



# 1 Methods

## 1.1 Computing Feature Vectors

## 1.2 Feature Normalization

# 2 Visualizations

## 2.1 Successful Segmentations

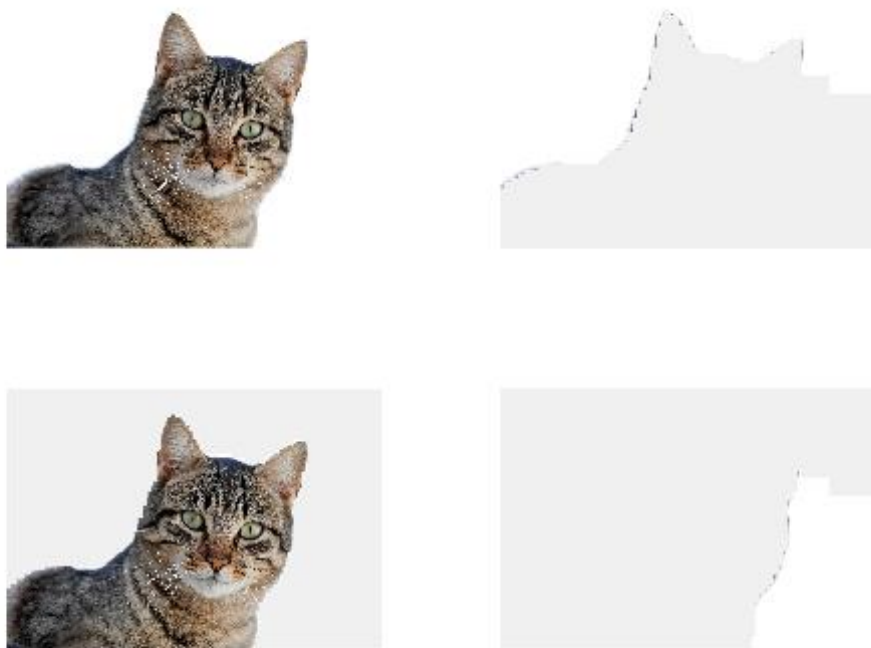


Figure 1: `cat_march.jpg`, using HAC with  $k = 3$ , position + color features, feature normalization, and a resize factor of 0.025.



Figure 2: `Cat_Bed.jpg`, using k-means clustering with  $k = 4$ , position + color features, and feature normalization.



Figure 3: `black_kitten_star.jpg`, using k-means clustering with  $k = 3$ , color features, and no feature normalization.

## 2.2 Unsuccessful Segmentations

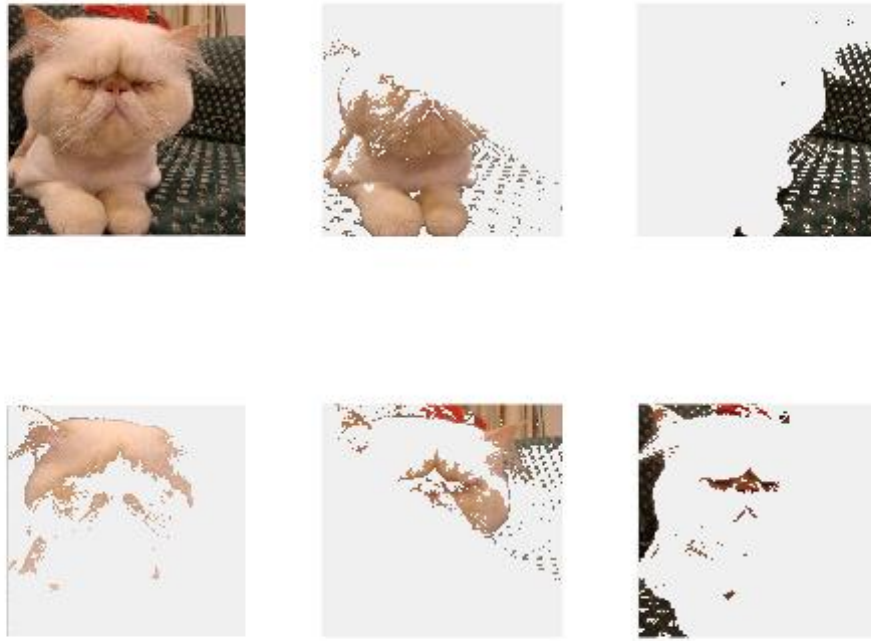


Figure 4: `cat_grumpy.jpg`, using k-means clustering with  $k = 5$ , position + color features, and no feature normalization.



Figure 5: `cat-jumping-running-grass.jpg`, using k-means clustering with  $k = 3$ , color features, and feature normalization.



Figure 6: kitten16.jpg, using HAC with  $k = 3$ , color features, feature normalization, and a resize factor of 0.25.

### 2.3 Composite Images

Using the script titled `GrabCat.m`, we were able to produce composite images by transferring segments from one image to another background image. This allowed us to create the two composite images shown below.



Figure 7: Input: `black_kitten_star.jpg`, `desert.jpg`, using k-means clustering with  $k = 3$ , color features, and feature normalization.



Figure 8: Input: `black_kitten.jpg`, `beach.jpg`, using HAC with  $k = 5$ , color features, feature normalization, and a resize factor of 0.2.



### 3 Evaluation