

Hands-on Exercises

We will now use the GTSAM library to solve the smoothing problem as above. The file hw4_data.mat contains three variables:

dposes - holding noisy odometry measurements,

traj3- a rough initial estimate for robot trajectory

poses3_gt - the actual ground truth poses

```
addpath(fullfile(pwd, 'gtsam_toolbox'));  
disp('added toolbox to path')
```

```
added toolbox to path
```

1. Load and display the initial trajectory.

Cell array traj3 contains robot 6dof poses given as 4x4 transformation matrices (T C G in course notations). Follow the steps below to display the trajectory using GTSAM. Add your plot to your homework pdf / report.

```
data=importdata('hw4_data.mat');
```

(a) Convert transformations to gtsam.Pose3 objects.

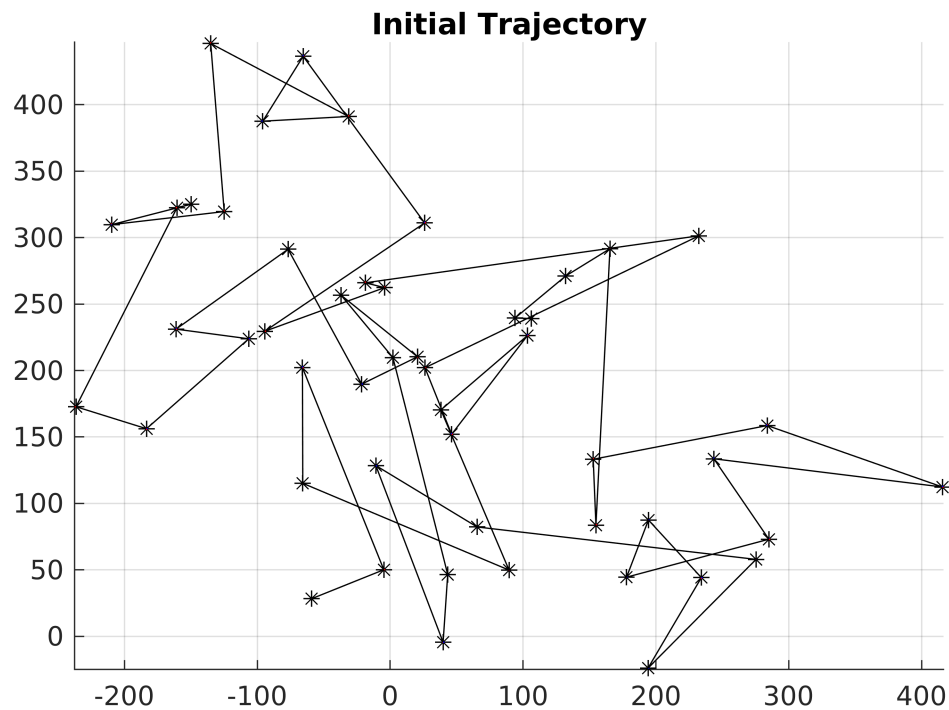
```
traj3Poses = cellfun(@(R) gtsam.Pose3(R), data.traj3);
```

(b) Store the above poses in a gtsam.Values object representing the initial estimate for robot trajectory.

```
trajectory = gtsam.Values;  
for ii = 1:length(traj3Poses)  
    trajectory.insert(gtsam.symbol('x',ii),traj3Poses(ii));  
end
```

(c) Use gtsam.plot3DTrajectory to display the list of poses. Include the plot in your report.

```
figure()  
gtsam.plot3DTrajectory(trajectory, '*-k',200)  
view([0,-1,0]); axis equal; axis tight; grid on;  
title('Initial Trajectory')
```

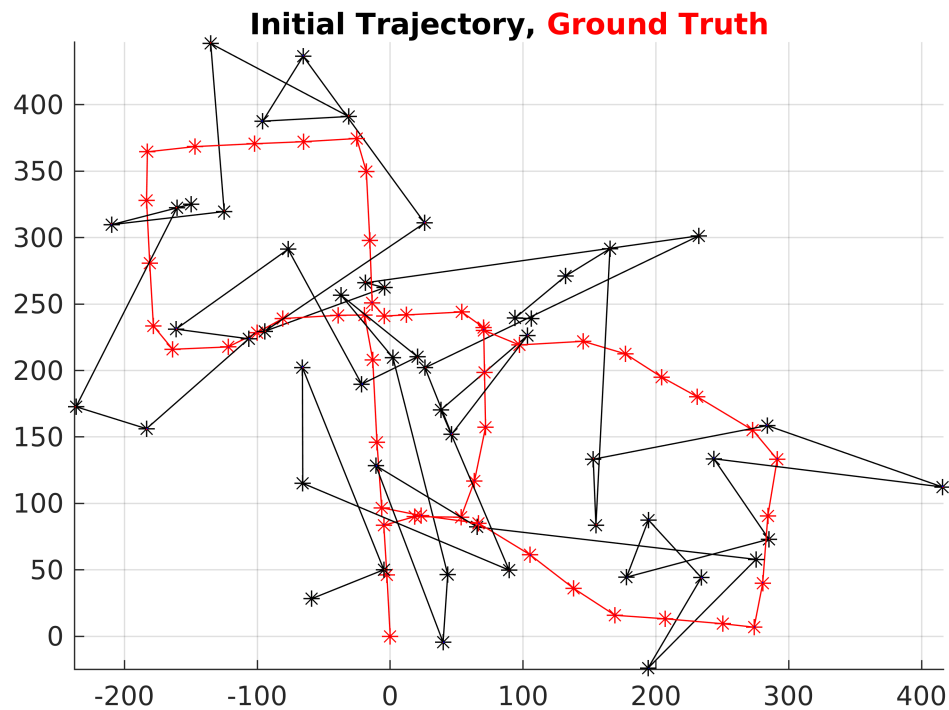


(d) Overlay in your plot the ground truth trajectory given in the variable `poses3_gt` (use a different color).

```
traj3Poses = cellfun(@(R) gtsam.Pose3(R), data.traj3);

gt3Poses = cellfun(@(R) gtsam.Pose3(R), data.poses3_gt);
gt = gtsam.Values;
for ii = 1:length(traj3Poses)
    gt.insert(gtsam.symbol('x',ii),gt3Poses(ii));
end

figure();
gtsam.plot3DTrajectory(trajjectory, '*-k',200);
view([0,-1,0]); axis equal; axis tight; grid on; hold on;
gtsam.plot3DTrajectory(gt, '*-r',200);
title('\color{black}Initial Trajectory, \color{red}Ground Truth')
```



2. Construct factor graph.

(a) `graph = gtsam.NonlinearFactorGraph` creates a general factor graph.

```
graph = gtsam.NonlinearFactorGraph;
```

(b) Add `gtsam.BetweenFactorPose3` factors for given "measured" relative poses.

```
Sr = 1e-3 * ones(3,1);
St = 0.1 * ones(3,1);
noiseModel = gtsam.noiseModel.Diagonal.Sigmas([Sr ; St ]);
zd3Poses = cellfun(@(R) gtsam.Pose3(R), data.dpose);

for ii = 1:length(zd3Poses)
    graph.add(gtsam.BetweenFactorPose3(gtsam.symbol('x',ii),...
        gtsam.symbol('x',ii+1),...
        zd3Poses(ii),...
        noiseModel));
end
```

(c) Assume robot is initially located at the origin, its axes aligned with the global reference frame. Use `gtsam.PriorFactorPose3` to incorporate this information into the factor graph.

```
Origin = gtsam.Pose3(eye(4));
```

```
graph.add(gtsam.PriorFactorPose3(gtsam.symbol('x',1),Origin,noiseModel));
```

3. Calculate and display the MAP trajectory estimate.

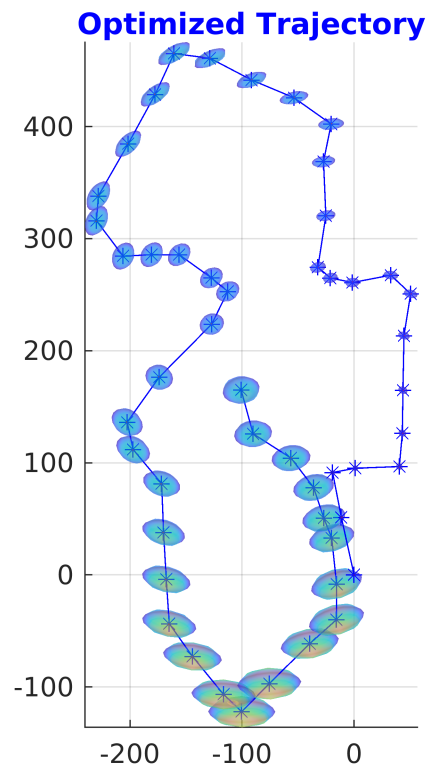
(a) Create and run an optimizer.

```
optimizer = gtsam.LevenbergMarquardtOptimizer(graph,trajectory);
result =optimizer.optimizeSafely();
```

(b) Display updated trajectory estimate.

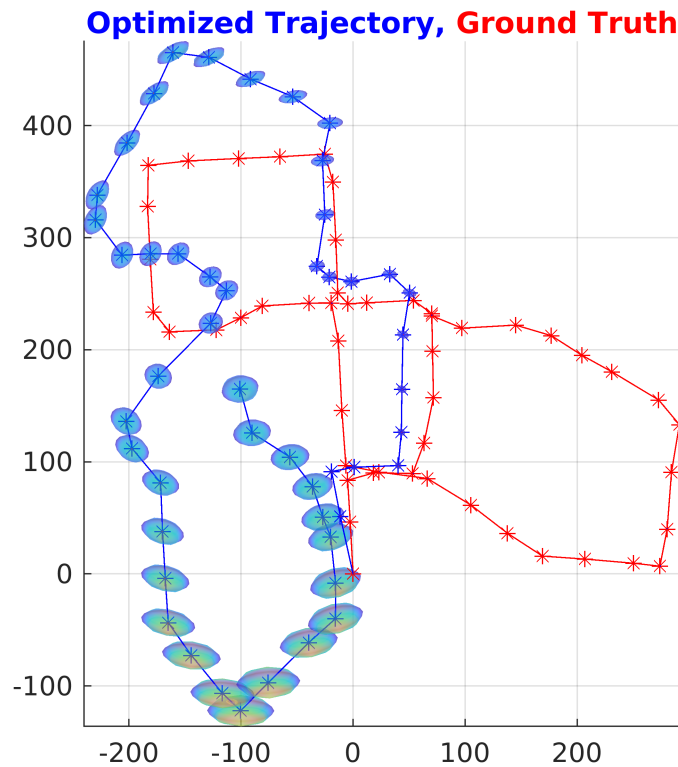
```
marginals = gtsam.Marginals(graph, result);

figure();
gtsam.plot3DTrajectory(result, '*-b' ,[],[],marginals)
view([0,-1,0]); axis equal; axis tight; grid on;
title('\color{blue}Optimized Trajectory');
```



(c) Overlay with the ground truth as in the 1st clause.

```
figure();
gtsam.plot3DTrajectory(result, '*-b' ,[],[],marginals)
gtsam.plot3DTrajectory(gt, '*-r' ,200);
view([0,-1,0]); axis equal; axis tight; grid on;
title('\color{blue}Optimized Trajectory, \color{red}Ground Truth')
```



4. Loop closure

```
R2t1=[0.330571768  0.0494690228  -0.942483486;
      0.0138000518 0.998265226   0.0572371968;
      0.943679959  -0.0319273223  0.329315626];
t_2t1_in1=[-24.1616858;
           -0.0747429903;
           275.434963];

t1 = 3;
t2 = 42;
```

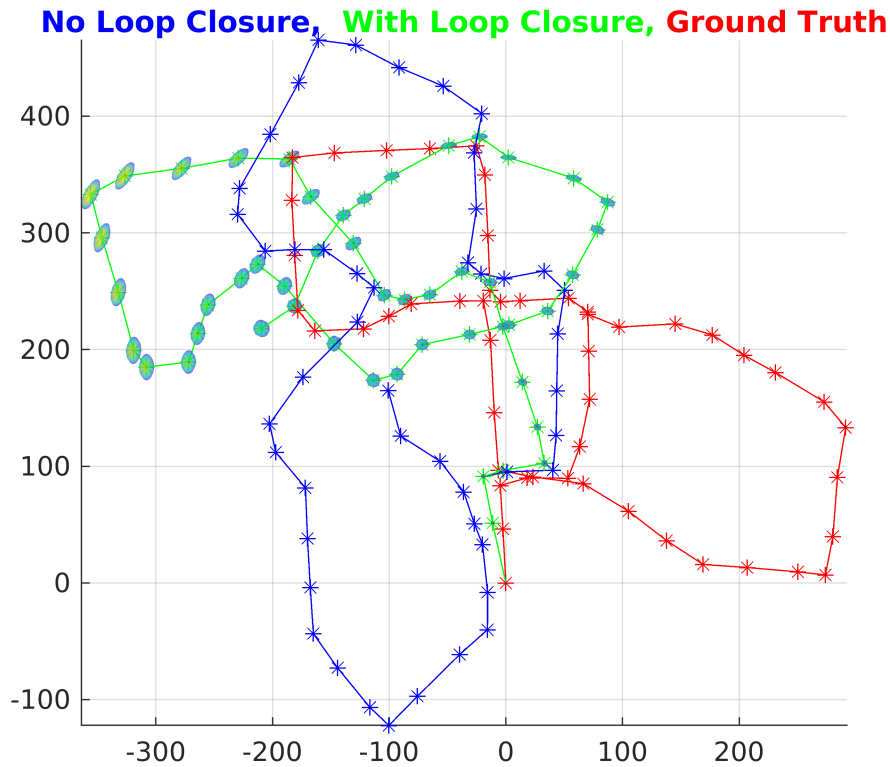
(a) Assuming the same noise model as before, show the estimated trajectory when incorporating this new information (as before, overlay it with ground truth).

```
dPose = gtsam.Pose3([R2t1,t_2t1_in1; [0 0 0 1]]);
graph.add(gtsam.BetweenFactorPose3(gtsam.symbol('x',t1),...
    gtsam.symbol('x',t2),...
    dPose,...
    noiseModel));

optimizer = gtsam.LevenbergMarquardtOptimizer(graph,trajectory);
result_LC =optimizer.optimizeSafely();
marginals = gtsam.Marginals(graph, result_LC);

figure();
gtsam.plot3DTrajectory(result, '*-b');
gtsam.plot3DTrajectory(result_LC, '*-g' ,[],[],marginals);
```

```
gtsam.plot3DTrajectory(gt, '*-r', 200);
view([0,-1,0]); axis equal; axis tight; grid on;
title('\color{blue}No Loop Closure, \color{green} With Loop Closure, \color{red}Ground
```



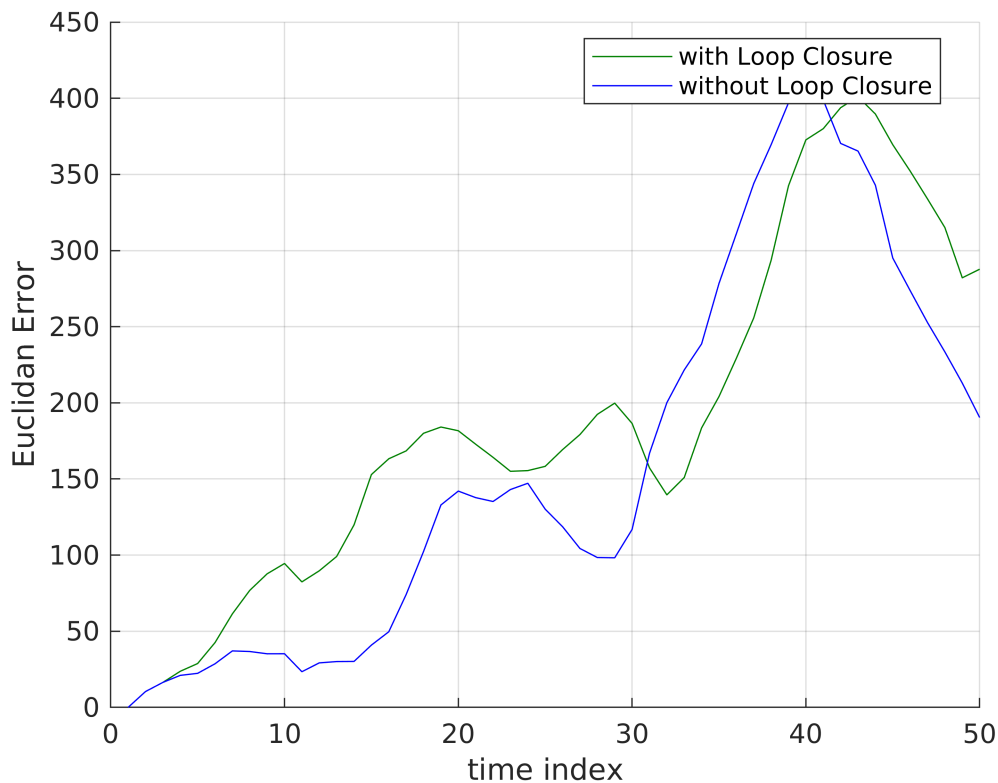
(b) Plot the localization error in metres (location difference only - Euclidean distance) over time, for the result without and with the loop closure.

```
poses_gt=gtsam.utilities.extractPose3(gt);
poses_nLC=gtsam.utilities.extractPose3(result);
poses_yLC=gtsam.utilities.extractPose3(result_LC);

%extract translation, motion is almost planar on the x-z plane.
t_poses_gt = poses_gt(:,10:12);
t_poses_nLC = poses_nLC(:,10:12);
t_poses_yLC = poses_yLC(:,10:12);

err_nLC = vecnorm(t_poses_nLC - t_poses_gt,2,2);
err_yLC = vecnorm(t_poses_yLC - t_poses_gt,2,2);

fig = figure('color',[1,1,1]);
ax = axes(fig);
grid(ax,'on'); hold(ax,'on'); xlabel(ax,'time index'); ylabel(ax,'Euclidan Error');
plot(ax,err_yLC,'color',[0 0.5,0]);
plot(ax,err_nLC,'b');
legend('with Loop Closure','without Loop Closure','location','best');
```



Bonus: Compute a good loop closure from poses in t_9 and t_{50} and repeat
 reminder:

$$R_1^2 = R_G^2 R_1^G = (R_2^G)^T R_1^G$$

$$t_{2 \rightarrow 1}^2 = R_G^2 (t_{2 \rightarrow G}^G - t_{1 \rightarrow G}^G) = (R_2^G)^T (t_{2 \rightarrow G}^G - t_{1 \rightarrow G}^G)$$

Reconstruct Graph

```
graph = gtsam.NonlinearFactorGraph;
Sr = 1e-3 * ones(3,1);
St = 0.1 * ones(3,1);
noiseModel = gtsam.noiseModel.Diagonal.Sigmas([Sr ; St]);
zd3Poses = cellfun(@(R) gtsam.Pose3(R), data.dpose);
for ii = 1:length(zd3Poses)
    graph.add(gtsam.BetweenFactorPose3(gtsam.symbol('x',ii),...
        gtsam.symbol('x',ii+1),...
        zd3Poses(ii),...
        noiseModel));
end
Origin = gtsam.Pose3(eye(4));
graph.add(gtsam.PriorFactorPose3(gtsam.symbol('x',1),Origin,noiseModel));
```

Connect new loop closure

```

t1 = 50;
t2 = 9;

pose1 = [reshape(poses_gt(t1,:),[3,4]);[0 0 0 1]];
pose2 = [reshape(poses_gt(t2,:),[3,4]);[0 0 0 1]];
R1tG = pose1(1:3,1:3);
R2tG = pose2(1:3,1:3);
t_1tG_inG = pose1(1:3,4);
t_2tG_inG = pose2(1:3,4);

R1t2 = R2tG'*R1tG;
t_2t1_in2 = R2tG'*(t_2tG_inG-t_1tG_inG);

dPose = gtsam.Pose3([R1t2,t_2t1_in2;[0 0 0 1]]);

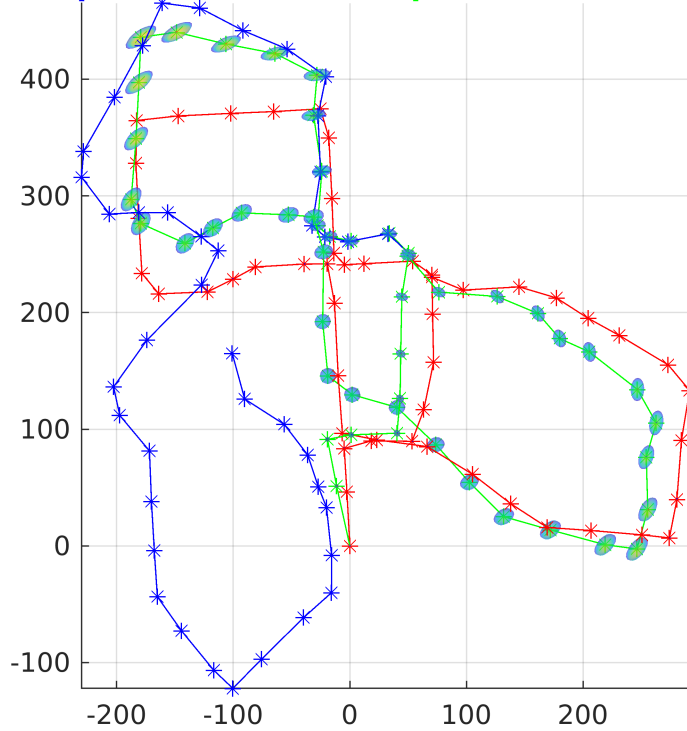
graph.add(gtsam.BetweenFactorPose3(gtsam.symbol('x',t1),...
    gtsam.symbol('x',t2),...
    dPose,...
    noiseModel));

optimizer = gtsam.LevenbergMarquardtOptimizer(graph,trajectory);
result_LC =optimizer.optimizeSafely();
marginals = gtsam.Marginals(graph, result_LC);

figure();
gtsam.plot3DTrajectory(result, '*-b');
gtsam.plot3DTrajectory(result_LC, '*-g' ,[],[],marginals);
gtsam.plot3DTrajectory(gt, '*-r' ,200);
view([0,-1,0]); axis equal; axis tight; grid on;
title('\color{blue}No Loop Closure, \color{green} With Loop Closure, \color{red}Ground

```


No Loop Closure, With Loop Closure, Ground Truth



Error Graphs for our new loop closure

```
poses_gt=gtsam.utilities.extractPose3(gt);
poses_nLC=gtsam.utilities.extractPose3(result);
poses_yLC=gtsam.utilities.extractPose3(result_LC);

%extract translation, motion is almost planar on the x-z plane.
t_poses_gt = poses_gt(:,10:12);
t_poses_nLC = poses_nLC(:,10:12);
t_poses_yLC = poses_yLC(:,10:12);

err_nLC = vecnorm(t_poses_nLC - t_poses_gt,2,2);
err_yLC = vecnorm(t_poses_yLC - t_poses_gt,2,2);

fig = figure('color',[1,1,1]);
ax = axes(fig);
grid(ax,'on'); hold(ax,'on'); xlabel(ax,'time index'); ylabel(ax,'Euclidan Error');
plot(ax,err_yLC,'color',[0 0.5,0]);
plot(ax,err_nLC,'b');
legend('with Loop Closure','without Loop Closure','location','best');
```

