
Homework 1

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Problem 2

```
%%% a) With eps = 1, gamma = 1 a minimum is not found. Instead, NaN's
      are returned

X=[1 0.2;0.1 1];
C=[0.1;0.5];

eps=1; % stopping criterion for gradient algorithm
gamma=1; % step for gradient algorithm
M=1000000;
[a_opt1] = gradient_test1(X,C,eps,gamma,M);

%%% b) eps = 0.1, gamma = 0.1 a minimum is found, but not the
      minimum. This is due
      % to eps being too small

eps=0.1; % stopping criterion for gradient algorithm
gamma=0.1; % step for gradient algorithm
[a_opt2] = gradient_test1(X,C,eps,gamma,M);

%%% c) eps = 0.001, gamma = 0.1 results in a solution that is closer
      to the actual
      % minimum

eps=0.001; % stopping criterion for gradient algorithm
gamma=0.1; % step for gradient algorithm
[a_opt3] = gradient_test1(X,C,eps,gamma,M);

%%% e) eps = 0.01, gamma = 0.01 results in a solution that is further
      from the
      % minimum than was found in the previous step. This shows that perhaps
      gamma
      % isn't the parameter that should be changed (in this case).

eps=0.001; % stopping criterion for gradient algorithm
gamma=0.01; % step for gradient algorithm
[a_opt4] = gradient_test1(X,C,eps,gamma,M);
```

Note

With each subsequent reduction of the size of ϵ , the number of steps required to be within that bound increases.

```
%%% e)  $\epsilon = 0.001$ ,  $\gamma = 0.1$  results in a solution that is very  
close to the  
% actual minimum.
```

```
 $\epsilon = 0.0001$ ; % stopping criterion for gradient algorithm  
 $\gamma = 0.1$ ; % step for gradient algorithm  
[a_opt5] = gradient_test1(X,C, $\epsilon$ , $\gamma$ ,M);
```

Problem 3

A solution that is reasonably close to the minimum can be obtained by having $\epsilon = 1e-9$. This choice of ϵ was determined by iteratively decreasing ϵ until a minimum that was within $1e-9$ of the actual minimum was reached.

```
 $\epsilon = 1e-9$ ; % stopping criterion for gradient algorithm  
 $\gamma = 0.1$ ; % step for gradient algorithm  
[a_opt6] = gradient_test1(X,C, $\epsilon$ , $\gamma$ ,M);
```

Problem 4

Adjusting the stopping criterion to be $f(a_{i+1}) - f(a_i) < \epsilon$ resulted in worse performance than in the previous problems. As shown below, γ had to be adjusted and the number of steps required to reach a minimum increased approximately 3 orders of magnitude.

```
 $\epsilon = 1e-9$ ; % stopping criterion for gradient algorithm  
 $\gamma = 1e-5$ ; % step for gradient algorithm  
[a_opt7] = gradient_test1(X,C, $\epsilon$ , $\gamma$ ,M);
```

Problem 5

Using the steepest descent algorithm results in a much quicker convergence and with greater accuracy than using a static step-size.

```
 $\epsilon = 1e-9$ ; % stopping criterion for gradient algorithm  
[a_opt8] = gradient_test2(X,C, $\epsilon$ ,M);
```

```
a_opt =
```

```
NaN  
NaN
```

```
steps =
```

```
1000001
```

```
a_opt =  
    0.0300  
    0.1040
```

```
steps =  
    3
```

```
a_opt =  
    0.0049  
    0.4951
```

```
steps =  
    27
```

```
a_opt =  
    0.0345  
    0.4456
```

```
steps =  
    119
```

```
a_opt =  
    0.0005  
    0.4995
```

```
steps =  
    42
```

```
a_opt =  
    0.0000  
    0.5000
```

```
steps =  
    116
```

```
a_opt =
```

```
0.0000
```

```
0.5000
```

```
steps =
```

```
587616
```

```
a_opt =
```

```
0.0000
```

```
0.5000
```

```
steps =
```

```
17
```

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