Tp OPtimisation

```
In [1]:
          import numpy as np
          import math
          import matplotlib.pyplot as plt
In [2]:
         # Definition de la fonction principale
          def f(p):
              return 1/2*math.pow(p[0], 2) + 7/2* math.pow(p[1], 2)
In [ ]:
In [3]:
          # gradiant de f
          def grad f(p):
              return np.array([p[0] ,7*p[1]])
In [ ]:
In [45]:
          def norme(v):
              return math.sqrt(pow(v[0] , 2) + math.pow(v[1],2))
In [ ]:
In [58]:
          def produit_scalaire(vect1 , vect2):
              if(len(vect1) != len(vect2)):
                  return "imposible"
              else:
                  for i in range(len(vect1)):
                      s = s + vect1[i]*vect2[i]
                  return s
In [ ]:
```

```
In [62]: # pour calculer la norme d'un vecteur
    def normeg(vect):
        s = 0
        for i in range(len(vect)):
             s = s+ math.pow(vect[i] , 2)
        return math.sqrt(s)
In []:
```

Algorthime de Descente 1

```
In []:
In [4]: # second critere d'optimalité Stagnation de la solution courante

def stagnation_valeur_courante(xk_1 , xk):
    return norme(xk_1 - xk)
```

Descente à pas fixe respectant le critère d'optimalité

```
# x0 point de ou commence l'algorithme
In [46]:
          # s represente le pas
          # e represente la precision
          def descente a_pas_fixe(e , x0, s):
              k=0
              xk = x0
              data =[]
              try:
                  while ( norme(grad_f(xk)) > e):
                      dk = -grad_f(xk)
                      xk = xk + s* dk
                      k = k+1
                      data.append([k , xk])
              except:
                  return "Dv"
              return data
```

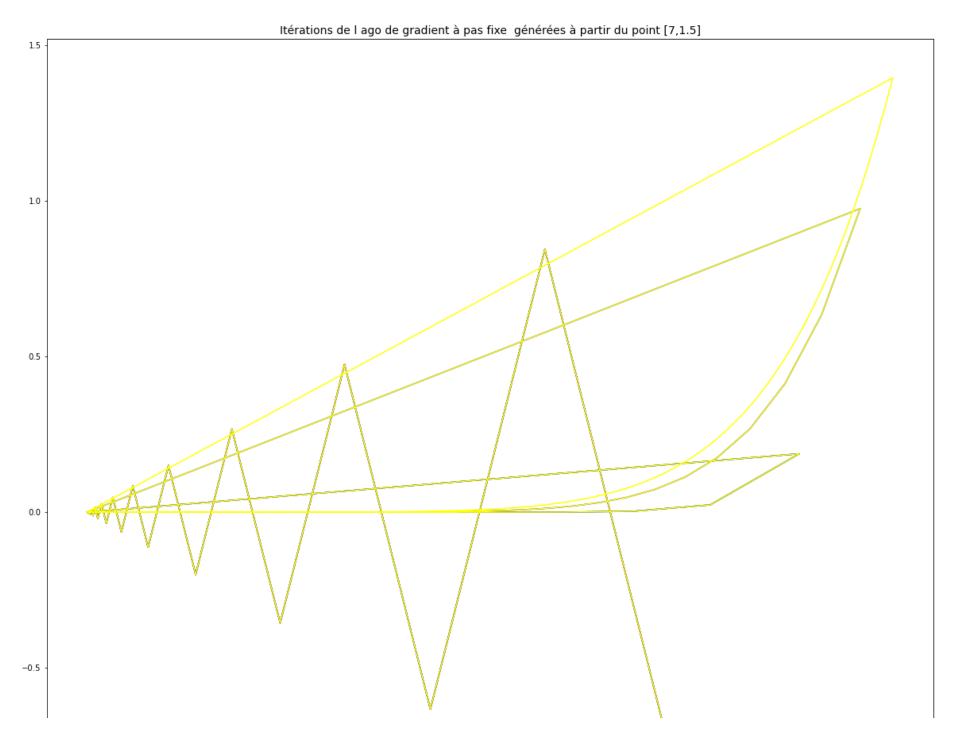
```
descente_a_pas_fixe(10e-5 , [7,1.5] , 0.125)
In [47]:
Out[47]: [[1, array([6.125, 0.1875])],
          [2, array([5.359375 , 0.0234375])],
          [3, array([4.68945312e+00, 2.92968750e-03])],
          [4, array([4.10327148e+00, 3.66210938e-04])],
          [5, array([3.59036255e+00, 4.57763672e-05])],
          [6, array([3.14156723e+00, 5.72204590e-06])],
          [7, array([2.74887133e+00, 7.15255737e-07])],
          [8, array([2.40526241e+00, 8.94069672e-08])],
          [9, array([2.10460461e+00, 1.11758709e-08])],
          [10, array([1.84152903e+00, 1.39698386e-09])],
          [11, array([1.61133790e+00, 1.74622983e-10])],
          [12, array([1.40992067e+00, 2.18278728e-11])],
          [13, array([1.23368058e+00, 2.72848411e-12])],
          [14, array([1.07947051e+00, 3.41060513e-13])],
          [15, array([9.44536696e-01, 4.26325641e-14])],
          [16, array([8.26469609e-01, 5.32907052e-15])],
          [17, array([7.23160908e-01, 6.66133815e-16])],
          [18, array([6.32765795e-01, 8.32667268e-17])],
          [19, array([5.53670070e-01, 1.04083409e-17])],
          [20, array([4.84461311e-01, 1.30104261e-18])],
          [21, array([4.23903647e-01, 1.62630326e-19])],
          [22, array([3.70915692e-01, 2.03287907e-20])],
          [23, array([3.24551230e-01, 2.54109884e-21])],
          [24, array([2.83982326e-01, 3.17637355e-22])],
          [25, array([2.48484536e-01, 3.97046694e-23])],
          [26, array([2.17423969e-01, 4.96308368e-24])],
          [27, array([1.90245973e-01, 6.20385459e-25])],
          [28, array([1.66465226e-01, 7.75481824e-26])],
          [29, array([1.45657073e-01, 9.69352280e-27])],
          [30, array([1.27449939e-01, 1.21169035e-27])],
          [31, array([1.11518696e-01, 1.51461294e-28])],
          [32, array([9.75788593e-02, 1.89326617e-29])],
          [33, array([8.53815019e-02, 2.36658272e-30])],
          [34, array([7.47088141e-02, 2.95822839e-31])],
          [35, array([6.53702124e-02, 3.69778549e-32])],
          [36, array([5.71989358e-02, 4.62223187e-33])],
          [37, array([5.00490688e-02, 5.77778983e-34])],
          [38, array([4.37929352e-02, 7.22223729e-35])],
          [39, array([3.83188183e-02, 9.02779661e-36])],
          [40, array([3.35289660e-02, 1.12847458e-36])],
```

```
[41, array([2.93378453e-02, 1.41059322e-37])],
[42, array([2.56706146e-02, 1.76324153e-38])],
[43, array([2.24617878e-02, 2.20405191e-39])],
[44, array([1.96540643e-02, 2.75506488e-40])],
[45, array([1.71973063e-02, 3.44383111e-41])],
[46, array([1.50476430e-02, 4.30478888e-42])],
[47, array([1.31666876e-02, 5.38098610e-43])],
[48, array([1.15208517e-02, 6.72623263e-44])],
[49, array([1.00807452e-02, 8.40779079e-45])],
[50, array([8.82065206e-03, 1.05097385e-45])],
[51, array([7.71807055e-03, 1.31371731e-46])],
[52, array([6.75331173e-03, 1.64214664e-47])],
[53, array([5.90914777e-03, 2.05268330e-48])],
[54, array([5.17050429e-03, 2.56585412e-49])],
[55, array([4.52419126e-03, 3.20731765e-50])],
[56, array([3.95866735e-03, 4.00914707e-51])],
[57, array([3.46383393e-03, 5.01143383e-52])],
[58, array([3.03085469e-03, 6.26429229e-53])],
[59, array([2.65199785e-03, 7.83036536e-54])],
[60, array([2.32049812e-03, 9.78795670e-55])],
[61, array([2.03043586e-03, 1.22349459e-55])],
[62, array([1.77663137e-03, 1.52936823e-56])],
[63, array([1.55455245e-03, 1.91171029e-57])],
[64, array([1.36023340e-03, 2.38963787e-58])],
[65, array([1.19020422e-03, 2.98704733e-59])],
[66, array([1.04142869e-03, 3.73380917e-60])],
[67, array([9.11250107e-04, 4.66726146e-61])],
[68, array([7.97343844e-04, 5.83407682e-62])],
[69, array([6.97675863e-04, 7.29259603e-63])],
[70, array([6.10466381e-04, 9.11574504e-64])],
[71. array([5.34158083e-04. 1.13946813e-64])].
[72. array([4.67388323e-04. 1.42433516e-65])].
[73, array([4.08964782e-04, 1.78041895e-66])],
[74, array([3.57844184e-04, 2.22552369e-67])],
[75, array([3.13113661e-04, 2.78190461e-68])],
[76, array([2.73974454e-04, 3.47738077e-69])],
[77. array([2.39727647e-04. 4.34672596e-70])].
[78, array([2.09761691e-04, 5.43340745e-71])],
[79, array([1.83541480e-04, 6.79175931e-72])],
[80, array([1.60598795e-04, 8.48969914e-73])],
[81, array([1.40523945e-04, 1.06121239e-73])],
[82, array([1.22958452e-04, 1.32651549e-74])],
[83, array([1.07588646e-04, 1.65814436e-75])],
```

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m sur} \ 12$ 07/12/2021, 09:14

```
In [ ]:
In [ ]:
        Tableau de test pas le pas fixe
        pasTest = [0.25, 0.125, 0.05, 0.01]
         pasTest = [0.25, 0.125, 0.05, 0.01]
In [15]:
         resultat= []
In [48]:
          for i in range(len(pasTest)):
              resultat.append(descente a pas fixe(10e-5 , np.array([7,1.5] ),pasTest[i]))
         def dataplot(datap):
In [52]:
              x=[]
              y=[]
              fig ,ax = plt.subplots()
              ax.set prop cycle(color=['red','green','blue','yellow'])
              for j in range(len(datap)):
                  for i in range(len(datap[j]) ):
                      x.append(datap[j][i][1][0])
                     y.append(datap[j][i][1][1])
                  plt.rcParams["figure.figsize"] = (20,20)
                  ax.set title('Itérations de l ago de gradient à pas fixe générées à partir du point [7,1.5]',
                 fontsize = 14)
                  plt.plot(x,y ,label = "itermanfbjkfbgfkjbg")
              plt.show()
In [ ]:
In [53]:
         dataplot(resultat)
```

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```
In [ ]:
In [ ]: #
In [ ]:
In [54]:
          datap= descente_a_pas_fixe(10e-5 , np.array([7,1.5] ),0.01)
          x=[]
          y=[]
          for i in range(len(datap) ):
              x.append(datap[i][1][0])
              y.append(datap[i][1][1])
          #datap[1][1][0]
          plt.rcParams["figure.figsize"] = (20,20)
          #plt.plot(x,y , color ="blue")
          #plt.show()
          print("ok")
         ok
```

Descente à pas Optimal respectant le critère d'optimalité

```
In [ ]:
```

```
#pas Optimal solution du probleme Minf(xk+s*dk) avec s positif
In [551:
          def s(x):
              return (pow(x[0],2) + pow(7,2)*pow(x[1], 2)) / (pow(x[0],2) + pow(7,3)*pow(x[1], 2))
In [ ]:
         # e : precision
          #x0 : point de départ
          def descentePasOptimal(e , x0):
              xk = x0
              k=0
              data = []
              \#norme(grad\ f(x0)) \le e
              while(norme(grad f(xk)) > e):
                  dk = -grad f(xk)
                  xk = xk + s(xk)*dk
                  k = k+1
                  data.append([k,xk])
              return data
         descentePasOptimal(10e-10 ,[7,1.5])
In [ ]:
```

Algroiithme 2 Amidjo et wolfe

```
f(x+sd) <= f(x) + ep1d(grad(f(x))T*d)
soit fi(s) <= fi(0) + ep1 d fi'(0)
In []:
In []:
In [56]: f(np.array([7,1.5]) + 0.5* np.array([-7, -10.5]))
Out[56]: 55.34375
```

```
In [ ]:
In [ ]:
In [59]: f(np.array([7,1.5]) )+0.125*10e-4* produit_scalaire(grad_f(np.array([7,1.5])) , np.array([-7, -10.5]))
Out[59]: 32.35509375
          def premiere condition de wolfe(x ,s, dk , ep1 ):
In [60]:
              if f(x + s*dk) \leftarrow f(x) + ep1*s* produit scalaire(grad <math>f(x), dk):
                  return True
              else:
                  return False
          premiere condition de wolfe(np.array([8,0]) , 1 ,np.array([1,0]),0.0)
In [61]:
Out[61]: False
In [66]:
          def deuxieme condition de wolfe(x,s,dk , ep2):
              if produit scalaire( grad f(x + s*dk) , dk)
                                                              < ep2 * produit scalaire(grad f(x), dk):</pre>
                  return False
              else:
                  True
          deuxieme condition de wolfe(np.array([8,19]) , 0.1 ,np.array([1,1]),0.8)
In [ ]:
In [ ]:
In [ ]:
          produit Scalaire(np.array([1,2]) , np.array([2,3,3]))
In [ ]:
In [63]:
           norme([2,1,4])
Out[63]: 2.23606797749979
In [ ]:
```

Out[67]: 0.25

```
In [ ]:
In [64]:
          def recherche lineaire de wolfe(x,ep1 ,ep2 ,s, d ):
              s_=0
              s_plus = float('inf')
              sk = s
              infini = float('inf')
              k = 0
              condition_wolfe = False
              while condition wolfe == False:
                  if premiere condition de wolfe(x ,sk, d , ep1 ) == False :#pas trop grad
                      s plus = sk
                      sk = (s_plus + s_)/2
                      condition wolfe = False
                  elif deuxieme_condition_de_wolfe(x,sk,d , ep2) == False :# pas trop petit
                      s = sk
                      if s plus < infini :</pre>
                          sk = (s_plus + s_)/2
                      else:
                          sk = 2*sk
                  else :
                      condition_wolfe = True
                  k = k+1
              return sk
          recherche_lineaire_de_wolfe(np.array([7,1.5]), 10e-4, 0.99,8, np.array([-7,-10.5]))
In [67]:
```

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```
def descente_gradiant_wolfe(e , x0 , ep1 = 10e-4 , ep2 = 0.99 ,s = 1000000) :
In [76]:
              xk = x0
              k=0
              data=[]
              #norme(grad_f(x0))<=e</pre>
              while(norme(grad f(xk)) > e):
                   dk = -(grad f(xk))
                   sk = recherche_lineaire_de_wolfe(xk,ep1 ,ep2 ,s, dk )
                   xk = xk + sk* \overline{d}k
                   k = k+1
                   print(k)
                   data.append([k,xk])
               return xk
          descente_gradiant_wolfe(10e-5 , np.array([7, 1.5]) )
In [78]:
          2
          9
          10
         11
         12
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          22
          23
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```

```
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34
35
36
37
           38
Out[78]: array([ 5.53091490e-05, -9.91407779e-06])
 In [ ]:
 In [ ]:
```