# Implementation of Morphology Filters and Segmentation for Improving the Simple Python Optical Character Recognition (OCR) System

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#### **Description**

In this assignment, I implemented a simple Python Optical Character Recognition (OCR) system using Tesseract. Then, I applied and compared several types of morphology filters and segmentation in the image to improve OCR results.

#### **Tools**

- Tesseract
- Python 3.7
- · Some Python libraries:
  - Open-CV
  - PyTesseract

#### Installation

#### **Tesseract**

Tesseract is an open-source OCR engine. It began as a Ph.D. research project in HP Labs Bristol, then developed by **HP** between 1984 and 1994. In 2005, HP released Tesseract as an open-source software. Since 2006 it is developed by **Google**.

1. Using Homebrew to install Tesseract in Mac

```
$ brew install tesseract

# This formula contains only the "eng", "osd", and "snum" language data files.

# If you need any other supported languages, run `brew install tesseract-lang`.

# To install other languages

$ brew install tesseract-lang
```

Installation of Tesseract in other OS will be different.

#### **Implementation**

```
In [1]:
```

```
# import libraries
from PIL import Image
import cv2
import pytesseract
import matplotlib.pyplot as plt
import numpy as np
import os
from math import ceil
```

#### In [2]:

```
# constant global variables

IMG_PATH = "img/"
IMG_NAME = "teks.png"

DEST_PATH = "preprocessed_img/" + IMG_NAME.split('.')[0] + "/"

OCR_PATH = "result/" + IMG_NAME.split('.')[0] + "/"
```

```
# OS Handler
def get list(path, list type='img'):
   my_list = []
   if list_type == 'img':
        valid_list = ['.jpg', '.png', '.jpeg']
    elif list_type == 'txt':
       valid list = ['.txt']
    # get all image name
    for f in os.listdir(path):
        ext = os.path.splitext(f)[1]
        if not (ext.lower() in valid_list):
            continue
       my list.append(f)
    return my_list
def check dir(path):
    # make directory
    if not os.path.exists(path):
        os.makedirs(path)
```

#### 1. Preprocessing

In [4]:

```
# Preprocessing
# get grayscale image
def get_grayscale(image):
   return cv2.cvtColor(image, cv2.COLOR_BGR2GRAY)
# # noise removal - using median filter
# def remove noise(image, kernel=5):
     return cv2.medianBlur(image, kernel)
# # 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)
```

#### 1.1 Morphology Filters

```
In [5]:
```

```
# 1. Dilation
def dilate(image, kernel=np.ones((5,5),np.uint8)):
    return cv2.dilate(image, kernel, iterations = 1)

# 2. Erosion
def erode(image, kernel=np.ones((5,5),np.uint8)):
    return cv2.erode(image, kernel, iterations = 1)

# 3. Opening - erosion followed by dilation
def opening(image, kernel=np.ones((5,5),np.uint8)):
    return cv2.morphologyEx(image, cv2.MORPH_OPEN, kernel)

# 4. Closing - dilation followed by erosion
def closing(image, kernel=np.ones((5,5),np.uint8)):
    return cv2.morphologyEx(image, cv2.MORPH_CLOSE, kernel)
```

```
# 5. Canny edge detection
def canny(image):
    return cv2.Canny(image, 100, 200)
```

#### 1.2 Segmentation - Thresholding

```
In [6]:
```

```
# 1. Thresholding
def thresholding(image):
    return cv2.threshold(image, 0, 255, cv2.THRESH_BINARY + cv2.THRESH_OTSU)[1]

# 2. Adaptive Thresholding using mean value
def adaptive_mean_thresholding(image):
    return cv2.adaptiveThreshold(image, 255, cv2.ADAPTIVE_THRESH_MEAN_C, cv2.THRESH_BINARY, 199, 5)

# 2. Adaptive Thresholding using gaussian value
def adaptive_gaussian_thresholding(image):
    return cv2.adaptiveThreshold(image, 255, cv2.ADAPTIVE_THRESH_GAUSSIAN_C, cv2.THRESH_BINARY, 199, 5)
```

#### 1.3 Applying Morphology and Segmentation

In [7]:

```
def apply_filter(image):
   gray_image = get_grayscale(image)
   scenarios = {
        'thresholding': thresholding,
        'adp-mean-thresholding': adaptive mean thresholding,
        'adp-gaussian-thresholding': adaptive_gaussian_thresholding,
        'opening': opening,
        'closing': closing,
        'dilate': dilate,
       'erode': erode,
       'canny': canny
   for key, function in scenarios.items():
        prep image = function(gray image)
       filename = key + '_' + IMG_NAME
        # save file
           cv2.imwrite(DEST_PATH + filename, prep_image[:,:,::-1])
       except IndexError:
            cv2.imwrite(DEST PATH + filename, prep image)
```

#### 2. Processing

#### 2.1 Optical Character Recognition (OCR)

```
In [8]:
```

```
def ocr(image, image_name):
    filename = image_name + '_output.txt'
    # Adding custom options
    custom_config = r'--oem 3 --psm 6'
    with open(OCR_PATH + filename, 'w') as f:
        f.write(pytesseract.image_to_string(image, config=custom_config))
```

#### 2.2 Displaying Results

```
In [9]:
```

```
def generate_plot(total_img, num_img_each_row=2):
    row = ceil(total_img / num_img_each_row) * 100
    col = num_img_each_row * 10
```

```
return row + col
def display img(dest path, fig size=(17, 20)):
   imgs_name = [IMG_NAME] + get_list(dest_path)
   # generate plot
   plot = generate plot(len(imgs name))
    # plotting
   fig = plt.figure(figsize=fig_size)
   for i in range(plot, plot + len(imgs_name)):
       title = imgs_name[i - plot].split('_')[0]
       img_path = IMG_PATH + imgs_name[i - plot] if i == plot else dest_path + imgs_name[i - plot]
       img = cv2.imread(img_path)[:,:,::-1]
       plt.subplot(i + 1), plt.imshow(img)
       plt.title(title)
       plt.xticks([]), plt.yticks([])
   plt.show()
def display_ocr_result():
   files = get list(OCR PATH, 'txt')
   for filename in files:
       f = open(OCR PATH + filename, 'r')
       file contents = f.read()
       print ('======')
       print (filename)
       print ('====="')
       print (file_contents)
       print ('====="')
       f.close()
# compare file
# files = get list(OCR PATH, 'txt')
# ori_file = 'original_' + IMG_NAME.split('.')[0] + '_output.txt'
# files.remove(ori_file)
# for file name in files:
     with open(OCR_PATH + ori_file, 'r') as file1:
#
         with open(OCR_PATH + file_name, 'r') as file2:
#
             same = set(file1).intersection(file2)
     same.discard('\n')
#
     with open(OCR_COMPARE_PATH + 'ori_' + file_name.split('_')[0] + '.txt', 'w') as file_out:
         total\_char = 0
#
         for line in same:
#
             file out.write(line)
#
             total_char += len(line)
#
         print('ori_{}.txt ===> {}'.format(file_name.split('_')[0], total_char))
```

#### 2.3 Main Program

```
In [10]:
```

```
# check and create all dirs
check_dir(DEST_PATH)
check_dir(OCR_PATH)

# preprocessing
image = cv2.imread(IMG_PATH + IMG_NAME)[:,:,::-1]
apply_filter(image)

# OCR original img
image = cv2.imread(IMG_PATH + IMG_NAME)[:,:,::-1]
ocr(image, 'original_' + IMG_NAME.split('.')[0])

# OCR preprocessed img
imgs = get_list(DEST_PATH)
for img_name in imgs:
    image = cv2.imread(DEST_PATH + img_name)[:,:,::-1]
    ocr(image, img_name.split('.')[0])
```

#### In [11]:

```
# display preprocessed image
display_img(DEST_PATH)
```

teks.png

## Region-based segmentation

Let us first determine markers of the coins and the background. These markers are pixels that we can label unambiguously as either object or background. Here, the markers are found at the two extreme parts of the histogram of grey values:

>>> markers = np.zeros\_like(coins)

dilate



opening

## Region-based segmentation

Let us first determine markers of the coins and the background. These markers are pixels that we can label unambiguously as either object or background. Here, the markers are found at the two extreme parts of the lastogram of grey values:

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# Region-based segmentation

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