## Quantum Advantages for Approximate Combinatorial Optimization

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## Combinatorial Optimization

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－Combinatorial optimization is hard
－Incredibly successful heuristics（for approximation）
－Can quantum computers help？
Certification of an optimal TSP tour through 85，900 cities
 Daniel G．Espinoza ${ }^{\text {e }}$ 『，Marcos Goycoolea ${ }^{\dagger}$ 『，Keld Helsgaun ${ }^{9}$ 『

## APPROXIMATION HARDNESS

－MAX－CUT is APX－hard
－Unless P＝NP，there exists no poly－time algorithm that computes a solution with more than $N=\frac{16}{17} N_{\text {opt }}$ cuts for any MAX－CUT instance［Håstad］


## FORMULA COLORING

－Generalization of graph coloring
－$\left(z_{1} \neq z_{2}\right) \wedge\left(\left(z_{1}=z_{3}\right) \rightarrow\left(z_{2}=z_{4}\right)\right)$
－NP－complete
－Even hard to approximate！［Kearns］

INTEGER LINEAR PROGRAM（ILP）

$$
\min _{x \in \mathbb{T}^{n}} \mathbf{c} \cdot \mathbf{x}
$$

subject to linear constraints

## A Provable Approximation Advantage

"A fault tolerant quantum computer can approximate certain combinatorial optimization problems super-polynomially more efficiently than a classical computer." [Pirnay]


## Computational Problems and Models



Private key must be hard coded!


Deterministic Finite Automaton (DFA)


POLYNOMIAL REDUCTION


## A bit of learning theory

CONCEPT CLASS


## CONSISTENCY PROBLEM

## Con( $C, H$ )

Instance: A set of labeled examples
$S=\{(x, c(x)) \mid x \in X\}$
Solution: Minimal-size $h \in H$ which is consistent with $S$
define: opt $_{\text {Con }}(S):=|h|$

## OCCAM'S RAZOR

## compression

parameter

For a sample set $S$ of size $|S|=\tilde{\mathcal{O}}\left(\frac{1}{\epsilon}+\left[\frac{n^{\alpha}}{\epsilon}\right]^{\frac{1}{1-\beta}}\right)$
"approximation gap"
 any $h \in H$ consistent with $S$ which also satisfies $|h| \leq \operatorname{opt}_{\text {Con }}(S)^{\alpha}|S|^{\beta}$ achieves $\operatorname{error}(h):=\mathbb{P}_{x}[h(x) \neq c(x)] \leq \epsilon$ with high probability. Where $\alpha \geq 1$ and $0 \leq \beta<1$
[Blumer]

## A Provable Approximation Advantage

STRATEGY

- Classical hardness of inverting RSA
- Hardness of approximation for Con(C-RSA, $H$ ) yia Occam's razor
- Approximation preserving reduction to
Con(DFA-RSA, DFA) and then FC-RSA [Kearns]
- approximation-preserving reduction to ILP-RSA [Pirnay]
- Efficient quantum algorithm for approximating ILP-RSA [Pirnay]

With sample size
$|S|=\operatorname{poly}\left(n, \epsilon^{-1}\right)$ any $h \in H$
that is consistent with $S$, s.t.
$|h| \leq \operatorname{opt}_{\mathrm{Con}}(S)^{\alpha}|S|^{\beta}$
achieves error $\leq \epsilon$.

- Learning C-RSA by $H$ can be seen as an approximation task: Approximate opt ${ }_{\text {Con }}(S)$
- Approximately learning a C-RSA circuit enables one to break RSA ! [Alexi]
- $|h| \mapsto$ (partitions)
- $S \mapsto$ FC-graph


```
min c
```

$x \in \mathbb{Z}^{n}$
subject to
constraints

## A Provable Approximation Advantage



Figure 5 in [Pirnay]

## ILP-RSA

By our construction, we get the integer linear programming problem ILP $_{F}$

## $\operatorname{minimize} \sum_{1 \leq i \leq M} w_{i}^{(\text {color } l \text { is being used) }}$

> indicator variable
subject to the following constraints,

$$
\begin{array}{lr}
\text { for all } u, i \in\{1, \ldots, M\}, & \left(x_{u, i}=1\right) \Longleftrightarrow\left(\hat{z}_{u}=i\right), \\
\text { for all } u \in\{1, \ldots, M\}, & \text { only one color per variable: } \sum_{i=1}^{M} x_{u, i}=1, \\
\text { for all } u, i \in\{1, \ldots, M\}, & \text { count colors: } x_{u, i} \leq w_{i}, \\
\text { for all } Q \text { clauses }\left(z_{u} \neq z_{v}\right) \text { and all } i \in\{1, \ldots, M\}, & x_{u, i}+x_{v, i} \leq 1, \\
\text { for all } R \text { clauses }\left(\left(z_{u} \neq z_{v}\right) \vee\left(z_{k}=z_{l}\right)\right) \text { with } j \in\{1, \ldots, R\}, & \left(a_{j}=1\right) \Longleftrightarrow\left(\hat{z}_{k}=\hat{z}_{l}\right),
\end{array} \quad \begin{aligned}
&\left(b_{j}=1\right) \Longleftrightarrow\left(\hat{z}_{u} \neq \hat{z}_{v}\right), \\
& s_{j}=\left(a_{j} \vee b_{j}\right), \\
& s_{j} \geq 1,
\end{aligned}
$$

and $w_{i}, x_{u, i}, a_{j}, b_{j}, s_{j} \in\{0,1\}$ and $1 \leq \hat{z}_{u}, \hat{z}_{v}, \hat{z}_{k}, \hat{z}_{l} \leq M$.

## Classical Hardness of Approximation

Theorem V. 12 (Classical hardness of approximation for integer linear programming). Assuming the hardness of inverting the RSA function, there exists no classical probabilistic polynomial-time algorithm that on input an instance ILP $_{F}$ of ILP-RSA finds an assignment of the variables in $\mathrm{ILP}_{F}$ which satisfies all constraints and approximates the size opt $\mathrm{ILP}^{\left(\mathrm{ILP}_{F}\right)}$ ) of the optimal solution by

$$
\begin{equation*}
\sum_{1 \leq i \leq M} w_{i} \leq \text { opt }_{\mathrm{ILP}}\left(\operatorname{ILP}_{F}\right)^{\alpha}\left|\operatorname{ILP}_{F}\right|^{\beta} \tag{46}
\end{equation*}
$$

for any $\alpha \geq 1$ and $0 \leq \beta<1 / 4$.

## An Efficient Quantum Algorithm

## Algorithm 1: Approximate the solution of Con(C-RSA, BC)

```
Input : A labeled sample S of C-RSA
Output : The description of a Boolean circuit consistent with S
Pick any example s\inS and read e,N from it;
Run Shor's algorithm [1] to factor N and retrieve p and q;
Run the extended Euclidean algorithm to compute d, such that d\timese=1 mod (p-1)(q-1);
// Note that at this point, d is the secret RSA exponent.
Output the description of a Boolean circuit that, on input binary (powers 
    together for which the bit d}\mp@subsup{d}{i}{}=1\mathrm{ (thereby hard-wiring d into the circuit), using the iterated products technique [33] and outputs the
    LSB of the result.
```

Move along the chain of reductions...

Theorem V. 16 (Quantum efficiency for ILP-RSA). There exists a polynomial-time quantum algorithm that, on input an instance $\operatorname{ILP}_{F_{S}}$ of ILP-RSA, finds a variable assignment A that satisfies all constraints and for which the objective function is bounded as

$$
\sum_{1 \leq i \leq M} w_{i} \leq o p t_{\mathrm{ILP}}\left(\operatorname{ILP}_{F_{S}}\right)^{\alpha}
$$

for all $\operatorname{ILP}_{F_{S}}$ and for some $\alpha \geq 1$.

## Conclusion

- Constructive quantum advantage for approximate optimization
- Opens up new problems to study with actual quantum optimization algorithms (QAOA)
- Alternative proofs via the PCP theorem possible [Szegedy]
- Opens up the path towards more practical advantagebearing instances


Slides at: frederikwil.de/hqcc2023

