
NEWS BROADCAST ANALYSIS

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1 Shot detection

A score is calculated for each frame using either the sum of absolute differences method or the histogram differences method. Given the mean μ and standard deviation σ of all the scores, a threshold is chosen as

$$\text{threshold} = \mu + k \times \sigma$$

where k is an integer. A shot is a series of frames taken by continuously by one camera. A shot change is declared when the value of the score becomes greater than the threshold and then becomes smaller than the threshold (similar to a peak above the threshold line).

Sum of absolute differences

A score is assigned to each frame by calculating the sum of the absolute differences between consecutive frames for every pixel. This value is then normalized by the size of the frame. When the score of a frame is greater than the threshold, a shot change is declared. This method works quite well with simple videos but it is not robust against movements and changes in lighting.

Histogram differences

Each frame is converted into a gray-scale image. A histogram with 256 bins, representing all the possible values of a pixel, is created for each frame. Then, a score is assigned by calculating the sum of the absolute differences between histograms of consecutive frames. This method is more robust against small changes in a frame.

Performance

Let C be the number of correctly identified cuts, M be the number of cuts that are not identified, and F be the number of falsely identified cuts. To evaluate how well the algorithm is detecting the shots, the recall (V), precision (P) and F_1 scores for each clip are calculated. They are defined as below.

$$\begin{aligned} V &= \frac{C}{C + M} \\ P &= \frac{C}{C + F} \\ F_1 &= \frac{2 \times P \times V}{P + V} \end{aligned} \tag{1}$$

Figure 1: Sum of absolute differences scores of clip 1

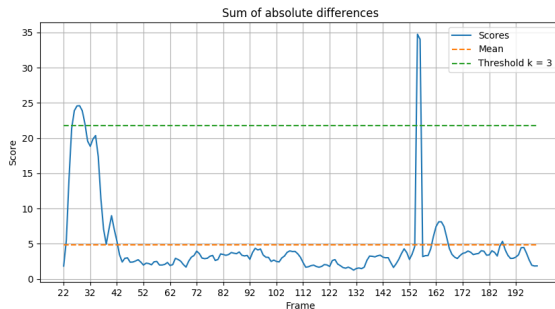


Figure 2: Histogram differences scores of clip 1

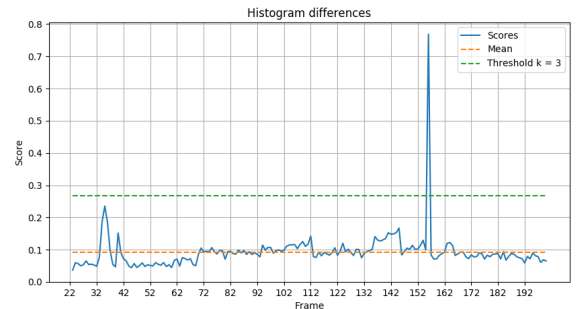


Figure 3: Sum of absolute differences scores of clip 2

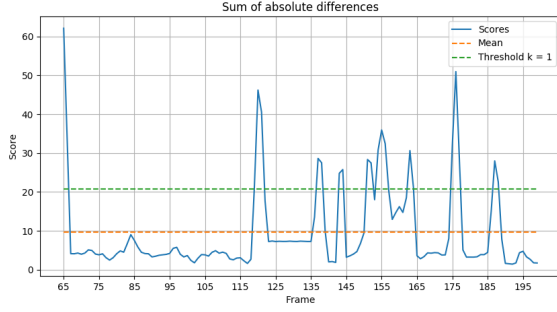


Figure 4: Histogram differences scores of clip 2

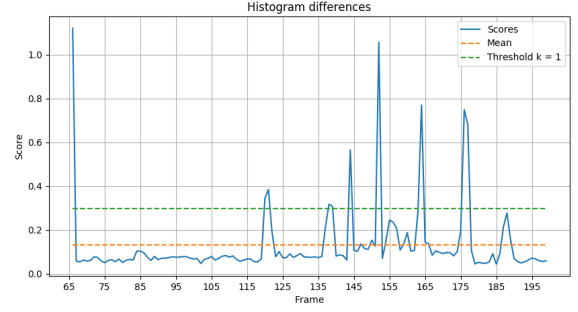


Figure 5: Sum of absolute differences scores of clip 3

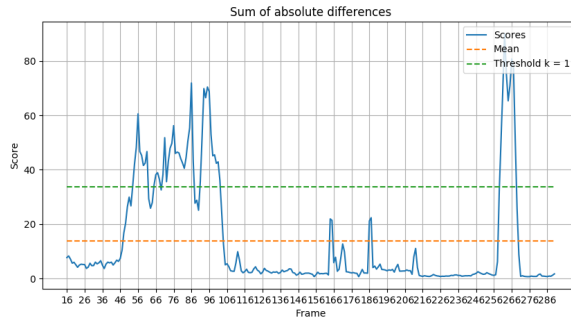


Figure 6: Histogram differences scores of clip 3

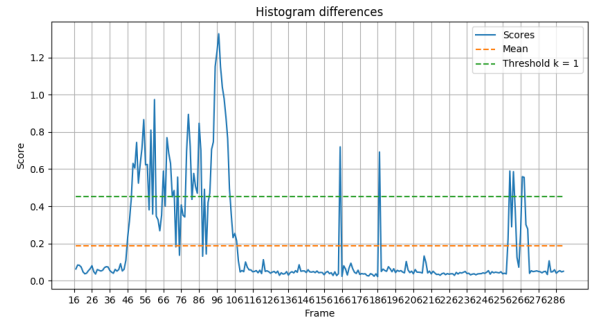


Table 1 and Table 2 shows the respective scores for both methods. On average, histogram differences achieve a higher score across all three metrics comparing to sum of absolute differences method. For example, the transition from frame 164 (Figure 7) to frame 165 (Figure 8) is not detected by the latter but it is by the former. It performs especially better in the montage section in clip 3, where there are frequent shot changes.

Table 1: Shot detection performance with sum of absolute differences

Clip	Correct (C)	Missed (M)	Falsely detected (F)	Recall (V)	Precision (P)	F_1
1	1	0	1	1	0.5	0.6667
2	7	0	2	1	0.7778	0.8750
3	5	16	0	0.2381	1	0.3846

Table 2: Shot detection performance with histogram differences

Clip	Correct (C)	Missed (M)	Falsely detected (F)	Recall (V)	Precision (P)	F_1
1	1	0	0	1	1	1
2	6	1	1	0.8571	0.8571	0.8571
3	15	6	1	0.7142	0.9375	0.8118

Figure 7: Frame 164 of clip 3



Figure 8: Frame 165 of clip 3



The relevant code is in `shot.py`. To get the graphs of shot detection, run the command below.

```
python3 run.py shot_detection -t <type> -i <path to frames>
```

To add shot numbers into frames, run the command below. The shot number will appear on the bottom left corner of each frame.

```
python3 run.py add_shot_numbe -i <input directory> -o <output directory>  
-k <k for thresholding>
```

2 Logo detection

Template matching is an object detection algorithm which is translation invariant but not scale or rotation invariant. As we are detecting the news company's logo, we can assume that the target of detection will be in a known orientation. Due to the possibility that there may be multiple occurrences of the logo in a frame, we cannot simply match SIFT features between the logo template and a frame. Template matching is run on templates of different sizes because the size of the logo in a frame is unknown. Then, a score is calculated for all the matches by normalized cross-correlation. The normalized version is chosen because brighter patches will not have a higher score. Also, the score obtained will be in the range $[0, 1]$ and this makes choosing a threshold more intuitive. As the logo in the template and the one in the frame might be slightly different, possibly due to the differences in resolutions or styles, a looser threshold is first used to filter out the irrelevant matches.

For each match, if its score is greater than the loose threshold, it is kept. Then, the algorithm checks if that particular match is a slight translation of a match we have already decided to keep. If it is, the match is discarded. This prevents having multiple boxes around one logo. Afterwards, SIFT descriptors are calculated for the remaining matches and a score is calculated using feature matching and Lowe's ratio test with the logo template. If a match's score is above a different threshold, it is declared as a match of the template and a box is put around the match.

Initially, I only did one pass with either normalized cross-correlation or SIFT feature matching. However, this led to fairly poor results, with the algorithm often unable to detect multiple logos and including many irrelevant matches. Therefore, the two passes approach, first with a looser threshold using normalized cross-correlation and then with another threshold with SIFT feature matching, is used. As can be seen in Figure 10, the algorithm can detect multiple logos. In hindsight, it would have been more efficient to scale down the images to multiple scales instead of scaling

down the templates. This is because sliding the template across the image takes much longer than resizing every frame to different scales.

Figure 9: Logo detection on frame 104 of clip 1



Figure 10: Logo detection on frame 52 of clip 1



The relevant code is in `logo.py`. To run logo detection, run the command below.

```
python3 run.py logo_detection -i <input directory> -o <output directory>
-d <logo path> -t <min threshold for NCC>
```

3 Face detection and tracking

There are 260 images in the female and male classes respectively. Each image is accompanied by a `.mat` file specifying the coordinates of the left eye, right eye, nose and mouth. The following rules are used to crop the images in order to obtain the faces.

$$\begin{aligned} start_x &= left_eye_x - 0.5 \times (right_eye_x - left_eye_x) \\ end_x &= right_eye_x + 0.5 \times (right_eye_x - left_eye_x) \\ start_y &= eyes_y - (mouth_y - eyes_y) \\ end_y &= mouth_y + (mouth_y - eyes_y) \end{aligned} \quad (2)$$

The relevant code for face cropping is in `crop_images()` in `face.py`.

The initial attempt to detect faces is to use skin detection - trying to filter out skin in images. Figure 11 and Figure 12 show the color distributions of faces in RGB and HSV color spaces. The HSV color space has narrower distributions, especially with hue. Although this method works sometimes as seen in Figure 13, it is not successful in general. It fails to detect a large area of the face of the man on the right in Figure 14 and includes a lot of the background. This model is especially poor when other things in the frame are very similar to human skin tone.

Figure 11: RGB distribution of fe male training images

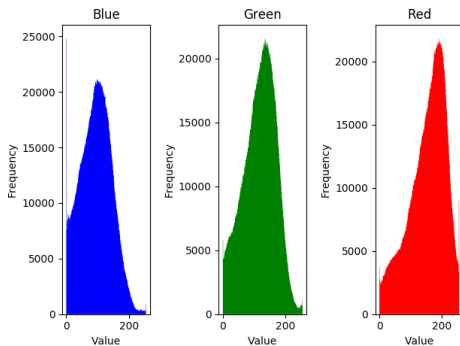


Figure 12: HSV distribution of female training images

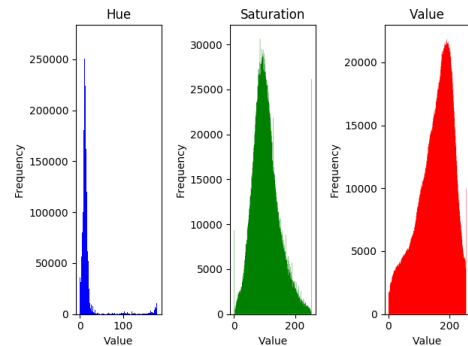


Figure 13: HSV colour detection on frame 160 of clip 1



Figure 14: HSV colour detection on frame 50 of clip 1



The relevant code for HSV face detection is in `visualize_distributions()` and `face_detection_hsv()` in `face.py`.

I ended up using `cv2.CascadeClassifier` for detecting faces. The full name of this classifier is Haar feature-based cascade classifier for object detection. It is a OpenCV pre-trained classifier for face stored in an XML file. It works fairly well. It has no trouble detecting multiple people in a frame or people of colour, like in Figure 15. However, it is usually unable to detect faces turned to the side. The detector was unable to detect the person on the left in any of the frames he appeared in that particular position in Figure 16.

After obtaining the faces in a frame, for each face in the current frame, the SIFT descriptors are found. Then, they are matched to the SIFT descriptors of all the faces in the previous frame using feature matching and Lowe's ratio test. Then, a score is obtained from the number of matches. If the score is above a set threshold, the algorithm declares that the face we are looking at is found in the previous frame and we will display the index assigned to that particular face in the previous frame. If the face is not found, we assign a new index to the face. The index of each face is labelled on top of the box.

Figure 15: Face detection on frame 69 of clip 2



Figure 16: Face detection on frame 110 of clip 1



4 Gender classification

90% of the images (234 images from each class) are used for training, whereas the other 10% (26 images from each class) are used for testing the accuracy of the model. The accuracy is defined as the percent of correctly identified faces, not SIFT descriptors, in the cases of support-vector machine and neural network.

SVM

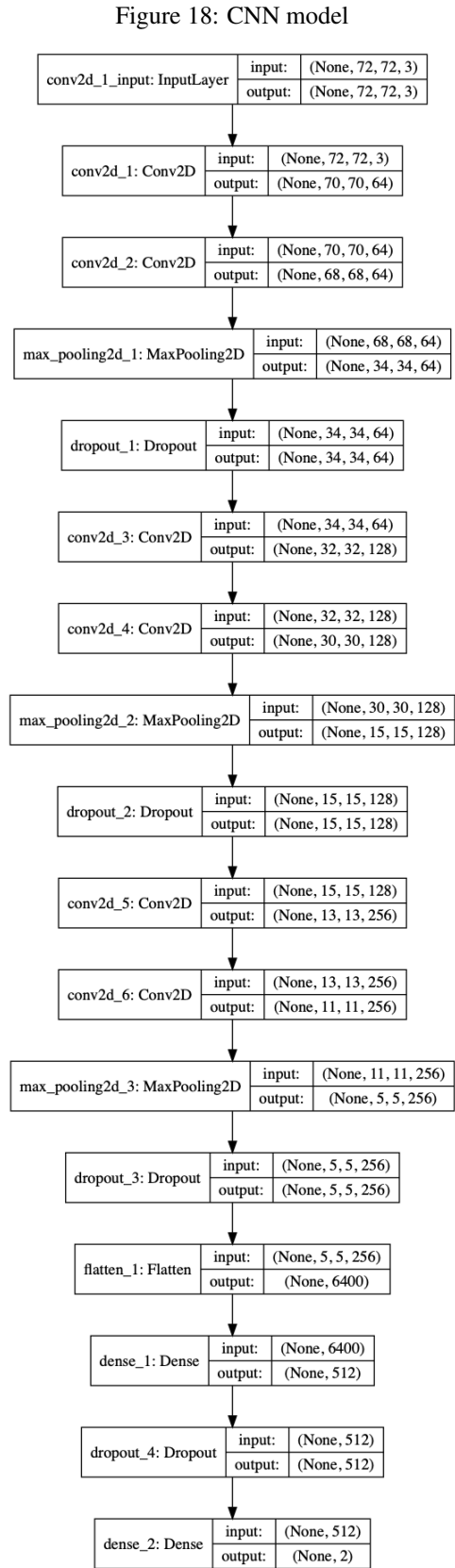
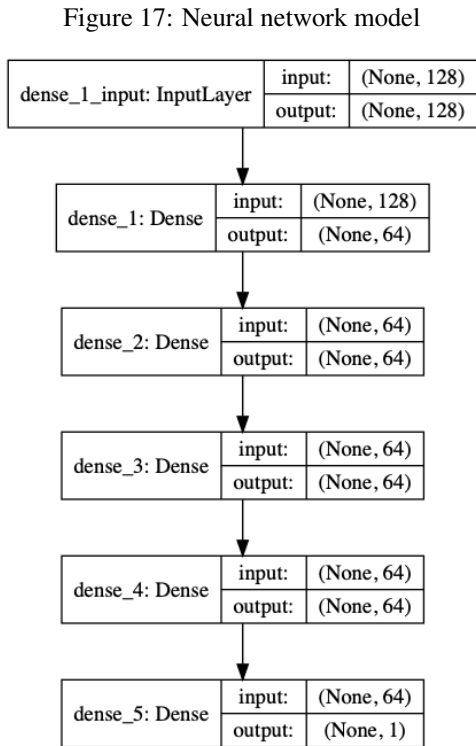
The SIFT descriptors of the training images are passed into the SVM model for training. Then, for each detected face in a frame, the SIFT descriptors are extracted and fed to the trained SVM model. A prediction for each descriptor is obtained. If more descriptors are predicted as female than male, the image is classified as female. If more descriptors are predicted as male than female, the image is classified as male. If there are equal number of descriptors being predicted as both female and male, the image is then classified as unknown.

Neural network

The same as SVM, except with a neural network model instead. The model uses a binary crossentropy loss, adam for optimization, and accuracy as the metric. The neural network used is shown in Figure 17.

CNN

All the training and testing images are padded with black borders to obtain a square shape and then resized to be 72 pixels by 72 pixels. The training images are then passed to the CNN model shown in Figure 18 for training. The model uses a binary crossentropy loss, stochastic gradient descent for optimization, and accuracy as the metric. Each detected face is padded to obtain a square shape. Then, the image is resized to be 72 pixels by 72 pixels. The resized image is then passed to the trained CNN model and a category prediction is obtained.



The relevant code for training any of the three models is in `train_model()` in `face.py`. To train a gender classification model, run the command below.

```
python3 run.py train -m <model path after training>
-c <classification model (SVM, NN_SIFT, or CNN)>
```

The relevant code for face detection (including gender classification and face tracking) is in `face_detection()` in `face.py`. Run the command below.

```
python3 run.py face_detection -i <input directory> -o <output directory>
-c <classification model (SVM, NN_SIFT, or CNN)> -m <trained model path>
-s <min size for face detection>
```

Performance

Table 3 shows the test accuracies of the three models. As expected, CNN performed poorly, achieving an accuracy that is equivalent to random guesses, due to the very small training data size of 468. It consistently classify all the faces as either female and male.

Table 3: Gender classification performance

Model	Description	Accuracy on test set
SVM	Using SIFT descriptors of faces	100.00%
Neural network	Using SIFT descriptors of faces	92.30%
CNN	Using cropped and resized faces	50.00%

5 Make video

The relevant code for combining all the frames to a video is in `make_video()` in `utils.py`. To do logo detection, face detection, face tracking and gender classification and make a video, run the command below.

```
python3 run.py run_all -i <input directory> -o <output directory>
-d <logo path> -t <min threshold for NCC> -m <trained model path>
-c <classification model (SVM, NN_SIFT, or CNN)> -v <name of output video>
-f <frame per second> -s <min size for face detection>
```

6 References

Video shot boundary detection based on color histogram

Wikipedia - Shot transition detection

Shot detection using pixel wise difference with adaptive threshold and color histogram method in compressed and uncompressed video

7 Code

Listing 1: shot.py

```
1 import cv2
2 import matplotlib.pyplot as plt
3 import numpy as np
4 import os
5
6 def add_shot_number(input_dir, output_dir, shots):
7     """
8     Add shot number using the shot changes specified in variable shots.
9     """
10    print("add_shot_number")
11    if not os.path.isdir(output_dir):
12        os.mkdir(output_dir)
13
14    exts = [".jpg", ".png"]
15    img_names = [img for img in os.listdir(input_dir) if img.endswith(tuple(exts))]
16    # Sort images by name in ascending order
17    img_names.sort(key=lambda img: int("".join(filter(str.isdigit, img))))
18
19    i = 0
20    shot = 0
21    for filename in img_names:
22        image_num = int("".join(filter(str.isdigit, filename)))
23        img = cv2.imread(os.path.join(input_dir, filename))
24        if i < len(shots) and image_num == shots[i]:
25            shot += 1
26            i += 1
27        # Add shot number to frame
28        text = str(shot)
29        # Get width and height of the text box
30        text_width, text_height = cv2.getTextSize(text, cv2.FONT_HERSHEY_PLAIN, fontScale
31        ↪ =1.5, thickness=2)[0]
32        # Set the text start position
33        text_x = 10
34        text_y = img.shape[0] - 10
35        box_coords = ((text_x, text_y + 2), (text_x + text_width - 2, text_y -
36        ↪ text_height - 4))
37        cv2.rectangle(img, box_coords[0], box_coords[1], (255, 255, 255), cv2.FILLED)
38        cv2.putText(img, text, (text_x, text_y), cv2.FONT_HERSHEY_PLAIN, fontScale=1.5,
39        ↪ color=(0, 0, 0), thickness=2)
40        # Save frame
41        cv2.imwrite(os.path.join(output_dir, filename), img)
42        # cv2.imshow("A box!", img)
43        # cv2.waitKey(0)
44
45 def shot_detection(input_dir, method, k):
46     """
47     Get graphs of scores of shot changes using the method specify in method.
48     Method should be either SAD2 or HD.
49     """
50    scores = []
51    exts = [".jpg", ".png"]
52    img_names = [img for img in os.listdir(input_dir) if img.endswith(tuple(exts))]
53    # Sort images by name in ascending order
54    img_names.sort(key=lambda img: int("".join(filter(str.isdigit, img))))
55    start_idx = int(img_names[0].find(".jpg"))
56    end_idx = int(img_names[-1].find(".jpg"))
```

```

54
55 # Sum of absolute differences
56 if method == "SAD2":
57     prev_img = None
58     for i in range(len(img_names)):
59         # Default type is numpy.uint64
60         curr_img = cv2.imread(os.path.join(input_dir, img_names[i])).astype(np.int64)
61         if i == 0:
62             r, c, d = curr_img.shape
63             next_img = cv2.imread(os.path.join(input_dir, img_names[i + 1])).astype(np.
64                 ↪ int64)
65             score = np.sum(np.abs(curr_img - next_img))
66         elif i == len(img_names) - 1:
67             score = np.sum(np.abs(curr_img - prev_img))
68         else:
69             next_img = cv2.imread(os.path.join(input_dir, img_names[i + 1])).astype(np.
70                 ↪ int64)
71             score = 0.5 * np.sum(np.abs(curr_img - prev_img)) + 0.5 * np.sum(np.abs(
72                 ↪ curr_img - next_img))
73             scores.append(score)
74             prev_img = curr_img
75     x = np.arange(start_idx, end_idx + 1)
76     scores = np.array(scores) / (r * c * d)
77     title = "Sum of absolute differences"
78     new_filename = "output/" + input_dir.name + "_score_sad2.png"
79
80 elif method == "SAD":
81     prev_img = None
82     for i in range(len(img_names)):
83         # Default type is numpy.uint64
84         curr_img = cv2.imread(os.path.join(input_dir, img_names[i])).astype(np.int64)
85         if i == 0:
86             r, c, d = curr_img.shape
87             prev_img = curr_img
88             continue
89             score = np.sum(np.abs(curr_img - prev_img))
90             scores.append(score)
91             prev_img = curr_img
92     x = np.arange(start_idx + 1, end_idx + 1)
93     scores = np.array(scores) / (r * c * d)
94     title = "Sum of absolute differences"
95     new_filename = "output/" + input_dir.name + "_score_sad2.png"
96
97 # Histogram differences
98 elif method == "HD":
99     prev_histogram = None
100     for i in range(len(img_names)):
101         # Default type is numpy.uint8
102         curr_img = cv2.imread(os.path.join(input_dir, img_names[i]))
103         # Default type is numpy.uint8
104         curr_img_g = cv2.cvtColor(curr_img, cv2.COLOR_BGR2GRAY).astype(np.int16)
105         histogram = np.histogram(np.ravel(curr_img_g), bins=np.arange(-1, 256))
106         if i == 0:
107             r, c = curr_img_g.shape
108             prev_histogram = histogram[0]
109             continue
110             score = np.sum(np.abs(histogram[0] - prev_histogram))
111             scores.append(score)
112             prev_histogram = histogram[0]

```

```

110
111     x = np.arange(start_idx + 1, end_idx + 1)
112     scores = np.array(scores) / (r * c)
113     title = "Histogram differences"
114     new_filename = "output/" + input_dir.name + "_score_hd.png"
115
116     else:
117         raise ValueError("Illegal method value")
118
119     y_mean = [np.mean(scores)] * len(x)
120     threshold = [np.mean(scores) + k * np.std(scores)] * len(x)
121     f = plt.figure(figsize=(10, 5))
122     ax = f.gca()
123     ax.set_xticks(np.arange(start_idx, end_idx, 10))
124     plt.title(title)
125     plt.xlabel("Frame")
126     plt.ylabel("Score")
127     plt.plot(x, scores, label="Scores")
128     # Plot the average line
129     plt.plot(x, y_mean, label="Mean", linestyle="--")
130     plt.plot(x, threshold, label="Threshold k = " + str(k), linestyle="--")
131     plt.legend(loc="upper right")
132     plt.grid()
133     f.savefig(new_filename)
134     print("Output saved to " + new_filename)

```

Listing 2: logo.py

```

1 import cv2
2 import imutils
3 import numpy as np
4 import os
5
6 def get_score(img1, img2):
7     """
8     Get similarity score between img1 and img2 using SIFT features matching and
9     Lowe's ratio testing.
10    """
11    sift = cv2.xfeatures2d.SIFT_create()
12    kp1, des1 = sift.detectAndCompute(img1, None)
13    kp2, des2 = sift.detectAndCompute(img2, None)
14    if len(kp1) < 2 or len(kp2) < 2:
15        return 0
16    index_params = dict(algorithm=0, trees=5)
17    flann = cv2.FlannBasedMatcher(index_params, None)
18    matches = flann.knnMatch(des1, des2, k=2)
19    good_matches = []
20    for m, n in matches:
21        if m.distance < 0.6*n.distance:
22            good_matches.append(m)
23    num_kps = 0
24    if len(kp1) <= len(kp2):
25        num_kps = len(kp1)
26    else:
27        num_kps = len(kp2)
28    score = len(good_matches) / num_kps * 100
29    return score
30
31 def logo_detection(input_dir, output_dir, logo_path, min_threshold):
32     print("detect_logo")

```

```

33
34 if not os.path.isdir(output_dir):
35     os.mkdir(output_dir)
36
37 template = cv2.imread(str(logo_path))
38 template_g = cv2.cvtColor(template, cv2.COLOR_BGR2GRAY)
39
40 w, h = template_g.shape[::-1]
41
42 # Compute template of different sizes
43 if template.shape[0] < 50:
44     # scales = np.linspace(0.8, 1.0, 6)[::-1]
45     scales = np.linspace(0.9, 1.0, 2)[::-1]
46 else:
47     # scales = np.linspace(0.1, 1.0, 25)[::-1]
48     scales = np.linspace(0.5, 1.0, 10)[::-1]
49 templates = []
50 ratios = []
51 for scale in scales:
52     resized = imutils.resize(template_g, width=int(template_g.shape[1] * scale))
53     templates.append(resized)
54     ratios.append(resized.shape[1] / float(template_g.shape[1]))
55
56 for img_name in os.listdir(input_dir):
57     if not img_name.endswith(".jpg"):
58         continue
59     img = cv2.imread(os.path.join(input_dir, img_name))
60     img_g = cv2.cvtColor(img, cv2.COLOR_BGR2GRAY)
61     img_canny = cv2.Canny(img_g, 100, 200)
62
63     p = -1
64     q = -1
65     matches = []
66     # Loop through the templates
67     for i in range(len(templates) - 1, -1, -1):
68         # Stop when template is bigger than image
69         if img_g.shape[0] < templates[i].shape[0] or img_g.shape[1] < templates[i].
            ↪ shape[1]:
70             break
71
72         # First pass – normalized cross correlation
73         match = cv2.matchTemplate(img_g, templates[i], cv2.TM_CCORR_NORMED) # img_g.shape
74             ↪ - template_g.shape + 1
75         if p == -1 and q == -1:
76             p, q = match.shape
77             m, n = match.shape
78             matches.append(np.pad(match, ((0, p - m), (0, q - n)), mode="constant",
            ↪ constant_values=0))
79
80 boxes = []
81
82 matches = np.array(matches)
83 r, max_y, max_x = np.unravel_index(np.argmax(matches), matches.shape)
84 r = ratios[len(ratios) - 1 - r]
85 max_val = np.max(matches)
86 max_thresh = max(max_val * 0.95, min_threshold)
87 # If the match with the highest score is smaller than the min threshold,
88 # there is no match in this image and just saves the input image.
89 if max_val < min_threshold:

```

```

89         cv2.imwrite(os.path.join(output_dir, img_name), img)
90         continue
91     start_x, start_y = max_x, max_y
92     end_x, end_y = int((max_x + w * r)), int((max_y + h * r))
93     match_score = get_score(template, img[start_y:end_y, start_x:end_x])
94     if match_score < 5:
95         cv2.imwrite(os.path.join(output_dir, img_name), img)
96         continue
97     boxes.append((r, max_y, max_x, 1))
98
99     # Matches obtained from first pass
100    match_locations = np.where(matches >= max_thresh)
101    for i in range(len(match_locations[0])):
102        r1, y1, x1 = ratios[len(ratios) - 1 - match_locations[0][i]], match_locations
103        ↪ [1][i], match_locations[2][i]
104        found = False
105        for j in range(len(boxes)):
106            r2, y2, x2, count = boxes[j]
107            # Check if two boxes of the same size overlap or
108            # if a smaller one is contained in the bigger one
109            if (r1 == r2 and np.abs(x1 - x2) < w * r1 and np.abs(y1 - y2) < h * r1) or
110            ↪ \
111            ((r1 < r2) and (x1 <= (x2 + w * r2) <= (x1 + w * r1)) and (y1 <= (y2 + h
112            ↪ * r2) <= (y1 + h * r1))) or \
113            (np.abs(x1 - x2) < 0.5 * w * r1 and np.abs((x1 + w * r1) - (x2 + w * r2
114            ↪ )) < 0.5 * w * r1 and \
115            np.abs(y1 - y2) < 0.5 * h * r1 and np.abs((y1 + h * r1) - (y2 + h * r2
116            ↪ )) < 0.5 * h * r1):
117                boxes[j] = (r2, y2, x2, count + 1)
118                found = True
119                break
120    if not found:
121        start_x, start_y = x1, y1
122        end_x, end_y = int((x1 + w * r1)), int((y1 + h * r1))
123        # Second pass - SIFT features matching
124        match_score = get_score(template, img[start_y:end_y, start_x:end_x])
125        if match_score > 5:
126            boxes.append((r1, y1, x1, 1))
127
128    for r, y, x, count in boxes:
129        start_x, start_y = x, y
130        end_x, end_y = int((x + w * r)), int((y + h * r))
131        cv2.rectangle(img, (start_x, start_y), (end_x, end_y), (0, 255, 0), thickness
132        ↪ =2)
133
134    cv2.imwrite(os.path.join(output_dir, img_name), img)
135    # cv2.imshow("img", img)
136    # cv2.waitKey(0)

```

Listing 3: face.py

```

1  import copy
2  import cv2
3  import matplotlib.pyplot as plt
4  import numpy as np
5  import os
6  import pickle
7  import scipy.io, scipy.misc
8  from skimage import io
9  from skimage import transform as tf

```

```

10 from skimage.color import rgb2hsv
11 from sklearn import svm
12 from utils import shuffle
13
14 def visualize_distributions(imgs_dir):
15     """
16     Visualize the distributions of values in HSV and BGR of pictures in imgs_dir.
17     """
18     H, S, V = None, None, None
19     B, G, R = None, None, None
20     for filename in os.listdir(imgs_dir):
21         if not filename.endswith(".png"):
22             continue
23         img = cv2.imread(imgs_dir + filename)
24         hsv_img = cv2.cvtColor(img, cv2.COLOR_BGR2HSV)
25         if H is None:
26             H = hsv_img[..., 0].flatten()
27             S = hsv_img[..., 1].flatten()
28             V = hsv_img[..., 2].flatten()
29             B = img[..., 0].flatten()
30             G = img[..., 1].flatten()
31             R = img[..., 2].flatten()
32         else:
33             H = np.concatenate([H, hsv_img[..., 0].flatten()])
34             S = np.concatenate([S, hsv_img[..., 1].flatten()])
35             V = np.concatenate([V, hsv_img[..., 2].flatten()])
36             B = np.concatenate([B, img[..., 0].flatten()])
37             G = np.concatenate([G, img[..., 1].flatten()])
38             R = np.concatenate([R, img[..., 2].flatten()])
39
40     # Plot
41     f = plt.figure()
42     ax1 = f.add_subplot(1, 3, 1)
43     ax1.hist(H, bins=180,
44             range=(0.0, 180.0), histtype="stepfilled", color="b", label="Hue")
45     plt.title("Hue")
46     plt.xlabel("Value")
47     plt.ylabel("Frequency")
48     ax2 = f.add_subplot(1, 3, 2)
49     ax2.hist(S, bins=256,
50             range=(0.0, 255.0), histtype="stepfilled", color="g", label="Saturation")
51     plt.title("Saturation")
52     plt.xlabel("Value")
53     plt.ylabel("Frequency")
54     ax3 = f.add_subplot(1, 3, 3)
55     ax3.hist(V, bins=256,
56             range=(0.0, 255.0), histtype="stepfilled", color="r", label="Value")
57     plt.title("Value")
58     plt.xlabel("Value")
59     plt.ylabel("Frequency")
60     f.tight_layout()
61     f.savefig(imgs_dir + "_hsv_distributions.png")
62     plt.show()
63
64     f = plt.figure()
65     ax1 = f.add_subplot(1, 3, 1)
66     ax1.hist(B, bins=256,
67             range=(0.0, 255.0), histtype="stepfilled", color="b", label="Blue")
68     plt.title("Blue")

```

```

69     plt.xlabel("Value")
70     plt.ylabel("Frequency")
71     ax2 = f.add_subplot(1, 3, 2)
72     ax2.hist(G, bins=256,
73             range=(0.0, 255.0), histtype="stepfilled", color="g", label="Saturation")
74     plt.title("Green")
75     plt.xlabel("Value")
76     plt.ylabel("Frequency")
77     ax3 = f.add_subplot(1, 3, 3)
78     ax3.hist(R, bins=256,
79             range=(0.0, 255.0), histtype="stepfilled", color="r", label="Red")
80     plt.title("Red")
81     plt.xlabel("Value")
82     plt.ylabel("Frequency")
83     f.tight_layout()
84     f.savefig(imgs_dir + "_rgb_distributions.png")
85     plt.show()
86
87 def crop_images(old_dir, new_dir):
88     """
89     Crop .jpg images in old_dir given coordinates of left eye, right eye,
90     nose and mouth in .mat files. Save cropped images in new_dir.
91     """
92     print("crop_images")
93     os.mkdir(new_dir)
94
95     H, S, V = None, None, None
96     for filename in os.listdir(old_dir):
97         if not filename.endswith(".jpg"):
98             continue
99
100         index = filename.find(".jpg")
101         name = filename[:index]
102
103         # Approximate coordinates of face
104         coords = scipy.io.loadmat(old_dir + name + ".mat")
105         start_x = int(coords["x"][0][0] - 0.5*(coords["x"][1][0] - coords["x"][0][0]))
106         end_x = int(coords["x"][1][0] + 0.5*(coords["x"][1][0] - coords["x"][0][0]))
107         start_y = int(coords["y"][0][0] - (coords["y"][3][0] - coords["y"][0][0]))
108         end_y = int(coords["y"][3][0] + (coords["y"][3][0] - coords["y"][2][0]))
109         img = io.imread(old_dir + filename)
110         face = img[start_y:end_y, start_x:end_x]
111         # Save cropped image
112         scipy.misc.imsave(new_dir + name + ".png", face)
113
114 def resize_images(input_dir, output_dir, size=72):
115     """
116     Resize images in input_dir to size x size.
117     """
118     print("resize_images")
119     if not os.path.isdir(output_dir):
120         os.mkdir(output_dir)
121
122     for filename in os.listdir(input_dir):
123         if not filename.endswith(".png"):
124             continue
125         img = cv2.imread(os.path.join(input_dir, filename))
126         x, y, d = img.shape
127         # Pad image to square

```

```

128     if x > y:
129         padded_img = np.pad(img, ((0, 0), (0, x - y), (0, 0)),
130                               mode="constant", constant_values=0)
131     else:
132         padded_img = np.pad(img, ((0, y - x), (0, 0), (0, 0)),
133                               mode="constant", constant_values=0)
134     resized_img = cv2.resize(padded_img, (size, size))
135     scipy.misc.imsave(output_dir + filename, cv2.cvtColor(resized_img, cv2.
        ↳ COLOR_BGR2RGB))

136
137 def get_features(input_dir):
138     """
139     Get keypoints and descriptors of the images in input_dir with SIFT.
140     """
141     kps, des = None, None
142     sift = cv2.xfeatures2d.SIFT_create()
143
144     for filename in os.listdir(input_dir):
145         if not filename.endswith(".png"):
146             continue
147         img = io.imread(input_dir + filename)
148         kp, d = sift.detectAndCompute(img, None)
149
150         if des is None:
151             kps = kp
152             des = d
153         else:
154             kps = np.concatenate([kps, kp], axis=0)
155             des = np.concatenate([des, d], axis=0)
156     return kps, des
157
158 def get_data(f_dir, m_dir):
159     """
160     Get the images of the F and M classes from their respective directories and shuffle.
161     """
162     import keras
163     x = []
164     y = [0] * len(os.listdir(f_dir)) + [1] * len(os.listdir(m_dir))
165     imgs = os.listdir(f_dir) + os.listdir(m_dir)
166     for filename in os.listdir(f_dir):
167         if not filename.endswith(".png"):
168             continue
169         x.append(cv2.imread(os.path.join(f_dir, filename)))
170     for filename in os.listdir(m_dir):
171         if not filename.endswith(".png"):
172             continue
173         x.append(cv2.imread(os.path.join(m_dir, filename)))
174
175     x, y = shuffle(x, y)
176     y = keras.utils.to_categorical(y, num_classes=2)
177     return x, y
178
179 def train_model(model_path, classification):
180     """
181     Train either a SVM or neural network model using SIFT features or
182     a CNN using face images.
183     """
184     OLD_F_DIR = "original_data/female/"
185     OLD_M_DIR = "original_data/male/"

```



```

186 F_TRAIN_DIR = "data/female_train/"
187 M_TRAIN_DIR = "data/male_train/"
188 F_TEST_DIR = "data/female_test/"
189 M_TEST_DIR = "data/male_test/"
190 F_TRAIN_CNN_DIR = "data/female_cnn_train/"
191 M_TRAIN_CNN_DIR = "data/male_cnn_train/"
192 F_TEST_CNN_DIR = "data/female_cnn_test/"
193 M_TEST_CNN_DIR = "data/male_cnn_test/"
194
195 if not os.path.isdir("data/"):
196     os.mkdir("data/")
197     crop_images(OLD_F_DIR, F_TRAIN_DIR)
198     crop_images(OLD_M_DIR, M_TRAIN_DIR)
199
200 if classification == "SVM":
201     f_train_kps, f_train_des = get_features(F_TRAIN_DIR)
202     m_train_kps, m_train_des = get_features(M_TRAIN_DIR)
203     x_train = np.concatenate([f_train_des, m_train_des], axis=0)
204     y_train = [-1] * len(f_train_des) + [1] * len(m_train_des)
205     model = svm.SVC(kernel="rbf", gamma="scale", C=10.0)
206
207     model.fit(x_train, y_train)
208     # Save model
209     pickle.dump(model, open(model_path, "wb"))
210     print("Model saved as " + str(model_path))
211
212 elif classification == "NN_SIFT":
213     import keras
214     from keras.layers import Conv2D, Dense, Dropout, Flatten, MaxPooling2D
215     from keras.models import Sequential
216     from keras.optimizers import Adam, SGD
217     from keras.utils import plot_model
218
219     f_train_kps, f_train_des = get_features(F_TRAIN_DIR)
220     m_train_kps, m_train_des = get_features(M_TRAIN_DIR)
221     x_train = np.concatenate([f_train_des, m_train_des], axis=0)
222     y_train = [0] * len(f_train_des) + [1] * len(m_train_des)
223     x_train, y_train = shuffle(x_train, y_train)
224
225     f_test_kps, f_test_des = get_features(F_TEST_DIR)
226     m_test_kps, m_test_des = get_features(M_TEST_DIR)
227     x_test = np.concatenate([f_test_des, m_test_des], axis=0)
228     y_test = [0] * len(f_test_des) + [1] * len(m_test_des)
229     x_test, y_test = shuffle(x_test, y_test)
230
231     model = Sequential()
232     model.add(Dense(64, activation='relu', kernel_initializer='random_normal', input_dim
        ↳ =128))
233     model.add(Dense(64, activation='relu', kernel_initializer='random_normal'))
234     model.add(Dense(64, activation='relu', kernel_initializer='random_normal'))
235     model.add(Dense(64, activation='relu', kernel_initializer='random_normal'))
236     model.add(Dense(1, activation='sigmoid', kernel_initializer='random_normal'))
237     model.compile(optimizer='adam', loss='binary_crossentropy', metrics=['accuracy'])
238
239     model.fit(x_train, y_train, batch_size=10, epochs=20)
240     score = model.evaluate(x_test, y_test)
241     print("score:", score)
242
243     model.save(model_path)

```

```

244     print("Model saved as " + str(model_path))
245     plot_model(model, to_file="output/nn_model.png", show_shapes=True)
246
247 elif classification == "CNN":
248     import keras
249     from keras.layers import Conv2D, Dense, Dropout, Flatten, MaxPooling2D
250     from keras.models import Sequential
251     from keras.optimizers import Adam, SGD
252     from keras.utils import plot_model
253
254     np.random.seed(123)
255     size = 72
256
257     # Pad and resize images to size by size
258     # resize_images(F_TRAIN_DIR, F_TRAIN_CNN_DIR, size)
259     # resize_images(M_TRAIN_DIR, M_TRAIN_CNN_DIR, size)
260     # resize_images(F_TEST_DIR, F_TEST_CNN_DIR, size)
261     # resize_images(M_TEST_DIR, M_TEST_CNN_DIR, size)
262
263     x_train, y_train = get_data(F_TRAIN_CNN_DIR, M_TRAIN_CNN_DIR)
264     x_test, y_test = get_data(F_TEST_CNN_DIR, M_TEST_CNN_DIR)
265
266     model = Sequential()
267     model.add(Conv2D(32, (3, 3), activation="relu", input_shape=(size, size, 3)))
268     model.add(Conv2D(32, (3, 3), activation="relu"))
269     model.add(MaxPooling2D(pool_size=(2, 2)))
270     model.add(Dropout(0.25))
271
272     model.add(Conv2D(64, (3, 3), activation="relu"))
273     model.add(Conv2D(64, (3, 3), activation="relu"))
274     model.add(MaxPooling2D(pool_size=(2, 2)))
275     model.add(Dropout(0.25))
276
277     model.add(Conv2D(128, (3, 3), activation="relu"))
278     model.add(Conv2D(128, (3, 3), activation="relu"))
279     model.add(MaxPooling2D(pool_size=(2, 2)))
280     model.add(Dropout(0.25))
281
282     model.add(Flatten())
283     model.add(Dense(256, activation="relu"))
284     model.add(Dropout(0.5))
285     model.add(Dense(2, activation="softmax"))
286
287     adam = Adam(lr=0.0005)
288     model.compile(loss="binary_crossentropy", optimizer=adam, metrics=["accuracy"])
289
290     model.fit(x_train, y_train, epochs=10)
291     score = model.evaluate(x_test, y_test)
292     print("score:", score)
293
294     model.save(model_path)
295     print("Model saved as " + str(model_path))
296     plot_model(model, to_file="output/cnn_model.png", show_shapes=True)
297
298 else:
299     raise ValueError("Illegal classification value")
300
301 def predict_model(model_path):
302     """

```

```

303     Get accuracy of test set for either the SVM or neural network model.
304     """
305     print("predict_model")
306     F_TEST_DIR = "data/female_test/"
307     M_TEST_DIR = "data/male_test/"
308
309     ext = os.path.splitext(model_path)[1]
310     if ext == ".sav":
311         model = pickle.load(open(model_path, "rb"))
312     elif ext == ".h5":
313         from keras.models import load_model
314         model = load_model(model_path)
315     else:
316         raise ValueError("Not valid model extension")
317
318     sift = cv2.xfeatures2d.SIFT_create()
319
320     correct = 0
321     for filename in os.listdir(F_TEST_DIR):
322         if not filename.endswith(".png"):
323             continue
324         img = cv2.imread(os.path.join(F_TEST_DIR, filename))
325         kps, des = sift.detectAndCompute(img, None)
326         result = model.predict(des)
327         if ext == ".sav":
328             f_count = np.count_nonzero(result == -1)
329             m_count = np.count_nonzero(result == 1)
330         elif ext == ".h5":
331             f_count = np.sum(result < 0.5)
332             m_count = np.sum(result > 0.5)
333         if f_count > m_count:
334             correct += 1
335
336     for filename in os.listdir(M_TEST_DIR):
337         if not filename.endswith(".png"):
338             continue
339         img = cv2.imread(os.path.join(M_TEST_DIR, filename))
340         kps, des = sift.detectAndCompute(img, None)
341         result = model.predict(des)
342         if ext == ".sav":
343             f_count = np.count_nonzero(result == -1)
344             m_count = np.count_nonzero(result == 1)
345         elif ext == ".h5":
346             f_count = np.sum(result < 0.5)
347             m_count = np.sum(result > 0.5)
348         if m_count > f_count:
349             correct += 1
350     score = correct / (len(os.listdir(F_TEST_DIR)) + len(os.listdir(M_TEST_DIR))) * 100
351     print(str(score) + "% of images are categorized correctly")
352
353 def face_detection_hsv(input_dir, output_dir):
354     """
355     Face detection with HSV color detection (does not work well).
356     """
357     print("face_detection_hsv")
358
359     if not os.path.isdir(output_dir):
360         os.mkdir(output_dir)
361     for filename in os.listdir(input_dir):

```

```

362     if not filename.endswith(".jpg"):
363         continue
364     img = cv2.imread(os.path.join(input_dir, filename))
365
366     # HSV color detection
367     hsv_img = cv2.cvtColor(img, cv2.COLOR_BGR2HSV)
368
369     lower1 = np.array([0, 48, 80])
370     upper1 = np.array([20, 255, 255])
371     mask1 = cv2.inRange(hsv_img, lower1, upper1) # img_hsv.shape
372     lower2 = np.array([170, 0, 0])
373     upper2 = np.array([180, 255, 255])
374     mask2 = cv2.inRange(hsv_img, lower2, upper2) # img_hsv.shape
375     mask = cv2.bitwise_or(mask1, mask2)
376
377     kernel = cv2.getStructuringElement(cv2.MORPH_ELLIPSE, (10, 10))
378     new_img = cv2.bitwise_and(img, img, mask=mask)
379     cv2.imwrite(os.path.join(output_dir, filename), new_img)
380
381 def face_detection_cascade(input_dir, output_dir, model_path, classification, min_size):
382     """
383     Face detection with classifier, face tracking, and gender classification
384     using the method specified in classification variable using model_path.
385     """
386     print("face_detection_cascade")
387     XML_FILENAME = "models/haarcascade_frontalface_default.xml"
388     F_TEXT = "Female"
389     M_TEXT = "Male"
390     O_TEXT = "Not sure"
391     F_COLOR = (0, 0, 255)
392     M_COLOR = (255, 0, 0)
393     O_COLOR = (0, 255, 0)
394
395     if classification == "SVM":
396         model = pickle.load(open(model_path, "rb"))
397     elif classification == "NN_SIFT":
398         from keras.models import load_model
399         model = load_model(model_path)
400     elif classification == "CNN":
401         from keras.models import load_model
402         model = load_model(model_path)
403         size = 72
404     else:
405         raise ValueError("Illegal classification value")
406
407     if not os.path.isdir(output_dir):
408         os.mkdir(output_dir)
409
410     exts = [".jpg", ".png"]
411     img_names = [img for img in os.listdir(input_dir) if img.endswith(tuple(exts))]
412     # Sort images by name in ascending order
413     img_names.sort(key=lambda img: int(''.join(filter(str.isdigit, img))))
414     sift = cv2.xfeatures2d.SIFT_create()
415     index_params = dict(algorithm=0, trees=5)
416     flann = cv2.FlannBasedMatcher(index_params, None)
417     face_count = 0
418     prev_faces = {}
419
420     for filename in img_names:

```

```

421 img = cv2.imread(os.path.join(input_dir, filename))
422 img_g = cv2.cvtColor(img, cv2.COLOR_BGR2GRAY)
423
424 #####
425 ### Cascade face detection ###
426 #####
427 face_cascade = cv2.CascadeClassifier(XML_FILENAME)
428 faces = face_cascade.detectMultiScale(img_g, scaleFactor=1.3,
429 minNeighbors=5, minSize=(min_size, min_size))
430 curr_faces = {}
431
432 for x, y, w, h in faces:
433     #####
434     ### Face tracking ###
435     #####
436     face = img[y:y+h, x:x+w]
437     kp1, des1 = sift.detectAndCompute(face, None)
438     found_face = False
439     for (x2, y2, w2, h2), index in prev_faces.items():
440         face2 = img[y2:y2+h, x2:x2+w]
441         kp2, des2 = sift.detectAndCompute(face2, None)
442         if len(kp1) < 2 or len(kp2) < 2:
443             continue
444         matches = flann.knnMatch(des1, des2, k=2)
445         good_matches = []
446         for m, n in matches:
447             if m.distance < 0.6*n.distance:
448                 good_matches.append(m)
449         num_kps = 0
450         if len(kp1) <= len(kp2):
451             num_kps = len(kp1)
452         else:
453             num_kps = len(kp2)
454         score = len(good_matches) / num_kps * 100
455         if score > 45:
456             found_face = True
457             face_number = index
458             curr_faces[(x, y, w, h)] = index
459     if not found_face:
460         face_number = face_count
461         curr_faces[(x, y, w, h)] = face_count
462         face_count += 1
463
464 #####
465 ### Gender classification ###
466 #####
467 if classification == "SVM":
468     # Get SIFT descriptors
469     kps, des = sift.detectAndCompute(face, None)
470     # Predict female or male
471     result = model.predict(des)
472     f_count = np.count_nonzero(result == -1)
473     m_count = np.count_nonzero(result == 1)
474     # Plot color box
475     if f_count > m_count:
476         text = str(face_number) + " " + F_TEXT + ": " + \
477             format(f_count/(f_count + m_count)*100, ".2f") + "%"
478         color = F_COLOR
479     elif m_count > f_count:

```

```

480         text = str(face_number) + " " + M_TEXT + ": " + \
481             format(m_count/(f_count + m_count)*100, ".2f") + "%"
482         color = M_COLOR
483     else:
484         text = str(face_number) + " " + O_TEXT
485         color = O_COLOR
486
487     elif classification == "NN_SIFT":
488         assert w == h
489         kps, des = sift.detectAndCompute(face, None)
490         prediction = model.predict(des)
491         f_count = np.sum(prediction < 0.5)
492         m_count = np.sum(prediction > 0.5)
493         if f_count > m_count:
494             text = str(face_number) + " " + F_TEXT + ": " + \
495                 format(f_count/(f_count + m_count)*100, ".2f") + "%"
496             color = F_COLOR
497         elif m_count > f_count:
498             text = str(face_number) + " " + M_TEXT + ": " + \
499                 format(m_count/(f_count + m_count)*100, ".2f") + "%"
500             color = M_COLOR
501         else:
502             text = str(face_number) + " " + O_TEXT
503             color = O_COLOR
504
505     elif classification == "CNN":
506         assert w == h
507         face = cv2.resize(face, (size, size))
508         prediction = model.predict(np.array([face]), verbose=0)
509         prediction = prediction[0]
510         if prediction[0] > 0.5:
511             text = str(face_number) + " " + F_TEXT + ": " + \
512                 format(prediction[0]*100, ".2f") + "%"
513             color = F_COLOR
514         elif prediction[0] < 0.5:
515             text = str(face_number) + " " + M_TEXT + ": " + \
516                 format(prediction[0]*100, ".2f") + "%"
517             color = M_COLOR
518         else:
519             text = str(face_number) + " " + O_TEXT
520             color = O_COLOR
521     cv2.rectangle(img, (x, y), (x+w, y+h), color, thickness=2)
522     cv2.putText(img, text, (x, y-10), color=color,
523                fontFace=cv2.FONT_HERSHEY_PLAIN, fontScale=1)
524
525     prev_faces = curr_faces
526     cv2.imwrite(os.path.join(output_dir, filename), img)
527     # cv2.imshow("img", img)
528     # cv2.waitKey(0)

```

Listing 4: utils.py

```

1  import cv2
2  import os
3  import numpy as np
4
5  def make_video(imgs_dir, vid_name, fps):
6      """
7      Make vid_name.mp4 using images in imgs_dir.
8      """

```

```

9     OUTPUT_DIR = "output/"
10    if not os.path.isdir(OUTPUT_DIR):
11        os.mkdir(OUTPUT_DIR)
12    exts = [".jpg", ".png"]
13    imgs = [img for img in os.listdir(imgs_dir) if img.endswith(tuple(exts))]
14    # Sort images by name in ascending order
15    imgs.sort(key=lambda img: int("".join(filter(str.isdigit, img))))
16    frame = cv2.imread(os.path.join(imgs_dir, imgs[0]))
17    h, w, _ = frame.shape
18
19    # Make sure vid_name does not include a file extension
20    index = vid_name.find(".")
21    if index != -1:
22        vid_name = vid_name[:index]
23    file_path = OUTPUT_DIR + vid_name + ".mp4"
24    vid = cv2.VideoWriter(file_path, cv2.VideoWriter_fourcc(*"MP4V"), fps, (w, h))
25    for img in imgs:
26        vid.write(cv2.imread(os.path.join(imgs_dir, img)))
27    vid.release()
28    print("Video is now in ", file_path)
29
30    def shuffle(x, y):
31        """
32        Shuffle data x and their labels y.
33        """
34        assert len(x) == len(y)
35        idx = np.random.permutation(len(x))
36        x, y = np.array(x)[idx], np.array(y)[idx]
37        return x, y

```

Listing 5: run.py

```

1  import click
2  from pathlib import Path
3
4  @click.group()
5  def main():
6      pass
7
8  @main.command()
9  # @click.option("--model_path", "-m", type=Path, default="models/cnn_model.h5")
10 @click.option("--model_path", "-m", type=Path, default="models/svm_model.sav")
11 @click.option("--classification", "-c", default="SVM", help="SVM or NN_SIFT or CNN")
12 def train(**kwargs):
13     from face import train_model
14     train_model(**kwargs)
15
16 @main.command()
17 # @click.option("--model_path", "-m", type=Path, default="models/nn_model.h5")
18 @click.option("--model_path", "-m", type=Path, default="models/svm_model.sav")
19 def predict(**kwargs):
20     from face import predict_model
21     predict_model(**kwargs)
22
23 @main.command()
24 @click.option("--model_path", "-m", type=Path, default="models/svm_model.sav")
25 # @click.option("--model_path", "-m", type=Path, default="models/nn_model.h5")
26 @click.option("--classification", "-c", default="SVM", help="SVM or NN_SIFT or CNN")
27 # Clip 1
28 @click.option("--input_dir", "-i", type=Path, default="original_data/clip_1/")

```



```

29 @click.option("--output_dir", "-o", type=Path, default="output/clip_1_face/")
30 @click.option("--min_size", "-s", default=70, type=int, help="Min size for face
    ↪ detection")
31 # Clip 2
32 # @click.option("--input_dir", "-i", type=Path, default="original_data/clip_2/")
33 # @click.option("--output_dir", "-o", type=Path, default="output/clip_2/")
34 # @click.option("--min_size", "-s", default=70, type=int, help="Min size for face detection")
35 def face_detection(**kwargs):
36     from face import face_detection_hsv, face_detection_cascade
37     # face_detection_hsv(**kwargs)
38     face_detection_cascade(**kwargs)
39
40 @main.command()
41 # Clip 1
42 # @click.option("--input_dir", "-i", type=Path, default="original_data/clip_1/")
43 # @click.option("--output_dir", "-o", type=Path, default="output/clip_1_logo/")
44 # @click.option("--logo_path", "-d", type=Path, default="data/clip_1_logo2.png")
45 # @click.option("--min_threshold", "-t", type=float, default=0.87)
46 # Clip 2
47 # @click.option("--input_dir", "-i", type=Path, default="original_data/clip_2/")
48 # @click.option("--output_dir", "-o", type=Path, default="output/clip_2/")
49 # @click.option("--logo_path", "-d", type=Path, default="data/clip_2_logo.png")
50 # @click.option("--min_threshold", "-t", type=float, default=0.82)
51 # Clip 3
52 @click.option("--input_dir", "-i", type=Path, default="original_data/clip_3/")
53 @click.option("--output_dir", "-o", type=Path, default="output/clip_3/")
54 @click.option("--logo_path", "-d", type=Path, default="data/clip_3_logo.png")
55 @click.option("--min_threshold", "-t", type=float, default=0.82)
56 def logo_detection(**kwargs):
57     from logo import logo_detection
58     logo_detection(**kwargs)
59
60 @main.command()
61 @click.option("--input_dir", "-i", type=Path, default="original_data/clip_3/")
62 @click.option("--method", "-m", default="SAD2", help="SAD2 or HD")
63 @click.option("--k", "-k", default=1, help="k used in threshold")
64 def shot_detection(**kwargs):
65     from shot import shot_detection
66     shot_detection(**kwargs)
67
68 @main.command()
69 @click.option("--imgs_dir", "-i", default="original_data/clip_1/", type=Path)
70 @click.option("--vid_name", "-v", default="clip_1")
71 @click.option("--fps", "-fps", default=6, type=int, help="Frame per second")
72 def make_video(**kwargs):
73     from utils import make_video
74     make_video(**kwargs)
75
76 @main.command()
77 # Clip 1
78 @click.option("--input_dir", "-i", type=Path, default="original_data/clip_1/")
79 @click.option("--output_dir", "-o", type=Path, default="output/clip_1_shots/")
80 @click.option("--shots", "-s", type=list, default=[22, 156])
81 # Clip 2
82 # @click.option("--input_dir", "-i", type=Path, default="original_data/clip_2/")
83 # @click.option("--output_dir", "-o", type=Path, default="output/clip_2_shots/")
84 # @click.option("--shots", "-s", type=list, default=[65, 120, 138, 144, 152, 164, 177, 188])
85 # Clip 3
86 # @click.option("--input_dir", "-i", type=Path, default="original_data/clip_3/")

```



```

87 # @click.option("--output_dir", "-o", type=Path, default="output/clip_3_shots/")
88 # @click.option("--shots", "-s", type=list, default=[16, 51, 57, 59, 61, 63, 65, 67, 69, 71, 72, 74, 78, 80, 84, 91,
    ↪ 95, 102, 165, 187, 260, 267])
89 def add_shot_number(**kwargs):
90     from shot import add_shot_number
91     add_shot_number(**kwargs)
92
93 @main.command()
94 # Clip 1
95 # @click.option("--input_dir", "-i", type=Path, default="output/clip_1_shots/")
96 # @click.option("--output_dir", "-o", type=Path, default="output/clip_1/")
97 # @click.option("--logo_path", "-d", type=Path, default="data/clip_1_logo.png")
98 # @click.option("--min_threshold", "-t", type=float, default=0.87)
99 # @click.option("--model_path", "-m", type=Path, default="models/svm_model.sav")
100 # @click.option("--classification", "-c", default="SVM", help="SVM or NN_SIFT or CNN")
101 # @click.option("--vid_name", "-v", default="clip_1")
102 # @click.option("--fps", "-fps", default=6, type=int, help="Frame per second")
103 # @click.option("--min_size", "-s", default=70, type=int, help="Min size for face detection")
104 # Clip 2
105 @click.option("--input_dir", "-i", type=Path, default="output/clip_2_shots/")
106 @click.option("--output_dir", "-o", type=Path, default="output/clip_2/")
107 @click.option("--logo_path", "-d", type=Path, default="data/clip_2_logo.png")
108 @click.option("--min_threshold", "-t", type=float, default=0.82)
109 @click.option("--model_path", "-m", type=Path, default="models/svm_model.sav")
110 @click.option("--classification", "-c", default="SVM", help="SVM or NN_SIFT or CNN")
111 @click.option("--vid_name", "-v", default="clip_2")
112 @click.option("--fps", "-fps", default=6, type=int, help="Frame per second")
113 @click.option("--min_size", "-s", default=70, type=int, help="Min size for face
    ↪ detection")
114 # Clip 3
115 # @click.option("--input_dir", "-i", type=Path, default="output/clip_3_shots/")
116 # @click.option("--output_dir", "-o", type=Path, default="output/clip_3/")
117 # @click.option("--logo_path", "-d", type=Path, default="data/clip_3_logo.png")
118 # @click.option("--min_threshold", "-t", type=float, default=0.82)
119 # @click.option("--model_path", "-m", type=Path, default="models/svm_model.sav")
120 # @click.option("--classification", "-c", default="SVM", help="SVM or NN_SIFT or CNN")
121 # @click.option("--vid_name", "-v", default="clip_3")
122 # @click.option("--fps", "-fps", default=8, type=int, help="Frame per second")
123 # @click.option("--min_size", "-s", default=40, type=int, help="Min size for face detection")
124 def run_all(input_dir, output_dir, logo_path, min_threshold, model_path, \
125             classification, vid_name, fps, min_size):
126     from face import face_detection_cascade
127     from logo import logo_detection
128     from utils import make_video
129     logo_detection(input_dir=input_dir, output_dir=output_dir, \
130                   logo_path=logo_path, min_threshold=min_threshold)
131     face_detection_cascade(input_dir=output_dir, output_dir=output_dir, \
132                           # face_detection_cascade(input_dir=input_dir, output_dir=output_dir, \
133                           model_path=model_path, classification=classification, min_size=min_size)
134     make_video(imgs_dir=output_dir, vid_name=vid_name, fps=fps)
135
136 if __name__ == "__main__":
137     main()

```