

Tutorial on the adversary method for quantum and classical lower bounds

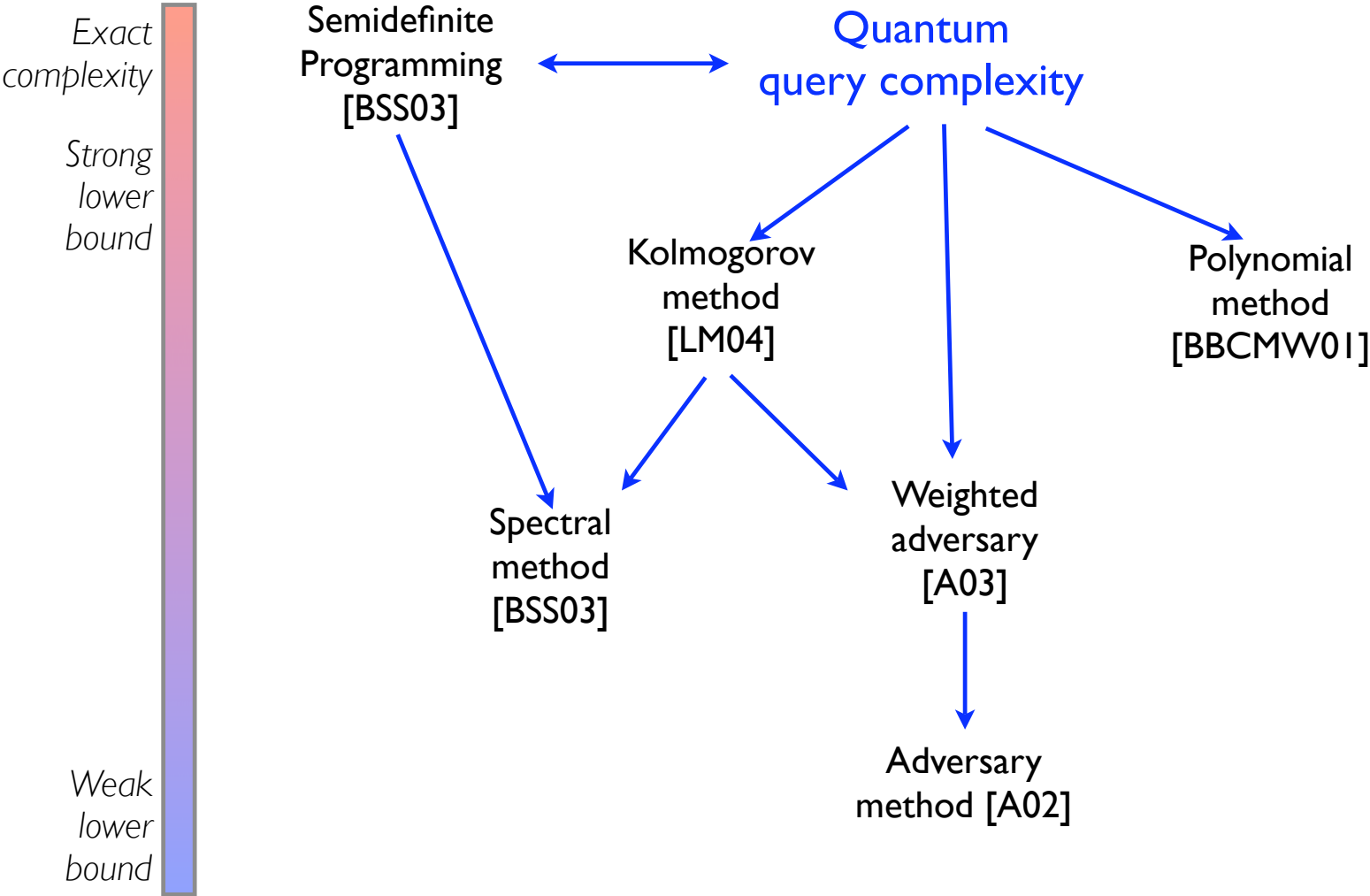
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Adversary method lower bounds



Summary of this talk

<p>Deterministic method</p> <ul style="list-style-type: none"> - General statement - Combinatorial bound 	$DT(f) \geq \frac{ R }{\max_i R_i }$ $\geq \max\left\{\frac{m}{l}, \frac{m'}{l'}\right\}$
<p>Quantum method</p> <ul style="list-style-type: none"> - General statement - Combinatorial bound [A02] 	$Q_\varepsilon(f) \geq \frac{c_\varepsilon R }{\sum_i Progress_t(i)}$ $\geq \frac{c_\varepsilon}{2} \sqrt{\frac{mm'}{ll'}}$
<p>Weighted method [A03, LM04]</p>	$Q_\varepsilon(f) \geq \frac{c_\varepsilon}{2} \min_{x,y,i} \frac{\sqrt{p(x)p(y)p'_{x,i}(y)p'_{y,i}(x)}}{q(x,y)}$
<p>Spectral method [BSS03]</p>	$Q_\varepsilon(f) \geq \frac{c_\varepsilon}{2} \frac{\ \Gamma\ }{\max_i \ \Gamma_i\ }$

Classical decision tree model

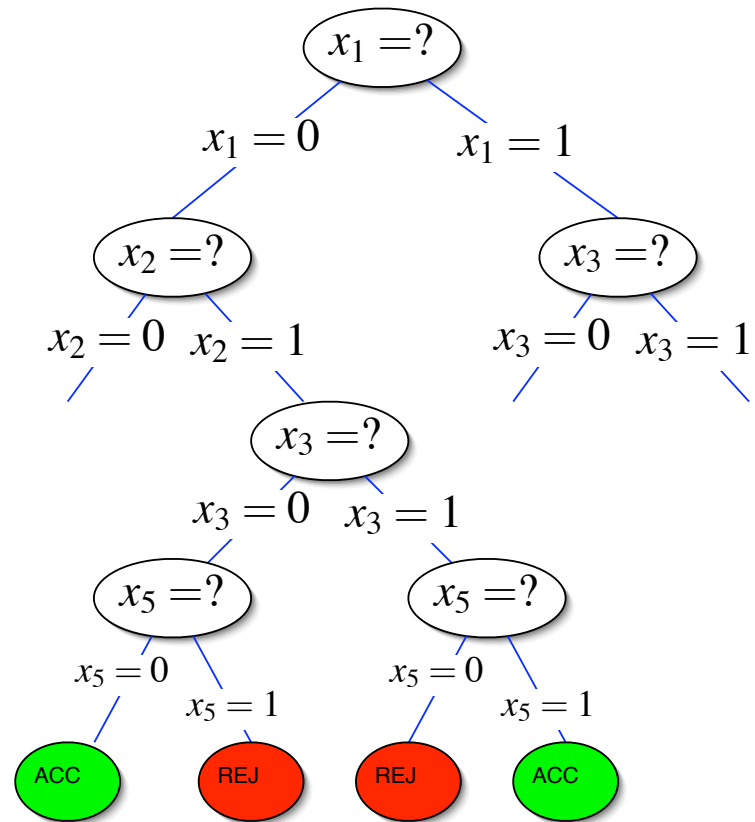
To compute a boolean function
 $f : \{0,1\}^n \rightarrow \{0,1\}$,

Model: decision tree

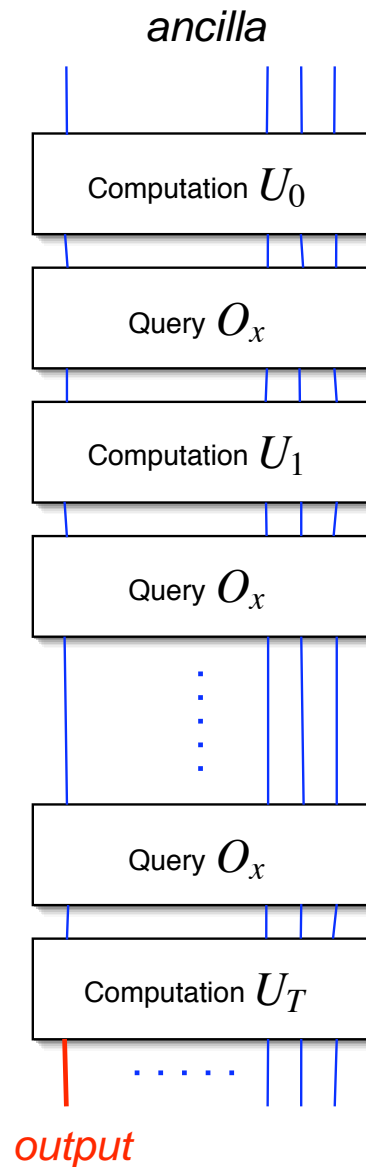
Cost: Number of queries to
input

Query complexity of f :

depth of shallowest decision
tree for f

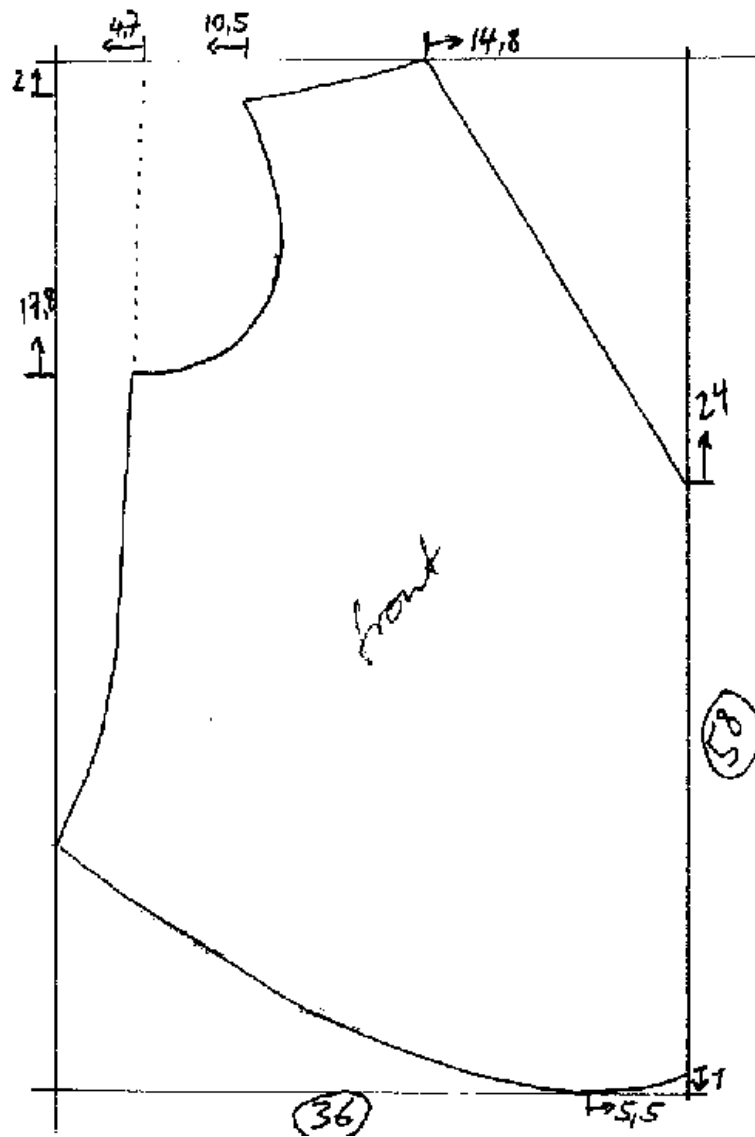


Quantum query model



- **Query:** Unitary transformation O_x that maps $|i, b\rangle$ to $|i, b \oplus x_i\rangle$
- **Computation :**
$$|\psi_T\rangle = U_T O_x \cdots O_x U_0 |0\rangle$$
- **Output:** Measure 1st qubit of $|\psi_T\rangle$
- Error probability at most $1/3$
- Same model with **stochastic matrices** for **randomized query complexity**

Estimating number of pieces by their size

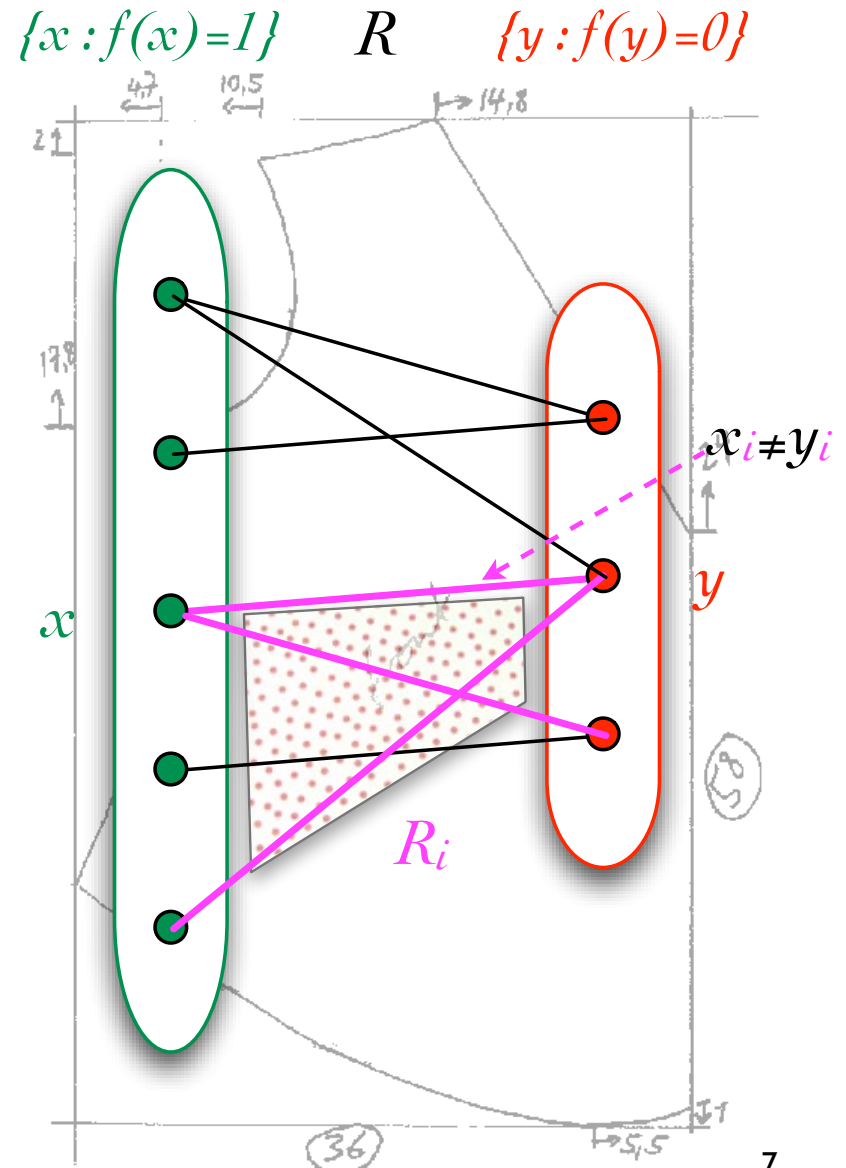


Adversary method, deterministic setting

Goal: separate $\{x : f(x)=1\}$ from $\{y : f(y)=0\}$ with minimum queries to input bits

- Subrelation R_i : pairs for which query i is useful
- Queries \geq number of R_i needed to cover R .

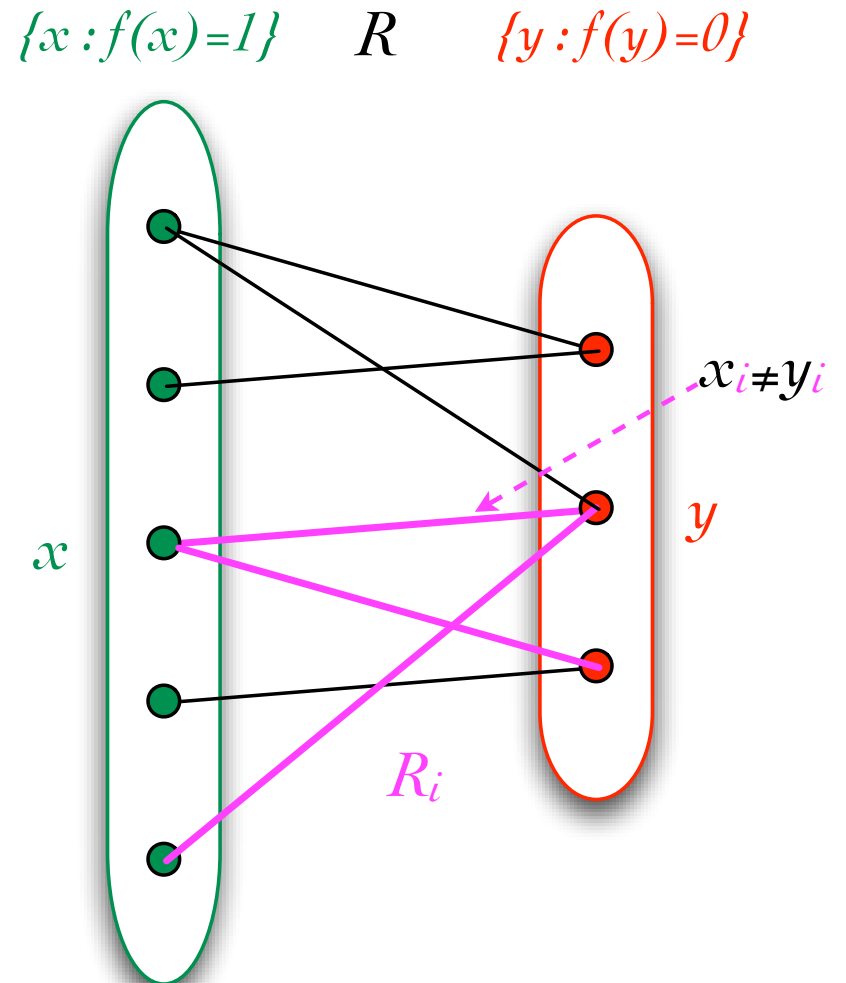
$$DT(f) \geq \frac{|R|}{\max_i |R_i|}$$



Deterministic lower bound

- **left degree** of $R \geq m$
right degree $\geq m'$
- **left degree** of all $R_i \leq l$
right degree $\leq l'$
- $|R| \geq m |X|, m' |Y|$
- $|R_i| \leq l |X|, l' |Y|$

$$\begin{aligned}
 DT(f) &\geq \frac{|R|}{\max_i |R_i|} \\
 &\geq \max\left\{\frac{m}{l}, \frac{m'}{l'}\right\}
 \end{aligned}$$



Progress on x, y, i after a query

At time t , progress towards distinguishing $(x, y) \in R_i$ by making one query :

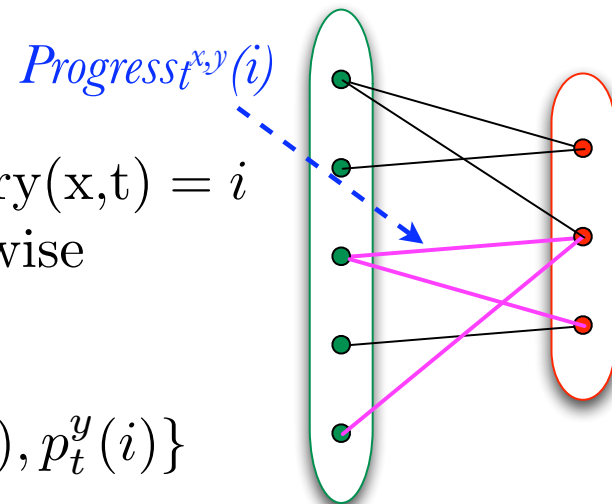
- Deterministic case

$$DProgress_t^{x,y}(i) = \begin{cases} 1 & \text{if query}(x,t) = i \\ 0 & \text{otherwise} \end{cases}$$

- Randomized case

$$RProgress_t^{x,y}(i) = 2 \min\{p_t^x(i), p_t^y(i)\}$$

- Quantum case ? (next slide)



Quantum progress

State of the system after query

t , on input x is written $|\psi_t^x\rangle$

Fix $(x, y) \in R$.

- At beginning, $\langle \psi_0^x | \psi_0^y \rangle = 1$

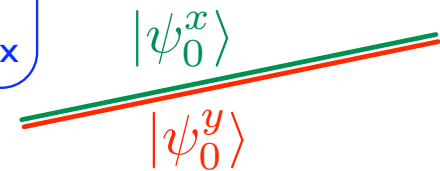
- At each time step,

$$|\langle \psi_t^x | \psi_t^y \rangle - \langle \psi_{t+1}^x | \psi_{t+1}^y \rangle| \leq 2 \sum_{i: x_i \neq y_i} \sqrt{p_t^x(i) p_t^y(i)}$$

- At the end, $|\langle \psi_T^x | \psi_T^y \rangle| \leq 2\sqrt{\varepsilon(1-\varepsilon)}$

$$T \cdot \sum_i \underbrace{2\sqrt{p_t^x(i) p_t^y(i)}}_{\text{Progress}_{x,y}^t(i)} \geq 1 - 2\sqrt{\varepsilon(1-\varepsilon)}$$

Amplitude² of $|i\rangle$
in query t , input x



C_ε

Quantum progress on x, y, i after a query

At time t , progress towards distinguishing $(x, y) \in R_i$ by making one query :

- Deterministic case

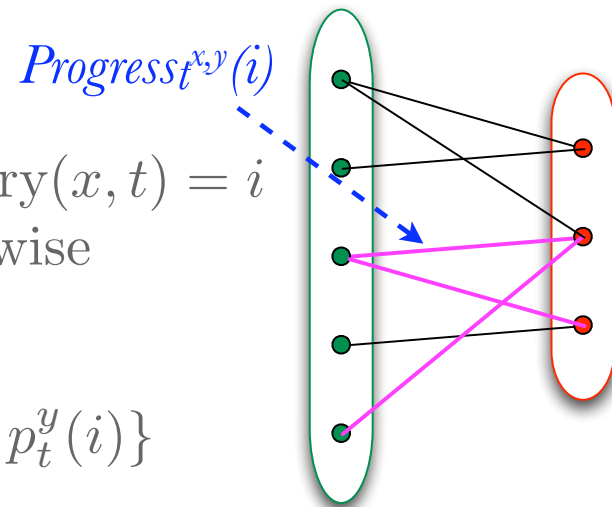
$$DProgress_t^{x,y}(i) = \begin{cases} 1 & \text{if query}(x, t) = i \\ 0 & \text{otherwise} \end{cases}$$

- Randomized case

$$RProgress_t^{x,y}(i) = 2 \min\{p_t^x(i), p_t^y(i)\}$$

- **Quantum case**

$$QProgress_t^{x,y}(i) = 2\sqrt{p_t^x(i)p_t^y(i)}$$



Ambainis' unweighted method

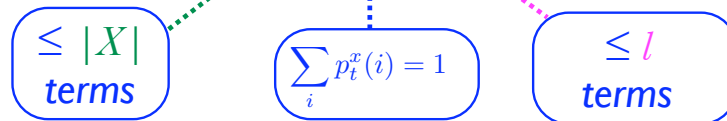
Claim: $\sum_i Progress_t(i) \leq 2\sqrt{l|X| \cdot l'|Y|}$

Proof: $\sum_i Progress_t(i)$

$$= \sum_{x,y,i} Progress_t^{x,y}(i)$$

$$= 2 \sum_{x,y,i} \sqrt{p_t^x(i)p_t^y(i)}$$

$$\leq 2 \sqrt{\sum_x \sum_i \sum_y p_t^x(i)} \sqrt{\sum_y \sum_i \sum_x p_t^y(i)} \quad \text{Cauchy Schwarz}$$



$$\leq 2\sqrt{l|X| \cdot l'|Y|}$$

Ambainis' unweighted method

Claim:
$$\sum_i Progress_t(i) \leq 2\sqrt{l|X| \cdot l'|Y|}$$

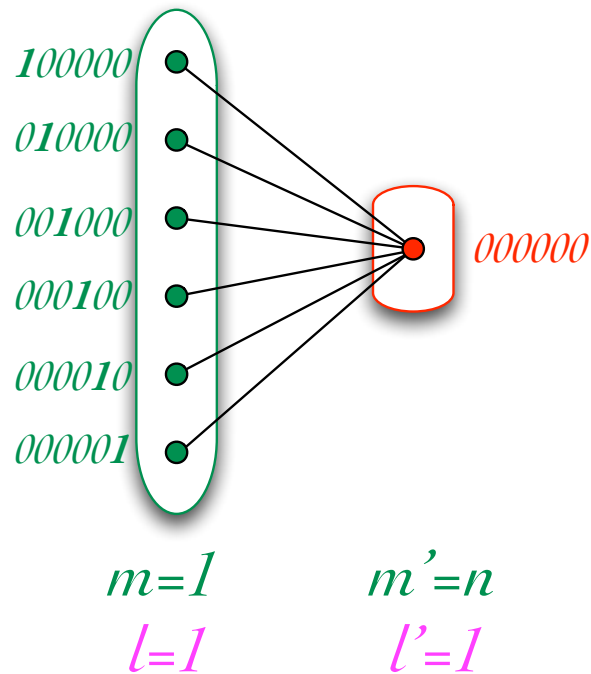
Corollary:
$$Q_\varepsilon(f) \geq \frac{c_\varepsilon |R|}{\sum_i Progress_t(i)}$$

$$\geq \frac{c_\varepsilon \sqrt{m|X| \cdot m'|Y|}}{2\sqrt{l|X| \cdot l'|Y|}}$$

$$\begin{aligned} |R| &\geq \max\{m|X|, m'|Y|\} \\ &\geq \sqrt{m|X| \cdot m'|Y|} \end{aligned}$$

$$Q_\varepsilon(f) \geq \frac{c_\varepsilon}{2} \sqrt{\frac{mm'}{ll'}}$$

Example: OR function



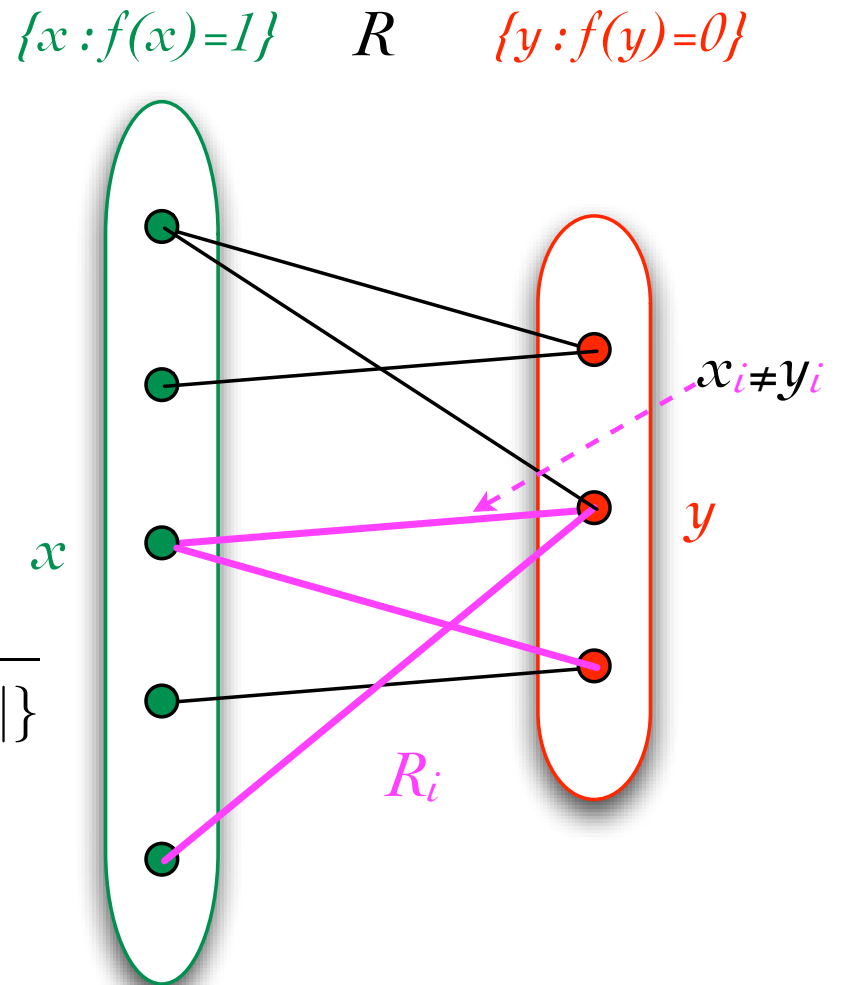
- **Deterministic queries**
 $\geq \max\{m/l, m'/l'\}$
 $= n$
- **Quantum queries**
 $\geq \sqrt{\frac{mm'}{ll'}}$
 $= \sqrt{n}$

Unweighted “ \mathcal{I}_{max} ” lower bound

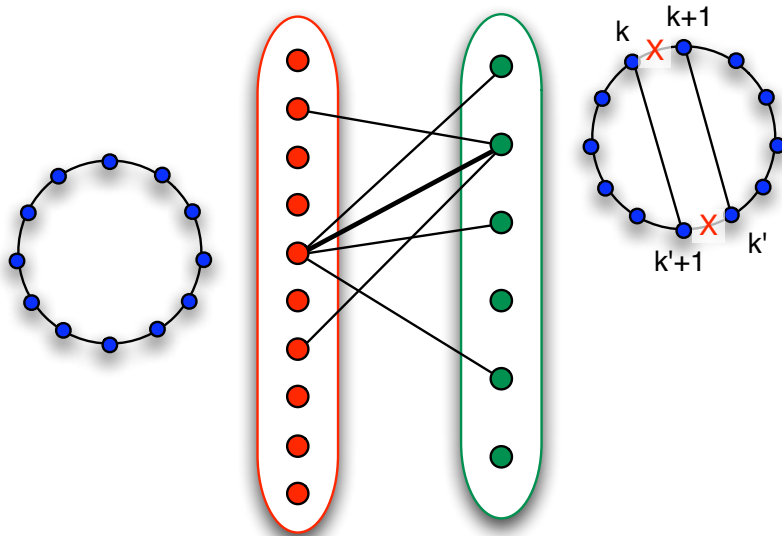
- left degree of $R \geq m$
right degree $\geq m'$
- left degree of $R_i \leq l_i$
right degree $\leq l'_i$
- $|R| \geq m |X|, m' |Y|$

$$\sum_i \text{Progress}_t(i) \leq 2\sqrt{\max_i \{l_i |X| \cdot l'_i |Y|\}}$$

$$Q_\varepsilon(f) \geq \frac{c_\varepsilon}{2} \sqrt{\frac{mm'}{\max_i \{l_i l'_i\}}}$$



Example: Connectivity



$$m \approx n^2 \quad m' \approx n^2$$

$$(x) \quad l_i \approx n \quad l'_i = 1$$

$$(I) \quad l_i \approx 1 \quad l'_i = n$$

- Deterministic queries

$$\geq \max \{m/l, m'/l'\}$$

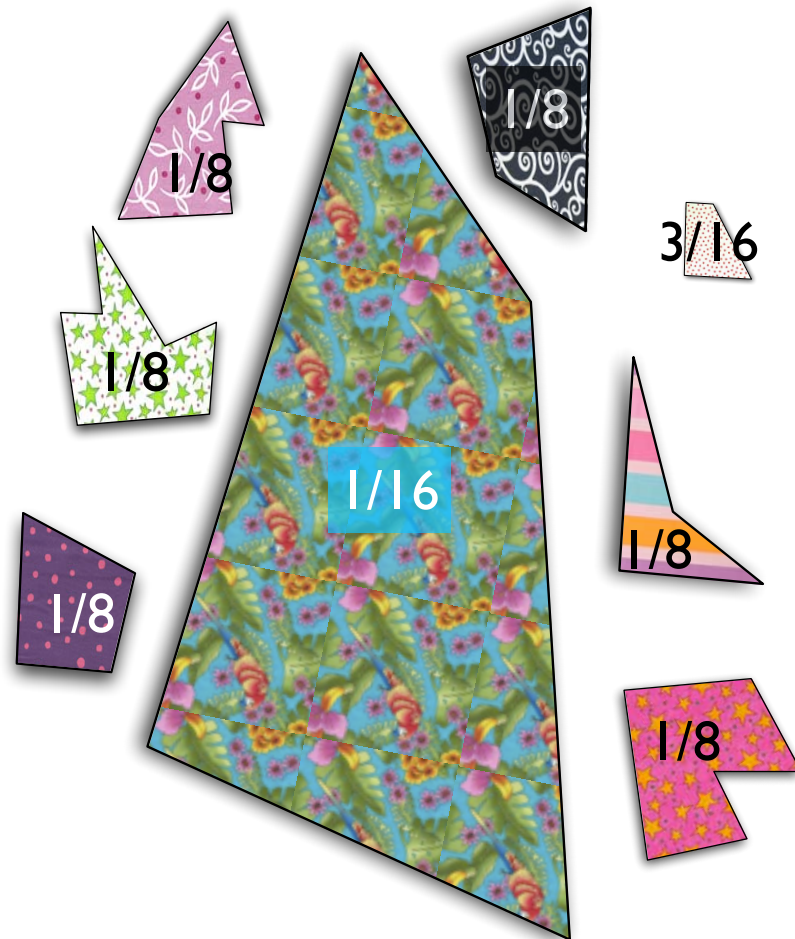
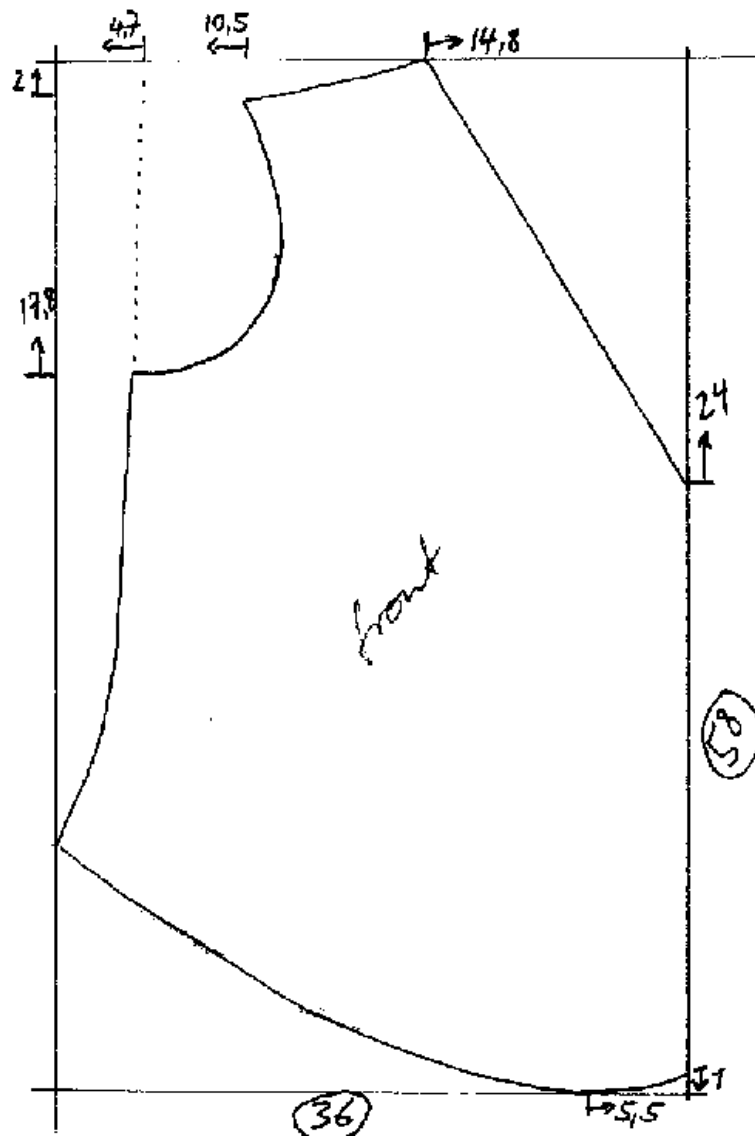
$$= n^2$$

- Quantum queries

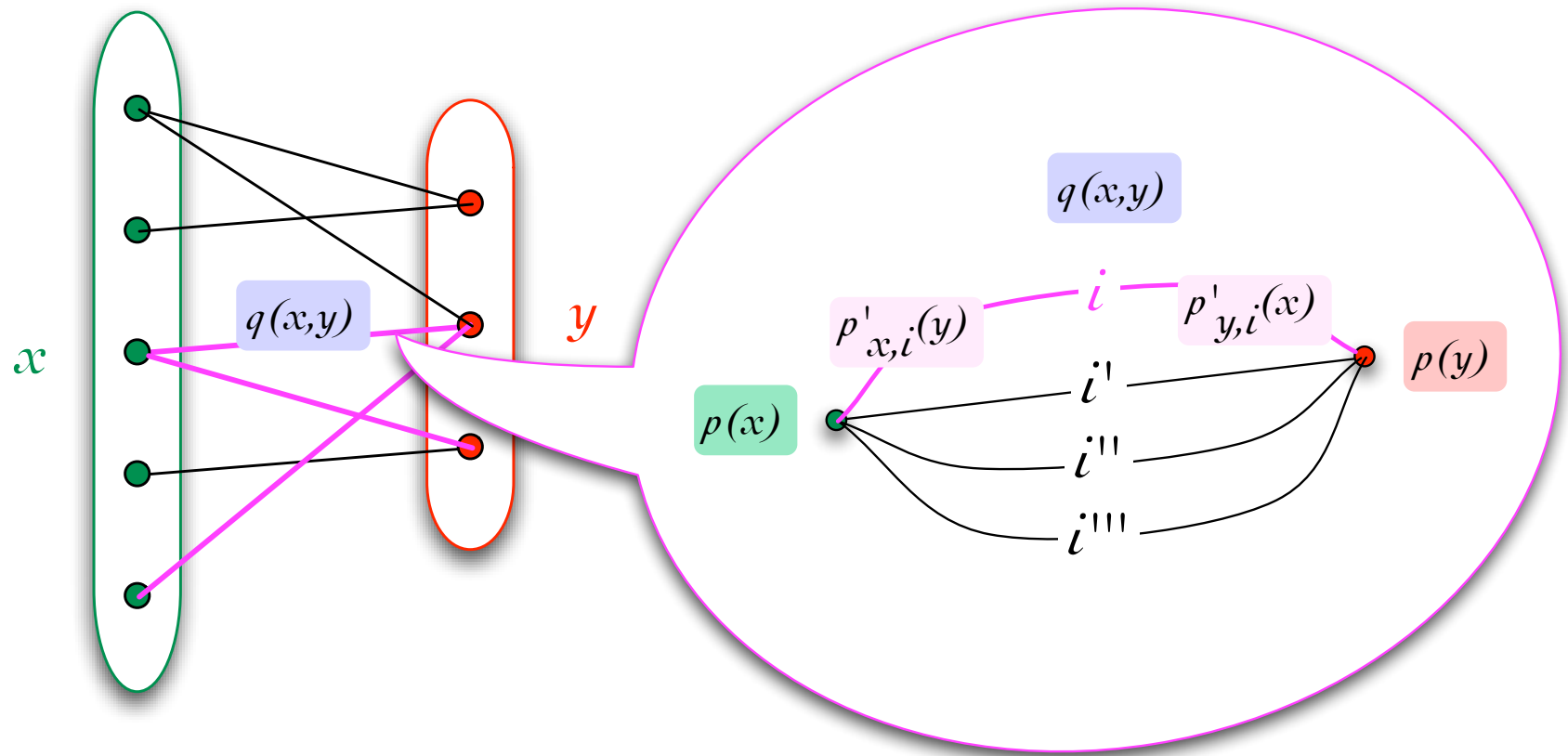
$$\geq \sqrt{\frac{mm'}{ll'}}$$

$$= n^{3/2}$$

Weighted method



Weight scheme for adversary method



$$\sum_{x,y \in R} q(x,y) = 1, \quad \sum_{x \in X} p(x) = 1, \quad \sum_{y \in Y} p'_{x,i}(y) = 1$$

Weighted adversary method

Consider three distributions over pairs $(x,y) \in R$

$$\begin{aligned}
 P(x,y) &= \sum_i p(x) p_t^x(i) p'_{x,i}(y), \\
 P'(x,y) &= \sum_i p(y) p_t^y(i) p'_{y,i}(x), \\
 Q(x,y) &= q(x,y)
 \end{aligned}$$

Claim: $\exists (x,y) \in R \sum_i \text{Progress}_t^{x,y}(i)$

$$\leq 2 \max_i \frac{q(x,y)}{\sqrt{p(x)p'_{x,i}(y)p(y)p'_{y,i}(x)}}$$

$$\begin{aligned}
 \sum_{x,y} \sqrt{P(x,y)P'(x,y)} &\leq \sqrt{\sum_{x,y} P(x,y) \sum_{x,y} P'(x,y)} \\
 &\leq 1 = \sum_{x,y} Q(x,y)
 \end{aligned}$$

Proof: $\exists x,y \sqrt{P(x,y)P'(x,y)} \leq Q(x,y)$

$$\exists x,y \sum_i \sqrt{p_t^x(i)p_t^y(i)} \leq \frac{q(x,y)}{\min_i \sqrt{p(x)p(y)p'_{x,i}(y)p'_{y,i}(x)}}$$

Weighted adversary method

Consider three distributions over pairs $(x, y) \in R$

$$\begin{aligned} P(x, y) &= \sum_i p(x) p_t^x(i) p'_{x,i}(y), \\ P'(x, y) &= \sum_i p(y) p_t^y(i) p'_{y,i}(x), \\ Q(x, y) &= q(x, y) \end{aligned}$$

Claim: $\exists (x, y) \in R \sum_i \text{Progress}_t^{x,y}(i)$

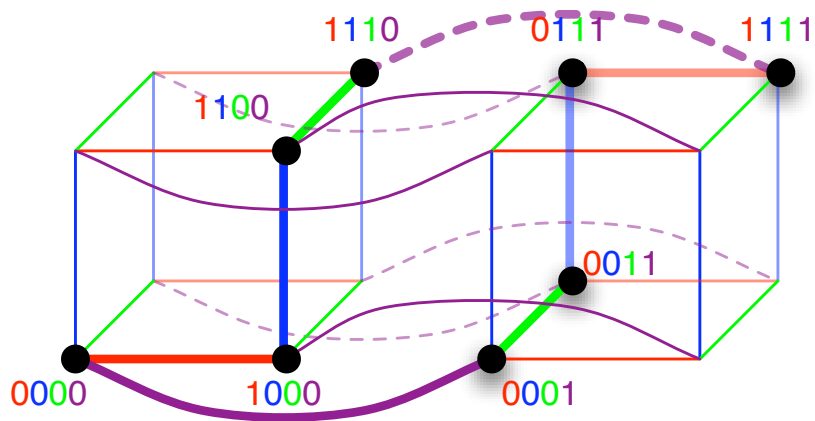
$$\leq 2 \max_i \frac{q(x, y)}{\sqrt{p(x) p'_{x,i}(y) p(y) p'_{y,i}(x)}}$$

Corollary:

Recall that $\forall (x, y) \in R \quad Q_\varepsilon(f) \geq \frac{c_\varepsilon}{\sum_i \text{Progress}_t^{x,y}(i)}$

$$Q_\varepsilon(f) \geq \frac{c_\varepsilon}{2} \min_{x,y,i} \frac{\sqrt{p(x) p(y) p'_{x,i}(y) p'_{y,i}(x)}}{q(x, y)}$$

Example: Ambainis' function



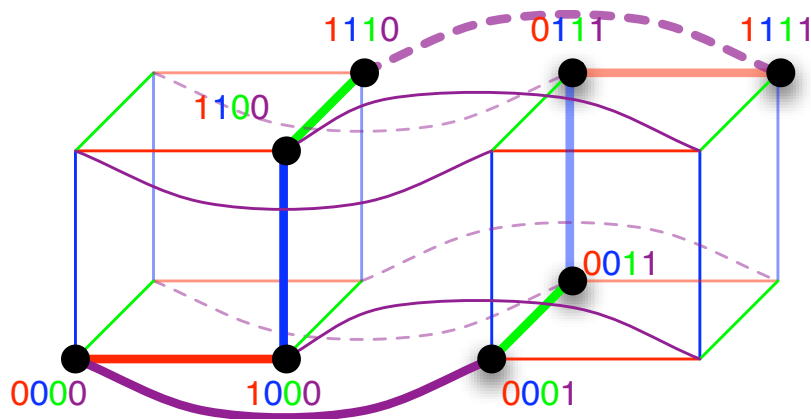
$$f(x_1x_2x_3x_4) = \begin{cases} 1 & \text{if } x_1 \leq x_2 \leq x_3 \leq x_4 \\ 1 & \text{if } x_1 \geq x_2 \geq x_3 \geq x_4 \\ 0 & \text{otherwise} \end{cases}$$

$$\begin{array}{l}
 \boxed{q(x,y)} \\
 \begin{array}{l}
 3/80 \text{ --- } | 0 0 | \\
 3/80 \text{ --- } 0 | 0 | \\
 1/40 \text{ --- } | | 0 | \\
 1/40 \text{ --- } 0 0 | 0
 \end{array} \\
 0 0 0 | \begin{array}{l} \diagup \\ \diagdown \\ \diagdown \\ \diagdown \end{array} \\
 !! \checkmark \checkmark
 \end{array}$$

$$\boxed{p(x)} = \boxed{p(y)} = 1/8$$

$$Q_\epsilon(f) \geq \frac{c_\epsilon}{2} \min_{x,y,i} \frac{\sqrt{p(x)p(y)p'_{x,i}(y)p'_{y,i}(x)}}{q(x,y)}$$

Example: Ambainis' function



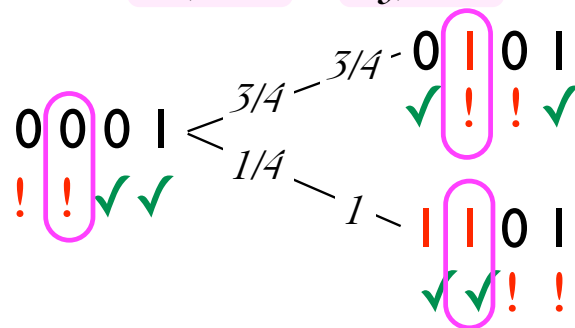
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$$p'_{x,i}(y)$$

$$p'_{y,i}(x)$$

$$q(x,y)$$

$$\frac{\sqrt{p'_{x,i}(y)p'_{y,i}(x)}}{q(x,y)} \sqrt{p(x)p(y)}$$



$$3/80$$

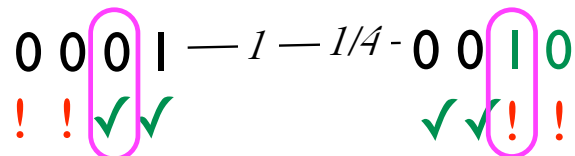
$$20$$

$$1/8$$

$$1/40$$

$$20$$

$$1/8$$



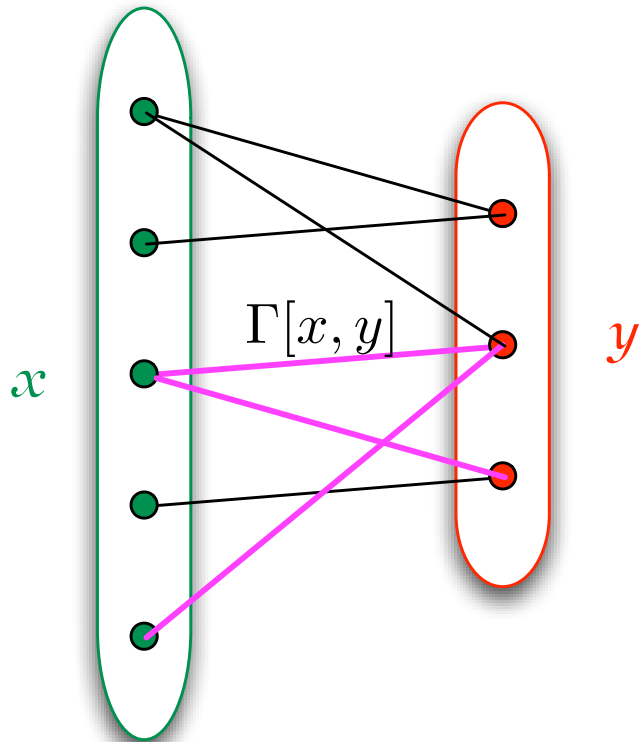
$$1/40$$

$$20$$

$$1/8$$

$$Q_\epsilon(f) \geq \frac{c_\epsilon}{2} \min_{x,y,i} \frac{\sqrt{p(x)p(y)p'_{x,i}(y)p'_{y,i}(x)}}{q(x,y)} = 20/8 = 5/2$$

Spectral method



Nonnegative weights matrix

$$\Gamma[x, y]$$

$$\Gamma[x, y] = 0 \text{ if } f(x) = f(y)$$

$$\Gamma_i[x, y] = \begin{cases} 0 & \text{if } x_i = y_i \\ \Gamma[x, y] & \text{otherwise} \end{cases}$$

$$\|\Gamma\| = \max_{\substack{u, v \\ |u|=|v|=1}} |u^* \Gamma v|$$

$$Q_\varepsilon(f) \geq \frac{c_\varepsilon}{2} \frac{\|\Gamma\|}{\max_i \|\Gamma_i\|}$$

Idea: $q(x, y)$ derived from

$p(x), p(y)$ derived from u, v maximizing $|u^* T v|$

$p'_{y,i}(x), p'_{x,i}(y)$ derived from u_i, v_i maximizing $|u_i^* T v_i|$