui B tilde:\n', d, '\n')

```

Matricea initiala:

```

[[2. 0. 0.]
 [0. 1. 0.]
 [0. 0. 4.]]

```

B tilde:

```

[[2. 1. 0.]
 [0. 5. 0.]
 [0. 2. 4.]]

```

Inversa lui B tilde:

```

[[ 0.5 -0.1  0. ]
 [ 0.   0.2  0. ]
 [ 0.  -0.1  0.25]]

```

### 3 TEMA 3 & TEMA 4

De implementat algoritmul simplex în varianta primală, respectiv duală.

Functii ajutatoare:



```
[13]: def pivot(matrice, linie_pivot, coloana_pivot):
    """ Executa un pas din reducerea Gaussiana pentru pivotul dat """
    matrice[linie_pivot] /= matrice[linie_pivot, coloana_pivot]

    for linie in range(len(matrice)):
        if linie != linie_pivot:
            matrice[linie] -= matrice[linie, coloana_pivot] *
↪matrice[linie_pivot]

def strCoeficienti(sir):
    """ Transcrie un sir de coeficienti ca formula """
    s = f'{sir[0]} * x1 '
    for i in range(1, len(sir)):
        if sir[i] >= 0:
            s += f'+ {sir[i]}'
        else:
            s += f'- {-1 * sir[i]}'
        s += f' * x{i + 1} '
    return s
```

Clasa cu rezolvarea problemei:

```
[14]: class Sistem:
    def __init__(self, matrice, semne, coeficienti, maxim=True):
        assert matrice.shape[0] == semne.shape[0] and matrice.shape[1] - 1 ==
↪coeficienti.shape[0], \
            'Dimensiunile datelor de intrare nu se potrivesc!'

        self.matrice = matrice                # matricea sistemului
        self.semne = semne                    # semnele (in)ecuatilor
        self.coeficienti = coeficienti        # coeficientii din max
        self.variabale_libere = 0              # variabilele adaugate pentru
↪aducerea la forma LP standard
        self.tip = 'max' if maxim else 'min'  # daca programul calculeaza
↪maximul / minimul

    def __str__(self):
        s = f'{self.tip} {strCoeficienti(self.coeficienti)}\n'
        for i in range(self.matrice.shape[0]):
            s += f'{strCoeficienti(self.matrice[i][: -1])} {self.semne[i]} {self.
↪matrice[i][ -1]}\n'
        return s

    def adaugaVariabila(self, coeficient, index):
        """ Adauga o variabila ajutatoare in matrice """
        termeni_liberi = self.matrice[:, -1]
```

```

        aux = np.array([coeficient if i == index else 0 for i in range(self.
↪matrice.shape[0])], dtype=float)
        self.matrice = np.concatenate((self.matrice[:, :-1], aux[:, None]),
↪axis=1)
        self.matrice = np.concatenate((self.matrice, termeni_liberi[:, None]),
↪axis=1)

    def adaugaTermeniLiberi(self, semn=-1):
        """ Aduaga termenii liberi in matricea sistemului """
        termeni_liberi = np.array([semn * c for c in self.coeficienti] + [0] *
↪(self.variabile_libere + 1), dtype=float)
        self.matrice = np.concatenate((self.matrice, termeni_liberi[None, :]))
        self.semne = np.append(self.semne, ['='])

    def extrageTermeniLiberi(self):
        """ Extrage termenii liberi din matricea sistemului si ii muta in
↪vectorul de coeficienti """
        self.coeficienti = self.matrice[-1][:-1]
        self.matrice = self.matrice[:-1]

    def standardizareLP(self):
        """ Aduce problema in forma LP standard """
        for i in range(len(self.semne)):
            if self.semne[i] == '=':
                continue
            elif self.semne[i] == '<=':
                self.adaugaVariabila(1, i)
            elif self.semne[i] == '>=':
                self.adaugaVariabila(-1, i)

            self.semne[i] = '='
            self.variabile_libere += 1

        self.adaugaTermeniLiberi()

    def simplexPrimal(self):
        """ Tema 3: Algoritmul simplex in varianta primala """
        self.standardizareLP()

        while True:
            coloana_pivot = -1
            for coloana, valoare in enumerate(self.matrice[-1]):
                if valoare < 0:
                    coloana_pivot = coloana
                    break

            if coloana_pivot == -1:

```

```

        break

    linie_pivot = -1
    raport_minim = np.inf
    for linie, valoare in enumerate(self.matrice[:-1, -1]):
        elem = self.matrice[linie, coloana_pivot]
        if elem > 0:
            raport = valoare / elem
            if raport < raport_minim:
                linie_pivot = linie
                raport_minim = raport

    if linie_pivot == -1:
        return 'Valoarea optima: infinit'

    pivot(self.matrice, linie_pivot, coloana_pivot)

s = f'Valoarea optima: {self.matrice[-1][-1]}\n'
for index in range(self.matrice.shape[1] - 1 - self.variabile_libere):
    s += f'x{index + 1} = {self.matrice[-1, index]}\n'
return s

def transformareDual(self):
    """ Transforma sistemul pentru aplicarea algoritmului dual """
    self.adaugaTermeniLiberi(1)
    self.matrice = self.matrice.T
    self.extrageTermeniLiberi()
    self.tip = 'min' if self.tip == 'max' else 'max'
    operator = '<=' if self.tip == 'max' else '>='
    self.semne = [operator] * self.matrice.shape[0]

def simplexDual(self):
    """ Tema 4: Algoritmul simplex in varianta duala """
    self.transformareDual()
    self.standardizareLP()
    self.matrice[-1] *= -1

    while True:
        linie_pivot = -1
        for linie, valoare in enumerate(self.matrice[:, -1]):
            if valoare < 0:
                linie_pivot = linie
                break

        if linie_pivot == -1:
            break

```

```

coloana_pivot = -1
raport_minim = np.inf
for coloana, valoare in enumerate(self.matrice[-1, :-1]):
    elem = self.matrice[linie_pivot][coloana]
    if elem < 0:
        raport = valoare / elem
        if raport < raport_minim:
            coloana_pivot = coloana
            raport_minim = raport

if coloana_pivot == -1:
    return 'Valoarea optima: infinit'

pivot(self.matrice, linie_pivot, coloana_pivot)

s = f'Valoarea optima: {self.matrice[-1][-1]}\n'
for index in range(self.matrice.shape[1] - 1 - self.variabale_libere):
    s += f'y{index + 1} = {-self.matrice[-1, self.variabale_libere + ↵
↵index]}\n'
return s

```

### 3.1 TEMA 3:

Algoritmul simplex în varianta primală

```

[18]: s = Sistem(np.array([[1, 3, 3], [2, 7, 1]], dtype=float),
               np.array(['>=', '<=']),
               np.array([3, 5], dtype=float))
print(f'Sistemul initial:\n{s}')
print(f'Rezolvarea cu simplex primal:\n{s.simplexPrimal()}')

```

Sistemul initial:

```

max 3.0 * x1 + 5.0 * x2
1.0 * x1 + 3.0 * x2  >= 3.0
2.0 * x1 + 7.0 * x2  <= 1.0

```

Rezolvarea cu simplex primal:

```

Valoarea optima: 1.5
x1 = 0.0
x2 = 5.5

```

### 3.2 TEMA 4:

Algoritmul simplex în varianta duală

```

[19]: s = Sistem(np.array([[1, -3, -2, 2], [-1, 4, -3, -1], [2, 1, -4, 1]], ↵
               ↵dtype=float),

```

```

        np.array(['<=', '<=', '<=']),
        np.array([-1, -2, -4], dtype=float))
print(f'Sistemul initial:\n{s}')
s.transformareDual()

print(f'Sistemul dupa transformarea duala:\n{s}')
print(f'Rezolvarea cu simplex dual:\n{s.simplexDual()}')

```

Sistemul initial:

```

max -1.0 * x1 - 2.0 * x2 - 4.0 * x3
1.0 * x1 - 3.0 * x2 - 2.0 * x3  <= 2.0
-1.0 * x1 + 4.0 * x2 - 3.0 * x3  <= -1.0
2.0 * x1 + 1.0 * x2 - 4.0 * x3  <= 1.0

```

Sistemul dupa transformarea duala:

```

min 2.0 * x1 - 1.0 * x2 + 1.0 * x3
1.0 * x1 - 1.0 * x2 + 2.0 * x3  >= -1.0
-3.0 * x1 + 4.0 * x2 + 1.0 * x3  >= -2.0
-2.0 * x1 - 3.0 * x2 - 4.0 * x3  >= -4.0

```

Rezolvarea cu simplex dual:

Valoarea optima: 1.1

y1 = 0.0

y2 = 1.2

y3 = 0.1