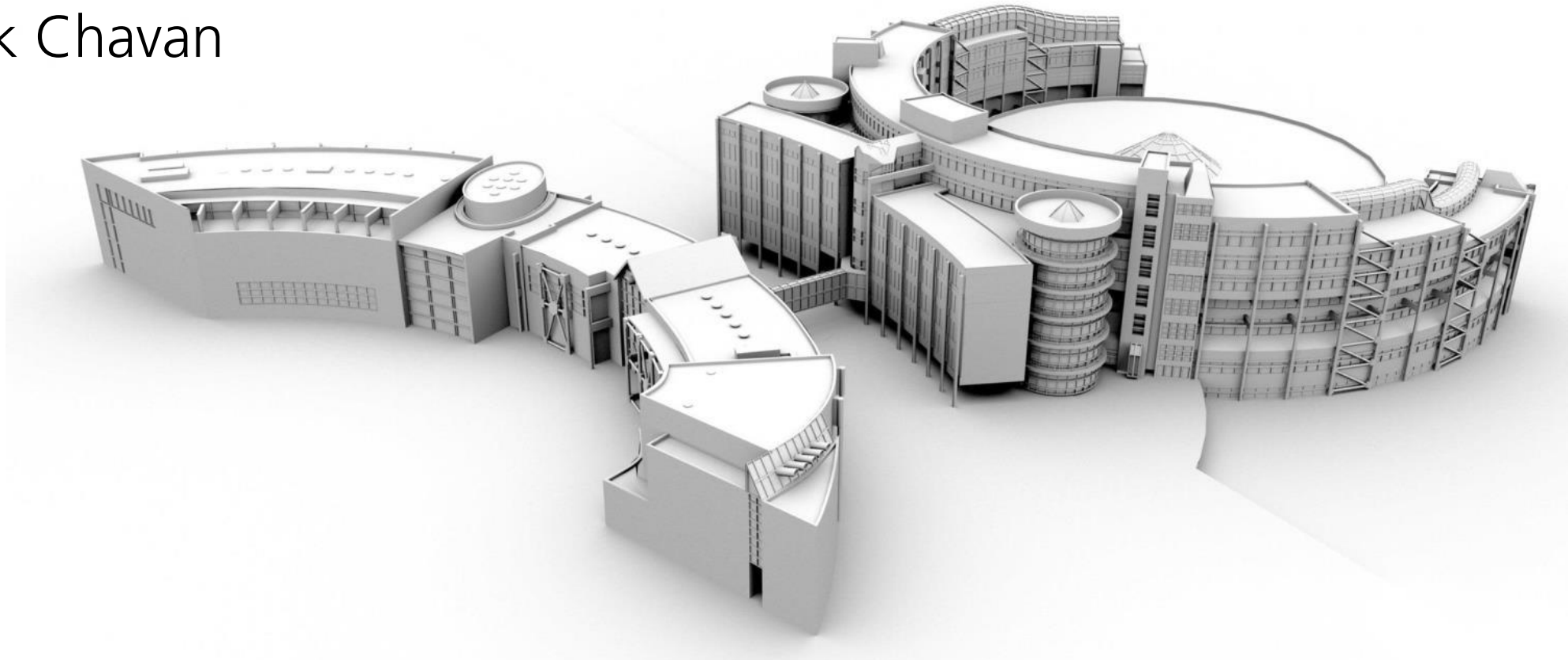


# GREEN INCREMENTAL LEARNING

## EIBA 10 Camera System: Dataset Sampling

-Vivek Chavan



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# AGENDA

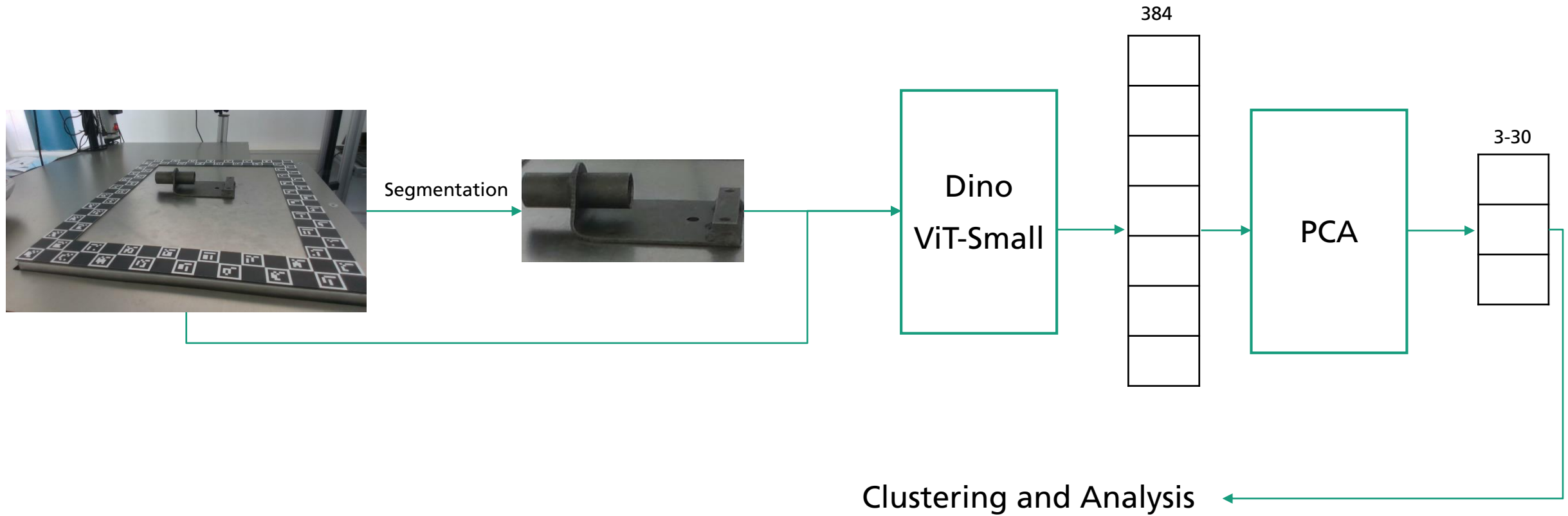
- I. 10 Camera & 12 rotation system
- II. Uncropped Image Analysis
- III. Segmented Image Analysis
- IV. Inferences for the Camera positions
- V. Inferences for the rotations
- VI. Outlier detection
- VII. Further work

# 10 CAMERA & 12 ROTATION SYSTEM

- $(10 \text{ cameras}) \times (12 \text{ rotations}) = 120$  images per object (only the primary orientation was considered)

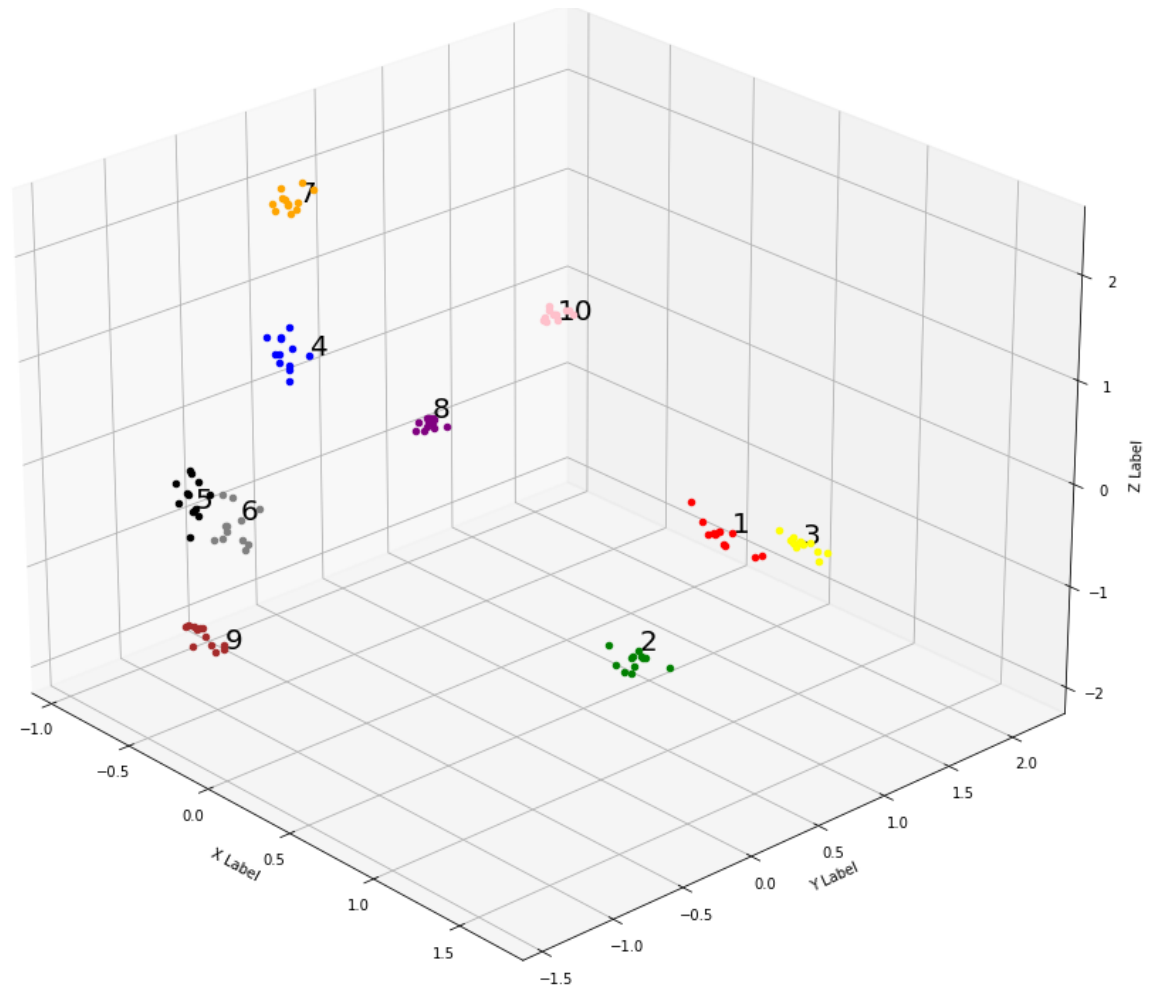
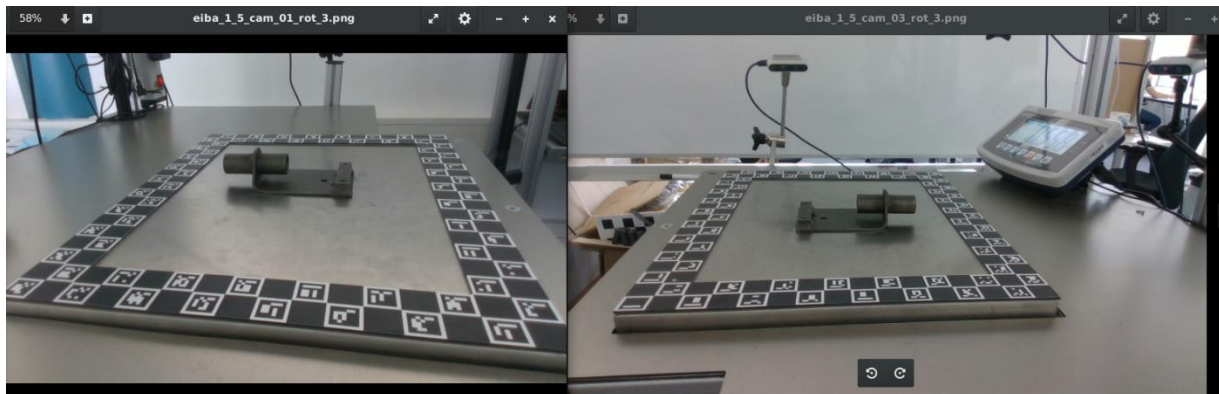


# APPROACH



# UNCROPPED IMAGE ANALYSIS

- Clear separation between the camera positions can be seen
- Cameras 1 and 3 produce close embeddings, along with cameras 5 and 6
  - Closer clusters correspond to similarity in the images and features represented



# UNCROPPED IMAGE ANALYSIS

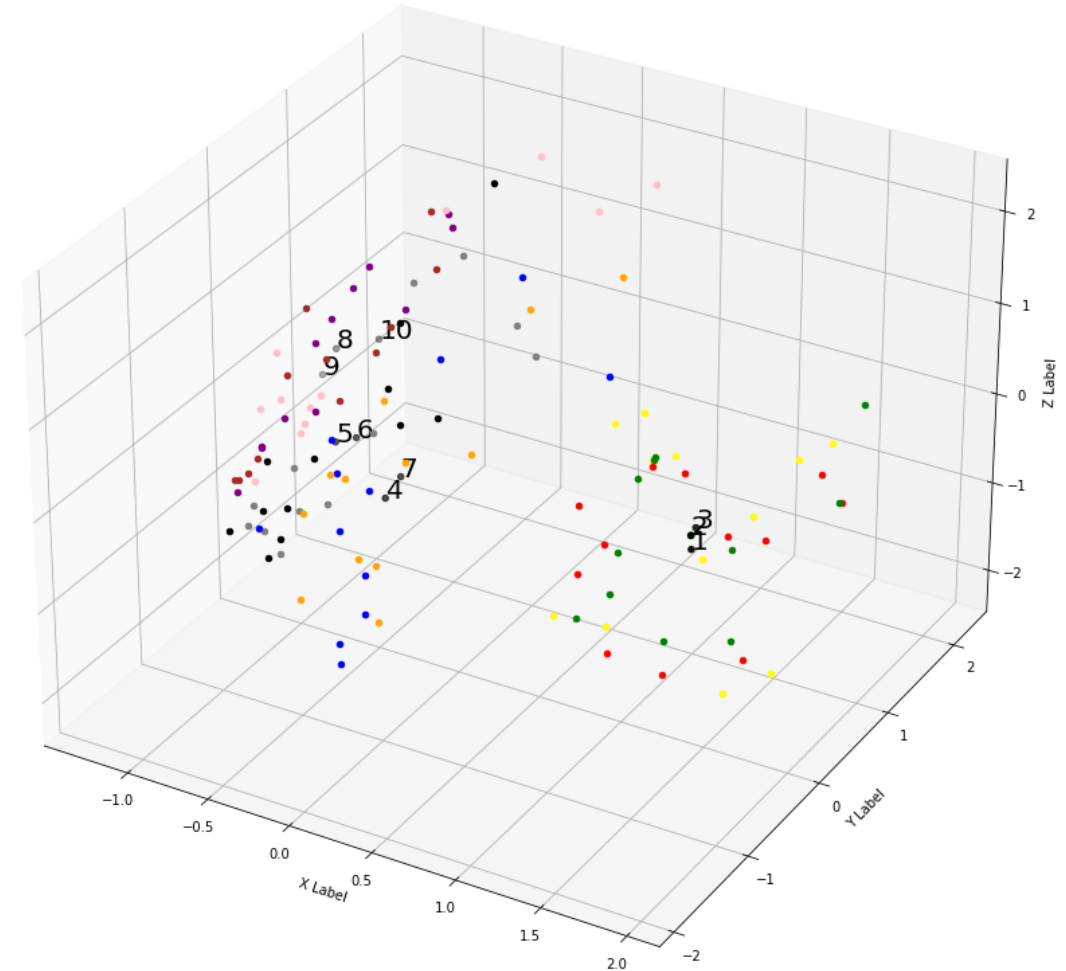
- Applications:
  - Finding and eliminating duplicate camera representations
    - E.g.: Only 1 camera could be selected out of Cameras 1, 2 and 3. All features are still captured given that the object is rotated 12x
    - E.g.: For symmetrical/axisymmetric objects, 12 rotations produce multiple duplicate images
  - Selecting a few cameras for maintaining maximum variance
    - Only 4 cameras: 1, 4, 6, 10
    - 7 cameras: 1, 4, 6, 7, 8, 9, 10





# SEGMENTED IMAGE ANALYSIS

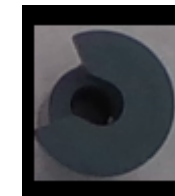
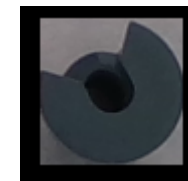
- The separation between different camera locations is less prominent
- The relative locations of different embeddings remains the same
- In the figure we see the labels at the center of the different camera locations
- Cameras 1, 2 and 3 are very close, which means that they produce very similar images (considering all 12 rotations)
- Cameras 5 and 6 are also close



# SEGMENTED IMAGE ANALYSIS

## APPLICATIONS:

- Embeddings that are very close to each other signify very similar images.
- These can belong to the same camera or different cameras
- Image 1: Cameras 1 and 3
- Image 2: Different rotations of camera 9

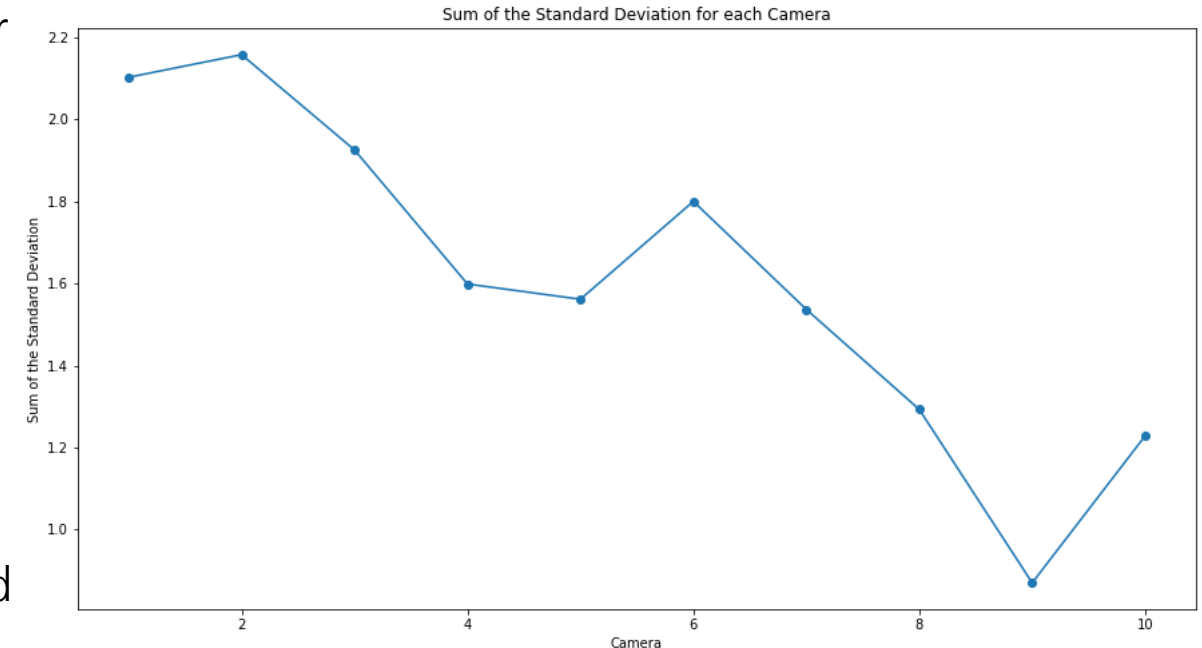




# SEGMENTED IMAGE ANALYSIS

## ROTATIONS AND FEATURE VARIANCE

- The plot shows the sum of the Standard Deviation for different camera locations
- Hypothesis:
  - A high score means that the 12 rotations of the object add a lot of variation to the captured features
  - A low score means that the 12 rotations do not add much in terms of new features.
  - The cameras with low scores (8, 9, 10) are located at the top of the setup, which is why more rotations often do not expose new features of the object



# SEGMENTED IMAGE ANALYSIS

## ROTATIONS AND FEATURE VARIANCE

### ■ Explanation:

- Sum of std is just reducing the 384 dim to 3 using PCA
- The standard deviation along the three dimensions is then summed up (for each camera) to get a single value.
- It is a hypothesis, that clusters with low std are closer together and have less variance in the data. Hence, rotating the object 12x did not add more variation to the images.
- We also know this instinctively that cameras using the top view will see the same perspective with each rotation



Rotations for camera 1

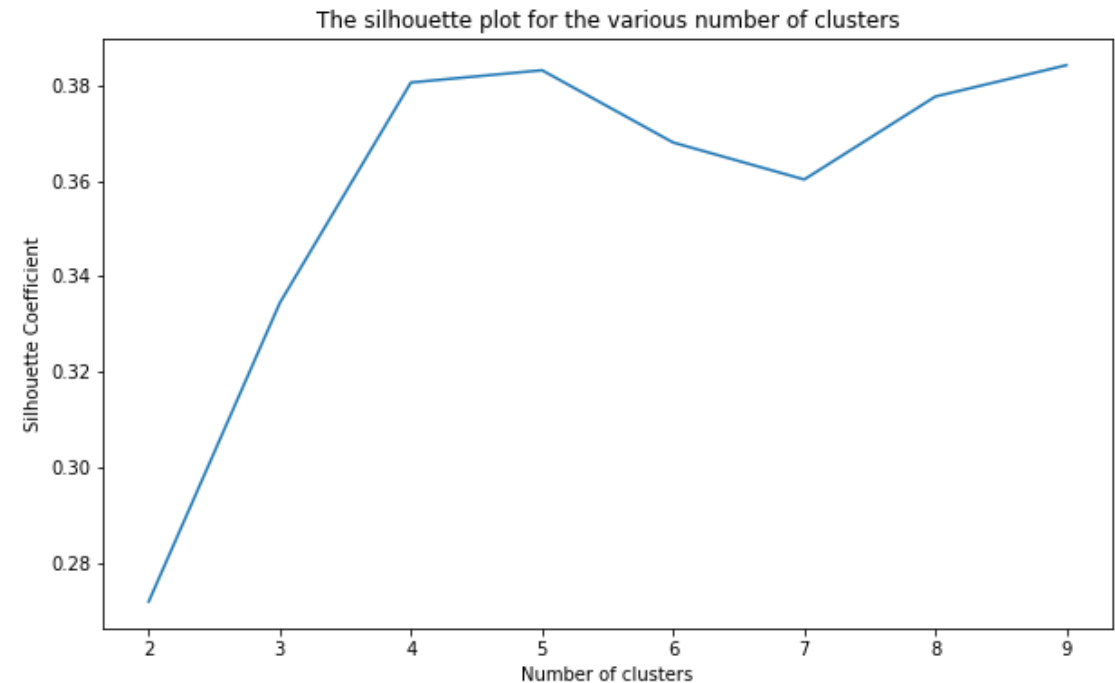


Rotations for camera 9

# SEGMENTED IMAGE ANALYSIS

## SILHOUETTE SCORE

- The Silhouette score indicates the best possible number of clusters
- Hypothesis:
  - It can also be used to get the optimal number of camera configurations to ensure the same feature representations with less cameras
  - E.g. For this object, 4/5 cameras can be used to get the same feature space representation
- The score distribution varies across different objects
- The hypothesis needs to be proven and verified

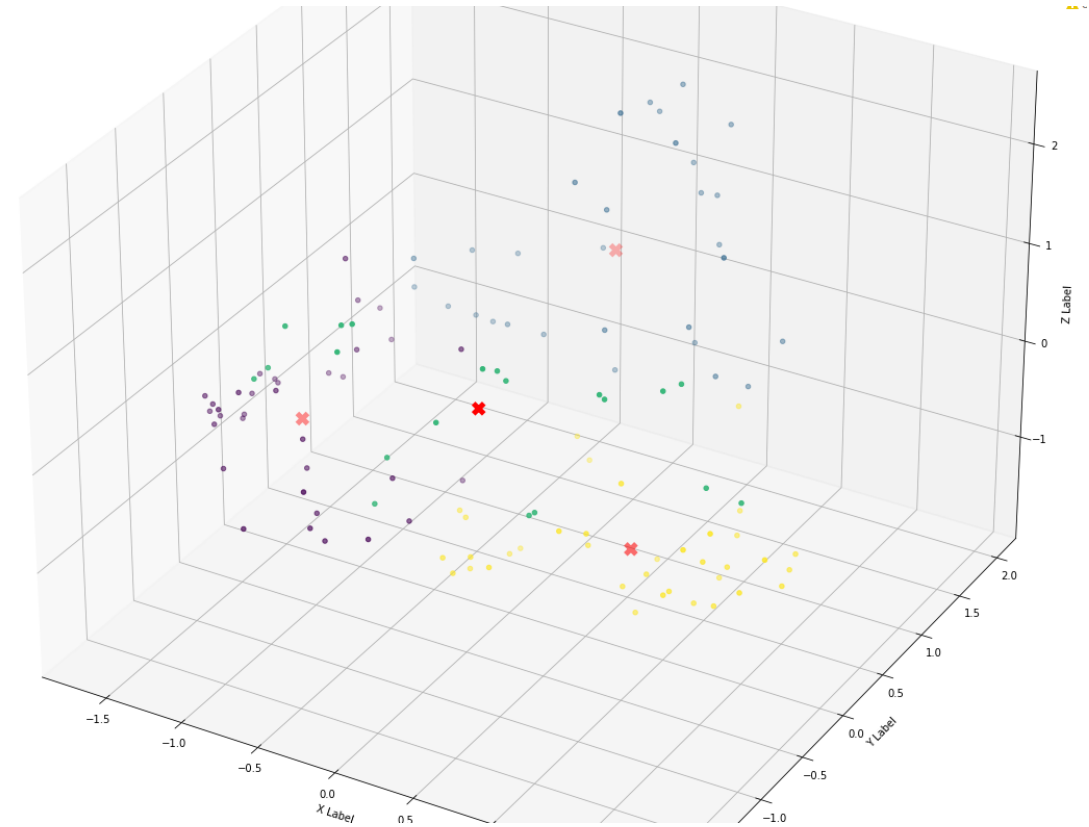


# SEGMENTED IMAGE ANALYSIS

## SILHOUETTE SCORE

### ■ Explanation

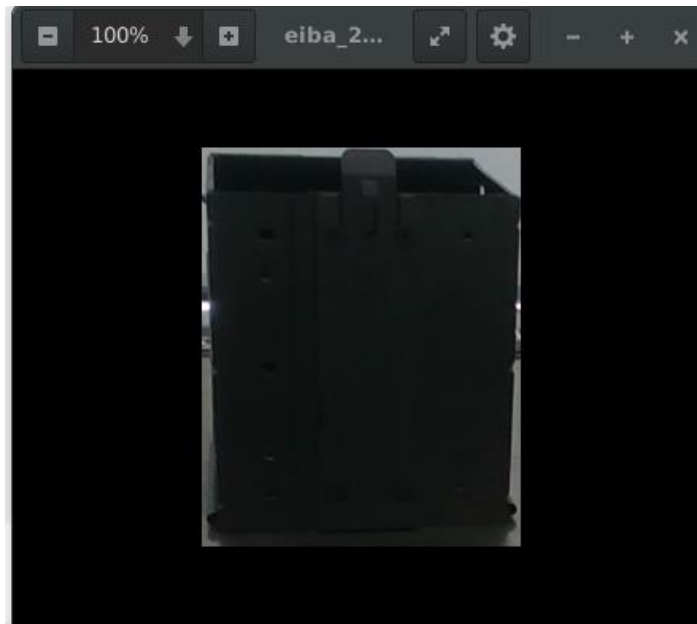
- Silhouette score is meant to get the optimal number of clusterings,
- We used it predict the minimum number of camera 'clusters' to still get the same variation with fewer cameras
- Silhouette score can potentially also be used to cluster images along rotation angles to derive further insights. This needs to be investigated using a larger dataset.



Only 4 clusters = Cameras: 1, 4, 6 and 10

# SEGMENTED IMAGE ANALYSIS

- Outlier identification: Gave accurate results for some objects
- The image below does not show the cutouts in the surfaces like other images



eiba\_2\_6\_cam\_01\_rot\_4.png



eiba\_2\_6\_cam\_01\_rot\_5.png



eiba\_2\_6\_cam\_01\_rot\_6.png



eiba\_2\_6\_cam\_01\_rot\_7.png



eiba\_2\_6\_cam\_01\_rot\_8.png

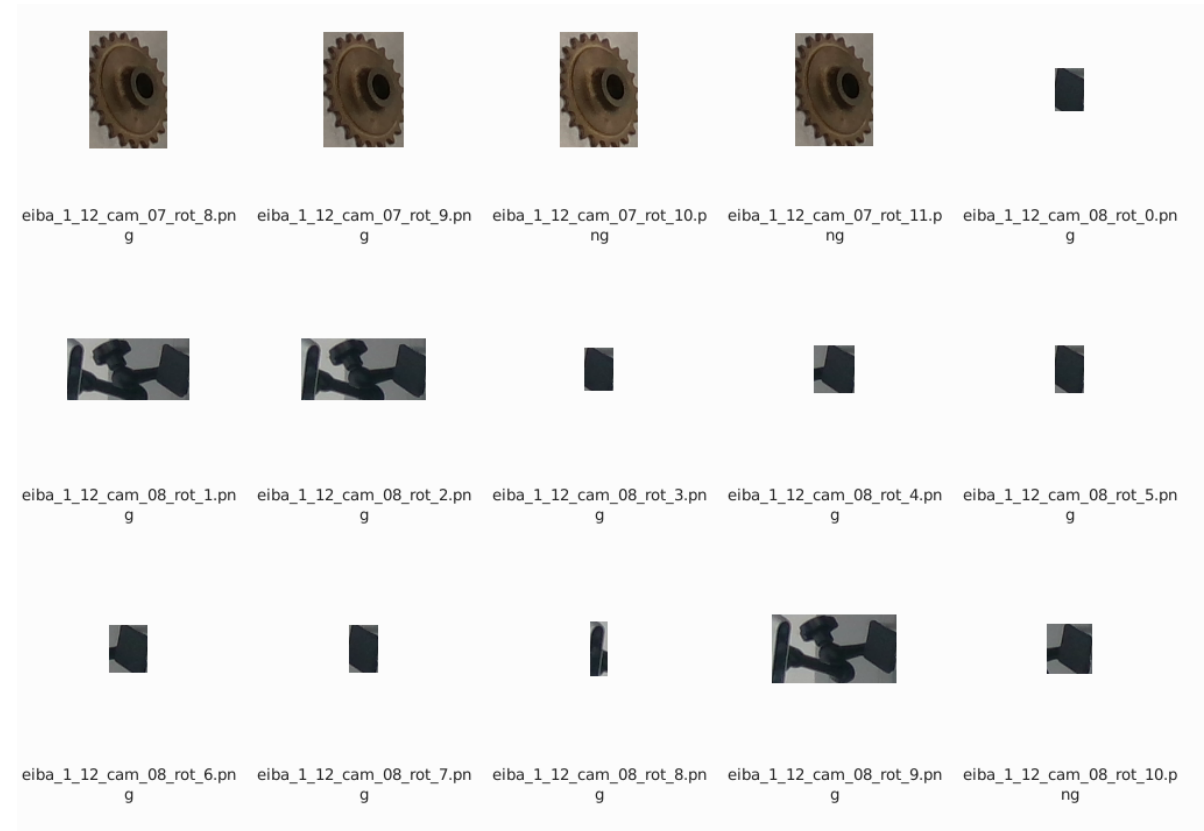


eiba\_2\_6\_cam\_01\_rot\_9.png



# SEGMENTED IMAGE ANALYSIS

- Outlier identification: Was not accurate for other objects. The segmentations were also not all accurate and may have contributed to this



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# FURTHER WORK

- Testing out the hypothesis
- Studying the approach on properly segmented images (many objects had incorrect segmentations)
- Deriving more insights