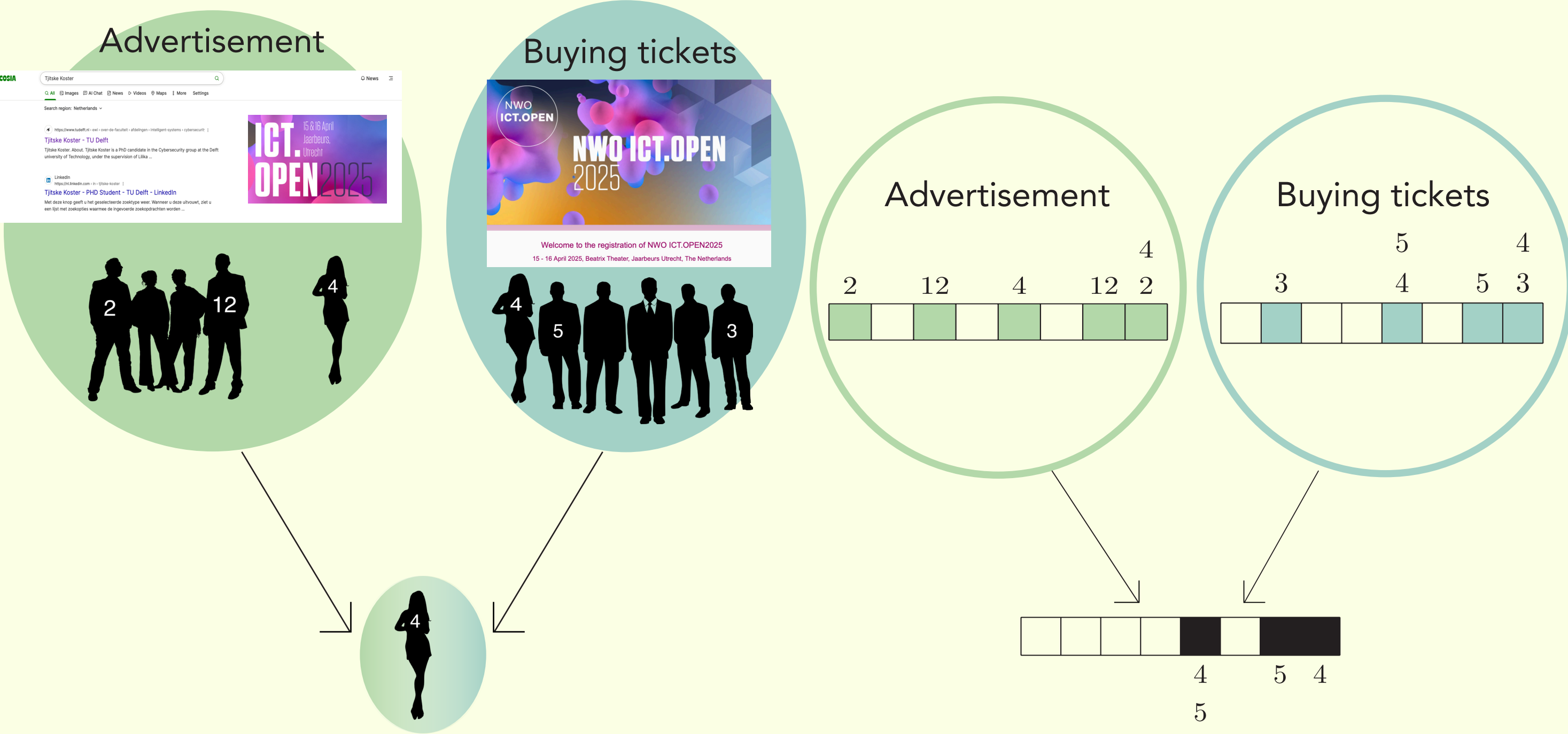
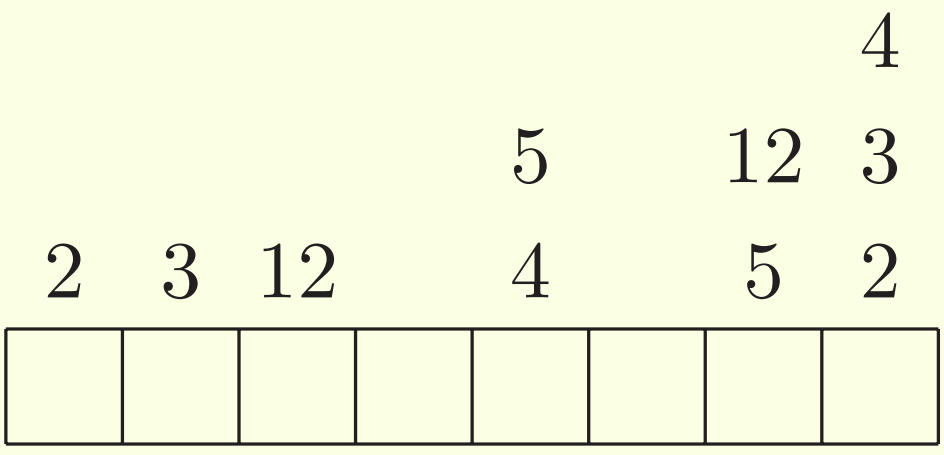


# On the Insecurity of Bloom Filter-Based Private Set Intersections

## Introduction

### Bloom filter

Hash each element twice into the filter.



## Why PSI?

We want to learn the intersection without revealing private data.

- Ads Conversion Measurement.
- Financial transactions.
- Comparison of no-fly lists.
- Private Contact Discovery.
- Password Breach Monitoring.

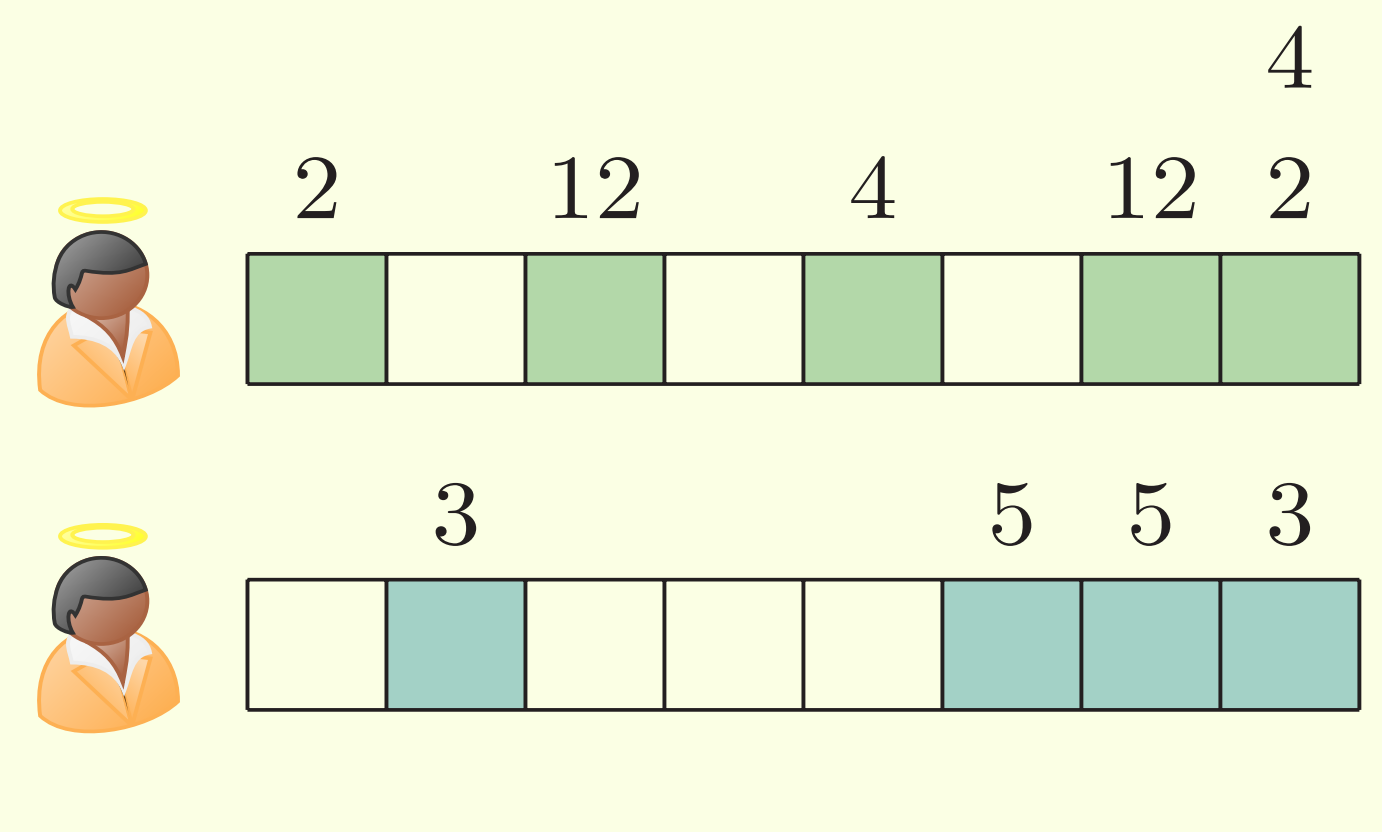
## Why Bloom filter-based PSI?

- Bloom filters are **small** compared to the input.
- Hash functions are **easy** to compute.
- The intersection is a **fast** logical AND operation.
- There might be false positives.
- No false negatives.

## Flaws in existing proofs

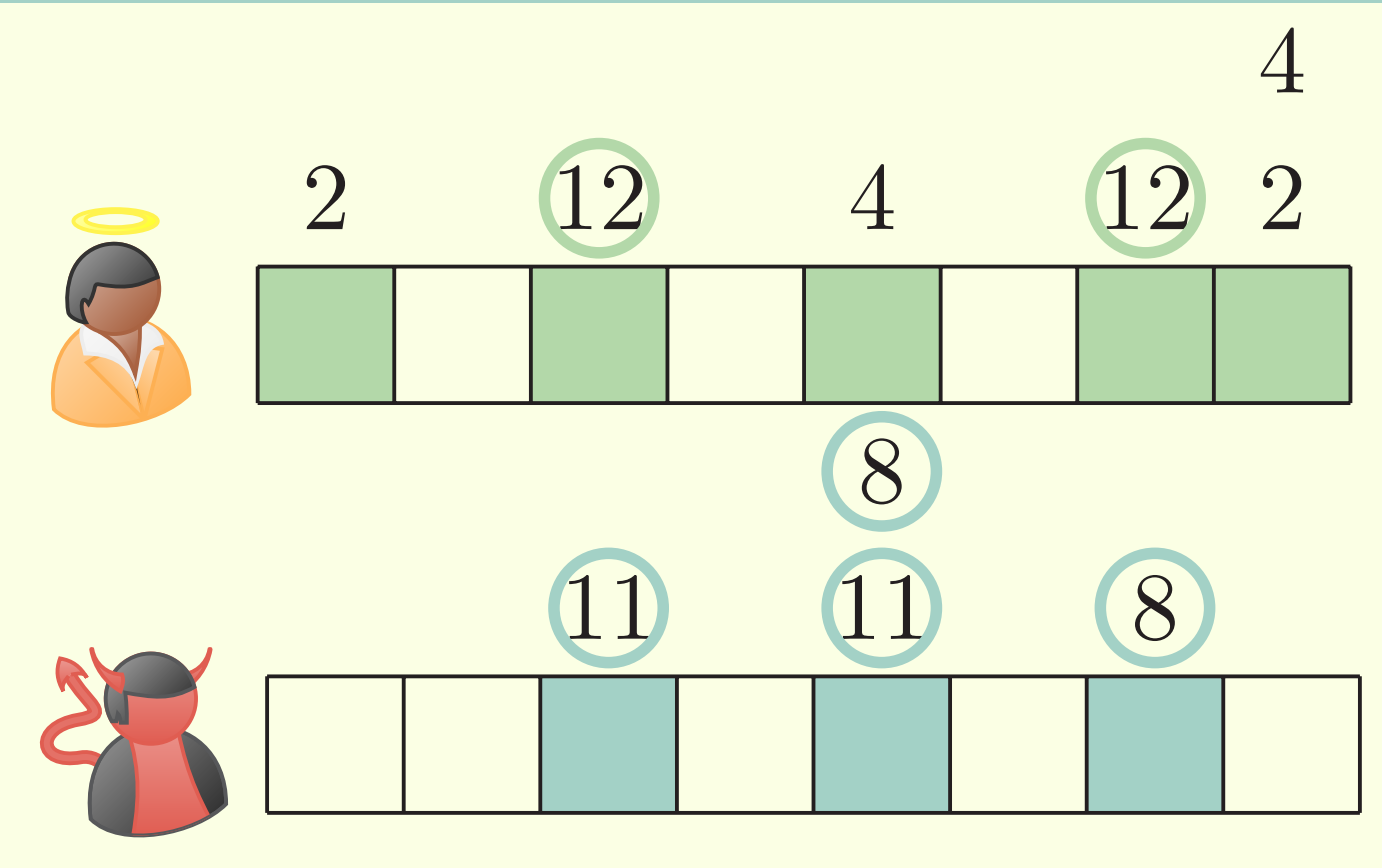
### Honest Bloom filter intersection

We expect no false positives, because the false positive rate is low.



### Input malicious Bloom filter intersection

Given a particular bloom filter we might evoke false positives.

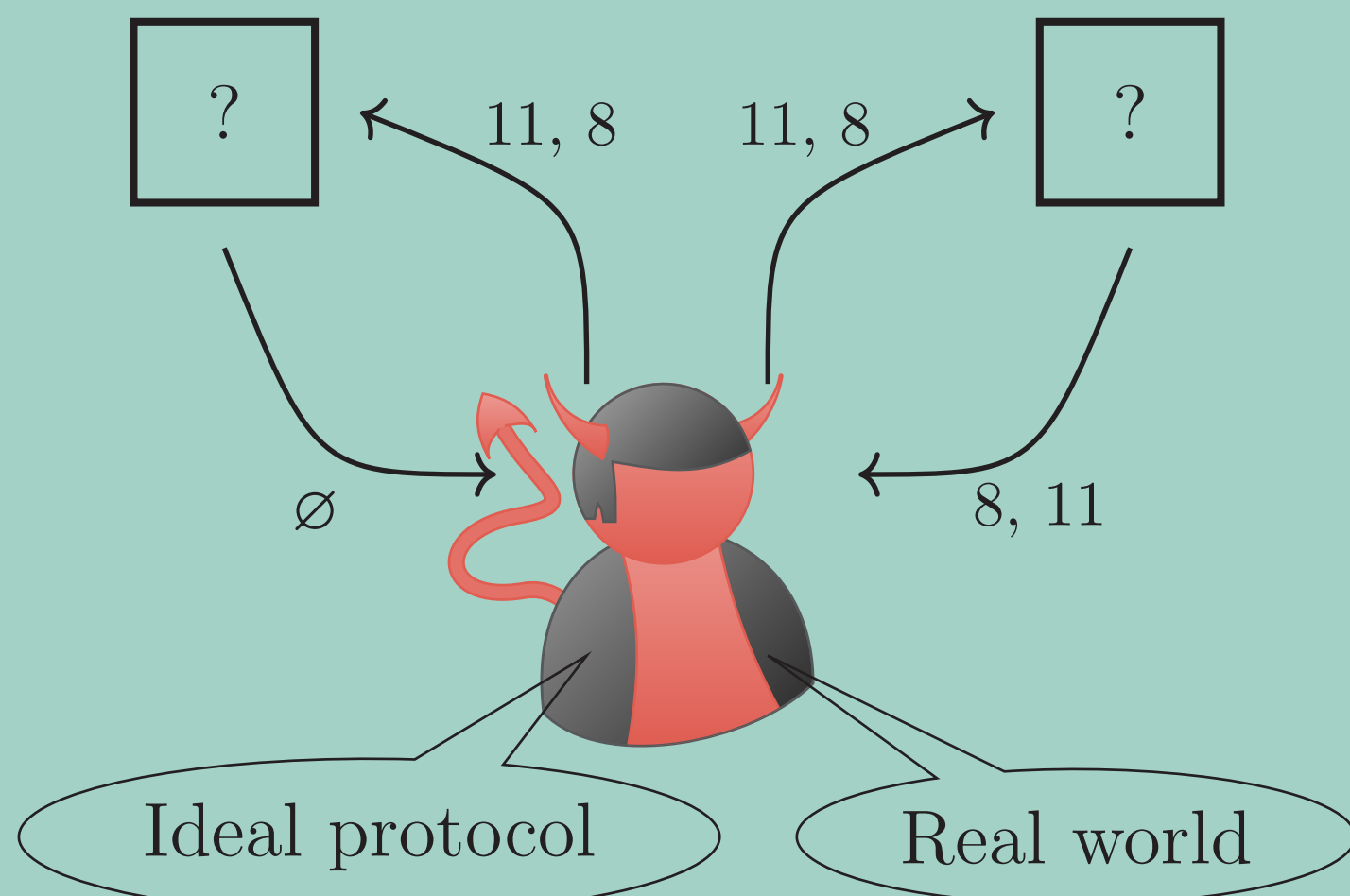


### Distinguisher real and ideal

Given a particular bloom filter we might evoke false positives.

In the **ideal functionality** the false positive rate is **low**. So, we expect no false positives.

In the **real world** we can trigger false positives, so we expect false positives.



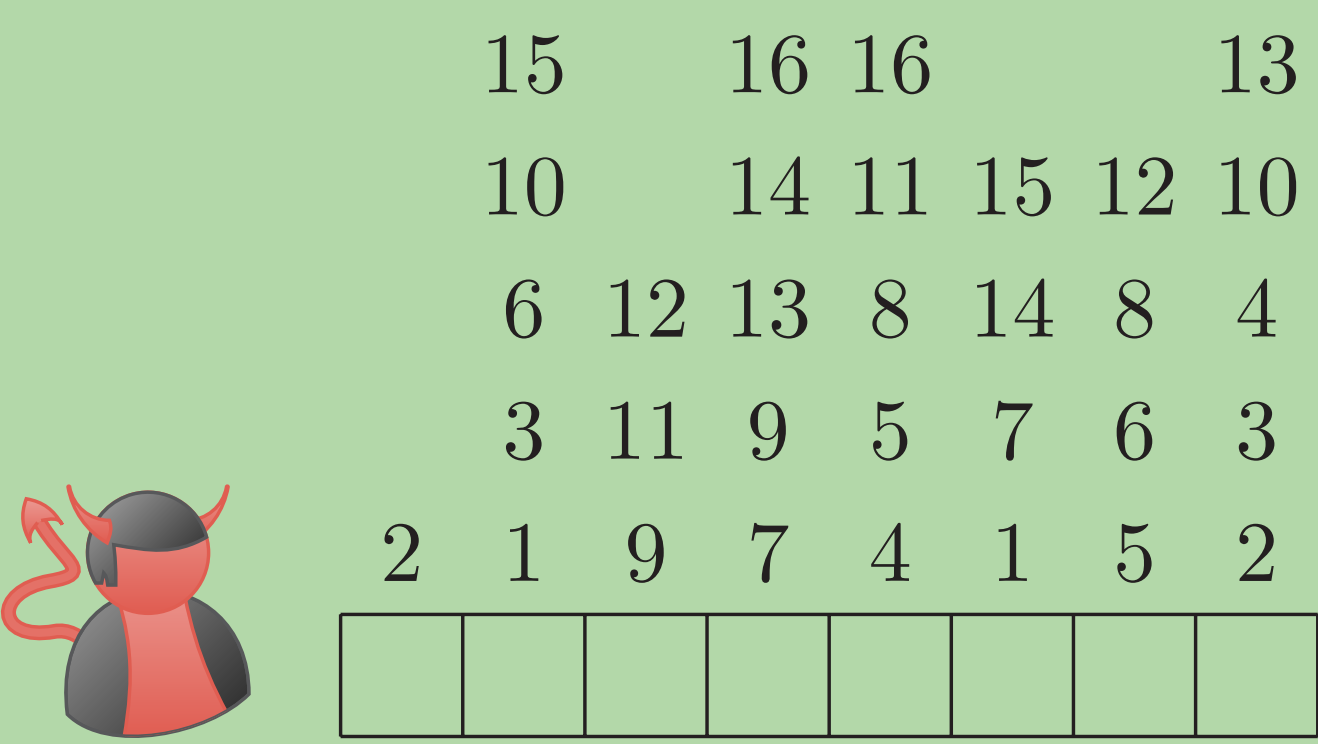
## Practical attack

### 1. Guess the input set of the other

We expect that Ecosia showed our add to 16 people.

### 2. Determine the Bloom filter

The Hash functions are public



### 3. Find target element T

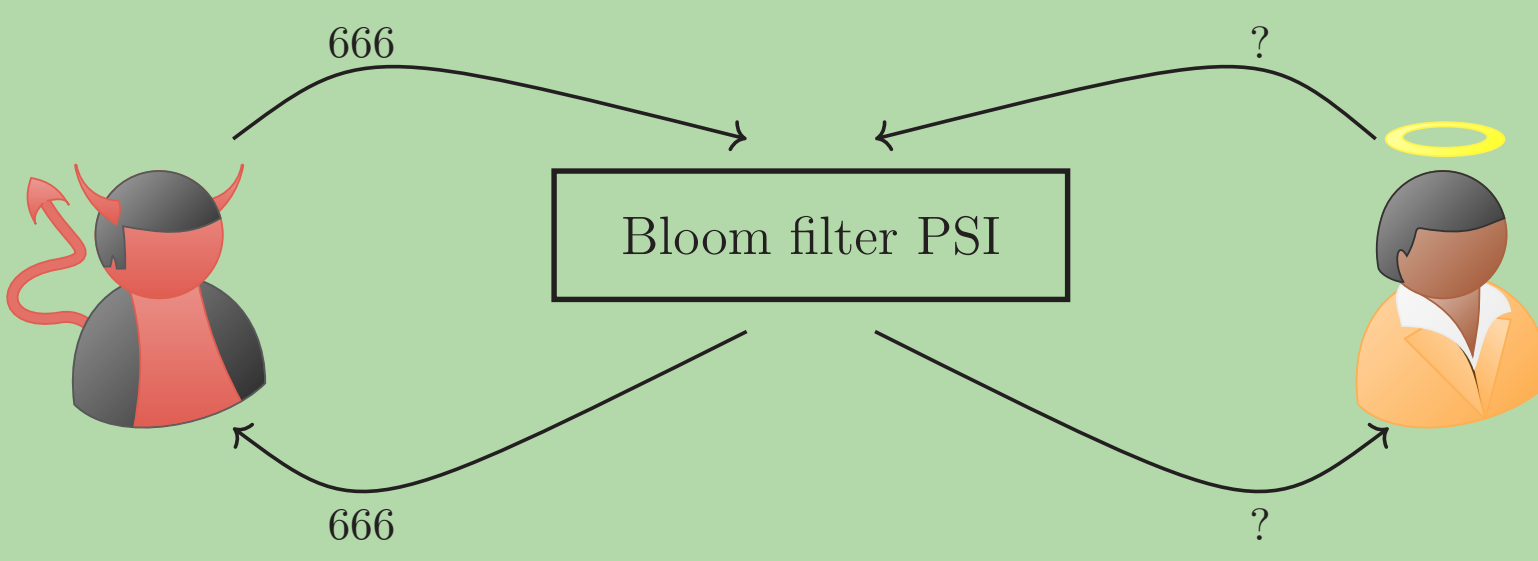
2 is an easy target, because it is alone in its bin

### 4. Find input revealing target T

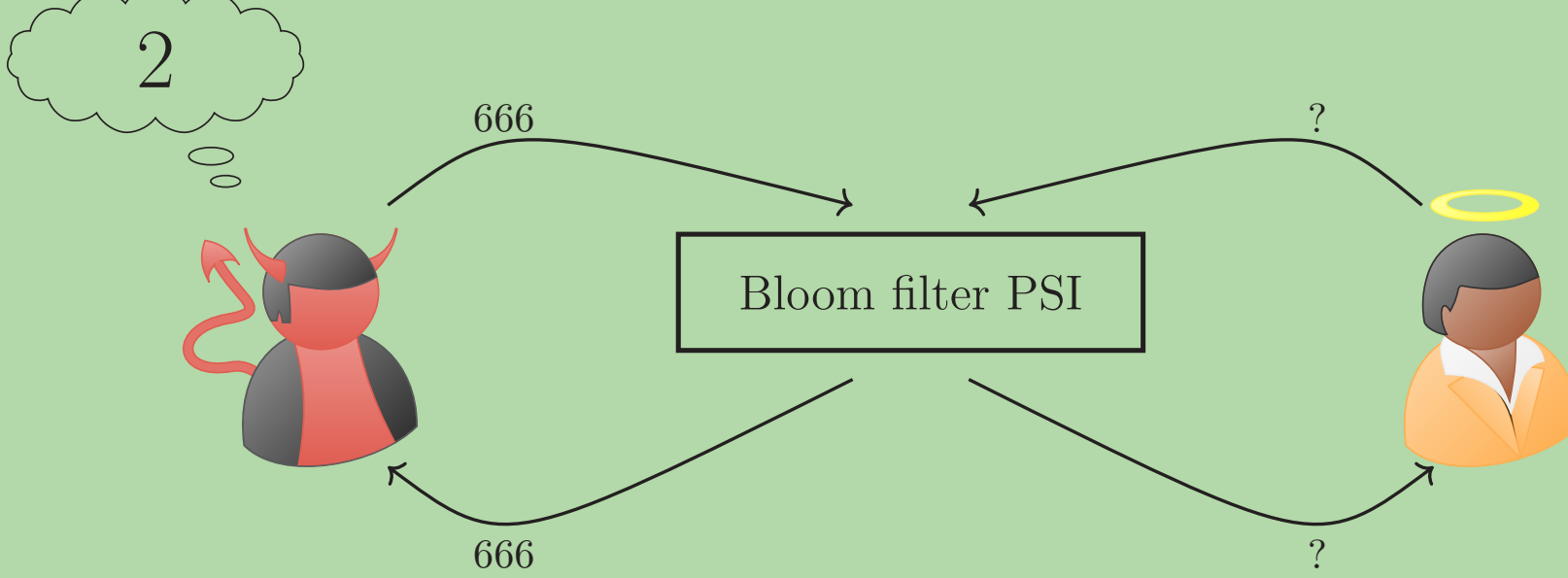
666 reveals 2 if the honest party has at least one of 4,5,8,11,16.



### 5. Execute the protocol



### 6. Conclude from intersection



## Attack in practice

### Success of the attack

- The attack succeeds for previously used parameters.
- We limit the search to 2 hours.

Setting	Parameters			Success percentage		
	set size	hash	filter size	Universe size		
false pos.	k			2k	3k	4k
$2^{-5}$	256	5	1,852	100%	100%	100%
	256	10	3,702	100%	82%	61%
	4,096	10	59,102	100%	-	-
$2^{-10}$	65,536	10	945,493	100%	-	-
	256	20	7,403	15%	1%	0.2%
	4,096	20	118,202	98%	-	-
$2^{-20}$	65,536	20	1,890,985	68%	-	-
	256	30	11,103	3%	$2^{-14}\%$	-
	4,096	30	177,302	2%	-	-
$2^{-30}$	65,536	30	2,836,477	$2^{-10}\%$	-	-

## Key takeaways

### Theoretical attack

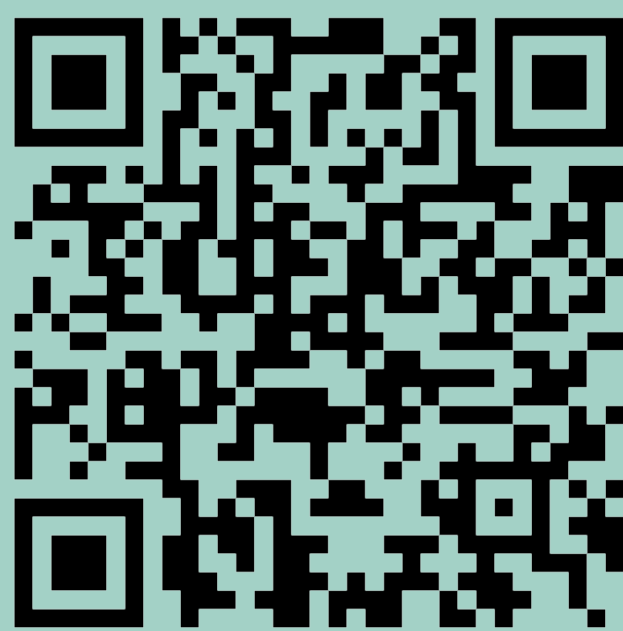
- Previous proofs were flawed.
- Bloom filters cannot be used to approximate the intersection.

### Practical attack

- Security parameters must be order of magnitudes larger;
- Or mitigations should be used.

### Mitigations

- Either take longer;
- Or more communication.



## Proposed mitigations

### How can we prevent this attack?

- Using larger parameters
- Using OPRFs
- Using PBKDFs
- Protocol takes longer
- More communication

Setting		State of the art			Mitigation 1: Large Bloom filters			Mitigation 2: OPRFs			Mitigation 3: PBKDFs		
Parties	Set size	Time	Rounds	Comm.	Time	Rounds	Comm.	Time	Rounds	Comm.	Time	Rounds	Comm.
2	256	0.25	5	0.27 MB	3.26	5	3.66 MB	0.31	7	0.30 MB	21.63	5	0.27 MB
	4096	3.87	5	4.36 MB	51.38	5	58.46 MB	4.89	7	4.86 MB	345.58	5	4.36 MB
	65536	60.74	5	69.71 MB	815.37	5	935.33 MB	78.21	7	77.71 MB	> 1h	5	69.71 MB
3	256	0.38	5	0.55 MB	5.29	5	7.32 MB	0.53	7	0.64 MB	21.91	5	0.55 MB
	4096	6.06	5	8.71 MB	82.59	5	116.93 MB	8.44	7	10.21 MB	351.11	5	8.71 MB
	65536	97.37	5	139.42 MB	1338.55	5	1.83 GB	138.80	7	163.42 MB	> 1h	5	139.42 MB