

# Problem class, ground truth annotations and reported metrics

To perform benchmarking, ground truth annotations should be encoded in a format that is specific to the associated problem class. BIA workflows are also expected to output results in the same format. Currently 9 problem classes are supported in BIAFLOWS and their respective annotation formats and computed benchmark metrics are described below.

**Note:** each problem class has a long name (explicit) and short name, for instance Object Segmentation (ObjSeg). The same hold for metrics, for instance DICE (DC).

See section **Benchmarking Metrics** for metrics description.

## Problem class: Object Segmentation (ObjSeg)

Task: Delineate objects or isolated regions

Object Encoding: 2D/3D label masks with foreground > 0 (one unique ID per object), background = 0

Reported metrics: DICE (DC), AVERAGE\_HAUSDORFF\_DISTANCE (AHD), computed by [VISCERAL](#) executable (archived [here](#)), Fraction overlap (FOVL) computed by [custom Python code](#), Mean Average Precision computed by [Data Science Bowl 2018](#) Python code.

## Problem class: Spot / object counting (SptCnt)

Task: Estimate the number of objects

Object Encoding: 2D/3D binary masks, exactly 1 spot/object per non null pixel

Reported metrics: RELATIVE\_ERROR\_COUNT (REC), computed by [custom Python code](#).

## Problem class: Spot / object detection (ObjDet)

Task: Detect objects in an image (e.g. nucleus)

Object Encoding: 2D/3D binary masks, exactly 1 object per non null pixel

Reported metrics: CONFUSION\_MATRIX (TP, FN, FP), F1\_SCORE (F1), PRECISION (PR), RECALL (RE), Distance RMSE (RMSE), computed by [Particle Tracking Challenge](#) metric Java code (particle matching only, [archived here](#) in bin / DetectionPerformance.jar)

### **Problem class: Pixel/Voxel Classification (PixCla)**

Task: Estimate pixels class

Object Encoding: 2D/3D class masks, gray level encodes pixel/voxel class, background = 0

Reported metrics: F1\_SCORE (F1), ACCURACY (ACC), PRECISION (PR), RECALL (RE), computed by [custom Python code](#)

### **Problem class: Filament Tree Tracing (TreTrc)**

Task: Estimate the medial axis of a connected filament tree network (one per image)

Object Encoding: [SWC file](#)

Reported metrics:

UNMATCHED\_VOXEL\_RATE (UVR), computed by [custom Python code](#)

NetMets metrics: Geometric False Negative rate (FNR), Geometric False Positive rate (FPR) computed by [NetMets Python code](#)

Metrics parameters:

GATING\_DIST (UVR): Maximum distance between skeleton voxels in reference and prediction skeletons to be considered as matched (default = 5 pix)

Sigma (NetMets): tolerance in centreline position (default: 5 pix)

### **Problem class: Filament Networks Tracing (LooTrc)**

Task: Estimate the medial axis of one or several connected filament network(s)

Object Encoding: 2D/3D skeleton binary masks with skeleton pixels > 0, background = 0

Reported metrics:

UNMATCHED\_VOXEL\_RATE (UVR), computed by [custom Python code](#).

NetMets metrics: Geometric False Negative rate (FNR), Geometric False Positive rate (FPR) computed by [NetMets Python code](#)

Metrics parameters:

GATING\_DIST (UVR): Maximum distance between skeleton voxels in reference and prediction skeletons to be considered as matched (default = 5 pix)

Sigma (NetMets): tolerance in centreline position (default: 5 pix)

Skeleton sampling distance (NetMets): skeletons are sampled to be converted to SWC models. (default: 3 voxels, default Z Ratio: 1)

### **Problem class: Landmark Detection (LndDet)**

Task: Estimate the position of specific feature points

Object Encoding: 2D/3D class masks, exactly 1 landmark per non null pixel, gray level encodes landmark class (1 to N, N is the number of landmarks)

Reported metrics: Number of reference / predicted landmarks (NREF, NPRED), Mean distance from predicted landmarks to closest reference landmarks with same class (MRE). All metrics computed by [custom Python code](#)

### **Problem class: Particle Tracking (PrtTrk)**

Task: Estimate the tracks followed by particles (no division)

Object Encoding: 2D/3D label masks, exactly 1 particle per non null pixel, gray level encodes particle track ID

Reported metrics: Normalized pairing score alpha (NPSA), Full normalized pairing score beta (FNPSB), Number of reference tracks (NRT), Number of candidate tracks (NCT), Jaccard Similarity Tracks (JST), Number of paired tracks (NPT), Number of missed tracks (NMT), Number of spurious tracks (NST), Number of reference detections (NRD), Number of candidate detections (NCD), Jaccard similarity detections (JSD), Number of paired detections (NPD), number of missed detections (NMD), Number of spurious detections (NSD)

All metrics computed by [Particle Tracking Challenge](#) Java code ([archived here](#))

Metrics parameters: GATING\_DIST (default = 5, maximum distance between particle detections in reference / prediction tracks to be considered as matched)

### **Problem class: Object Tracking (ObjTrk)**

Task: Estimate object tracks and segmentation masks (with possible divisions)

Encoding: 2D/3D TIFF label masks, gray level encodes object ID + division text file (see [Cell Tracking Challenge](#) format)

Reported metrics: Segmentation measure (SEG), Tracking measure (TRA). All computed from [Cell Tracking Challenge](#) metric command-line executables (archived [here](#) and [here](#))