ASAM OpenLABEL V1.0.0 Webinar Labeling Geometries and Methods

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Association for Standardization of Automation and Measuring Systems

Agenda

OpenLABEL V1.0.0 Release Webinar

1	Process towards a joint consensus
2	Incorporating best practices from an engineering perspective
3	Annotation types covered in the initial release V1.0.0



Process towards a joint consensus



Finding a joint consensus, driven by the industry demand

Process





Types of annotation covered

Raw data sources: Images, videos, pointclouds...

2D Bounding Box





3D Bounding Box (cuboids)

Semantic Segmentation (2D)





Point cloud segmentation (3D)



Incorporating best practices from an engineering perspective



Coordinate system, spatial orientations and rotations

Including best practices to reduce common errors and ensuring compatibility







Annotation types covered in the initial release V1.0.0



2D Bounding Box

2D

Bounding boxes are geometric entities which enclose the shape of an object in Cartesian coordinates.

Bounding boxes define minimum and maximum limits at each dimension so the entire object lies within the specified limits.



A 2D bounding box is defined as a 4-dimensional vector [x, y, w, h], where [x, y] is the center of the bounding box and [w, h] represent the width (horizontal, x-coordinate dimension) and height (vertical, y-coordinate dimension), respectively.



Rotated 2D Bounding Box

2D

2D rotated bounding box: enclosing an entire object defined by its center position (in pixels), its width and height, and the rotation angle



A 2D rotated bounding box is defined as a 5-dimensional vector [x, y, w, h, alpha], where [x, y] is the center of the bounding box and [w, h] represent the width (horizontal, x-coordinate dimension) and height (vertical, y-coordinate dimension), respectively.



3D Bounding Box (Cuboid)

3D

A 3D bounding box is a cuboid in 3D Euclidean space.

It is defined by position (of its center), rotation, and size.

Position and size are defined as 3-vectors, while rotation can be expressed in two alternative forms,

- using 4-vector quaternion notation or
- 3-vector Euler notation.





A standardized taxonomy for semantic segmentation

Partial scene segmentation

There are some pixels that have no classes associated with them

Full scene segmentation

All pixels have a class associated. In this case *D* coincides with *P*. Note that in the use case, despite the class unlabeled or other indicating all pixels outside of the real classes of interest, there is still a form of full scene segmentation performed.

Single-class per pixel segmentation

This is the case when each labeled pixel is associated with exactly one class.

Multi-class per pixel segmentation

This is the case when at least one labeled pixel is associated with more than one class.



Describing the 2D image with more granular annotations





Segmenting a subset of the image -> Partial scene segmentation

Partial scene







Segmenting a subset of the image -> Partial scene segmentation + instance aware



Partial scene





Segmenting the whole image -> Full scene segmentation



Partial scene

Full scene





Segmenting the whole image -> Full scene segmentation + instance aware





Full scene



Point cloud segmentation

Representing the world from different sensor modalities





Point cloud segmentation

ASAM OpenLABEL V1.0.0 supports 3D point cloud segmentations.





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A standardized taxonomy for semantic segmentation

Formal definition

• Partial scene segmentation

$$\exists p_x \in P : (p_x, c_y) \notin R_{seg}$$

Full scene segmentation

 $orall p_x \in P, \exists c_y \in C: (p_x, c_y) \in R_{seg}$

- Single-class per pixel segmentation
- $orall p_x \in D, \exists ! c_y \in C: (p_x, c_y) \in R_{seg}$

Multi-class per pixel segmentation

$$\exists p_x \in D, \exists c_1, c_2 \ldots c_k \in C: (p_x, c_1), (p_x, c_2), \ldots (p_x, c_k) \in R_{seg}.$$

Pixels: $P = p_1, p_2, \dots p_p$ Classes: $C = c_1, c_2, \dots c_c$ Seg.: $R_{seg} \subset P \times C$, where $P \times C = (p_1, c_1), (p_1, c_2), \dots (p_n, c_m)$