# Communication Complexity of Set-disjointness for All Probabilities 

Mika Göös \& Thomas Watson<br>University of Toronto

## Communication complexity?



## Communication complexity?



## Set-disjointness: $x \cap y=\varnothing$ ?

## Communication complexity?

## Alice

$$
x \subseteq[n]
$$

## Bob

$$
y \subseteq[n]
$$

## Set-disjointness: $x \cap y=\varnothing$ ?

[Kalyanasundaram-Schnitger'92], [Razborov'92], [Bar-Yossef et al.'04] ...

## Main result

## Bounded-error model:

- yes-inputs accepted with prob. $\geq 99 \%$
- no-inputs accepted with prob. $\leq 1 \%$


## Main result

Our focus: Arbitrary probabilities $\alpha(n)>\beta(n)$ :

- yes-inputs accepted with prob. $\geq \alpha$
- no-inputs accepted with prob. $\leq \beta$


## Main result

Our focus: Arbitrary probabilities $\alpha(n)>\beta(n)$ :

- yes-inputs accepted with prob. $\geq \alpha$
- no-inputs accepted with prob. $\leq \beta$

■ Public vs. private coins

Tight bound: $\quad \Theta(n \cdot(1-\beta / \alpha))$

Simplifies: [Braun et al., FOCS'12]: EFs for max-clique
[Braverman-Moitra, STOC'13]: $\alpha=1 / 2+\epsilon, \beta=1 / 2-\epsilon$

## Main result

Our focus: Arbitrary probabilities $\alpha(n)>\beta(n)$ :

- yes-inputs accepted with prob. $\geq \alpha$
- no-inputs accepted with prob. $\leq \beta$
- Public vs. private coins

Tight bound: $\quad \Theta(n \cdot(1-\beta / \alpha))$

Simplifies: [Braun et al., FOCS'12]: EFs for max-clique
[Braverman-Moitra, STOC'13]: $\alpha=1 / 2+\epsilon, \beta=1 / 2-\epsilon$

Key insight: $\quad$ Suffices to understand case $\beta=\alpha / 2$

## SBP: Case $\beta=\alpha / 2$

## SBP: Small bounded-error computations

- yes-inputs accepted with prob. $\geq \alpha$
- no-inputs accepted with prob. $\leq \alpha / 2$


## SBP: Case $\beta=\alpha / 2$

## SBP: Small bounded-error computations

- yes-inputs accepted with prob. $\geq \alpha$
- no-inputs accepted with prob. $\leq \alpha / 2$

New: $\quad \mathbf{S B P}(f)=\min _{\alpha(n)>0} \mathrm{R}_{\alpha, \alpha / 2}^{\text {pub }}(f)+\log (1 / \alpha)$

## SBP: Case $\beta=\alpha / 2$

## SBP: Small bounded-error computations

■ yes-inputs accepted with prob. $\geq \alpha$

- no-inputs accepted with prob. $\leq \alpha / 2$

New: $\quad \mathbf{S B P}(f)=\min _{\alpha(n)>0} \mathrm{R}_{\alpha, \alpha / 2}^{\mathrm{pub}}(f)+\log (1 / \alpha)$

Compare: $\quad \mathbf{P P}(f)=\min _{\epsilon(n)>0} \mathrm{R}_{1 / 2+\epsilon, 1 / 2-\epsilon}^{\text {pub }}(f)+\log (1 / \epsilon)$

## SBP: Case $\beta=\alpha / 2$

## SBP: Small bounded-error computations

- yes-inputs accepted with prob. $\geq \alpha$
- no-inputs accepted with prob. $\leq \alpha / 2$

$$
\text { New: } \quad \begin{aligned}
\quad \operatorname{SBP}(f) & =\min _{\alpha(n)>0} \mathrm{R}_{\alpha, \alpha / 2}^{\mathrm{pub}}(f)+\log (1 / \alpha) \\
\mathbf{U S B P}(f) & =\min _{\alpha(n)>0} \mathrm{R}_{\alpha, \alpha / 2}^{\text {priv }}(f) \\
\text { Compare: } \quad \operatorname{PP}(f) & =\min _{\epsilon(n)>0} \mathrm{R}_{1 / 2+\epsilon, 1 / 2-\epsilon}^{\text {pub }}(f)+\log (1 / \epsilon) \\
\operatorname{UPP}(f) & =\min _{\epsilon(n)>0} \mathrm{R}_{1 / 2+\epsilon, 1 / 2-\epsilon}^{\text {priv }}(f)
\end{aligned}
$$

## SBP in context


[Klauck'07]: $\quad$ PP $=$ Disc
[Klauck'03]: $\quad \mathbf{M A} \subseteq$ Corr $\subseteq \mathbf{A M}$

## SBP in context


[Klauck'07]: $\quad$ PP $=$ Disc
[Klauck'03]: $\quad \mathbf{M A} \subseteq$ Corr $\subseteq \mathbf{A M}$

## Results for SBP and USBP

## Theorem: $\quad \mathbf{S B P}(f)=\Theta(\operatorname{Corr}(f))$

## Results for SBP and USBP

## Theorem: $\quad \mathbf{S B P}(f)=\Theta(\operatorname{Corr}(f))$

## Corruption bound:

- Let $\mu_{\text {yes }}$ and $\mu_{\text {no }}$ be supported on $f^{-1}(1)$ and $f^{-1}(0)$
- Rectangle $R$ is 1 -biased iff $\mu_{\text {yes }}(R) \geq 2 \cdot \mu_{\mathrm{no}}(R)$
$■ \operatorname{Corr}\left(f, \mu_{\text {yes }}, \mu_{\mathrm{no}}\right)=\max \Delta$ such that all 1-biased $R$ have size $\mu_{\text {yes }}(R) \leq 2^{-\Delta}$


## Results for SBP and USBP

## Theorem: $\quad \mathbf{S B P}(f)=\Theta(\operatorname{Corr}(f))$

## Corruption bound:

- Let $\mu_{\text {yes }}$ and $\mu_{\text {no }}$ be supported on $f^{-1}(1)$ and $f^{-1}(0)$
- Rectangle $R$ is 1 -biased iff $\mu_{\text {yes }}(R) \geq 2 \cdot \mu_{\mathrm{no}}(R)$
$■ \operatorname{Corr}\left(f, \mu_{\text {yes }}, \mu_{\mathrm{no}}\right)=\max \Delta$ such that all 1-biased $R$ have size $\mu_{\text {yes }}(R) \leq 2^{-\Delta}$
$■ \operatorname{Corr}(f)=\max _{\mu_{\mathrm{yes}}, \mu_{\mathrm{no}}} \operatorname{Corr}\left(f, \mu_{\mathrm{yes}}, \mu_{\mathrm{no}}\right)$


## Results for SBP and USBP

## Theorem: $\quad \mathbf{S B P}(f)=\Theta(\operatorname{Corr}(f))$

## Corollary: $\operatorname{SBP}($ Disj $)=\Omega(n)$

## Results for SBP and USBP

## Theorem: $\quad \mathbf{S B P}(f)=\Theta(\operatorname{Corr}(f))$

## Corollary: $\operatorname{SBP}($ Disj $)=\Omega(n)$

## Theorem: $\operatorname{USBP}(\operatorname{Disj})=\Omega(n)$

## Simple proof of main theorem

Proof of $\Omega(n \cdot(1-\beta / \alpha))$
1 Start with $(\alpha, \beta)$-protocol $\Pi$
2 And-amplify into $\left(\alpha^{k}, \beta^{k}\right)$-protocol $\Pi^{k}$
3 Then $\Pi^{k}$ is an SBP protocol for $k=(1-\beta / \alpha)^{-1}$
4 Hence $\left|\Pi^{k}\right| \geq \Omega(n)$
5 Hence $|\Pi| \geq \Omega(n / k)=\Omega(n \cdot(1-\beta / \alpha))$

## Simple proof of main theorem

## Proof of $\Omega(n \cdot(1-\beta / \alpha))$

1 Start with $(\alpha, \beta)$-protocol $\Pi$
2 And-amplify into $\left(\alpha^{k}, \beta^{k}\right)$-protocol $\Pi^{k}$
3 Then $\Pi^{k}$ is an SBP protocol for $k=(1-\beta / \alpha)^{-1}$
4 Hence $\left|\Pi^{k}\right| \geq \Omega(n)$
5 Hence $|\Pi| \geq \Omega(n / k)=\Omega(n \cdot(1-\beta / \alpha))$

Note: And-amplification for nonnegative rank
1 Start with nonnegative matrix $M$
2 Raise entries to power $k$
3 Basic fact: $\operatorname{rank}_{+}\left(M^{(k)}\right) \leq \operatorname{rank}_{+}(M)^{k}$

## Proof ideas (for experts)

## Theorem: $\quad \mathbf{S B P}(f)=\Theta(\operatorname{Corr}(f))$

- Analogous to [Klauck'07]
- Uses minimax


## Proof ideas (for experts)

## Theorem: $\quad \operatorname{USBP}(\operatorname{Disj})=\Omega(n)$

■ Information complexity framework [Bar-Yossef et al.'04]

- New challenge: Transcript useless $1-\alpha$ of the time Solution: Study transcripts conditioned on acceptance
- Cannot prove $\Omega(1)$ info lower bound for 2-bit NAND function!


## Proof ideas (for experts)

## Theorem: $\quad \operatorname{USBP}(\mathrm{Disj})=\Omega(n)$

- Information complexity framework [Bar-Yossef et al.'04]

■ New challenge: Transcript useless $1-\alpha$ of the time Solution: Study transcripts conditioned on acceptance

- Cannot prove $\Omega(1)$ info lower bound for 2-bit NAND function!



## Proof ideas (for experts)

## Theorem: $\quad \operatorname{USBP}(\mathrm{Disj})=\Omega(n)$

■ Information complexity framework [Bar-Yossef et al.'04]

- New challenge: Transcript useless $1-\alpha$ of the time Solution: Study transcripts conditioned on acceptance
- Cannot prove $\Omega(1)$ info lower bound for 2-bit NAND function! Solution: Use a different gadget


|  | 12 |  |
| :---: | :---: | :---: |
| 0 | 1 | 1 |
| 1 | 0 | 1 |
| 2 | 1 | 0 |

## Summary



## Future work:

- WIP: Separating MA and SBP ?

■ No ideas: Separating SBP and USBP ?
■ Long standing: Lower bounds for $\mathbf{A M}$ ?

## Summary



## Future work:

- WIP: Separating MA and SBP ?

■ No ideas: Separating SBP and USBP ?
■ Long standing: Lower bounds for $\mathbf{A M}$ ?

## Cheers!

