

CS284 HW2 Report: Pose Graph Optimization

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Abstract—This report covers a basic description of the implementation, performance, and Q&A of the pose graph optimization in HW2.

I. PROJECT

A. Description of the pose graph optimization algorithm

The pose graph optimization algorithm can be summarized as follow:

ALGORITHM 1: Graph Optimization Algorithm

```
1: while !converged do
2:    $(H, b) = \text{buildLinearSystem}(x)$ 
3:    $\Delta x = \text{solveSparse}(H\Delta x = -b)$ 
4:    $x = x + \Delta x$ 
5: end while
6: return  $x$ 
```

I use *python* to implement the algorithm and visualize the result.

B. Structure of the project

CS284_hw2_data contains the raw data.

g2o_data contains the data in g2o format.

g2opy contains the 3rd library.

main.py is the implementation and visualization of the pose graph optimization algorithm.

Mode	LastWriteTime	Length	Name
d----	2021/10/17 16:53	1	.idea
d----	2021/10/17 1:16	1	CS284_hw2_data
d----	2021/10/17 16:38	1	g2opy
d----	2021/10/17 16:41	1	g2o_data
-a----	2021/10/17 16:43	5.83KB	main.py

Fig. 1: Structure of the project

C. Dependencies

python $\geq 3.8.11$

numpy $\geq 1.20.3$

D. Instructions

For task k , run `python main.py --task_index=k`

II. PERFORMANCE

All the results can be Visualized by run *main.py*.

A. Efficiency

The time consumed for Task1 is 0.51043s;

The time consumed for Task2 is 0.51268s;

The time consumed for Task3 is 0.53842s;

The time calculated includes the process of reading raw data, the main optimization part, printing, plotting and so on.

B. Accuracy and Visualization

Accuracy

I take the ideal pose positions as reference(i.e., $[0, 0, 0]$, $[0, 0, \pi/6]$, $[0, 0, \pi/3]$, ..., $[0, 0, 2\pi]$) and calculate the sum of error for t_x, t_y, θ independently.

	t_x	t_y	θ
original	11.73150	7.28932	0.61755
task2	12.84017	7.15501	0.61756
task3	12.16079	7.46702	0.23592

TABLE I: Error for all the tasks

Visualization

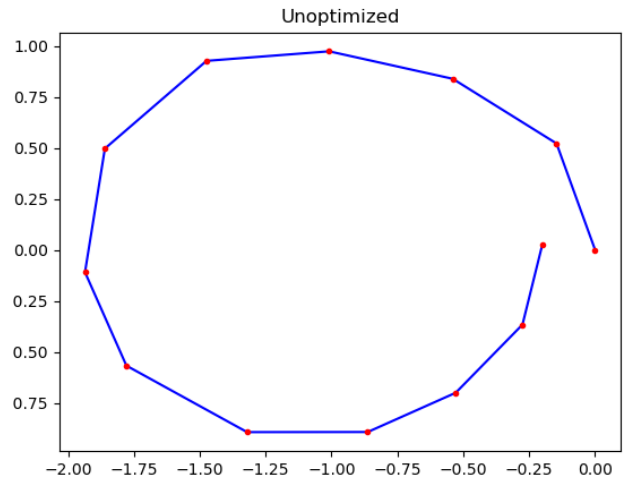


Fig. 2: Results for Task1

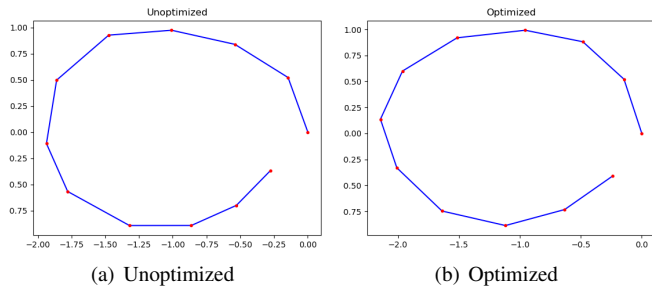


Fig. 3: Results for Task2

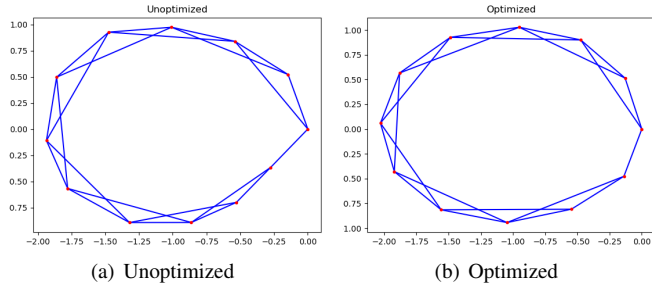


Fig. 4: Results for Task3

III. Q&A

Q1: For task1, what can you observe? In particular, is there anything to say about the thirteenth pose?

A1: As shown in Fig. 2, the distance between 13th and 1st pose is obvious (i.e., The Euclidean distance is 0.20182). In other word, the cumulative error is large.

Q2: For task2, Describe your observations.

A2: Without the edge between 13th and 1st pose, the effect of pose graph optimization is weak. The interpret $T_{12,1}$ is $[-0.00247, 0.45857, 0.64978]$.

Q3: For task3, Describe your observations.

A3: The effect of pose graph optimization is obvious. The 12 absolute poses are close to the ideal circular trajectory of 1m radius.

Q4: Bonus questions: Visualize and explain what the information matrix looks like.

A4: I set the information matrix as identity in this project. The information matrix is the inverse of covariance matrix, so the visualization of it is Symmetrical.