

# 第四次作业

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## 2图像去畸变

```

int main(int argc, char **argv)
{
    // 本程序需要你自己实现去畸变部分的代码。尽管我们可以调用OpenCV的去畸变，但自己实现一遍有助于理解。
    // 畸变参数
    double k1 = -0.28340811, k2 = 0.07395907, p1 = 0.00019359, p2 = 1.76187114e-05;
    // 内参
    double fx = 458.654, fy = 457.296, cx = 367.215, cy = 248.375;
    cv::Mat image = cv::imread(image_file,0);    // 图像是灰度图，CV_8UC1
    int rows = image.rows; //x
    int cols = image.cols; //y
    cout<<"rows: "<<rows<<" cols: "<<cols<<endl;
    cv::imshow("image undistorted", image);
    cv::waitKey();
    cv::Mat image_undistort = cv::Mat(rows, cols, CV_8UC1);    // 去畸变以后的图
    // 计算去畸变后图像的内容
    for (int v = 0; v < rows; v++)
    {
        for (int u = 0; u < cols; u++)
        {
            double u_distorted = 0, v_distorted = 0;
            // TODO 按照公式，计算点(u,v)对应到畸变图像中的坐标(u_distorted, v_distorted) (~6 lines)
            // start your code here
            double x_distorted = 0, y_distorted = 0;
            double x = 0, y = 0, r = 0;
            x = (v - cy)/fy;
            y = (u - cx)/fx;
            r = sqrt(x*x + y*y);

            x_distorted = x * ( 1 + k1*r*r + k2*pow(r,4) ) + 2*p1*x*y + p2*(r*r + 2*x*x); //row
            y_distorted = y * ( 1 + k1*r*r + k2*pow(r,4) ) + p1*(r*r + 2*y*y) + 2*p2*x*y; //col

            v_distorted = fx * x_distorted + cx; //x
            u_distorted = fy * y_distorted + cy; //y
            // end your code here

            // 赋值（最近邻插值）
            if (u_distorted >= 0 && v_distorted >= 0 && u_distorted < cols && v_distorted < rows)
            {
                image_undistort.at<uchar>(v, u) = image.at<uchar>((int) v_distorted, (int) u_distorted);
            }
            else
            {
                image_undistort.at<uchar>(v, u) = 0;
            }
        }
    }
}

```

```
    }  
    }  
}  
// 画图去畸变后图像  
cv::imshow("image undistorted", image_undistort);  
cv::waitKey();  
return 0;  
}
```

### 3双目视差的使用

```

int main(int argc, char **argv)
{

    // 内参
    double fx = 718.856, fy = 718.856, cx = 607.1928, cy = 185.2157;

    // 间距
    double d = 0.573;

    // 读取图像
    cv::Mat left = cv::imread(left_file, 0);
    cv::Mat right = cv::imread(right_file, 0);
    cv::Mat disparity = cv::imread(disparity_file, 0); // disparity 为CV_8U,单位为像素

    // 生成点云
    vector<Vector4d, Eigen::aligned_allocator<Vector4d> > pointcloud; //

    // TODO 根据双目模型计算点云
    // 如果你的机器慢, 请把后面的v++和u++改成v+=2, u+=2
    for (int v = 0; v < left.rows; v++) // y
    {
        for (int u = 0; u < left.cols; u++) // x
        {
            //mat.at q
            Vector4d point(0, 0, 0, left.at<uchar>(v, u) / 255.0); // 前三
            维为xyz,第四维为颜色

            // start your code here (~6 lines)
            // 根据双目模型计算 point 的位置

            // unsigned char di = disparity.ptr<unsigned short> (v)[u]; //
            视差图

            int di = disparity.ptr<uchar> (v)[u]; // 视差图
            // if ( d==0 ) continue; // 为0表示没有测量到
            // z = f *b / d
            point[2] =fx*d/di; //z ;d为间距
            point[0] = (u-cx)*point[2]/fx; //x
            point[1] = (v-cy)*point[2]/fy; //y
            pointcloud.push_back(point);
            // end your code here
        }
    }
    // 画出点云
    showPointCloud(pointcloud);
    return 0;
}

```

## 4矩阵微分

1.  $d(Ax)/dx$ 是实值向量函数的行向量偏导数，称之为向量函数 $f(x)$ 在 $x$ 处的Jacobian矩阵
2.  $d(x^T Ax)/dx$ 是实值标量函数的行向量偏导数，称之为实值标量函数 $f(x)$ 在 $x$ 处的梯度向量

## 5高斯牛顿

```

int main(int argc, char **argv)
{
    double ar = 1.0, br = 2.0, cr = 1.0;           // 真实参数值
    double ae = 2.0, be = -1.0, ce = 5.0;         // 估计参数值
    int N = 100;                                    // 数据点
    double w_sigma = 1.0;                           // 噪声Sigma值
    cv::RNG rng;                                     // OpenCV随机数产生器

    vector<double> x_data, y_data;                  // 数据
    for (int i = 0; i < N; i++)
    {
        double x = i / 100.0;
        x_data.push_back(x);
        y_data.push_back(exp(ar * x * x + br * x + cr) + rng.gaussian(w_sigma));
    }

    // 开始Gauss-Newton迭代
    int iterations = 100;                          // 迭代次数
    double cost = 0, lastCost = 0;                 // 本次迭代的cost和上一次迭代的cost

    for (int iter = 0; iter < iterations; iter++) //迭代 100 times
    {
        //H
        Matrix3d H = Matrix3d::Zero();             // Hessian = J^T J in Gauss-Newton
        //g
        Vector3d b = Vector3d::Zero();             // bias
        cost = 0;

        for (int i = 0; i < N; i++) //N个数据点
        {
            double xi = x_data[i], yi = y_data[i]; // 第i个数据点
            // start your code here

            double error = 0;                       // 第i个数据点的计算误差, f(x)
            // 填写计算error的表达式
            error = yi - (ae * xi * xi + be * xi + ce); // @liyubo
            Vector3d J; // 雅可比矩阵 f()的导数
            //分别求导
            J[0] = -exp(ae * xi * xi + be * xi + ce) * xi; // de/da
            J[1] = -exp(ae * xi * xi + be * xi + ce); // de/db
            J[2] = -exp(ae * xi * xi + be * xi + ce); // de/dc

            H += J * J.transpose(); // GN近似的H, transpose求转置
            b += -error * J;

            // end your code here
            cost += error * error; //本次迭代
        }
    }
}

```

```

    }
    // 求解线性方程  $Hx=b$ , 建议用ldlt
    // start your code here
    Vector3d dx;
    dx = H.ldlt().solve(b);    //ldlt Cholesky来解方程

    // end your code here

    if (isnan(dx[0]))
    {
        cout << "result is nan!" << endl;
        break;
    }
    if (iter > 0 && cost > lastCost)
    {
        // 误差增长了, 说明近似的不够好
        cout << "cost: " << cost << ", last cost: " << lastCost << endl;
        break;
    }
    // 更新abc估计值
    ae += dx[0];
    be += dx[1];
    ce += dx[2];

    lastCost = cost;
    cout<<"ae: "<<ae<<" be:"<<be<<" ce:"<<ce<<endl;
    cout << "total cost: " << cost << endl;
}

cout << "estimated abc = " << ae << ", " << be << ", " << ce << endl;
return 0;

```