Edge Detection & Morphology

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使用sobel kernel對原圖進行kernel運算,分別計算出gx, gy,最後將兩維度之梯度各取平方相加開根號得到梯度值magnitude,即為本圖。

```
def cal_gradient(image, operator_x, operator_y):
    height, width = image.shape
    ksize = 3
    #image = image.astype(np.float32)
    pad_size = ksize//2
    padded_image = np.pad(image, pad_size, mode = 'reflect')
    gx = np.zeros(image.shape, dtype=np.float32)
    gy = np.zeros(image.shape, dtype=np.float32)

for y in range(height):
    for x in range(width):
        gx[y, x] = np.sum(padded_image[y:y+ksize, x:x+ksize]*operator_x)
        gy[y, x] = np.sum(padded_image[y:y+ksize, x:x+ksize]*operator_y)
    return gx, gy
```

Gradient計算



Sobel

與sobel相同作法,但不同之處在於使用的是 prewitt kernel

```
def cal_gradient(image, operator_x, operator_y):
    height, width = image.shape
    ksize = 3
    #image = image.astype(np.float32)
    pad_size = ksize//2
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    gx = np.zeros(image.shape, dtype=np.float32)
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for y in range(height):
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        gx[y, x] = np.sum(padded_image[y:y+ksize, x:x+ksize]*operator_x)
        gy[y, x] = np.sum(padded_image[y:y+ksize, x:x+ksize]*operator_y)
    return gx, gy
```

Gradient計算



Prewitt

- 1. 先使用高斯filter降低雜訊圖形後計算圖形之梯度 值和角度
- 2. 接著依照梯度方向和鄰居比較,若小於其中一個鄰居則梯度值設為0(non-maxima suppression)
- 3. 接著透過高/低threshold去判斷強弱邊
- 4. 最後檢查弱邊周圍有強邊能連接,有的話則加入邊緣



Canny

```
for y in range(1, H-1):
    for x in range(1, W-1):
    pixel = G_theta[y,x]
    if((pixel < 67.5 and pixel >= 22.5) or (pixel >= -157.5 and pixel < -112.5)):
        G_dir[y, x] = DIRECTION.SLASH
    elif(pixel < 112.5 and pixel >= 67.5 or (pixel < -67.5 and pixel >= -112.5)):
        G_dir[y, x] = DIRECTION.VERTICAL
    elif(pixel < -22.5 and pixel >= -67.5 or (pixel < 157.5 and pixel >= 112.5)):
        G_dir[y, x] = DIRECTION.BACK_SLASH
    else:
        G dir[y, x] = DIRECTION.HORIZON
```

判斷梯度方向

強弱邊連接

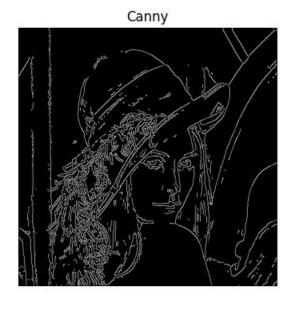
```
#做non-maxima supression
for y in range(1, H-1):
 for x in range(1, W-1):
   cur dir = G dir[y, x]
   cur_magnitude = G_magnitude[y, x]
   v1 = 0
   v2 = 0 #鄰居
   if(cur dir == DIRECTION.VERTICAL):
     v1 = G_magnitude[y+1, x]
     v2 = G magnitude[y-1, x]
    elif(cur dir == DIRECTION.HORIZON):
     v1 = G magnitude[y, x+1]
     v2 = G magnitude[y, x-1]
    elif(cur dir == DIRECTION.SLASH):
     v1 = G magnitude[y+1, x+1]
     v2 = G magnitude[y-1, x-1]
    elif(cur_dir == DIRECTION.BACK_SLASH):
     v1 = G_magnitude[y-1, x+1]
     v2 = G magnitude[y+1, x-1]
   else:
     v1 = 255
     v2 = 255
    if(cur magnitude < v1 or cur magnitude < v2):</pre>
     G_{supressed}[y, x] = 0
   else:
     G supressed[v, x] = cur magnitude
```

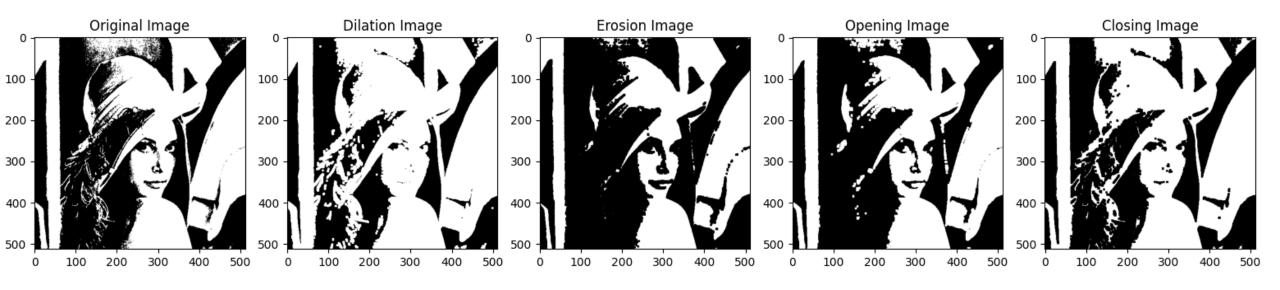
Non-maxima supression











*structure element為全為1(255)的3*3圖形

將structure element去掃一遍圖,只要和原圖中白色(255)的部分有交集(有碰到),皆會被轉換成1



Dilation

Erosion實作起來則較dilation複雜,需要整個 structure element都在對上原圖都符合才可保留點

```
-開始設定result image全部為黑的
#後續判斷時若"kernel有整個在圖內則kernel該點會是255(白的),而其他部分被erode掉(黑的)'
#new_image = np.full(image.shape, 0)
new image = np.full((image.shape), 0, dtype=np.uint8)
pad size = max(len(arr) for arr in kernel)
pad_size = max(pad_size, kernel.shape[0])
pad_size = pad_size//2
pixels = []
for y in range(H):
  for x in range(W):
    flag = True
    for i in range(kernel.shape[0]):
     cur_y = y+i-pad_size
      for j in range(kernel[i].size):
       cur x = x+j-pad size
       if cury < 0 or cury >= H or curx < 0 or curx >= W:
         continue
       if image[cur_y][cur_x] != kernel[i][j]:
         flag = False
         break
    if flag == True:
      pixels.append((y,x))
for y, x in pixels:
  new_image[y][x] = 255
```



Erosion

即先做erosion後再做dilation,可以用於將不小心因 雜訊連接起來的物件分開



Opening

即先做dilation後再做erosion,可以用於填補破洞密集的一直在0和255切換的區塊,如這張原圖帽子上方的部分



Opening