Introduction

Our project is based on 'A Stereo Matching Algorithm with an Adaptive Window: Theory and Experiment' by Takeo Kanade and Masatoshi Okutomi.

Link for the same is:

https://www.researchgate.net/publication/224749504_A_stereo_matching_algorithm_with_an_adaptive_window_Theory_and_experiment

Link for our code

https://github.com/mssaboo/Stereo-Matching

Iterative Stereo Algorithm with an Adaptive Window :

- 1. Start with and initial disparity estimate $d_0(x, y)$.
- For each point (x, y), we want to choose a window that provides the estimate
 of disparity increment having the lowest uncertainty. For the chosen window,
 calculate the disparity increment(Δd) first and update the disparity estimate by
 d_{i+1} = d_i + Δd.
- and width can be controlled independently giving least disparity uncertainty.

 (a) Place a small 3 x 3 window centered at the pixel, and compute the

Now we need to select a window which would be rectangular and its height

uncertainty.

(b) Expand the window by one pixel in one direction, e.g., to the right x+, for

trial, and compute the uncertainty for the expanded window. If the expansion increases the uncertainty, the direction is prohibited from further expansions.

- Repeat the same process for each of the four directions z+, x--, y+, and y(excluding the already prohibited ones).
- (c) Compare the uncertainties for all the directions tried and choose the direction which produces the minimum uncertainty.
- (d) Expand the window by one pixel in the chosen direction.(e) Iterate steps (b) to (d) until all directions become prohibited from expansion or until the window size reaches a limit that is previously set.

Our Implementation : First we initialised our node. We give our parameters which are our left and right images giving suitable path.

Then we create image publisher object using in-built image_transport library.

Next we read the image and convert it into a ROS message. This step we done twice for both left and right image.

We convert our left and right CV images into grayscale images.

Next we define two matrices, for each image and one more for disparity which is initialized to zero.

We set our several parameters which are minimum and maximum disparity and half-block size. We define our disparity range as the difference of maximum disparity and minimum disparity.

The pseudo-code for calculating SSD is as shown:

```
ssd = 0
for i = 0 to height - 1
   for j = 0 to width - 1
        diff = A[i][j] - B[i][j]
        ssd += diff * diff
```

The general idea is that for matching images the SSD will be small. If you're trying to match two images, where one image is translated by some amount, then you would typically do a brute force approach where you calculate the SSD over a range of x, y displacements and then identify the minimum SSD value, which should then correspond to the best alignment offset.

SSD is generally only used due to its simplicity.

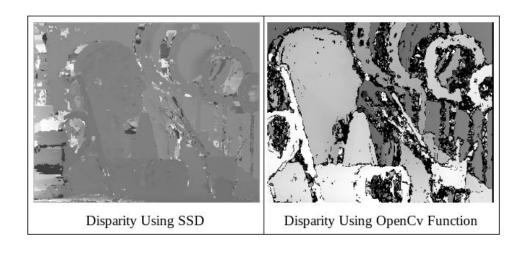
We calculate SSD on values of RGB in corresponding pixels in left and right images and then store it in matrix of SSD value if it is less than the previous value and also update the disparity matrix with the current range of disparity. We then normalise our matrix of disparity values.

This matrix was converted into image using OpenCV and then was published as ROS message. This was the implementation of SSD in our code for getting disparity matrix.

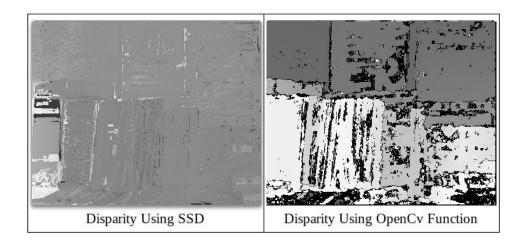
We also used the inbuilt OpenCV library to get disparity map and compared the results.

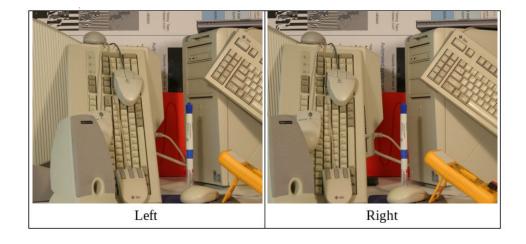
Image	Topic
Left image	/image_l
Right image	/image_r
Disparity using SSD	/image_disp_ssd
Disparity using inbuilt function	/image_disp_ib

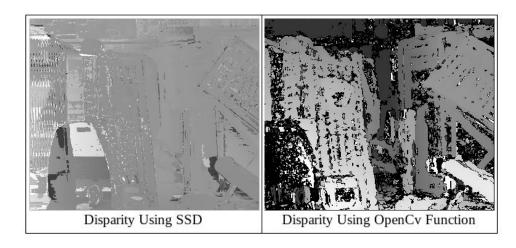




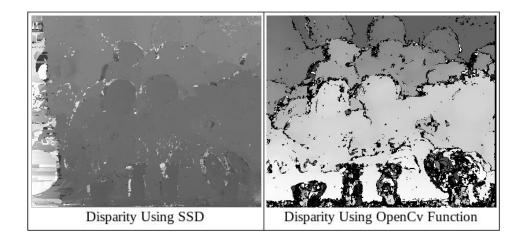




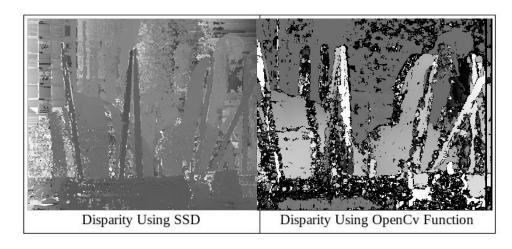




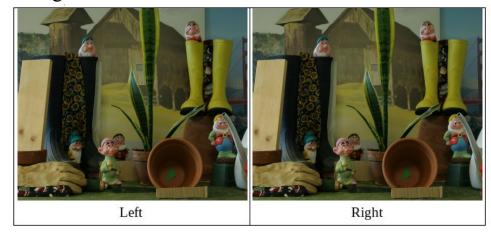


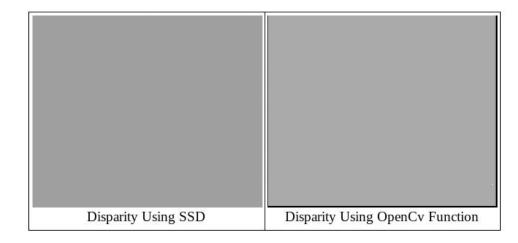






Results: 6 When both images are same





Thank You

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