# Intelligent Learning and Analysis Systems SS19, Exercise Sheet 1

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## 5. The $\phi$ -HH Problem: Lower Bounds

Suppose  $\Sigma_1 = \Sigma_2$ . Since  $S_1 \neq S_2$ , there is an  $x \in [n]$  such that w.l.o.g.  $x \in S_1, x \notin S_2$ .

Now let  $\sigma_1$  be the concatenation of  $\langle S_1 \rangle$  and  $\langle x \rangle$ , and let  $\sigma_2$  be the concatenation of  $\langle S_2 \rangle$  and  $\langle x \rangle$ . Since the stream lengths of  $\sigma_1$  and  $\sigma_2$  are m+1, we might have more than m' bits at our disposal, but this is not important.

Let  $\phi = \frac{2}{m+1}$ . Then every item which appears at least twice in  $\sigma_1$  or  $\sigma_2$  is a  $\phi$ -heavy hitter. Since  $S_1, S_2$  are sets, there is no item that appears twice in  $\langle S_1 \rangle$  or  $\langle S_2 \rangle$ .

Since  $x \in S_1$  but  $x \notin S_2$ , x is a heavy hitter of  $\sigma_1$ , but x is not a heavy hitter of  $\sigma_2$ . But before we processed the last element of the streams, the respective states of the storages  $\Sigma_1$  and  $\Sigma_2$  were equal by assumption. And since the last elements of both  $\sigma_1$  and  $\sigma_2$  are identical, the algorithm should have identical output for both  $\sigma_1$  and  $\sigma_2$ . But this is not the case. So we contradicted our assumption, which means  $\Sigma_1$  and  $\Sigma_2$  have to be different.

### 2. Identifying the Majority

Let  $\sigma = \langle a_1, \dots, a_m \rangle$  and  $a_i \in [n]$ .

#### **Algorithm 1** Algorithm for finding the majority of $\sigma$ , if it exists

```
2: for all i \in [m] do
         if c = 0 then
 3:
             s \leftarrow a_i
 4:
             c \leftarrow 1
 5:
        end if
 6:
         if s = a_i then
 7:
             c += 1
 8:
         else
 9:
             c -= 1
10:
11:
         end if
12: end for
13: \mathbf{return} \ s
```

Let  $A(\sigma)$  be the output of the algorithm.

Claim:  $A(\sigma)$  is the majority of  $\sigma$ , if it exists.

<u>Proof:</u> Let k be the number of times c is set to 0 in the for loop. We prove the claim by induction over k.

#### Base case:

Let k = 0. Since c is never set to 0 during the loop, we know that s never changes after iteration 1, i.e.  $A(\sigma) = a_1$ . Also, c increases at least one time more often than it decreases, so  $A(\sigma) = a_1$  has to be the majority of  $\sigma$ .

### Step case:

Let k > 0. Suppose the claim is true for all streams for which c is set to 0 at most k-1 times in the for loop. Let d be the majority of  $\sigma$ . Let j be the first iteration where c is set to 0. Side note: When c is set to 0 in an iteration j', j' has to be even. d can occur at most  $\frac{j}{2}$  times in  $\tau := \langle a_1, \ldots, a_j \rangle$ , otherwise d would be the majority of  $\tau$  and c couldn't be set to 0 in iteration j. Consequently d will occur at least

$$\left\lfloor \frac{m}{2} \right\rfloor + 1 - \frac{j}{2} = \left\lfloor \frac{m-j}{2} \right\rfloor + 1$$

times in  $\sigma' = \langle a_{j+1}, \ldots, a_m \rangle$ . Because  $\sigma'$  is a stream of length m-j, d is the majority of  $\sigma'$ . Since the number of zeros in  $\sigma'$  is smaller than k, we can apply the inductive hypothesis to see that  $A(\sigma') = d$ . Since in iteration j of computing  $A(\sigma)$ , c is set to 0, applying the algorithm to  $\sigma$  is the same as applying it to  $\tau$  and then to  $\sigma'$ . Hence,  $A(\sigma) = d$ .