task

April 15, 2021

1 Tarefa 2: Álgebra Linear e Otimização para ML - MO431A

Universidade Estadual de Campinas (UNICAMP), Instituto de Computação (IC)

Prof. Jacques Wainer, 2021s1

```
[1]: # RA & Name

print('265673: ' + 'Gabriel Luciano Gomes')

print('192880: ' + 'Lucas Borges Rondon')

print('265674: ' + 'Paulo Júnio Reis Rodrigues')
```

265673: Gabriel Luciano Gomes 192880: Lucas Borges Rondon

265674: Paulo Júnio Reis Rodrigues

1.0.1 Imports necessários

```
[2]: import numpy as np
import matplotlib.pyplot as plt
from matplotlib.colors import LogNorm
import tensorflow as tf
```

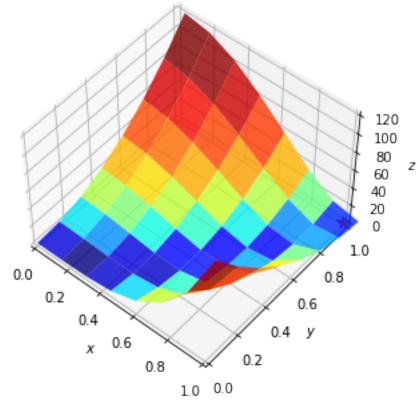
1.1 Equação Rosenbrock 2D

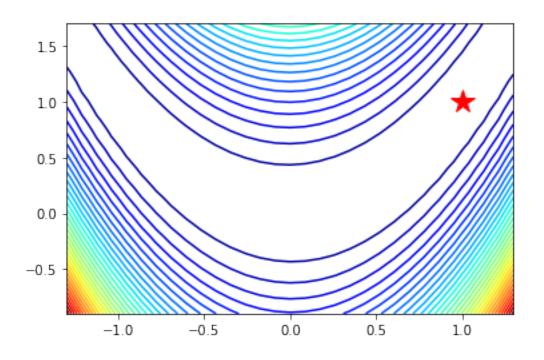
```
[3]: # função fx de Rosenbrock em 2D
def fx(x1, x2):
return ((1 - x1)**2) + 100*((x2 - x1**2)**2)
```

1.1.1 Formato da função

```
minima_ = minima.reshape(-1, 1)
fig = plt.figure(figsize=(8, 5))
ax = plt.axes(projection='3d', elev=50, azim=-50)
plt.title(r'Rosenbrock Function: f(x,y) = (1-x)^2 + 100(y-x^2)^2')
ax.plot_surface(x, y, z, norm=LogNorm(), rstride=1, cstride=1,
                edgecolor='none', alpha=.8, cmap=plt.cm.jet)
ax.plot(*minima_, fx(*minima_), 'r*', markersize=10)
ax.set_xlabel('$x$')
ax.set_ylabel('$y$')
ax.set_zlabel('$z$')
ax.set_xlim((xmin, xmax))
ax.set_ylim((ymin, ymax))
plt.show()
_, ax = plt.subplots(1, 1)
# make a contour plot of the rosenbrock function surface.
X, Y = np.meshgrid(np.linspace(-1.3, 1.3, 31), np.linspace(-0.9, 1.7, 31))
Z = fx(X, Y)
ax.plot(*minima_, 'r*', markersize=18)
ax.contour(X, Y, Z, 40, cmap=plt.cm.jet)
ax.set_xlim(-1.3, 1.3)
ax.set_ylim(-0.9, 1.7)
plt.show()
```

Rosenbrock Function: $f(x, y) = (1 - x)^2 + 100(y - x^2)^2$





1.1.2 Função gradiente de f

```
[5]: # derivada parcial de fx em x

def dx(x1, x2):
    return 2 * (200*x1**3 - 200*x1*x2 + x1 -1)

# derivada parcial de fx em y

def dy(x1, x2):
    return 200*(x2 - x1**2)
```

1.1.3 Plot das informações

```
[6]: def plotInfo(points, fx_values, steps):
         newList = [(elem1, elem2) for elem1, elem2 in points]
         plt.title(r'Values of $f(x)$ X # of iterations')
         plt.plot(fx_values)
         plt.ylabel(r'values of $f(x)$')
         plt.xlabel(r'# of iterations')
         plt.show()
         path = np.array(newList).T
         _, ax = plt.subplots(1, 1)
         # make a contour plot of the rosenbrock function surface.
         X, Y = \text{np.meshgrid}(\text{np.linspace}(-1.3, 1.3, 31), \text{np.linspace}(-0.9, 1.7, 31))
         Z = fx(X, Y)
         ax.plot(*minima_, 'r*', markersize=18)
         ax.contour(X, Y, Z, 40, cmap=plt.cm.jet)
         ax.quiver(path[0,:-1], path[1,:-1], path[0,1:]-path[0,:-1], path[1,1:]
      \rightarrow]-path[1,:-1],
             scale_units='xy', angles='xy', scale=2, color='k')
         ax.set_xlim(-1.3, 1.3)
         ax.set_ylim(-0.9, 1.7)
         plt.show()
         print(f'Best value of x: {points[len(points)-1]}')
         print(f'Number os steps: {steps}')
```

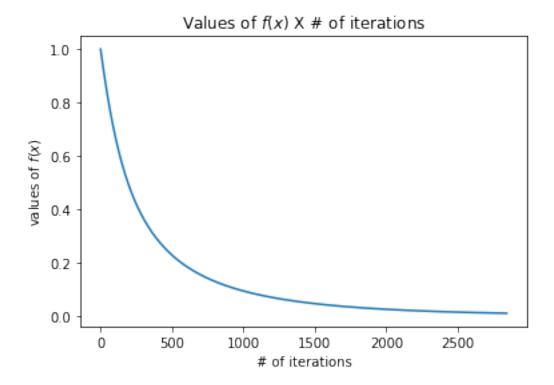
1.1.4 L.r = 1.e-3

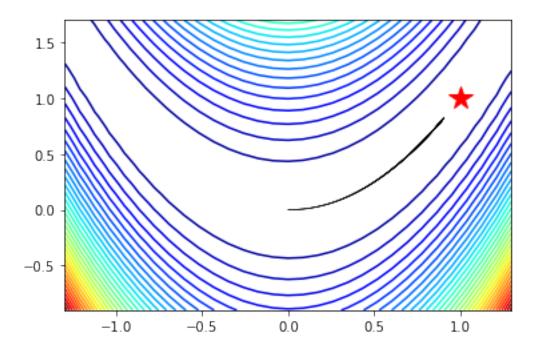
```
[7]: x = 0.0
y = 0.0
points = [np.array([x, y])]
```

```
fx_values = [fx(x,y)]
tol = 1
steps = 0
lr = 10**-3
while (tol > 10**-5) and (steps < 50000):
   # Compute function to old point
   f_old = fx(x, y)
    # Compute gradient and new point
   x = lr * dx(x, y)
    y = lr * dy(x, y)
    # Compute function to new point
   f_{new} = fx(x, y)
    # Compute tolerancy
    tol = np.abs(f_new - f_old)
    # Append current point to list
    points.append(np.array([x, y]))
    fx_values.append(f_new)
    steps+=1
print(f'Valor da tolerância: {tol}, Valor de passos: {steps}')
```

Valor da tolerância: 9.998940970952844e-06, Valor de passos: 2842

```
[8]: plotInfo(points, fx_values, steps)
```





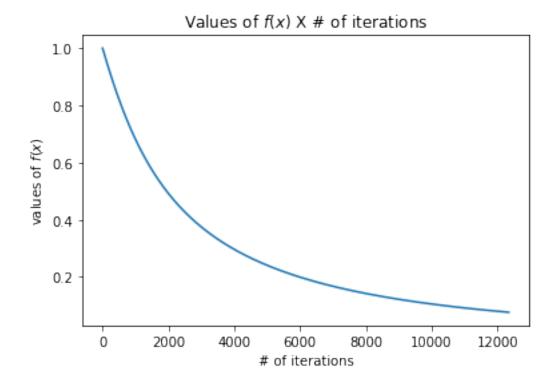
Best value of x: [0.90489266 0.81845007]

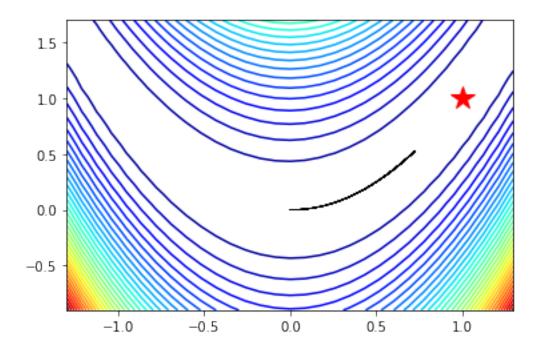
1.1.5 L.r = 1.e-4

```
[9]: x = 0.0
    y = 0.0
     points = [np.array([x, y])]
     fx_values = [fx(x,y)]
     tol = 1
     steps = 0
     lr = 10**-4
     while (tol > 10**-5) and (steps < 50000):
         # Compute function to old point
         f_old = fx(x, y)
         # Compute gradient and new point
         x = lr * dx(x, y)
         y = lr * dy(x, y)
         # Compute function to new point
         f_{new} = fx(x, y)
         # Compute tolerancy
         tol = np.abs(f_new - f_old)
         # Append current point to list
         points.append(np.array([x, y]))
         fx_values.append(f_new)
         steps+=1
     print(f'Valor da tolerância: {tol}, Valor de passos: {steps}')
```

Valor da tolerância: 9.999994764961495e-06, Valor de passos: 12346

```
[10]: plotInfo(points, fx_values, steps)
```





Best value of x: [0.72373005 0.5225001]

1.1.6 L.r grande

```
[11]: x = 0.0
      y = 0.0
      points = [np.array([x, y])]
      fx_values = [fx(x,y)]
      tol = 1
      steps = 0
      lr = 8**-2
      while (tol > 10**-5) and (steps < 50000):
          # Compute function to old point
          f_old = fx(x, y)
          # Compute gradient and new point
          x = lr * dx(x, y)
          y = lr * dy(x, y)
          # Compute function to new point
          f_{new} = fx(x, y)
          # Compute tolerancy
          tol = np.abs(f_new - f_old)
          # Append current point to list
          points.append(np.array([x, y]))
          fx_values.append(f_new)
          steps+=1
      print(f'Valor da tolerância: {tol}, Valor de passos: {steps}')
```

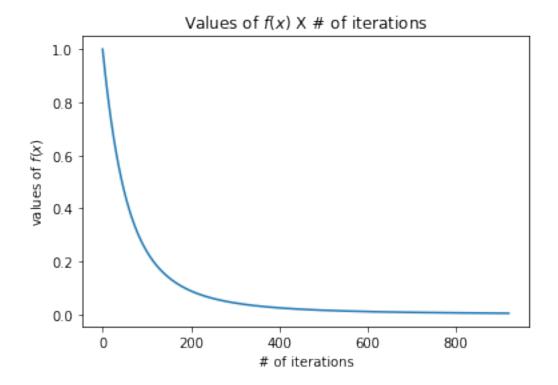
```
OverflowError: (34, 'Result too large')
```

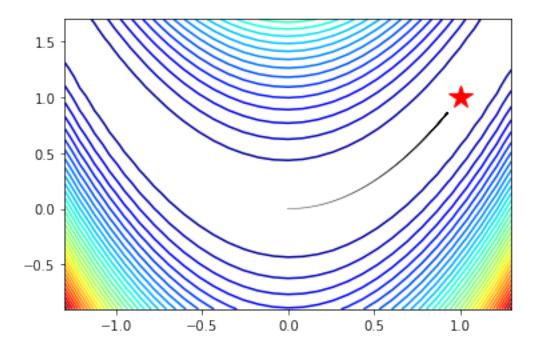
1.1.7 Politica de redução do l.r

```
[12]: x = 0.0
      y = 0.0
      points = [np.array([x, y])]
      fx_values = [fx(x,y)]
      tol = 1
      steps = 0
      lr = 3**-5
      while (tol > 10**-5) and (steps < 50000):
          # Compute function to old point
          f_old = fx(x, y)
          # Compute gradient and new point
          x = lr * dx(x, y)
          y = lr * dy(x, y)
          # Compute function to new point
          f_{new} = fx(x, y)
          # Compute tolerancy
          tol = np.abs(f_new - f_old)
          # Append current point to list
          points.append(np.array([x, y]))
          fx_values.append(f_new)
          # Updating learning rate
          lr = lr * 0.999
          steps+=1
      print(f'Valor da tolerância: {tol}, Valor de passos: {steps}')
```

Valor da tolerância: 9.985739491158961e-06, Valor de passos: 919

```
[13]: plotInfo(points, fx_values, steps)
```



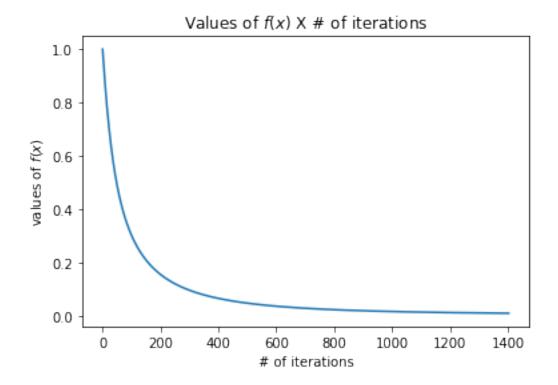


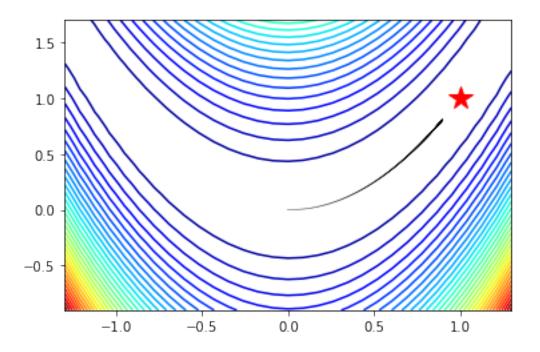
Best value of x: [0.92901798 0.86280725]

1.1.8 Utilizando Tensorflow para calcular o gradiente

```
[14]: points = [np.array([0.0, 0.0])]
      x = tf.Variable(0.0)
      y = tf.Variable(0.0)
      fx_values = [fx(x,y).numpy()]
      tol = 1
      steps = 0
      lr = 3**-5
      while (tol > 10**-5) and (steps < 50000):
          # Compute function to old point
          f_old = fx(x, y).numpy()
          # Computing gradient with TensorFlow
          with tf.GradientTape(persistent=True) as g:
            g.watch([x,y])
            # Rosenbrock function
            z = (1-x)**2 + 100*((y-x**2)**2)
          # Generatin new point
          # Multiply gradient by learning rate
          x.assign_sub(lr * g.gradient(z, x).numpy())
          y.assign_sub(lr * g.gradient(z, y).numpy())
          # Compute function to new point
          f_{new} = fx(x, y).numpy()
          # Compute tolerancy
          tol = abs(f_new - f_old)
          # Append current point to list
          points.append(np.array([x, y]))
          fx_values.append(f_new)
          # Updating learning rate
          lr = lr * 0.999
          steps += 1
```

```
[15]: plotInfo(points, fx_values, steps)
```





Best value of x: [0.8979067 0.8058023]