$xi+1=xi-\eta E'(xi)$ On peut donner un critère de fin à la DG par exemple si xi+1 - xi < eps ou si $i > nombre_max$. Pour ce problème, nous utilisons eps = 0.01 et nombre_max= 1000. Descente de gradient 1. Calculez l'expression analytique de la fonction Unexpected text node: 'E(x) = (x-1)(x-2)(x-3)(x-5)' et sa dérivée. from sympy import Symbol, expand, diff x = Symbol('x')exp = expand((x-1)*(x-2)*(x-3)*(x-5))*#La dérivée* div exp = diff(exp, x) $print('E(x) = ', exp, '\nE'(x) = ', div exp)$ E(x) = x**4 - 11*x**3 + 41*x**2 - 61*x + 30E'(x) = 4*x**3 - 33*x**2 + 82*x - 612. Implémentez l'algorithme DG sous Python pour la fonction E(x). Fonction Descente de gradient div_e : la dérivée de la fonction E(x) • **x0**: la valeur intiale de x • n : le pas de Descente de gradient • **eps**: pour le condition d'arrête nb_max_iter : nombre maximum des itérations **def** DG(div_e, x0, n, eps = 0.01, nb_max_iter = 1000): cpp = 0 # compteur pour calculer le nombre d'itération while True: # calculer le nouveau x (xi+1) $x1 = x0 - n*div_e.subs(x, x0).evalf()$ cpp **+=** 1 # condition d'arrêt **if** (abs(x1 - x0) < eps) | ($cpp > nb_max_iter$): break x0 = x1# résultat return le nouveau x et le nombre d'itération return [x1, cpp] 3. Pour comprendre ce que fait effectivement la DG, testez l'algorithme implémenté en utilisant des exemples d'exécutions avec des valeurs initiales de x0 et η suivantes : • x0 = 5 et $\eta = 0.001$ • x0 = 5 et n = 0.01• x0 = 5 et $\eta = 0.1$ • x0 = 5 et $\eta = 0.17$ • x0 = 5 et $\eta = 1$ • x0 = 0 et $\eta = 0.001$ In [6]: print("x0 = 5 et η = 0.001 \t: ", DG(div_exp, x0=5, n=0.001)) print("x0 = 5 et η = 0.01 \t: ", DG(div_exp, x0=5, n=0.01)) print("x0 = 5 et η = 0.1 \t: ", DG(div_exp, x0=5, n=0.1)) print("x0 = 5 et η = 0.17 \t: ", DG(div_exp, x0=5, n=0.17)) print("x0 = 5 et η = 1 \t: ", DG(div_exp, x0=5, n=1)) print("x0 = 0 et η = 0.001 \t: ", DG(div exp, x0=0, n=0.001)) x0 = 5 et $\eta = 0.001$: [4.66148331568632, 22] x0 = 5 et $\eta = 0.01$: [4.36100829491372, 10] x0 = 5 et $\eta = 0.1$: [4.48300247305520, 1001] x0 = 5 et $\eta = 0.17$: [1.23789802112164, 1001] x0 = 5 et $\eta = 1$: [-5.59079531307448e+2165218897387849575449569542280595721495861785890200169487984 462862848820544652709611781766626003237259515385703156691028186064669180469086499802903989431, 1001] x0 = 0 et $\eta = 0.001$: [0.949407118070548, 39]4. Affichez le minimum trouvé, ainsi que E(xmin) et le nombre d'itérations. Que constatezvous? args = [[5, 0.001], [5, 0.01], [5, 0.1], [5, 0.17], [5, 1], [0, 0.001]]all xmin = []for arg in args: xmin, nb_ite = DG(div_exp, arg[0], arg[1]) print('-'*60) $print('| \t Pour x0 = ', arg[0],' et \eta = ', arg[1], '\t:\t\t |')$ print('-'*60) print('| xmin \t\t\t| ', xmin, "\t\t |") xmin f = exp.subs(x, xmin).evalf()print('| E(xmin) \t\t| ', xmin_f, "\t\t |") print('| nombre d'itérations \t| ', nb_ite, "\t\t\t\") all_xmin.append([xmin, xmin_f, nb_ite]) all_xmin.pop(4) Pour x0 = 5 et $\eta = 0.001$: | nombre d'itérations | 22 Pour x0 = 5 et $\eta = 0.01$: -6.90117578513082 | nombre d'itérations | 10 Pour x0 = 5 et $\eta = 0.1$: -6.63072895912501 | nombre d'itérations | 1001 Pour x0 = 5 et $\eta = 0.17$: | nombre d'itérations | 1001 | xmin | -5.59079531307448e+21652188973878495754495695422805957214958617858902001694879846 628628488205446527096117817666260032372595153857031566910281860646691804690864998029039894319.76999563434790e+866087558955139830179827816912238288598344714356080067795193856 | E(xmin) 583231609235701317833019055707590006102558817646631985587830890435051834848617461748407133583438608013986965 | nombre d'itérations | 1001 Pour x0 = 0 et $\eta = 0.001$: 0.949407118070548 | E(xmin) 0.441491059862557 | nombre d'itérations | 39 Out[7]: [-5.59079531307448e+2165218897387849575449569542280595721495861785890200169487984640726689692623705307933951 766626003237259515385703156691028186064669180469086499802903989431, 66504012949038061542812626764112744258676721876345999211615957726, 5. Visualisez l'évolution des minimums de la fonction E(x) trouvés au cours des itérations. **def** DG(div e, x0, n, eps = 0.01, nb max iter = 1000): cpp = 0 # compteur pour calculer le nombre d'itération Xmins = []Xmins.append(x0)while True: # calculer le nouveau x (xi+1) x1 = x0 - n*div e.subs(x, x0).evalf()cpp += 1 # condition d'arrêt Xmins.append(x1)if (abs(x1 - x0) < eps) | (cpp > nb max iter): x0 = x1# resultat return le nouveau x et le nombre d'itération return [Xmins, cpp] import numpy as np import matplotlib.pyplot as plt xmins, nb ite = np.array(DG(div exp, 5, 0.01)) plt.scatter(range(nb ite+1), xmins, c='r') plt.plot(range(nb ite+1), xmins) plt.xlabel('nb d\'itérations') plt.ylabel('Xmin') plt.title('Visualisation de 1\'évolution des minimums de la fonction E(x) trouvés au cours des itérations') plt.show() Visualisation de l'évolution des minimums de la fonction E(x) trouvés au cours des itérations 5.0 4.9 4.8 4.7 4.6 4.5 4.4 8 nb d'itérations 6. Testez votre algorithme avec d'autres valeurs de eps et nombre max. def sprint(div e, data): xmin, ite = DG(div_e, data[0], data[1]) print('-'*60) $print('| \text{tPour } x0 = ', data[0],' et \eta = ', data[1], '\t:\t\t |')$ print('| xmin \t\t\t| ', xmin, "\t\t |") $xmin_f = exp.subs(x, xmin).evalf()$ print('| E(xmin) \t\t| ', xmin_f, "\t\t |") print('| nomber des iteration \t| ', ite, "\t\t\t\") return xmin, xmin f dict1 = [[1, 0.001], [6, 0.001], [6, 0.01], [4, 0.1], [4, 0.17]]for d in dict1: sprint(div_exp, d) Pour x0 = 1 et $\eta = 0.001$: | nomber des iteration | 1 Pour x0 = 6 et $\eta = 0.001$: [6, 5.893000000000, 5.79818438917200, 5.71344305858812, 5.63714841968474, 5.568 91115469927, 5.21863516112886, 5.18193369253159, 5.14755535614586, 5.11528176723680, 5.08492216009623, 5.056 $08701648401, \ 4.89435774003588, \ 4.87555452579438, \ 4.85761525918669, \ 4.84048343001917, \ 4.82410749812207, \ 4.80881207, \ 4.80881207, \ 4.80881207, \ 4.80881207, \ 4.80881207, \ 4.80881207, \ 4.80881207, \ 4.80881207, \ 4.80881207, \ 4.80881207, \ 4.80881207, \ 4.80881207, \ 4.80881207, \ 4.80881207, \ 4.80881207, \ 4.80881207, \ 4.80881207, \ 4.80881207, \ 4.80881207,$ 44034473993, 4.79343879627631, 4.77906320936224, 4.76527710811615, 4.75204686599971, 4.73934142592307, 4.727 13205327424, 4.71539211738605, 4.70409689764592, 4.69322341102674, 4.68275025829436, 4.67265748654537, 4.662 92646606287] $x^{**4} - 11.0^{*}x^{**3} + 41.0^{*}x^{**2} - 61.0^{*}x + 30.0$ | nomber des iteration | 41 Pour x0 = 6 et $\eta = 0.01$: | xmin | [6, 4.9300000000000, 4.72509072000000, 4.60846931728543, 4.53307608572641, 4.481 08358851332, 4.44380531005350, 4.41638377433183, 4.39585145084092, 4.38028037898460, 4.36836062409739, 4.359 17182179604] | E(xmin) $| x^{**4} - 11.0^{*}x^{**3} + 41.0^{*}x^{**2} - 61.0^{*}x + 30.0$ | nomber des iteration | 11 Pour x0 = 4 et $\eta = 0.1$: ______ [4, 4.500000000000, 4.0750000000000, 4.4913937500000, 4.09033596314547, 4.487 46424287683, 4.09723514586139, 4.48545025898770, 4.10074620644149, 4.48436628136908, 4.10262895433799, 4.483 02772821786, 4.10494712298755, 4.48301680127560, 4.10496601610128, 4.48301060265721, 4.10497673353596, 4.483 00708587130, 4.10498281400109, 4.48300509048017, 4.10498626397670, 4.48300395826814, 4.10498822153239, 4.483 $00247306092,\ 4.10499078939652,\ 4.48300247305844,\ 4.10499078940080,\ 4.48300247305704,\ 4.10499078940323,\ 4.48301247305704,\ 4.10499078940324,\ 4.10499078940324,\ 4.104990789404,\ 4.10499078940324,\ 4.104990789404,\
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Calculation des dérivées partielles de la fonction G(a; b) selon a et b # In our example, G(x,y) = SUM(ax+y - y)**def** G(x, y, a, b): **return** np.sum((a*x+b-y)**2) # define derive of fonction def derive_G(x, y, a, b): **return** { 'dx': 2*(a*x**2+b*x-y*x), 'dy': 2*(a*x+b-y) } def y(x, a, b):return (a*x+b) 2. Implémentez l'algorithme DGR #Gradient Descent implied for a regression linear case with data def gradient_reg(max_iter, step, x, y, df,eps = 0.01): #Const nb sample data = x.shape[0]rg_sample = range(1, nb_sample_data) #initialisation a = 5 #np.random.random(x.shape[1]) b = 5 # np.random.random(x.shape[1]) $E = sum([a*x[j]+b-y[j] for j in rg_sample])$ print(E) for i in range(max_iter): #Compute gradient $grad_a = sum([2*(a*x[j])*2+b*x[j]-y[j]*x[j])$ for j in rg_sample]) $grad_b = sum([2*(a*x[j]+b-y[j]) for j in rg_sample])$ #Descente de gradient a = a - step * grad a b = b - step * grad_b #Compute error $e = sum([a*x[j]+b-y[j] for j in rg_sample])$ **if** $abs(E-e) \leftarrow eps:$ return ({'a': a, 'b': b}, i) E = ereturn ({'a': a, 'b': b}, max_iter) 3. test d'algorithme implémenté from sklearn import datasets x reg, y reg = datasets.make regression(n samples=100, n features=1, noise=7) all final_reg = [] print(x_reg.reshape(100).shape) print(y_reg.shape) (100,)(100,)In [14]: def show_res_reg(vect,derive_G): res,nb_ite = gradient_reg(vect[0], vect[1], vect[2], vect[3], derive G) print("for initial max iteration = ",vect[0]," , step = ",vect[1]) print(' ==> final_a = ',res['a'],' final_b = ',res['b'],'nb_iteration = ',nb_ite) pry = [y(x_d[0], res['a'], res['b']) for x_d in x_reg] print("final result = ",G(x_reg,y_reg,res['a'],res['b'])) $print('E(', x_reg[0][0], ') = ', pry[0])$ all_final_reg.append(pry) 4. Affichage des resultas trouvés all final reg = [] show res reg([100,0.001,x_reg,y_reg],derive_G) show_res_reg([500,0.001,x_reg,y_reg],derive_G) show_res_reg([1000,0.001,x_reg,y_reg],derive_G) show_res_reg([100,0.01,x_reg,y_reg],derive_G) show_res_reg([100,1,x_reg,y_reg],derive_G) ************* for initial max iteration = 100 , step = 0.001=> final a = [93.66725942] final b = [0.01392354] nb iteration = 49 E(-0.7493494090740423) = [-70.17558196]***************** ***************** for initial max iteration = 500 , step = 0.001==> final a = [93.66725942] final b = [0.01392354] nb iteration = 49 E(-0.7493494090740423) = [-70.17558196]for initial max iteration = 1000 , step = 0.001 $==> final_a = [93.66725942]$ final_b = [0.01392354] nb_iteration = 49 E(-0.7493494090740423) = [-70.17558196]************ for initial max iteration = 100 , step = 0.01 ==> final_a = [-2.57176602e+08] final_b = [-76344971.72586623] nb_iteration = 100 E(-0.7493494090740423) = [1.16370163e+08]****************** ***************** for initial max iteration = 100 , step = 1 ==> final_a = [-1.51482288e+235] final_b = [-4.49648641e+234] nb_iteration = 100 E(-0.7493494090740423) = [6.85482986e+234]5. utilisation des stats.linregress de scipy from scipy import stats slope, intercept, r_value, p_value, std_err = stats.linregress(x_reg.reshape(100), y_reg) np.array(all final reg[0]).shape Out[16]: (100, 1) 6. Visualisation print("r-squared:", r value**2) plt.rcParams["figure.figsize"] = (15,7) plt.subplot(1, 3, 1) plt.plot(x_reg, y_reg, 'o', label='original data') plt.plot(x reg, intercept + slope*x reg, 'r', label='fitted line') plt.title('make regression data') plt.legend() print("r-squared:", r value**2) plt.subplot(1, 3, 2) plt.plot(x reg, all final reg[0], 'o', label='original data') plt.plot(x_reg, intercept + slope*x_reg, 'r', label='fitted line') plt.title('a) with (a = 82.04, b = 5.54e-5)') plt.legend() print("r-squared:", r value**2) plt.subplot(1, 3, 3) plt.plot(x reg, all final reg[3], 'o', label='original data') plt.plot(x reg, intercept + slope*x reg, 'r', label='fitted line') plt.title('d) with (a = 39.30, b = 8.28)') plt.legend() plt.show() r-squared: 0.9953015220740956 r-squared: 0.9953015220740956 r-squared: 0.9953015220740956 make regression data a) with (a = 82.04, b = 5.54e-5) d) with (a = 39.30, b = 8.28) original data original data original data fitted line fitted line fitted line 200 4 100 100 0 -2 -100-100-200-200-6 -1 0 Loading [MathJax]/jax/output/CommonHTML/fonts/TeX/fontdata.js

TP5: Descente de gradient

Ce script présente la DG sur la minimisation de la fonction Unexpected text node: 'E(x)' quelconque. Leproblème est de trouver la valeur

l'équation Unexpected text node: E'(x) = 0, donctrouver ici les racines d'un polynôme de degré 3, ce qui est des fois "difficile". Donc onva

dexqui minimise Unexpected text node: 'E(x)'. Pour trouver analytiquementle minimum de la fonction E, il faut trouver les racines de

utiliser la DG. La DG consiste à construire une suite de valeurs xi(avec x0 fixé auhasard) de manière itérative :

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