TME n°3 - Introduction à l'optimisation

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Exercice 6 - Fonction de Rosenbrock et méthode de Newton

1.

```
1 % x1 et x2 sont des variables symboliques
2 syms x1 x2
3
4 f = 100*(x2 - x1^2)^2 + (1 - x1)^2;
6 H = hessian(f, [x1, x2]);
7 G = gradient(f, [x1, x2]);
```

2.

 x^* est un minimum de f si sa matrice Hessienne est définie-positive. Une matrice est définie-positive si chacune de ses valeurs propres est positive.

```
1 x1_star = 1;
2 x2_star = 1;
4 x1 = x1_star;
5 x2 = x2_star;
7 eigenvalues = eig(double(subs(H)));
8 nb_eigenvalues = size(eigenvalues,1);
10 positive = true;
11
12 for i = 1:nb_eigenvalues
   if(eigenvalues(i) < 0)</pre>
         positive = false;
14
15
           break;
     end
16
17 end
19 if (positive)
20
      disp('x* est bien un minimum local de f');
21 else
     disp('x* n est pas un minimum local de f');
22
23 end
```

```
1 x_0 = [-1; -2];
2 x_star = [1; 1];
3
4 x = []; % liste des abscisses des points du chemin parcouru
5 y = []; % liste des ordonn es des points du chemin parcouru
6 errors = []; % liste des normes de l'erreur
7
8 x_i = x_0;
```

```
9 for i=1:5
10
      x1 = x_i(1); \% abscisse
      x2 = x_i(2); % ordonn e
11
12
      H_x = double(subs(H)); % Hessienne
13
      G_x = double(subs(G)); % Gradient
14
15
      L = chol(H_x); % d composition de Choleski : <math>H_x = LL'
16
17
      s_i = L(L'(-G_x)); % on r soud LL' s_i = -G_x
18
19
      x_i = x_i + s_i; % calcul de l'it ration
20
21
22
      % on sauvegarde le chemin parcouru
23
      x = [x, x1];
24
      y = [y, x2];
25
      % on sauvegarde l'erreur
26
27
       errors = [errors, norm(x_i - x_star)];
  end
28
29
30 % courbes de niveau + it r s
31 ezcontour(f, [-1.5;2;-3;3]);
32 hold on;
33 plot(x, y, '-*');
35 % courbe de l'erreur en fct du nb d'it ration
36 figure();
37 plot(errors);
```

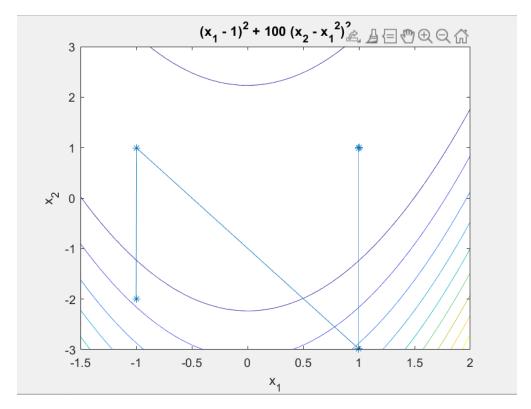


Figure 1: Affichage des itérés sur le graphe des lignes de niveau de la fonction f

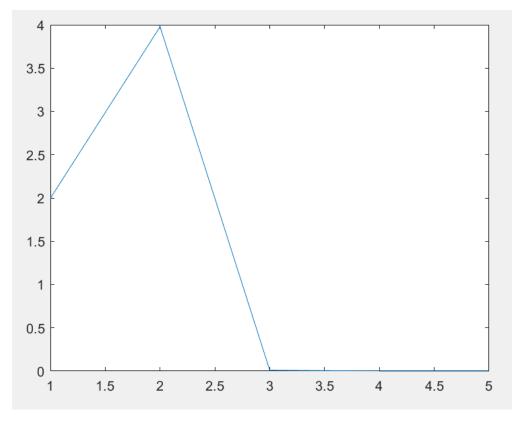
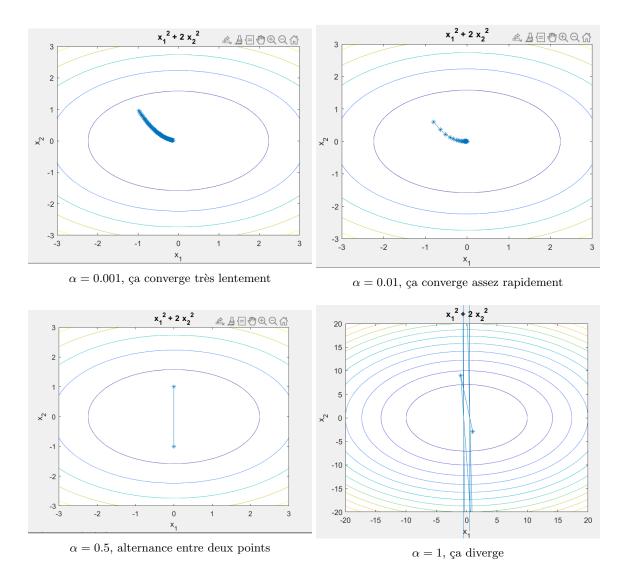


Figure 2: Affichage de la courbe de la norme de l'erreur en fonction des itérations

Exercice 7 - Méthode de gradient à pas optimal et méthode de Wolfe



Tests avec plusieurs pas de descente

```
syms x1 x2; % variables symboliques

f = x1^2 + 2*x2^2; % fonction minimiser

nb_iter = 100;
x_0 = [-1;1]; % point initial
alpha = 1; % pas

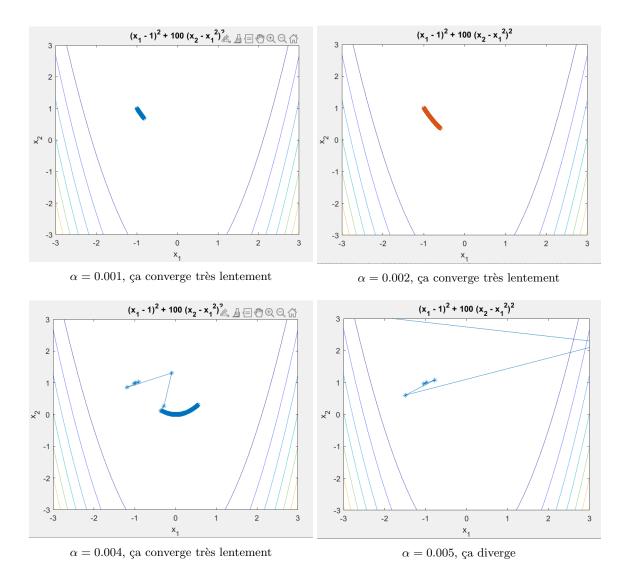
[x_star, path] = gd(gradient(f),nb_iter,[-1;1],alpha); %descente de gradient
ezcontour(f, [-20;20;-20;20]); % lignes de niveau

hold on;
plot(path(1,:), path(2,:), '-*'); % on trace le chemin parcouru

symboliques

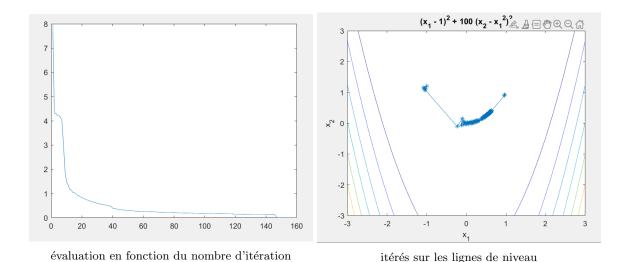
plot(path(1,:), path(2,:), '-*'); % on trace le chemin parcouru
```

```
function [x_i, path] = gd(df,nb_iter,x_0,alpha)
14
15
            x_i = x_0; % point initial
16
17
            path = [];
18
            for i = 1:nb_iter
19
20
                    x1 = x_i(1); % abscisse
21
                    x2 = x_i(2); % ordonn e
22
23
                    \label{eq:G} \begin{array}{lll} \texttt{G} & = & \texttt{double(subs(df));} & \texttt{\%} & \texttt{calcul} & \texttt{du} & \texttt{Gradient} \\ \texttt{x\_i} & = & \texttt{x\_i} & - & \texttt{alpha*G;} & \texttt{\%} & \texttt{calcul} & \texttt{de} & \texttt{l'it} & \texttt{ration} \end{array}
24
25
26
                    path = [path, x_i]; % chemin parcouru
27
28
29
30
    end
```



Tests avec plusieurs pas de descente

```
1 syms x1 x2; % variables symboliques
3 \text{ rosenbrock} = 100*(x2 - x1^2)^2 + (1 - x1)^2; \% \text{ fonction}
                                                                 minimiser
4 nb_iter = 200;
x_0 = [-1; 1.2]; \% point initial
6 alpha = 0.004; % pas
8 [x_star, path] = gd(gradient(rosenbrock), nb_iter,[-1;1], alpha); %descente de gradient
9 ezcontour(rosenbrock, [-3;3;-3;3]); % lignes de niveau
10
11 hold on;
plot(path(1,:), path(2,:), '-*'); % on trace le chemin parcouru
13
  function [x_i, path] = gd(df,nb_iter,x_0,alpha)
14
15
      x_i = x_0; % point initial
16
      path = [];
17
18
19
      for i = 1:nb_iter
20
21
           x1 = x_i(1); \% abscisse
           x2 = x_i(2); % ordonn e
22
23
           G = double(subs(df)); % calcul du Gradient
24
           x_i = x_i - alpha*G; % calcul de l'it ration
25
26
           path = [path, x_i]; % chemin parcouru
27
28
29
30
  end
```



On atteint le minimum en 147 itérations

```
syms x1 x2;
rosenbrock = 100*(x2 - x1^2)^2 + (1 - x1)^2;
f = x1^2 + 2*x2^2;

[xstar, path, iter, f_value] = gd_opt(rosenbrock, gradient(rosenbrock), [-1;1.2], 160);
ezcontour(rosenbrock, [-3;3;-3;3]);
hold on;
plot(path(1, :), path(2, :), '-*');
```

```
11
12 figure();
plot(iter,f_value);
15 function [x_i, path, iter, f_value] = gd_opt(f, df, x_0, nb_iter)
      x_i = x_0;
path = [];
16
17
       iter = [];
18
       f_value = [];
19
20
21
       for i = 1:nb_iter
           iter = [iter, i];
22
23
24
           x1 = x_i(1);
           x2 = x_i(2);
25
26
           d = -double(subs(df));
27
28
29
           tg = 0; td = Inf; t = 1; m1 = 0.1; m2 = 0.9;
30
31
           g0 = double(subs(f));
32
33
           f_value = [f_value, g0];
34
35
           gp0 = double(d.*subs(df));
36
           while abs(tg-td)>1e-3
37
               path = [path, x_i];
38
39
                x = x_i + t*d;
40
                x1 = x(1);
41
               x2 = x(2);
42
43
               gt = double(subs(f));
44
                gpt = double(subs(df).*d);
45
46
                if(gt <= (g0 + m1*t*gp0))</pre>
47
48
                    if (gpt >= m2*gp0)
49
                         break;
50
                         tg = t;
51
52
53
                        if(td == Inf)
                            t = 10*tg;
54
55
                         else
                             t = (td + tg)/2;
56
                         end
57
58
                    end
                else
59
60
                    td = t;
61
                    if(td == Inf)
62
                        t = 10*tg;
63
                    else
64
                         t = (td + tg)/2;
65
                    end
66
67
                end
           end
68
           x_i = x_i + t*d;
69
70
71 end
```