

## Run Mask Detect on live video

### Imports, paths, and info

```
In [1]: 1 import cv2
        2 import mediapipe as mp
        3 import os
        4 import numpy as np
        5 import PIL
        6 from PIL import Image, ImageOps, ImageDraw
        7 import torch
        8 from torch import nn
        9 from typing import Sequence
       10

In [2]: 1 mp_face_detection = mp.solutions.face_detection
        2 mp_drawing = mp.solutions.drawing_utils
        3
        4 target_h = 112
        5 target_w = target_h # enforce square
        6
        7 # detection constants
        8 BAD_COLOR = (0, 0, 255) # red
        9 GOOD_COLOR = (0, 128, 0) # green
       10 BOX_LINE_THICKNESS = 4

In [3]: 1 video_fp = r'C:\Users\Andrew\Documents\2022 Summer\Data Mining\Project\Webcam\videos\surgical mask.mov'
        2 #video_fp = r'C:\Users\Andrew\Documents\2022 Summer\Data Mining\Project\Webcam\videos\cloth mask.mov'
        3
        4 processed_video_fp = r'C:\Users\Andrew\Documents\2022 Summer\Data Mining\Project\Webcam\videos\processed.avi'
        5
        6 full_model_path = r'C:\Users\Andrew\Documents\2022 Summer\Data Mining\Project\results\upgrade\full_model_best'
        7 debug_output_dir = r'D:\data\face_mask\webcam\debug'
```

### Model and Functions

```

In [4]: 1 class CNN(nn.Module):
2     def __init__(
3         self,
4         input_size: Sequence[int] = (3, 112, 112),
5         num_classes: int = 2,
6         channels: Sequence[int] = (8, 16, 32),
7         kernel_sizes: Sequence[int] = (10, 10, 10, 10),
8         linear_units: Sequence[int] = (100, 10),
9         lr: float = 0.001,
10        epochs: int = 10
11    ):
12        super(CNN, self).__init__()
13
14        self.input_size = input_size
15        self.num_classes = num_classes
16        self.channels = input_size[0:1] + channels
17        self.kernel_sizes = kernel_sizes
18        self.linear_units = linear_units
19        self.lr = lr
20        self.epochs = epochs
21
22        self.flatten = nn.Flatten()
23        self.pool = partial(nn.MaxPool2d, kernel_size=2, stride=2) # first 2 is for 2x2 kernel, second is stride length
24        self.dropout = nn.Dropout
25        self.activation = nn.ReLU
26        self.accuracy = torchmetrics.functional.accuracy
27        self.conf_matrix = torchmetrics.functional.confusion_matrix
28
29        # optional, define batch norm here
30
31        # build the convolutional layers
32        conv_layers = list()
33        for in_channels, out_channels, kernel_size in zip(
34            self.channels[:-2], self.channels[1:-1], self.kernel_sizes[:-1]
35        ):
36            conv_layers.append(
37                nn.Conv2d(
38                    in_channels=in_channels,
39                    out_channels=out_channels,
40                    kernel_size=kernel_size,
41                    #stride=2,
42                    #padding='same',
43                )
44            )
45            conv_layers.append(self.activation())
46            conv_layers.append(self.pool())
47        # add final layer to convolutions
48        conv_layers.append(
49            nn.Conv2d(
50                in_channels=self.channels[-2],
51                out_channels=self.channels[-1],
52                kernel_size=self.kernel_sizes[-1],
53                stride=2,
54                #padding='same',
55            )
56        )
57        conv_layers.append(self.activation())
58        conv_layers.append(self.pool())
59
60
61        # turn list into layers
62        self.conv_net = nn.Sequential(*conv_layers)
63
64        # Linear Layers
65        linear_layers = list()
66        prev_linear_size = self.channels[-1] * 9 # const scale it correctly
67        for dense_layer_size in self.linear_units:
68            linear_layers.append(
69                nn.Linear(
70                    in_features=prev_linear_size,
71                    out_features=dense_layer_size,
72                )
73            )
74            linear_layers.append(self.activation())
75            prev_linear_size=dense_layer_size
76
77        self.penultimate_dense = nn.Sequential(*linear_layers)
78        self.ultimate_dense = nn.Linear(
79            in_features=self.linear_units[-1],
80            out_features=self.num_classes
81        )
82
83
84        def forward(self, x: torch.Tensor) -> torch.Tensor:
85            x = self.conv_net(x)
86            x = self.flatten(x)
87            # may need to expand dense entry since flatten
88            x = self.penultimate_dense(x)
89            x = self.ultimate_dense(x)
90            return x
91
92
93        def train(dataloader, model, loss_fn, optimizer, verbose=False):
94            #model = model.float() # sometime fixes random obscure type error
95            model.train() # configures for training, grad on, dropout if there is dropout
96            size = len(dataloader.dataset)
97
98            for batch, (X, y) in enumerate(dataloader):
99                optimizer.zero_grad()
100
101                # compute prediction loss
102                preds = model(X)
103                loss = loss_fn(preds, y)
104
105                # backprop
106                loss.backward()
107                optimizer.step()
108
109                if batch % 5 == 0 and verbose:
110                    loss, current = loss.item(), batch * len(X)
111                    print(f"loss: {loss:>7f} [{current:>5d}/{size:>5d}]")
112            return loss
113
114        # for evaluating on validation data too

```

```

115 def test(dataloader, model, loss_fn, verbose=False):
116     model.eval()
117     test_loss, correct = 0, 0
118     size = len(dataloader.dataset)
119     num_batches = len(dataloader)
120
121     with torch.no_grad():
122         for X, y in dataloader:
123
124             pred = model(X.float())
125             test_loss += loss_fn(pred, y).item()
126             correct += (pred.argmax(1) == y).type(torch.float).sum().item()
127
128     test_loss /= num_batches
129     correct /= size
130     if verbose:
131         print(f"Results: \n Accuracy: {(100*correct):>0.1f}%, Avg loss: {test_loss:>8f} \n")
132     return correct, test_loss
133

```

```

In [5]: 1 def correct_crop(xl, xr, yt, yb, w, h):
2         if yt < 0:
3             diff = abs(yt)
4             yt = 0
5             expand_left = int(diff / 2)
6             expand_right = diff - expand_left
7             xl = xl - expand_left
8             xr = xr + expand_right
9         if xl < 0:
10            diff = abs(xl)
11            xl = 0
12            expand_down = int(diff / 2)
13            expand_up = diff - expand_down
14            yb = yb + expand_down
15            yt = yt - expand_up
16        if yb > h:
17            diff = yb - h
18            yb = h
19            expand_left = int(diff / 2)
20            expand_right = diff - expand_left
21            xl = xl - expand_left
22            xr = xr + expand_right
23        if xr > w:
24            diff = xr - w
25            xr = w
26            expand_down = int(diff / 2)
27            expand_up = diff - expand_down
28            yb = yb + expand_down
29            yt = yt - expand_up
30        if yt < 0 or xl < 0 or yb > h or xr > w:
31            print('coords error after correction')
32        return xl, xr, yt, yb

```

```

In [6]: 1 def rect_square_expansion(xl, xr, yt, yb, w, h):
2         bbh = yb - yt
3         bbw = xr - xl
4         if bbh > bbw:
5             diff = bbh - bbw
6             expand_left = int(diff/2)
7             expand_right = diff - expand_left
8             xl = xl - expand_left
9             xr = xr + expand_right
10        elif bbw > bbh:
11            diff = bbw - bbh
12            expand_down = int(diff/2)
13            expand_up = diff - expand_down
14            yb = yb + expand_down
15            yt = yt - expand_up
16
17        return xl, xr, yt, yb

```

In [7]:

```
1  # for debug
2  counter = 0
3
4  # classification smoothing
5  classification = 0.5 # init to no confidence in either direction
6  smoothing_adaptability = .1 # 1.0 means display most recent detection,
7    # 0.0 means do not update classification at all
8
9
10
11
12 # Load classification model
13 model = torch.load(full_model_path)
14 model.eval()
15
16 #cap = cv2.VideoCapture(0)
17 cap = cv2.VideoCapture(video_fp)
18
19 frame_width = int(cap.get(3))
20 frame_height = int(cap.get(4))
21 size = (frame_width, frame_height)
22
23 output_vid_writer = cv2.VideoWriter(processed_video_fp,
24    cv2.VideoWriter_fourcc('MJPG'),
25    10, size)
26
27 output_vid_writer = cv2.VideoWriter(processed_video_fp,
28    cv2.VideoWriter_fourcc('MJPG'),
29    10, size)
30
31 with mp_face_detection.FaceDetection(
32     model_selection=0, min_detection_confidence=0.5) as face_detection:
33     while cap.isOpened():
34         success, image = cap.read()
35         if not success:
36             print("Ignoring empty camera frame.")
37             # If loading a video, use 'break' instead of 'continue'.
38             break
39
40         # To improve performance, optionally mark the image as not writeable to
41         # pass by reference.
42         image.flags.writeable = False
43         image = cv2.cvtColor(image, cv2.COLOR_BGR2RGB)
44         results = face_detection.process(image)
45
46         # Draw the face detection annotations on the image.
47         image.flags.writeable = True
48         image = cv2.cvtColor(image, cv2.COLOR_RGB2BGR)
49         if results.detections:
50             for detection in results.detections:
51
52                 # get height, width and depth of image frame
53                 h, w, d = image.shape
54
55                 # draw face
56                 #mp_drawing.draw_detection(image, detection)
57
58                 # get bounding box and transform normalized coords to pixel coords
59                 rbb = detection.location_data.relative_bounding_box
60                 rect_start_point = mp_drawing._normalized_to_pixel_coordinates(
61                     rbb.xmin, rbb.ymin, w, h)
62                 rect_end_point = mp_drawing._normalized_to_pixel_coordinates(
63                     rbb.xmin + rbb.width, rbb.ymin + rbb.height, w, h)
64
65                 if rect_start_point is not None and rect_end_point is not None:
66                     # get individual coordinates from the tuples and create the square
67                     x1, yt = rect_start_point
68                     xr, yb = rect_end_point
69                     x1, xr, yt, yb = rect_square_expansion(x1, xr, yt, yb, w, h)
70
71                     # expand if nessisary
72                     expansion = .125
73                     bbh = yb - yt
74                     bbw = xr - x1
75                     amt_to_add = int(expansion * max(bbh, bbw))
76                     yt = yt - amt_to_add
77                     yb = yb + amt_to_add
78                     x1 = x1 - amt_to_add
79                     xr = xr + amt_to_add
80
81                     x1, xr, yt, yb = correct_crop(x1, xr, yt, yb, w, h)
82                     x1, xr, yt, yb = correct_crop(x1, xr, yt, yb, w, h)
83
84                     # crop frame to face
85                     pil_img = Image.fromarray(cv2.cvtColor(image, cv2.COLOR_BGR2RGB))
86                     # PIL crop format: Left, top, right, bottom
87                     crop = [x1, yt, xr, yb]
88                     pil_crop = pil_img.crop(crop)
89
90                     # resize
91                     pil_crop = pil_crop.resize((target_h, target_w), resample=PIL.Image.Resampling.HAMMING)
92
93                     # debug
94                     #out_path = os.path.join(debug_output_dir, '{}.png'.format(counter))
95                     #pil_crop.save(out_path)
96
97                     # turn image into array
98                     im_arr = np.array(pil_crop)
99                     im_arr = im_arr.reshape((3, 112, 112))
100                    im_arr = im_arr.reshape((1, 3, 112, 112))
101
102                    # norm
103                    im_arr = im_arr / 255
104
105                    model_input = torch.Tensor(im_arr)
106                    raw_pred = model(model_input.float()) # need to add .float()
107                    mask_class = raw_pred.argmax(1).item()
108                    #print(mask_class)
109
110                    rect_start_point_classification = (x1, yt)
111                    rect_end_point_classification = (xr, yb)
112
113                    # smooth out detection
114                    classification = smoothing_adaptability * mask_class + (1-smoothing_adaptability) * classification
```

```

115
116         if classification < 0.5: # no mask/incorrect mask
117             cv2.rectangle(image, rect_start_point_classification, rect_end_point_classification,
118                           BAD_COLOR, BOX_LINE_THICKNESS)
119         else:
120             cv2.rectangle(image, rect_start_point_classification, rect_end_point_classification,
121                           GOOD_COLOR, BOX_LINE_THICKNESS)
122
123
124
125         counter += 1
126
127
128         output_vid_writer.write(image)
129
130         # Flip the image horizontally for a selfie-view display.
131         cv2.imshow('MediaPipe Face Detection', cv2.flip(image, 1))
132         if cv2.waitKey(5) & 0xFF == 27: # if wait 5 miliseconds and 0xFF == 00011011 (always false)??
133             break
134     cap.release()
135     output_vid_writer.release()

```

Ignoring empty camera frame.

```

In [8]: 1 #cap.release()
        2 output_vid_writer.release()
        3
        4 # Closes all the frames
        5 cv2.destroyAllWindows()

```

```

In [ ]: 1

```