

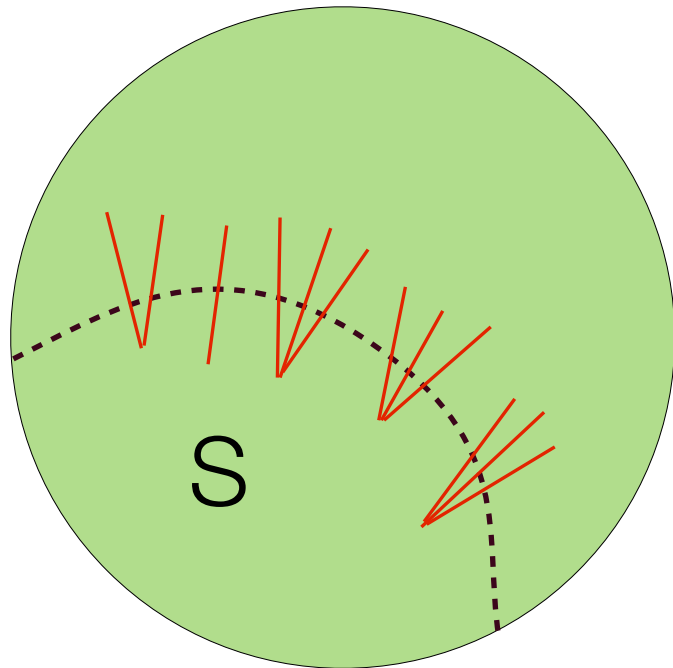
Reductions Between Expansion Problems

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MSR New England

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Princeton University

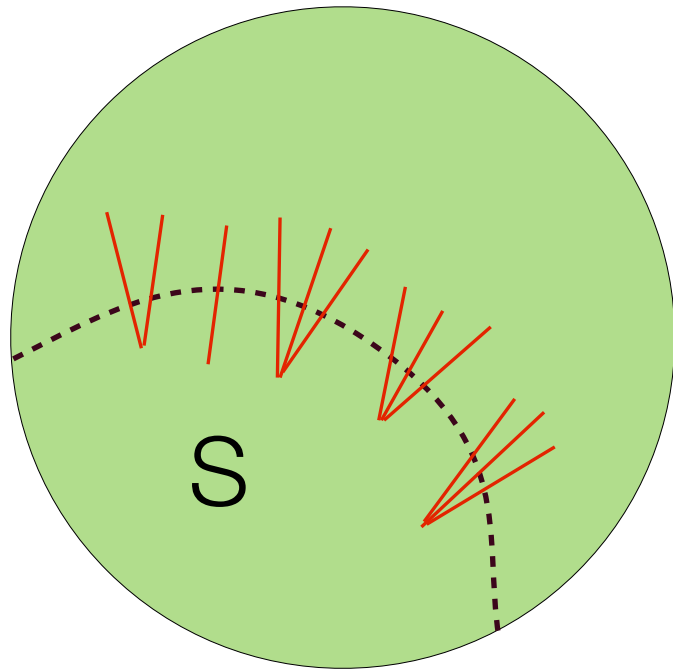
Graph Expansion



$G = (V, E)$

d-regular

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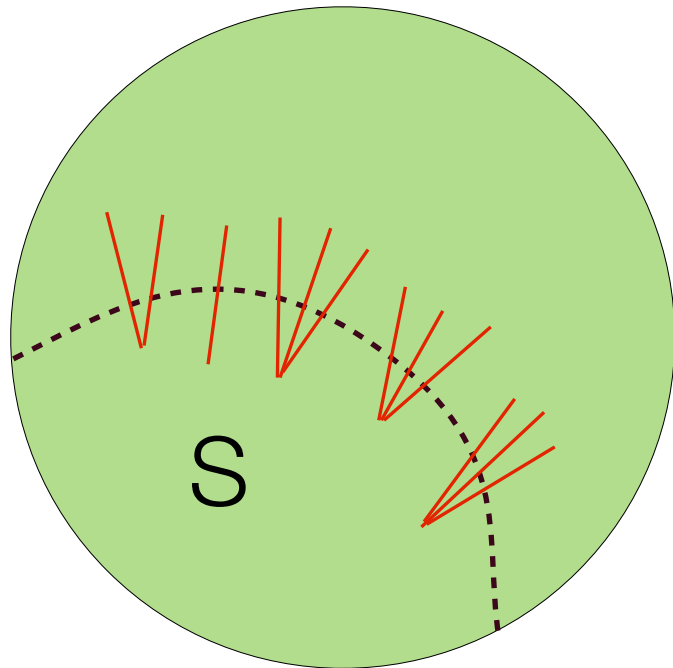
d-regular

$$\Phi_G(S) = \frac{E(S, V \setminus S)}{d|S|}$$

$$= \frac{\mathbf{P}_{(x,y) \in E} (x \in S, y \notin S)}{\mathbf{P}_{(x,y) \in E} (x \in S)}$$

$$\Phi_G = \min_{|S| \leq n/2} \{\Phi_G(S)\}$$

Graph Expansion



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d-regular

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- Expansion measures the probability of a random edge crossing a set S .
- Approximating the expansion of a graph is important for algorithms and also a fundamental problem in complexity.

Expansion at different scales

Expansion at different scales

- Define the **measure** of a set by the fraction of edges landing in it.

$$\begin{aligned}\mu(S) &= \mathbf{P}_{(x,y) \in E} (x \in S) \\ &= |S|/n \text{ (for regular graphs)}\end{aligned}$$

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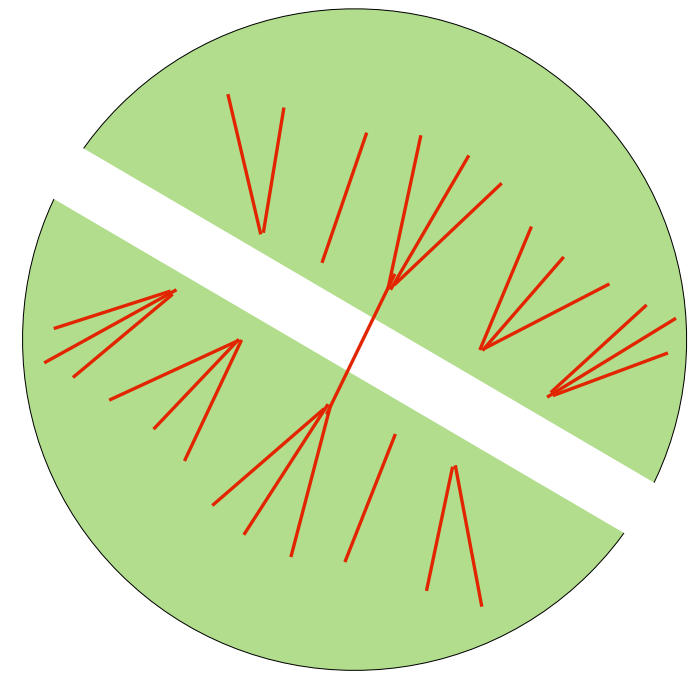
- $\Phi_G(\delta) = \min_{\mu(S) = \delta} \{\Phi_G(S)\}$
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Expansion at different scales

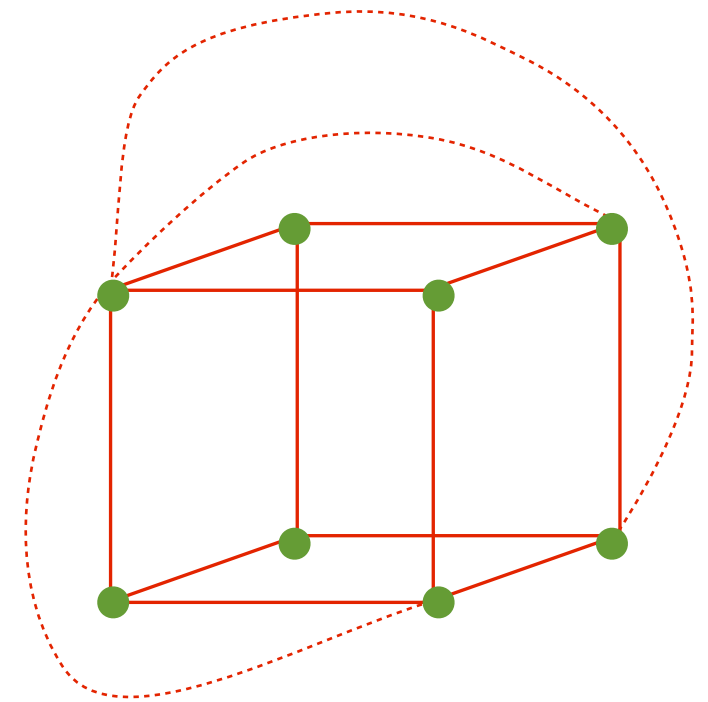
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- $\Phi_G(\delta) = \min_{\mu(S) = \delta} \{\Phi_G(S)\}$
- Φ_G measures the minimum expansion over all scales.
- The expansion profile of a graph may look very different at different scales.

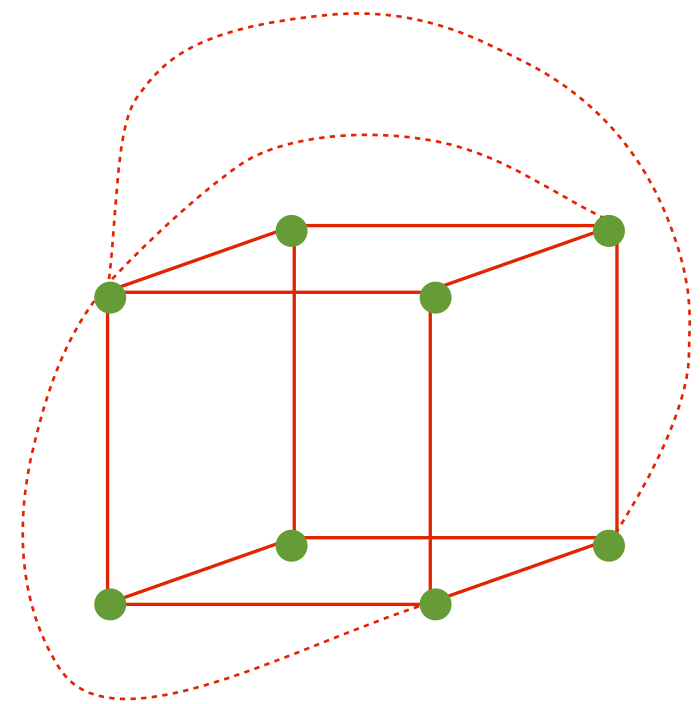


An example: the noise graph on $\{0, 1\}^n$



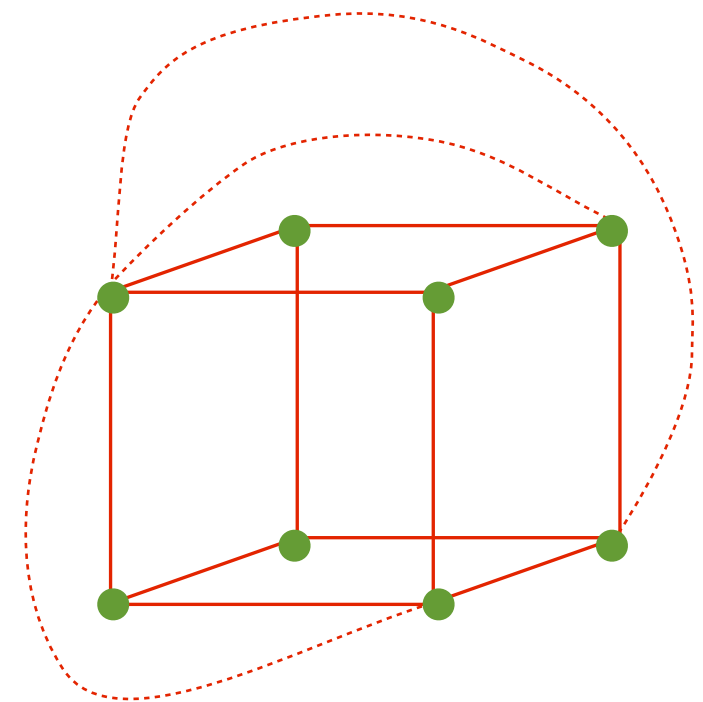
An example: the noise graph on $\{0,1\}^n$

- Connect $x, y \in \{0,1\}^n$ with weight $\epsilon^{H(x,y)} (1-\epsilon)^{n-H(x,y)}$



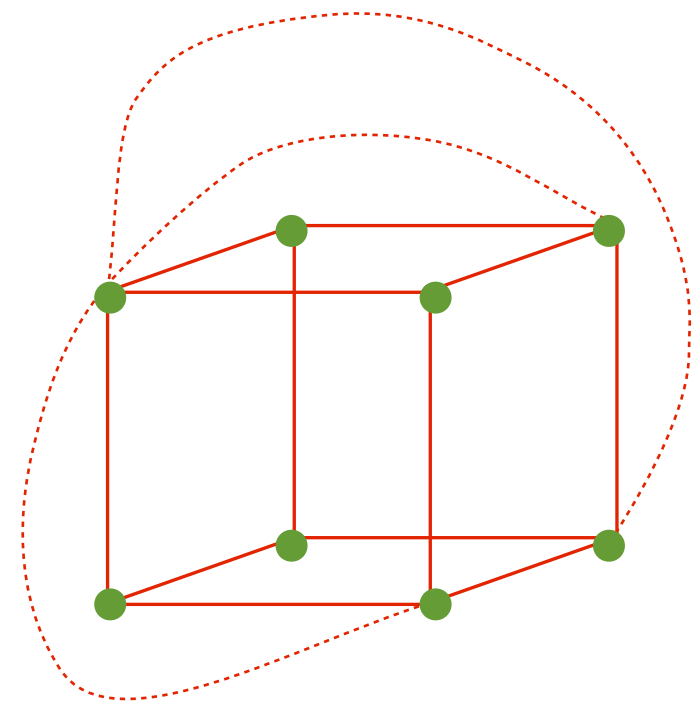
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- Connect $x, y \in \{0,1\}^n$ with weight $\epsilon^{H(x,y)} (1-\epsilon)^{n-H(x,y)}$
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 $\Phi_G(S) = \epsilon$

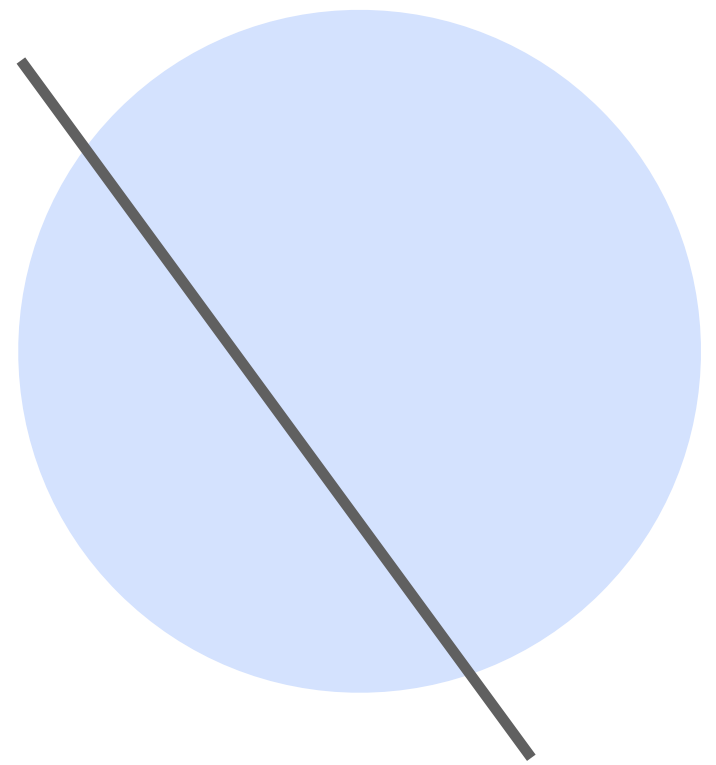


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 $\Phi_G(S) = \epsilon$
- Best cut with $\mu(S) = \delta$ is subcube of dimension $n - \log(1/\delta)$.
 $\Phi_G(S) \approx \epsilon \log(1/\delta)$

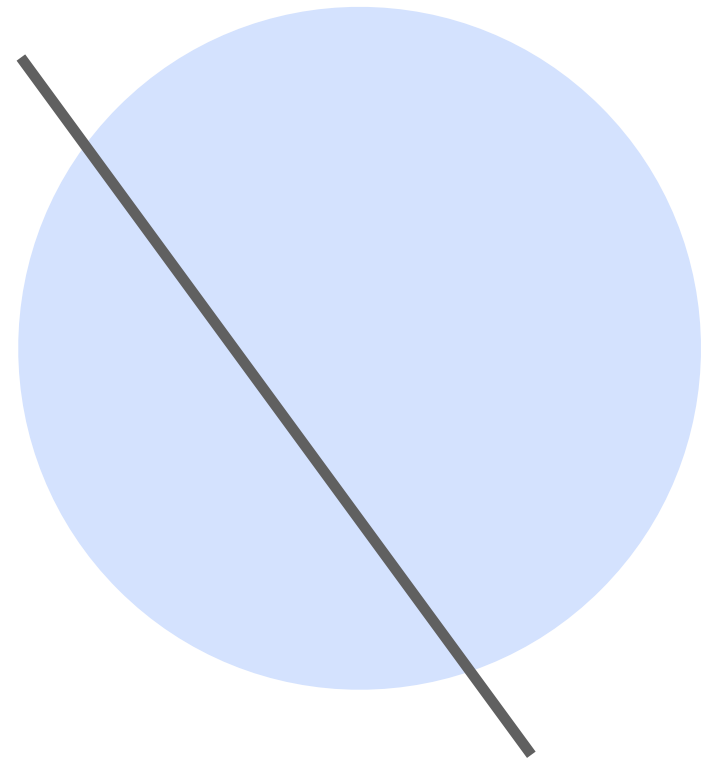


Cuts not aligned with any direction



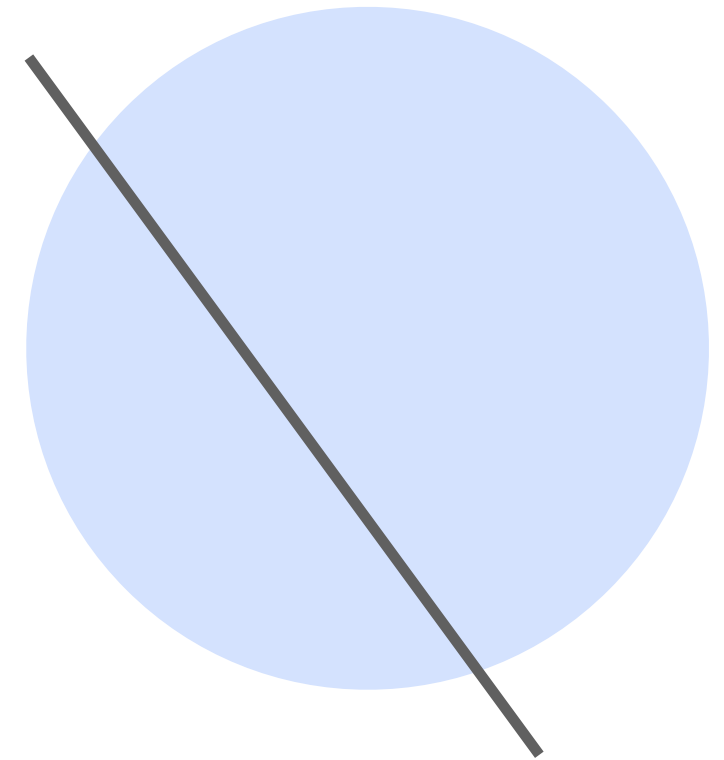
Cuts not aligned with any direction

- Cuts “far from direction cuts” are like cuts on the sphere.



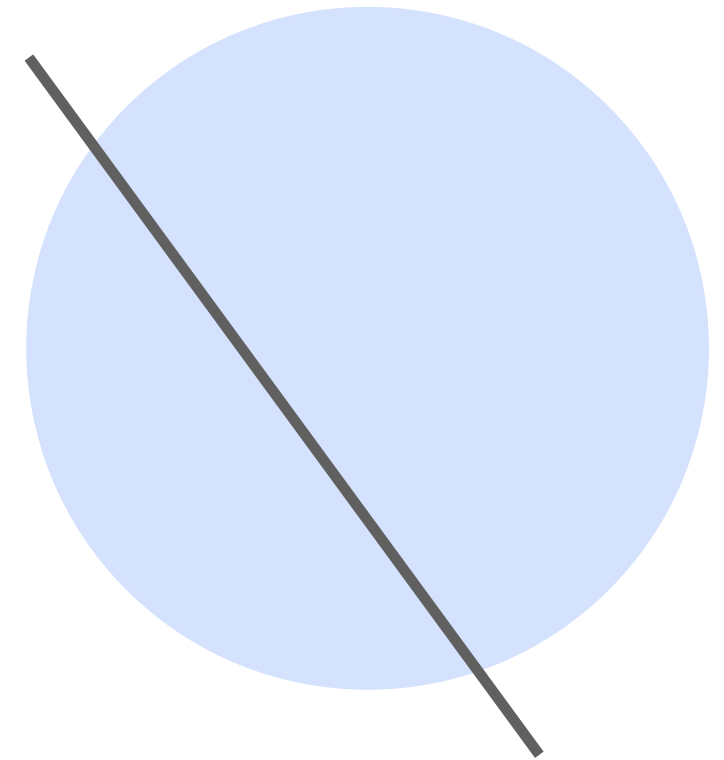
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- Best cut with $\mu(S) = 1/2$ is majority [MOO 05]
 $\Phi_G(S) = \Omega(\epsilon^{1/2})$
- Best cut with $\mu(S) = \delta$ is threshold function with mean δ .
 $\Phi_G(S) \approx 1 - (1/\delta) \Gamma_{1-2\epsilon}(\delta) \approx 1 - \delta^{\epsilon/(1-\epsilon)}$



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- Showed a reduction **to** Unique Games.

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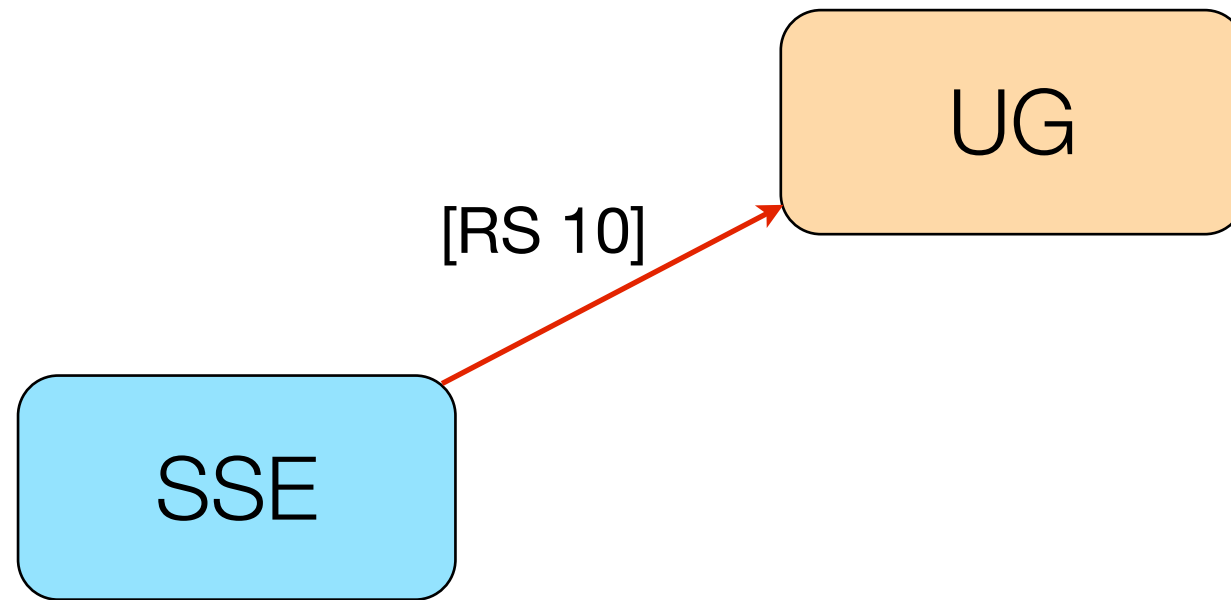
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- SSE(η, δ): Given graph G , distinguish between the following cases:
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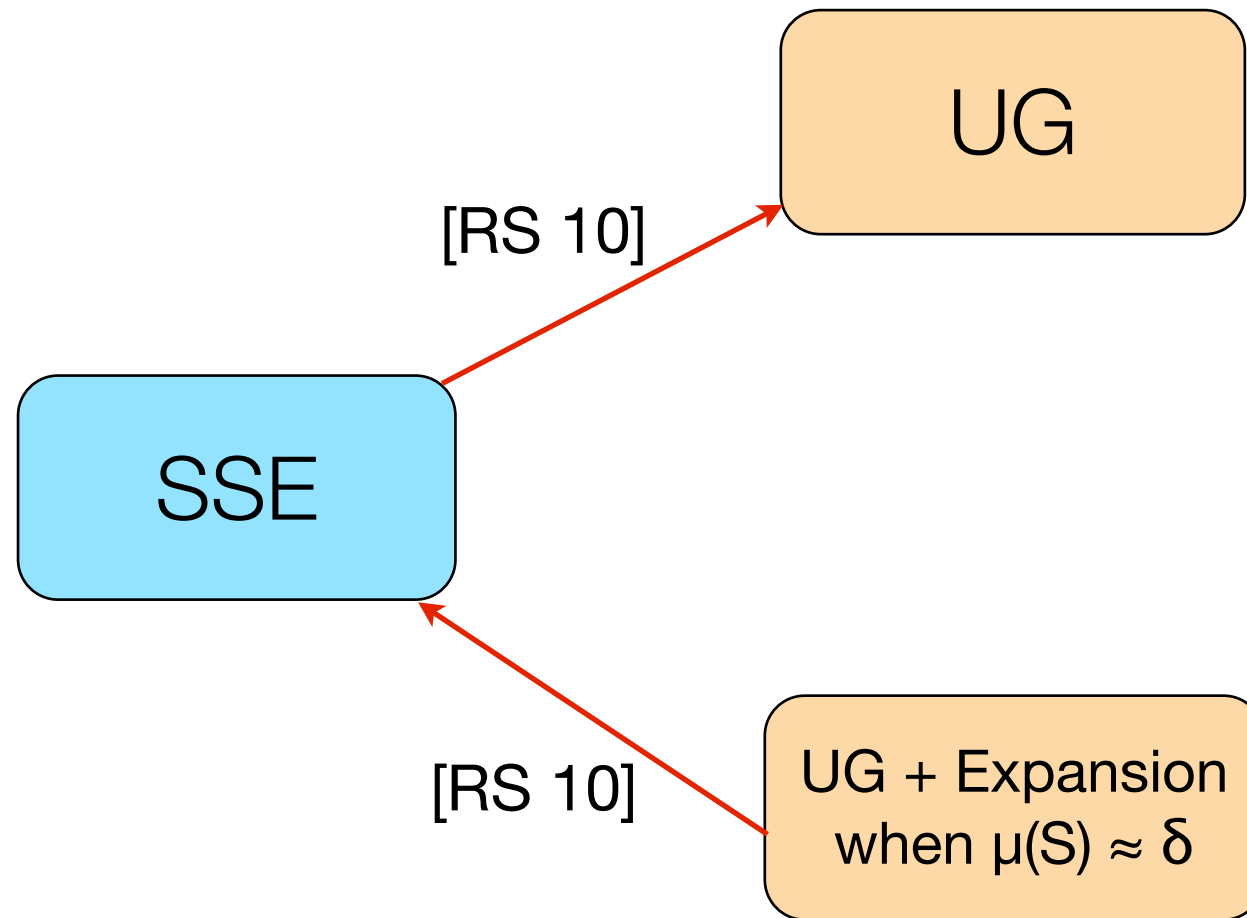
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- SSE-Conjecture: $\forall \eta \exists \delta$ such that SSE(η, δ) is hard.

The big-picture

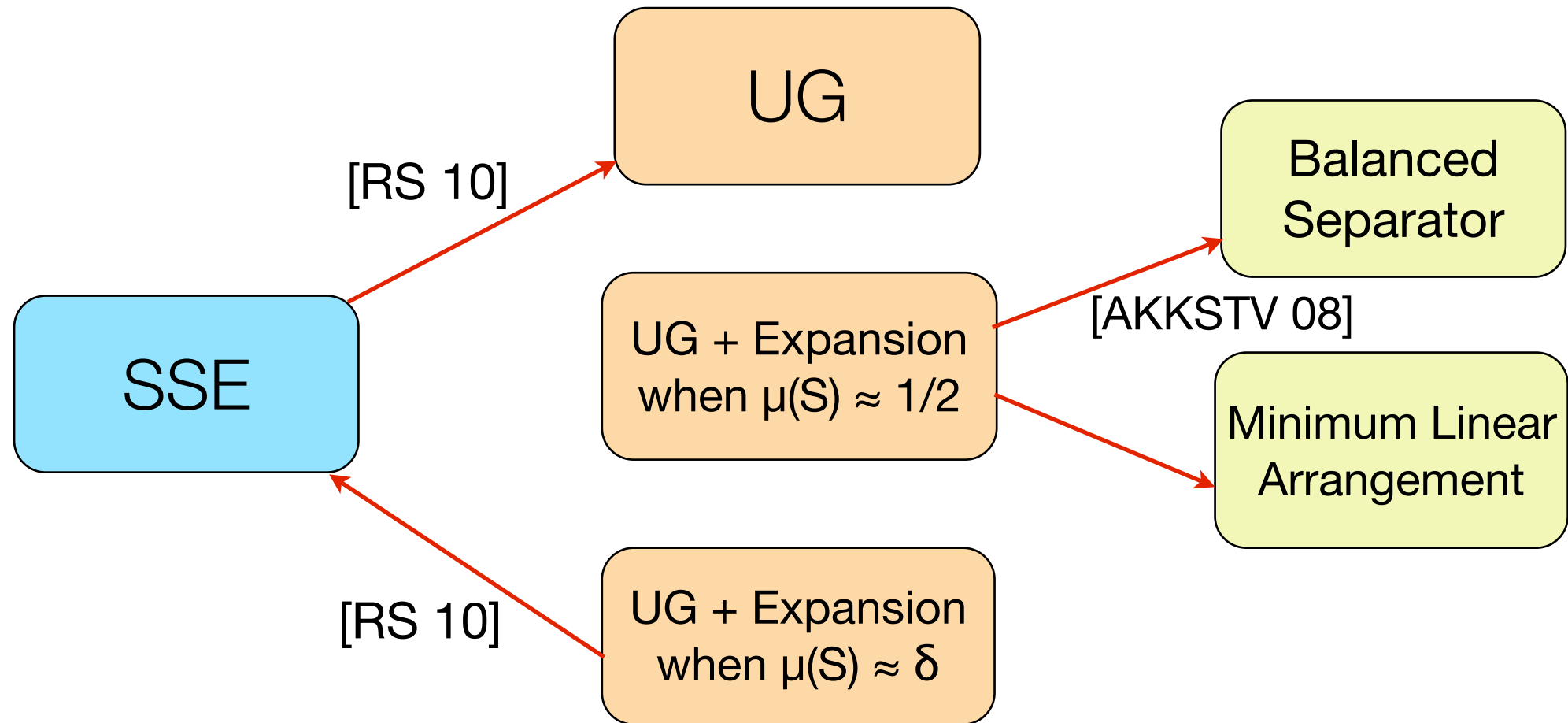
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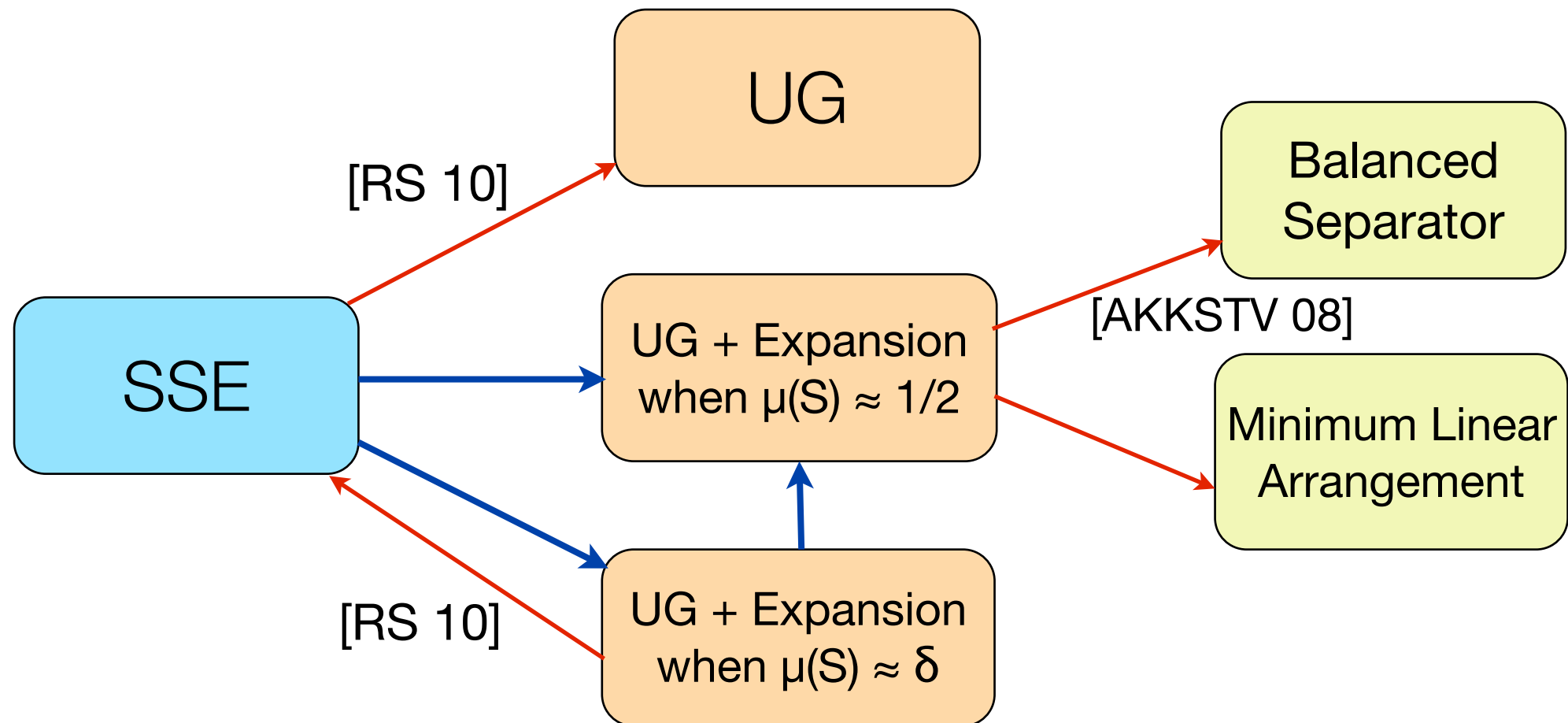
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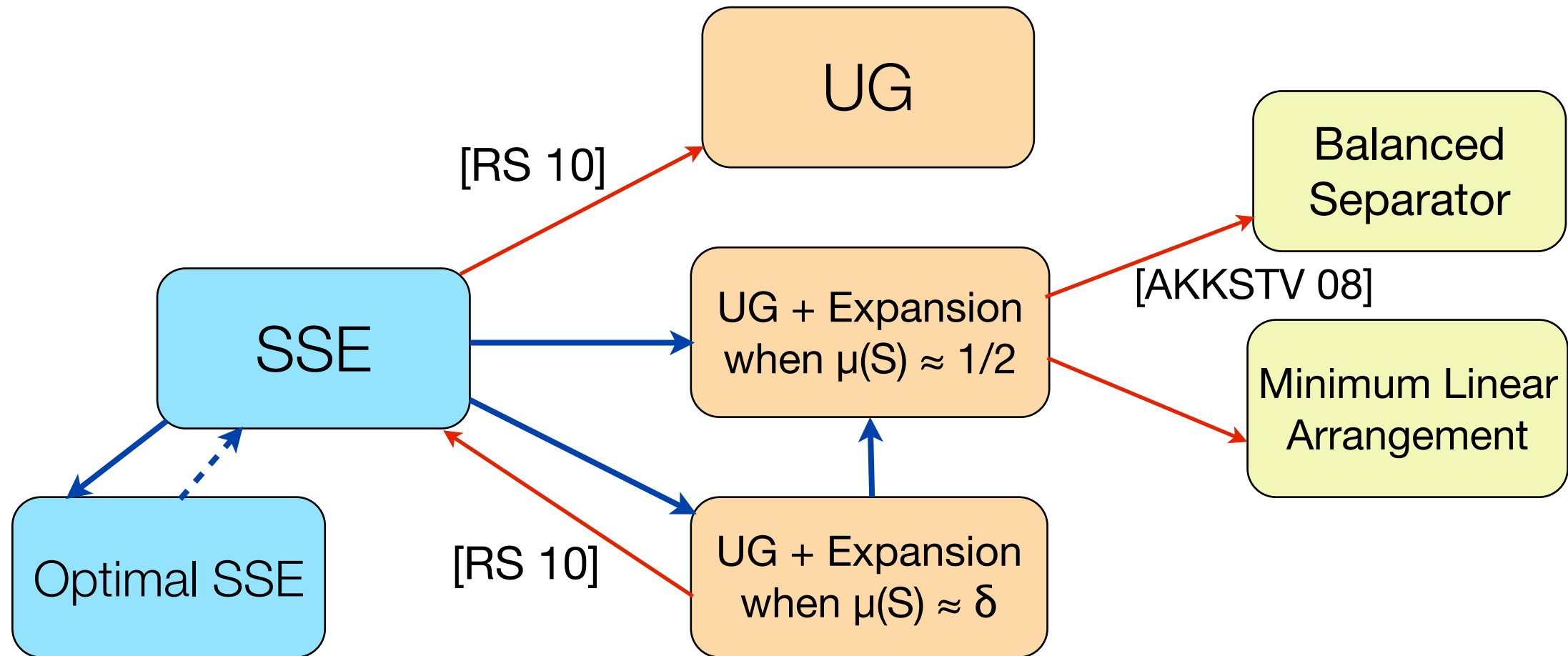
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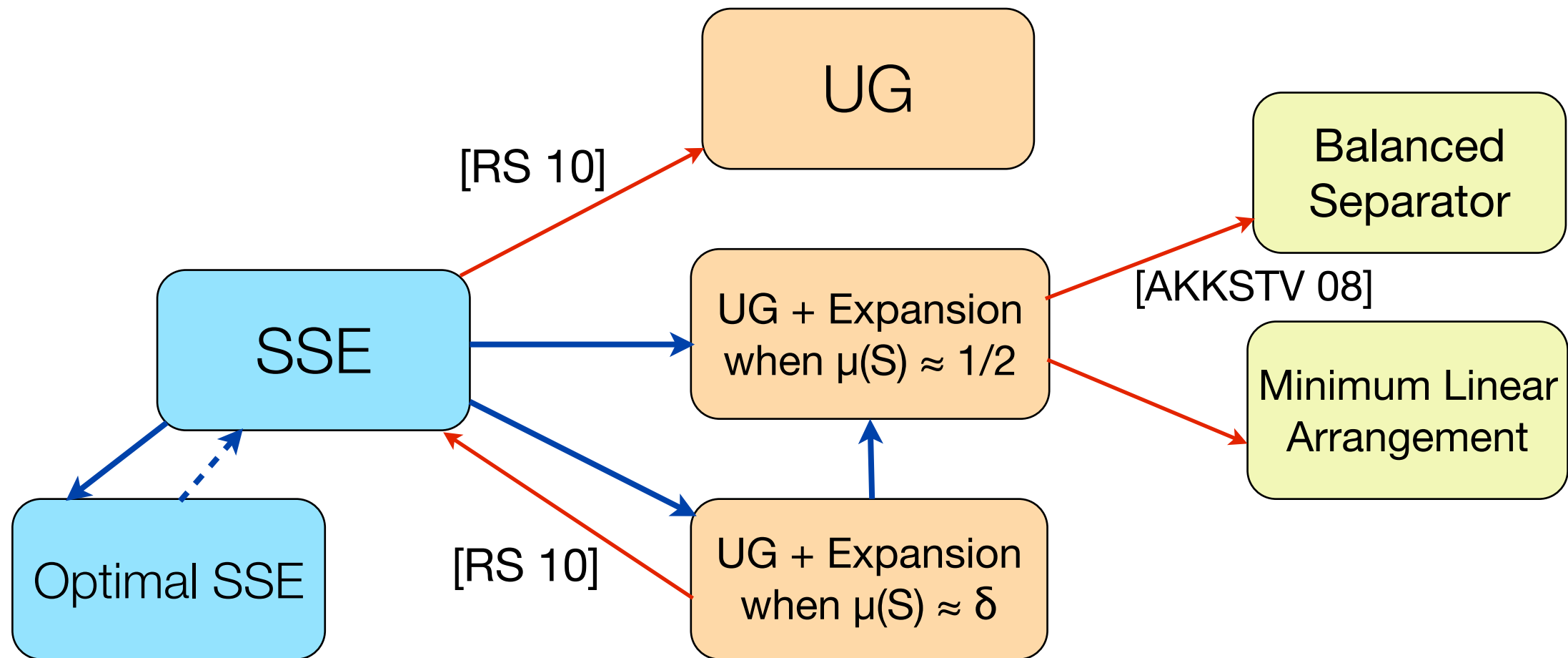
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The big-picture



- Can reduce problem about expansion at scale δ to any scale $> \delta$.
- Similar to alphabet reduction for Unique Games by [KKMO 04] (UG alphabet $\approx 1/\delta$).

Main result

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For all $\epsilon, \delta > 0$ and $q \in \mathbb{N}$, the following is SSE-hard.

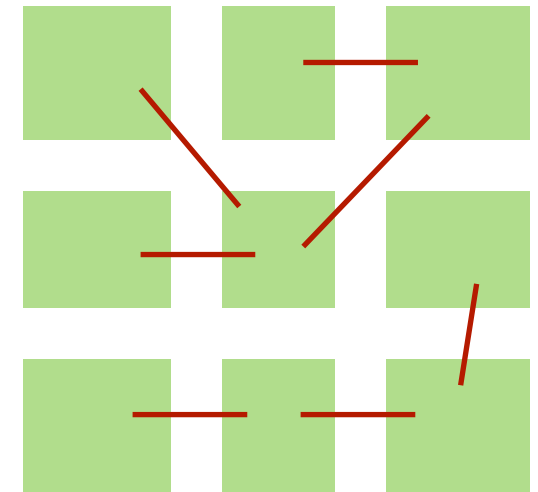
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Main result

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Given $H = (V_H, E_H)$, distinguish between the cases:

- V_H can be partitioned into equal pieces S_1, \dots, S_q
s.t. $\forall t \quad \Phi_G(S_t) \leq \epsilon + o(\epsilon)$.

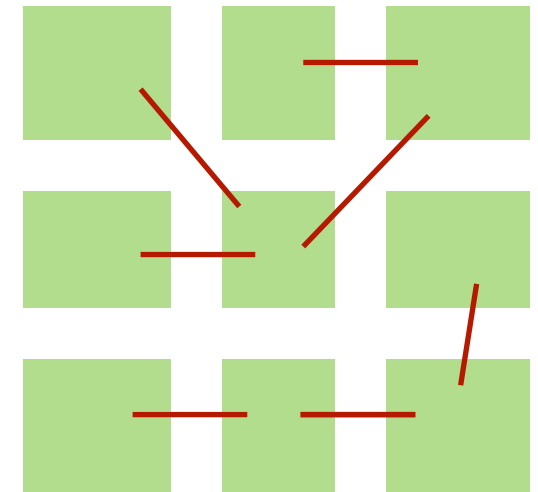


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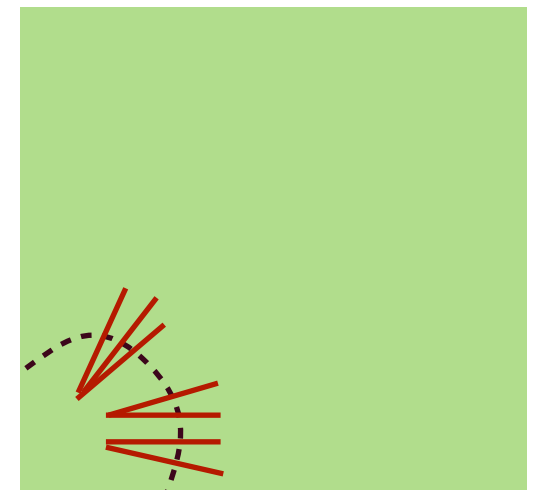
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- For all $S \subseteq V_H$, with $\mu(S) \geq \delta$
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vs.



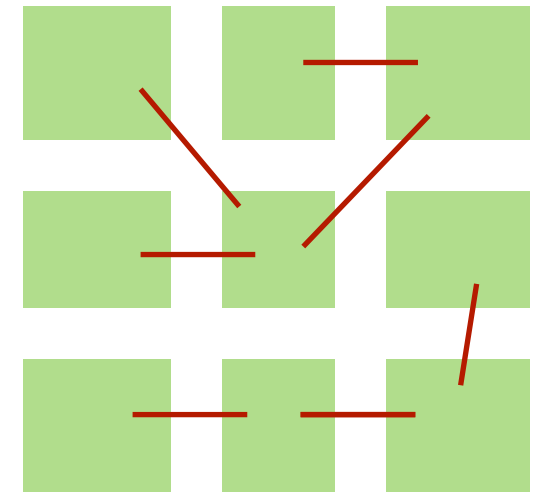
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Matches (upto constants) the algorithmic guarantee of [RST 10]



vs.



Application to Balanced Separator

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Simply taking $q = 2$ and $\delta = 1/4$ gives the hardness of distinguishing for $H = (V_H, E_H)$:

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Also gives hardness for Minimum Bisection since $\mu(S) = 1/2$ in the first case.

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Problem: Find an ordering π of the vertices to minimize the average length of an edge = $E_{(x,y) \in E} |\pi(x) - \pi(y)|$.

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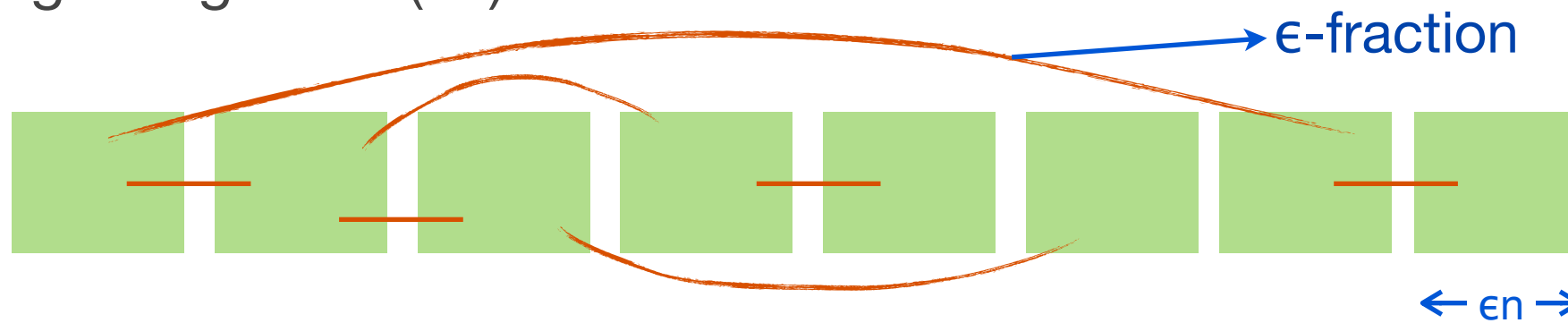
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Average length = $O(\epsilon n)$.

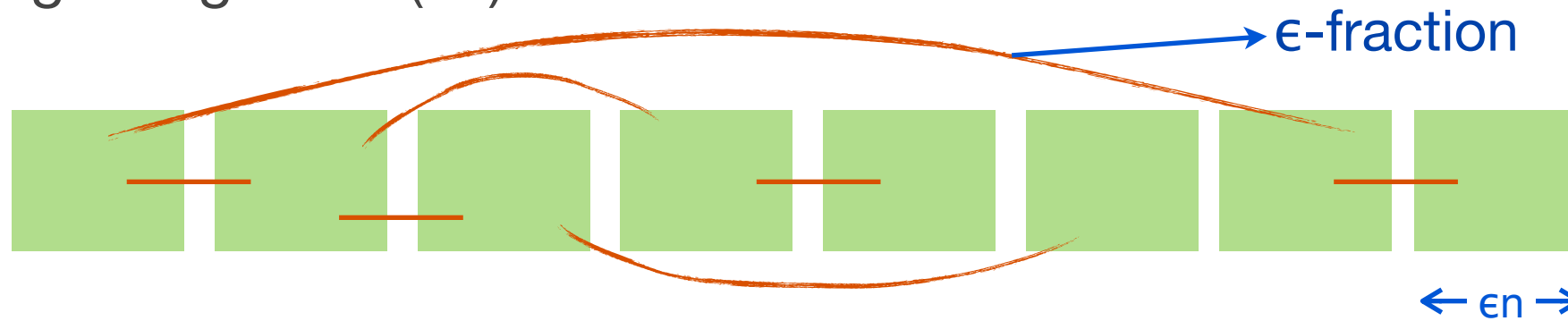


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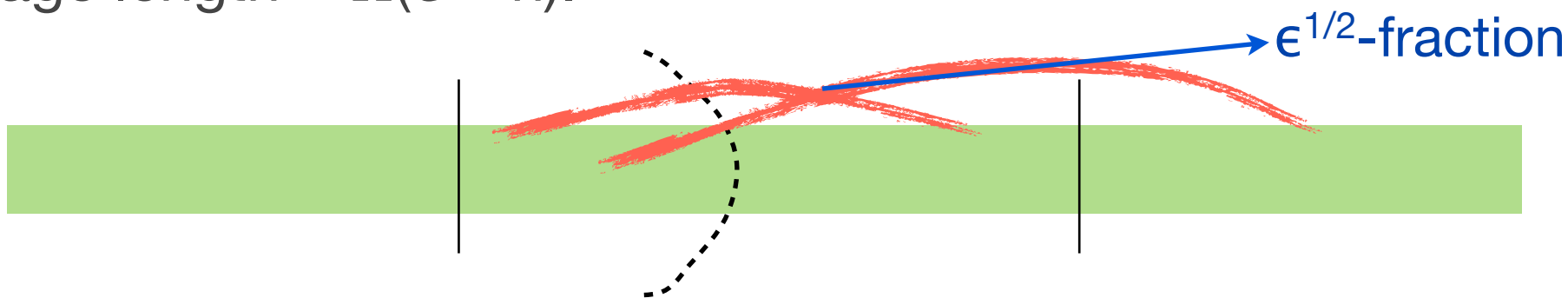
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- $\Omega(\epsilon^{1/2})$ fraction of edges crossing every cut on middle $n/3$ vertices.
Average length = $\Omega(\epsilon^{1/2} n)$.



Reducing to an expansion problem

Unique Games


Unique Games

- **Given:** Graph $G=(V,E)$, alphabet R and for each edge (u,v) , a permutation constraint $\pi_{uv} : [R] \rightarrow [R]$.
Find: A labeling $L : V \rightarrow [R]$ satisfying for as many edges (u,v) as possible
 $\pi_{uv}(L(u)) = L(v)$
- Can also be viewed as a two-prover game (Alice and Bob get one vertex each from a randomly chosen edge).

Unique Games

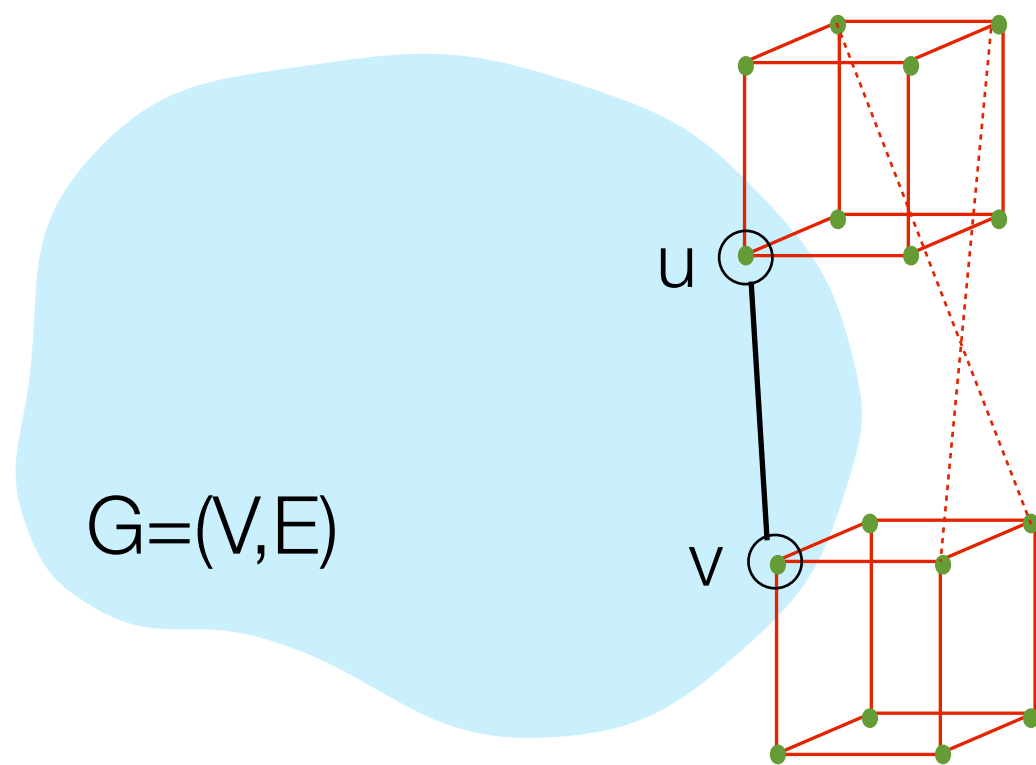
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- Can also be viewed as a two-prover game (Alice and Bob get one vertex each from a randomly chosen edge).
- **UGC:** Hard to distinguish between the cases when
 - At least $1-\eta$ fraction of edges can be satisfied
 - At most η fraction of edges can be satisfied.

(Attempted) Reduction to Balanced Separator



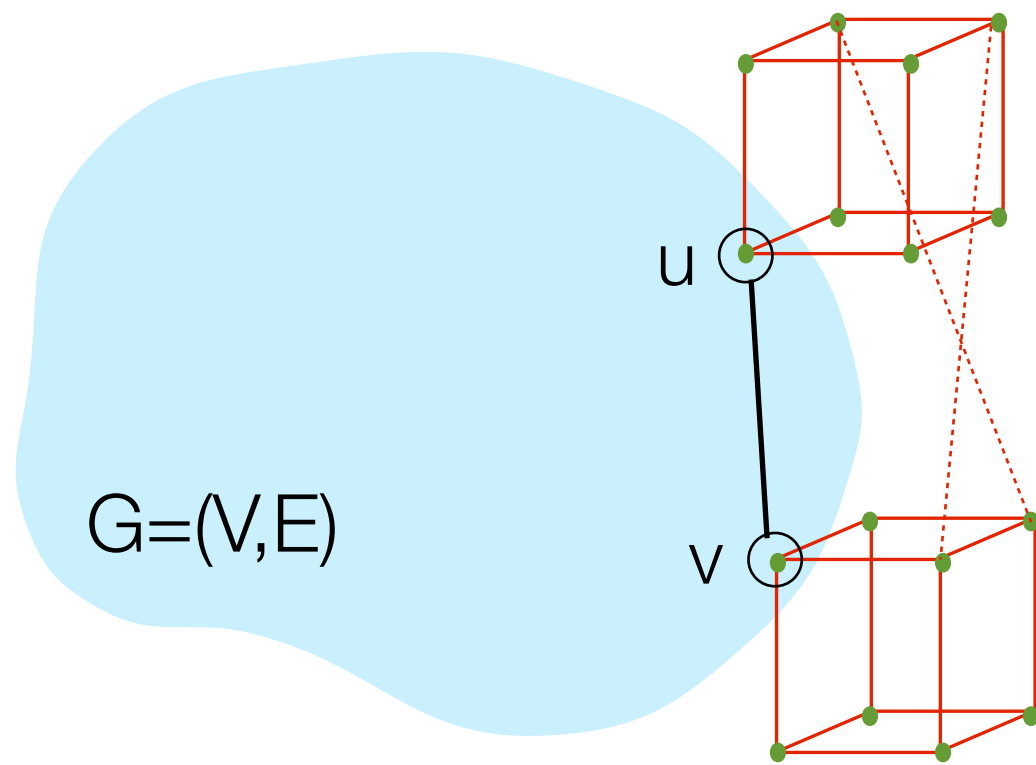
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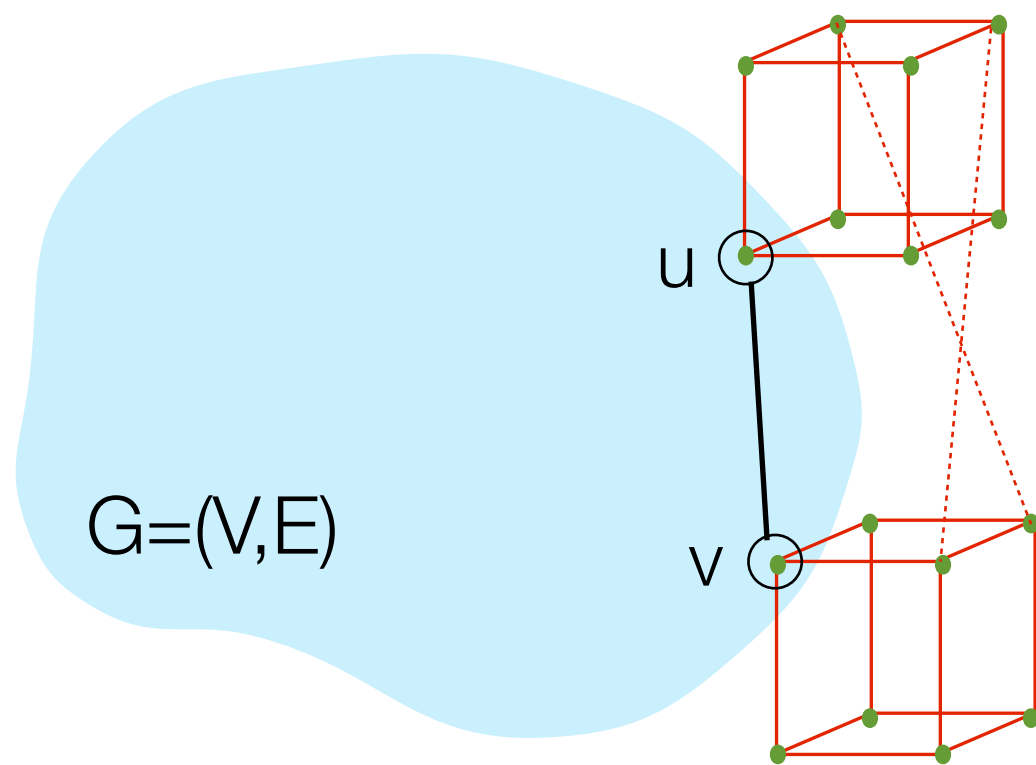
- Attach $\{0,1\}^R$ to each vertex of G
- Pick $x, y \in \{0,1\}^R$ with ϵ -noise
(weight = $\epsilon^{H(x,y)} (1-\epsilon)^{R-H(x,y)}$)
- Connect $(u, \pi_{uv}(x))$ to (v,y) with above weight. (send $(u, \pi_{uv}(x))$ to Alice, (v,y) to Bob)
- Ask for a balanced cut on new graph.

(Attempted) Reduction to Balanced Separator



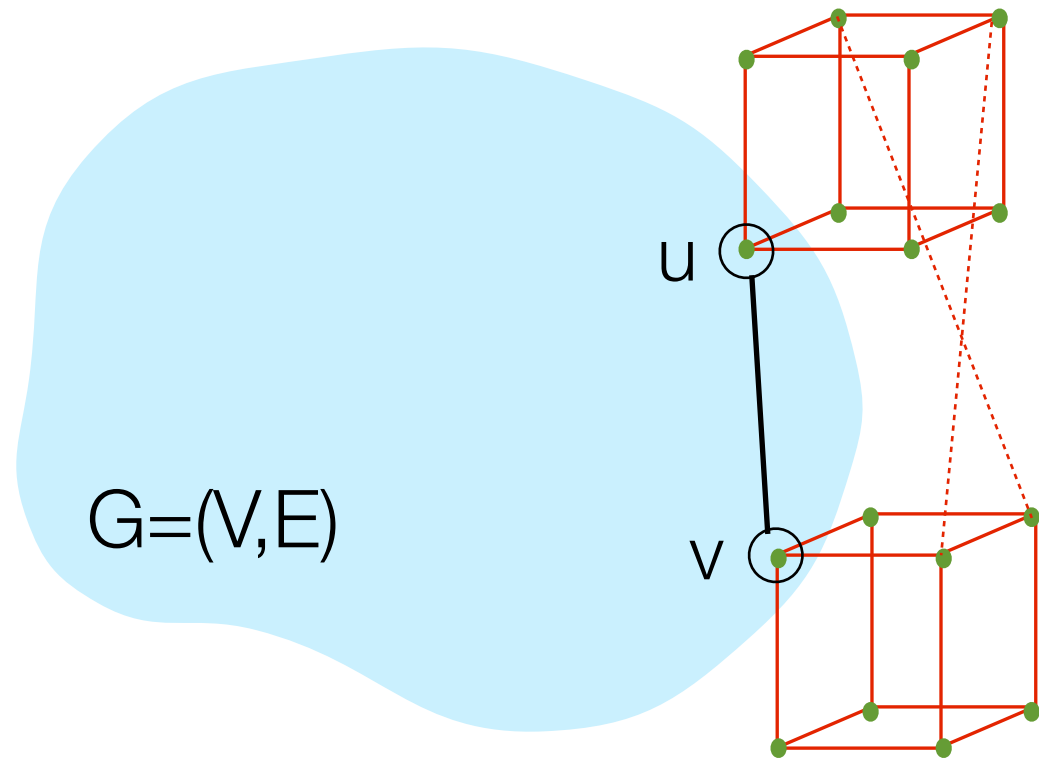
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 - If $\pi_{uv}(L(u)) = L(v)$, then only ϵ -fraction of edges between cubes of u and v are cut.

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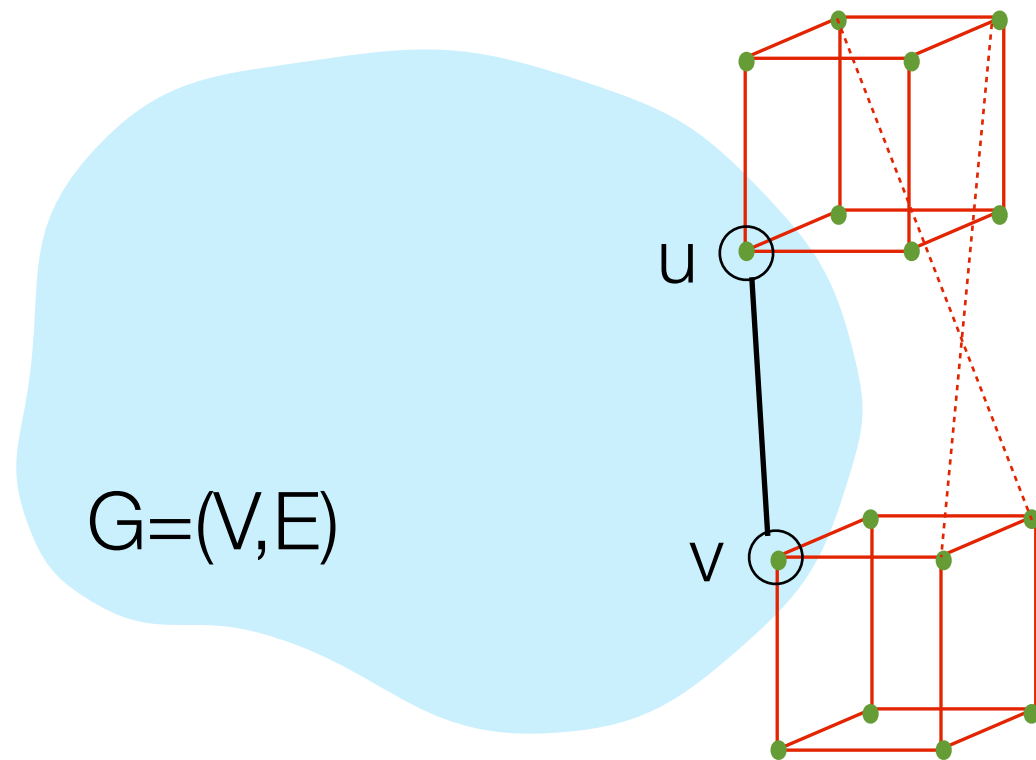


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 - If labeling satisfies $1-\eta$ fraction, then value of cut in graph is $\leq \epsilon + \eta$.

Problems with the reduction

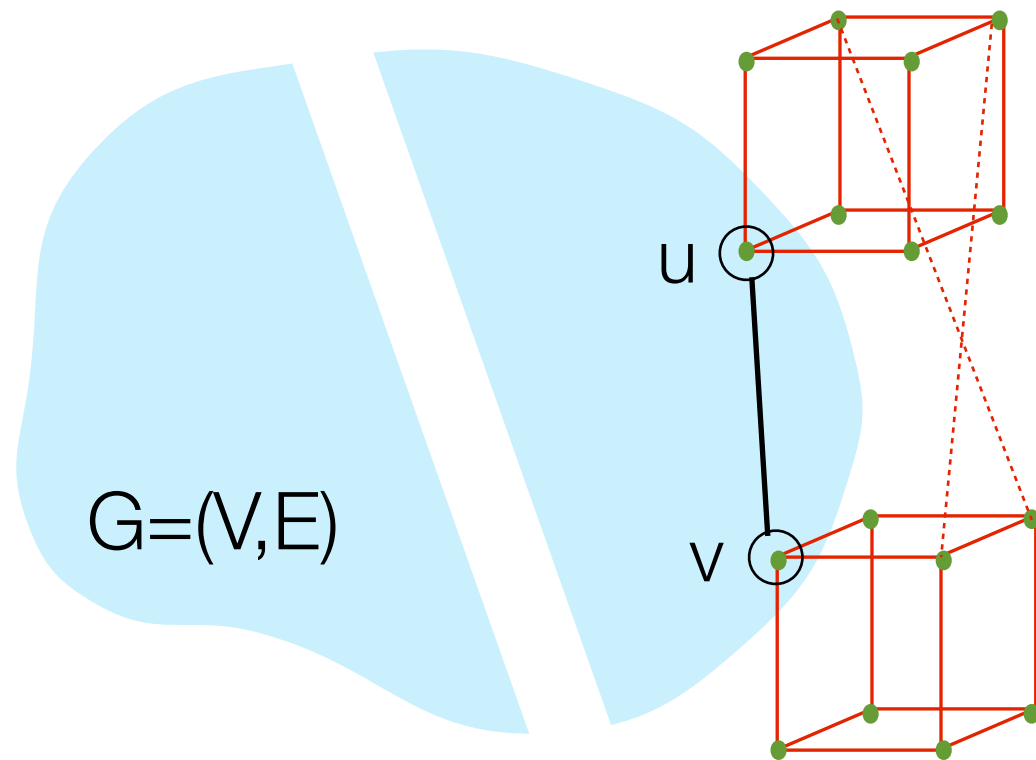


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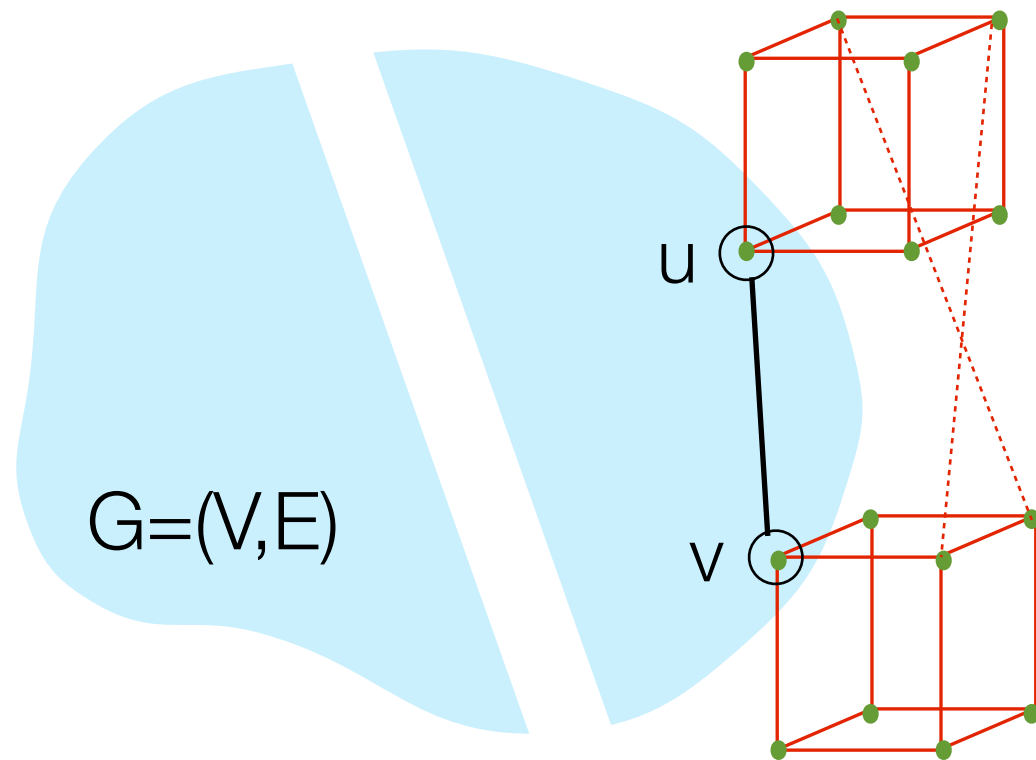
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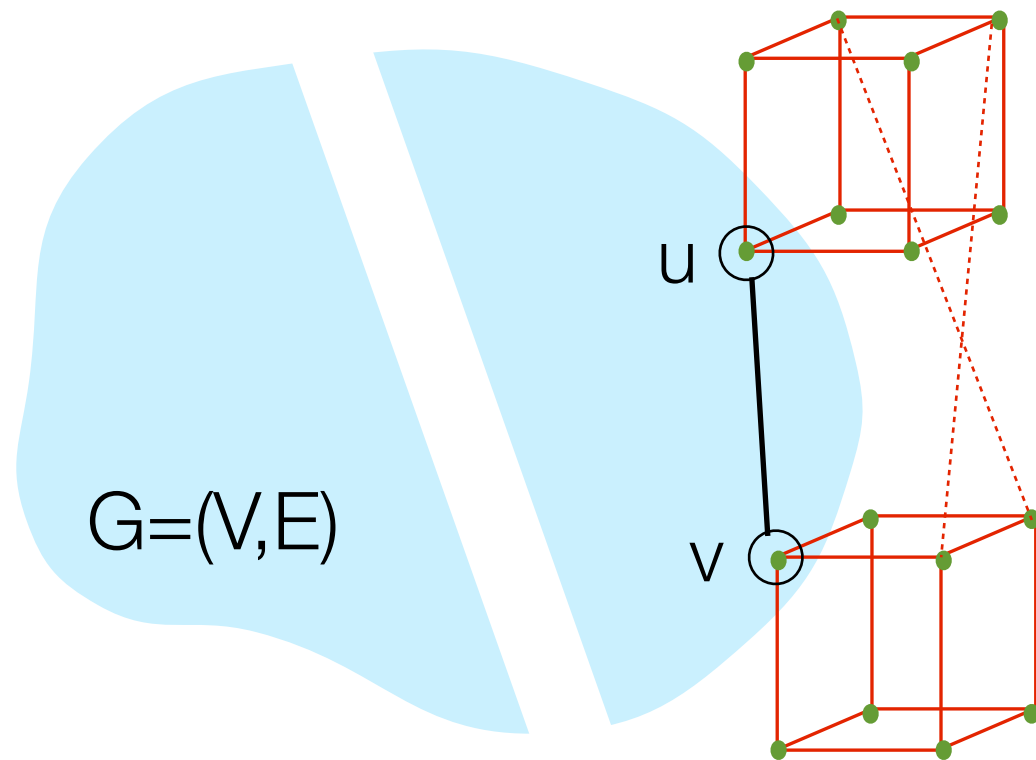
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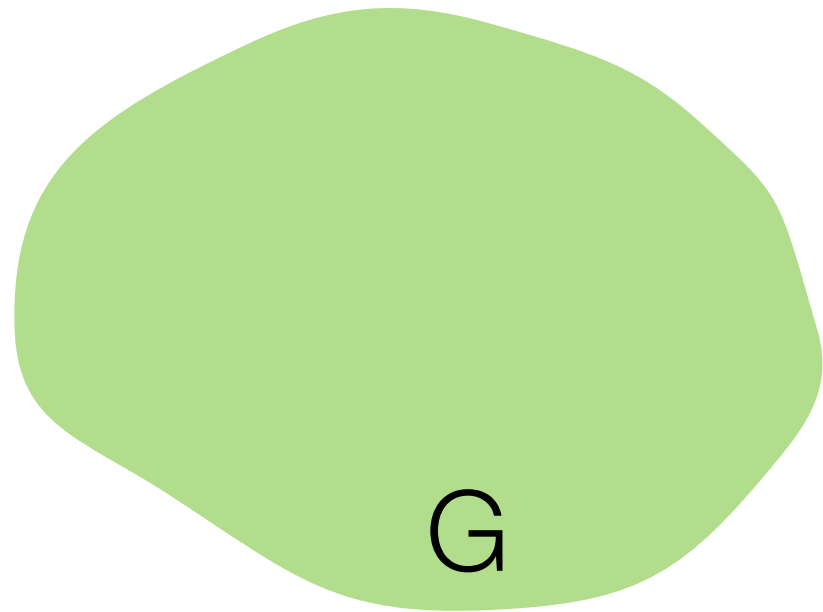
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- Reduction preserves structure of starting graph.
- If starting UG does not have expansion, neither does the new graph.

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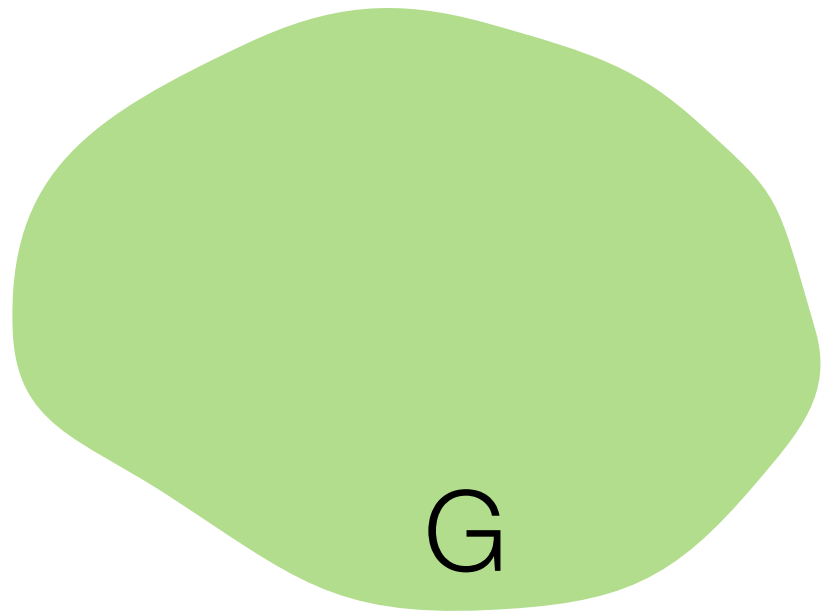


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- For UGs obtained by reducing from Small-Set Expansion, can introduce a special folding operation, which **does not preserve structure** of UG-graph.
 - Even though the graph for UG may not have expansion, the graph for Balanced Separator does.

From SSE to Unique Games [RS 10]

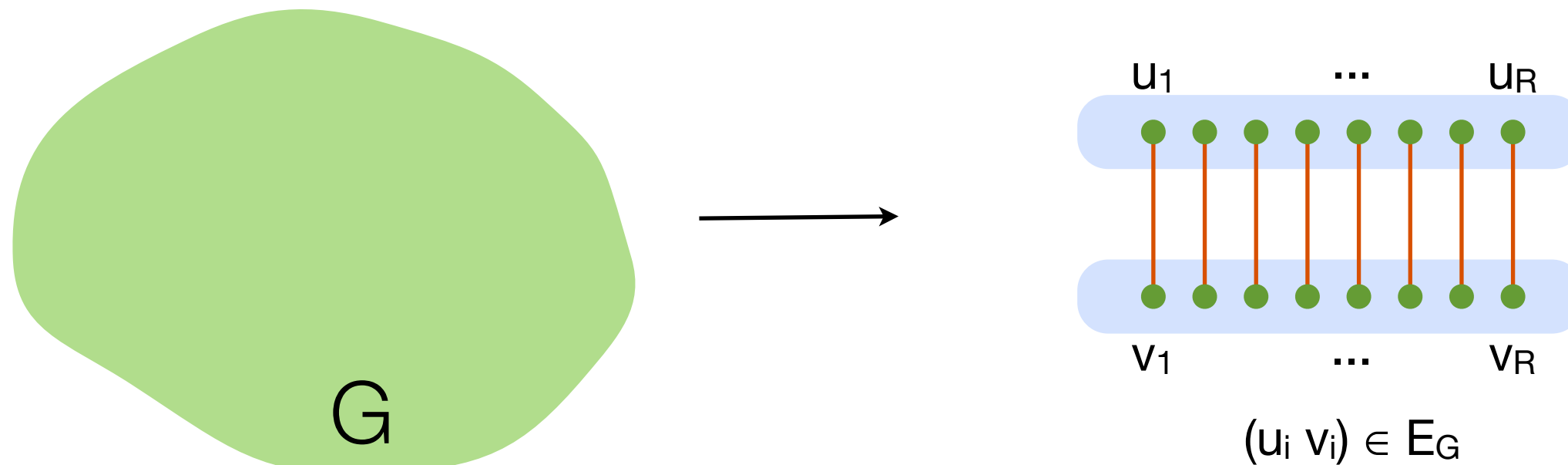


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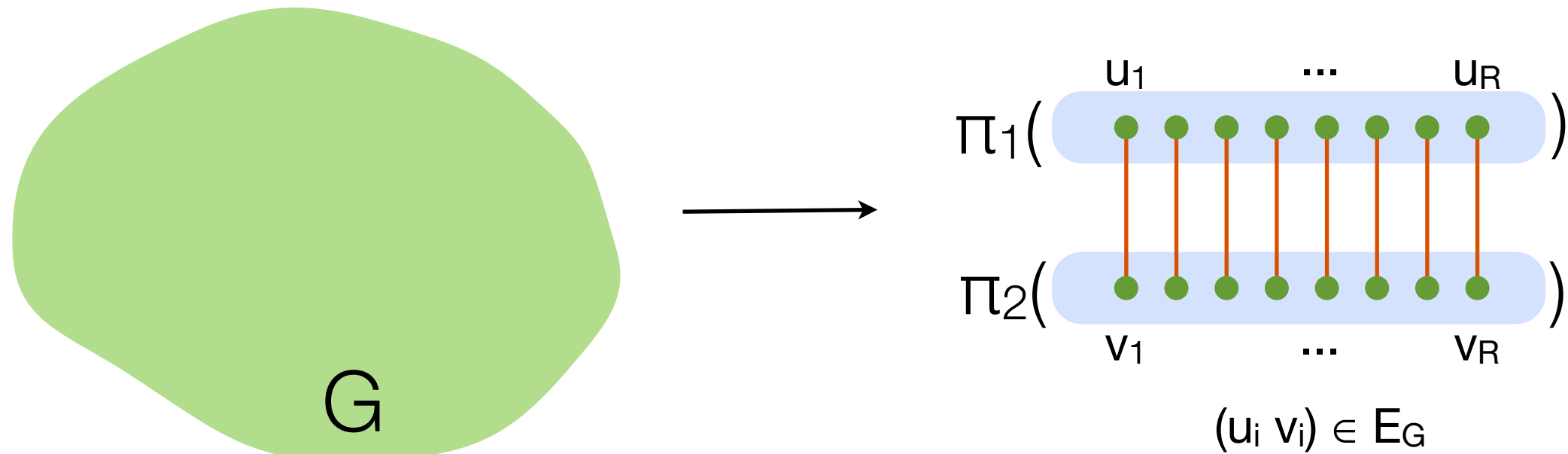
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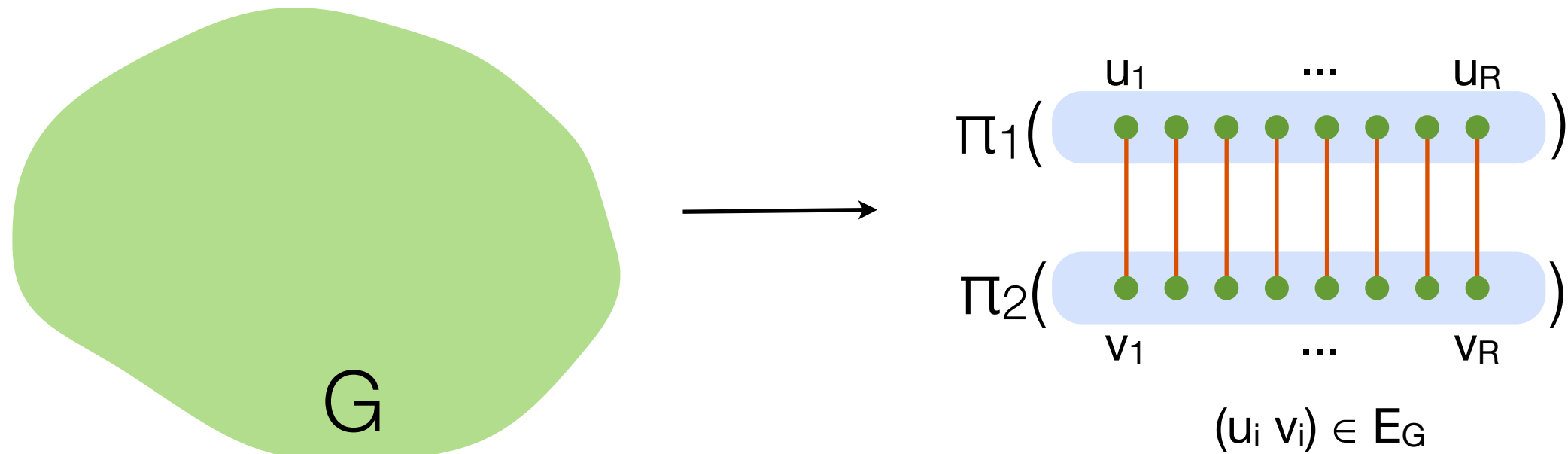
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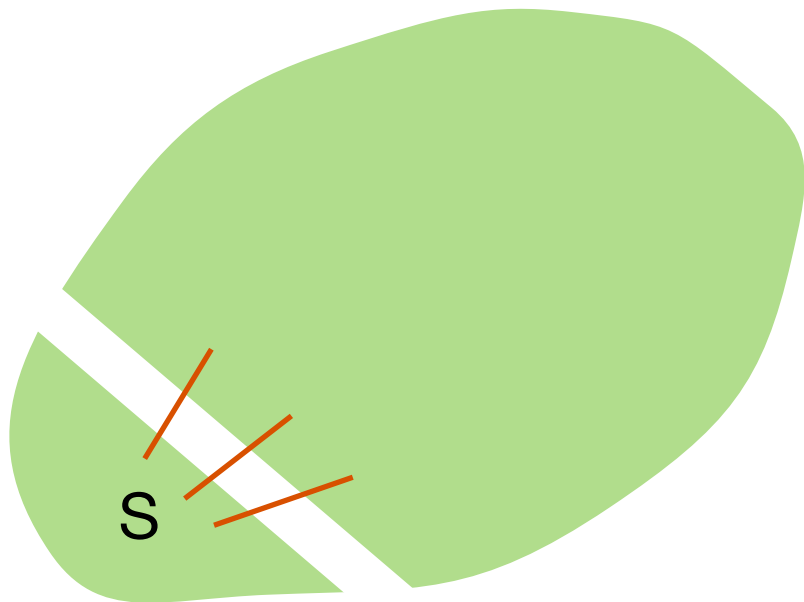
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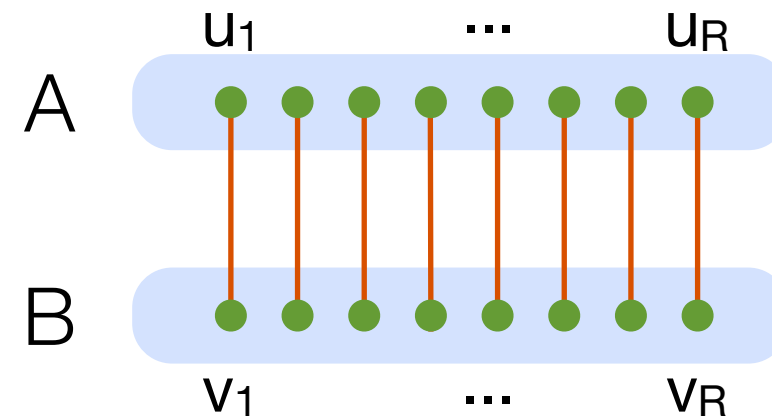


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- Connect $\pi_1(A)$ to $\pi_2(B)$. Call them u, v . (send $\pi_1(A)$ to Alice, $\pi_2(B)$ to Bob)
- Labeling is required to find an edge in G . Constraint on (u, v) is
 $\pi_1^{-1}(L(u)) = \pi_2^{-1}(L(v))$

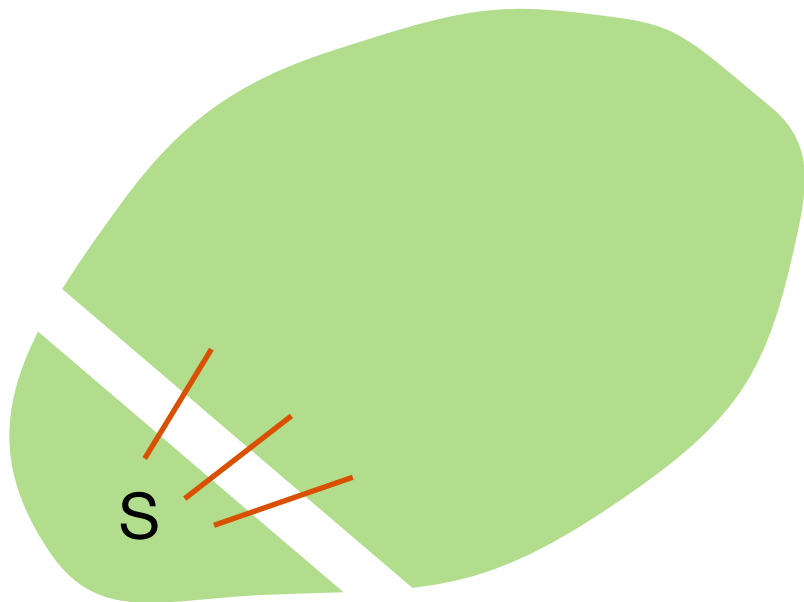
Why the SSE reduction works



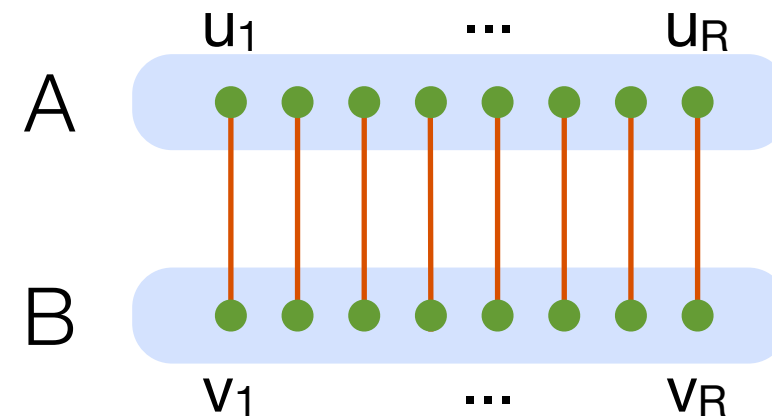
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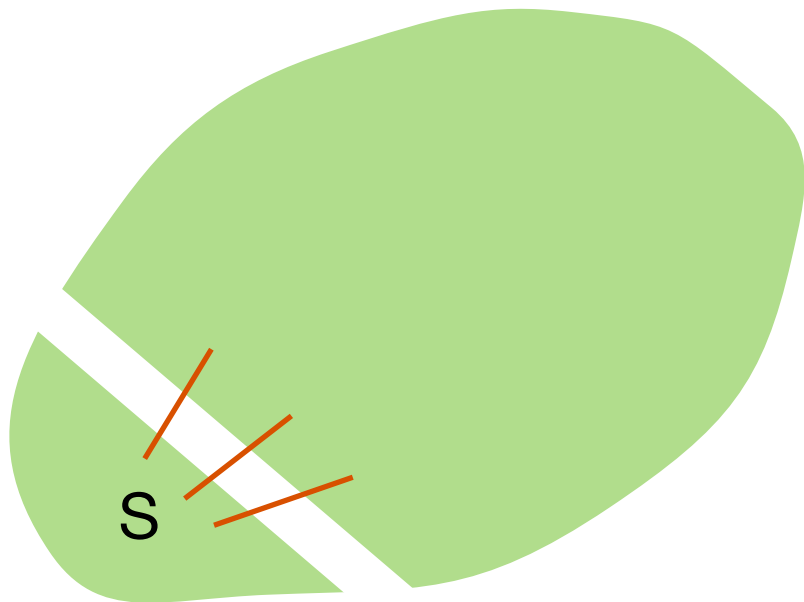


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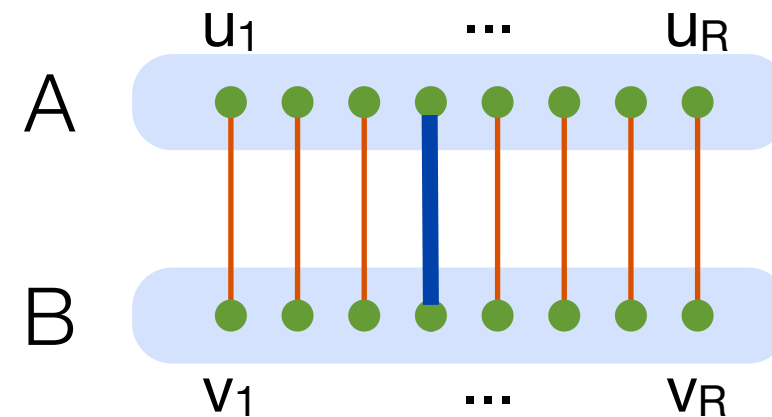


- $P(\text{A has exactly one S-vertex}) = (1 - 1/R)^{R-1} \approx 1/e$

Why the SSE reduction works

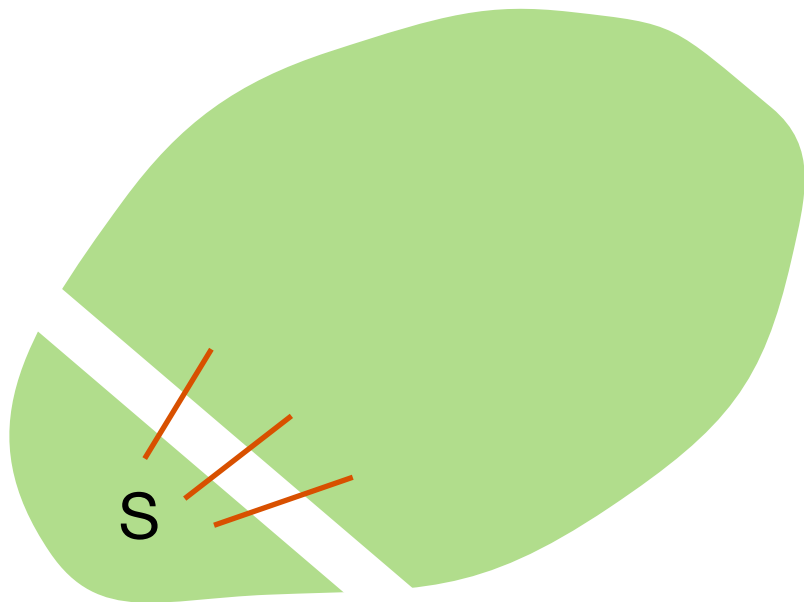


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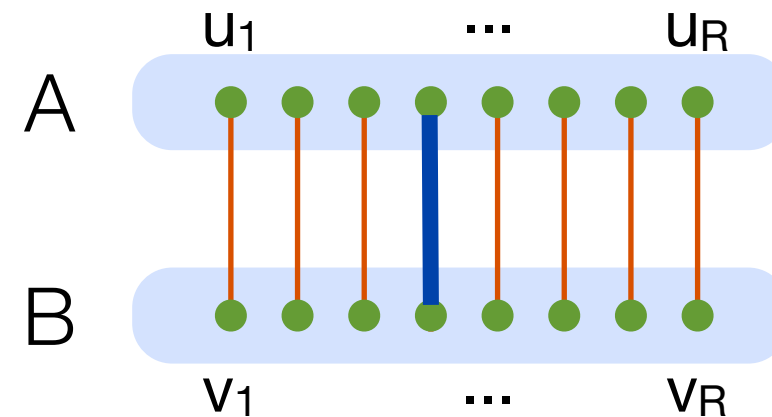


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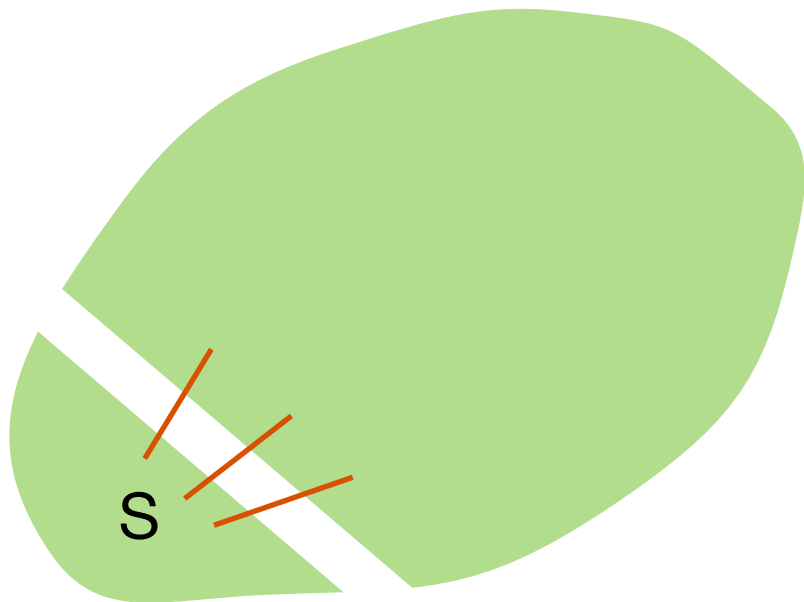


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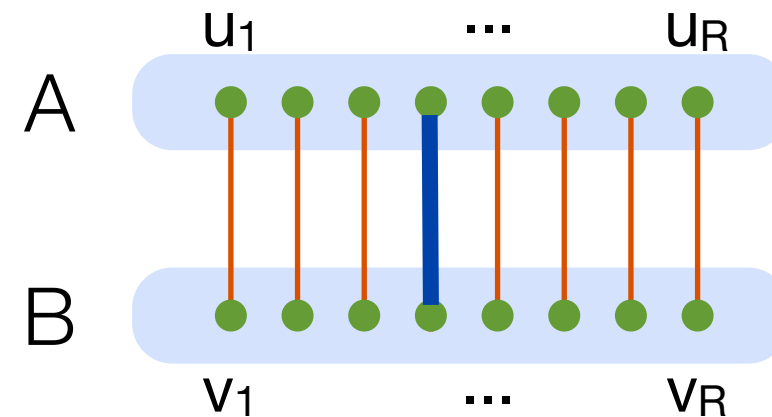


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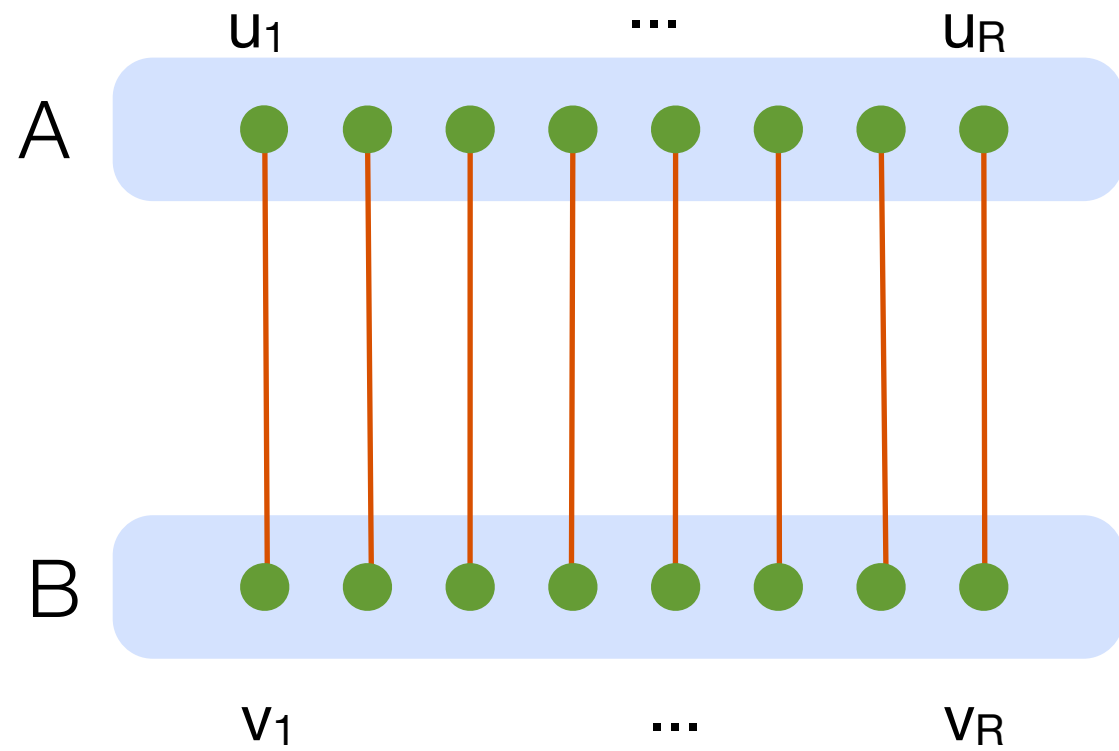


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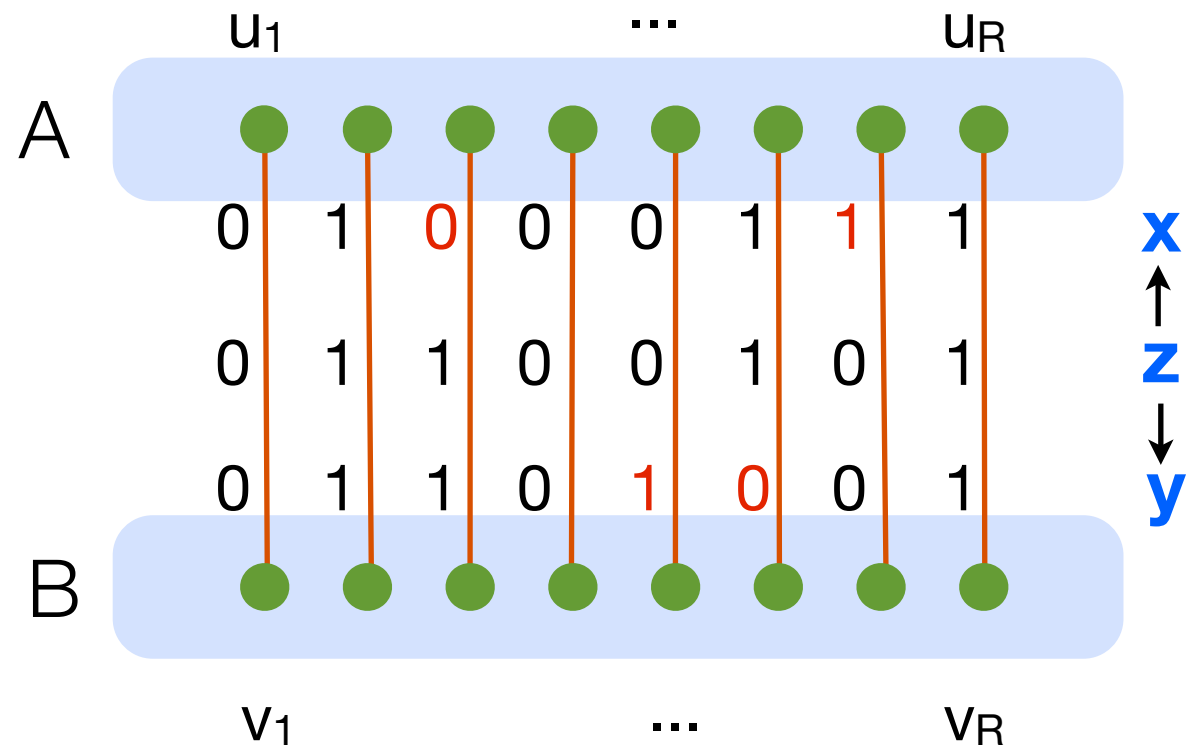


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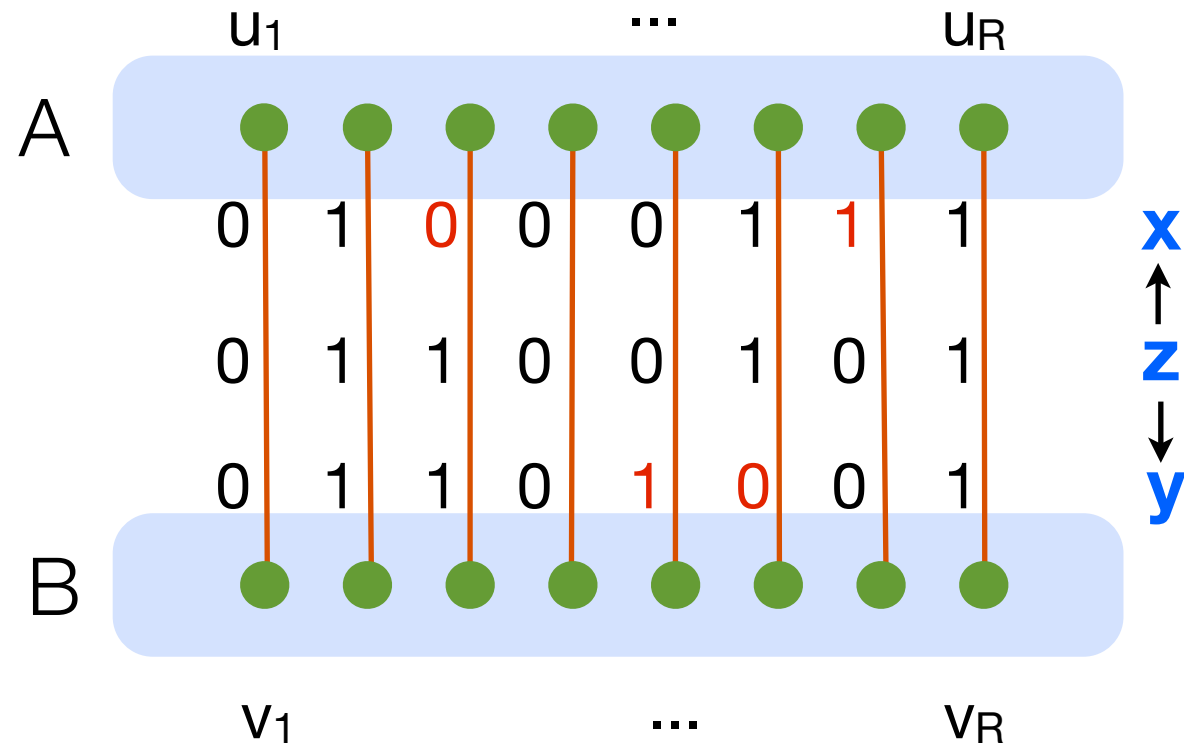
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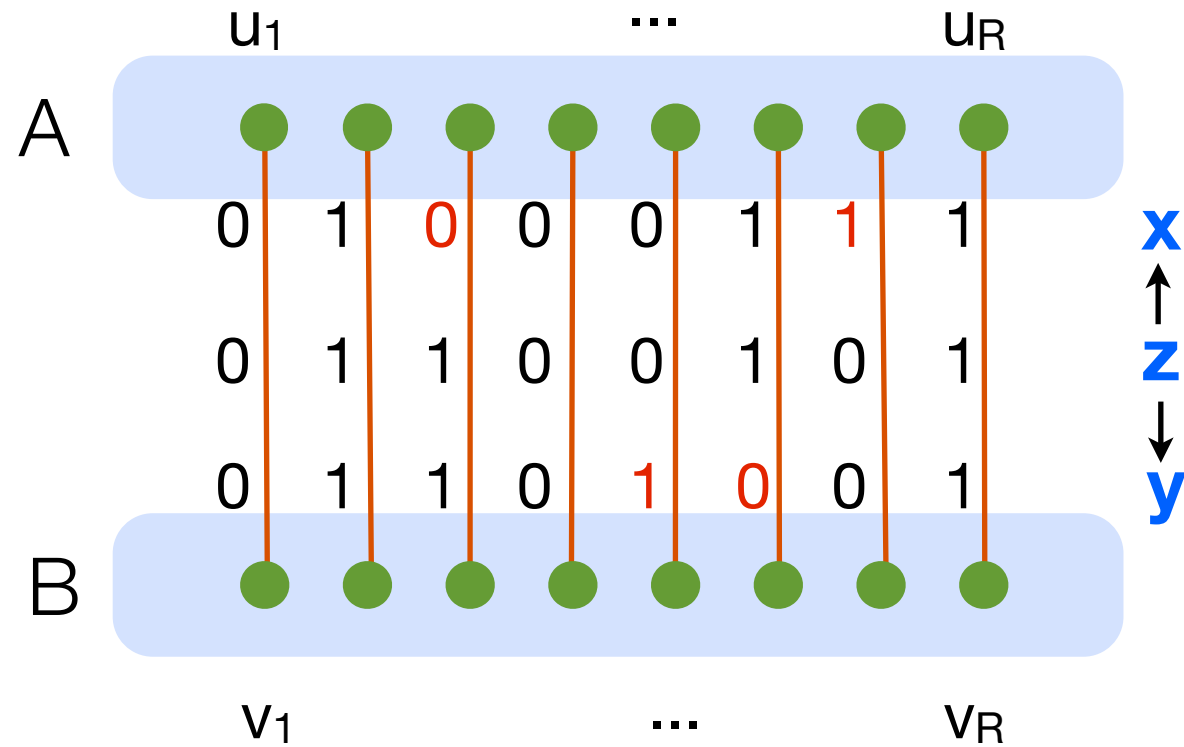


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Generate x, y from z using independent ϵ -noise.

Connect $(\pi_1(A), \pi_1(x))$ to $(\pi_2(B), \pi_2(y))$.

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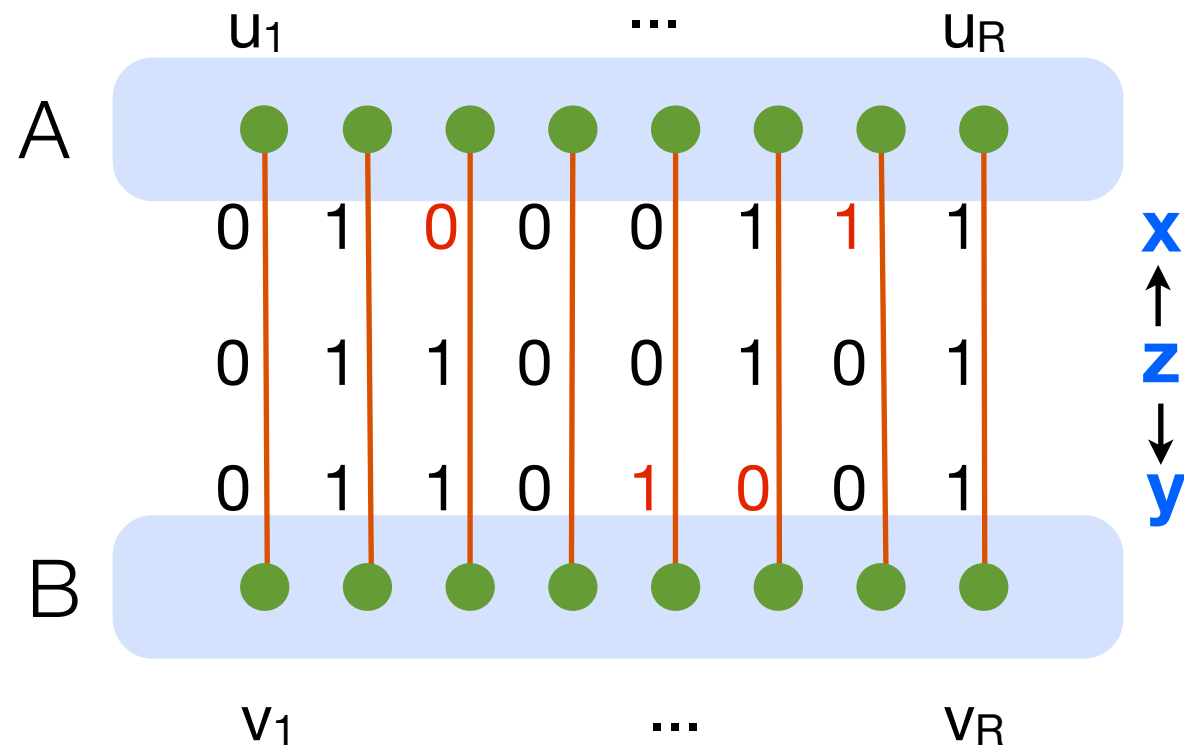
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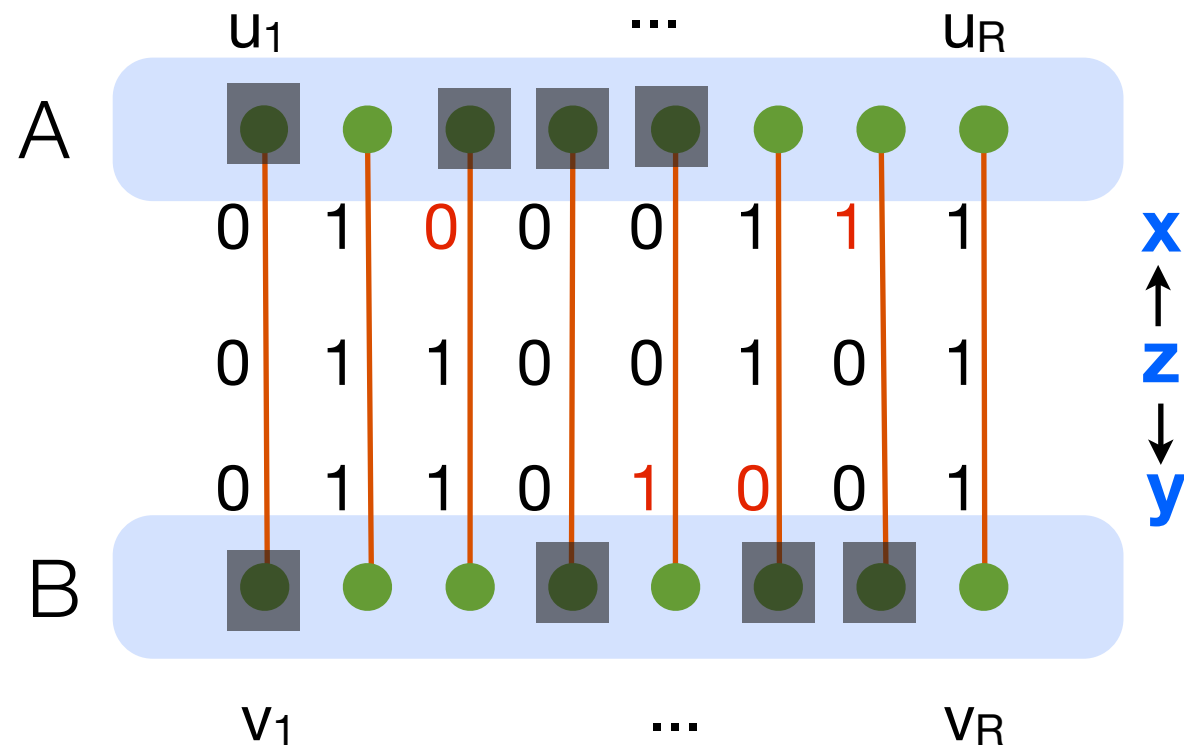
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- Corresponds to a folding operation on the graph obtained after reduction.

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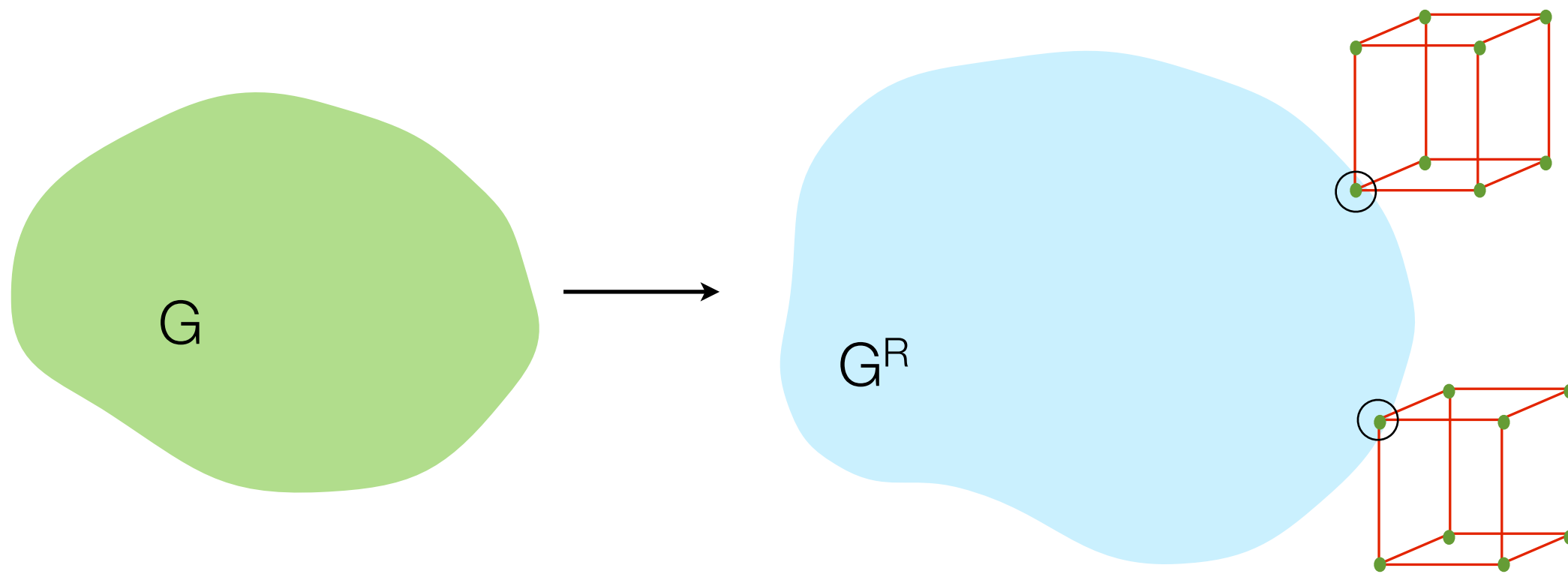
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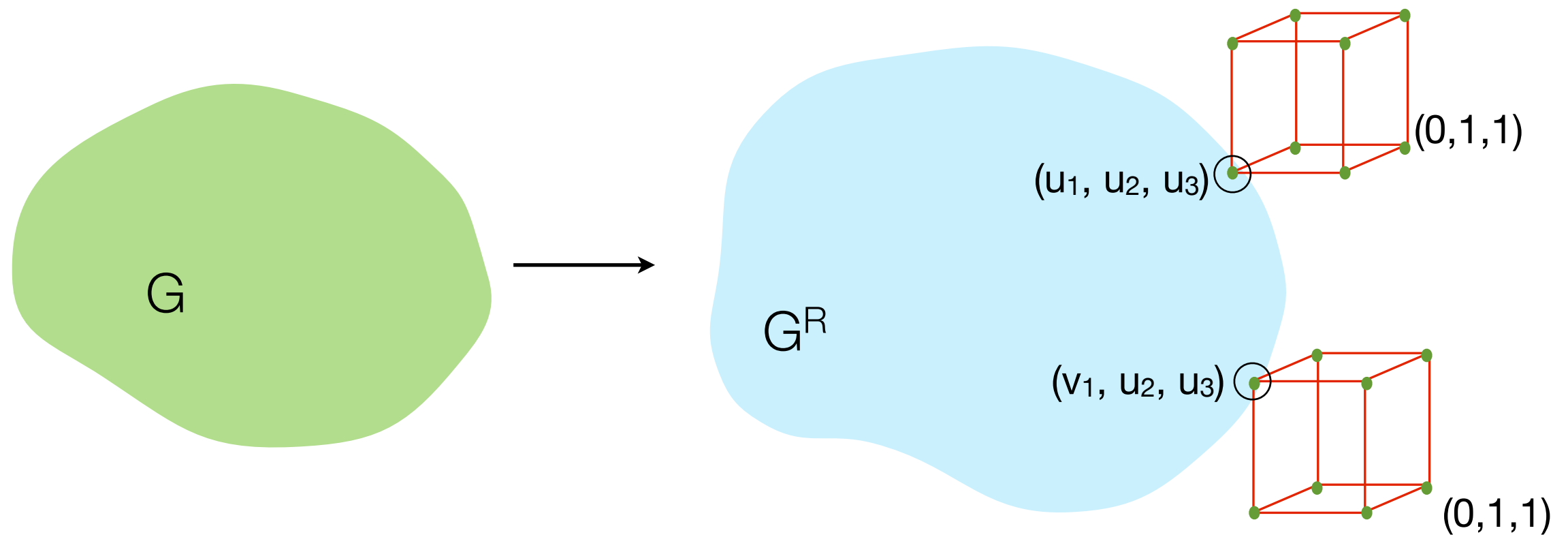
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- Modification only uses the fact that the strategy is about membership in a **set**.

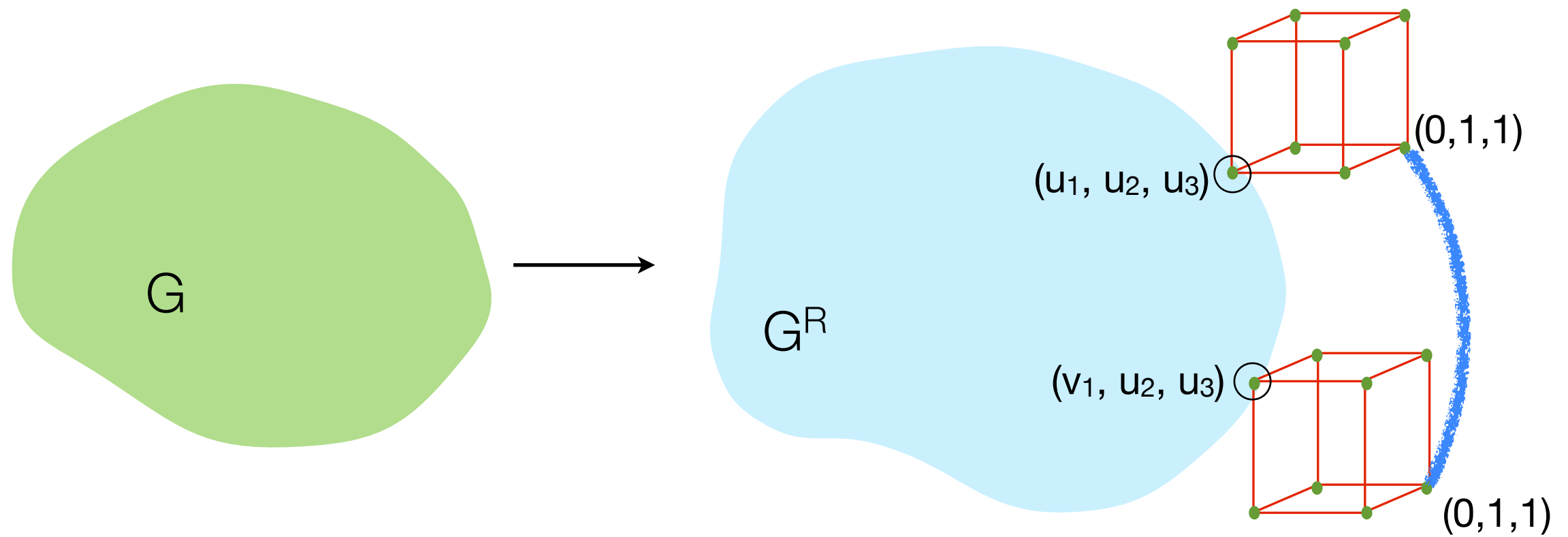
The resulting graph



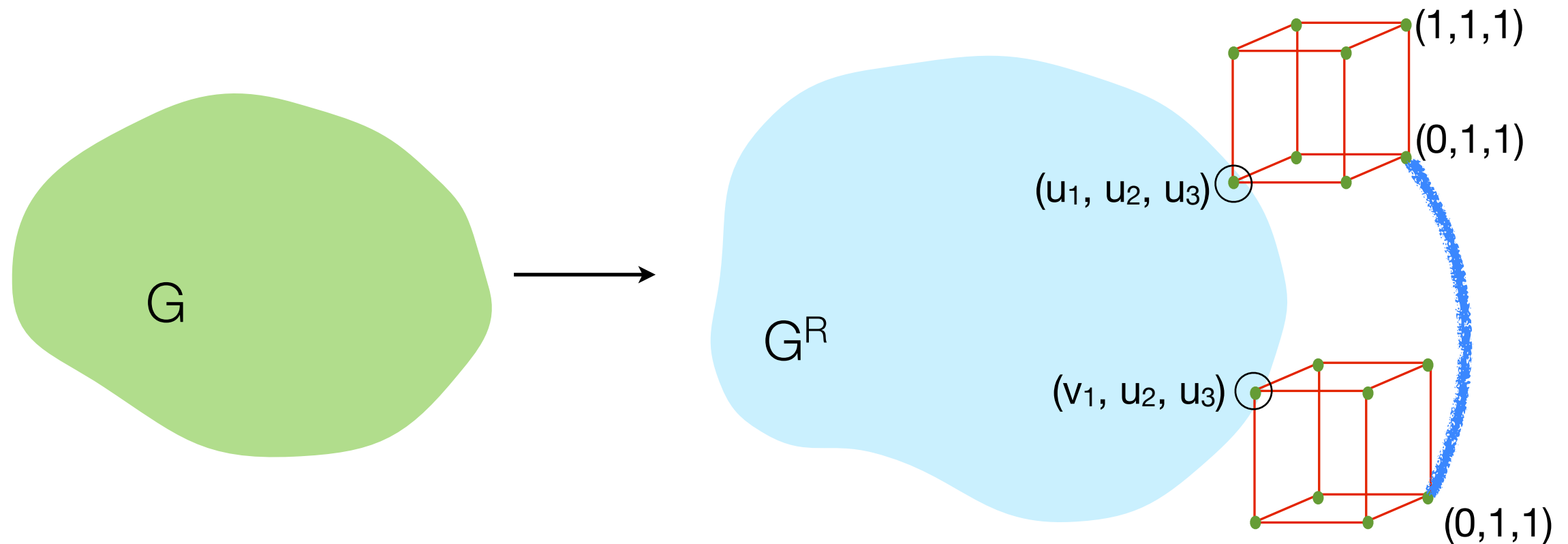
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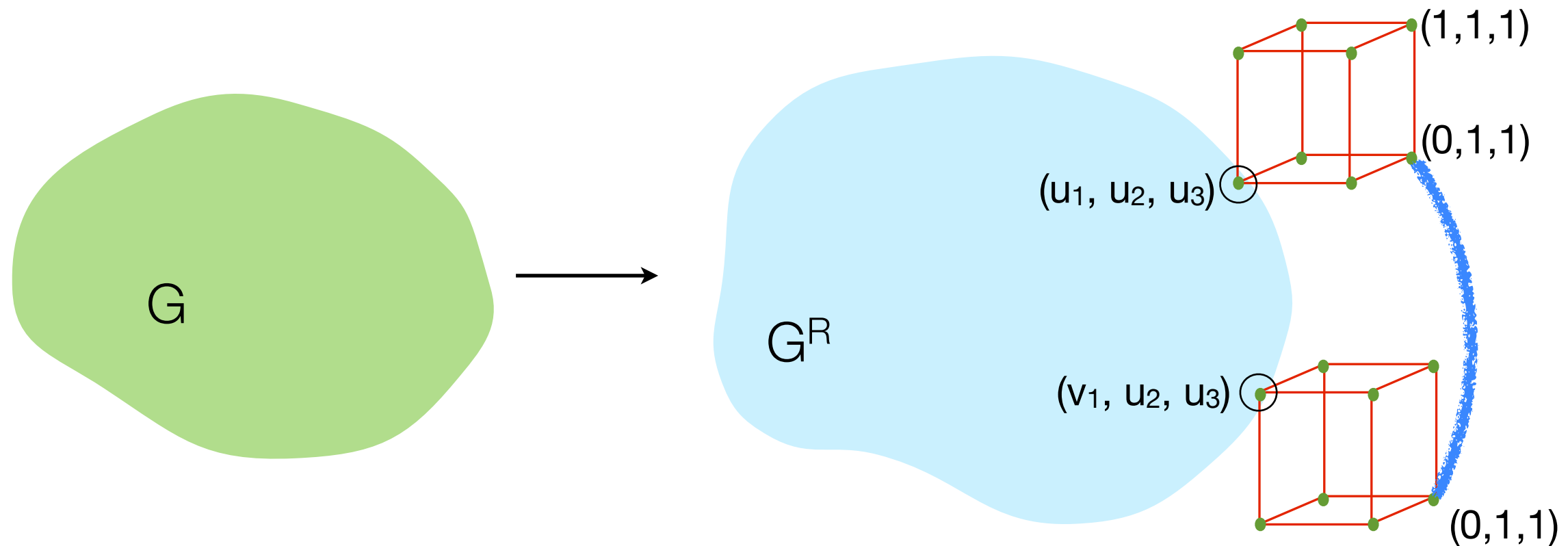


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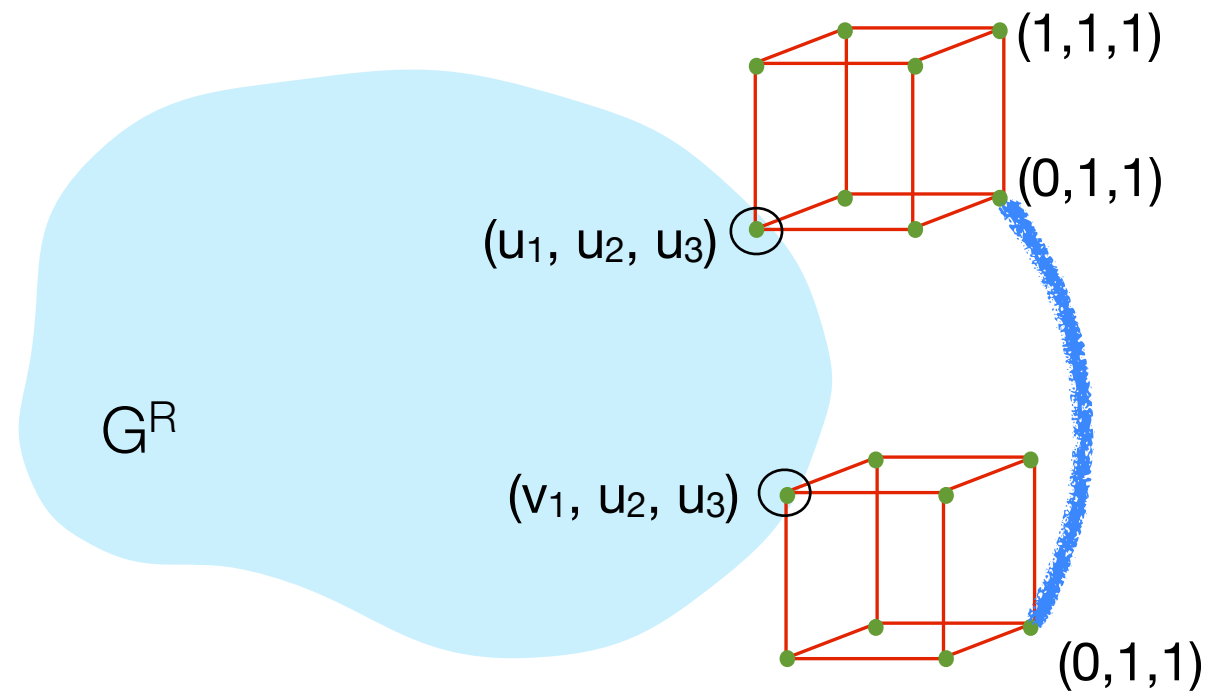
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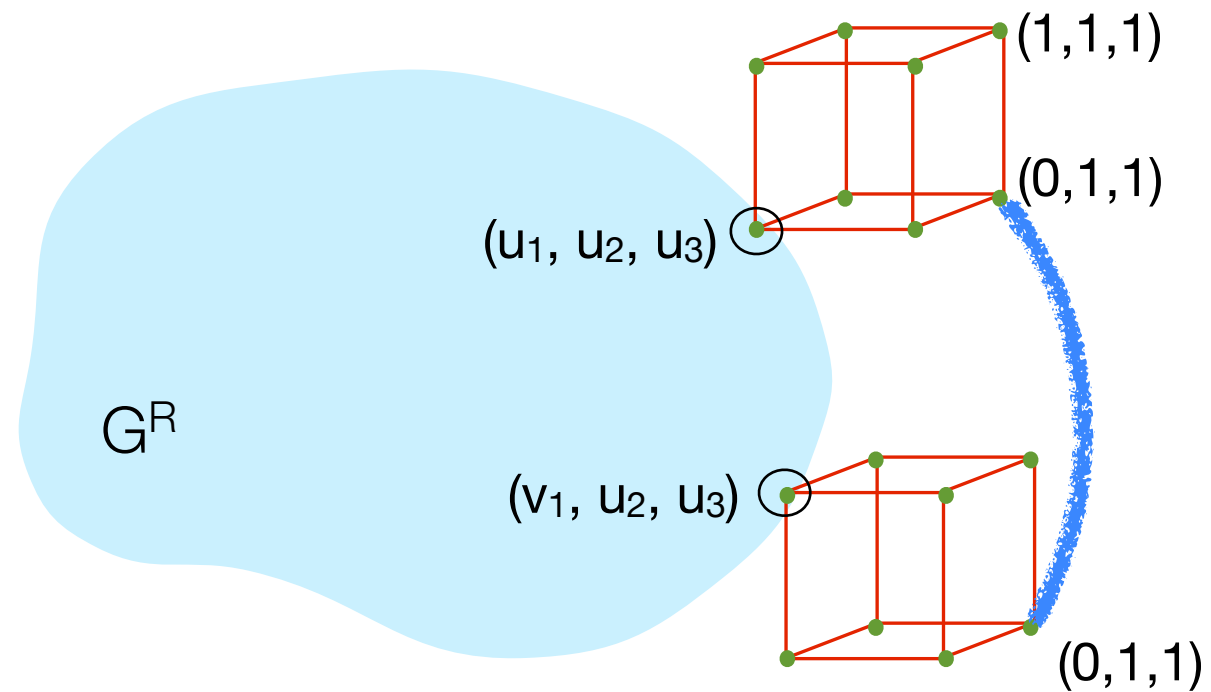


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- Completely destroys the gadget-based nature of the reduction.

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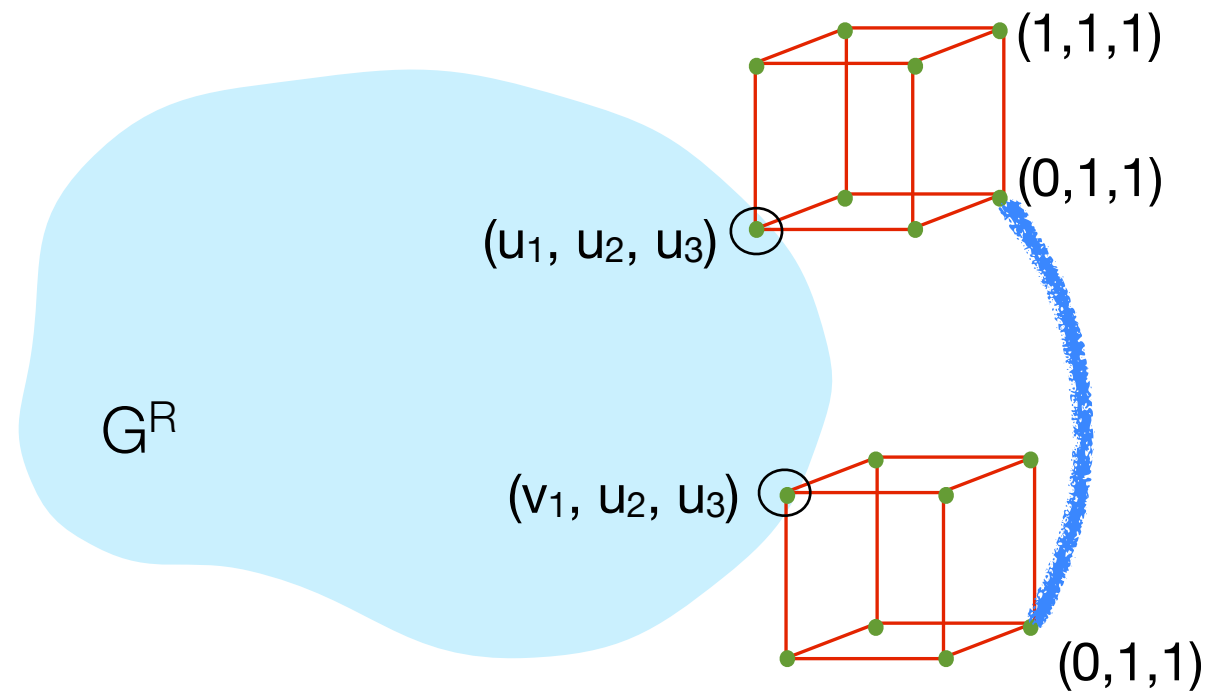


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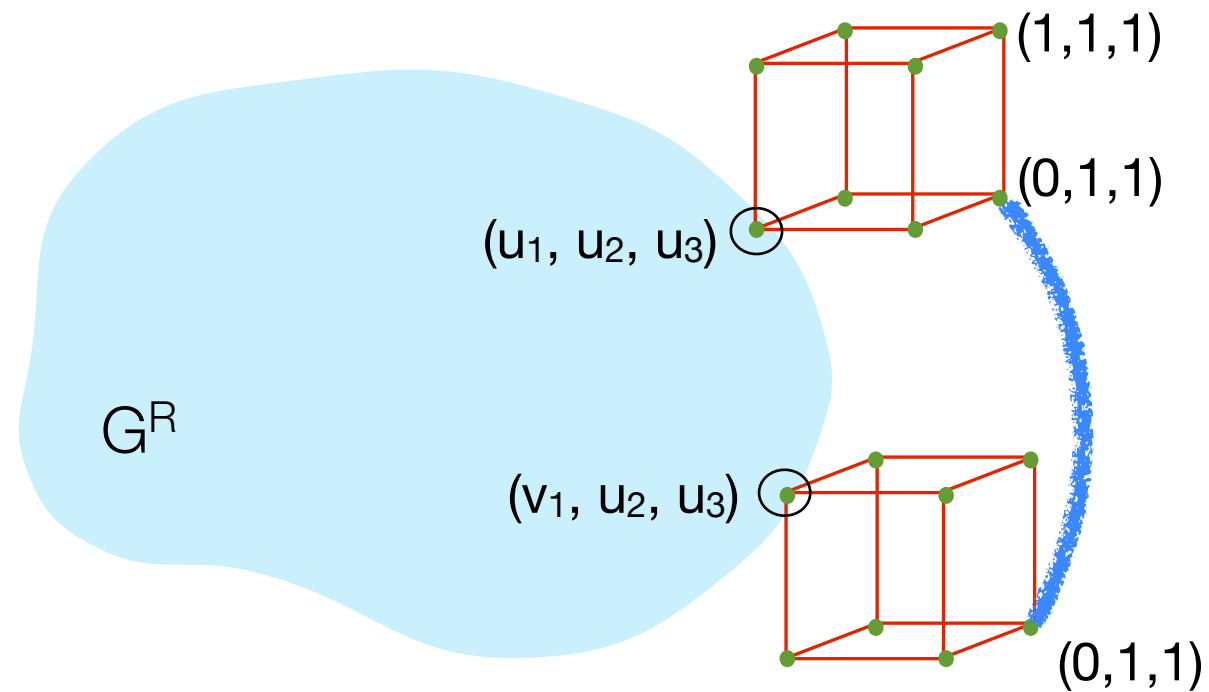
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- Show that any such cut must have large variance in most cubes i.e. cannot be a cut of the underlying graph.
- Obtain full result by using q -ary cubes and more general folding.

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Conclusions

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- Can the folding operation be generalized to other classes of Unique Games where deleting part of the question still preserves a Unique Game.
- Reduction between scales in the other direction (Balanced Separator to SSE)?

Thank You

Questions?