### **Travail pratique 4**

INF600F - Traitement d'images (H2022, UQÀM)

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```
In [53]: from scipy import ndimage as ndi
         from skimage import (
             color, feature, filters, measure, morphology, segmentation, util
         import imageio
         import numpy as np
         import matplotlib.pyplot as plt
         import random
         from skimage import color
         from skimage import io
         from scipy.ndimage.morphology import distance transform edt
         from skimage.morphology import binary opening, thin, label
         from sklearn.cluster import KMeans
         from skimage.feature import hog, peak local max
         from skimage.morphology import reconstruction
         import skimage.measure as measure
         from skimage.transform import resize
         from skimage.filters import threshold otsu
         from skimage.segmentation import watershed
```

### Exercice 1 : Reconnaissance de texte

### Partie 1

```
In [54]: # Lire 1'image
         img = imageio.imread('tp4 ex1.png')
         # Prétraitement
         # Convertir en ton de gris + convertir les pixels en floats
         img gris = img.mean(axis=2).astype(float)
         # Normaliser l'intensité
         img norm = img gris / np.max(img gris)
         # Inverser les intensités
         img inverse = np.abs(img norm -1)
         plt.figure(figsize=(20,20))
         plt.subplot(141)
         plt.title('Image originale', fontsize=15)
         plt.imshow(img, cmap='gray')
         plt.subplot(142)
         plt.title('Image en ton de gris', fontsize=15)
         plt.imshow(img gris, cmap='gray')
         plt.subplot(143)
         plt.title('Intensités normalisées', fontsize=15)
         plt.imshow(img norm, cmap='gray')
         plt.subplot(144)
         plt.title('Intensités inversées', fontsize=15)
         plt.imshow(img inverse, cmap = 'gray');
         plt.show()
         # Segmentation
         # Retirer l'arriere plan de l'image + retirer les bordures
         source = np.copy(img inverse)
         source[1:-1, 1:-1] = img inverse.min()
         masque = img inverse
         dilatation = reconstruction(source, masque, method='dilation')
         img noBG = img inverse - dilatation
         # Otsu
         otsu = threshold otsu(img noBG)
         img otsu=(img noBG > otsu)
         # Étiquetage de l'image
         img etiquette = measure.label(img otsu)
```

```
plt.figure(figsize=(20,20))
plt.subplot(131)
plt.title('Image sans arrière-plan', fontsize=15)
plt.imshow(img noBG, cmap='gray')
plt.subplot(132)
plt.title('Image segmentée Otsu', fontsize=15)
plt.imshow(img otsu, cmap='gray')
plt.subplot(133)
plt.title('Image étiquetée', fontsize=15)
plt.imshow(img etiquette, cmap='gray')
plt.show()
# Détecter les lettres
lettres = []
for region in measure.regionprops(img etiquette):
    minr, minc, maxr, maxc = region.bbox
    lettres.append(img otsu[minr:maxr, minc:maxc])
for i in range(len(lettres)):
    lettres[i] = np.pad(lettres[i], pad_width=3, mode='constant', constant_values=0)
    lettres[i] = resize(lettres[i], (24,24))
```

#### Image originale

The brightness levels in a scene or image, together with the spatial relationships among these levels, comprise the total input and output data available to the digital image processing operation, these are the elements of the image processing operation, these are the elements of the image which are being manipulated and which represent the major concern of the image processing professional. However, many other factors affect image brightness relations and can distort or obscure the outcome of any image processing experiment. These factors, which consist of a long chain of transmitters, transducers, signal conditioners and processors, are in aggregate commonly called the image chain. An understanding of the image chain is essential to the design of image processing systems.

0 250 500 750 1000 1250 1500 1750

#### lmage en ton de gris

The brightness levels in a scene or image, together with the spatial relationships among these levels, comprise the total input and output data available to the digital image processing operation, these are the elements of the image processing operation, these are the elements of the image processing operation of the image processing operation of the image processing professional. However, many other factors affect image brightness relations and can distort or obscure the outcome of any image processing experiment. These factors, which consist of a long chain of transmitters, transducers, signal conditioners and processors, are in aggregate commonly called the image chain. An understanding of the image chain is essential to the design of image processing systems.

0 250 500 750 1000 1250 1500 1750

#### Intensités normalisées

The brightness levels in a scene or image, together with the spatial relationships among these levels, comprise the total input and output data available to the digital image processing operation, these are the elements of the image which are being manipulated and which represent the major concern of the image processing professional. However, many other factors affect image brightness relations and can distort or obscure the outcome of any image processing experiment. These factors, which consist of a long chain of transmitters, transducers, signal conditioners and processors, are in aggregate commonly called the image chain. An understanding of the image chain is essential to the design of image processing systems.

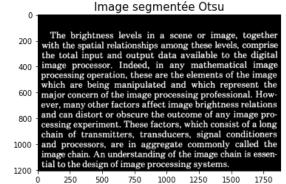
0 250 500 750 1000 1250 1500 1750

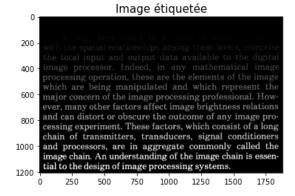
#### Intensités inversées

The brightness levels in a scene or image, together with the spatial relationships among these levels, comprise the total input and output data available to the digital image processor. Indeed, in any mathematical image processing operation, these are the elements of the image which are being manipulated and which represent the major concern of the image processing professional. However, many other factors affect image brightness relations and and distort or obscure the outcome of any image processing experiment. These factors, which consist of a long chain of transmitters, transducers, signal conditioners and processors, are in aggregate commonly called the image chain. An understanding of the image chain is essential to the design of image processing systems.

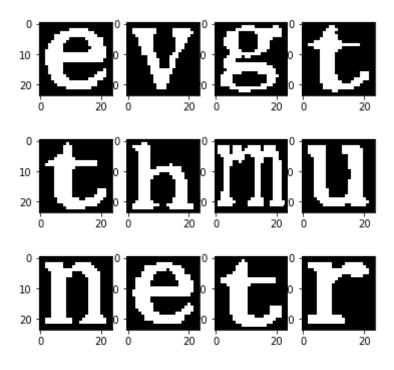
0 250 500 750 1000 1250 1500 1750

			Imag	je san	s arriè	ere-pla	an	
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200	with	the sp	htness atial rela	ationshi	os amon	g these	levels, o	omprise
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600	maj	or conc	being n ern of th other fa	e image	process	sing pro	fessiona	l. How-
800	cess	ing exp	stort or o periment ransmitt	. These	factors,	which c	onsist o	f a long
1000	and	proces ge chair	sors, ar	e in ag derstand	gregate ling of t	commo	only cal e chain i	led the
1200	tial	to the c	lesign of	ımage p	rocessii	ng syste	ms.	
	0	250	500	750	1000	1250	1500	1750





```
In [55]: # Visualiser au hasard les lettres
         def display randomImages(images ar):
             selection = []
             for i in range(12):
                 x = random.randint(0, len(images ar))
                 selection.append(images ar[x])
             plt.subplot(141)
             plt.imshow(selection[0], cmap='gray')
             plt.subplot(142)
             plt.imshow(selection[1], cmap='gray')
             plt.subplot(143)
             plt.imshow(selection[2], cmap='gray')
             plt.subplot(144)
             plt.imshow(selection[3], cmap='gray')
             plt.show()
             plt.subplot(141)
             plt.imshow(selection[4], cmap='gray')
             plt.subplot(142)
             plt.imshow(selection[5], cmap='gray')
             plt.subplot(143)
             plt.imshow(selection[6], cmap='gray')
             plt.subplot(144)
             plt.imshow(selection[7], cmap='gray')
             plt.show()
             plt.subplot(141)
             plt.imshow(selection[8], cmap='gray')
             plt.subplot(142)
             plt.imshow(selection[9], cmap='gray')
             plt.subplot(143)
             plt.imshow(selection[10], cmap='gray')
             plt.subplot(144)
             plt.imshow(selection[11], cmap='gray')
             plt.show()
             return selection
         # Nombre de lettres
         rand = display randomImages(lettres)
         print(len(lettres))
```



724

# Question : Combien de lettres avez-vous détectées dans cette image ?

Nous avons détecté 724 lettres dans l'image.

Question: En observant les lettres segmentées, pouvez-vous identifier des situations pour lesquelles le pipeline de traitement d'images proposé plus haut échoue à extraire une lettre isolée? Que suggérez-vous pour améliorer les performances de l'extraction des lettres?

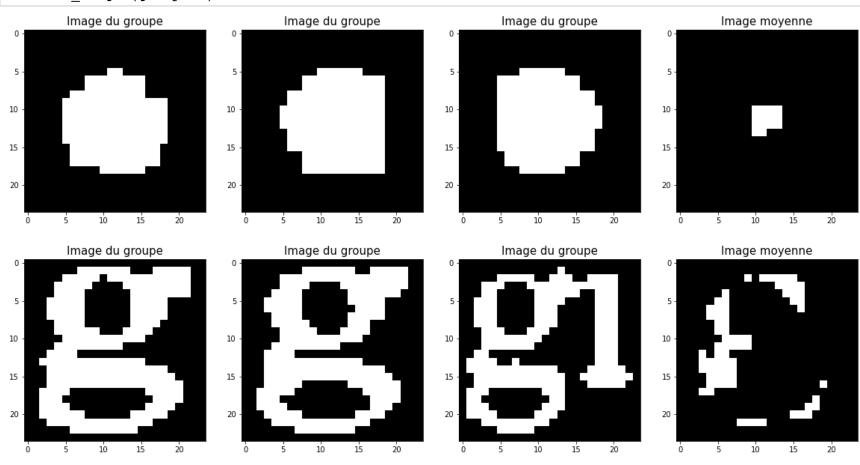
Certaines lettres sont trop collées sur leurs voisines, elles sont donc comptées comme une seule lettre dans le pipeline. Les signes de ponctuation sont également détecté comme une lettre, ce qui influence le résultat de visualisation. Pour améliorer les performances, la segmentation par ligne de partage des eaux (Watershed) permetterait de régler la problématique des lettres distincte.

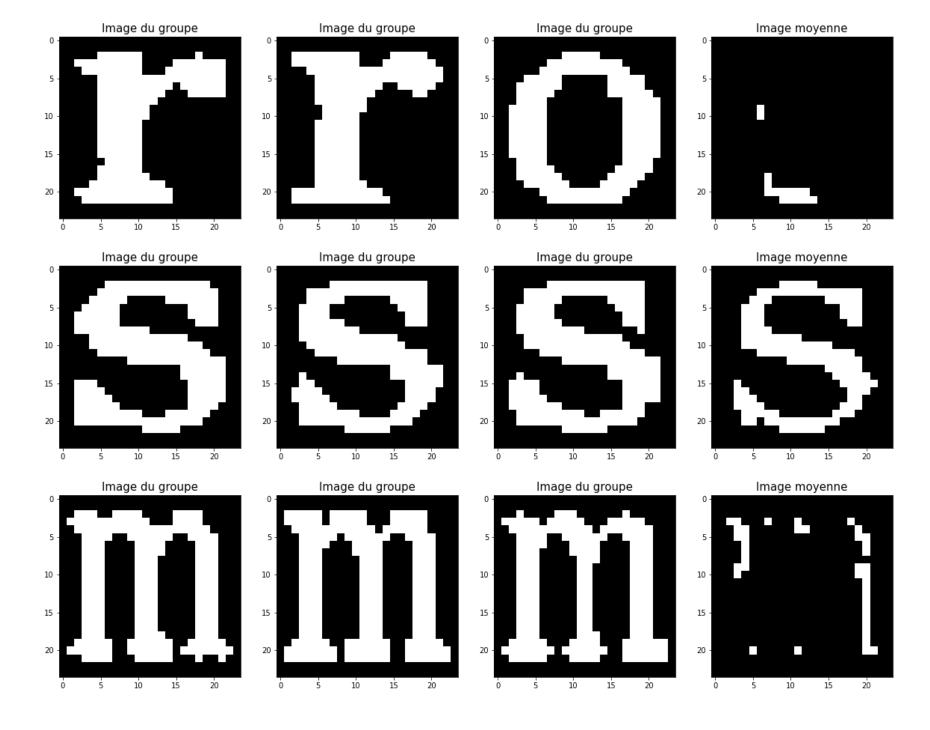
### Partie 2

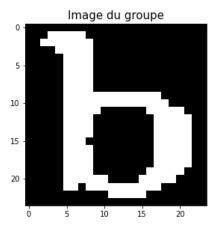
```
In [56]: f, hog l = hog(lettres[0], orientations=8, pixels_per_cell=(16, 16),
                             cells per block=(1, 1), visualize=True, multichannel=False)
         X = np.array(hog l.flatten())
         for i in range (1, len(lettres)):
             fd, hog image = hog(lettres[i], orientations=8, pixels per cell=(16, 16),
                             cells per block=(1, 1), visualize=True, multichannel=False)
             nouv = np.array(hog image.flatten())
             X = np.vstack((X, nouv))
         modele = KMeans(n clusters=26).fit(X)
         def init groupe(no groupe, modele):
             arr = []
             for i in range (len(modele)):
                 if modele[i] == no groupe:
                     arr.append(i)
             return arr
         groupe1 = init groupe(0, modele.labels )
         groupe2 = init groupe(1, modele.labels )
         groupe3 = init groupe(2, modele.labels )
         groupe4 = init groupe(3, modele.labels )
         groupe5 = init groupe(4, modele.labels )
         groupe6 = init groupe(5, modele.labels )
         groupe7 = init groupe(6, modele.labels )
         groupe8 = init groupe(7, modele.labels )
         groupe9 = init_groupe(8, modele.labels )
         groupe10 = init groupe(9, modele.labels )
         groupe11 = init groupe(10, modele.labels )
         groupe12 = init groupe(11, modele.labels )
         groupe13 = init groupe(12, modele.labels )
         groupe14 = init groupe(13, modele.labels )
         groupe15 = init groupe(14, modele.labels )
         groupe16 = init groupe(15, modele.labels )
         groupe17 = init groupe(16, modele.labels )
         groupe18 = init groupe(17, modele.labels )
         groupe19 = init groupe(18, modele.labels )
         groupe20 = init groupe(19, modele.labels )
         groupe21 = init_groupe(20, modele.labels )
         groupe22 = init groupe(21, modele.labels )
         groupe23 = init groupe(22, modele.labels )
```

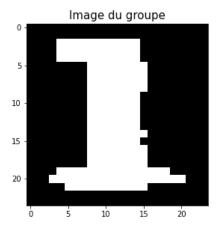
```
groupe24 = init groupe(23, modele.labels )
groupe25 = init groupe(24, modele.labels )
groupe26 = init groupe(25, modele.labels )
def afficher images (groupe):
    compte = 0
    moy = lettres[groupe[0]]
    for i in range(1, len(groupe)):
        moy = moy + lettres[groupe[i]]
        compte += 1
    moy = moy // compte
    plt.figure(figsize=(20,20))
    plt.subplot(141)
    plt.title('Image du groupe', fontsize=15)
    plt.imshow(lettres[groupe[0]], cmap='gray')
    plt.subplot(142)
    plt.title('Image du groupe', fontsize=15)
    plt.imshow(lettres[groupe[1]], cmap='gray')
    plt.subplot(143)
    plt.title('Image du groupe', fontsize=15)
    plt.imshow(lettres[groupe[2]], cmap='gray')
    plt.subplot(144)
    plt.title('Image moyenne', fontsize=15)
    plt.imshow(moy, cmap='gray')
    plt.show()
afficher images(groupe1)
afficher_images(groupe2)
afficher_images(groupe3)
afficher images(groupe4)
afficher images(groupe5)
afficher_images(groupe6)
afficher_images(groupe7)
afficher_images(groupe8)
afficher_images(groupe9)
afficher_images(groupe10)
afficher images(groupe11)
afficher images(groupe12)
afficher images(groupe13)
afficher images(groupe14)
afficher images(groupe15)
afficher images(groupe16)
afficher images(groupe17)
```

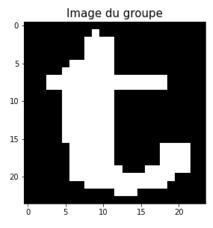
afficher\_images(groupe18) afficher\_images(groupe19) afficher\_images(groupe20) afficher\_images(groupe21) afficher\_images(groupe22) afficher\_images(groupe23) afficher\_images(groupe24) afficher\_images(groupe25) afficher\_images(groupe26)

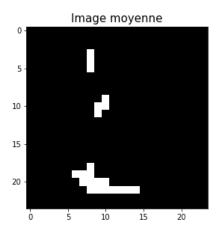


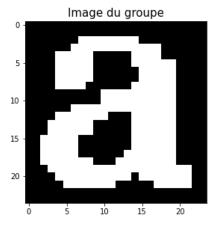


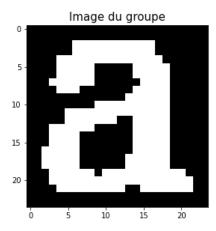


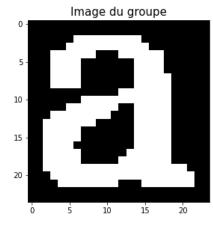


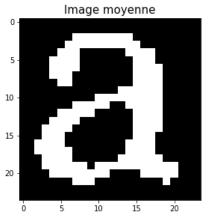


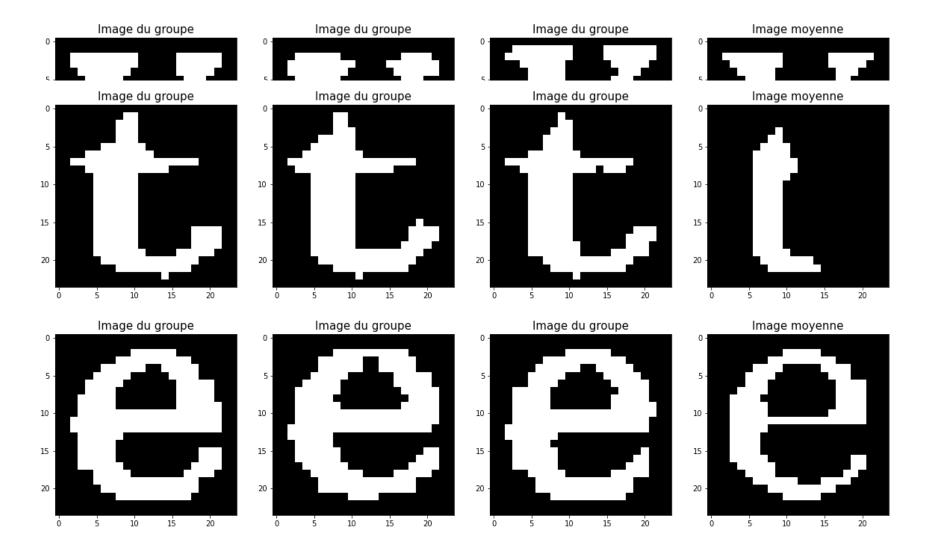


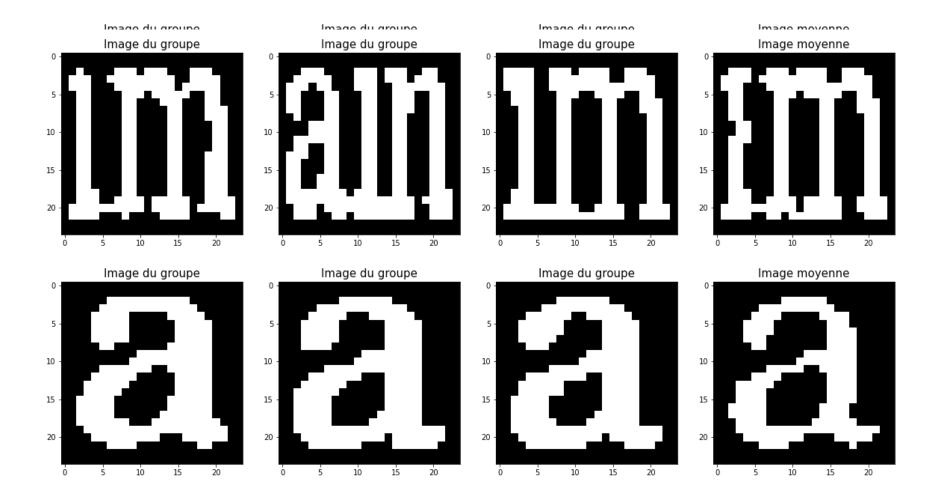


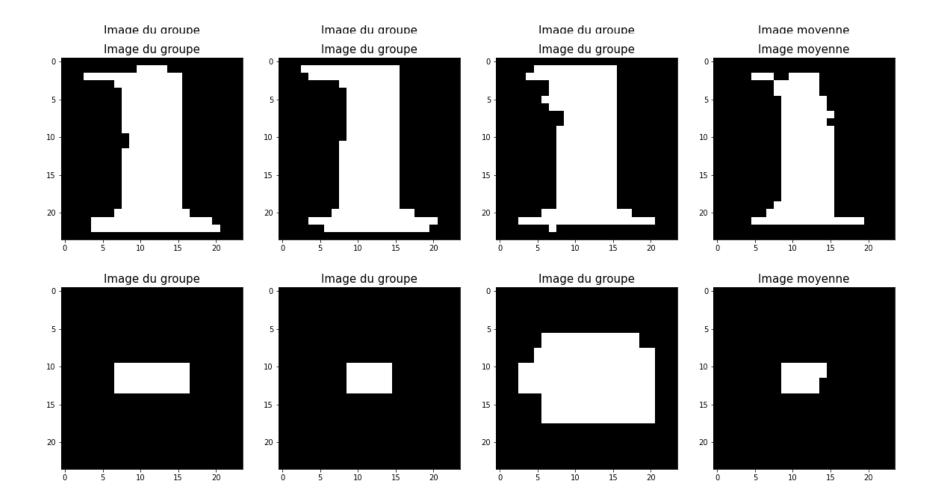


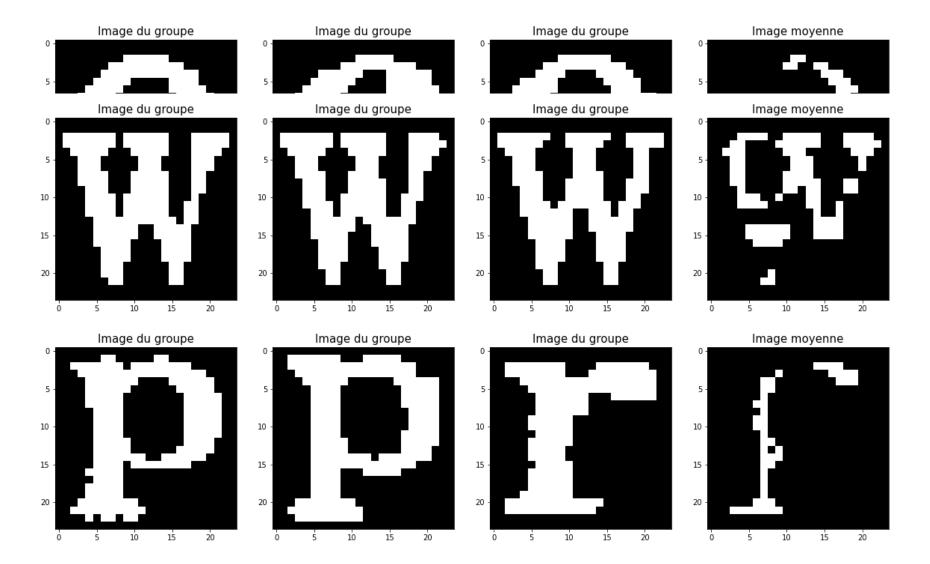


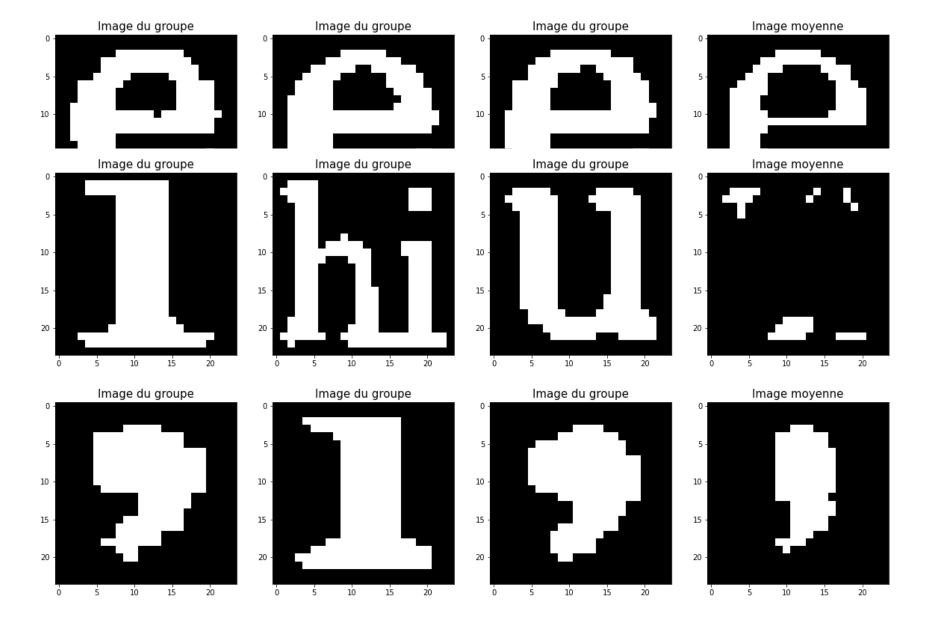


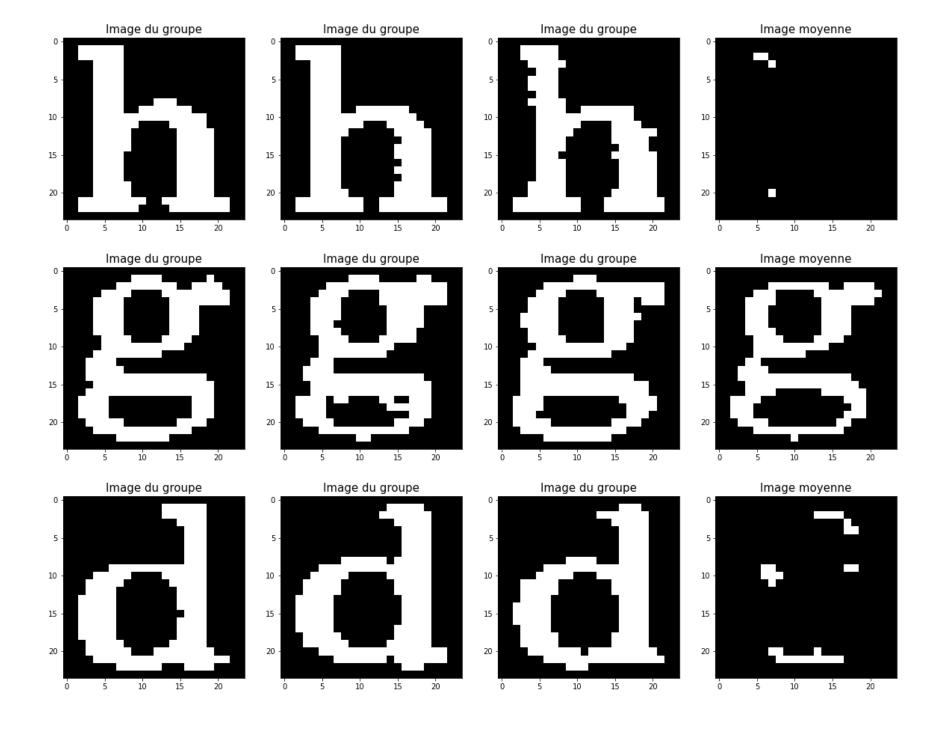


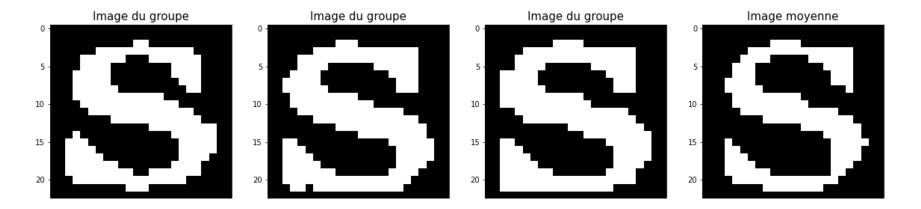












### **Discussion**

Les lettres identiques ou simillaires sont dans le même groupe. Pour ce qui est de l'image moyenne, elle représente les pixels en commun des images du groupe.

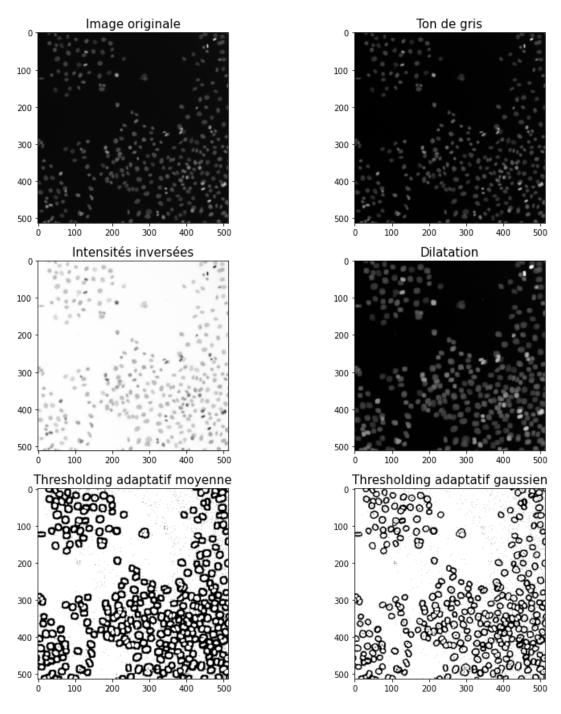
## **Exercice 2 : Compter des cellules**

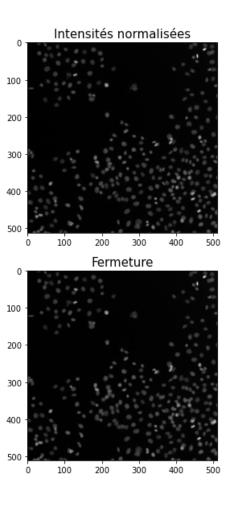
```
In [57]: # Lire 1'image
         cellules = cv2.imread('tp4 ex2.tif')
         # Convertir l'image en tons de gris
         cellules gris = cv2.cvtColor(cellules, cv2.COLOR_BGR2GRAY)
         # Normaliser les intensités
         cellules norm = cellules gris / np.max(cellules gris)
         # Inverser les intensités
         cellules inv = np.abs(cellules norm-1)
         # Oppératios morphologiques
         kernel = np.ones((5, 5), np.uint8)
         dilatation = cv2.dilate(cellules gris, kernel, iterations = 1)
         fermeture = cv2.morphologyEx(cellules_gris, cv2.MORPH CLOSE, kernel)
         # Thresholding adaptatif sur la moyenne et le filtre gaussien
         th moy = cv2.adaptiveThreshold(cellules_gris, 255, cv2.ADAPTIVE_THRESH_MEAN_C, cv2.THRESH_BINARY, 11, 2)
         th gauss = cv2.adaptiveThreshold(cellules gris, 255, cv2.ADAPTIVE THRESH GAUSSIAN C, cv2.THRESH BINARY,
         plt.figure(figsize=(20,20))
         plt.subplot(431)
         plt.title('Image originale', fontsize=15)
         plt.imshow(cellules, cmap='gray')
         plt.subplot(432)
         plt.title('Ton de gris', fontsize=15)
         plt.imshow(cellules gris, cmap='gray')
         plt.subplot(433)
         plt.title('Intensités normalisées', fontsize=15)
         plt.imshow(cellules norm, cmap='gray')
         plt.subplot(434)
         plt.title('Intensités inversées', fontsize=15)
         plt.imshow(cellules inv, cmap='gray')
         plt.subplot(435)
         plt.title('Dilatation', fontsize=15)
         plt.imshow(dilatation, cmap='gray')
         plt.subplot(436)
         plt.title('Fermeture', fontsize=15)
         plt.imshow(fermeture, cmap='gray')
         plt.subplot(437)
         plt.title('Thresholding adaptatif movenne', fontsize=15)
```

```
plt.imshow(th_moy, cmap='gray')
plt.subplot(438)
plt.title('Thresholding adaptatif gaussien', fontsize=15)
plt.imshow(th_gauss, cmap='gray')
plt.show()

# Compter les cellules
thresholds = filters.threshold_multiotsu(cellules_gris, classes=3)
regions = np.digitize(image, bins=thresholds)
cells = image > thresholds[0]
labeled_cells = measure.label(cells)

print('Nombre total de cellules:', labeled_cells.max())
```





Nombre total de cellules: 288