```
% Tests for LS-GA, LS, 11, P1, P2(1) methods without noise
% clear workspace and close all figures
close all;
clearvars;
% Setting parameters:
k = 16; % Number of sensors
m = 4; % Size of observation vectors b
n = 20; % Size of unknown vector x
reliable sensors list = [6 8 10 12 14]; % Number of consistent ✓
sensors
delta = 1e-6; % Concave approximation related constant
threshold = 1e-4; % Threshold to recover reliable sensors
MCexperiments = 1000;
methods = 5; % Methods being studied (LS-GA, LS, 11, P1, P2(1))
% save results
results = zeros(methods, length(reliable sensors list));
for s index = 1:length(reliable sensors list)
    s = reliable sensors list(s index);
    reliable_sensors = [ones(1, s) zeros(1, k-s)];
    fprintf('Considered %d reliable sensors. ', s);
    parfor j=1:MCexperiments
        %preallocations
        bi = zeros(m, 1, k);
        % unknown vector is modeled as x0 \sim N(0, n^{-1/2})In)
        x0 = mvnrnd(zeros(1, n), n^{-1}*eye(n))';
        % Entries of matrix A are drawn independently from N(0, 1)
        Ai = randn(m, n, k);
        % generate sensor observation data b
```

```
% consistent observations
        for i=1:s
            bi(:, :, i) = Ai(:, :, i)*x0;
        end
        % unreliable sensors
        for i=s+1:k
            bi(:, : , i) = mvnrnd(zeros(1, m), eye(m))';
        end
        % Rearrange arrays and matrices
        b = bi(:);
        C = permute(Ai, [1 3 2]);
        A = reshape(C, [], size(Ai, 2), 1);
        b ga = b(1:m*s);
        A ga = A(1:m*s,:)
        % LS-GA method
        x ls ga = ls method(A ga, b ga, n);
        sensor results ls ga = sensor validation(Ai, bi, x ls ga, ✓
threshold, k, s);
        results_ls_ga(j, s_index) = isequal(reliable sensors, ✓
sensor results ls ga);
        % LS method
        x ls = ls method(A, b, n);
        sensor results ls = sensor validation(Ai, bi, x ls, ✓
threshold, k, s);
        results ls(j, s index) = isequal(reliable sensors, ✓
sensor_results ls);
        % 11 method
        x 11 = 11 method(A, b, n);
        sensor results l1 = sensor validation(Ai, bi, x l1, ✓
threshold, k, s);
        results_l1(j, s_index) = isequal(reliable sensors, ✓
sensor results 11);
        % P1 method
        x p1 = p1 method(Ai, bi, n, k);
```

```
sensor results p1 = sensor validation(Ai, bi, x p1, ✓
threshold, k, s);
        results p1(j, s index) = isequal(reliable sensors, ∠
sensor results p1);
        % P2(1) method
        x p2 1 = p2 1 method(Ai, bi, n, k, x p1, delta);
        sensor results p2 1 = sensor validation(Ai, bi, x p2 1, \checkmark
threshold, k, s);
        results_p2_1(j, s_index) = isequal(reliable_sensors, ✓
sensor results p2 1);
    end
end
results(1,:) = sum(results ls ga, 1);
results(2,:) = sum(results ls, 1);
results(3,:) = sum(results l1, 1);
results(4,:) = sum(results p1, 1);
results(5,:) = sum(results p2 1, 1);
results = (results./MCexperiments).*100;
% Print results
f = figure('Position',[440 500 500 140]);
% Create the column and row names in cell arrays
cnames = {'s=6','s=8','s=10', 's=12', 's=14'};
rnames = {'LS-GA', 'LS','L1','P1', 'P2(1)'};
% Create the uitable
t = uitable(f, 'Data', results,...
            'ColumnName', cnames, ...
            'RowName', rnames);
% Set width and height
t.Position(3) = t.Extent(3);
t.Position(4) = t.Extent(4);
```

```
% Tests for LS-GA, LS, 11, P1, P2(1) methods with noise
% clear workspace and close all figures
close all;
clearvars;
% Setting parameters:
k = 16; % Number of sensors
m = 4; % Size of observation vectors b
n = 20; % Size of unknown vector x
s = 14; % Number of consistent sensors
delta = 1e-6; % Concave approximation related constant
methods = 4; % Methods being studied (LS, 11, P1, P2(1) )
SNR = [5 10 15 20 25]; % SNR wanted
noise levels sigma = (10.^{-SNR/20});
MCexperiments = 1000;
% save all MSE values for all methods
results mse = zeros(length(noise levels sigma), methods);
for noise index = 1:length(noise levels sigma)
    noise sigma = noise levels sigma(noise index);
    fprintf('Considered SNR: %d. ', SNR(noise index));
    parfor j=1:MCexperiments
        %preallocations
        bi = zeros(m, 1, k);
        % unknown vector is modeled as x0 \sim N(0, n^{-1/2})In
        x0 = mvnrnd(zeros(1, n), n^{-1}*eye(n))';
        % Entries of matrix A are drawn independently from N(0, 1)
        Ai = randn(m, n, k);
        for i=1:s
            % reliable sensors measures
```

```
vi = mvnrnd(zeros(1, m), (noise_sigma^2)*eye(m))';
            bi(:, :, i) = Ai(:, :, i)*x0 + vi;
        end
        for i=s+1:k
            % unreliable sensors measures
            bi(:, : , i) = mvnrnd(zeros(1, m), (1+noise\_sigma^2) \checkmark
*eye(m))';
        end
        % Rearrange arrays and matrices
        b = bi(:);
        C = permute(Ai, [1 3 2]);
        A = reshape(C, [], size(Ai, 2), 1);
        b ga = b(1:m*s);
        A ga = A(1:m*s,:)
        % LS-GA method
        x ls ga = ls method(A ga, b ga, n);
        results noise ls ga(j, noise index) = norm(x0-x ls ga)^2;
        % LS method
        x ls = ls method(A, b, n);
        results noise ls(j, noise index) = norm(x0-x ls)^2;
        % 11 method
        x 11 = 11 method(A, b, n);
        results noise l1(j, noise index) = norm(x0-x l1)^2;
        % P1 method
        x p1 = p1 method(Ai, bi, n, k);
        results noise p1(j, noise index) = norm(x0-x p1)^2;
        % P2(1) method
        x p2 1 = p2 1 method(Ai, bi, n, k, x p1, delta);
        results noise p2 1(j, noise index) = norm(x0-x p2 1)^2;
    end
```

end

```
results_mse(:,1) = mean(results_noise_ls_ga, 1);
results mse(:,2) = mean(results noise ls, 1);
results mse(:,3) = mean(results noise l1, 1);
results mse(:,4) = mean(results noise p1, 1);
results mse(:,5) = mean(results noise p2 1, 1);
% plot data and add pretty stuff
semilogy(results mse, '.-', 'MarkerSize',20, 'LineWidth', 1.5)
title('MSE variation with SNR')
xlabel('SNR [dB]')
ylabel('MSE')
legend('LS-GA', 'LS', 'L_1', 'P_1', 'P_2(1)', 'Location', ✓
'southwest');
ax = gca;
ax.XTick = [1 2 3 4 5];
ax.XTickLabel = SNR;
grid on;
```

```
% concave approximation method, cvx aided
function [ x ] = p2_1_method( A, b, n, k, x0, delta )
    for i=1:k
        w(i) = ( norm( b(:,:,i)-A(:,:,i)*x0 ) + delta )^(-1);
    end
    cvx_begin quiet
        variable x(n, 1)
        % define cost function
        f = 0;
        for i=1:k
            f = f + w(i)*norm( b(:,:,i)-A(:,:,i)*x );
        end
        minimize(f)
    cvx_end
end
```

```
% Tests for Matching Solutions method without noise with 2∠
iterations
close all;
clearvars;
k = 16; % Number of sensors
m = 4; % Size of observation vectors b
n = 20; % Size of unknown vector x
reliable sensors list = [6 8 10 12 14]; % Number of consistent ✓
sensors
restriction delta = 10^-10;
threshold = 10^-4;
MCexperiments = 1000;
for s index = 1:length(reliable sensors list)
    s = reliable sensors list(s index);
    fprintf('Considered %d reliable sensors. ', s);
    reliable sensors = [ones(1, s) zeros(1, k-s)];
    parfor j=1:MCexperiments
        %preallocations
        bi = zeros(m, 1, k);
        % unknown vector is modeled as x0 \sim N(0, n^{-1/2})In
        x0 = mvnrnd(zeros(1, n), n^{-1}*eye(n))';
        % Entries of matrix A are drawn independently from N(0, 1)
        Ai = randn(m, n, k);
        % generate sensor observation data b
        % consistent observations
        for i=1:s
            bi(:, :, i) = Ai(:, : ,i)*x0;
        end
```

```
% unreliable sensors
        for i=s+1:k
            bi(:, : , i) = mvnrnd(zeros(1, m), eye(m))';
        end
        x = st = 0 = randn(n, k) / sqrt(n);
        lambda est 0 = mean(x_est_0, 2);
        % lambda est = matching solutions( Ai, bi, n, k, ✓
restriction delta, x0, L0);
        [x est 1, lambda est 1] = matching solutions miter( Ai, bi, ✓
n, k, restriction delta, x est 0, lambda est 0);
        [x est 2, lambda est 2] = matching solutions miter( Ai, bi, ✓
n, k, restriction delta, x est 1, lambda est 1);
        % check for reliable sensors
        method reliable sensors iter1 = sensor validation( Ai, bi, ∠
lambda est 1, threshold, k, s);
        method reliable sensors iter2 = sensor validation( Ai, bi, ✓
lambda est 2, threshold, k, s);
        list_results_iter1(j, s_index) = isequal(reliable_sensors, ✓
method reliable sensors iter1);
        list results iter2(j, s index) = isequal(reliable sensors, ✓
method reliable sensors iter2);
    end
end
results(1, :) = sum(list_results_iter1, 1);
results(2, :) = sum(list results iter2, 1);
results = (results/MCexperiments) * 100;
% Print results
f = figure('Position',[440 500 500 140]);
% Create the column and row names in cell arrays
cnames = \{ 's=6', 's=8', 's=10', 's=12', 's=14' \};
rnames = \{'MS(1)', 'MS(2)'\};
```

```
% Create the uitable
t = uitable(f,'Data',results,...
'ColumnName',cnames,...
'RowName',rnames);

% Set width and height
t.Position(3) = t.Extent(3);
t.Position(4) = t.Extent(4);
```

```
% Tests for Matching Solutions method with noise
close all;
clearvars;
k = 16; % Number of sensors
m = 4; % Size of observation vectors b
n = 20; % Size of unknown vector x
s = 14;
SNR = [5 10 15 20 25]; % SNR wanted
noise_levels_sigma = (10.^(-SNR/20));
restriction delta = 10^-10;
threshold = 10^-4;
MCexperiments = 1000;
for noise index = 1:length(noise levels sigma)
    noise sigma = noise levels sigma(noise index);
    reliable sensors = [ones(1, s) zeros(1, k-s)];
    fprintf('Considered SNR: %d. ', SNR(noise index));
    parfor j=1:MCexperiments
        %preallocations
        bi = zeros(m, 1, k);
        % unknown vector is modeled as x0 \sim N(0, n^{-1/2})In
        x0 = mvnrnd(zeros(1, n), n^{-1}*eye(n))';
        % Entries of matrix A are drawn independently from N(0, 1)
        Ai = randn(m, n, k);
        for i=1:s
            % reliable sensors measures
            vi = mvnrnd(zeros(1, m), (noise_sigma^2)*eye(m))';
            bi(:, :, i) = Ai(:, :, i)*x0 + vi;
        end
        for i=s+1:k
```

```
% unreliable sensors measures
            bi(:, : , i) = mvnrnd(zeros(1, m), (1+noise sigma^2) \checkmark
*eye(m))';
        end
        x_iter0 = randn(n, k) / sqrt(n);
        lambda iter0 = mean(x_iter0, 2);
        lambda iter1 = matching solutions( Ai, bi, n, k, ∠
restriction delta, x iter0, lambda iter0);
        results noise ms(j, noise index) = norm(x0-lambda iter1)^2;
    end
end
results mse = mean(results noise ms, 1);
% plot data and add pretty stuff
semilogy(results_mse, '.-', 'MarkerSize',20, 'LineWidth', 1.5)
title('MSE variation with SNR')
xlabel('SNR [dB]')
ylabel('MSE')
legend('MS(1)', 'Location', 'southwest');
ax = gca;
ax.XTick = [1 2 3 4 5];
ax.XTickLabel = SNR;
grid on;
```

```
% Matching solutions method
function [ x, L ] = matching_solutions_miter( A, b, n, k, delta, ∠
x0, L0)
    cvx_begin quiet
        variable x(n, k)
        variable L(n, 1)
        % define cost function
        for i=1:k
            f(i) = norm(x(:,i)-L) / norm(x0(:,i)-L0);
        end
        minimize(sum(f))
        subject to
        for i=1:k
            (b(:,:,i)-A(:,:,i)*x(:,i))'*(b(:,:,i)-A(:,:,i)*x(:,i)) \checkmark
<= delta;
        end
    cvx_end
end
```