

THE UNIVERSITY OF TEXAS AT AUSTIN

EE381V LARGE SCALE OPTIMIZATION

Problem Set 2

Edited by \LaTeX

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1 Matlab and Computational Assignment

1.1 Five flavors for Eq. (9.20) in B & V

1.1.1 Standard Gradient Descent with Backtracking

Command to be executed in matlab:

```
>> x_init = [1 1]'; alpha = 0.3; bta = 0.8;
>> [x, iter, all_costs] = gd_btls(x_init, @func, @func_grad, alpha, bta);
```

Dump

Minima

```
x = [-0.3379, -0.0031], obj = 2.559267
```

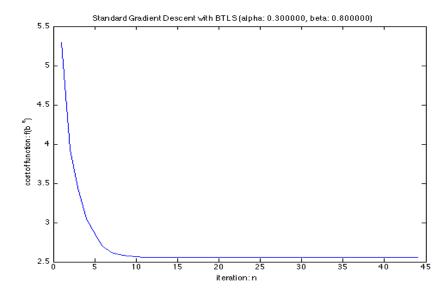


Figure 1: Standard gradient descent with BTLS on Eq. 9.20 with $\alpha = 0.3$ and $\beta = 0.8$

1.1.2 Steepest Descent with P_1

Command to be executed in matlab:

```
>> P1 = [8 0; 0 2];
>> x_init = [1 1]'; alpha = 0.3; bta = 0.8;
>> [x, iter, all_costs] = sd_btls(x_init, @func, @func_grad, P1, alpha, bta);
```

Dump

```
Iter: 1, Cost: 4.109800e+00, Conv_Rate: 0.082430, gamma: 0.035184
Iter: 2, Cost: 2.574134e+00, Conv_Rate: 0.626340, gamma: 0.409600
Iter: 3, Cost: 2.561394e+00, Conv_Rate: 0.995051, gamma: 1.000000
Iter: 4, Cost: 2.559319e+00, Conv_Rate: 0.999190, gamma: 0.800000
Iter: 5, Cost: 2.559268e+00, Conv_Rate: 0.999980, gamma: 0.800000
Iter: 6, Cost: 2.559267e+00, Conv_Rate: 1.000000, gamma: 0.800000
Iter: 7, Cost: 2.559267e+00, Conv_Rate: 1.000000, gamma: 0.800000
Iter: 8, Cost: 2.559267e+00, Conv_Rate: 1.000000, gamma: 0.800000
Iter: 9, Cost: 2.559267e+00, Conv_Rate: 1.000000, gamma: 0.800000
Iter: 10, Cost: 2.559267e+00, Conv_Rate: 1.000000, gamma: 0.800000
Iter: 11, Cost: 2.559267e+00, Conv_Rate: 1.000000, gamma: 0.800000
Iter: 12, Cost: 2.559267e+00, Conv_Rate: 1.000000, gamma: 0.800000
Iter: 12, Cost: 2.559267e+00, Conv_Rate: 1.000000, gamma: 0.800000
```

Minima

```
x = [-0.3466 - 0.0000], obj = 2.559267
```

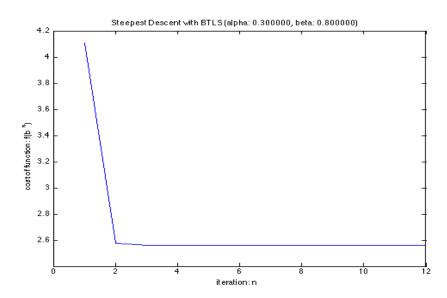


Figure 2: Steepest Descent with BTLS on Eq. 9.20 with P_1 , $\alpha = 0.3$ and $\beta = 0.8$

1.1.3 Steepest Descent with P_2

Command to be executed in matlab:

```
>> P2 = [2 0; 0 8];
>> x_init = [1 1]'; alpha = 0.3; bta = 0.8;
>> [x, iter, all_costs] = sd_btls(x_init, @func, @func_grad, P2, alpha, bta);
```

Dump

```
Iter: 1, Cost: 5.397521e+00, Conv_Rate: 0.108258, gamma: 0.011529
Iter: 2, Cost: 4.867283e+00, Conv_Rate: 0.901763, gamma: 0.068719
Iter: 3, Cost: 4.673428e+00, Conv_Rate: 0.960172, gamma: 0.107374
Iter: 4, Cost: 4.447617e+00, Conv_Rate: 0.951682, gamma: 0.085899
Iter: 5, Cost: 4.276423e+00, Conv_Rate: 0.961509, gamma: 0.134218
Iter: 6, Cost: 4.100684e+00, Conv_Rate: 0.958905, gamma: 0.107374
Iter: 7, Cost: 3.955941e+00, Conv_Rate: 0.964703, gamma: 0.134218
Iter: 8, Cost: 3.820349e+00, Conv_Rate: 0.965724, gamma: 0.134218
Iter: 9, Cost: 3.687908e+00, Conv_Rate: 0.965333, gamma: 0.134218
Iter: 10, Cost: 3.560463e+00, Conv_Rate: 0.965442, gamma: 0.134218
Iter: 11, Cost: 3.444372e+00, Conv_Rate: 0.967394, gamma: 0.134218
Iter: 12, Cost: 3.347890e+00, Conv_Rate: 0.971989, gamma: 0.167772
Iter: 13, Cost: 3.259404e+00, Conv_Rate: 0.973570, gamma: 0.167772
Iter: 164, Cost: 2.559267e+00, Conv_Rate: 1.000000, gamma: 0.262144
Iter: 165, Cost: 2.559267e+00, Conv_Rate: 1.000000, gamma: 0.409600
Iter: 166, Cost: 2.559267e+00, Conv_Rate: 1.000000, gamma: 0.327680
Iter: 167, Cost: 2.559267e+00, Conv_Rate: 1.000000, gamma: 0.262144
```

Iter: 168, Cost: 2.559267e+00, Conv_Rate: 1.000000, gamma: 0.167772

Minima

Convergence reached!

```
x = [-0.3466 - 0.0000], obj = 2.559267
```

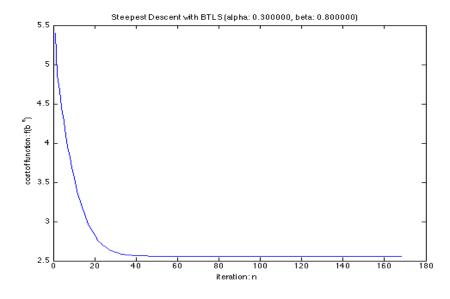


Figure 3: Steepest Descent with BTLS on Eq. 9.20 with P_2 , $\alpha = 0.3$ and $\beta = 0.8$

1.1.4 Cyclic Coordinate Descent

Command to be executed in matlab:

```
>> x_init = [1 1]'; alpha = 0.3; bta = 0.8;
>> [x, iter, all_costs] = ccd_btls(x_init, @func, @func_grad, alpha, bta);
```

Dump

```
Iter: 1, Cost: 1.217687e+01, Conv_Rate: 0.244232, gamma: 0.043980
Iter: 2, Cost: 3.551842e+00, Conv_Rate: 0.071239, gamma: 0.035184
Iter: 3, Cost: 3.064142e+00, Conv_Rate: 0.862691, gamma: 0.209715
Iter: 4, Cost: 2.732279e+00, Conv_Rate: 0.769257, gamma: 0.134218
Iter: 5, Cost: 2.664867e+00, Conv_Rate: 0.975328, gamma: 0.209715
Iter: 6, Cost: 2.600923e+00, Conv_Rate: 0.951925, gamma: 0.134218
Iter: 7, Cost: 2.585633e+00, Conv_Rate: 0.994121, gamma: 0.262144
Iter: 8, Cost: 2.563802e+00, Conv_Rate: 0.985727, gamma: 0.107374
Iter: 9, Cost: 2.561867e+00, Conv_Rate: 0.999245, gamma: 0.209715
Iter: 10, Cost: 2.560171e+00, Conv_Rate: 0.998584, gamma: 0.107374
Iter: 11, Cost: 2.559836e+00, Conv_Rate: 0.999869, gamma: 0.167772
Iter: 12, Cost: 2.559551e+00, Conv_Rate: 0.999758, gamma: 0.107374
Iter: 13, Cost: 2.559478e+00, Conv_Rate: 0.999971, gamma: 0.167772
Iter: 60, Cost: 2.559267e+00, Conv_Rate: 1.000000, gamma: 0.134218
Iter: 61, Cost: 2.559267e+00, Conv_Rate: 1.000000, gamma: 0.262144
Iter: 62, Cost: 2.559267e+00, Conv_Rate: 1.000000, gamma: 0.085899
Iter: 63, Cost: 2.559267e+00, Conv_Rate: 1.000000, gamma: 0.327680
Iter: 64, Cost: 2.559267e+00, Conv_Rate: 1.000000, gamma: 0.134218
Convergence reached!
```

Minima

```
x = [-0.3466 \ 0.0000], obj = 2.559267
```

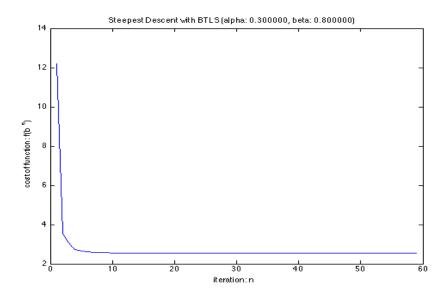


Figure 4: Cyclical Coordinate Descent with BTLS on Eq. 9.20 with $\alpha = 0.3$ and $\beta = 0.8$

1.1.5 Greedy Coordinate Descent

Command to be executed in matlab:

```
>> x_init = [1 1]'; alpha = 0.3; bta = 0.8;
>> [x, iter, all_costs] = gcd_btls(x_init, @func, @func_grad, alpha, bta);
```

Dump

```
Iter: 1, Cost: 5.615225e+00, Conv_Rate: 0.112625, gamma: 0.005903
Iter: 2, Cost: 4.227075e+00, Conv_Rate: 0.752788, gamma: 0.028147
Iter: 3, Cost: 3.306101e+00, Conv_Rate: 0.782125, gamma: 0.035184
Iter: 4, Cost: 2.823271e+00, Conv_Rate: 0.853958, gamma: 0.054976
Iter: 5, Cost: 2.666299e+00, Conv_Rate: 0.944401, gamma: 0.068719
Iter: 6, Cost: 2.577867e+00, Conv_Rate: 0.966834, gamma: 0.107374
Iter: 7, Cost: 2.566546e+00, Conv_Rate: 0.995608, gamma: 0.107374
Iter: 8, Cost: 2.561822e+00, Conv_Rate: 0.998160, gamma: 0.107374
Iter: 9, Cost: 2.560570e+00, Conv_Rate: 0.999511, gamma: 0.107374
Iter: 10, Cost: 2.559369e+00, Conv_Rate: 0.999531, gamma: 0.107374
Iter: 11, Cost: 2.559298e+00, Conv_Rate: 0.999972, gamma: 0.107374
Iter: 12, Cost: 2.559277e+00, Conv_Rate: 0.999992, gamma: 0.107374
Iter: 32, Cost: 2.559267e+00, Conv_Rate: 1.000000, gamma: 0.107374
Iter: 33, Cost: 2.559267e+00, Conv_Rate: 1.000000, gamma: 0.107374
Iter: 34, Cost: 2.559267e+00, Conv_Rate: 1.000000, gamma: 0.107374
Iter: 35, Cost: 2.559267e+00, Conv_Rate: 1.000000, gamma: 0.107374
Convergence reached!
```

Minima

```
x = [-0.3466 - 0.0000], obj = 2.559267
```

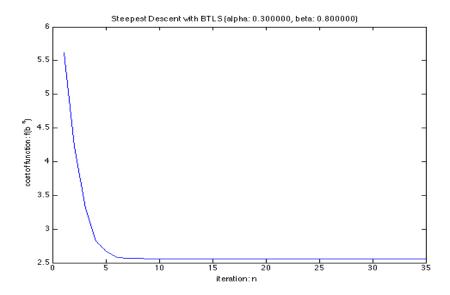


Figure 5: Greedy Coordinate Descent with BTLS on Eq. 9.20 with $\alpha = 0.3$ and $\beta = 0.8$

2 Written Problems

A Codes Printout

A.1 Eq. 20 and its gradient

```
$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ
%%% objective function for optimization
% Usage:
y = func(x)
% Parameter:
                     x: input vector, must be column vector
function y = func(x)
assert (size(x, 1) == 2);
assert (size(x, 2) == 1);
x_1 = x(1);
x_2 = x(2);
term1 = x_1 + 3 * x_2 - 0.1;
term2 = x_1 - 3 * x_2 - 0.1;
term3 = -1 * x_1
                                                                                                                -0.1;
y = \exp(term1) + \exp(term2) + \exp(term3);
end
$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ$\circ
%%% gradient of function EQ. 9.20 in B & V
% Usage:
                   gradient = func_grad (x)
% Parameter:
                    b: variable vector
function gradient = func_grad (x)
assert(all(size(x) == [2 1]))
x_{-1} = x(1);
x_2 = x(2);
grad_1 = exp(x_1+3*x_2-0.1) + exp(x_1-3*x_2-0.1) -1*exp(-1*x_1-0.1);
grad_2 = 3 \times \exp(x_1 + 3 \times x_2 - 0.1) - 3 \times \exp(x_1 - 3 \times x_2 - 0.1);
gradient = [grad_1 grad_2]';
end
```

A.2 Standard Gradient Descent with BackTracking Line Search

```
%%% HW2: Gradient Descent with Backtrack Line Search
% Usage:
    [b, iter, all_costs] = gd_btls (b_init, f, fgrad, alpha, discount)
% Parameters:
    b_init: inital value of variable
   f: objective function
   fgrad: gradient of objective function
   alpha: parameter for evaluating "OK" step size
    discount (beta): discounting factor for not-OK step size
%% Note that the distinguisable period eps = 10e-16
function [b, iter, all_costs] = gd_btls (b_init, f, fgrad, alpha, discount)
assert (discount > 0 && discount < 1);
assert (alpha > 0 \&\& alpha < 0.5);
iter = 1; % iteration count
b = b_init; % variable vector
last_cost = func(b); % value of objective function in most recent iteration
all_costs = []; % all values of objective function, for plotting
while true,
   %% compute essential numerics and do gradient descent
   gradient = fgrad(b);
   delta_b = -1.0 * gradient / norm(gradient);
   %% do backtrack line search
   gamma = 1.0; % step size
   while f(b+gamma*delta_b) > f(b) + alpha*gamma*gradient'*delta_b,
       gamma = discount * gamma;
       % disp(sprintf('BTLS: new gamma is %f', gamma));
   end
   b = b + gamma * delta_b;
   cost = func(b);
   rate = (cost / last_cost);
   all_costs = [all_costs cost];
   %% output numeric information of this iteration
   disp (sprintf('Iter: %d, Cost: %e, Conv_Rate: %f, gamma: %f',iter,cost,rate,gamma));
   %% quadratic optimization converges to zero
   if abs((cost - last_cost) / last_cost) < eps,</pre>
       disp('Convergence reached!')
       break
   end
   %% prepare for next iteration
   last_cost = cost;
   iter = iter + 1;
%% uncomment following code for plotting individual gradient descent run
% plot f(b^(n)) with regard to n
plot (1:iter, all_costs)
title (sprintf ('Standard Gradient Descent with BTLS (alpha: %f, beta: %f)',alpha,discount));
xlabel ('iteration: n');
ylabel ('cost of function: f(b^n)');
end
```

A.3 Steepest Descent with BackTracking Line Search

```
%%% Steepest Descent Algorithm with Backtrack Line Search
% Usage:
    [b, iter, all_costs] = sd_btls (b_init, f, fgrad, P, alpha, discount)
% Parameters:
    b_init: starting search point of variable
    f: objective function
응
    fgrad: gradient of objective function
    P: matrix that defines norm of steepest descent
응
    alpha: parameter for evaluating "OK" step size
    discount (beta): discounting factor for not-OK step size
%% Note that the distinguisable period eps = 10e-16
function [b, iter, all_costs] = sd_btls (b_init, f, fgrad, P, alpha, discount)
assert (discount > 0 && discount < 1);
assert (alpha > 0 \&\& alpha < 0.5);
iter = 1; % iteration count
b = b_init; % variable vector
last_cost = func(b); % value of objective function in most recent iteration
all_costs = []; % all values of objective function, for plotting
while true.
   %% compute essential numerics and do gradient descent
   gradient = fgrad(b);
   delta_b = -1.0 * inv(P) * gradient;
   %% do backtrack line search
   gamma = 1.0;
   while f(b+gamma*delta_b) > f(b) + alpha*gamma*gradient'*delta_b,
       gamma = discount * gamma;
       % disp(sprintf('BTLS: new gamma is %f', gamma));
   end
   b = b + gamma * delta_b;
   cost = func(b);
   rate = (cost / last_cost);
   all_costs = [all_costs cost];
   %% output numeric information of this iteration
   disp (sprintf('Iter: %d, Cost: %e, Conv_Rate: %f, gamma: %f',iter,cost,rate,gamma));
   %% quadratic optimization converges to zero
   if abs((cost - last_cost) / last_cost) < eps,</pre>
       disp('Convergence reached!')
       break
   end
   %% prepare for next iteration
   last_cost = cost;
   iter = iter + 1;
end
%% uncomment following code for plotting individual gradient descent run
% plot f(b^(n)) with regard to n
plot (1:iter, all_costs)
title (sprintf ('Steepest Descent with BTLS (alpha: %f, beta: %f)',alpha,discount));
xlabel ('iteration: n');
ylabel ('cost of function: f(b^n)');
end
```

A.4 Cyclic Coordinate Descent

```
%%% Cyclic Coordinate Descent Algorithm with Backtrack Line Search
% Usage:
    [b, iter, all_costs] = ccd_btls (b_init, f, fgrad, alpha, discount)
% Parameters:
응
    b_init: starting search point of variable
    f: objective function
응
    fgrad: gradient of objective function
응
    P: matrix that defines norm of steepest descent
    alpha: parameter for evaluating "OK" step size
    discount (beta): discounting factor for not-OK step size
% Note:
응
    a) the minimal distinguisable value eps = 10e-16
    b) coordinate descent in cyclical epoch is one iteration
function [b, iter, all_costs] = ccd_btls (b_init, f, fgrad, alpha, discount)
assert (discount > 0 && discount < 1);
assert (alpha > 0 \&\& alpha < 0.5);
iter = 1; % iteration count
b = b_init; % variable vector
ndim = size(b, 1);
last_cost = func(b); % value of objective function in most recent iteration
all_costs = []; % all values of objective function, for plotting
while true,
   gradient = fgrad(b);
    for d = 1:ndim,
       %% compute essential numerics and do gradient descent
       delta_b = zeros (d, 1);
       delta_b(d) = -1.0 * gradient(d);
       %% do backtrack line search
       gamma = 1.0;
       while f(b+gamma*delta_b) > f(b) + alpha*gamma*gradient'*delta_b,
           gamma = discount * gamma;
       end
       b = b + gamma * delta_b;
   end
   cost = func(b);
   rate = (cost / last_cost);
    all_costs = [all_costs cost];
    %% output numeric information of this iteration
   disp (sprintf('Iter: %d, Cost: %e, Conv_Rate: %f, gamma: %f',iter,cost,rate,gamma));
    %% quadratic optimization converges to zero
   if abs((cost - last_cost) / last_cost) < eps,</pre>
       disp('Convergence reached!')
       break
   end
    %% prepare for next iteration
   iter = iter + 1;
    last_cost = cost;
%% uncomment following code for plotting individual gradient descent run
% plot f(b^(n)) with regard to n
plot (1:iter, all_costs)
title (sprintf ('Steepest Descent with BTLS (alpha: %f, beta: %f)',alpha,discount));
xlabel ('iteration: n');
ylabel ('cost of function: f(b^n)');
end
```

A.5 Greedy Coordinate Descent

```
%%% Greedy Coordinate Descent Algorithm with Backtrack Line Search
% Usage:
    [b, iter, all_costs] = gcd_btls (b_init, f, fgrad, alpha, discount)
% Parameters:
    b_init: starting search point of variable
    f: objective function
응
    fgrad: gradient of objective function
    P: matrix that defines norm of steepest descent
    alpha: parameter for evaluating "OK" step size
    discount (beta): discounting factor for not-OK step size
% Note:
    a) the minimal distinguisable value eps = 10e-16
function [b, iter, all_costs] = gcd_btls (b_init, f, fgrad, alpha, discount)
assert (discount > 0 && discount < 1);
assert (alpha > 0 \&\& alpha < 0.5);
iter = 1; % iteration count
b = b_init; % variable vector
ndim = size(b, 1);
last_cost = func(b); % value of objective function in most recent iteration
all_costs = []; % all values of objective function, for plotting
while true.
   gradient = fgrad(b);
   waitlist = zeros(ndim, 1);
   gammalist = zeros(ndim, 1);
    for d = 1:ndim,
       %% compute essential numerics and do gradient descent
       delta_b = zeros (d, 1);
       delta_b(d) = -1.0 * gradient(d);
       %% do backtrack line search
       gamma = 1.0;
       while f(b+gamma*delta_b) > f(b) + alpha*gamma*gradient'*delta_b,
           gamma = discount * gamma;
       end
       waitlist(d) = f(b+gamma*delta_b);
       gammalist(d) = gamma;
   end
    [min_value, min_index] = min(waitlist);
    delta_b = zeros (d, 1);
   delta_b(min_index) = -1.0 * gradient(min_index);
   b = b + gammalist(min_index) * delta_b;
   cost = func(b);
   rate = (cost / last_cost);
   all_costs = [all_costs cost];
   %% output numeric information of this iteration
   disp (sprintf('Iter: %d, Cost: %e, Conv_Rate: %f, gamma: %f',iter,cost,rate,gamma));
    %% quadratic optimization converges to zero
    if abs((cost - last_cost) / last_cost) < eps,</pre>
       disp('Convergence reached!')
       break
   end
   %% prepare for next iteration
   iter = iter + 1;
   last_cost = cost;
%% uncomment following code for plotting individual gradient descent run
% plot f(b^(n)) with regard to n
plot (1:iter, all_costs)
title (sprintf ('Steepest Descent with BTLS (alpha: %f, beta: %f)',alpha,discount));
xlabel ('iteration: n');
ylabel ('cost of function: f(b^n)');
end
```