Manual to Berlin-APC: A Privacy-Friendly Dataset for Automated Passenger Counting in Public Transport

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I. INTRODUCTION

This document provides a short technical introduction to the *Berlin-APC* dataset (version 1). The dataset consists of two files, a HDF5 file which contains the image sequences, and a CSV file which contains the labels. The CSV file has three columns: (1) the image sequence name; (2) the number of boarding passengers in that image sequence; (3) the number of alighting passengers in that image sequence. The image sequence names also serve as keys in the HDF5 file. The HDF5 file's datasets (indexed by the aforementioned sequence names) are float16 arrays of the shape (number of frames, 20, 25), the pixel values range between 0–1, and the frame rate is 10 frames per second.

1 \$ head berlin-apc.csv 00000_000_01_06,2,0 3 00001_000_01_07,0,2 000002_000_01_09,1,0 5 00003_000_01_09,1,0 000004_000_01_11,2,0 7 000005_000_01_11,5,1 000006_000_01_12,0,7

The image sequence name is structured as follows:

<serial number>_<recording day id>_<recording month>_<recording hour>

For example, the entry 012958_030_09_15, 0, 3 indicates the 12,958th image sequence of the dataset (counting starts at zero),¹ which was recorded on a day in September 2017, between 3.00 and 4.00 pm. That day is the 30th recording day, i.e., all image sequence with a recording day id of 30 were recorded on the same day. According to manual counting, this image sequence shows three alighting passengers, and no boarding passengers.

II. WORKING WITH THE HDF5 AND CSV FILE

We now give an example how to work with the dataset in Python:

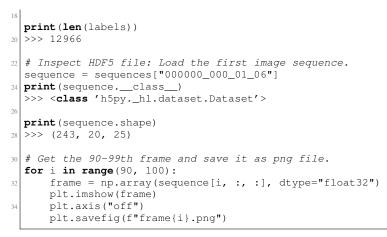
```
import pandas as pd
 import numpy as np
  import h5py as h5
 import matplotlib.pyplot as plt
  # Load HDF5 and CSV file.
6
  sequences = h5.File("berlin-apc.h5", mode="r")
 labels = pd.read_csv("berlin-apc.csv", names=["sequence_name", "n_boarding", "n_alighting"])
  # Inspect CSV file: Print first 5 labels.
 print(labels.head())
 >>>
            sequence_name n_boarding n_alighting
         000000_000_01_06
  >>> 0
                                    2
                                                  0
 >>> 1
        000001_000_01_07
                                    0
                                                  2
 >>> 2
        000002_000_01_09
                                    1
                                                  0
                                                  0
 >>> 3
        000003_000_01_09
                                    1
16
 >>> 4
        000004_000_01_11
                                    2
                                                  0
```

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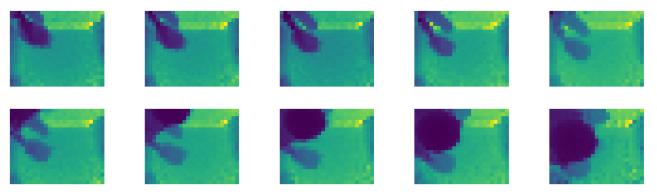
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¹Note that the serial numbers are not consistent between different versions of this dataset.



The frames 90-99 from the previous example look like this:



III. ANNOTATION DETAILS AND SOFTWARE

The video annotation software *VisualCount* (developed by INTERAUTOMATION Deutschland GmbH, Berlin, Germany; see screenshot below) was used to perform the manual counts. The labels for this dataset have been created using both low resolution 3D videos (contained in the HDF5 file) as well as high resolution 2D videos (not included in this dataset) to make the scenes more comprehensible for human annotators. *VisualCount* renders the 3D information as overlay into the 2D footage (see the left part of the screenshot below, as well as the color scheme right of the 2D/3D overlay). The height information from 3D videos was used to distinguish between children and adults with a threshold height of 1.20 meters. In the screenshot, the small person in the center of the frame would be considered an adult, while the person in the left a child. A game controller is used to navigate through the video with playback or rewind speeds corresponding to the pressure applied to the buttons to obtain both slow motion for crowded scenes as well as high speeds (10x and more) to skip through long idle timespans (see bottom right of the screenshot for the assignment of keys).



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V. LICENSE

The entire dataset consisting of the CSV file, the HDF5 file and this document is licensed under a Creative Commons Attribution-NonCommercial-NoDerivatives 4.0 International License, see https://creativecommons.org/licenses/by-nc-nd/4.0/.



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