



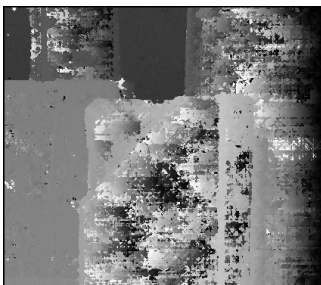
Semi-Global Matching Result & Report

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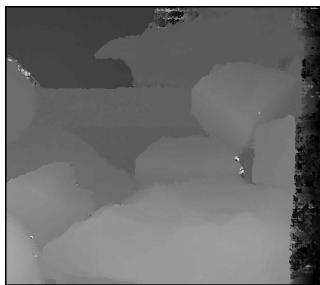
Results

To begin with, here I'm gonna show the results obtained before and after the refinement procedure.

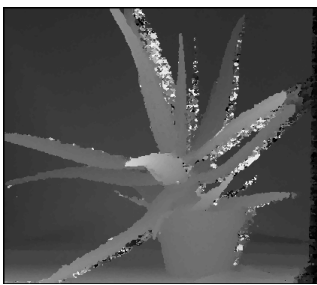
- **Results before refinement:**



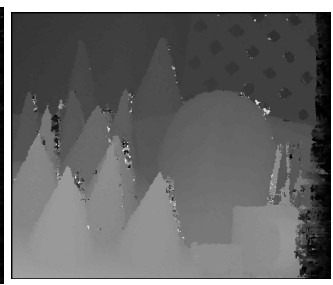
Plastic MSE: 820.049



Rocks MSE: 557.735

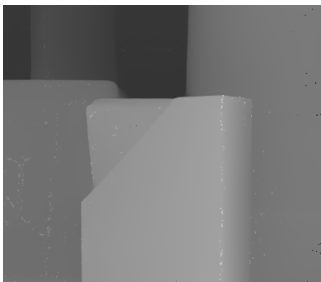


Aloe MSE: 122.452

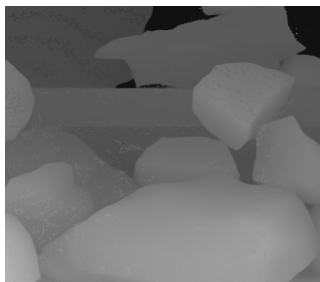


Cones MSE: 475.159

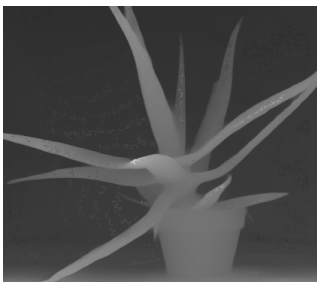
- **Results after refinement:**



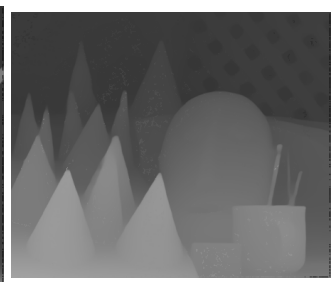
Plastic MSE: 348.223



Rocks MSE: 34.6984



Aloe MSE: 13.7204



Cones MSE: 17.382

Implementation

Calculate Path Cost

The function `compute_path_cost()` can be decomposed in two phases:

- I The first part checks if the pixel we are processing is the first one. If that is the case, we use the following equation:

$$E_{smooth}(p, q) = E(q, f_q)$$

where f_q is the disparity level of pixel q.

- II The second part computes the smoothness cost of the current pixel with the following equation:

$$E_{smooth}(p, q) = \min \begin{cases} E(q, f_q) & \text{if } f_p = f_q \\ E(q, f_q) + c_1 & \text{if } |f_p - f_q| = 1 \\ \min_{0 \leq \Delta \leq d_{max}} E(q, \Delta) + c_2(p, q) & \text{if } |f_p - f_q| > 1 \end{cases}$$

Aggregation

In the function `aggregation()` I've initialized the variables according to the desired direction and then allowed the computation of the path cost. The reasoning has been the following:

- If the x has direction equal to -1, we want to go from east to west with a step of -1.
- If the x has direction equal to +1, we want to go from west to east with a step of +1.
- If the y has direction equal to -1, we want to go from south to north with a step of -1.
- If the y has direction equal to +1, we want to go from north to south with a step of +1.

Compute Disparity

In the `compute_disparity()` I've implemented the following:

- I In the first part, we store the good disparities from the initial guess and from the disparity computed in the previous steps.
- II Then, in the second part, using the solution for non homogeneous systems, I've retrieved the h and k factors of the following equation:

$$d_{sgm} = h \times d_{mono} + k$$

Once Obtained the values, I've scaled and improved only the values that had a small confidence.

Code Snippets

```

if (cur_y == pw._north || cur_y == pw._south || cur_x == pw._east || cur_x == pw._west) {
    //Please fill me!
    for (unsigned int d = 0; d < disparity_range_; d++) {
        path_cost[cur_path][cur_y][cur_x][d] = cost[cur_y][cur_x][d];
    }
} else {
    //Please fill me!
    best_prev_cost = path_cost[cur_path][cur_y-direction_y][cur_x-direction_x][0];
    for(int d = 0; d < disparity_range_; d++){
        prev_cost = path_cost[cur_path][cur_y - direction_y][cur_x - direction_x][d];
        no_penalty_cost = prev_cost;
        if (d == 0) {
            small_penalty_cost = path_cost[cur_path][cur_y - direction_y][cur_x - direction_x][d + 1] + p1_;
        } else if (d == disparity_range_ - 1) {
            small_penalty_cost = path_cost[cur_path][cur_y - direction_y][cur_x - direction_x][d - 1] + p1_;
        } else {
            if (path_cost[cur_path][cur_y - direction_y][cur_x - direction_x][d + 1] >=
                path_cost[cur_path][cur_y - direction_y][cur_x - direction_x][d - 1])
                small_penalty_cost = path_cost[cur_path][cur_y - direction_y][cur_x - direction_x][d - 1] + p1_;
            else
                small_penalty_cost = path_cost[cur_path][cur_y - direction_y][cur_x - direction_x][d + 1] + p1_;
        }
        best_prev_cost = 100000;
        for (unsigned int dd = 0; dd < disparity_range_; dd++) {
            if (int(dd - d) >= 2 || int(dd - d) <= -2) {
                if (best_prev_cost > path_cost[cur_path][cur_y - direction_y][cur_x - direction_x][dd])
                    best_prev_cost = path_cost[cur_path][cur_y - direction_y][cur_x - direction_x][dd];
            }
        }
        big_penalty_cost = best_prev_cost + p2_;
        unsigned long temp = min(no_penalty_cost, small_penalty_cost);
        path_cost[cur_path][cur_y][cur_x][d] = cost[cur_y][cur_x][d] + min(temp, big_penalty_cost);
    }
}

```

CODE TO BE IMPLEMENTED 1/4

```

switch (dir_x)
{
    case -1:
        start_x=pw._east;
        end_x=pw._west;
        step_x=-1;
        break;
    default:
        start_x=pw._west;
        end_x=pw._east;
        step_x=1;
        break;
}
switch (dir_y)
{
    case -1:
        start_y=pw._south;
        end_y=pw._north;
        step_y=-1;
        break;
    default:
        start_y=pw._north;
        end_y=pw._south;
        step_y=1;
        break;
}
for (int y = start_y; y != end_y; y += step_y) {
    for (int x = start_x; x != end_x; x += step_x) {
        compute_path_cost(dir_y, dir_x, y, x, cur_path);
    }
}

```

CODE TO BE IMPLEMENTED 2/4

```

d_mono.push_back(float(mono_.at<uchar>(row, col)));
d_sgm.push_back(smallest_disparity);

```

CODE TO BE IMPLEMENTED 3/4

```

float h, k;
Mat A, transpose_A, b(d_sgm);
const Mat vector_of_ones = Mat_<float>::ones(1, d_mono.size());
cv::vconcat(d_mono, vector_of_ones, transpose_A);
A = transpose_A.t();
Mat temp_x, temp_x_2, x;
temp_x = transpose_A * A;
temp_x_2 = temp_x.inv() * transpose_A;
x = temp_x_2 * b;
h = x.at<float>(0,0);
k = x.at<float>(0,1);
for (int row = 0; row < height; ++row) {
    for (int col = 0; col < width; ++col) {
        if (inv_confidence[row][col] <= 0 || inv_confidence[row][col] >= conf_thresh_) {
            disp_.at<uchar>(row,col) = (mono_.at<uchar>(row,col)* h +k)* 255 / disparity_range_;
        }
    }
}

```

CODE TO BE IMPLEMENTED 4/4