

# Differential Privacy

		Statistics
Alice	P	① How many passed?
Bob	P	② Did Dave pass? <u>private</u>
Charlie	P	③ How many girls passed?
Dave	P	grey area...
Eve	F	
Fred	P	
Gina	F	

Private Information

ESTR 2018 Grades

① Anonymize. This could work in some cases.

Name	Weight	Smoker
Bob	57kg	Yes

side-info ← "How many 57kg heavy smokers are there?" → can still infer some info from this.

Doesn't work in general

② Set up rules: e.g. in total 100 students, only answer queries that concern 50 or more students.

Still has problems.

Ask 2 questions: "How many passed?"  
"How many not named Dave passed?" X

③ "Lie" a little bit (☺)

Eg. 100 students, 85 passed.

"How many passed?"

"How many  $\neq$  Dave passed?"

True      My answer

85              88

84              82

How to "lie"?

A mechanism is a probabilistic algorithm for answering queries.

On every question, lie independently.

Alice      P      Take ans  
P      P      Flip a coin with prob.  $p$ ,  
change a grade

Bob      P      P      Say  $p=0.2$ ,

Charlie      P      F      Let  $M$  be the answer now  
which is a r.v.

Dave      P      P

Eve      F      P

Fred      P      P

Gina      F      F

# Passes =  $a$

true ans

$\hat{A}$

r.v.

$$E[\hat{A}] = (1-p)a + p(n-a)$$

$$= (1-2p)a + pn$$

We want  $E[M] = a$ .

$$\Rightarrow M \text{ outputs } \frac{\hat{A} - pn}{1-2p}$$

Utility  $\equiv$  Privacy

Typical difference

$$\rightarrow E[(M-a)^2] = \text{Var}[M]$$

$$\rightarrow \text{So let } U(M) = \frac{1}{\delta(M)} = \frac{1}{\sqrt{\text{Var}[M]}}$$

If  $M=a$ , w. p. 1,  $\text{Var}[M]=0$ ,  $U(M) \rightarrow \infty$ , but no privacy.

Privacy.

You have some beliefs about the grades.  $P_1, P_2, \dots, P_n$ . e.g.,  $P(P_{\text{Dave}}=1 \text{ AND } P_{\text{Bob}}=0) = 0.2$ .

Knowing info from  $M$  should not change my belief, otherwise it's leaking information.

$$\text{So, } P(P_{\text{Dave}}=1 | M=m, P_{\text{Dave}}=P) = P(P_{\text{Dave}}=1 | P_{\text{Dave}}=P)$$

But this is not quite right.

(nothing can change your "prejudice")

$P_{\text{-Dave}}$  = grades for all other students.

This should hold for all students  $\Rightarrow$

$M$  is independent of the true answers

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$\frac{1}{\epsilon}$  - Differential Privacy:

For  $\forall$  two databases that differ in 1 row for every  $m$ ,

$$e^{-\epsilon} \leq \frac{P(M=m)}{P(M'=m)} \leq e^{\epsilon}, \quad \epsilon \rightarrow 0.$$

This means:

$$P(P_{\text{Dave}}=1 | M=m, P_{\text{Dave}}=P) = \frac{P(M=m | P_D=1, -) \cdot P(P_D=1 | -)}{P(M=m | -)}$$

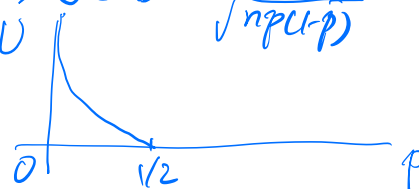
change by at most  $e^{\pm \epsilon}$ ,  $\approx 1 \pm \epsilon$ .

Utility of  $M$ :  $\frac{1}{\delta(M)}$

$\frac{\hat{A} - pn}{1-2p}$   $\hat{A}$ : flip each row w.p.  $p$ .

$$\text{Var}[A] = \text{Var}[\text{Binomial}(n, p)] = np(1-p)$$

$$\Rightarrow \delta(M) = \frac{\sqrt{np(1-p)}}{1-2p} \Rightarrow U(M) = \frac{1-2p}{\sqrt{np(1-p)}}$$



Privacy of  $M$ :  $\frac{1}{\epsilon}$

Draw  $p$

Draw  $F$

$$\frac{p(\text{outcome})}{p(\text{outcome}')} = \frac{p}{1-p} \text{ or } \frac{1-p}{p}$$

$\in e^{\pm \epsilon}$  where

$$\epsilon = \ln \frac{1-p}{p}$$

