

【Experiment name】

Conjugate gradient method to solve Rosen brock function

【Purpose】

1. Master the solution of Rosenbrock function by FR conjugate gradient method and PRP conjugate gradient method respectively.
2. Master and call the precise one - dimensional search golden section method

【Experimental principle】

F R Conjugate Gradient Algorithm

算法2 FR 共轭梯度法
给定控制误差 ε .

step1 给定初始点 \mathbf{x}_0 及初始下降方向 $\mathbf{p}_0 = -\mathbf{g}_0$, 令 $k = 0$.

step2 做精确一维搜索, 求步长 α_k

$$f(\mathbf{x}_k + \alpha_k \mathbf{p}_k) = \min_{\alpha \geq 0} f(\mathbf{x}_k + \alpha \mathbf{p}_k)$$

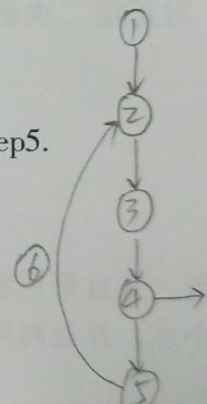
step3 令 $\mathbf{x}_{k+1} = \mathbf{x}_k + \alpha_k \mathbf{p}_k$. 求得 \mathbf{g}_{k+1}

step4 若 $\|\mathbf{g}_{k+1}\| \leq \varepsilon$, 则 $\mathbf{x}^* = \mathbf{x}_{k+1}$, 停止计算; 否则, 转 step5.

step5 取搜索方向 \mathbf{p}_{k+1} 为

$$\mathbf{p}_{k+1} = -\mathbf{g}_{k+1} + \beta_k \mathbf{p}_k$$
$$\beta_k = \frac{\mathbf{g}_{k+1}^T \mathbf{g}_{k+1}}{\mathbf{g}_k^T \mathbf{g}_k}$$

step6 令 $k = k + 1$, 转 step2.



PRP Conjugate Gradient Method Algorithm

算法3 PRP 共轭梯度法

在算法2的 step 5 中，用 PRP 公式

$$\beta_k = \frac{\mathbf{g}_{k+1}^T (\mathbf{g}_{k+1} - \mathbf{g}_k)}{\mathbf{g}_k^T \mathbf{g}_k}$$

代替 FR 公式，就得到 PRP 共轭梯度法。

对于正定的二次函数，FR 共轭梯度法与 PRP 共轭梯度法等价。但是对于一般函数，二者是不同的，因为对于目标函数的 Hesse 阵不是常数矩阵，所以迭代过程中所产生的方向不再是共轭方向了。不过若采用精确一维搜索，两个算法所产生的搜索方向都满足

$$\mathbf{p}_k^T \mathbf{g}_k = (-\mathbf{g}_k + \beta_{k-1} \mathbf{p}_{k-1})^T \mathbf{g}_k = -\mathbf{g}_k^T \mathbf{g}_k < 0$$

故两者都是下降算法。

PRP 算法与 FR 算法都是常用的共轭梯度法。从一些实际计算的结果发现，PRP 算法一般优于 FR 算法。

Example 1:

例3 用 FR 共轭梯度法求解 Rosenbrock 函数

$$\min f(\mathbf{x}) = 100(x_2 - x_1^2)^2 + (1 - x_1)^2$$

问题有唯一极小点，精确解为 $\mathbf{x}^* = (1, 1)^T$, $f(\mathbf{x}^*) = 0$.

Example 2:

例5 利用 PRP 共轭梯度法求解 Rosenbrock 函数

$$\min f(\mathbf{x}) = 100(x_2 - x_1^2)^2 + (1 - x_1)^2$$

问题有唯一极小点，精确解为 $\mathbf{x}^* = (1, 1)^T$, $f(\mathbf{x}^*) = 0$.

【Experimental content】

I use the matlab software to write the experimental code, and successfully demonstrate the algorithm process of the two conjugate gradient methods and print them.

FR method experimental results: (epsilon takes 1 e-6)

The golden section takes the interval [0,5]				
k	x1	x2	f(x1,x2)	gk ^2
1	0.999932	1.000000	0.000002	0.061098
2	0.999986	0.999973	0.000000	0.000121
3	0.999987	0.999973	0.000000	0.000012
4	0.999987	0.999973	0.000000	0.000052
5	0.999998	0.999995	0.000000	0.000249
6	1.000001	1.000001	0.000000	0.000010
7	1.000001	1.000001	0.000000	0.000001

The golden section takes the interval [0,0.1]				
k	x1	x2	f(x1,x2)	gk ^2
1	0.999932	1.000000	0.000002	0.061098
2	0.999936	0.999872	0.000000	0.000139
3	0.999936	0.999873	0.000000	0.000082
4	0.999941	0.999883	0.000000	0.000197
5	0.999943	0.999885	0.000000	0.000264
6	0.999944	0.999887	0.000000	0.000529
7	1.000001	1.000001	0.000000	0.000341
8	1.000000	1.000000	0.000000	0.000004
9	1.000000	1.000000	0.000000	0.000004
10	1.000000	1.000000	0.000000	0.000000

P RP experiment results: (epsilon takes 1 e-6)

The golden section takes the interval [0,5]				
k	x1	x2	f(x1,x2)	gk ^2
1	0.999932	1.000000	0.000002	0.061098
2	0.999936	0.999872	0.000000	0.000150
3	0.999936	0.999872	0.000000	0.000057
4	0.999990	0.999979	0.000000	0.001059
5	1.000000	1.000000	0.000000	0.000006
6	1.000000	1.000000	0.000000	0.000006
7	1.000000	1.000000	0.000000	0.000000

The golden section takes the interval [0,0.1]				
k	x1	x2	f(x1,x2)	gk ^2
1	0.999932	1.000000	0.000002	0.061098
2	0.999936	0.999872	0.000000	0.000139
3	0.999936	0.999873	0.000000	0.000082
4	0.999941	0.999883	0.000000	0.000197
5	0.999943	0.999885	0.000000	0.000264
6	0.999944	0.999887	0.000000	0.000529
7	1.000001	1.000001	0.000000	0.000341
8	1.000000	1.000000	0.000000	0.000004
9	1.000000	1.000000	0.000000	0.000004
10	1.000000	1.000000	0.000000	0.000000

Other function test cases (eg: $f=(1/2)*x_1^2+(9/2)*x_2^2$; $x_0=[9 \ 1]$;

FR method: (epsilon takes 1 e-4)

The golden section takes the interval [0,0.1]				
k	x1	x2	f(x1,x2)	gk ^2
1	7.199976	-0.800024	28.800000	10.182475

2	5.904438	-0.656071	19.368123	8.350277
3	4.442956	0.030986	9.874249	4.451699
4	3.583445	0.198382	6.597640	4.003607
5	2.530044	0.155297	3.309089	2.890434
6	1.728076	-0.006876	1.493336	1.729184
7	1.268311	-0.058731	0.819828	1.374047
8	0.851220	-0.038635	0.369004	0.919500
9	0.579349	0.005123	0.167941	0.581181
10	0.412822	0.017995	0.086668	0.443455
11	0.274602	0.009300	0.038092	0.287074
12	0.189228	-0.002711	0.017937	0.190794
13	0.132601	-0.005577	0.008931	0.141784
14	0.088074	-0.002142	0.003899	0.090160
15	0.061265	0.001174	0.001883	0.062169
16	0.042394	0.001695	0.000912	0.045054
17	0.028245	0.000443	0.000400	0.028526
18	0.019750	-0.000457	0.000196	0.020174
19	0.013527	-0.000496	0.000093	0.014245
20	0.009072	-0.000069	0.000041	0.009093
21	0.006349	0.000167	0.000020	0.006525
22	0.004313	0.000138	0.000009	0.004489
23	0.002918	0.000000	0.000004	0.002918
24	0.002037	-0.000058	0.000002	0.002103
25	0.001375	-0.000036	0.000001	0.001414
26	0.000939	0.000006	0.000000	0.000941
27	0.000652	0.000019	0.000000	0.000675
28	0.000439	0.000009	0.000000	0.000446
29	0.000302	-0.000004	0.000000	0.000304
30	0.000208	-0.000006	0.000000	0.000216
31	0.000140	-0.000002	0.000000	0.000141
32	0.000097	0.000002	0.000000	0.000098

PRP method: (epsilon takes 1 e-4)

The golden section takes the interval [0,0.3]				
k	x1	x2	f(x1,x2)	gk ^2
1	7.199976	-0.800024	28.800000	10.182475
2	3.312444	-0.368102	6.095891	4.684841
3	3.300493	0.011568	5.447229	3.302135
4	2.310082	-0.006886	2.668452	2.310913
5	1.824031	0.015558	1.664633	1.829397
6	1.355016	-0.030050	0.922097	1.381741
7	1.022075	0.058221	0.537572	1.148565

8	0.702244	-0.095441	0.287564	1.109490
9	0.258053	0.050006	0.044548	0.518784
10	0.047801	-0.016743	0.002404	0.158088
11	-0.007284	-0.003326	0.000076	0.030810
12	0.000002	-0.000000	0.000000	0.000004

Attachment: Experimental code (4 files in total)

```
test.m
% Conjugate gradient method test area

% define several functions
syms x1 x2;
x={x1,x2};
% function f
f = 100 * (x2 - x1^2)^2 + (1 - x1)^2;
% initial point x0
x0=[-1 1];
% set epsilon
e=1e-6;

% function call
[result_x] = FR(f, x, x0, e); % FR method test
[result_x] = PRP(f, x, x0, e); % PRP method test

% other test data
%f=(3/2)*x1^2+(1/2)*x2^2-x1*x2-2*x1;
%x0=[0 0];
%f=(1/2)*x1^2+(9/2)*x2^2;
%x0=[9 1];
```

```
gold.m
function [result] = gold(f, a, b)
% golden section method
%Input f: function a, b: [a, b] operation interval

% Default epsilon=1e-4 (can be adjusted below)
syms x x2 f1 f2 xx ah;
flag=1; ratio=0.618; e=1e-4;%epsilon
for i = 1:100
%step1:
if flag == 1
x1 = a + (1-ratio) * (b - a);
f1 = subs(f, ah, x1);
x2 = a + ratio * (b - a);
```

```

f2 = subs(f, ah, x2);

flag = 2;
end
%step2
if flag == 2
if (abs(b - a) <= e)
xx = (a + b) / 2;
break;
else
flag = 3;
end
end

%step3
if flag == 3
if f1 < f2
b = x2;
x2 = x1;
f2 = f1;
x1 = a + 0.382 * (b - a);
f1 = subs(f, ah, x1);

flag = 2; % flag update

elseif f1 == f2
a = x1;
b = x2;

flag = 1; % flag update

elseif f1 > f2
a = x1;
x1 = x2;
f1 = f2;
x2 = a + 0.618 * (b - a);
f2 = subs(f, ah, x2);

flag = 2; % flag update
end
end
end
result = xx;
end

```

F _

```
function [result_x] = FR(f,x,x0,e)
% Using the "FR conjugate method" of the exact one-dimensional search golden section
method
% Input f: complete function x: independent variable set x0: initial point e: epsilon
% output result_x: exact solution approximation result x*

% Define the alpha used in the step size; k is used for counting
syms ah;
k=0;
% Create an n-dimensional cell array (only x1, x2 in the example, ie n=2)
n = length(x);
gx = cell(1, n);
for i = 1 : n
    gx{i} = diff(f, x{i});
end
% data initialization
g = subs(gx, x, x0); % partial derivative solution g(x)
g_pre = g;
xx = x0; % Substitute xx for x(i+1) and x_pre for x(i) in the later stage
x_pre=x0;
p = - g; % p0 reverse direction
% first iteration
xx = xx+ah*p;
phi = subs(f, x, xx);
nphi = double(gold(phi, 0, 2));
aa = nphi;
xx = double(x_pre+aa*p);
g = subs(gx, x, xx);
% Data output format [k x1 x2 f(x1,x2) ||gk||^2]
fprintf('k\tx1\tx2\tf(x1,x2)\t||gk||^2\n');

while (norm(g) > e)
% FR core algorithm step5
beta=(sumsqr(g)/sumsqr(g_pre));
p= - g+beta*p;

% Temporary data update
g_pre=g;
x_pre=xx;
k=k+1;
% data output
ff=double(subs(f, x, xx)); % f(x1,x2)
```

```

g2=double(norm(g)); % ||gk||^2
fprintf('%d\t\t',k);fprintf('%f\t',xx);fprintf('%f\t%f\n',ff, g2);

% (k-1)th iteration
xx = xx+ah*p;
phi = subs(f, x, xx);
nphi = double(gold(phi, 0, 5)); % golden section search 5
%nphi = double(gold(phi, 0, 0.1)); % Golden section search 0.1
aa = nphi;
xx = double(x_pre+aa*p);
g = subs(gx, x, xx);

end

% Last line of data output
ff=double(subs(f, x, xx));
g2=double(norm(g));
fprintf('%d\t\t',k+1);fprintf('%f\t',xx);fprintf('%f\t%f\n',ff, g2);
%The function returns the solution result x*
result_x=xx;
end

```

P RP.m

```

function [result_x] = PRP(f,x,x0,e)
% Using the "FR conjugate method" of the exact one-dimensional search golden section
method
% Input f: complete function x: independent variable set x0: initial point e: epsilon
% output result_x: exact solution approximation result x*

% Define the alpha used in the step size; k is used for counting
syms ah;
k=0;
% Create an n-dimensional cell array (only x1, x2 in the example, ie n=2)
n = length(x);
gx = cell(1, n);
for i = 1 : n
gx{i} = diff(f, x{i});
end
% data initialization
g = subs(gx, x, x0); % partial derivative solution g(x)
g_pre = g;
xx = x0; % Substitute xx for x(i+1) and x_pre for x(i) in the later stage
x_pre=x0;
p = - g; % p0 reverse direction

```



```
% first iteration
xx = xx+ah*p;
phi = subs(f, x, xx);
nphi = double(gold(phi, 0, 2));
aa = nphi;
xx = double(x_pre+aa*p);
g = subs(gx, x, xx);
% Data output format [k x1 x2 f(x1,x2) ||gk||^2]
fprintf('k\tx1\tx2\tf(x1,x2)\t||gk||^2\n');

while (norm(g) > e)
% PRP core algorithm step5
g1=double((g-g_pre)*(g'));
g2=double(g_pre*(g_pre'));
beta=(g1/g2);
p= - g+beta*p;

% Temporary data update
g_pre=g;
x_pre=xx;
k=k+1;

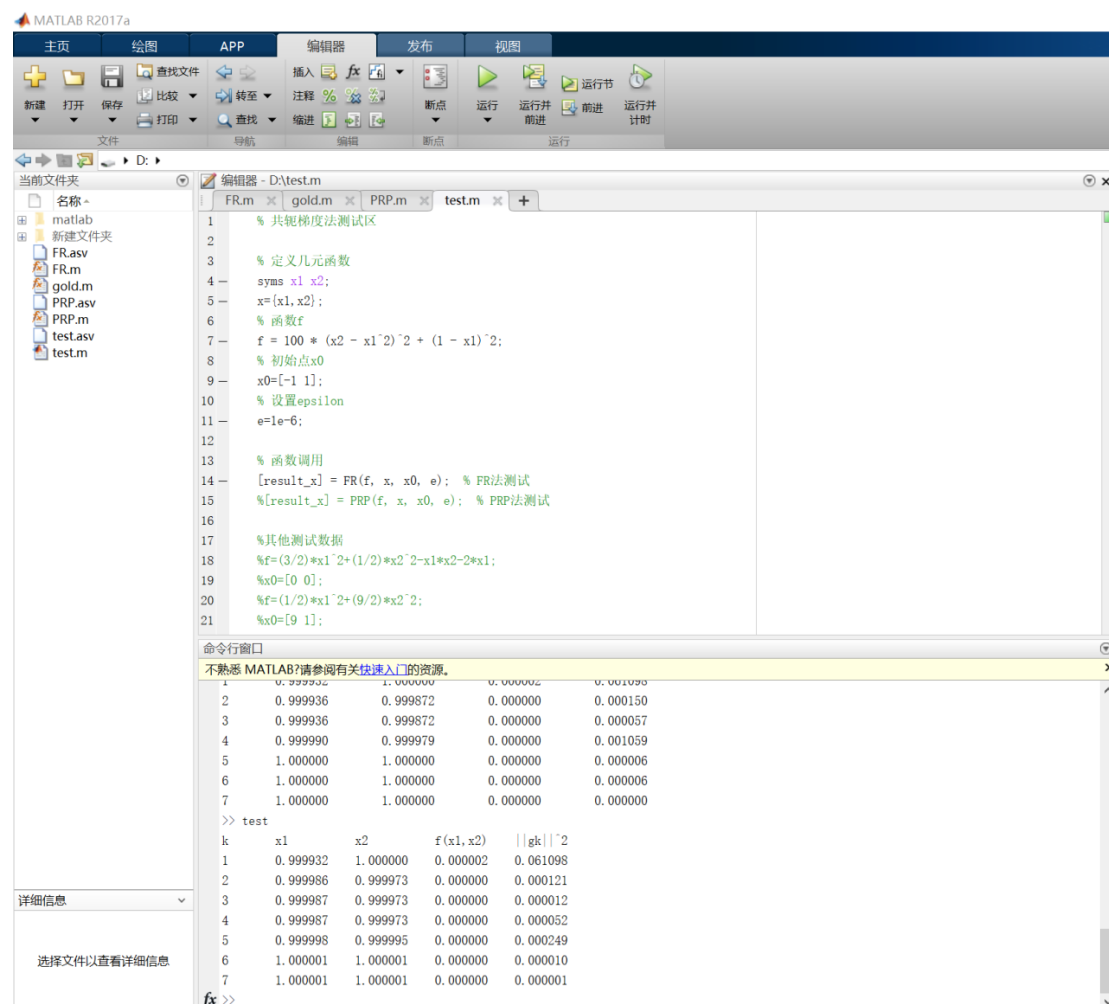
% data output
ff=double(subs(f, x, xx)); % f(x1,x2)
g2=double(norm(g)); % ||gk||^2
fprintf('%d\t\t',k);fprintf('%f\t',xx);fprintf('%f\t%f\n',ff, g2);

% (k-1)th iteration
xx = xx+ah*p;
phi = subs(f, x, xx);
nphi = double(gold(phi, 0, 5)); % golden section search
%nphi = double(gold(phi, 0, 0.1)); % golden section search
aa = nphi;
xx = double(x_pre+aa*p);
g = subs(gx, x, xx);

end

% Last line of data output
ff=double(subs(f, x, xx));
g2=double(norm(g));
fprintf('%d\t\t',k+1);fprintf('%f\t',xx);fprintf('%f\t%f\n',ff, g2);
%The function returns the solution result x*
result_x=xx;
end
```

Matlab experiment interface;



The image shows the MATLAB R2017a interface. The main window displays a script named 'test.m' with the following code:

```
1 % 共轭梯度法测试区
2
3 % 定义二元函数
4 syms x1 x2;
5 x=[x1,x2];
6 % 函数f
7 f = 100 * (x2 - x1^2)^2 + (1 - x1)^2;
8 % 初始点x0
9 x0=[-1 1];
10 % 设置epsilon
11 e=1e-6;
12
13 % 函数调用
14 [result_x] = FR(f, x, x0, e); % FR法测试
15 [result_x] = PRP(f, x, x0, e); % PRP法测试
16
17 %其他测试数据
18 %f=(3/2)*x1^2+(1/2)*x2^2-x1*x2-2*x1;
19 %x0=[0 0];
20 %f=(1/2)*x1^2+(9/2)*x2^2;
21 %x0=[9 1];
```

The command window shows the output of the script, including a table of results for the FR and PRP methods.

k	x1	x2	f(x1,x2)	gk ^2
1	0.999932	1.000000	0.000002	0.061098
2	0.999936	0.999872	0.000000	0.000150
3	0.999936	0.999872	0.000000	0.000057
4	0.999990	0.999979	0.000000	0.001059
5	1.000000	1.000000	0.000000	0.000006
6	1.000000	1.000000	0.000000	0.000006
7	1.000000	1.000000	0.000000	0.000000

【Experiment Summary】

Through this experiment, I have a better understanding of the use of the FR conjugate gradient method and the PRP conjugate gradient method in the unconstrained optimization numerical algorithm. I have clarified the algorithm idea by writing the algorithm code, and I have observed and understood the numerical optimization through specific experiments. The process, and the golden section one-dimensional search algorithm is well applied to it, which deepens the learning and understanding of our "Optimization Method" course.