## Exercises in Tracking & Detection

In this exercise we will implement a random forest for Hough voting. Luckily the training of the forest has been done already and you will only need to implement the testing. The forest has been trained to locate the center of airplanes from the Pascal VOC 2012 Dataset.

### Exercise 1 Loading the forest

The forest consists of 10 trees, each conveniently saved in a text-based format. One file per tree. Download them on the website.

- The first line of the file tells you how many internal nodes there are. We will call this number n. Then, the next n lines have to following format:  $i c_L c_R t x_0 y_0 z_0 x_1 y_1 z_1 s$ .
- i is the id of the node.  $c_L$  and  $c_R$  denote the id of the node's children. if  $c_R < 1$  (or  $c_L$  respectively) then a leaf node with id  $|c_R|$  is reached. t is the threshold.  $x_i, y_i, z_i$  with  $i \in \{0, 1\}$  denotes the first i = 0 or second i = 1 box feature center location offset, with  $z_i$  selecting the color channel. Finally s is the size of the two boxes.
- after n nodes the next line contains m the number of leaves. Again the next m lines have the form j  $p_x$   $p_y$ . j is the ID of the leaf node and  $p_x$   $p_y$  is the mean vote offset stored in that leaf.
- write a Matlab script that reads and stores a tree form such a file.

#### Exercise 2 Integral Images

To evaluate the Haar-Like features, we will need integral images. Write a function that takes a standard RGB image and computes three integral images, one for each color channel. Do not normalize the color values in any way, otherwise they do not fit the thresholds stored in the tree!

#### Exercise 3 Testing

Finally, it is time to run the forest on an image. The nodes inside the tree perform feature tests  $f(x, y, t, x_0, y_0, z_0, x_1, y_1, z_1) = b(x + x_0, y + y_0, z_0, s) - b(x + x_1, y + y_1, z_1, s) < t$ . Here (x, y) is the current pixel's location in the image. Comparing two box evaluations with a threshold. The box evaluation runs on the integral image and needs to access four values to compute the sum of all intensities in the box centered at  $(x + x_i, y + y_i)$  with size s and on channel  $z_i$ . That mean, that you compute the sum of all intensities in the square defined by  $(x + x_i - s, y + y_i - s)$  and  $(x + x_i + s, y + y_i + s)$ . If f < 0, proceed to the left child  $c_L$ , otherwise to the right child  $c_R$ .

Once you reach a leaf the you get a prediction in the form of an offset vector  $(p_x, p_y)$ . Compute the mean of the predictions from all the trees to have a final vote for the current pixel (x, y).

Create a heat-map of all predictions from all the pixels in the image. If everything is working correctly most votes should accumulate in the center of the object. Visualize the heat map and the maximum. Once you find the maximum also visualize all pixels that voted for this location.

# Frequently Asked Questions

- The color channel indiced range from 0 to 3, what do we do? That is a small inaccuracy that happened during training. The Mapping from the index to RGB is as follows: 0:B 1:G 2:R 3:R.
- Should we compute the average or just the sum of the pixels inside the boxes? In general the average makes more sense if the two boxes had different sizes. For this exercise just compute the sum and do not divide by the area.
- How do we compute the sum with integral images? Suppose your integral image is called I. Compute the feature as follows:  $b(x + x_i, y + y_i, z_i, s) = I(x + x_i + s, y + y_i + s) I(x + x_i s, y + y_i + s) I(x + x_i + s, y + y_i s) + I(x + x_i s, y + y_i s)$
- What do we do if a box extends over the image boundaries? Just assume the image is padded with zeros. That is the same as truncating the box at image boundaries. Since you do note compute the average just sum the pixels in the smaller box as well.
- How exactly do we choose to go left or right? If  $b(x + x_0, y + y_0, z_0, s) b(x + x_1, y + y_1, z_1, s) < t$  go left otherwise right.