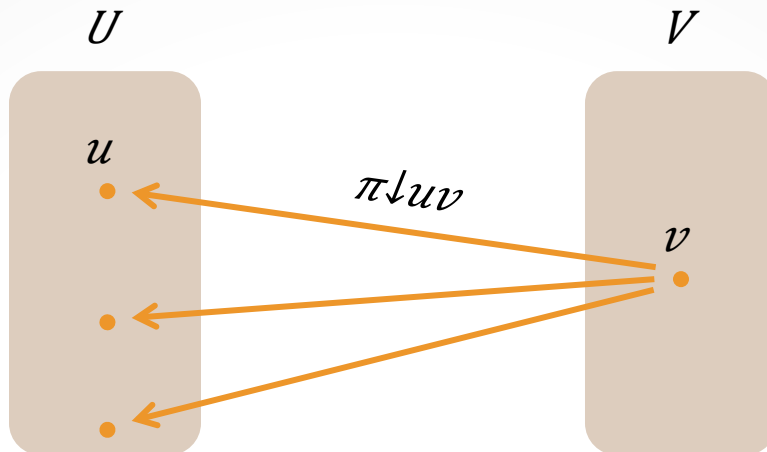


# Parallel Repetition and Direct Products

Irit Dinur  
Weizmann

• Joint work with David Steurer and Thomas Vidick •

constraint graph



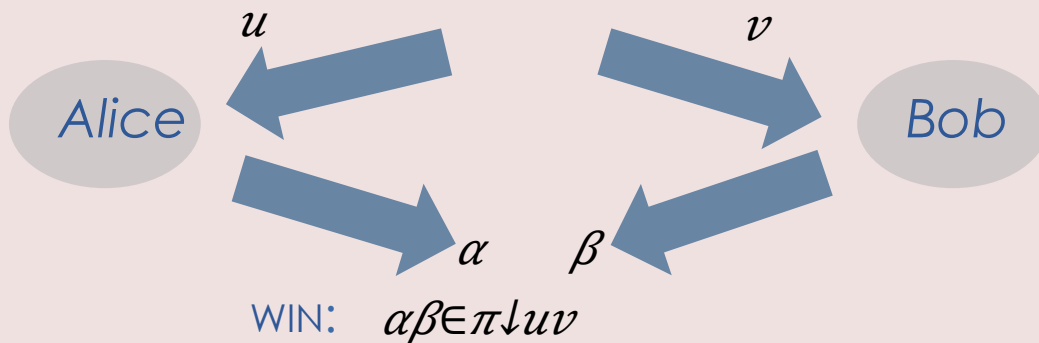
bipartite  
regular  
(for simplicity)

game  $G$

random:  
 $uv$

no communication  
between A & B

strategy  
 $g: U \rightarrow \Sigma$



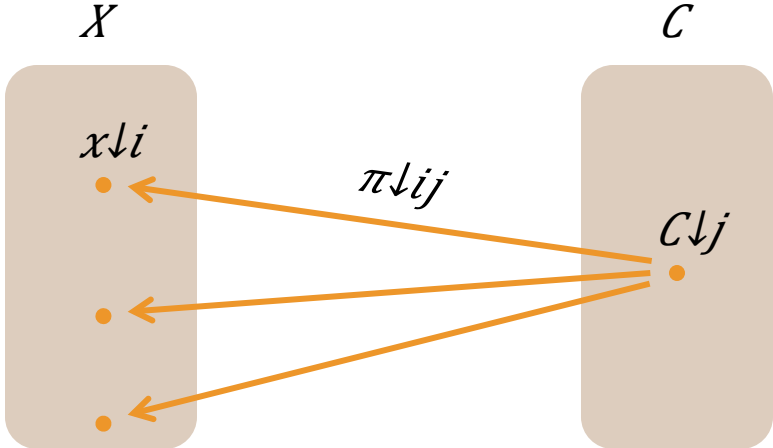
strategy  
 $f: V \rightarrow \Sigma$

$$\text{value}(G) := \max_{\tau, f, g} \mathbb{P} \downarrow uv \{ (g(u), f(v)) \in \pi \downarrow uv \}$$

# Example: the 3SAT game

Given a 3SAT formula with variables  $X = \{x_1, \dots, x_n\}$  and clauses  $C = \{C_1, \dots, C_m\}$

Clauses vs. variable constraint graph



**game  $G$**

random:  
 $ij$

strategy  
 $g: X \rightarrow \{0,1\}$

Alice

$x_i$

$C_j = \neg x_i \vee x_i' \vee x_i''$

Bob

strategy  
 $f: C \rightarrow \{1,2,\dots,7\}$

0 010



## Games and CSPs

Label Cover = the problem of finding the value of a given game  
( this is a 2-local Constraint Satisfaction Problem - CSP )

Every CSP gives rise to a clause vs. variable game  
(or to a multi-player game)

MIP = games

PCP = CSPs (fixed proof = assignment;  
randomized verification = clauses)

(classically equivalent, but the quantum analogs are not)

### COMPLEXITY OF LABEL-COVER

NP-hard to decide if  $val(G)=1$  or  $val(G)<1$  (Cook-Levin)

NP-hard to decide if  $val(G)=1$  or  $val(G)<1-\epsilon$  (PCP)

NP-hard to decide if  $val(G)=1$  or  $val(G)<\epsilon$  (strong PCP)

# The PCP Theorem [AS, ALMSS '91]

strong PCP theorem: “gap-label-cover is NP-hard”  
Proof: By reduction from tiny gap to constant gap,  
aka amplification

Start with label-cover  $G$  and end up with  $G'$ , s.t.

If $val(G) = 1$ then $val(G') = 1$	then $val(G' \uparrow \otimes k) = 1$
If $val(G) < 1 - \epsilon$	then $val(G' \uparrow \otimes k) < \epsilon$

How?

- by algebraic encoding [AS, ALMSS 1991]; or
- By “derandomized parallel repetition” (D. 2007)

# Repetition

**Sequential** repetition: run the game  $t$  rounds

Probability of not catching an error =  $\exp(-t)$

CSP perspective: Number of bits read from proof =  $O(t)$

Not a 2-local CSP (label-cover) anymore\*.  $2t$ -local.

Remark: derandomizing this is easy (e.g. via “expander walks”)

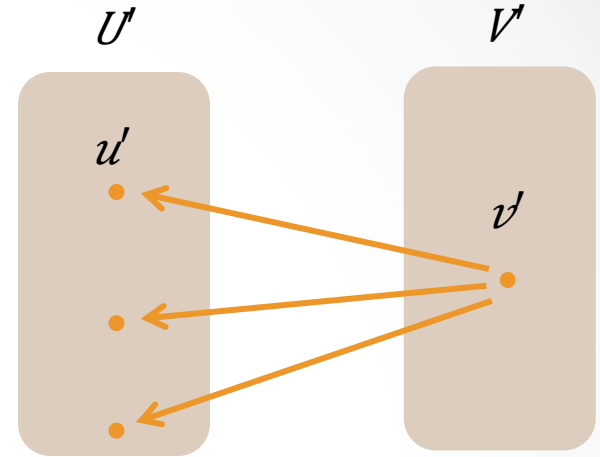
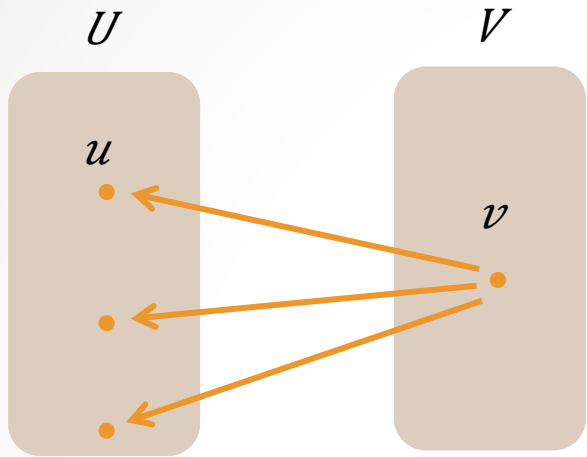
So we can plug in  $t = \log n$  and get the sequential analog of the sliding scale conjecture

Want: A 2-local CSP with the same properties

Why? Useful for inapprox, 2-local is simulatable by graph problems

Approach: **Parallel** Repetition

# Parallel Repetition



**G**

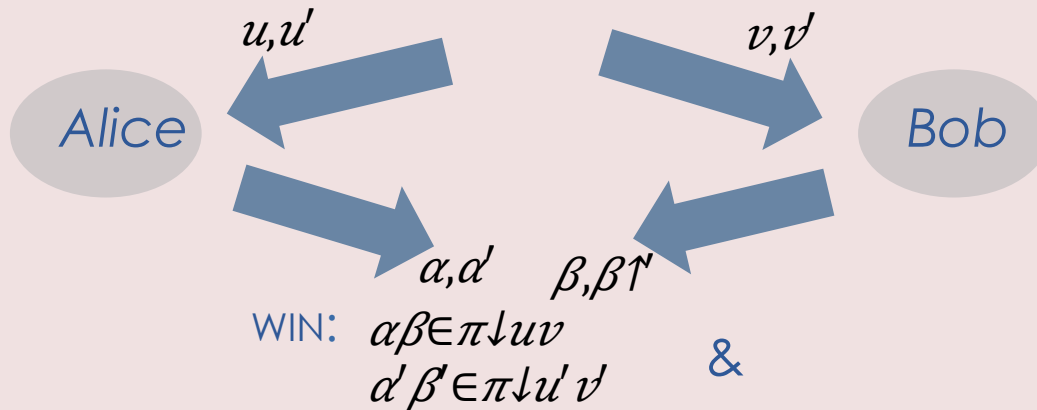
**G'**

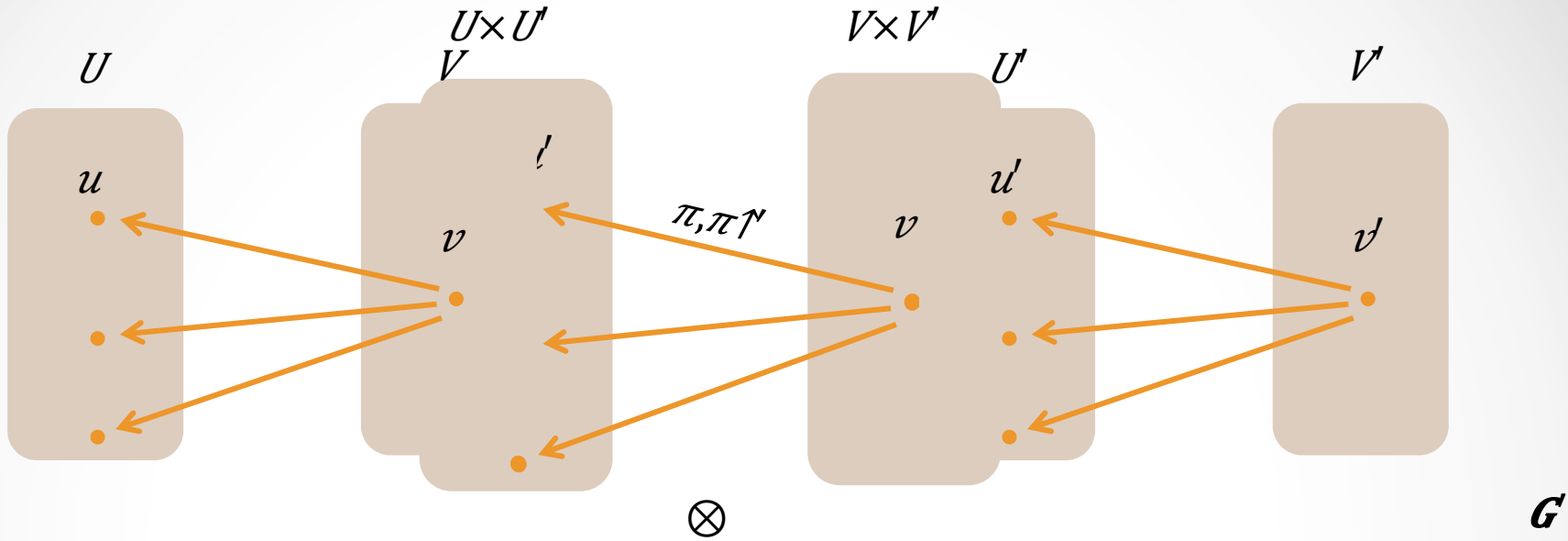
## Playing $G, G'$ in parallel

random:  
 $uv, u'v'$

no communication  
between A & B

strategy  
 $g: U \times U' \rightarrow \Sigma \times \Sigma'$



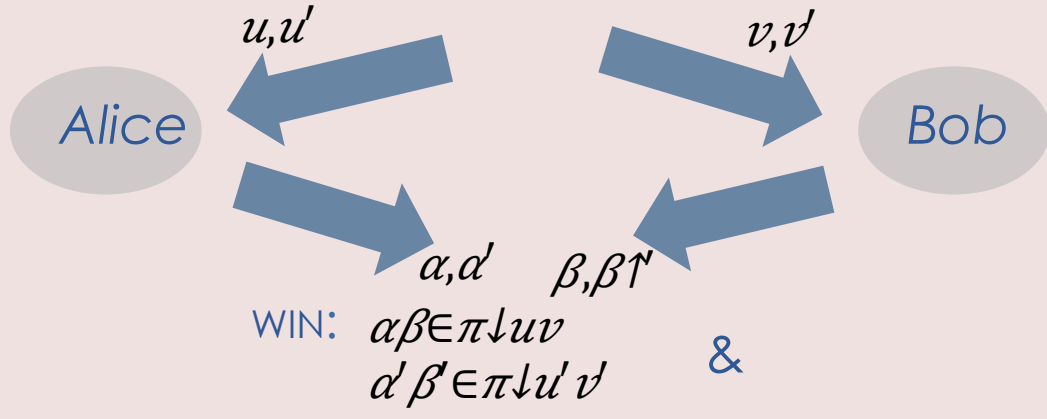


**game  $G \otimes G$**

random:  
 $uv, u'v'$

no communication  
between A & B

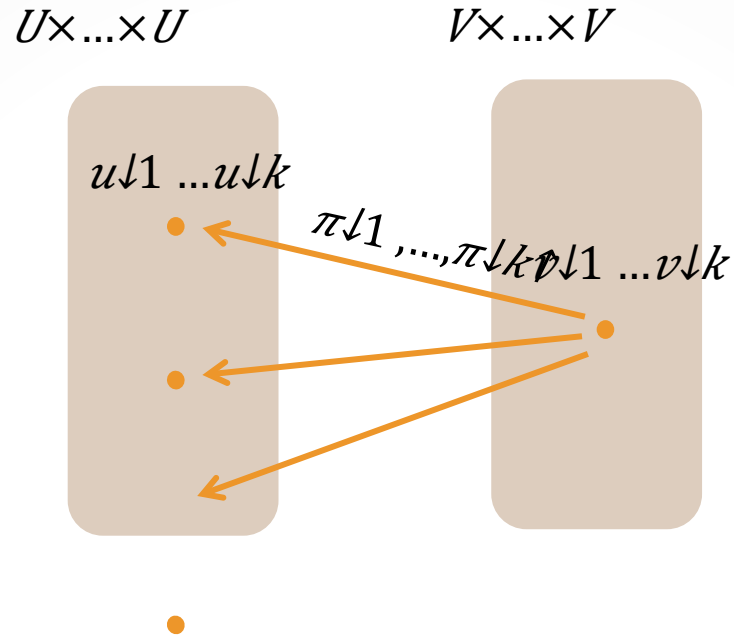
strategy  
 $g: U \times U' \rightarrow \Sigma \times \Sigma'$



strategy  
 $f: V \times V' \rightarrow \Sigma \times \Sigma'$

WIN:  $\alpha\beta \in \pi \downarrow uv$   
 $\alpha'\beta' \in \pi' \downarrow u'v'$  &





THIS IS A "PRODUCT" OPERATION ON GAMES

Product of Games:  $G \otimes H$

Parallel Repetition:  $G \uparrow k$

Q1: If  $value(G1) = \omega_1$  and  $value(G2) = \omega_2$

then what is  $value(G1 \otimes G2)$  ?

Q2: If  $value(G) = \alpha$ , then what is  $value(G \uparrow \otimes k)$  for  $k > 1$  ?

One obvious candidate is the direct product strategy.

But it is not, in general, the best strategy.

## previous bounds

(long history, notoriety)

[... Verbitsky'94, Feige-Kilian'94]

*parallel repetition theorem*

[Raz'95, improved: Holenstein'07, Rao'08]

(for projection games)

$$\text{value}(G) \leq 1 - \varepsilon \implies \text{value}(G^{\otimes k}) \leq (1 - \varepsilon^{32/C})^k$$

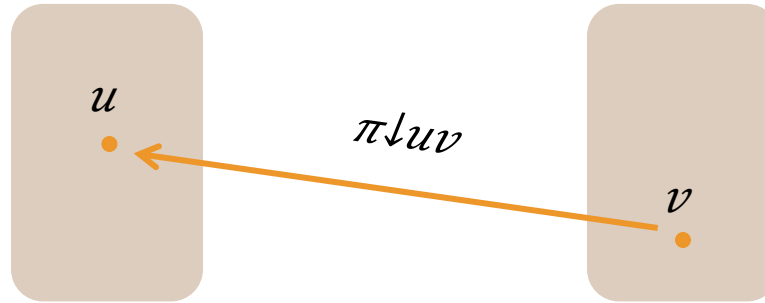
(tight even for games with XOR constraints) [Raz'08]

*main application: hardness amplification for LABEL COVER*

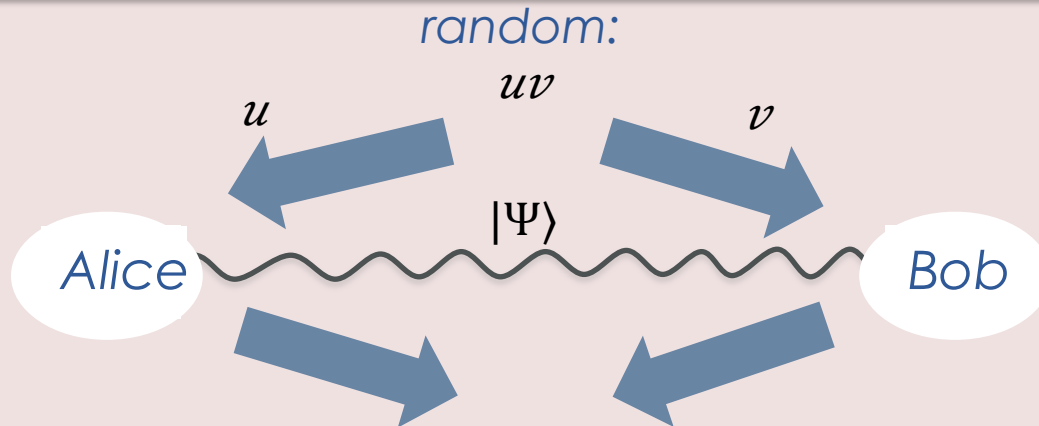
1 vs  $\delta$  approximation is NP-hard (basis of inapproximability results)

# Parallel repetition for entangled value

constraint graph



game  $G$



A & B also share an entangled state  $|\Psi\rangle$

Alice's strategy: For each  $u$ , Alice has a set of measurements  $\{A_u \uparrow a\} \uparrow a$  s.t.  $\sum a \uparrow \dots$

Bob's strategy: For each  $v$ , Bob has a set of measurements  $\{B_v \uparrow b\} \uparrow b$  s.t.  $\sum b \uparrow \dots$

## Parallel repetition for entangled value

[Holenstein '06] – parallel repetition for non-signalling strategies

[Cleve, Slofstra, Unger, Upadhyaya '08] – perfect parallel repetition for entangled value of XOR games

[Kempe Regev Toner '08] - unique games

[Kempe-Vidick '11] : polynomially decaying bounds for parallel repetition of entangled value

Extending the Feige-Kilian proof to the quantum setting

(adding consistency queries to force provers to play a direct product strategy)

[Chailloux, Scarpa '14] – parallel repetition for general games, bound depends on size of game

[Jain, Pereszlényi, Yao '14] – product-distribution games

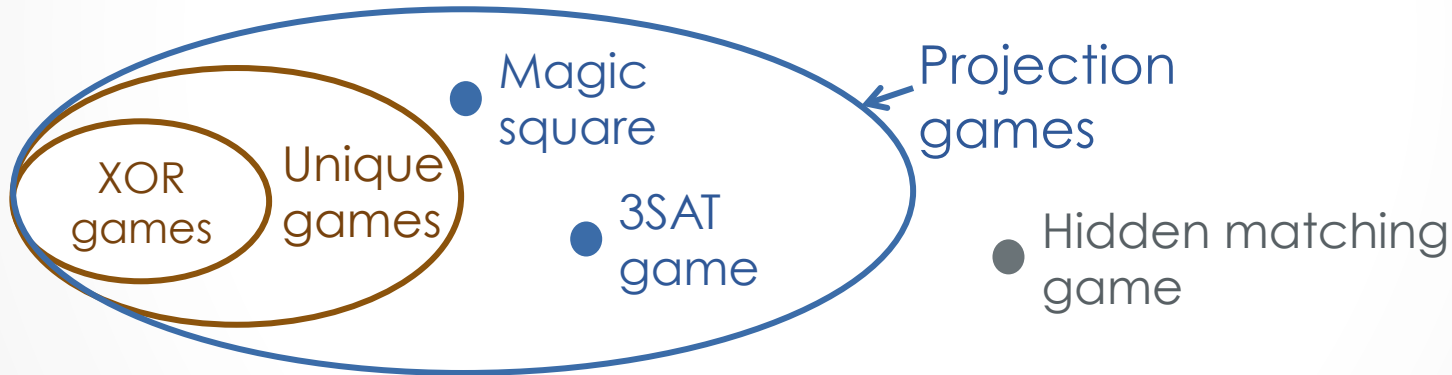
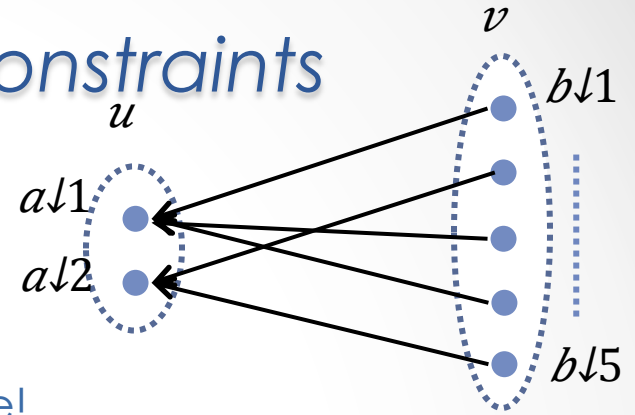
[D.-Steurer-Vidick, to appear in CCC'14]:

Let  $G$  be a projection-constraint game, then

If  $val^*(G) < 1 - \epsilon$  then  $val^*(G^{\otimes k}) < (1 - poly(\epsilon))^k$

# Games with projection constraints

- Projection game: for every pair of questions, any answer from B determines unique valid answer from A
- Your favorite two-player game is a projection game!



- Exists universal transformation  $G \rightarrow G'$  such that  $G'$  projection game and  $\omega(G) \approx \omega(G')$
- ...but could have  $\omega^*(G') \ll \omega^*(G)$

Recent but earlier work with David Steurer, (to appear in STOC '14) :

analytical framework to analyze parallel repetition  
for projection constraints  
(contrast to previous information-theoretic approach)

**new bounds**

low value:  $\text{value}(G) \leq \rho \Rightarrow \text{value}(G^{\uparrow k}) \leq (2\rho)^{\uparrow k/2}$   
(for projection constraints)

few repetitions:  $\text{value}(G) \leq 1 - \varepsilon \Rightarrow \text{value}(G^{\uparrow k}) \leq (1 - \varepsilon)^{\uparrow \sqrt{k}}$   
(for projection constraints,  $k \ll 1/\varepsilon^2$ )

optimal NP-hardness for SET COVER (and better NP-hardness for LABEL COVER)  
 $(1 - \varepsilon) \ln n$ -approximation, via [Feige, Moshkovitz–Raz, Moshkovitz]

Raz's parallel-repetition counterexample tight even for small  $k$

some  $G$  have  $\text{value} \leq 1 - \varepsilon$  but  $\text{value}(G^{\uparrow k}) \geq 1 - \varepsilon \sqrt{k}$   
(answers question of O'Donnell)

Analytical framework of DS extends nicely to entangled value setup

Plan:

Describe the “analytical / linear-algebraic” proof of [DS], show how to generalize to the entangled value setup



# V [DS] proof overview



- I. Analytical Setup View a projection game as a linear operator acting on (Bob)-assignments
- II. Prelim Step: move to “collision value”  $\|\cdot\|$  instead of  $\text{value}(G)$

$$\text{value}(G) \leq \|G\| \leq \text{value}(G) \uparrow 1/2 \quad \text{for all } G \quad (\text{easy})$$

- III. Main Step: further relax  $\|\cdot\|$  to  $\text{val}\downarrow+$  and prove

- 1.  $\text{val}\downarrow+(G \otimes H) = \text{val}\downarrow+(G) \cdot \text{val}\downarrow+(H)$  for all  $G, H$  (multiplicativity)
- 2.  $\text{val}\downarrow+(G) \approx \text{value}(G)$  for all  $G$  (approximation)

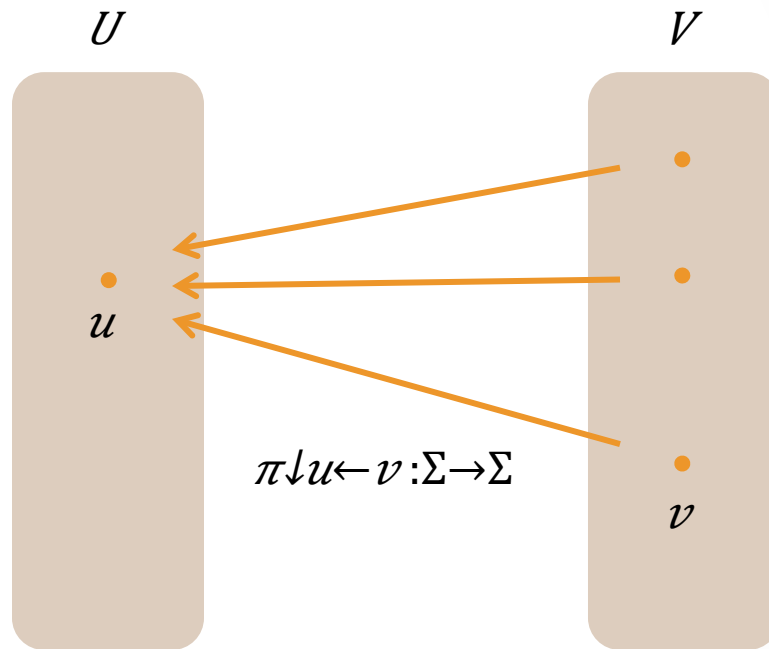
*proof of parallel-repetition bound*

$$\text{value}(G \uparrow k) \stackrel{\text{II}}{\leq} \|G \uparrow k\| \stackrel{\text{III}}{\leq} \text{val}\downarrow+(G \uparrow k) \stackrel{1.}{=} \text{val}\downarrow+(G) \uparrow k \stackrel{2.}{\approx} \text{value}(G) \uparrow k$$



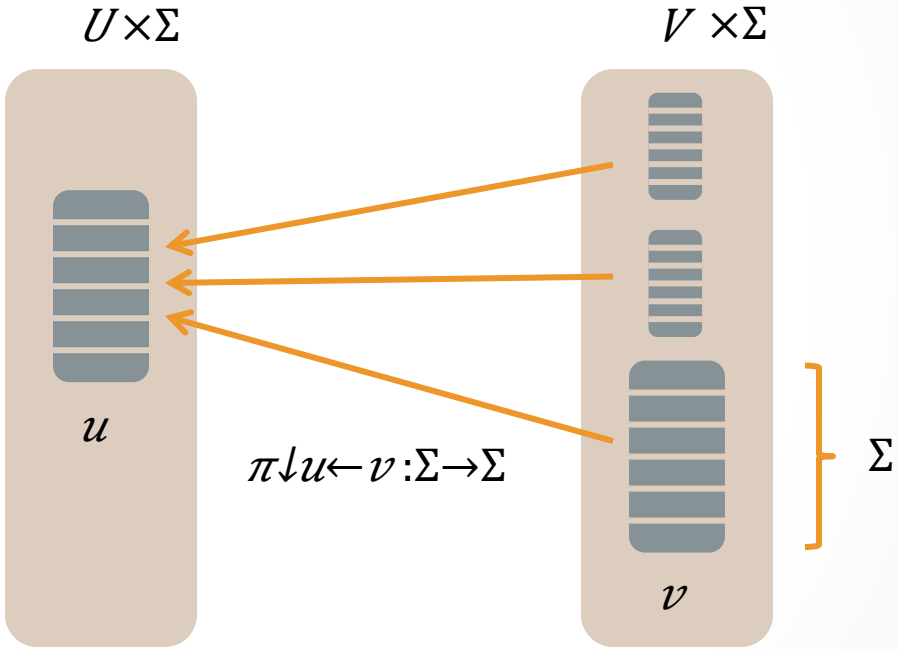
# analytical setup

constraint graph



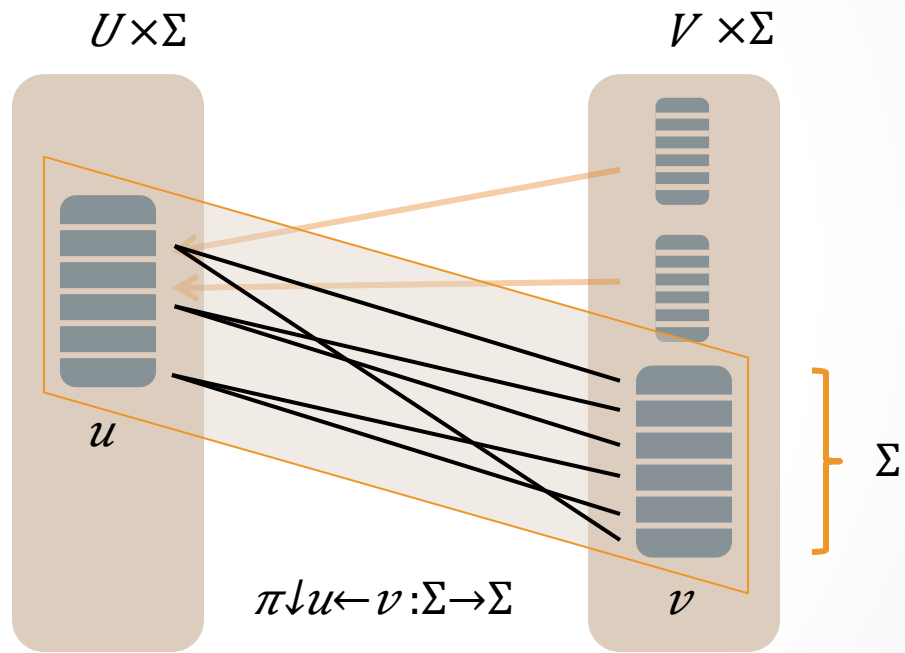
# analytical setup

constraint graph



# analytical setup

label-extended graph  
~~constraint graph~~



# analytical setup

label-extended graph  $G$

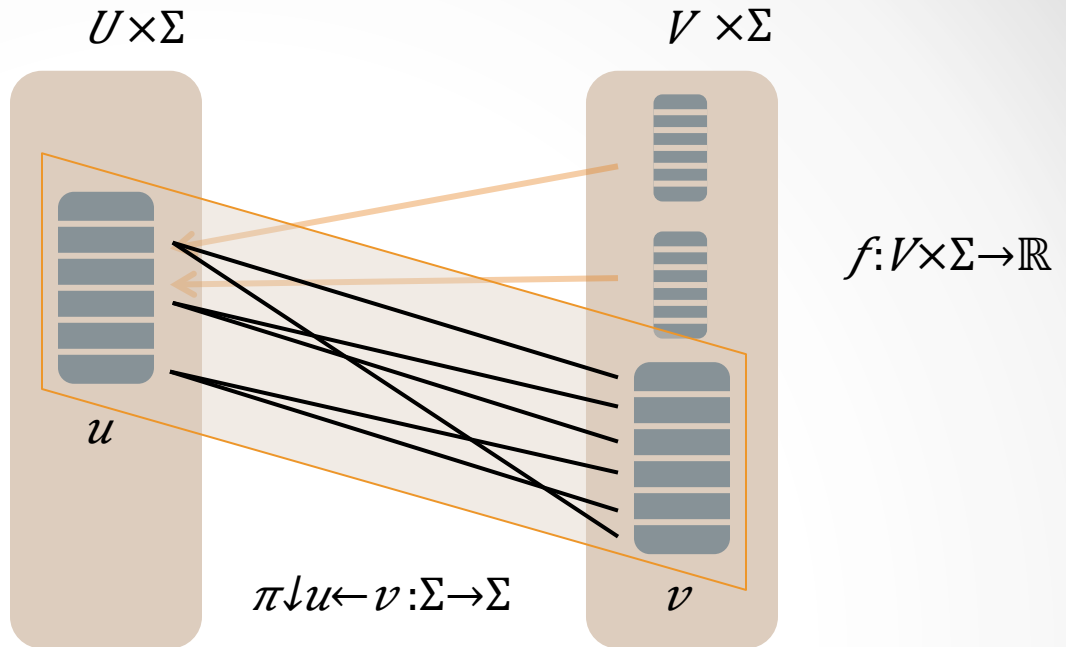
$$g: U \times \Sigma \rightarrow \mathbb{R}$$

$f: V \times \Sigma \rightarrow \mathbb{R}$  is an assignment if  $f \geq 0$  and  $\sum_{\beta \in \Sigma} f(v, \beta) = 1$  for all  $v \in V$

linear operator

= adjacency matrix of label-extended graph

bilinear form of  $G$



$$G: \mathbb{R}^{\uparrow V \times \Sigma} \rightarrow \mathbb{R}^{\uparrow U \times \Sigma}$$

For assignment  $f$ ,  
 $Gf(u, \alpha) = \text{prob. that random } u\text{-neighbor "demands" } \alpha$

$$Gf(u, \alpha) := \mathbb{E} \downarrow v: u \leftarrow v \sum_{\beta \rightarrow \alpha} f(v, \beta)$$

success probability for assignments  $f, g$

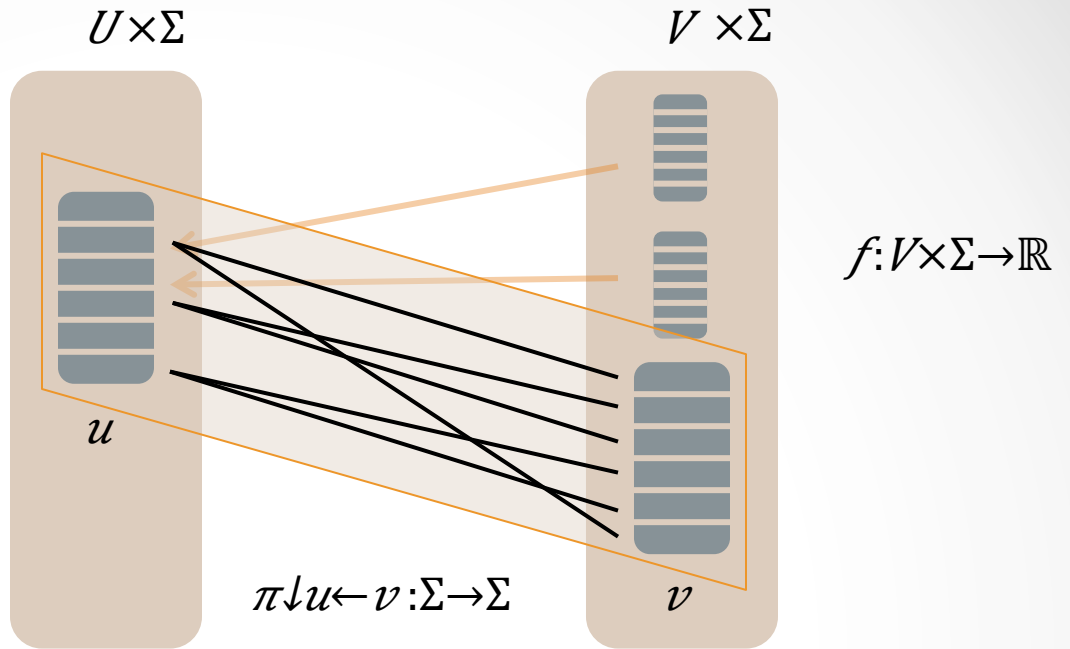
$$\langle Gf, g \rangle := \mathbb{E} \downarrow u \sum_{\alpha} Gf(u, \alpha) \cdot g(u, \alpha)$$

$$\text{value}(G) = \max \langle Gf, g \rangle \text{ over assignments } f, g$$

# analytical setup

label-extended graph  $G$

$$g: U \times \Sigma \rightarrow \mathbb{R}$$



$$G: \mathbb{R} \uparrow V \times \Sigma \rightarrow \mathbb{R} \uparrow U \times \Sigma$$

$$H: \mathbb{R} \uparrow V' \times \Sigma' \rightarrow \mathbb{R} \uparrow U' \times \Sigma'$$

tensor product

$$G \otimes H: \mathbb{R} \uparrow V \times V' \times \Sigma \times \Sigma' \rightarrow \mathbb{R} \uparrow U \times U' \times \Sigma \times \Sigma'$$

= parallel repetition

$$(G \otimes H) f(u, u', \alpha, \alpha') := \mathbb{E} \downarrow \blacksquare v \leftarrow u v' \leftarrow u' \quad \Sigma \blacksquare \beta \rightarrow \alpha \beta' \rightarrow \alpha'$$

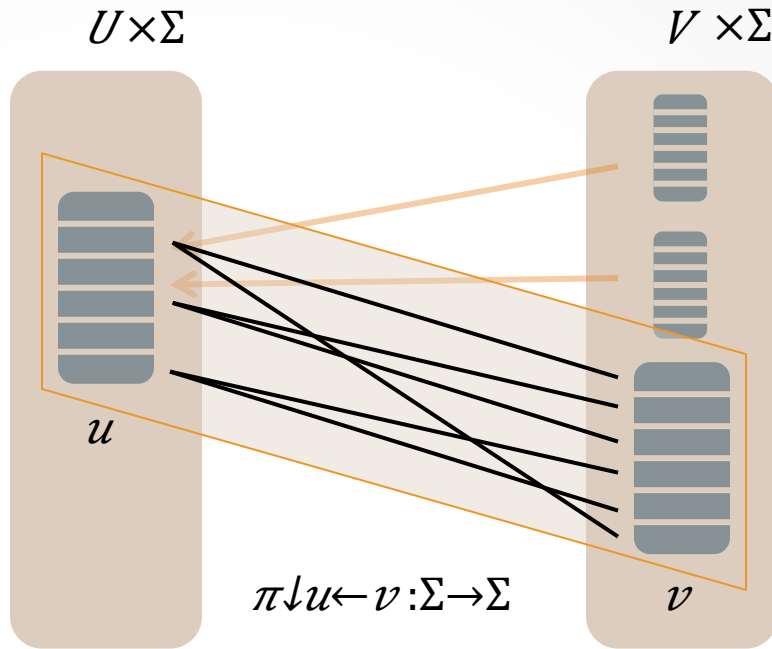
$$\uparrow \blacksquare f(v, v', \beta, \beta')$$

# analytical setup\*

label-extended graph  $G$

~~$$g: U \times \Sigma \rightarrow \mathbb{R}$$~~

$$g: U \times \Sigma \rightarrow \mathcal{L}(\mathbb{C} \uparrow d)$$



~~$$f: V \times \Sigma \rightarrow \mathbb{R}$$~~

$$f: V \times \Sigma \rightarrow \mathcal{L}(\mathbb{C} \uparrow d)$$

$$\sum b \uparrow \# f(v, b) = Id$$

$$G: \mathcal{L}(\mathbb{C} \uparrow d) \uparrow V \times \Sigma \rightarrow \mathcal{L}(\mathbb{C} \uparrow d) \uparrow U \times \Sigma$$

$$Gf(u, \alpha) := \mathbb{E} \downarrow v. u \leftarrow v \sum \beta \rightarrow \alpha \uparrow \# f(v, \beta)$$

$val \uparrow^* (G) = \sup_{(g, G) \downarrow |\Psi\rangle} \langle \Psi | g(u, \alpha) \otimes Gf(u, \alpha) | \Psi \rangle$

$$:= \mathbb{E} \downarrow u \sum \alpha \uparrow \# \langle \Psi | g(u, \alpha) \otimes Gf(u, \alpha) | \Psi \rangle$$

$$:= \mathbb{E} \downarrow u \sim v \sum \alpha \sim \beta \uparrow \# \langle \Psi | g(u, \alpha) \otimes f(v, \beta) | \Psi \rangle$$

## Collision value

II. Prelim Step: move to “collision value” instead of  $value(G)$

$$value(G) = \sup_{\tau f, g} \langle Gf, g \rangle \approx \sup_{\tau f} \langle Gf, Gf \rangle^{\uparrow} =: col-val(G)$$



Simple Cauchy-Schwarz

$$value^{\uparrow*}(G) = \sup_{\tau f, g, \Psi} \langle Gf, g \rangle_{\Psi} \approx \sup_{\tau f, \Psi} \langle Gf, Gf \rangle_{\Psi}^{\uparrow} =: col-val^{\uparrow*}(G)$$



Generalized Cauchy-Schwarz due to Haagerup



## Vector relaxation – $val_{\downarrow+}(G)$

III. Define  $val_{\downarrow+}(G)$  by replacing  $f(v,b)$  by a vector  $f(v,b,\omega)$

$$val_{\downarrow+}(G) = \sup_{\tau, f \geq 0} \langle Gf, Gf \rangle^{1/2} = \sup_{\tau, f} \mathbb{E}_{\downarrow u} \sum_{a \uparrow} \mathbb{E}_{\downarrow \omega} Gf(u, a, \omega)^2$$

Where the sup is over vector strategies  $f$ , with proper normalization

$$\forall v, \|\sum_{b \uparrow} f(v, b)\|_2 \leq 1$$

Why vector strategies ?

A strategy for  $G \otimes H$  is automatically a vector strategy, if viewed as a strategy for  $G$  alone.

The normalization “eliminates” the effects of the game  $H$

This facilitates proving  $val_{\downarrow+}(G \otimes H) = val_{\downarrow+}(G) \cdot val_{\downarrow+}(H)$

## Vector relaxation – $val_{\downarrow+}(G)$

III. Define  $val_{\downarrow+}^*(G)$  by replacing  $f(v,b)$  by a vector  $f(v,b,\omega)$

$$val_{\downarrow+}^*(G) \uparrow = \sup_{\tau, f, \Psi} \langle Gf, Gf \rangle_{\Psi} \uparrow = \sup_{\tau, f, \Psi} \mathbb{E}_{\downarrow u} \sum_{a \uparrow} \mathbb{E}_{\downarrow \omega} \langle \Psi | Gf(u, a, \omega) \otimes Gf(u, a, \omega) \rangle$$



Where the sup is over vector strategies  $f$ , with proper normalization

$$\sup_{\tau, \Psi} \sum_{b, b'} \mathbb{E}_{\downarrow \omega} \langle \Psi | f(v, b, \omega) \otimes f(v, b', \omega) | \Psi \rangle \leq 1 \uparrow$$

Why vector strategies ?

A strategy for  $G \otimes H$  is automatically a vector strategy, if viewed as a strategy for  $G$  alone.

The normalization “eliminates” the effects of the game  $H$

This facilitates proving  $val_{\downarrow+}(G \otimes H) = val_{\downarrow+}(G) \cdot val_{\downarrow+}(H)$

## Vector relaxation – $\text{val}_{\downarrow+}(G)$ –approximation

### III. Remaining goal: prove that $\text{val}_{\downarrow+}(G) \approx \text{val}(G)$

Given a vector strategy, derive a standard strategy that has similar value

Naïvely: focus on one coordinate  $\omega$  and hope for the best

Generally: combine different coordinates through correlated sampling



## [DSV] proof overview

I. Analytical Setup View a projection game as a linear operator acting on (Bob)-assignments

II. Prelim Step: move to “collision value”  $\|\cdot\|_*$  instead of  $\text{value}(G)$

$$\text{value}_*(G) \leq \|G\|_* \leq \text{value}(G) \cdot \frac{1}{2} \quad \text{for all } G \quad (\text{easy})$$

III. Main Step: further relax  $\|\cdot\|_*$  to  $\text{val}_{\downarrow+}_*$  and prove

1.  $\text{val}_{\downarrow+}_*(G \otimes H) = \text{val}_{\downarrow+}_*(G) \cdot \text{val}_{\downarrow+}_*(H)$  for all  $G, H$  (multiplicative)

2.  $\text{val}_{\downarrow+}_*(G) \approx \text{value}_*(G)$  for all  $G$  (approximate)

*proof of parallel-repetition bound*

$$\text{value}_*(G \uparrow \otimes k) \stackrel{\text{II}}{\leq} \|G \uparrow \otimes k\|_* \stackrel{\text{III}}{\leq} \text{val}_{\downarrow+}_*(G \uparrow \otimes k) \stackrel{1.}{=} \text{val}_{\downarrow+}_*(G) \uparrow k \stackrel{2.}{\approx} \text{value}_*(G) \uparrow k$$

Main source of trouble: non-product strategies that are too good

Why are they too good?

How good can they be ?

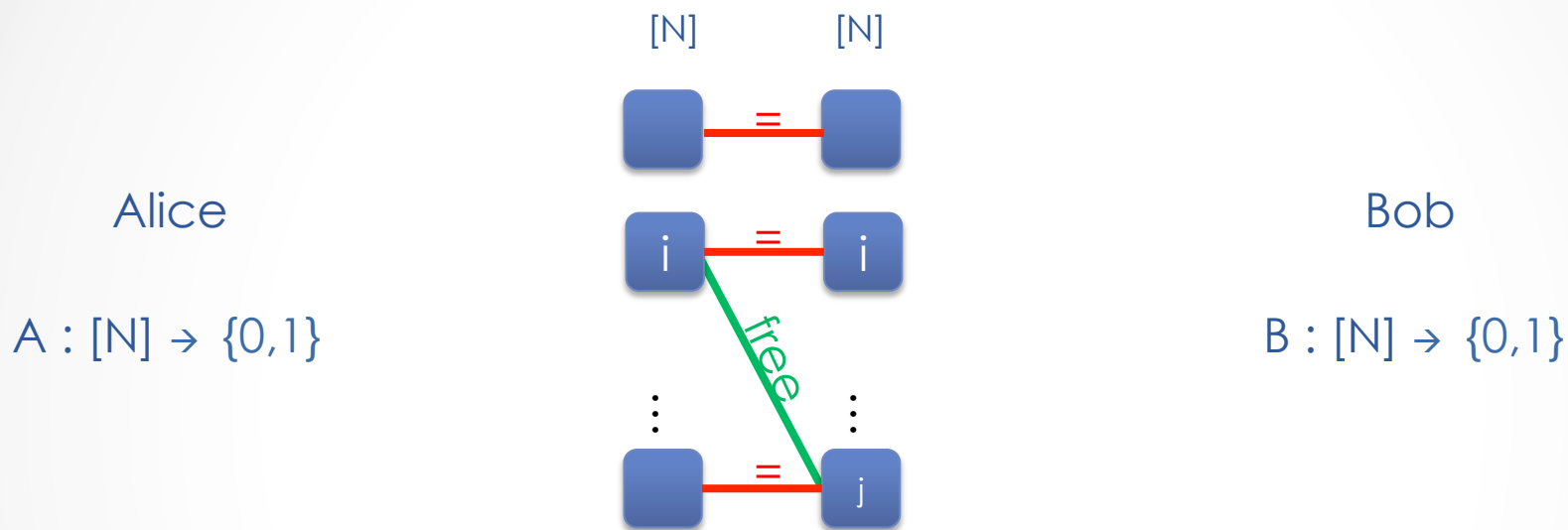
How does this depend on the game ?

Back to the basics – design a simple game in which this can be studied

THE CONFUSE & COMPARE GAME

\*[Feige & Kilian]

The Confuse & Compare game, with parameters  $p$ ,  $N$

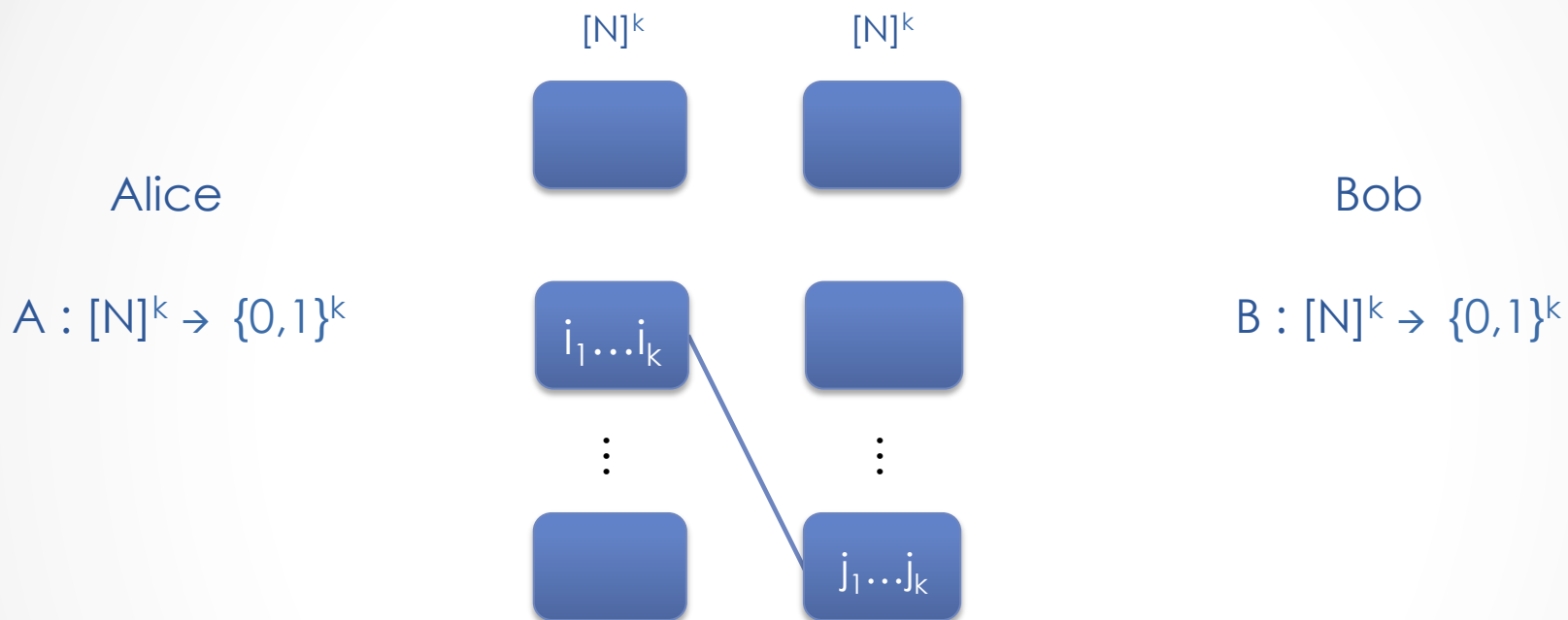


$p$  fraction of edges are equality edges  $(i,i)$  "compare"  
 $(1-p)$  fraction of edges are free edges  $(i,j)$  "confuse"

If  $A = B$  then win with probability 1

$k$ -fold parallel repetition: choose  $k$  independent edges  $i_1 j_1 \dots i_k j_k$   
 Send  $i_1 \dots i_k$  to Alice and  $j_1 \dots j_k$  to Bob

# The Confuse & Compare\* game, k-fold direct product



Clearly, a product strategy  $A^k$  obtains value 1

Question : can something be said about  $A$  if  $\text{val} > 0.99$  ?  $\text{val} > 0.001$  ?

# Direct Product Testing

The symmetrized version of this game is known as “direct product testing”. Both players use a strategy

$$A:[N]^k \rightarrow [M]^k$$

and the goal is to prove

THM: If Confuse-Compare-value (  $A$  )  $> \alpha$ ,  
then  $A$  is close to being direct-product

For some values of  $p$ ,  $N$ ,  $M$  this question is solved,  
[Goldreich-Safra, D.-Reingold, D.-Goldenberg, Impagliazzo-Kabanets-Wigderson, D.-Steurer]

but much is open

This is a clean question, can also be formulated in the entangled setting, but little is known there...



# Summary

- Parallel repetition is good for amplification
- Analytical approach useful for analyzing parallel repeated projection games
- Direct product testing