EE 610: Image Processing Final Project

Automatic Image Stitching

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Understanding the Problem

Why to stitch?

- matching of panoramic image - despite rotation, zoom and illumination

Problems of image Stitching:

- To Register images so their features align
- Matching different images to stitch into one
- Blending images to hide stitching marks

Algorithm: Panoramic Recognition

I. Extract SIFT features from all n images

Input: n unordered images

II. Find k nearest-neighbours for each feature using a k-d tree

III. For each image:

 Select m candidate matching images (with the maximum number of feature matches to this image) (ii) Find geometrically conssistent feature matches

using RANSAC to solve for the homography between pairs of images

(iii) Verify image matches using probabilistic model

IV. Find connected components of image matches

V. For each connected component:

(i) Perform bundle adjustment to solve for the rota-

tion $\theta_1, \theta_2, \theta_3$ and focal length f of all cameras (ii) Render panorama using multi-band blending

Output: Panoramic image(s)

Feature Matching: SIFT

SIFT:

Extract and match SIFT [Low04] features between all of the images.

Characteristic scale and orientation is established at each feature location

K-d tree:

Each feature is matched to its k nearest neighbours in feature space (we use k = 2)

$$\tilde{\mathbf{u}}_i = \mathbf{H}_{ij}\tilde{\mathbf{u}}_j$$

SIFT





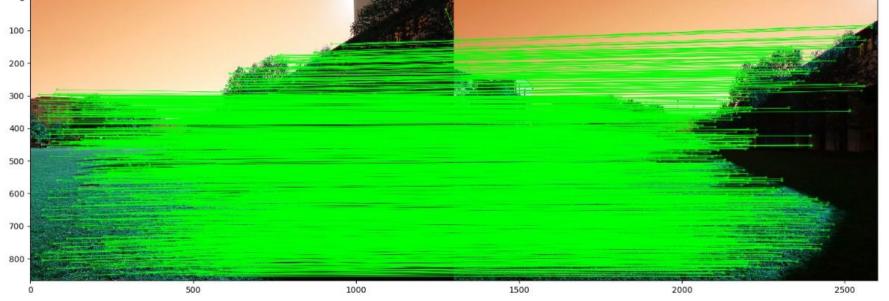


Image Matching - RANSAC

RANSAC is essentially a sampling approach to estimating H

$$p(\mathbf{H} \ is \ correct) = 1 - (1 - (p_i)^r)^n$$

N - no of trials

Pi - probability that a feature match is correct between a pair of matching images

R- feature correspondances(=2)

After 500 trials, $p = 10^{-14}$







(b) Image 2



(c) SIFT matches 1



(e) RANSAC inliers 1

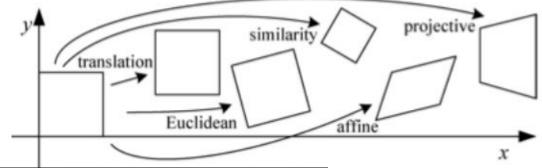


(d) SIFT matches 2



(f) RANSAC inliers 2

Homography



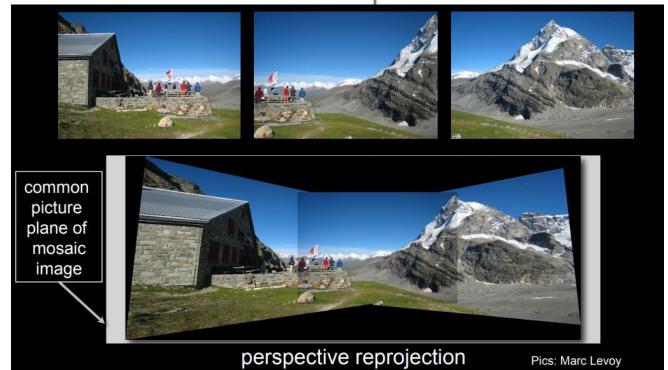


Image Matching - Probabilistic Model

Condition:

$$\frac{B(n_i; n_f, p_1)p(m=1)}{B(n_i; n_f, p_0)p(m=0)} \underset{reject}{\overset{accept}{\geq}} \frac{1}{\frac{1}{p_{min}} - 1}.$$

$$p(f^{(1:n_f)} | m = 1) = B(n_i; n_f, p_1)$$

$$p(f^{(1:n_f)} | m = 0) = B(n_i; n_f, p_0)$$

$$B(x; n, p) = \frac{n!}{x!(n-x)!} p^{x} (1-p)^{n-x}.$$

Choosing values $p(m = 1) = 10^{-6}$ and $p_{min} = 0.999$ gives the condition

$$n_i > \alpha + \beta n_f \tag{13}$$

for a correct image match, where $\alpha = 8.0$ and $\beta = 0.3$.



Multi-band Blending

Given two images we decide the overlapping regions.



Laplacian Pyramid is formed of the two images to be overlapped by finding

the DoG
$$B^i_{(k+1)\sigma} = I^i_{k\sigma} - I^i_{(k+1)\sigma}$$

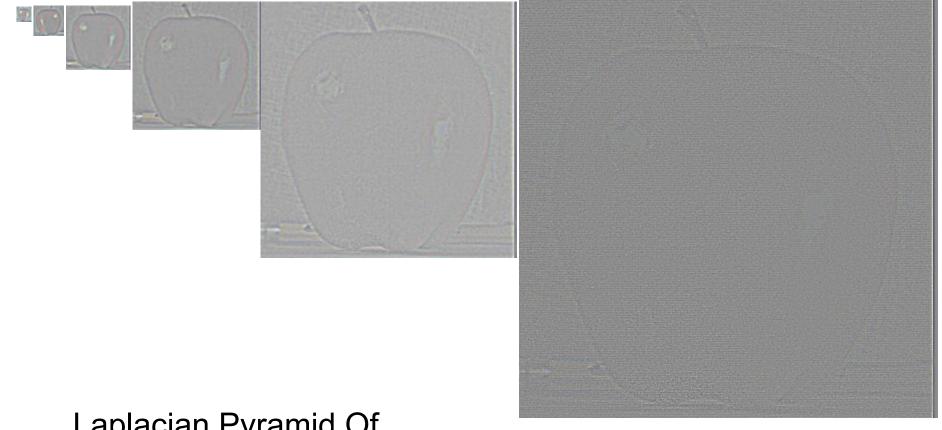
$$I^i_{(k+1)\sigma} = I^i_{k\sigma} * g_{\sigma'}$$

$$W^i_{(k+1)\sigma} = W^i_{k\sigma} * g_{\sigma'}$$

$$W^i_{max}(\theta,\phi) = \begin{cases} 1 & \text{if } W^i(\theta,\phi) = \arg\max_j W^j(\theta,\phi) \\ 0 & \text{otherwise} \end{cases}$$

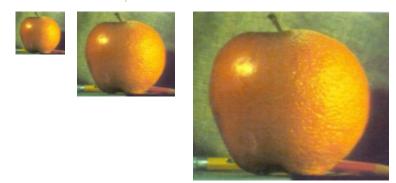
Gaussian Pyramid of the mask is formed

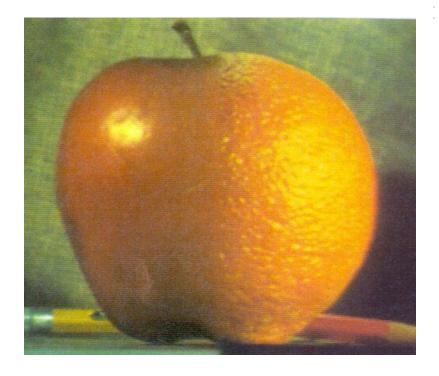
Laplacian Pyramid Gaussian Pyramid expand expand

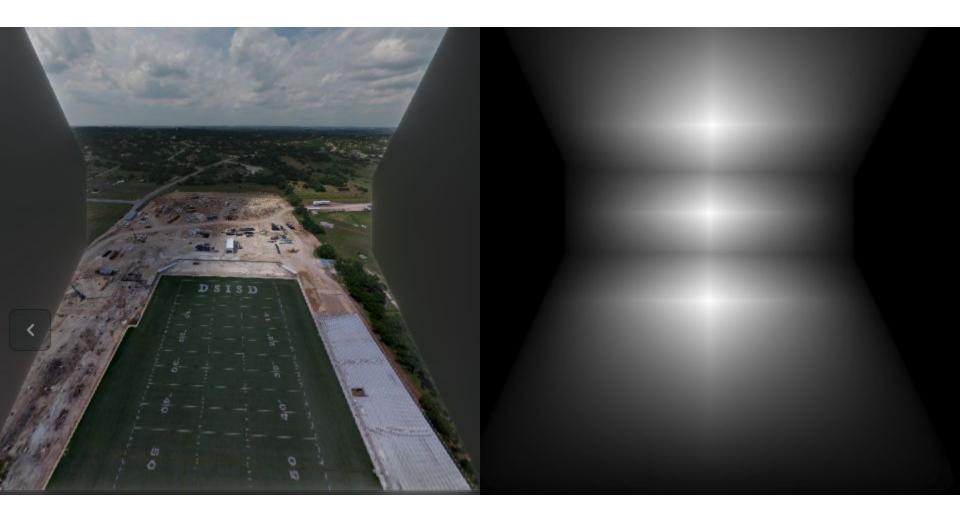


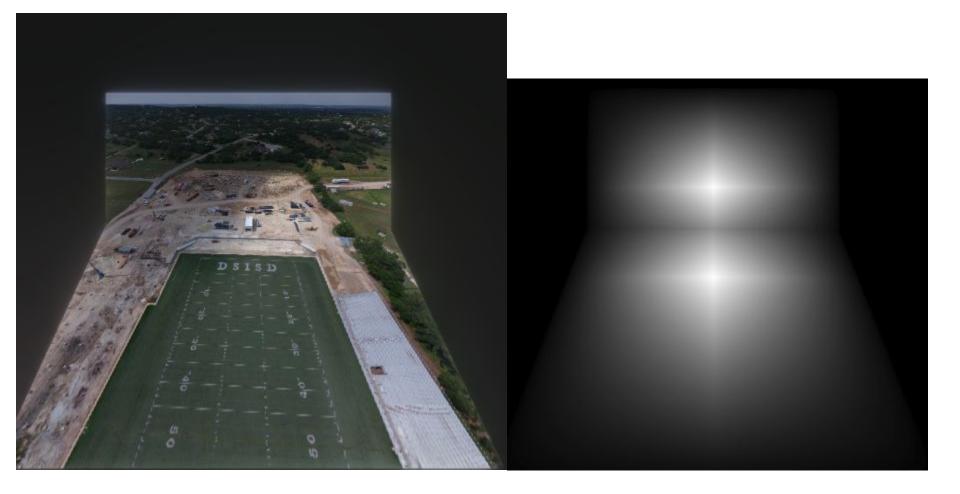
Laplacian Pyramid Of Apple images

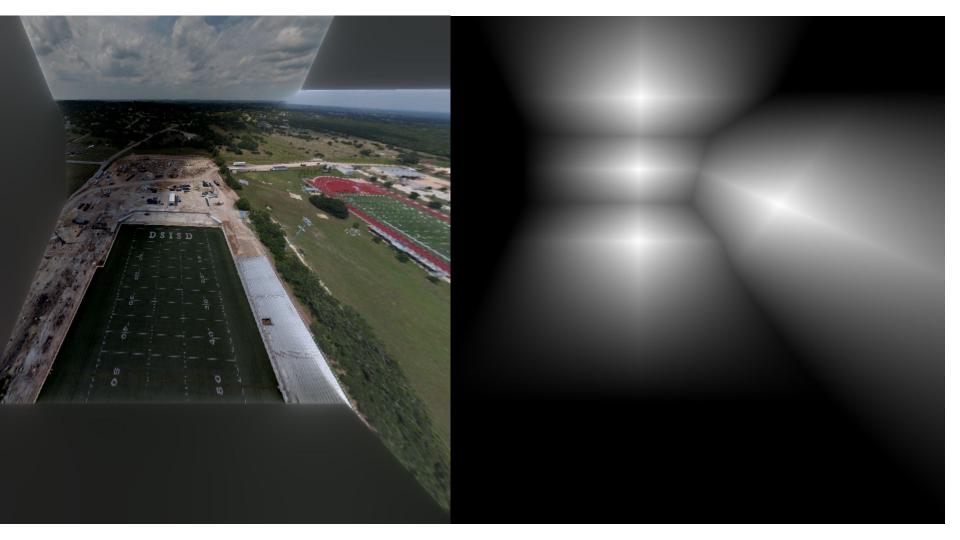
- Each pyramid is Blended by weighting the pyramids with the gaussian pyramids of the mask.
- Final image is formed by reconstructing the images from the pyramid.

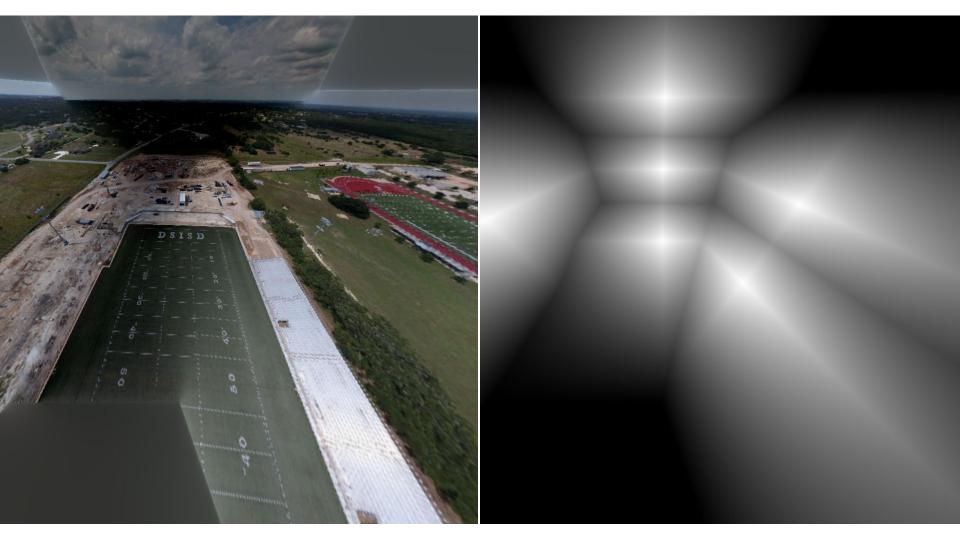


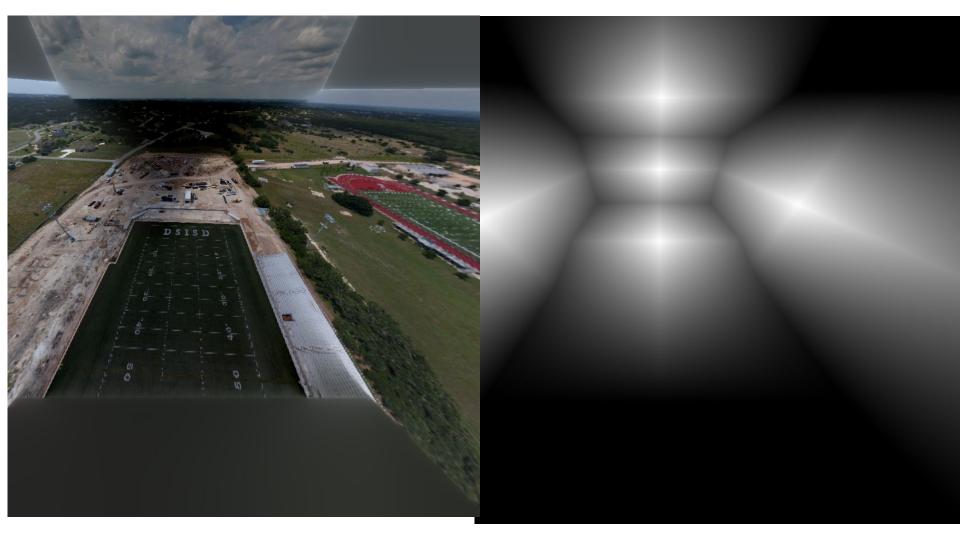


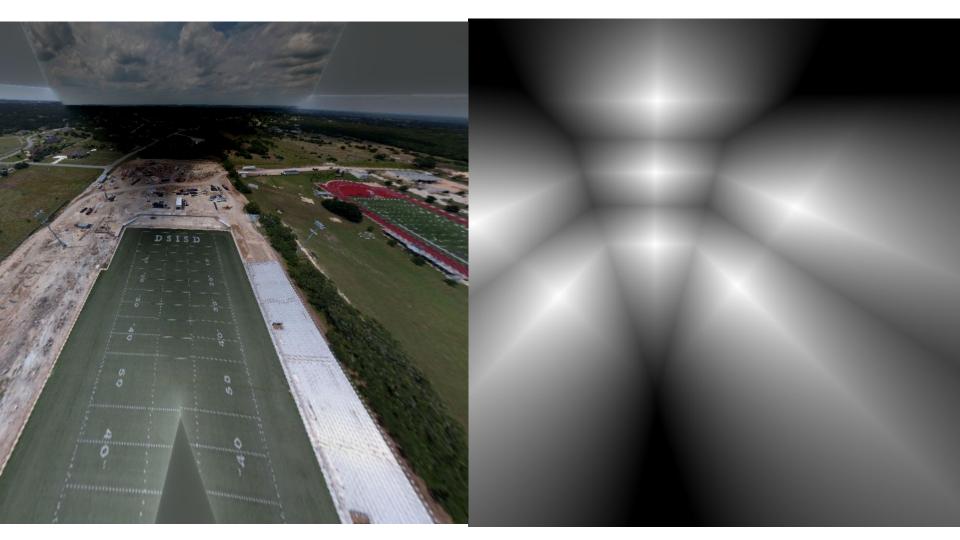


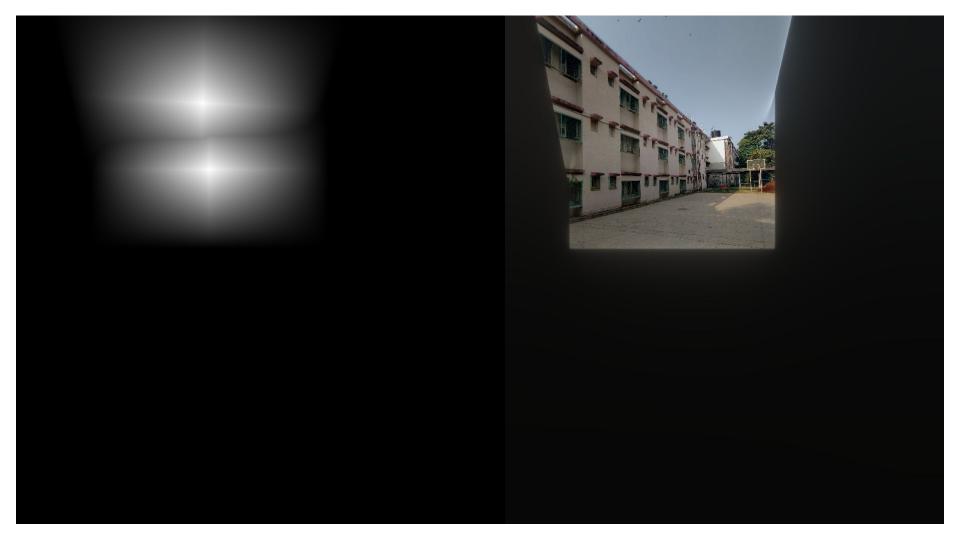


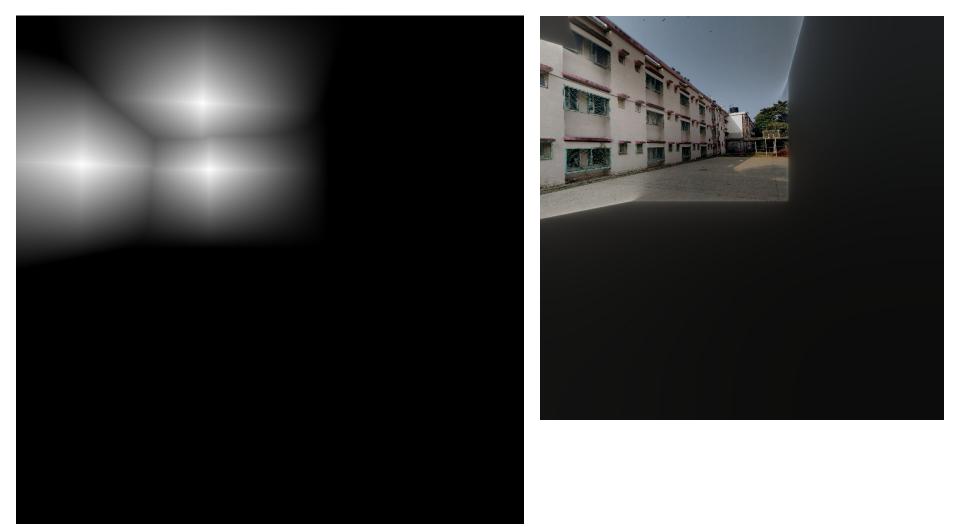


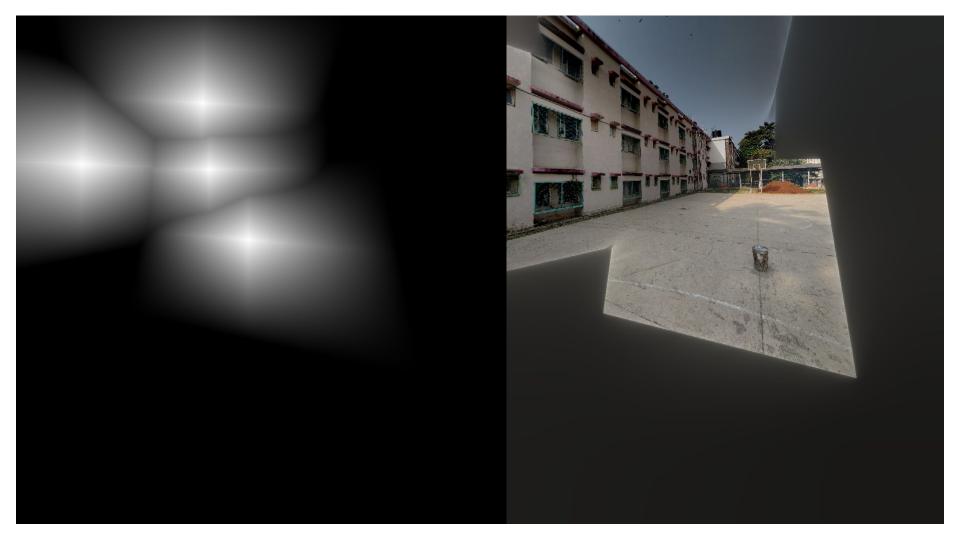


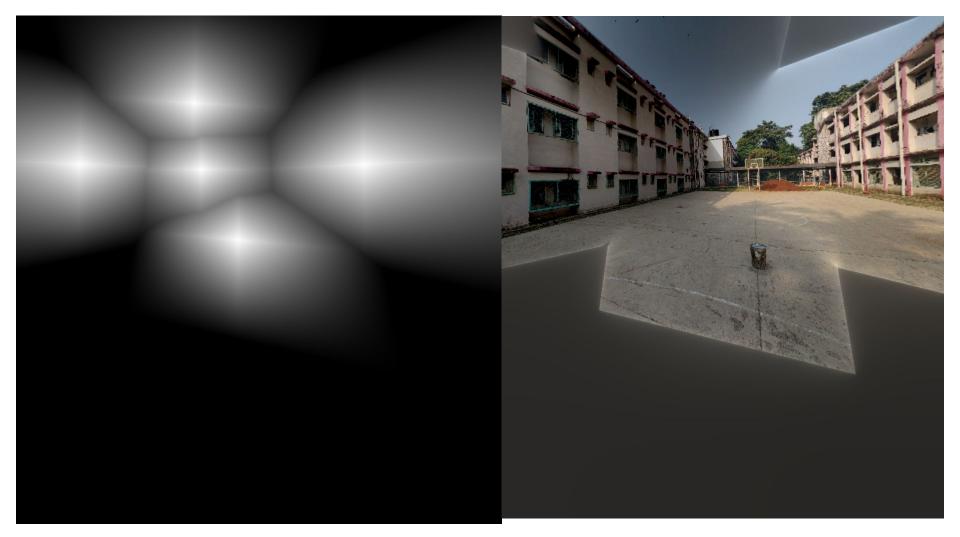


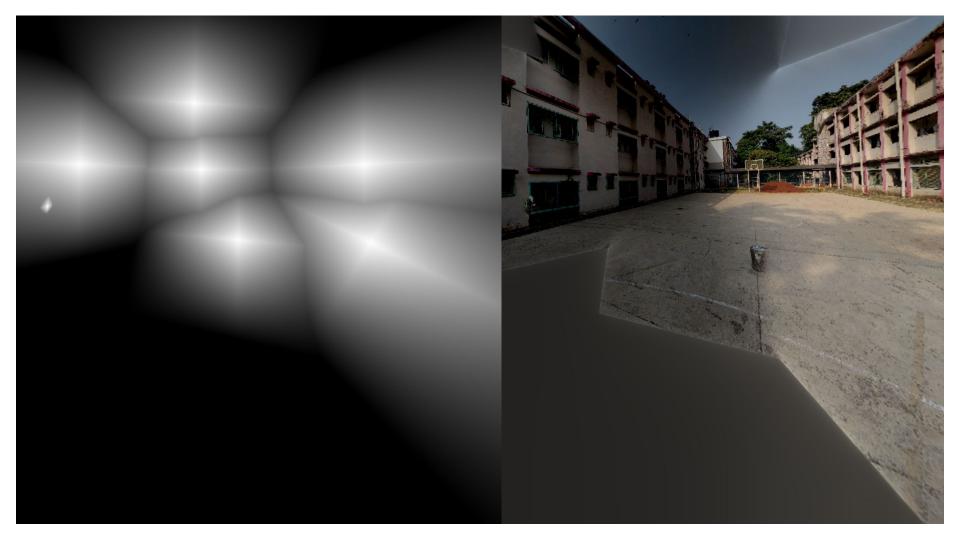






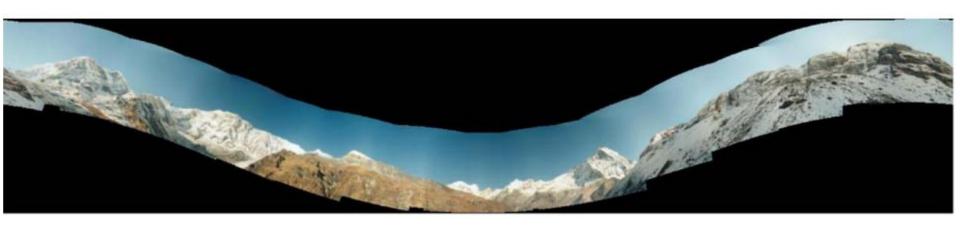






Future scope

- Scene motion
- Automatic panoramic straightening



Photometric modelling (decrease in intensity towards image edges)

Learning Points

- Implemented theoretical concepts of image processing.
- Understanding papers and referencing multiple papers.
- Harnessing synergy of team members towards a common goal.

Thank You