tesseract ocr

May 3, 2023

0.1 EDA - Tesseract OCR

Purpose: To investigate the utility in using Tesseract to identify odometer distances

General Plan: - Throw an image at tesseract and see if it can pick up a 6 digit number. - Maybe try adding some standard pre-processing - See how many photos it can do this for

Code is taken from/heavily inspired by: https://nanonets.com/blog/ocr-with-tesseract/

```
[]: # Imports
import cv2
import pytesseract
from pytesseract import Output
import matplotlib.pyplot as plt
import numpy as np
import copy
```

0.1.1 Try a image



```
[]: # Adding custom options custom_config = r"--psm 12" # psm = page segmentation mode; 12 = Sparse text

→with OSD.
```

```
[]: text = pytesseract.image_to_string(img, config=custom_config)
text
```

[]: 'Cua\n\n~~\n\n6 BO\n\n"gezil OV1+.\n\n2\n\nuly\n\n4d80-Z\n\n10\n\nLy OL f 7 A\n\n5°)\n\nSHERRIE,\n\n7\n\naa?\n'

```
(x, y, w, h) = (d["left"][i], d["top"][i], d["width"][i],

img = cv2.rectangle(img, (x, y), (x + w, y + h), (0, 255, 0), 2)

plt.imshow(img, cmap="gray")
plt.show()
```



What did we learn?

Tesseract seemingly doesn't work very well when trying to extract odometer values.

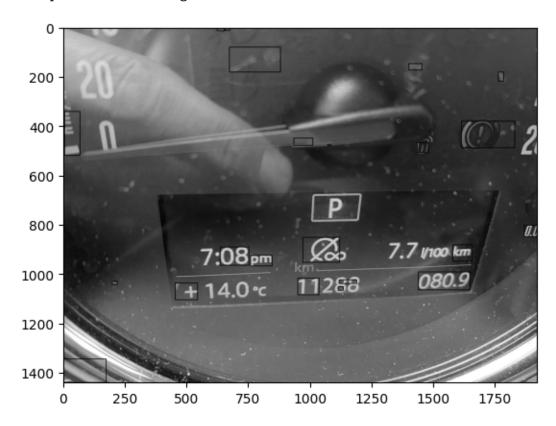
0.1.2 Try some image pre-processing

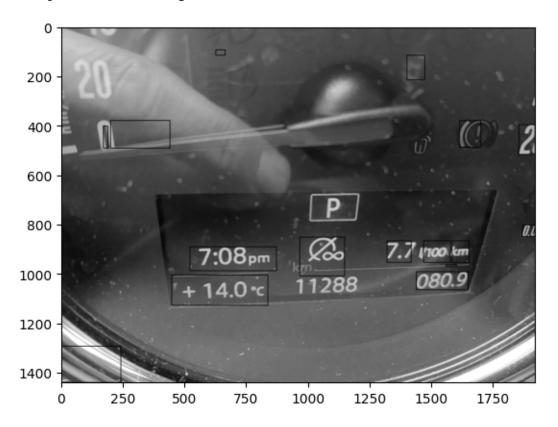
```
[]: def get_grayscale(image):
    return cv2.cvtColor(image, cv2.COLOR_BGR2GRAY)

# noise removal
def remove_noise(image):
    return cv2.medianBlur(image, 5)
```

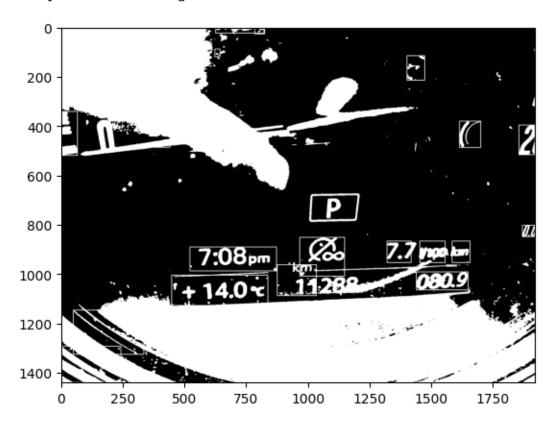
```
# thresholding
def thresholding(image):
    return cv2.threshold(image, 0, 255, cv2.THRESH_BINARY + cv2.THRESH_OTSU)[1]
# dilation
def dilate(image):
   kernel = np.ones((5, 5), np.uint8)
    return cv2.dilate(image, kernel, iterations=1)
# erosion
def erode(image):
    kernel = np.ones((5, 5), np.uint8)
    return cv2.erode(image, kernel, iterations=1)
# opening - erosion followed by dilation
def opening(image):
   kernel = np.ones((5, 5), np.uint8)
    return cv2.morphologyEx(image, cv2.MORPH_OPEN, kernel)
# canny edge detection
def canny(image):
    return cv2.Canny(image, 100, 200)
# skew correction
def deskew(image):
    coords = np.column_stack(np.where(image > 0))
    angle = cv2.minAreaRect(coords)[-1]
    if angle < -45:
        angle = -(90 + angle)
    else:
        angle = -angle
    (h, w) = image.shape[:2]
    center = (w // 2, h // 2)
    M = cv2.getRotationMatrix2D(center, angle, 1.0)
    rotated = cv2.warpAffine(
        image, M, (w, h), flags=cv2.INTER_CUBIC, borderMode=cv2.BORDER_REPLICATE
    return rotated
```

```
# template matching
def match_template(image, template):
    return cv2.matchTemplate(image, template, cv2.TM_CCOEFF_NORMED)
```

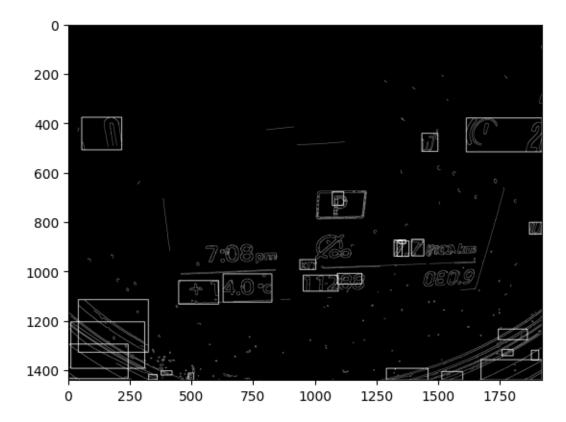


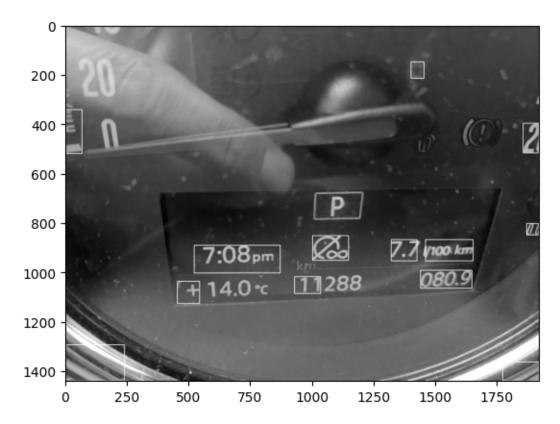


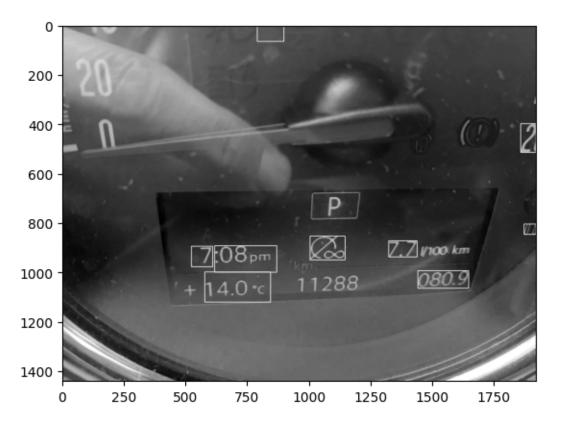
```
[]: img_pp = remove_noise(image)
img_pp = get_grayscale(img_pp)
```



dict_keys(['level', 'page_num', 'block_num', 'par_num', 'line_num', 'word_num',
'left', 'top', 'width', 'height', 'conf', 'text'])







Basically, preprocessing makes the image better but the performance is still quite bad.

0.2 EAST (Efficient accurate scene text detector)

Took this code from https://medium.com/technovators/scene-text-detection-in-python-with-east-and-craft-cbe03dda35d5

Found the model ('frozen_east_text_detection.pb) from a dropbox link in OpenCV_exra GitHub repo

adapted the code a bit

```
[]: import os, sys
     import numpy as np
     import cv2
     import time
     import re
     from imutils.object_detection import non_max_suppression
     def east_detect(image, newSize=320):
         layerNames = ["feature_fusion/Conv_7/Sigmoid", "feature_fusion/concat_3"]
         orig = image.copy()
         if len(image.shape) == 2:
             image = cv2.cvtColor(image, cv2.COLOR GRAY2RGB)
         (H, W) = image.shape[:2]
         # set the new width and height and then determine the ratio in change
         # for both the width and height: Should be multiple of 32
         (newW, newH) = (newSize, newSize)
         rW = W / float(newW)
         rH = H / float(newH)
         # resize the image and grab the new image dimensions
         image = cv2.resize(image, (newW, newH))
         (H, W) = image.shape[:2]
         net = cv2.dnn.readNet(
             "/Users/roanraina/LOCAL/ICBC/ICBC-Odometer-Recognition/eda/tesseract/

¬frozen_east_text_detection.pb"

         )
         blob = cv2.dnn.blobFromImage(
             image, 1.0, (W, H), (123.68, 116.78, 103.94), swapRB=True, crop=False
```

```
start = time.time()
net.setInput(blob)
(scores, geometry) = net.forward(layerNames)
(numRows, numCols) = scores.shape[2:4]
rects = []
confidences = []
# loop over the number of rows
for y in range(0, numRows):
    # extract the scores (probabilities), followed by the geometrical
    # data used to derive potential bounding box coordinates that
    # surround text
    scoresData = scores[0, 0, y]
    xData0 = geometry[0, 0, y]
    xData1 = geometry[0, 1, y]
    xData2 = geometry[0, 2, y]
    xData3 = geometry[0, 3, y]
    anglesData = geometry[0, 4, y]
    for x in range(0, numCols):
        # if our score does not have sufficient probability, ignore it
        # Set minimum confidence as required
        if scoresData[x] < 0.5:</pre>
            continue
        # compute the offset factor as our resulting feature maps will
        # x smaller than the input image
        (offsetX, offsetY) = (x * 4.0, y * 4.0)
        # extract the rotation angle for the prediction and then
        # compute the sin and cosine
        angle = anglesData[x]
        cos = np.cos(angle)
        sin = np.sin(angle)
        # use the geometry volume to derive the width and height of
        # the bounding box
        h = xData0[x] + xData2[x]
        w = xData1[x] + xData3[x]
        # compute both the starting and ending (x, y)-coordinates for
        # the text prediction bounding box
        endX = int(offsetX + (cos * xData1[x]) + (sin * xData2[x]))
        endY = int(offsetY - (sin * xData1[x]) + (cos * xData2[x]))
        startX = int(endX - w)
        startY = int(endY - h)
        # add the bounding box coordinates and probability score to
```

```
# our respective lists
                 rects.append((startX, startY, endX, endY))
                 confidences.append(scoresData[x])
         boxes = non_max_suppression(np.array(rects), probs=confidences)
         # loop over the bounding boxes
         for startX, startY, endX, endY in boxes:
             # scale the bounding box coordinates based on the respective
             # ratios
             startX = int(startX * rW)
             startY = int(startY * rH)
             endX = int(endX * rW)
             endY = int(endY * rH)
             # draw the bounding box on the image
             cv2.rectangle(orig, (startX, startY), (endX, endY), (0, 255, 0), 2)
         print(time.time() - start)
         # Loop over the detected text regions and extract the text
         texts = \Pi
         for box in boxes:
             (x, y, w, h) = box
             # Scale the coordinates back to the original dimensions
             x = int(x * rW)
             y = int(y * rH)
             w = int(w * rW)
             h = int(h * rH)
             # Crop the text region from the original image
             crop_img = orig[y : y + h, x : x + w]
             # Convert the cropped region to grayscale and apply thresholding
             gray = cv2.cvtColor(crop_img, cv2.COLOR_BGR2GRAY)
             thresh = cv2.threshold(gray, 0, 255, cv2.THRESH_BINARY_INV + cv2.
      →THRESH_OTSU) [1]
             # Perform OCR using Tesseract
             text = pytesseract.image_to_string(
                 thresh, config="--psm 11 -c tessedit_char_whitelist=0123456789"
             # Print the extracted text
             text = re.findall(r"\d+", text)
             texts += text
         return orig, texts
[]: image = cv2.imread(
         "/Users/roanraina/LOCAL/ICBC/ICBC-Odometer-Recognition/tmp/processed_April_
```

⇔2021_229.jpg"

```
image = cv2.cvtColor(image, cv2.COLOR_BGR2RGB)

out_image, text = east_detect(image)

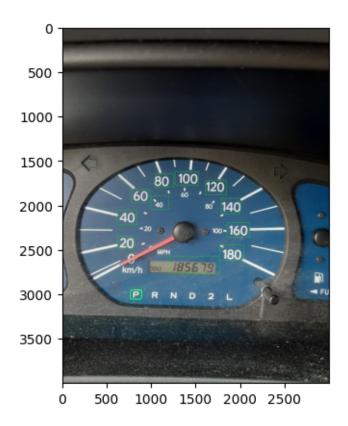
plt.imshow(out_image)
plt.show()

print(text)
```



```
['080', '7', '1', '1288', '080', '9', '100', '080', '9']
```

```
plt.show()
print(text)
```



```
image = cv2.cvtColor(image, cv2.COLOR_BGR2RGB)

out_image, text = east_detect(image)

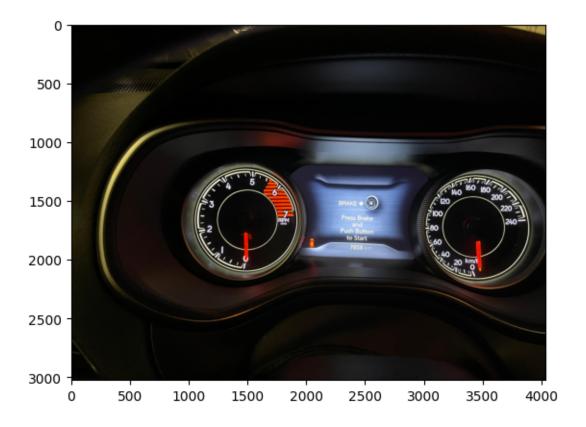
plt.imshow(out_image)
plt.show()

print(text)
```



```
['100', '1', '4', '1', '8', '1', '24', '7', '18', '60', '3', '275', '2', '1000', '8', '5', '3', '22', '3', '3', '1', '75', '7', '550', '4', '4', '8', '4', '75', '7', '550', '5', '4', '4', '14', '4', '550', '4', '4', '16', '3', '3', '97', '3', '407', '166', '100', '20', '1', '20', '120', '2']
```

```
out_image, text = east_detect(image)
plt.imshow(out_image)
plt.show()
print(text)
```



[]

Nope, not at all. None of them seem to work.

It seems EAST is better at identifying where text/numbers are located compared to tesseract but tesseract still cant read the numbers.

Looks like for the last two, the image quality is just too low.... which i realize can be fixed

```
out_image, text = east_detect(image, newSize=1280)
plt.imshow(out_image)
plt.show()
print(text)
```

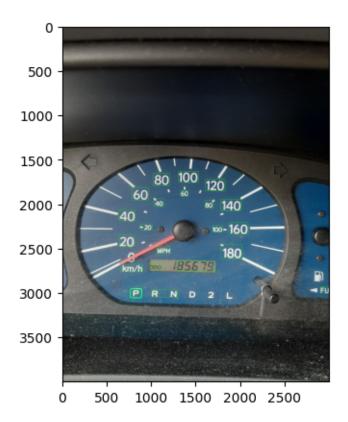


['00', '7', '11288', '080', '9', '140', '0', '11288']

```
[]: image = cv2.imread(
    "/Users/roanraina/LOCAL/ICBC/ICBC-Odometer-Recognition/tmp/processed_April_
    \( \times 2021_104.jpg" \)
    image = cv2.cvtColor(image, cv2.COLOR_BGR2RGB)

out_image, text = east_detect(image, newSize=1280)

plt.imshow(out_image)
    plt.show()
```



['75', '4', '2', '4', '3', '4', '0', '18', '7', '45', '9', '3', '5', '2', '7', '8', '4', '3', '5', '1', '7', '7', '7', '7', '2', '7', '0', '3', '7', '4', '78', '8', '3', '0', '4', '4', '2', '4', '2', '4', '2', '3', '2', '2', '3', '8', '1', '4', '3', '7', '4', '1', '6', '8', '4', '8', '4', '4', '2', '4', '4', '7', '2', '5', '2', '7', '1', '8', '4', '5', '7', '7', '3', '4', '4', '3', '5', '4', '2', '1', '5', '0', '4', '4', '2', '8', '2', '1', '2', '5', '7', '4', '6', '1', '7', '4', '2', '2', '2', '4', '1', '15', '74', '1', '4', '45', '1', '8', '4', '7', '4', '2', '1', '4', '2', '7', '4', '3', '40', '4', '2', '4', '2', '7', '7', '7', '5', '100', '60', '120', '80', '140', '100', '160', '2', '000', '120', '40', '80', '140', '100160', '7', '80', '23', '3', '2', '1', '3', '4', '36', '5', '1', '7', '7', '4', '1', '8', '4', '47', '8', '8', '4', '4', '7', '2', '2', '5', '3', '9', '7', '4', '2', '8', '5', '4', '3', '2', '0', '3', '7', '8', '7', '42', '1', '0', '4', '4', '1', '2', '12', '74', '4', '1', '6', '7', '4', '7', '2', '7', '4', '2', '5', '6', '7', '3', '3', '4', '7', '7', '7', '4', '3', '8', '4', '8', '1', '7', '7', '4', '4', '2', '3', '3', '3', '0', '4', '2', '2', '4', '03', '2', '4', '4', '9', '7', '1', '4', '4', '2', '3', '8', '1', '1', '2', '4', '3', '2', '7', '4', '8', '4', '1', '2', '2', '2', '4', '4', '7', '7', '7', '1', '1', '2', '5', '23', '7', '7', '3', '1', '2', '1', '1', '4', '2', '4', '0', '3', '3', '4', '4', '7', '3', '7', '3', '7', '3', '2', '3', '3', '3', '3', '0', '4',

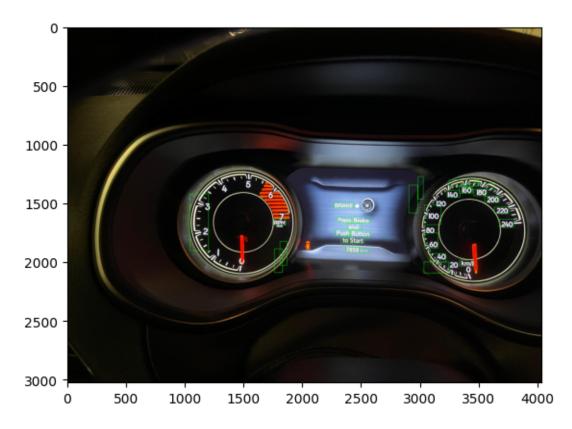
```
'057', '1', '20', '8', '4', '4', '2']
```



['4', '45', '1', '4', '4', '3', '4', '15', '8', '72', '5', '1', '1', '1', '5', '2', '5', '4', '24', '818', '2', '3', '4', '41', '3', '1', '28', '275', '4', '28', '7', '44', '40', '160', '00', '20', '10', '180', '120', '200', '8', '2', '7', '7', '7', '7', '46', '3', '3', '8', '1', '8', '275', '3', '4', '7', '22', '1', '4', '9', '1', '2', '21', '4', '7', '1', '4', '6', '7', '3', '1', '1', '6', '3', '53', '8', '8', '72', '3', '4', '1', '48', '2', '2', '3', '163', '4',

```
'1', '3', '8', '2', '0', '2', '4', '4', '4', '4', '3', '2', '5', '3', '74',
    '72', '4', '140', '3', '1607', '120', '7', '41', '8', '18', '48', '8', '4', '2',
    '8', '2', '7', '2', '2', '18', '2', '178', '275', '3', '24', '537', '4', '4',
    '4', '4', '4', '72', '2', '40', '20', '20', '0', '416', '2', '275', '1', '3',
    '2', '18', '1', '3', '275', '5', '4', '2', '2', '9', '2', '1', '4', '7', '4',
    '4', '8', '8', '4', '3', '4', '4', '4', '38', '1', '3', '2', '8', '4', '3', '3',
    '1', '4', '1', '4', '78', '6', '4', '3', '3', '3', '1', '3', '35', '4', '4',
    '4', '54', '8', '1', '275', '3', '7', '4', '1', '1', '7', '1', '7', '596',
    '300', '5', '9', '0', '1', '4']
[]: image = cv2.imread(
         "/Users/roanraina/LOCAL/ICBC/ICBC-Odometer-Recognition/tmp/processed\_BEU_{\sqcup}
     →Completed_1.jpg"
     image = cv2.cvtColor(image, cv2.COLOR_BGR2RGB)
     out_image, text = east_detect(image, newSize=1280)
     plt.imshow(out_image)
    plt.show()
```

print(text)



```
['2', '09', '240', '60', '0', '20', '240', '60', '0', '20', '0', '7', '240', '60', '0', '1', '20', '60', '10', '200', '220', '240', '4', '200', '220', '240', '1', '60', '0', '2', '200', '240', '40', '4', '60', '0', '240', '240', '0', '20', '240', '0', '20', '240', '0', '20', '20', '240', '0', '20', '20', '240', '0', '0', '0', '0', '20', '20', '240', '200', '2', '09', '240', '60', '0', '60', '0', '60', '0', '20', '240', '20', '40', '20', '20', '240', '20', '20', '240', '20', '20', '20', '20', '20', '20', '20', '20', '20', '20', '20', '20', '20', '20', '20', '20', '20', '20', '20', '20', '20', '20', '20', '20', '20', '20', '20', '20', '20', '20', '20', '20', '20', '20', '20', '20', '20', '20', '20', '20', '20', '20', '20', '20', '20', '20', '20', '20', '20', '20', '20', '20', '20', '20', '20', '20', '20', '20', '20', '20', '20', '20', '20', '20', '20', '20', '20', '20', '20', '20', '20', '20', '20', '20', '20', '20', '20', '20', '20', '20', '20', '20', '20', '20', '20', '20', '20', '20', '20', '20', '20', '20', '20', '20', '20', '20', '20', '20', '20', '20', '20', '20', '20', '20', '20', '20', '20', '20', '20', '20', '20', '20', '20', '20', '20', '20', '20', '20', '20', '20', '20', '20', '20', '20', '20', '20', '20', '20', '20', '20', '20', '20', '20', '20', '20', '20', '20', '20', '20', '20', '20', '20', '20', '20', '20', '20', '20', '20', '20', '20', '20', '20', '20', '20', '20', '20', '20', '20', '20', '20', '20', '20', '20', '20', '20', '20', '20', '20', '20', '20', '20', '20', '20', '20', '20', '20', '20', '20', '20', '20', '20', '20', '20', '20', '20', '20', '20', '20', '20', '20', '20', '20', '20', '20', '20', '20', '20', '20', '20', '20', '20', '20', '20', '20', '20', '20', '20', '20', '20', '20', '20', '20', '20', '20', '20', '20', '20', '20', '20', '20', '20', '20', '20', '20', '20', '20', '20', '20', '20', '20', '20', '20', '20', '20', '20', '20', '20', '20', '20', '20', '20', '20', '20', '20', '20', '20', '20', '20', '20', '20', '20', '20', '20', '20', '20', '20', '20', '20', '20', '20', '20', '20', '
```

So increasing the size of the image seems to improve EAST performance (but takes longer) across the board. This in turn seems to improve Tesseract's performance as shown in the first example. However, the remaining three examples still fail when it comes to transcribing the number.