

The Hardness of Being Private

Anil Ada, Arkadev Chattopadhyay, Stephen Cook,
Lila Fontes, Michal Koucký, Toniann Pitassi

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Communication complexity

Two-player model

- each player has a private input (Alice has $x \in X$, Bob has $y \in Y$)
- players communicate over a channel
- players follow a protocol to compute $f : X \times Y \rightarrow Z$
- the last message sent is the value $f(x, y) = z$

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The **communication cost** of a protocol is the worst-case length of the full transcript.

Matrix M_f has entries $M_f[x, y] = f(x, y)$.

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Vickrey auction

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$$X = Y = [2^n], Z = [2^{n+1}] \text{ and } f(x, y) = \begin{cases} (x, B), & \text{if } x \leq y \\ (y, A) & \text{if } y < x \end{cases}$$

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	1	2	3	4	...	$2^n - 1$	2^n
1	(1, B)	(1, B)	(1, B)	(1, B)	...	(1, B)	(1, B)
2	(1, A)	(2, B)	(2, B)	(2, B)	...	(2, B)	(2, B)
3	(1, A)	(2, A)	(3, B)	(3, B)	...	(3, B)	(3, B)
4	(1, A)	(2, A)	(3, A)	(4, B)	...	(4, B)	(4, B)
\vdots	\vdots	\vdots	\vdots	\vdots	\ddots	\vdots	\vdots
$2^n - 1$	(1, A)	(2, A)	(3, A)	(4, A)	...	$(2^n - 1, B)$	$(2^n - 1, B)$
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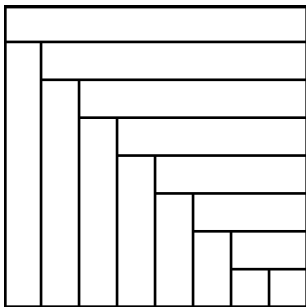
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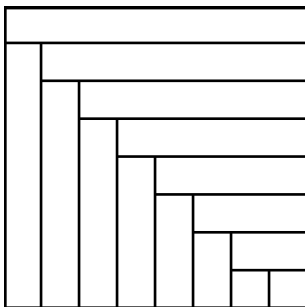
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Regions (preimages)

$$\text{region } R_{x,y} = \{(x', y') \in X \times Y \mid f(x, y) = f(x', y')\}$$

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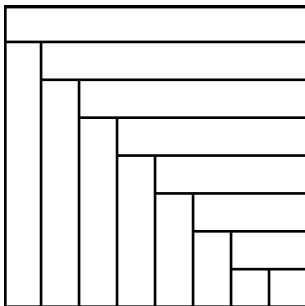
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Rectangles

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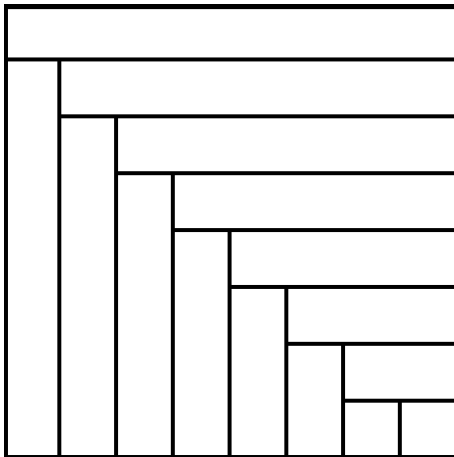
defined by **protocol**

Privacy against eavesdroppers

Can an eavesdropper learn about x and y , aside from $z = f(x, y)$?

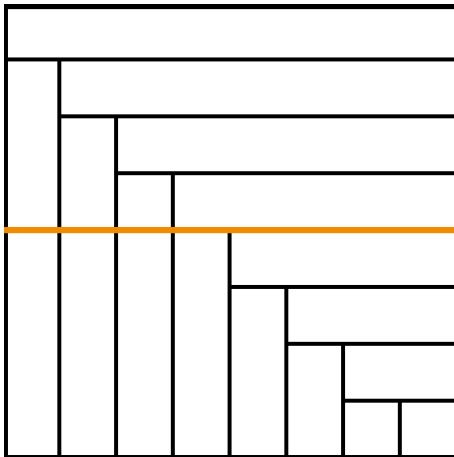
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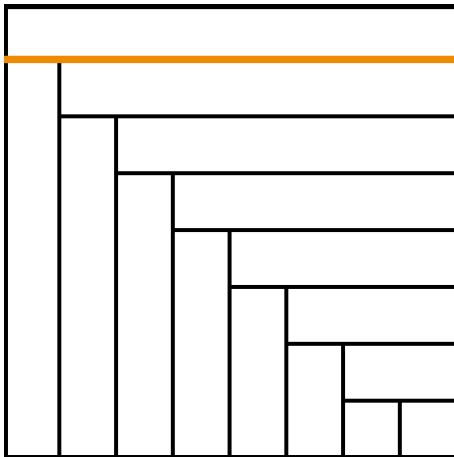
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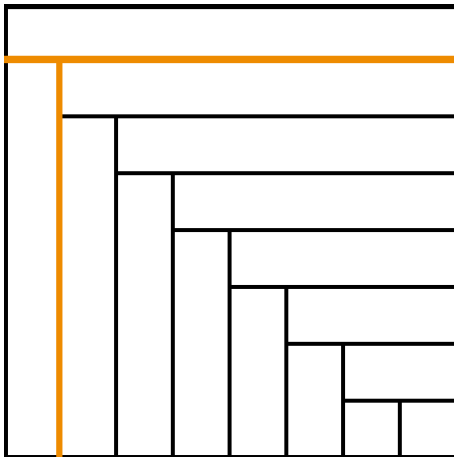
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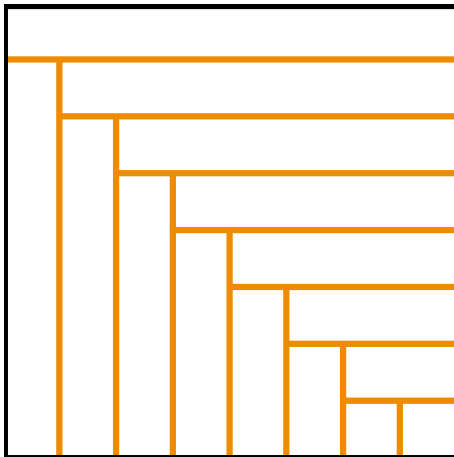
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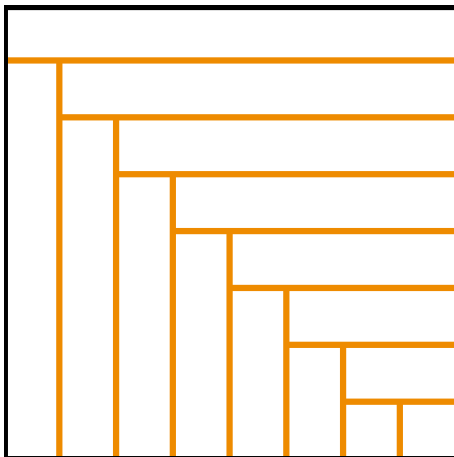
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Ascending English bidding.

Perfect privacy

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Characterizing perfect privacy (Kushilevitz '89)

The perfectly private functions of 2 inputs are fully characterized combinatorially. A private deterministic protocol for such functions is given by “decomposing” M_f .

Approximate privacy

Privacy approximation ratio (Feigenbaum Jaggard Schapira '10)

A protocol for f has **worst-case privacy approximation ratio**:

$$\text{worst-case PAR} = \max_{(x,y)} \frac{|R_{x,y}|}{|P_{x,y}|}$$

$$\text{average-case PAR} = \mathbb{E}_{(x,y)} \frac{|R_{x,y}|}{|P_{x,y}|} \text{ over distribution } \mathcal{U}$$

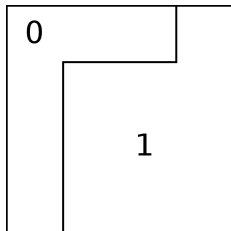
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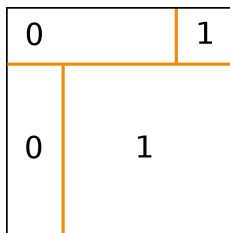
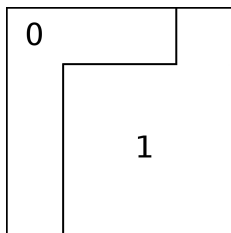
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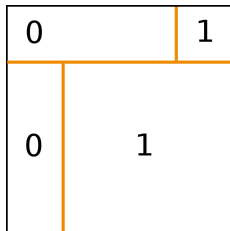
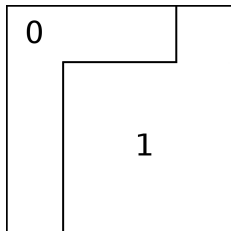
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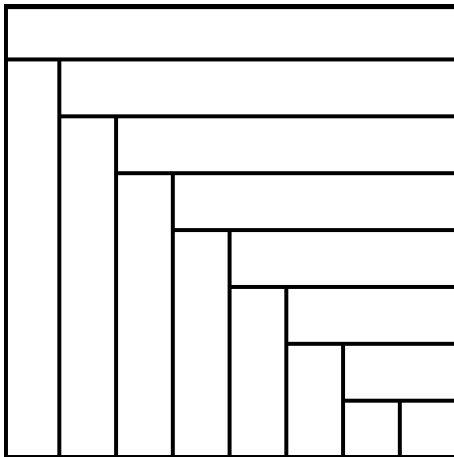
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worst-case PAR = 10
average-case PAR = 2

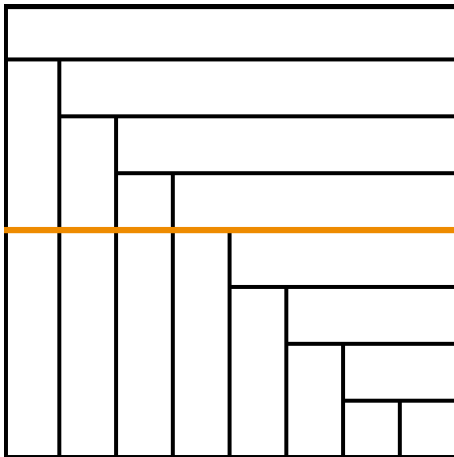
Two-player Vickrey auction

Bisection protocol



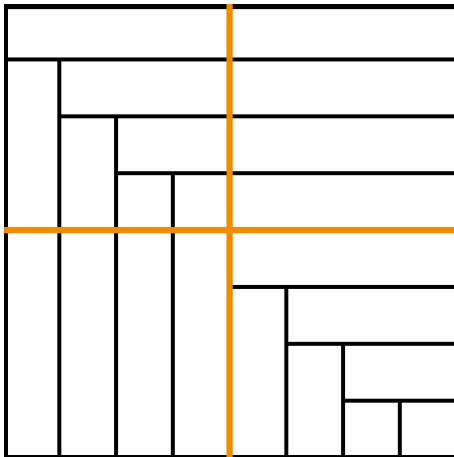
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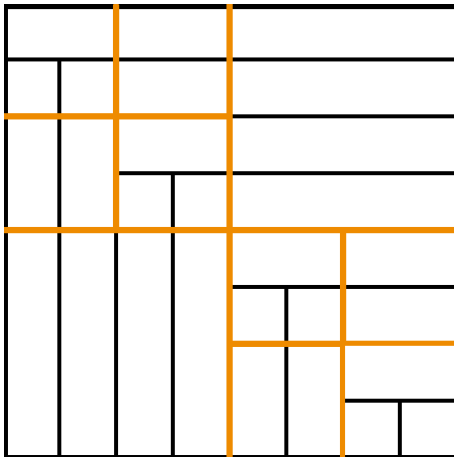
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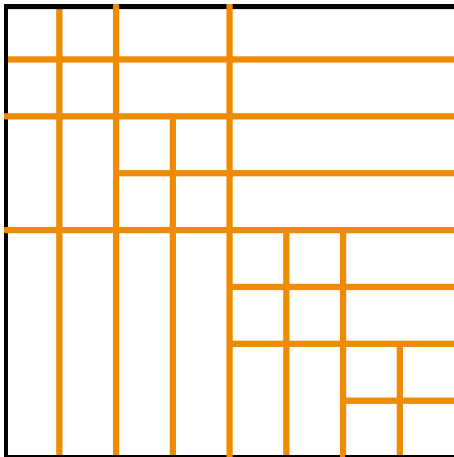
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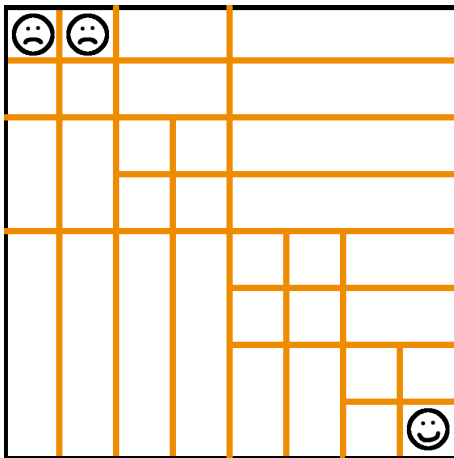
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Upper bounds (Feigenbaum Jaggarid Schapira '10)

	English bidding	bisection protocol
communication cost	2^n	$O(n)$
worst-case PAR	1	2^n
average-case PAR	1	$O(1)$

Theorem 1: worst-case lower bound

For all n , for all p , $2 \leq p \leq n/4$, any deterministic protocol for the n -bit two-player Vickrey auction obtaining PAR less than 2^{p-2} has length at least $2^{n/4p}$.

Our contributions

Theorem 1: worst-case lower bound

For all n , for all p , $2 \leq p \leq n/4$, any deterministic protocol for the n -bit two-player Vickrey auction obtaining PAR less than 2^{p-2} has length at least $2^{n/4p}$.

Theorem 2: average-case lower bound

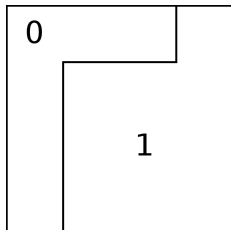
For all $n, r \geq 1$, any deterministic protocol of length at most r for the n -bit two-player Vickrey auction has average-case PAR greater than $\Omega\left(\frac{n}{\log(r/n)}\right)$.

Privacy against players

Can Bob learn anything about Alice's private input x , beyond the fact that $z = f(x, y)$? Can Alice learn anything about Bob's private input y ?

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Subjective regions

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defined by **function**
Alice sees

0	
	1

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Subjective rectangles

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Subjective privacy approximation ratio (Feigenbaum Jaggard Schapira '10)

$$\text{average-case PAR}^{\text{sub}} = \max_{v=A,B} \mathbb{E}_{(x,y)} \frac{|R_{x,y}^v|}{|P_{x,y}^v|}$$

Theorem (Braverman '11): $IC(DISJ) = \Omega(n)$.

information cost IC

$$IC = I(\mathbf{X} : \Pi_P(\mathbf{X}, \mathbf{Y}) | \mathbf{Y}) + I(\mathbf{Y} : \Pi_P(\mathbf{X}, \mathbf{Y}) | \mathbf{X})$$

informational privacy $PRIV_D$ (Klauck '02)

$$PRIV_D(P) = \max\{I(\mathbf{X} : \Pi_P(\mathbf{X}, \mathbf{Y}) | \mathbf{Y}, f(\mathbf{X}, \mathbf{Y})), I(\mathbf{Y} : \Pi_P(\mathbf{X}, \mathbf{Y}) | \mathbf{X}, f(\mathbf{X}, \mathbf{Y}))\}$$

Theorem: $PRIV_D - \log |Z| \leq IC \leq 2(PRIV_D + \log |Z|)$

Theorem: $PRIV_D(P) \leq \log(\text{avg}_D \text{PAR}^{sub}(P))$

Theorem 3

Any protocol P computing the n -bit Set Intersection $INTERSEC_n$ has exponential average-case subjective PAR:

$$\text{avg}_{\mathcal{U}} \text{PAR}^{sub}(P) = 2^{\Omega(n)}$$

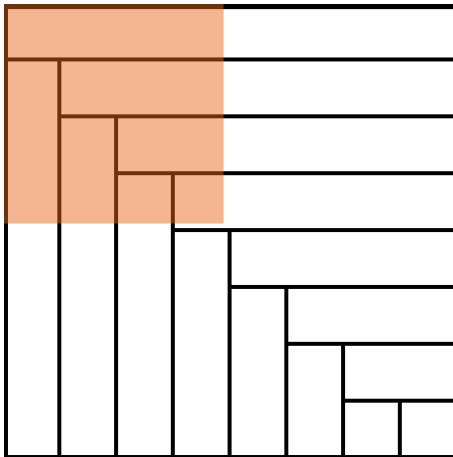
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progress:
steps that
look like
bisection.



useless:
steps that
look like
English
bidding.

Future directions

- a “good” unified definition of privacy
- length-privacy tradeoffs for other functions
- general results for length-privacy tradeoffs
- randomized protocols
- protocols with error
- approximate privacy hierarchy?
- more than 2 players
- privacy against coalitions?

Ball Partition Problem

For integers N and $r \geq 1$, there are N balls and r rounds. All of the balls begin in one big set. In each round, the balls in each current set are partitioned into (at most) two new sets. The cost of partitioning the balls in any set S into sets S_1 and S_2 is $\min(|S_1|, |S_2|)$. After r rounds, each of the N balls shall be in a singleton set. The total cost of the game is the sum of the cost, over all r rounds, of every partition made during each round. We denote the minimal possible cost by $B(N, r)$.

Theorem 17

For the Ball Partition Problem, $B(N, r) \geq \frac{N \log N}{4 \log(\frac{4r}{\log N})}$.

Average-case PAR

We define it slightly differently.

Theorem 2: average-case lower bound

For all $n, r \geq 1$, any deterministic protocol of length at most r for the n -bit two-player 2^n -Vickrey auction problem has average-case PAR greater than $\Omega\left(\frac{n}{\log(r/n)}\right)$ (over the uniform distribution of inputs).

Proof: The Ball Partition problem simplifies the analysis of arbitrary protocols to an analysis of protocol trees and probability.

Answered Feigenbaum conjecture about set intersection:

Theorem 3

For all $n \geq 1$, and any protocol P computing the Set Intersection INTERSEC_n the average-case subjective PAR is exponential in n under the uniform distribution:

$$\text{avg}_{\mathcal{U}} \text{PAR}^{\text{sub}}(P) = 2^{\Omega(n)}$$

Relating PAR to info measures. Definitions of mutual information measures of privacy (nice because info=0 corresponds to perfect privacy) [Kla02, Bra11]. *Theorem 21* Info theoretic privacy \leq log of average PAR.

$$\text{PRIV}_D(P) \leq \log(\text{avg}_D \text{PAR}^{\text{sub}}(P))$$



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