

CERIAS

The Center for Education and Research in Information Assurance and Security

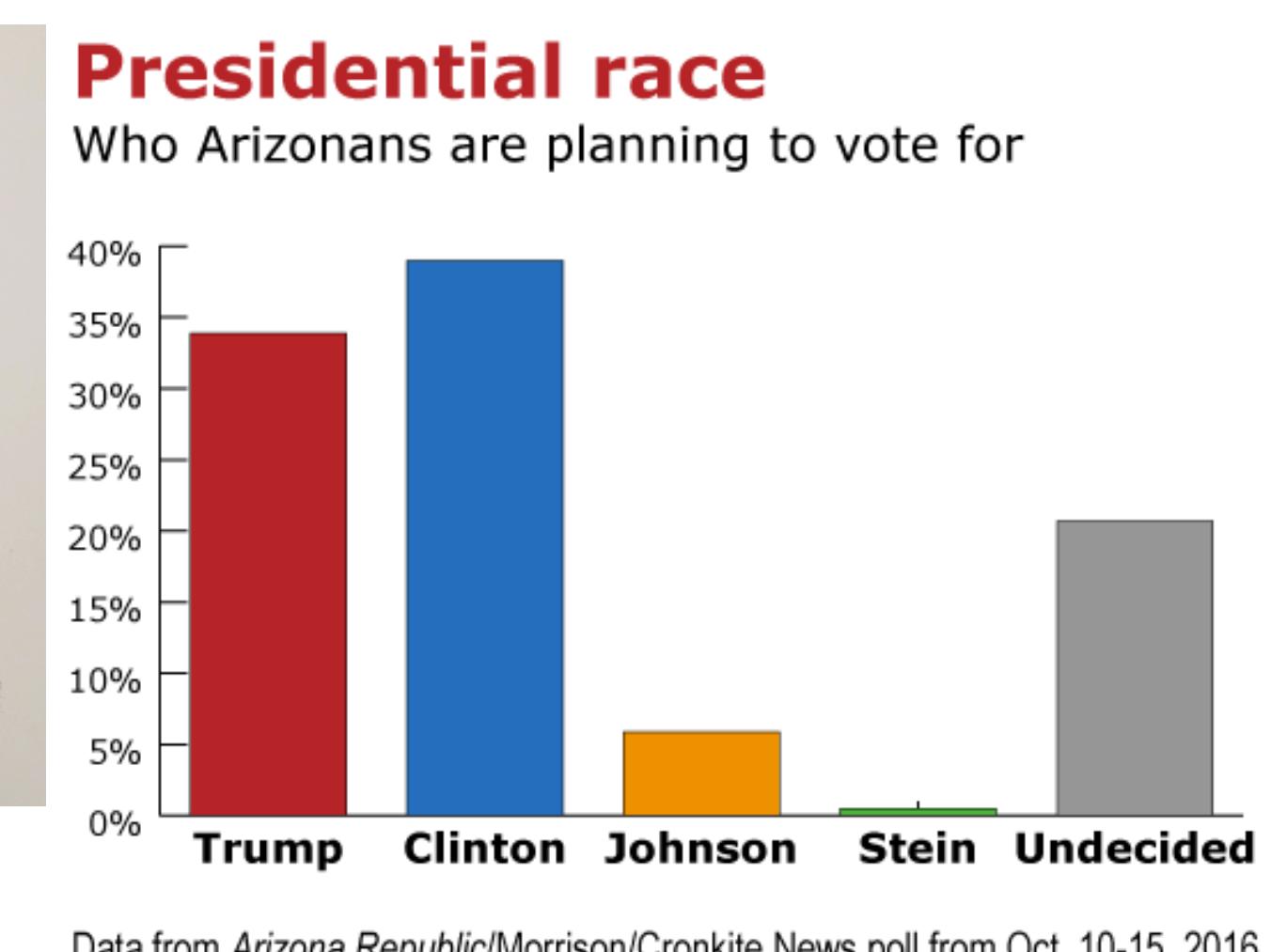
Differentially Private Heavy Hitter Problem

1. An embarrassing question



Q true answer
I like false answer

2. Toss a coin



3. With many people



4. Good estimation

III. Proposed Method

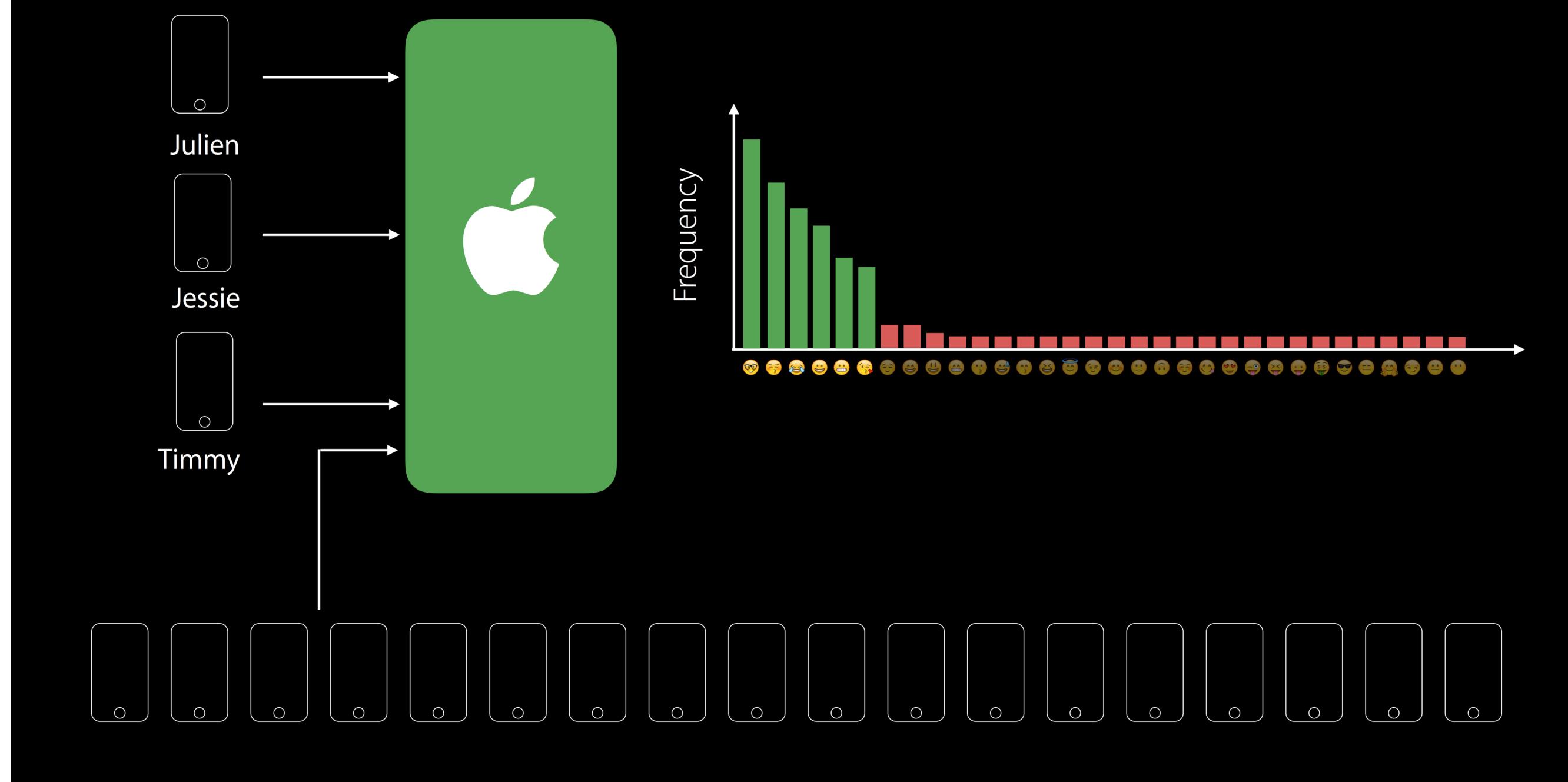
I. Differential Privacy in the Local setting

- Each user perturbs her own data before sending it to the aggregator

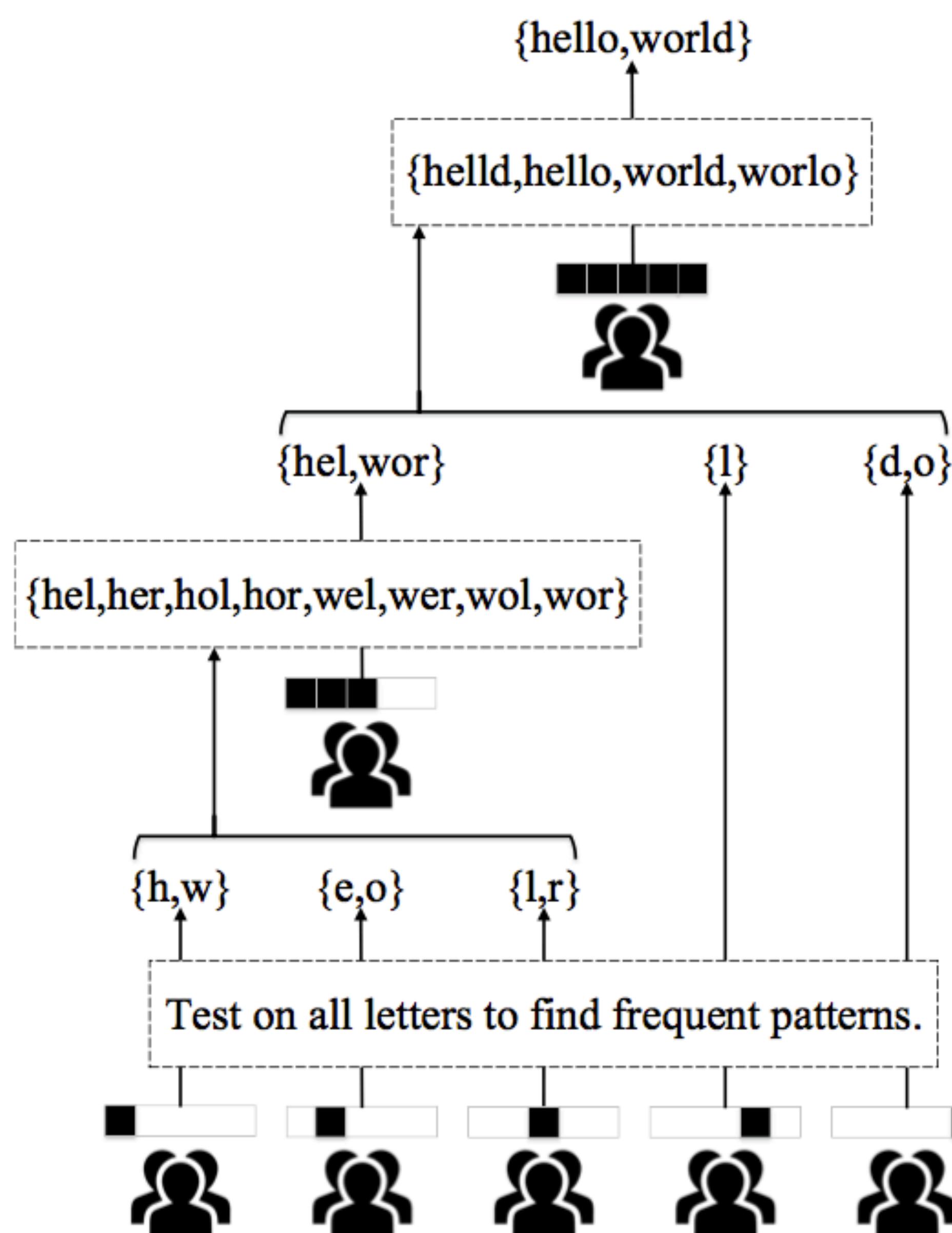


II. Frequency Estimator under LDP

Learning Popular Emojis with Privacy



- Good for small domain
- Inefficient and inaccurate for large domain -> focus of this research



IV. Results

ϵ	method	$n = 10^6$	
		$m = 64$	$m = 256$
2	HieProd	19.8(1.2) ; 2.9(0.1)	1.7(0.9) ; 0.4(0.2)
	SPM	0.3(0.6) ; 0.1(0.1)	0.0(0.0) ; 0.0(0.0)
	IterProd	20.1(1.1) ; 2.9(0.1)	2.1(0.7) ; 0.5(0.2)
	HASH	0.0(0.0) ; 0.0(0.0)	0.0(0.0) ; 0.0(0.0)
	SimHash	16.0(1.2) ; 2.6(0.1)	0.6(0.5) ; 0.1(0.1)
4	HieProd	63.4(2.6) ; 5.1(0.1)	24.9(0.7) ; 3.4(0.1)
	SPM	11.8(1.0) ; 2.1(0.1)	0.0(0.0) ; 0.0(0.0)
	IterProd	62.6(1.6) ; 5.1(0.1)	24.5(1.3) ; 3.3(0.1)
	HASH	0.0(0.0) ; 0.0(0.0)	0.0(0.0) ; 0.0(0.0)
	SimHash	51.5(2.4) ; 4.7(0.1)	21.8(1.3) ; 3.1(0.1)