

多传感器融合第一次作业讲评





第一题



- ◆ 第一题主要涉及下载数据集和配置环境
- → rviz的命令为 rviz -d display_bag.rviz
- ◆ 如果环境存在问题可以使用课程提供的docker镜像



- →在第一题的基础上修改代码。仿照老师写好的NDT匹配写一个PCL的ICP就可以。
- →注意事项 : 需要使用头文件 <pcl/registration/icp.h>
- →在 lidar_localization/src/models/registration/里 和 lidar_localization/include/lidar_localization/models/registration/ 仿照 老师提供的NDT匹配写一个ICP匹配的.hpp和.cpp
- →修改config.yaml 仿照老师提供的NDT参数写ICP的参数
- →在lidar_localization/src/front_end/front_end.cpp 中 FrontEnd::InitRegistration 添加ICP匹配



```
#ifndef LIDAR_LOCALIZATION_MODELS_REGISTRATION_ICP_REGISTRATION_HPP_
#define LIDAR_LOCALIZATION_MODELS_REGISTRATION_ICP_REGISTRATION_HPP_
#include "lidar_localization/models/registration/registration_interface.hpp"
#include <pcl/registration/icp.h>
namespace lidar_localization {
    class ICPRegistraion: public RegistrationInterface {
        ICPRegistraion(const YAML::Node& node);
        ICPRegistraion(float max_dist, float trans_eps, float eculi_eps, int max_iter);
        bool SetInputTarget(const CloudData::CLOUD_PTR& input_target) override;
        bool ScanMatch(const CloudData::CLOUD PTR& input source,
                       const Eigen::Matrix4f& predict_pose,
                       CloudData::CLOUD_PTR& result_cloud_ptr,
                       Eigen::Matrix4f& result_pose) override;
        bool SetRegistrationParam(float max dist, float trans eps, float eculi eps, int max iter);
        pcl::IterativeClosestPoint<CloudData::POINT,CloudData::POINT>::Ptr icp_ptr_;
```



```
#include "glog/logging.h"
namespace lidar_localization {
   ICPRegistraion::ICPRegistraion(const YAML::Node &node)
        :icp_ptr_(new pcl::IterativeClosestPoint<CloudData::P0INT,CloudData::P0INT>()){
       float max_dist = node["max_dist"].as<float>();
       float trans_eps = node["trans_eps"].as<float>();
       float eculi_eps = node["eculi_eps"].as<float>();
       int max_iter = node["max_iter"].as<int>();
       SetRegistrationParam(max_dist, trans_eps, eculi_eps, max_iter);
   ICPRegistraion::ICPRegistraion(float max_dist, float trans_eps, float eculi_eps, int max_iter)
       :icp_ptr_(new pcl::IterativeClosestPoint<CloudData::POINT,CloudData::POINT>()){
       SetRegistrationParam(max_dist,trans_eps,eculi_eps,max_iter);
   bool ICPRegistraion::SetRegistrationParam(float max_dist, float trans_eps, float eculi_eps, int max_iter) {
       icp_ptr_->setMaxCorrespondenceDistance(max_dist);
       icp_ptr_->setTransformationEpsilon(trans_eps);
       icp_ptr_->setEuclideanFitnessEpsilon(eculi_eps);
       icp_ptr_->setMaximumIterations(max_iter);
       LOG(INFO) << "ICP 的匹配参数为; " << std::endl
                 << "max_dist: " << max_dist << ","
                 << "trans_eps: " << trans_eps << ","
                 << "eculi eps: " << eculi eps << ","
                  << "max_iter: " << max_iter << ","
```



```
bool ICPRegistraion::SetInputTarget(const CloudData::CLOUD_PTR &input_target) {
   icp_ptr_->setInputTarget(input_target);
   return true;
bool ICPRegistraion::ScanMatch(const CloudData::CLOUD_PTR &input_source,
                               const Eigen::Matrix4f &predict_pose,
                               CloudData::CLOUD_PTR &result_cloud_ptr,
                               Eigen::Matrix4f &result_pose) {
   icp_ptr_->setInputSource(input_source);
   icp_ptr_->align(*result_cloud_ptr,predict_pose);
   result_pose = icp_ptr_->getFinalTransformation();
   return true;
```

第二題



```
bool FrontEnd::InitRegistration(std::shared_ptr<RegistrationInterface>& registration_ptr, const YA
    std::string registration_method = config_node["registration_method"].as<std::string>();
    LOG(INFO) << "点云匹配方式为: " << registration_method;
    if (registration method == "NDT") {
       registration_ptr = std::make_shared<NDTRegistration>(config_node[registration_method]);
    H
    else if (registration_method == "ICP") {
       registration_ptr = std::make_shared<ICPRegistraion>(config_node[registration_method]);
    else {
       LOG(ERROR) << "没找到与 " << registration_method << " 相对应的点云匹配方式!";
       return false;
    return true;
```



```
TCP:

max_dist : 1

trans_eps : 0.01

eculi_eps : 0.01

max_iter : 30
```

第三题



这里介绍来自LZM_HIT同学的优秀作业

3.1 作业思路

考虑到PCL_ICP中默认用的是SVD分解的方法,这里选择使用高斯牛顿的迭代方法求解ICP,自己手动推导一下相关的过程,步骤如下:

- a. 输入目标点云 目标点云用来构建一个kdtree,方便待匹配的点云寻找最近点。
- b. 输入待匹配的点云以及初始位姿R, t
- 1.初始化Hessian<6,6>=Zero,B<6,1>=Zero,对待匹配点云中的每个点 ${m p}_i$,利用初始位姿转换到目标点云坐标系下的到 ${m p}_i'={m R}*{m p}_i+{m t}$
- 2.根据kdtree在目标点云中搜索到离 \mathbf{p}_i' 最近的点 \mathbf{q}_j' ,计算距离误差 $f=\mathbf{p}_i'-\mathbf{q}_i'$,若f.norm小于设置的距离上限,执行下一步,否则,该点对舍弃。
- 3.对找到的每一个点对,计算 Jacobian<3,6> ,Hessian_i ,B_i

第三题



3.对找到的每一个点对,计算 Jacobian<3,6>, Hessian_i , B_i

对平移的求导
$$\dfrac{\partial oldsymbol{f}}{\partial oldsymbol{t}} = oldsymbol{I}$$
对旋转的求导,右导数 $\dfrac{\partial oldsymbol{f}}{\partial oldsymbol{R}} = -oldsymbol{R} * oldsymbol{p_i^{\Lambda}}$
 $Jacobian = [\dfrac{\partial oldsymbol{f}}{\partial oldsymbol{t}}, \dfrac{\partial oldsymbol{f}}{\partial oldsymbol{R}}]$
 $Hessian_i = Jacobian^T Jacobian$
 $B_i = -Jacobian^T * oldsymbol{f}$

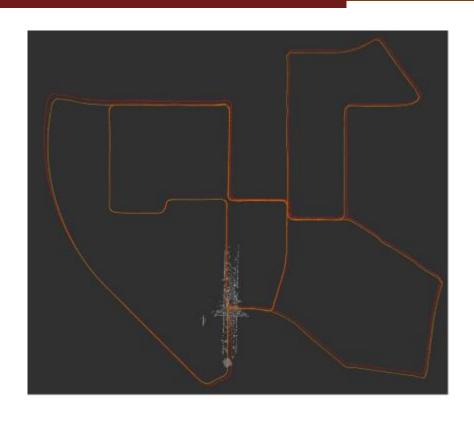
4.计算所有点对计算的 $Hessian_i$, B_i 求和,计算状态增量 δx

$$Hessian = \Sigma Hessian_i \ B = \Sigma B_i \ oldsymbol{\delta x} = Hessian.inverse()*B$$

5.利用增量 δx 计算新的R,t,返回第一步计算,直到达到收敛条件或者迭代次数

第三题





3.3 对比结果

分段统计精度	max	mean	median	min	rmse	sse	std
PCL_NDT	1.852103	0.786486	0.692123	0.211582	0.875525	34.494444	0.384686
PCL_ICP	1759.394213	274.894422	76.649318	0.655707	512.77679	11832301.6618	432.866138
ICP_MANUAL	8.473569	3.405850	3.357728	0.151904	3.981459	713.340758	2.062087

整体轨迹误差	max	mean	median	min	rmse	sse	std
PCL_NDT	68.038200	22.565926	18.100196	0.000001	30.304401	4161992.585209	20.227103
PCL_ICP	3541.7586	526.18734	105.68439	0.000001	1059.34568	5081381713.0555	919.423817
ICP_MANUAL	25.627234	12.309445	11.004779	0.000001	14.704755	979088.602983	8.044090

从结果可以看出:

对于**分段统计精度**而言,精度从高到低是: PCL_NDT>ICP_MANUAL>PCL_ICP

对于整体轨迹误差而言,精度从高到低是: ICP_MANUAL>PCL_NDT>PCL_ICP

```
CloudData::CLOUD_PTR transformed_cloud(new CloudData::CLOUD);
int knn = 1;
int iterator_num = 0;
while (iterator_num < max_iterator_)
 pcl::transformPointCloud(*input_source, *transformed_cloud, transformation_);
 Eigen::Matrix<float, 6, 6> Hessian;
 Eigen::Matrix<float, 6, 1> B;
Hessian.setZero();
B.setZero();
for (size ti=0; i < transformed cloud->size(); ++i)
 auto ori_point = input_source->at(i);
  if (!pcl::isFinite(ori_point))
  auto transformed_point = transformed_cloud->at(i);
  std::vector<float> distasnces;
 std::vector<int> indexs;
  kdtree_ptr_->nearestKSearch(transformed_point, knn, indexs, distasnces);
  if (distasnces[0] > max_correspond_distance_)
  Eigen::Vector3f closet_point = Eigen::Vector3f(target_cloud_->at(indexs[0]).x, target_cloud_->at(indexs[0]).y,
                        target_cloud_->at(indexs[0]).z);
  Eigen::Vector3f err dis =
   Eigen::Vector3f(transformed_point.x, transformed_point.y, transformed_point.z) - closet_point;
  Eigen::Matrix<float, 3, 6> Jacobian(Eigen::Matrix<float, 3, 6>::Zero());
  Jacobian.leftCols<3>() = Eigen::Matrix3f::Identity();
  Jacobian.rightCols<3>() =
    -rotation_matrix_ * Sophus::SO3f::hat(Eigen::Vector3f(ori_point.x, ori_point.y, ori_point.z));
```

void ICPRegistrationManual::calculateTrans(const CloudData::CLOUD_PTR& input_source)

```
Hessian += Jacobian.transpose() * Jacobian;
 B += -Jacobian.transpose() * err dis;
iterator_num++;
if (Hessian.determinant() == 0)
 continue;
Eigen::Matrix<float, 6, 1> delta_x = Hessian.inverse() * B;
translation_ += delta_x.head<3>();
auto delta_rotation = Sophus::SO3f::exp(delta_x.tail<3>());
rotation_matrix_ *= delta_rotation.matrix();
transformation_.block<3, 3>(0, 0) = rotation_matrix_;
transformation_.block<3, 1>(0, 3) = translation_;
// namespace lidar_localization
```

在线问答







感谢各位聆听

Thanks for Listening



