

# SimCol Evaluation Details

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August 24, 2022

## Abstract

This document aims to give a more detailed description of the submission format for the SimCol-to-3D 2022 challenge. The document describes the expected submission *zip*-folder and the evaluation metrics used for the leaderboard. Further information can be found on the challenge website <https://www.synapse.org/#!Synapse:syn28548633/wiki/617126> and the challenge GitHub project <https://github.com/anitarau/simcol>.

## 1 Submission Format

The challenge consists of three sub-tasks. Participants can choose to compete in one, two, or all three tasks.

### 1.1 Submission .zip File

Independently of the number of tasks that a group submits to, only one .zip file should be submitted. If all three tasks are tackled, the structure of this zip files should be as follows:

```
SimCol_submission
├── SyntheticColon_I_Test
│   ├── Frames_S5.OP
│   │   ├── pose
│   │   │   ├── FrameBuffer_0000_to_FrameBuffer_0001.txt
│   │   │   ├── ...
│   │   └── depth
│   │       ├── FrameBuffer_0000.npy
│   │       ├── ...
│   ├── Frames_S10.OP
│   │   ├── pose
│   │   │   ├── FrameBuffer_0000_to_FrameBuffer_0001.txt
│   │   │   ├── ...
│   │   └── depth
│   │       ├── FrameBuffer_0000.npy
│   │       ├── ...
│   ├── Frames_S15.OP
│   │   ├── pose
│   │   │   ├── FrameBuffer_0000_to_FrameBuffer_0001.txt
│   │   │   ├── ...
│   │   └── depth
│   │       ├── FrameBuffer_0000.npy
│   │       ├── ...
│   └── ...
├── SyntheticColon_II_Test
│   ├── Frames_B5.OP
│   │   ├── pose
│   │   │   ├── FrameBuffer_0000_to_FrameBuffer_0001.txt
```

```

├── ...
├── depth
│   ├── FrameBuffer_0000.npy
│   └── ...
├── Frames_B10.OP
│   ├── pose
│   │   ├── FrameBuffer_0000_to_FrameBuffer_0001.txt
│   │   └── ...
│   ├── depth
│   │   ├── FrameBuffer_0000.npy
│   │   └── ...
│   └── Frames_B15.OP
│       ├── pose
│       │   ├── FrameBuffer_0000_to_FrameBuffer_0001.txt
│       │   └── ...
│       ├── depth
│       │   ├── FrameBuffer_0000.npy
│       │   └── ...
├── SyntheticColon_III_Test
│   ├── Frames_01.OP
│   │   ├── pose
│   │   │   ├── FrameBuffer_0000_to_FrameBuffer_0001.txt
│   │   │   └── ...
│   │   ├── depth
│   │   │   ├── FrameBuffer_0000.npy
│   │   │   └── ...
│   │   └── Frames_02.OP
│   │       ├── pose
│   │       │   ├── FrameBuffer_0000_to_FrameBuffer_0001.txt
│   │       │   └── ...
│   │       ├── depth
│   │       │   ├── FrameBuffer_0000.npy
│   │       │   └── ...
│   │       └── Frames_03.OP
│   │           ├── pose
│   │           │   ├── FrameBuffer_0000_to_FrameBuffer_0001.txt
│   │           │   └── ...
│   │           ├── depth
│   │           │   ├── FrameBuffer_0000.npy
│   │           │   └── ...
│   └── RealColon_082
│       ├── ims_082_6.OP
│       │   ├── pose
│       │   │   ├── out...to_out...txt
│       │   │   └── ...
│       │   └── ims_082_12.OP
│       │       ├── pose
│       │       │   ├── out...to_out...txt
│       │       │   └── ...
│       │       └── ims_082_25.OP
│       │           ├── pose
│       │           │   ├── out...to_out...txt
│       │           │   └── ...
│       └── RealColon_084
│           ├── ims_084_25.OP
│           │   ├── pose

```



participants familiarize themselves with the COLMAP labels provided for Sequences 001 and 002 and use them for validation of their methods.

## 2 Evaluation Metrics

Please refer to the evaluation scripts provided on the project GitHub.

### 2.1 Subtask 1

Let  $Y$  be one ground truth depth map, and let  $Y'$  be its prediction. Further, let  $D$  be the number of pixels  $d$  in the depth map and let  $\mu()$  denote the median of all  $d$ . Then we use the following three evaluation metrics:

$$L_1 = \frac{1}{D} \sum_d \|Y(d) - Y'(d)\|_1 \quad (1)$$

$$L_{\text{rel}} = \mu_d(\| \frac{Y(d) - Y'(d)}{Y(d)} \|_1) \quad (2)$$

$$L_{\text{RMSE}} = \sqrt{\frac{1}{D} \sum_d (Y(d) - Y'(d))^2} \quad (3)$$

At test time the mean over all  $N$  depth maps in a sub dataset will be reported. Note that the predicted depth maps will be scaled before evaluation. For each subsequence we compute the scale factor  $s$ . Let  $\bar{d}$  denote the average depth value of a depth map  $Y$ .

$$s = \frac{\sum_n \bar{d} \cdot \bar{d}'}{\sum_n \bar{d}' \cdot \bar{d}'} \quad (4)$$

### 2.2 Subtasks 2 and 3

There are three metrics to evaluate the accuracy of the predicted poses: the Absolute Translation Error (ATE), the Relative Translation Error (RTE), and the ROTation error (ROT). Let  $\Omega$  denote relative poses, and let  $P$  denote absolute poses. As before,  $'$  denotes the predictions. Then the losses are defined as

$$RTE = \mu_\tau(\|trans(\Omega_\tau^{-1}\Omega'_\tau)\|), \quad (5)$$

$$ATE = \mu_\tau(\|trans(P_\tau) - trans(P'_\tau)\|), \quad \text{and} \quad (6)$$

$$ROT = \mu_\tau(\frac{trace(Rot(\Omega_\tau^{-1}\Omega'_\tau)) - 1}{2} \cdot \frac{180}{\pi}), \quad (7)$$

where  $\mu_\tau$  denotes the median over all steps  $\tau$ ,  $trans$  and  $Rot$  denote the translation and rotation components of a projection matrix, and  $\|\cdot\|$  denotes the 2-norm.

As for the depth task, we scale the predicted poses before evaluation. We compute the scale  $s$  as:

$$s = \frac{\sum_\tau trans(\Omega_\tau)^T \cdot trans(\Omega'_\tau)}{\sum_\tau trans(\Omega'_\tau)^T \cdot trans(\Omega'_\tau)} \quad (8)$$