

# 3D Processing and Analysis with ImageJ

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## ABSTRACT

Modern microscopical techniques yields 3D and more generally multi-dimensional images, such as LSCM, and even 4D data when including time, and even 5D when including numerous channels. Important developments have been done for visualisation of such images in ImageJ such as VolumeViewer, Image5D or Image3DViewer. However few plugins focused on the 3D information present in the images, generally most processing are done in a slice by slice manner, and few plugins (including ObjectCounter3D) performs 3D measurements. This workshop will emphasise on 3D tools for visualisation, processing and measurements of volume data using ImageJ.

**Keywords:** Image Processing, Image Analysis, 3D, 4D, 5D, LSCM, Microscopy

## 1. INTRODUCTION

Last decade has seen the development of major microscopic techniques in biology such as LSCM in fluorescence microscopy and tomography for electron microscopy, not mentioning MRI or CT-Scan in medical imaging. These techniques enable a better resolution but also permits to obtain true 3D volumetric data. The volumetric data can also be combined with time lapse and multi-channel acquisition leading to what is called 4D and 5D data (see fig. 1) .

In order to visualise, process and analyse these new type of data, new softwares had to be developed. Numerous commercial softwares exist such as Amira or Volocity. They allow to perform 3D visualisation of the data, but they often provide only basic processing and analysis features. Furthermore 3D programming is not possible in an easy and portable way.

We present here basics of 3D imaging including visualisation, processing, analysis and programming within ImageJ.

## 2. 3D VISUALISATION

Visualisation of 3D data may sometimes be the most complicated task of the analysis compared to 2D data. Thanks to the development of powerful computers and 3D graphics cards, the visualisation can now be performed easily on every computer. Within ImageJ basic but powerful 3D manipulation and visualisation plugins are available such as *VolumeViewer* or *Image3DViewer*.

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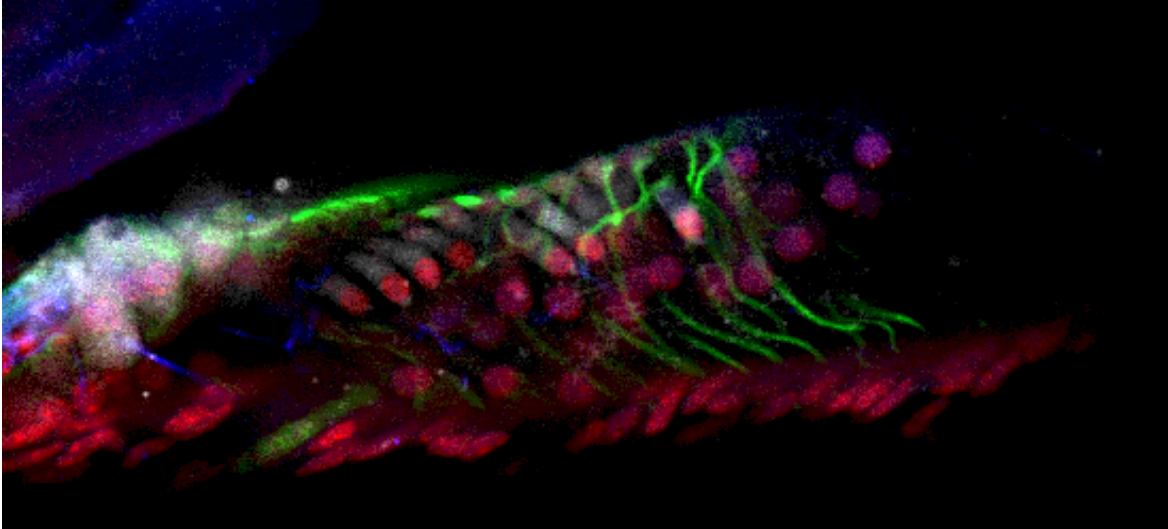


Figure 1. Slice 8 from the 4D stack Organ of Corti that can be found as a sample image in ImageJ.

### 3D manipulation and visualisation with ImageJ

ImageJ allows the manipulation of volume data via the *Stacks* submenu. It allows adding or removing slices, and animate the data. It allows also projections of the data in the *Z* direction, projection permits to visualise all the 3D data on one slice and increase signal to noise ratio. Volume image are represented as a succession of 2D slices in the *Z* direction, the data can be resliced in *X* and *Y* direction with the *Reslice* command. For reslicing it is important to take into account the spacing, *i.e* the size in *Z* of the voxel compared to its size in *X – Y*. Isotropic data are usually preferred for 3D visualisation and analysis, however generally the ratio between *Z* spacing and *X – Y* spacing is high, in order to reduce this ratio it is possible to reslice the data using the *Reslice* command by specifying the input and output spacing accordingly. A volume image with additional interpolated slices will be obtained, of course these interpolated slices do not increase the qualitative resolution in the *Z* direction.

In order to have an idea of the 3D presence of objects inside the volume the simplest way is to make series of projections of the data. This will create series of projections of the data as if one was turning around the data and view it transparent at each of these angles, thanks to a animation our brain will reconstruct the whole 3D data. This can be done using the *3D Project* command with the *mean value* and *Y axis* options.

### The VolumeViewer plugin

The *VolumeViewer* plugin allows more complex and interactive volume visualisation. Slices in any direction can be performed inside the volume, an orientation is chosen with the mouse and then one can scroll trough the volume adjusting the *distance* view. For volume visualisation a *threshold* must be chosen, the voxels above this value will be set as transparent, for other voxels the higher values the more opaque. The voxels that are visualised are set with the *depth* slider.

### The Image3DViewer plugin

The *ImageJ3DViewer* plugin requires the installation of Java3D and the presence of a 3D Graphics card. It allows the visualisation at the same time of volume data, orthogonal slices and surfaces. In order to build

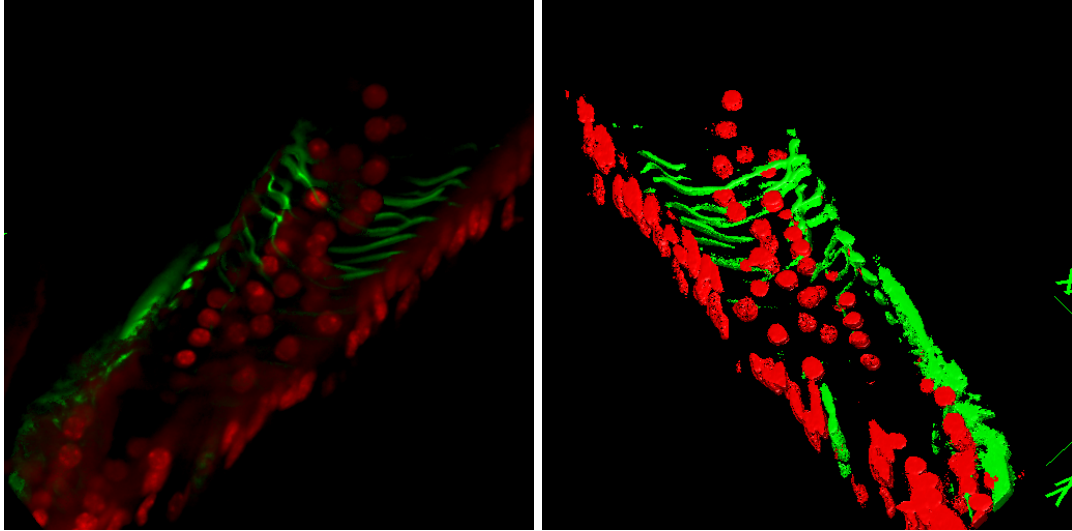


Figure 2. Two views with the *Image3DViewer* plugin from the red and green canal of 4D stack of Fig. 1. Left : volume rendering, all voxels are visualised. Right : surface rendering, only thresholded voxels are seen, note that no variation of intensities is available.

surfaces (meshes) a threshold must be chosen, a binary volume is build and voxels at the border of the binarized objects will be displayed. Transparencies between the various representations can be adjusted allowing a complete interactive view of multiple objects coming from the same or different volumes. Finally an animation of the 3D visualisation can be constructed interactively and recored as a AVI movie.

### HyperStacks manipulation : 4D and 5D

Volume data incorporating time or multiple channels are called *HyperStacks* and can be manipulated within ImageJ. The hyperStack can be projected in *Z* for all channels and times, however 4D or 5D visualisation is not (yet) possible with the *3D Project* command. For 4D visualisation, that is to say animation of several 3D dataset, the *ImageJ3DViewer* plugin can be used as well as other free softwares such as Voxx2.

## 3. 3D PROCESSING

Volmetric data may present a signal to noise ratio lower than 2D data, hence some processing may be necessary to reduce noise and enhance signal. The theory of processing is essentially the same for 2D and 3D data, it is based on the extraction of neighboring pixel values, the computation of the processed value from these values, and the application of this value to the central pixel of the neighbourhood. Basically in 2D the neighbourhood may be any symmetrical shape, usually a square or a circle. For 3D images any symmetrical shape like brick or ellipsoid is used.

The number of values to process in 3D neighbourhood are quite larger than in 2D, for a neighbourhood of radius 3, 9 values are present in a 2D square and 27 in a 3D cube. The time to process an image is then considerably increased from 2D to 3D. Implementations procedure should be optimised and JNI programming can be used. JNI programming consists in writing time-consuming procedures in other languages like C, that is quite faster than Java.

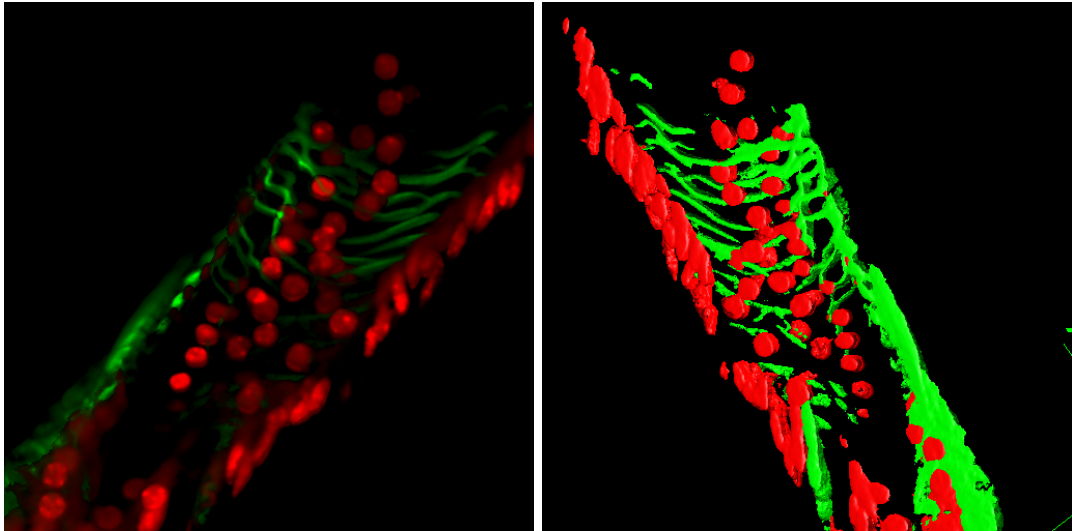


Figure 3. “ The stack of Fig. 1 was denoised by a 3D median filter of radius 1. Left : volume rendering. Right : surface rendering, note that surfaces are smoother.

In this section we describe commonly used filtering and morphological operators, but first a introduction to contrast adjustment.

### Brightness and contrast

Brightness and contrast may be adjusted slice by slice to improve the visualisation of objects of interest. However changes in contrast may be present from slice to slice due to the acquisition procedure, like decrease of fluorescence for LSCM or increase of thickness in tilted series for electron tomography. These changes in contrast can be modelled and the contrast can be corrected based on the model. The contrast can be also automatically adjusted from slice to slice thanks to the *Normalise All slice* command of the *Enhance Contrast* function. This procedure will process the data so that the the slices will have roughly the same mean and standard deviation, correcting hence any changes in contrast trough the data.

### Noise reduction

Volume data can be 2D-filtered in a slice by slice manner, this can produce fast satisfactory results but in some cases, where the signal to noise ratio is quite low, a 3D filtering should be favoured. Very weak signals are present in 3D, that is to say on adjacent slices, they may however be almost indistinguishable in 2D due to noise. Only a 3D filtering can take into account the 3D information of the signal and reduce the noise. Classical filters used for reducing noise are mean, gaussian or median filters. Median filter can be favoured since it is more robust to noise and preserve the edges of the objects, it is specially powerful on noisy confocal data images.

The radius of filtering should be chosen accordingly to the size of the objects of interest since filtering reduces noise but also may suppress small objects, the smaller the objects the lower the radius. The choice of the radius obeys the same rule in 2D as in 3D. However due to the acquisition systems, objects tend to be a bit elongated in  $Z$  compared to  $X - Y$ , furthermore the spacing is generally different in  $Z$  than in  $X - Y$  so round objects do not have the same radius in pixels in  $Z$  than in  $X - Y$ , the radius of filtering should hence be different in the two directions. The radii of the neighborhood kernel can also be optimally computed for each voxel in order to

fit inside a probable object and avoiding averaging between signals and background. Such procedures are part of anisotropic filtering,

## Morphology

Mathematical morphology is originally based on complex mathematical operations between sets, however applied to images it can be seen as a particular type of filtering. Morphological operations are to be applied to binarized images and work on the objects segmented by the binarization. However many operations can be extended to grey-level images. The basic operation of mathematical morphology is called *dilatation* and consist in superimposing a kernel to every pixel of the image, and if any of the pixel in the kernel is part of an object then the centre of the kernel turns to the objects value. Contrary to filtering the kernel does not need to be symmetrical, however usually a symmetrical kernel is used. The inverse operation of dilatation is called *erosion*, the centre of the kernel turns to the background value if any of the pixel inside the kernel is part of the background. For 3D morphology the kernel is a 3D shape, usually a ellipsoid or a 3D brick.

Actually useful morphological operations are a combination of the two basics operation, for instance *closing* consists in a *dilatation* followed by an *erosion* and permits to make the objects more compact and fills small holes inside objects. The twin operation *opening*, a *erosion* followed by a *dilatation*, permits to remove small spots in the background. Other useful operations are *fill holes* to suppress any hole present within an object, *distance map* to compute for any pixel its distance to the closest edge or *tophat* to enhance spots. This operation consists in computing a local background image by removing the objects, and then to perform the difference between the original image and the computed background. In the case of bright spots the background is computed by performing a minimum filter, equivalent to an erosion, hence removing any bright signal. The size of the kernel must therefore be larger that the object size. This procedure is quite powerful in 3D to detect fluorescence signals like F.I.S.H. spots.

## Other processing

Any procedure that processes a 2D image via a kernel can be quite directly adapted to 3D data, however the main concern is time and memory requirements since a 3D image can be hundred times bigger than a 2D image.

Another purpose of filtering is edge detection, in 2D a classical way to detect edges is to compute edges in both  $X$  and  $Y$  directions and then combine them to find the norm of the gradient corresponding to the strength of the edges, the angle of the edges can also be computed from the  $X$  and  $Y$  differentials. For 3D edges the differential in  $Z$  is added to the  $X$  and  $Y$  ones.

$$\delta x(i, j, k) \approx I(i + 1, j, k) - I(i - 1, j, k),$$

$$\delta y(i, j, k) \approx I(i, j + 1, k) - I(i, j - 1, k),$$

$$\delta z(i, j, k) \approx I(i, j, k + 1) - I(i, j, k - 1),$$

$$edge = \sqrt{\delta x^2 + \delta y^2 + \delta z^2}$$

## Fourier processing

The above cited processing were performed on the real images, other processing can be performed on the Fourier transform of the image. The Fourier transform of the image corresponds to the frequencies inside the real image, so, in another way, to the various sizes of objects present in the image. In order to reduce noise objects with very small sizes, hence high frequencies, can be deleted using a low pass filter. To enhance objects with a particular

size band pass filtering can be performed, see the *Bandpass Filter* in the *FFT* menu. The Fourier Transform can also be computed in 3D, however the computation time can be quite long if non power of 2 images are used. Other applications of Fourier Transform are signal deconvolution and correlation computations, that permits to find the best displacement in  $X$ ,  $Y$  and  $Z$  between two volumes.

For stack alignment the computation of the 3D translation between two volumes can be easily computed in Fourier Space, however for computation of 3D rotation the problem is quite more complex than in 2D since 3 angles, around each axis, must be computed. Complex optimisation algorithm should be used, iterative procedures including translation and rotations estimations are generally used, see the *StackAlignment* plugin.

## 4. 3D SEGMENTATION

Segmentation is the central procedure in image analysis. It permits to give a meaning to an image by extracting the objects, that are actually no more than a set of numeric values. The labelling of pixels to a numbered object or to background is the most difficult task of image analysis since we have to teach the computer what are objects, for instance what is a cell, what is a nucleus, what is an axon and so on.

The task is quite more complicated in 3D since it is more difficult to describe a 3D shape than a 2D one. To perform the labelling of objects, two main approaches can be used; first objects can either be drawn manually either extracted based on their voxels values. Active contours model, known as snakes, is a semi-automatic approach where the user draw a rough initial shape and the computer will try to deform this shape to best fit the surrounding contour.

### Thresholding

If the objects have homogeneous pixel values and if the values are significantly different from the background, a thresholding can be applied. In the case of an inhomogeneous illumination of the image the *Subtract Background* procedure should be used. The *Adjust Threshold* tool helps in determining the threshold, an interval of values is manually selected and the thresholded pixels, corresponding to the objects, are displayed in red colour. Automatic threshold can be computed using various algorithms like *IsoData*, *Maximum of Entropy*, *Otsu*,

Thresholding in 3D obeys the same rules, however the object may present intensity variation in the  $Z$  direction. This variation can be corrected by a normalisation procedure (see 3). An automatic calculation of the threshold can also be computed in each individual slices, but due to normal variation of objects intensities in  $Z$  this procedure is not well adapted for 3D segmentation.

After thresholding, objects should be labelled, that is to say numbered. The *Analyse Particles* procedure perform this task in 2D, it can be applied to 3D but it will label objects in each slices not taking into account the third dimension. Furthermore objects that look separated in 2D may be part of the same object in 3D, for instance a simple banana shape may present its two ends separated in a 2D section. Hence a true 3D labelling should be performed, the plugin *ObjectCounter3D* will both threshold and label the objects in 3D. It also computes geometrical measures like volume and surfaces.

### Manual segmentation

In some cases the objects can not be thresholded in a satisfying way. Morphology procedures (see 3) may improve the binarization result, but for some complex objects, as ones observed by electron microscopy, a manual segmentation is required. The user have to draw in each slices the contours of the different objects.

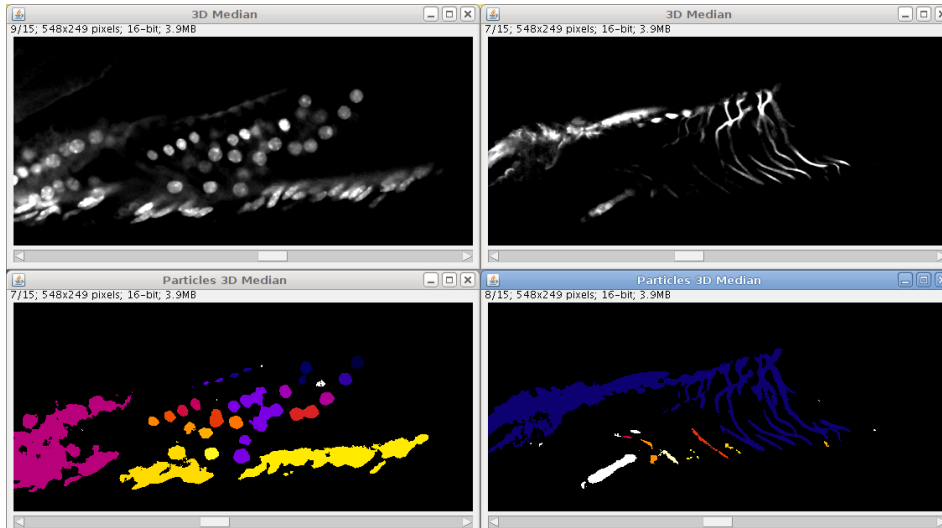


Figure 4. Output of the sample data from fig. 1 with the *ObjectCounter3D* plugin, objects are segmented in 3D and labelled with an unique index, displayed here with a look-up table. Left : red channel, note that some objects are fused. Right : green channels, note all that fibers form only one object.

Some plugins can help the user in this task such as *QuickvolIII* mainly for MRI, *TrackEM2* (see Workshop on this subject) or *SegmentationEditor*, other free tools exist like *IMOD* mainly specialised in electron microscopy data.

### Semi-automatic segmentation : active contours

Active contours models, known as *snakes*, are semi automatic techniques that can help the user automate a manual segmentation. The user draws the initial model close to the object he wants to segment, the algorithm will then try to deform this initial shape so that it fits onto the contours of the image. Many algorithms are available as plugins such as *IVUSnake*, *Snakuscule*, or *ABSsnake*.

Snakes model, though developed for 2D data, can be used in 3D in a iterative manner. The user draws an initial shape on the first slice of the volume, the algorithm will then segment the object, this segmentation on the first slice can be used as the initial shape for the next slice, and so on. However this method does not take into account the 3D information and the links that exist between adjacent slices. Furthermore in the case of noisy data or problem in acquisition an object can be missing or deformed in a particular slice. The slice by slice approach can not alleviate this problem, a real 3D procedure that take into account data in adjacent slices, as it is done in  $X - Y$  should be used. The plugin *ABSsnake3D* uses such an approach, the model is a set of points lying in the different slices. For each iteration the point will move towards the contours of the image while being constrained in  $X - Y$  by its two neighboring points and also in  $Z$  by the two closest points in the upper and lower slices.

## 5. 3D ANALYSIS

Once the objects have been segmented the analysis can be performed. This part will compute geometrical and pixel-based measures for each individual object. These measures must be performed using the real units of the image. The image should hence be calibrated in the 3 dimensions thanks to the *Image/Properties* window.

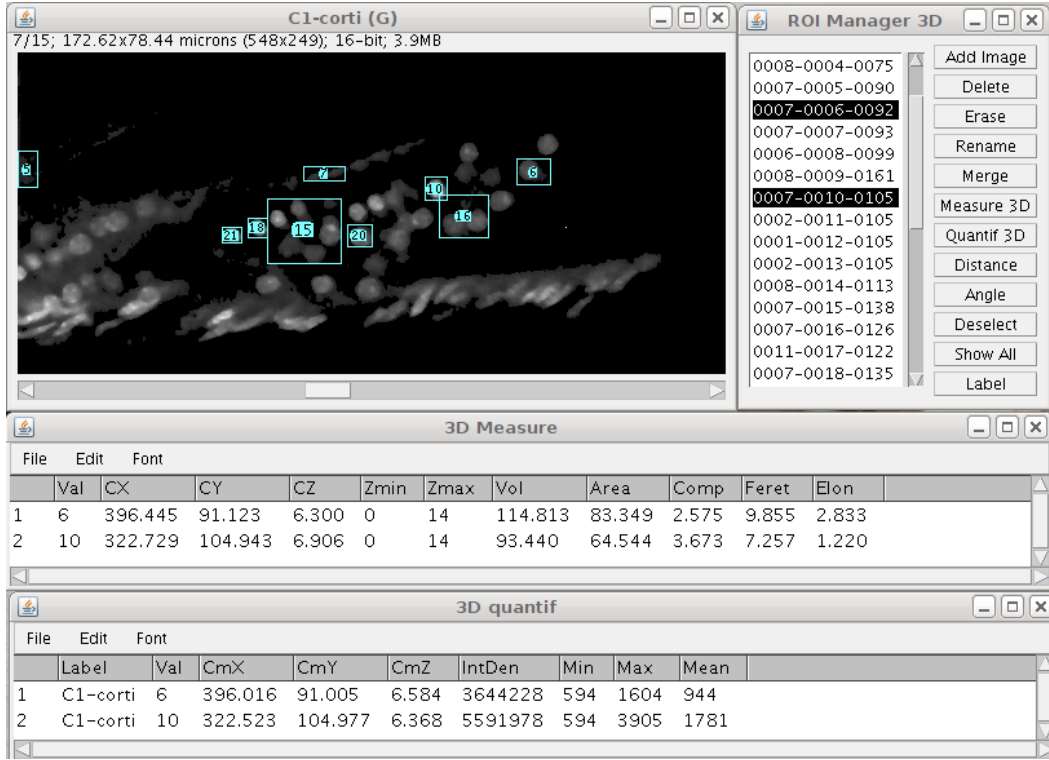


Figure 5. The *3DManager* plugin in action. Objects are added to the manager via the segmented image of fig. 4. The objects 6 and 10 were selected and measured. Geometrical measures appear in the *3D Measure* results window, the canal 1 window was selected and intensity measures from this image appear in the *3D quantif* results window.

## Geometrical features

The geometrical features concern the shape of the object without taking into account its intensity values. Basic measures are volume and area. Volume is actually the number of voxels belonging to the object and can be computed as the sum of 2D areas multiplied by the  $Z$  spacing. The area of a 3D object corresponds to the number of voxels at its border, however the border can be defined in several ways, for instance allowing diagonal voxels or not, and hence the computation of area may vary from a program to another. From this basic measures one can compute the 3D sphericity which is an extension of 2D circularity, and is computed from the ratio of volume over area. The sphericity, as well as the circularity, is maximal and equals 1 for a sphere :  $S^3 = \frac{36 \cdot \pi \cdot V^2}{A^3}$ , with  $S$  being the sphericity,  $V$  the volume and  $A$  the area.

Other measures like Feret diameter or ellipsoid fitting can not be computed from 2D slices, contrary to volume or area. Feret diameter is the longest distance from two contour points of the 3D object. In order to have an idea of the global orientation of the object the fitting a 3D ellipsoid can be used. Actually the main axes of the 3D shape can be computed as the eigen vectors of the following moments matrix :

$$\begin{pmatrix} sxx & sxy & sxz \\ syx & syy & syz \\ szx & szy & szz \end{pmatrix}$$

with :



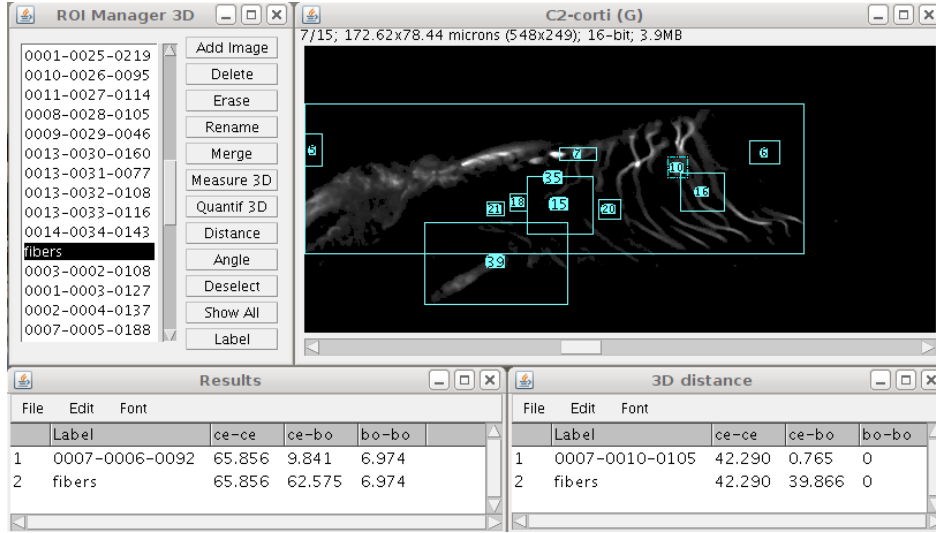


Figure 6. Distance computation with the *3DManager* plugin. The objects from red and green channels were added to the manager. The distances between objects 6 and 10 from red canal and object 1 (representing the fibers) from green canal were computed. From border to border distances one can see that object 10 touches the fibers.

$$sxx = \sum (x - cx).(x - cx),$$

$$syy = \sum (y - cy).(y - cy),$$

$$szz = \sum (z - cz).(z - cz),$$

$$sxy = syx = \sum (x - cx).(y - cy),$$

$$sxz = szx = \sum (x - cx).(z - cz),$$

$$syz = szy = \sum (y - cy).(z - cz),$$

$x$ ,  $y$  and  $z$  being the coordinates of the points belonging to the object,  $cx$ ,  $cy$ ,  $cz$  being the coordinates of the barycentre of the object. The 3 eigen vectors will correspond to the 3 main axes of the shape. An elongation factor can be computed as the ratio between the two eigen values of the two first main axes.

Other geometrical measures can give valuable information like distances between objects. The distances can be computed from centre to centre, or from border to border giving the minimal distance between objects. One may also be interested by the distance from the centre of an object to the border of another, or the distance along a particular direction. One may also be interested in the percentage of inclusion of an object into another, and so on.

Finally the shape of the envelope of the object can be studied, in 2D this envelope correspond to the perimeter of the object and can be mathematically analysed as a curve, hence information like curvature can be computed. In 3D such analysis is quite more complex since the border corresponds to a mathematical surface, and for instance for each point of a surface there are two curvatures instead of one.

## Intensity features

The above geometrical measures do not take into account the intensity of the voxels inside the object. In the case of an inhomogeneous distribution of the gray-levels inside the objects, intensity-based measure may be needed. For instance *Integrated Density*, which is the volume multiplied by the mean voxel value inside the object, and correspond to the sum of intensities inside the object. Mass centres are the barycenter of the voxels

weighted by the pixel intensities :  $MCx = \frac{\sum I_p \cdot x}{\sum I_p}$ , where  $I_p$  is the intensity at point  $p$  belonging to the object and  $x$  its x-coordinate, same applies for  $y$  and  $z$ .

Main axes can also be computed by weighting all coordinates by the voxel value. Note that in most cases, due to the segmentation process, the object present homogeneous intensities and hence intensity-based barycenter and axes does not differ much from those computed in the above paragraph where actually all voxels are weighted by a factor of 1.

## Morphological analysis

A first application of mathematical morphology is the computation of 3D local thickness based on the distance map. For each point inside the object its minimum distance to the border can be used as an estimate to the maximal sphere that can be fitted at this point inside the object.

Mathematical morphology can be used to fill holes inside objects or make them more compact, it can also be used to have a estimate of the sizes of the objects present in the volume without to segment them. The technique, called *granulometry*, is based on the idea that objects processed by an erosion greater than the radius of the object will disappear. For grey-levels volumes, an erosion is actually a minimum filter, hence if bright objects are present in the image, and if the radius of the minimum filter is greater than the radius of the object, the minimum value inside the neighbourhood will be in the background. This will occur for any pixel inside the object and hence the object will tend to disappear. Therefore computing series of *opening*, an erosion followed by a dilatation, will produce series of filtered images, and the image that will differ much from the original will correspond to the disappearing of many objects, the corresponding radius of opening will correspond to the radius of some objects present in the volume. If many objects are present with different radii, for each particular radius the image will change. The same can be done with series of closing, dilatation followed by an erosion, in this case, radius of background will be detected, that is to say distances between objects.

## 6. 3D PROGRAMMING

Many, if not all, operations implemented in ImageJ can be performed on 3D data on a slice by slice manner. Hence macro recording should be sufficient to record series of processing performed on a stack by either ImageJ or plugins. For 3D analysis access to all the slices is often required. This can be done in macros using the slices commands, and the `ImageStack` class in plugins.

### Macros

In order to manipulate a stack, the different slices corresponding to different  $z$  should be accessed. This can be done using the macro command `setSlice(n)` where  $n$  is the slice number between 1 and the total number of slices. The current displayed slice can be retrieved using the command `getSliceNumber()`, `nSlices` is a built-in variable corresponding to the total number of slices. In order to scroll trough the slices a loop is required, a loop is constructed by the `for` command : `for(index=start;index<=end;index=index+step)` where `start` and `end` are the starting and ending value for the variable `index`, `step` is the increment of `index` between each iteration, usually set to 1. Here are an example of a macro that finds the slice where the 3D object has the largest area :

```
start=getSliceNumber();
end=nSlices;
```

```

areamax=0;
slicemax=start;
run("Set Measurements...", "area redirect=None decimal=3");
for(slice=start;slice<=end;slice=slice+1){
    setSlice(slice);
    run("Analyze Particles...", "size=0-Infinity circularity=0.00-1.00
        show=Nothing display exclude record slice");
    area=getResult("Area");
    if(area>areamax){
        areamax=area;
        slicemax=slice;
    }
}
write("Max area at slice "+slicemax);
getVoxelSize(sizeX,sizeY,sizeZ,unit);
write("Max area at z "+slicemax*sizeZ+" "+unit);

```

For 4D and 5D data, the macro functions `Stack.` should be used. The various dimensions, that is to say the width, height, number of channels, number of slices and number of times, can be retrieved using the function `Stack.getDimensions(width,height,channels,slices,frames)`. The equivalent of `setSlice(n)` is simply `Stack.setSlice(n)` for HyperStacks, similarly the functions `Stack.setChannel(n)` and `Stack setFrame(n)` will change the channel and timepoint. The different information can be set using one function `Stack.setPosition(channel,slice,frame)` and retrieved by the function `Stack.getPosition(channel,slice,frame)`. The same macro is rewritten supposing a 4D data :

```

Stack.getPosition(channel,slice,frame);
startTime=frame;
startZ=slice;
Stack.getDimensions(w,h,channels,slices,frames);
endTime=frames;
endZ=slices;
run("Set Measurements...", "area redirect=None decimal=3");
getVoxelSize(sizeX,sizeY,sizeZ,unit);
for(time=startTime;time<=endTime;time=time+1){
    Stack.setFrame(time);
    areamax=0;
    slicemax=startZ;
    for(slice=startZ;slice<=endZ;slice=slice+1){
        Stack.setSlice(slice);
        run("Analyze Particles...", "size=0-Infinity circularity=0.00-1.00
            show=Nothing display exclude record slice");
        area=getResult("Area");
        if(area>areamax){

```

```

        areamax=area;
        slicemax=slice;
    }
}
write("For time "+time+" Max area at slice "+slicemax);
write("For time "+time+" Max area at z "+slicemax*sizez+" "+unit);
}

```

## Plugins

Plugins are more adapted than macros to implement new processing procedures since the pixels access is quite faster. The Java class that stores stacks is called `ImageStack`, the stack can be retrieved from the current `ImagePlus` window :

```

// get current image and get its stack
ImagePlus plus=IJ.getImage();
ImageStack stack=plus.getStack();

```

The dimensions of the stack can be obtained using the methods `getWidth()`, `getHeight()` and `getSize()`:

```

int w=stack.getWidth();
int h=stack.getHeight();
int d=stack.getSize();
IJ.write("dimensions : "+w+" "+h+" "+d);

```

In order to access the pixels values one has to access first the slice of interest and then the pixel inside this slice :

```

int x,y,z;
x=w/2; y=h/2; z=d/2;
ImageProcessor slice=stack.getProcessor(z+1);
int val=slice.getPixel(x,y);
IJ.write("The central pixel has value : "+val);

```

Note that the slice number in the `getProcessor` method starts at 1 and not 0. Following is an example of a 3D mean filtering :

```

int p0,p1,p2,p3,p4,p5,p6,mean;
ImageStack filtered=new ImageStack(w,h);
ImageProcessor filteredSlice;
ImageProcessor slice, sliceUp, sliceDown;
for(int k=2;k<=d-1;k++){
    slice=stack.getProcessor(k);
    sliceUp=stack.getProcessor(k+1);
    sliceDown=stack.getProcessor(k-1);

```

```

        filteredSlice=new ByteProcessor(w,h);
        for(int i=1;i<w-1;i++){
        for(int j=1;j<h-1;j++){
            p0=slice.getPixel(i,j);
            p1=slice.getPixel(i+1,j);
            p2=slice.getPixel(i-1,j);
            p3=slice.getPixel(i,j+1);
            p4=slice.getPixel(i,j-1);
            p5=sliceDown.getPixel(i,j);
            p6=sliceUp.getPixel(i,j);
            mean=(p0+p1+p2+p3+p4+p5+p6)/6;
            filteredSlice.putPixel(i,j,mean);
        } }
        filtered.addSlice("",filteredSlice);
    }
    ImagePlus filteredPlus = new ImagePlus("Filtered", filtered);
    filteredPlus.show();

```

Note that the result slices `filteredSlice` are built one by one and added to the result stack `filtered`.

## Using libraries

In order to manipulate more easily 3D volume data, specifically designed classes for 3D data can be used such as *ij3d* for processing and analysis, or classes coming from *Image3DViewer* or *VolumeViewer* for visualisation. With the *ij3d* package an 3D image is created from a stack, and pixels can be accessed directly :

```

// build a Image3D for 8-bits or 16-bits data
IntImage3D image3d = new IntImage3D(imp.getStack());
// get the sizes
int w = image3d.getSizex();
int h = image3d.getSizey();
int d = image3d.getSizez();
// get the value of the central pixel
int pix = image3d.getPixel(w/2,h/2,d/2);

```

Classical filters are implemented like median, mean, edges, morphological filters, and so on. The result can be visualised retrieving the underlying `ImageStack` :

```

// define the radii or filtering
IntImage3D filtered = image3d.meanFilter(2, 2, 1);
// display the filtered stack
new ImagePlus("3D mean", filtered.getStack()).show();

```

Objects can also be segmented in 3D and analysed using the class `Object3D`, basic 3D calculation can be performed using the class `Vector3D`, basic shapes can be draw in 3D using the class `ObjectCreator3D`, ....

```

// stack is a stack containing segmented objects, each one having a particular grey-level
// for instance coming from ObjectCounter3D
IntImage3D seg = new IntImage3D(stack);
// create the object 1, which corresponds to segmented voxels having values 1
obj = new Object3D(seg, 1);
// compute barycenter
obj.computeCenter();
// compute borders
obj.computeContours();
// get barycenter
float bx = obj.getCenterX();
float by = obj.getCenterY();
float bz = obj.getCenterZ();
// get the feret's diameter
double feret = obj.getFeret();
// construct an sphere with pixel value 255 centered on barycenter and with radius=feret
ObjectCreator3D sphere = new ObjectCreator3D(w, h, d);
obj.createEllipsoide(bx, by, bz, feret, feret, feret, 255);
new ImagePlus("Sphere", sphere.getStack()).show();

```

## APPENDIX A. LIST OF PLUGINS

- VolumeViewer : <http://rsbweb.nih.gov/ij/plugins/volume-viewer.html>
- ImageJ3DViewer and 3D Segmentation Editor : <http://www.neurofly.de/>
- Image5D : <http://rsbweb.nih.gov/ij/plugins/image5d.html>
- IJ3D processing and analysis : <http://imagejdocu.tudor.lu/doku.php>
- StackAlignment : <http://www.med.harvard.edu/JPNM/ij/plugins/Align3TP.html>
- ObjectCounter3D : <http://rsbweb.nih.gov/ij/plugins/track/objects.html>
- ABSnake : [http://imagejdocu.tudor.lu/doku.php?id=plugin:segmentation:active\\_contour:start](http://imagejdocu.tudor.lu/doku.php?id=plugin:segmentation:active_contour:start)
- LiveWire and IVUSnake : <http://ivussnakes.sourceforge.net/>
- Snakuscule : <http://bigwww.epfl.ch/thevenaz/snakuscule/>
- TrackEM2 : <http://www.ini.uzh.ch/~acardona/trakem2.html>
- QuickVol II : <http://www.quickvol.com/>

## APPENDIX B. OTHER PROGRAMS

- IMOD 3D segmentation and visualisation : <http://bio3d.colorado.edu/imod/>
- Vox2 3D and 4D visualisation : <http://www.nephrology.iupui.edu/imaging/voxx/>
- 3D molecular visualisation : <http://www.cgl.ucsf.edu/chimera/>
- Etdips 3D medical analysis and visualisation : <http://www.cc.nih.gov/cip/software/etdips/>

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