# 1 Python correction

### 1.1 Pyramidal decomposition and reconstruction

### 1.1.1 Decomposition

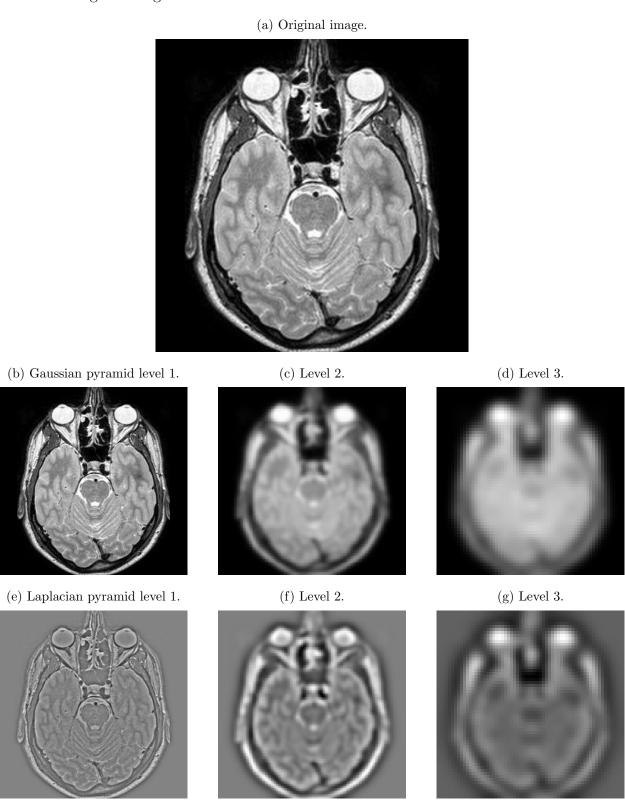
The following function makes the decomposition of the Laplacian and Gaussian pyramids at the same time. The Laplacian pyramid can be reconstructed without any additional information. This is illustrated in Fig. 1.

```
1 from skimage.transform import rescale, resize
3 def LaplacianPyramidDecomposition(Image, levels):
      Laplacian / Gaussian Pyramid
      The last image of the laplacian pyramid allows a full reconstruction
         \hookrightarrow of the original image.
      Image: original image, float32
      levels: number of levels of decomposition
9
      returns: pyrL, pyrG: Laplacian and Gaussian pyramids, respectively,
         \hookrightarrow as a list of arrays
      pyrL = [];
      pyrG = [];
      sigma = 3.;
      for 1 in range(levels):
17
          prevImage = Image.copy();
          g = ndimage.gaussian_filter(Image, sigma);
19
          print(g.dtype)
          Image = rescale(g, .5);
          primeImage= resize(Image, prevImage.shape);
          pyrL.append(prevImage - primeImage);
          pyrG.append(prevImage);
      pyrL . append ( Image ) ;
      pyrG.append(Image);
      return pyrL, pyrG;
```



rescale and resize functions have the same goal but use a scale or a shape as a parameter.

Figure 1: Gaussian and Laplacian pyramids, for 3 levels of decomposition. The Laplacian pyramid in addition to the last level of the Gaussian pyramid is required to exactly reconstruct the original image.

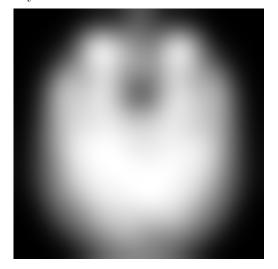


#### 1.1.2 Reconstruction

The reconstruction is straighforward and exact because of the construction of the residue. The details can be filtered (removed for example), thus giving the following result Fig. 2.

Figure 2: Reconstruction of the Laplacian pyramid.

(a) Reconstruction of the pyramid without any detail.



(b) Reconstruction of the pyramid with all the details.



## 1.2 Scale-space decomposition and multiscale filtering

The pyramid of erosions and dilations is illustrated in Fig.3.

```
def morphoMultiscale(I, levels):
    """

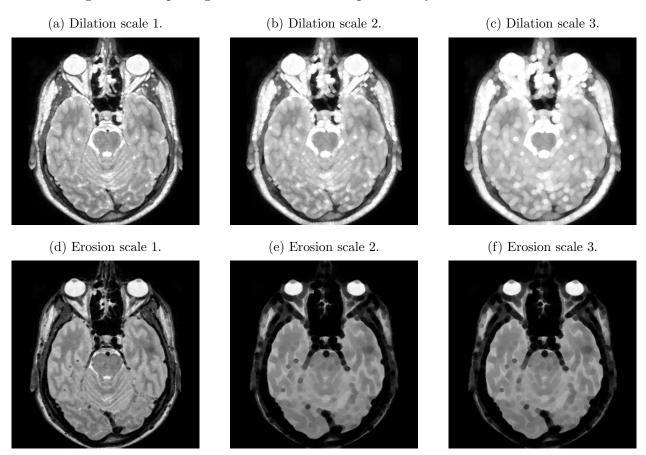
Morphological multiscale decomposition
    I: original image, float32
    levels: number of levels, int

returns: pyrD, pyrE: pyramid of Dilations/Erosions, respectively

pyrD=[];
    pyrE=[];
    for r in np.arange(1,levels):
        se = morphology.disk(r);
        pyrD.append( morphology.dilation(I, selem=se));
        pyrE.append( morphology.erosion(I, selem=se));

return pyrD, pyrE;
```

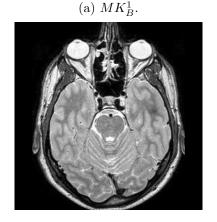
Figure 3: Morphological multiscale decomposition by dilation and erosion.



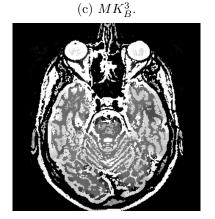
## 1.3 Kramer and Bruckner multiscale decomposition

The results are illustrated in Fig.4.

Figure 4: Kramer and Bruckner multiscale decomposition, with r = 5.







```
def kb(I, r):
    """

Elementary Kramer/Bruckner filter. Also called toggle filter.
    I: image
    r: radius of structuring element (disk), for max/min evaluation

se = morphology.disk(r);
    D=morphology.dilation(I, selem=se);
    E=morphology.erosion(I, selem=se);
    difbool = D-I < I-E;
    k = D*difbool + E * (~difbool);
    return k;</pre>
```

```
def KBmultiscale(I, levels, r=1):

"""

Kramer and Bruckner multiscale decomposition

I: original image, float32

pyrD: pyramid of Dilations
pyrE: pyramid of Erosions

returns: MKB: Kramer/Bruckner filters

"""

MKB = [];

MKB.append(I);
for i in range(levels):

MKB.append(kb(MKB[i-1], r));
return MKB
```