Mean Shift

Aim

The aim of this project is to implement Mean Shift clustering for color images, where the bandwidth is the only input provided by the user. The algorithm groups similar colors in the RGB space using a Gaussian kernel and outputs a final clustered version of the image. Additionally, the implementation includes:

- Visualization I: Animation of the clustering process in both 2D image space and 3D RGB color space as a gif
- Visualization II: Visualization of the 3D color density topography of an image to see the color distribution.

Implementation

Core Algorithm

mean_shift_color_pixel performs the actual mean shift procedure.

- Input:
 - flattened array of RGB pixel values
 - o bandwidth, which controls the size of the neighborhood
- For every iteration:
 - the color of every pixel is shifted towards the weighted average of similar colored pixels
 - the weights are determined by gaussian_weight, which calculates the weight using the Gaussian kernel
 - this is repeated until maximum shift is less than a threshold or the maximum number of iterations is reached
 - if the record_frames flag is set to True, the function also stores the frames of every iteration for later visualization
- Output:
 - the final color of each pixel after clustering
 - o optionally the frames of the clustering process for visualization

Once the pixel colors have converged, the function assign_clusters groups the pixels into their clusters:

- Each pixel is assign to an existing cluster if its distance to the cluster center is below a threshold (distance is calculated with colour dist)
- If no cluster is found, a new cluster is created

Visualization I

The function create_visualization_frames is used in mean_shift_color_pixel to create the frames for the animation. Creates:

- a 2D image from current pixel values reshaped back to image format
- a 3D RGB scatter plot showing the distribution of pixel colors

This is later used to create gifs using matplotlib.animation

Visualization II

The function color_density_3D visualizes the overall color distribution in RGB space as a 3D surface plot, using Gaussian Kernel Density Estimation. To enable a 3D visualization, the blue channel is fixed to a constant value (e.g., B = 128), allowing a two-dimensional grid to be created over the red and green channels. The resulting surface shows the estimated density of colors for each [R, G] combination, indicating which regions in color space are most frequently represented in the image.

Main

For each example image:

- runs the 3D color density topography
- iterates through some bandwidth values
- runs mean shift clustering and creates animations
- clusters results and saves the resulting image

Performance

To speed up the process, the code uses Numba JIT compilation to accelerate the mean shift algorithm.

Results

Note:

• To view the animations, the gifs themselves (located in the result folder) or the markdown file need to be opened. I only show screenshots of the a few animation iterations for the Pepper image.

• The original images were downsized for faster processing.

Peppers

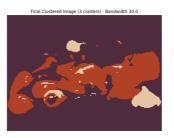
Original Image



Result with different bandwidths

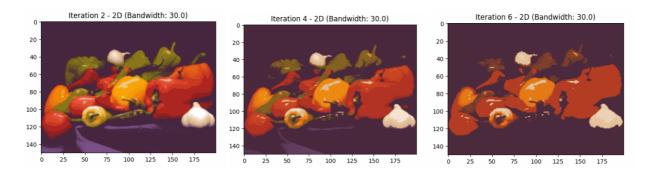




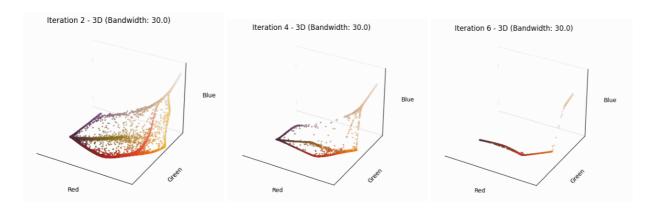


The bandwidth has a significant impact on the clustering result. It determines the size of the neighborhood used for averaging, where a small bandwidth leads to many small, detailed clusters, while a large bandwidth results in fewer, broader clusters by smoothing over fine color differences.

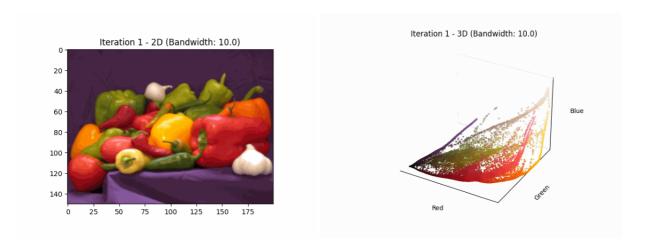
Visualization I: 2D Animation Iteration



Visualization I: 3D Animation Iteration

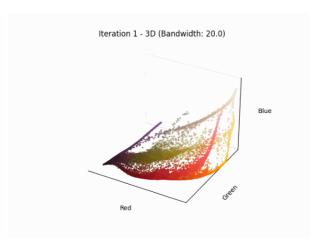


Visualization I for Bandwidth 10 (due to pdf convertion, the gifs are not animated)

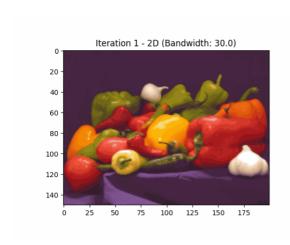


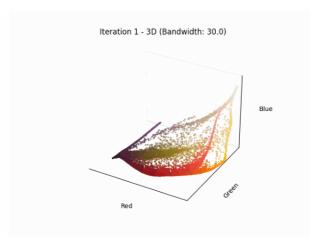
Visualization I for Bandwidth 20 (due to pdf convertion, the gifs are not animated)





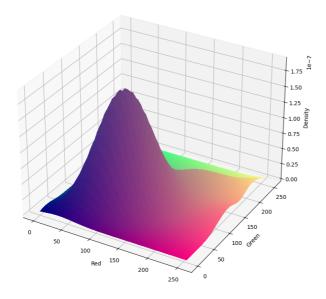
Visualization I for Bandwidth 30 (due to pdf convertion, the gifs are not animated)





Visualization II 3D Color Density Topography

3D Color Density Topography - RGB slice at Blue=128



Gas Station

Original Image



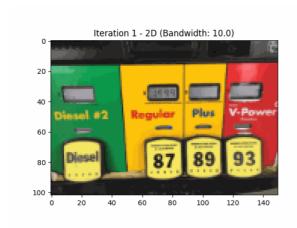
Result with different bandwidths

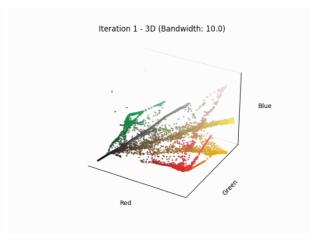




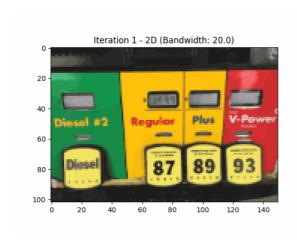


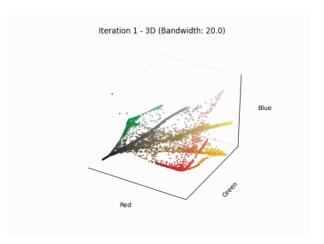
Visualization I for Bandwidth 10 (due to pdf convertion, the gifs are not animated)



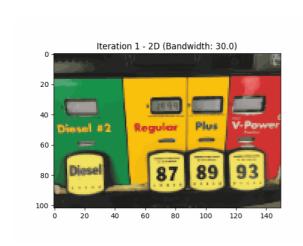


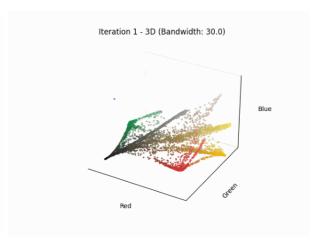
Visualization I for Bandwidth 20 (due to pdf convertion, the gifs are not animated)





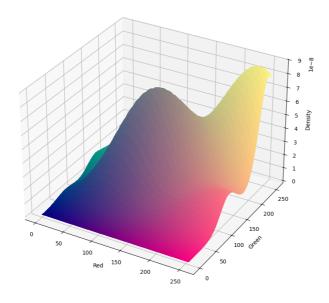
Visualization I for Bandwidth 30 (due to pdf convertion, the gifs are not animated)





Visualization II 3D Color Density Topography

3D Color Density Topography - RGB slice at Blue=128



Blue Bird

Original Image



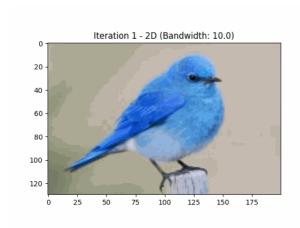
Result with different bandwidths

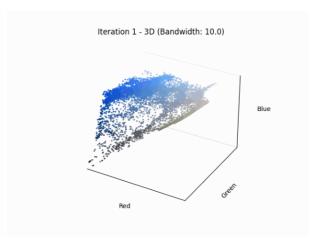




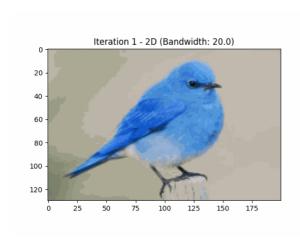


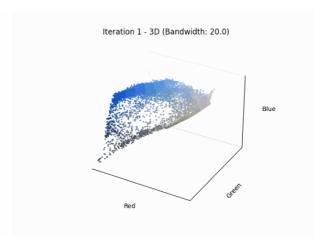
Visualization I for Bandwidth 10 (due to pdf convertion, the gifs are not animated)



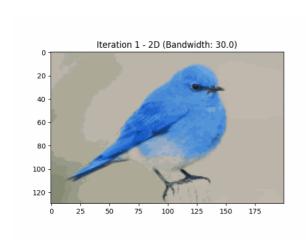


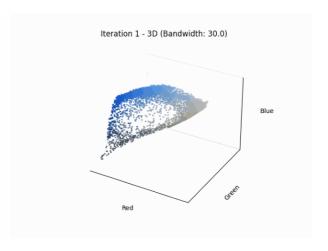
Visualization I for Bandwidth 20 (due to pdf convertion, the gifs are not animated)





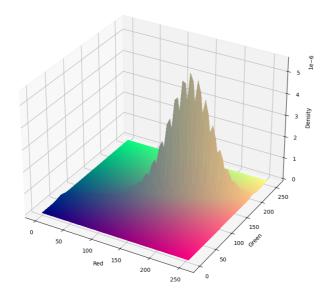
Visualization I for Bandwidth 30 (due to pdf convertion, the gifs are not animated)





Visualization II 3D Color Density Topography

3D Color Density Topography - RGB slice at Blue=128



Used Libraries

- NumPy numerical operations
- OpenCV image loading and saving
- Matplotlib visualization (2D/3D, animation)
- Numba performance optimization via JIT
- SciPy KDE density estimation
- os, io file handling and in-memory image buffers