# CV: Lab 06 Writeup

#### **Benedict Armstrong**

benedict.armstrong@inf.ethz.ch

## **Condensation Tracker**

This weeks task was to implement a condensation tracker.

## **Implementation**

The implementation of the missing functions was relatively straight forward.

#### **Color Histogram**

As suggested in the task description i started with implementing color\_histogram(). This is relatively simple with numpy's builtin histogramdd() function. The output of histogramdd() is then stacked into one big histogram.

#### Matrix A

Next step was deriving matrix a from the following equation.

$$s_t^{(n)} = As_{t-1}^{(n)} + w_{t-1}^{(n)}$$

There are two cases to consider. The first case is when we only use some noise on the particle position and ignore the velocity. In this case A is the Identity. The second case is when we use the velocity to help predict the next position. In this case

$$A = \begin{pmatrix} 1 & 0 & 1 & 0 \\ 0 & 1 & 0 & 1 \\ 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & 1 \end{pmatrix}$$

and every

 $s_t = \begin{pmatrix} x & y & v_x & v_y \end{pmatrix}^T$ 

#### **Propagation**

The propagation is done by simply multiplying the particles with the matrix A and adding some noise. The noise is sampled from a gaussian distribution with mean 0 and a parameter for the variance. Noise is added both to the velocities and the positions.

#### Observation

Here we compute the histogram for every particle and compare it to a reference histogram. We then assign a new weight to every histogram we see based on the distance to the reference. The distance is computed using the  $\chi^2$  distance. The weights are then normalized.

If all histograms have very low weight we might hit the limit of floats and get nan values. To avoid this we check if  $np.sum(particles_w) == 0$  and if so we reset all weights. This usually indicates that our tracked object has left the scene.

#### **Estimation**

The estimation is done by computing the weighted mean of the particles. This is done by multiplying every particle with its weight and then summing up all particles.

### Resampling

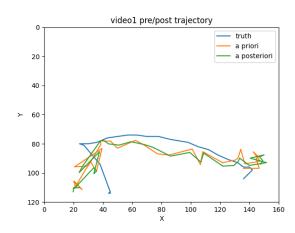
The resampling is done by sampling from the current particles with the weights as probabilities. This is done using np.random.choice().

## **Experiment**

Now for the fun part: trying it out!

To better visualize the results I manually tracked the objects to create a ground truth reference. This can then be used to calculate the difference for each run.

#### Video 1



video1 pre/post dist. avg: 13.61/13.01, norm: 86.70/84.66

a priori
a posteriori

o 5 10 15 20 25 30

Frame

Figure 1: Trajectories pre/post vs. truth

Figure 2: Distance from truth pre/post

With settings:

"hist\_bin": 16,
"alpha": 0.3,
"sigma\_observe": 0.2,
"model": 0,
"num\_particles": 40,
"sigma\_position": 15,
"sigma\_velocity": 5,
"initial\_velocity": (0, 0),

As we can see in the figures above we get a pretty good result. The trajectories look similar to the ground truth and the distance is pretty low. Seems like it's working. A gif of the tracking can be found on github.

#### Video 3

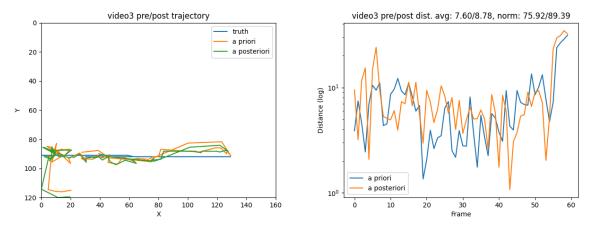


Figure 3: Trajectories pre/post vs. truth (Model 0) Figure 4: Distance from truth pre/post (Model 0)

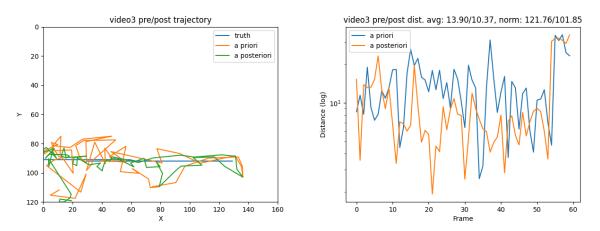


Figure 5: Trajectories pre/post vs. truth (Model 1) Figure 6: Distance from truth pre/post (Model 1) With settings:

```
"draw_plots": 1,

"hist_bin": 16,

"alpha": 0,

"sigma_observe": 0.2,

"model": 0/1,

"num_particles": 40,

"sigma_position": 10,

"sigma_velocity": 10,

"initial_velocity": (10, 0),
```

As you can see adding the extra noise with the position leads to worse results. The ball is moving at a contant speed until it changes direction. The model with the velocity struggles to follow the abrupt change in direction. The model without velocity is able to follow the ball better.

# Additional questions

Using fewer particles leads to worse results in my testing. Below 10 particles the tracker is not able to follow the object at all. Above 60 or so seems to also be too much and leads to worse results.

Using more bins seems to result in better results:

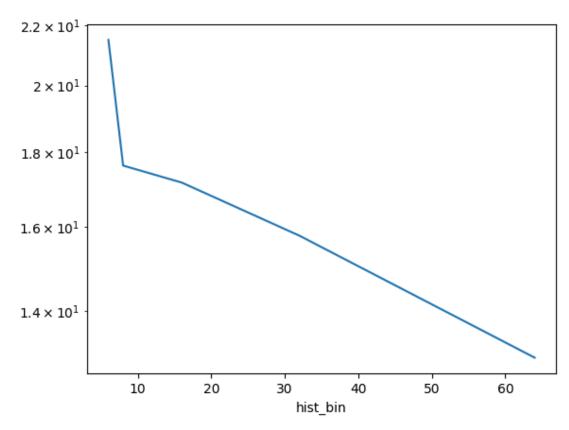


Figure 7: Distance from truth pre/post with different number of bins

Allowing the model to update is good if lighting changes (video 1 and 2) but leads to worse results for video 3 because the object is fairly constant.