

Super-Polynomial Lower Bound of TSP Extended Formula

Soham Chatterjee

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Introduction

Definition (Travelling Salesman)

Given a graph $G = (V, E)$, $S \subseteq V$ and weights $w : E \rightarrow \mathbb{R}$ find minimum weight cycle which visits every vertex of S exactly once.

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We will focus on $S = V$.

- We know Traveling Salesman Problem is NP-complete.
- In [Yannakakis, 1988, STOC] he proved every symmetric LP for the TSP has exponential size.
- Here we will show TSP admits no polynomial-size LP.
- This proof also shows unconditional super-polynomial lower bound on the number of inequalities.
- Therefore it is impossible to prove $P = NP$ by means of a polynomial size LP.

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Preliminaries

Definitions

Let $P = \{x \in \mathbb{R}^n \mid Ax \leq b\} = \text{conv}(V)$ is a polytope with $A \in \mathbb{R}^{m \times d}$, $b \in \mathbb{R}^m$ and $V \subseteq \mathbb{R}^d$. We will consider V as the characteristic vector for all hamiltonian paths.

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Lemma

Let P, Q and F be polytopes. Then the following holds:

- (i) If F is an extension of P then $xc(F) \geq xc(P)$.*
- (ii) If F is a face of Q then $xc(Q) \geq xc(F)$.*

Slack Matrix

Definition

Let $P = \{x \in \mathbb{R}^n \mid Ax \leq b\} = \text{conv}(V)$ is a polytope with $A \in \mathbb{R}^{m \times d}$, $b \in \mathbb{R}^m$ and $V \subseteq \mathbb{R}^d$. Let $V = \{v_1, \dots, v_n\}$. Then $S \in \mathbb{R}_0^{m \times n}$ is called the slack matrix of P wrt $Ax \leq b$ and V where

$$S(i,j) = b_i - A_i v_j$$

Some times we may refer to the submatrix of slack matrix induced by rows corresponding to facets as the slack matrix of P denoted by $S(P)$.

Some Polytopes

- $TSP(n)$ is the traveling salesman polytope for $K_n = (V_n, E_n)$. Let $C \subseteq E_n$ denotes a tour of K_n . Then χ^C denotes the characteristic vector of C . Then

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$$IND(G) := \text{conv}\{\chi^S \mid S \text{ is independent set of } G\}$$

- The correlation polytope $COR(n)$ is

$$COR(n) := \text{conv}\{bb^T \mid b \in \{0, 1\}^n\}$$

Proof Flow

Theorem

$$xc(TSP(n)) = 2^{\Omega(n^{\frac{1}{4}})}$$

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Step 2: For all n , $\exists$ graph G_n with n vertices such that $xc(IND(G_n)) \geq xc(COR(n'))$
where $n' = n^{\frac{1}{d}}$ for some $d > 1$.

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Step 1: First we will prove $xc(COR(n)) = 2^{\Omega(n)}$

Step 2: For all n , $\exists$ graph G_n with n vertices such that $xc(IND(G_n)) \geq xc(COR(n'))$
where $n' = n^{\frac{1}{d}}$ for some $d > 1$.

Step 3: For any n -vertex graph G , $IND(G)$ is linear projection of a face of $TSP(k)$
where $k = O(n^2)$.

Covering Bound of Matrix and Non-negative Factorization

Covering Bound of Matrix with Rectangles

- Let $M \in \{0,1\}^{n \times n}$ matrix.

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- A monochromatic rectangle R in M means a submatrix N of M whose all entries are 1.
- A collection of rectangles $\mathcal{C}$ covers M if their union covers all the nonzero entries of M .
- $|\mathcal{C}|$ is called a covering bound of M . $\text{Cov}(X) = \min\{|\mathcal{C}| : \mathcal{C} \text{ covers } M\}$

Covering Bound of Simple Matrix

Consider A matrix X of dimension $2^n \times 2^n$ where the rows and columns are indexed by strings from $\{0, 1\}^n$. Let $X(a, b) = (1 - a^T b)^2$ where $a, b \in \{0, 1\}^n$.

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Theorem (Yannakakis, 1988, STOC)

Every monochromatic rectangle cover of $\text{suppmat}(X)$ has size $2^{\Omega(n)}$ i.e.

$$\text{Cov}(\text{suppmat}(X)) \geq 2^{\Omega(n)}$$

Non-negative Factorization

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Theorem (Factorization Theorem)

For a polytope $P = \{x \mid Ax \leq b\}$ where S is the slack matrix of P the following are equivalent:

- (i) S has non-negative rank at most r .*
- (ii) P has an extension of size at most r .*
- (iii) P has an EF of size at most r .*

We get $\text{xc}(P) = \text{rank}_+(S)$.

Factorization and Covering Bound Relation

For any matrix $M \in \mathbb{R}^{m \times n}$ let $\text{suppmat}(M) \in \{0, 1\}^{m \times n}$ is a matrix where the $(i, j)^{\text{th}}$ element is 1 if $M(i, j) \neq 0$ and otherwise 0.

Theorem (Yannakakis, 1988, STOC)

Let M be any matrix with non-negative real entries. Then

$$\text{rank}_+(M) \geq \text{Cov}(\text{suppmat}(M))$$

Correlation Polytope Lower Bound

Polytope equations

Consider the inner product of two matrices $A, B \in \mathbb{R}^{m \times n}$ be $\langle A, B \rangle = \text{Tr}(A^T B)$

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For all $a \in \{0, 1\}^n$ there are some $b \in \{0, 1\}^n$ such that

$$\langle 2\text{diag}(a) - aa^T, bb^T \rangle = 1$$

$$\begin{aligned} 1 - \langle \text{diag}(a) - aa^T, bb^T \rangle &= 1 - 2\langle \text{diag}(a), bb^T \rangle + \langle aa^T, bb^T \rangle \\ &= 1 - 2a^T b + (a^T b)^2 = (1 - a^T b)^2 \end{aligned}$$

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Remark

Because of above prove for all $b \in \text{COR}(n)$, for all $a \in \{0, 1\}^n$, $\langle 2\text{diag}(a) - aa^T, bb^T \rangle \leq 1$.

Hence let A, b be such that $\text{COR}(n) = \{x \mid Ax \leq b\}$ where (A, b) includes these inequities. So the slack matrix S of $\text{COR}(n)$ contains X .

Lower Bound

Let S is the slack matrix of $COR(n)$. Then S contains the matrix X .

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- By Factorization Theorem $xc(COR(n)) = \text{rank}_+(S)$.
- Since X is submatrix of S we have $\text{rank}_+(S) \geq \text{rank}_+(X)$.
- By Covering-Factorization Relation $\text{rank}_+(X) \geq \text{Cov}(\text{suppmat}(X)) \geq 2^{\Omega(n)}$.

Theorem

$$xc(COR(n)) = 2^{\Omega(n)}.$$

Independent Set Polytope Lower Bound

New Graph Construction

Let fix an n . Now consider the complete graph K_n . Now we will construct a graph $H_n = (V_n, E_n)$ with $O(n^2)$ vertices.

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- Each edge $(i, j) \in K_n$
 - There is a 4-clique on the vertices $\{ij, \hat{ij}, \hat{\hat{i}}j, \hat{\hat{j}}i\}$.
 - The additional edges

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Let F is the face of $IND(H_n)$ containing independent sets which have exactly one vertex from each vertex-clique and one vertex from each edge-clique

$COR(n)$ Inside Independent Set Polytope

Take the linear map $\pi : \mathbb{R}^{V_n} \rightarrow \mathbb{R}^{n \times n}$. Let $\pi(x) = y$. Then

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- S contains a vertex ii if $b_i = 1$ and S contains $\hat{i}$ if $b_i = 0$.

$\chi^S \in F$. So $\pi(\chi^S) = bb^T$. So

$$\pi(F) = COR(n)$$

So $COR(n)$ is a face of $IND(H_n)$.

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- $IND(H_p)$ isomorphic to $IND(G_n)$

$$xc(IND(G_n)) = xc(IND(H_p)) \geq xc(COR(p)) \geq 2^{\Omega(p)} = 2^{\Omega(n^{\frac{1}{2}})}$$

Theorem

For all $n \in \mathbb{N}$ there exists graph G_n , $xc(IND(G_n)) = 2^{\Omega(n^{\frac{1}{2}})}$



TSP Polytope Lower Bound

Theorem (Yannakakis, 1988, STOC)

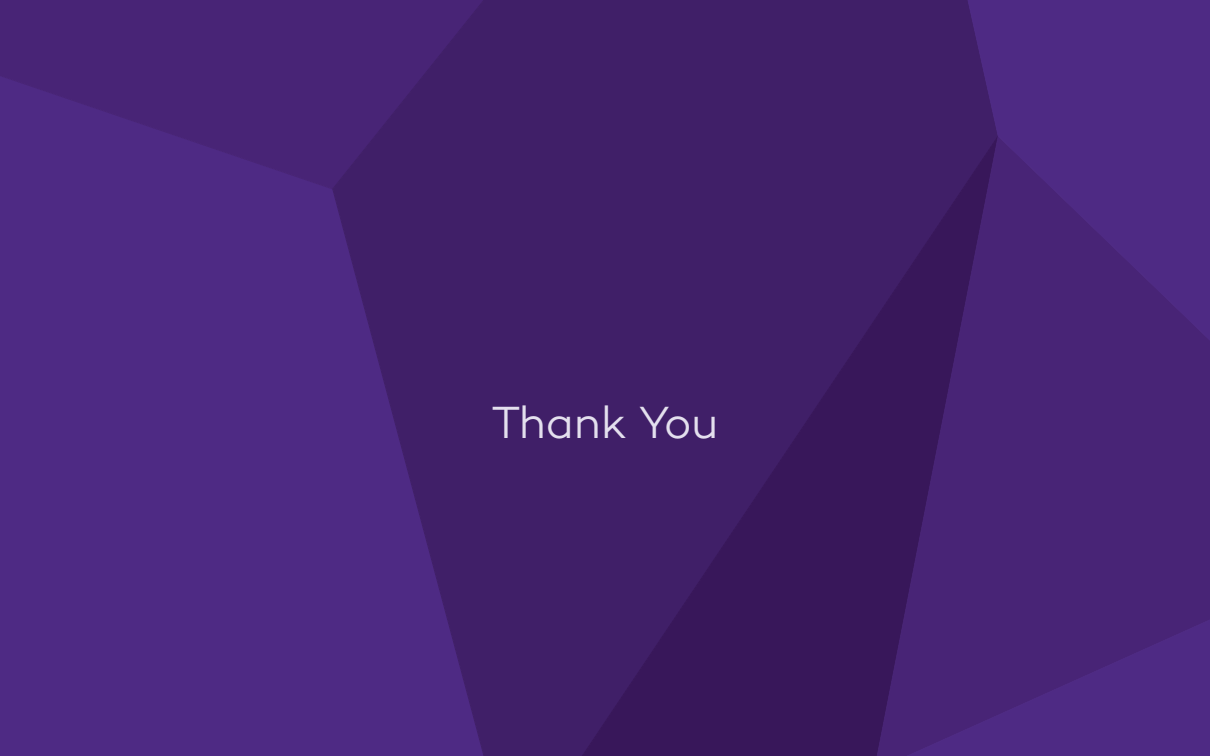
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Theorem (Yannakakis, 1988, STOC)

Every p -vertex graph G , $IND(G)$ is the linear projection of a face of $TSP(n)$ with $n = O(p^2)$.

Therefore

$$xc(TSP(n)) \geq xc(IND(G_p)) = 2^{\Omega(p^{\frac{1}{2}})} = 2^{\Omega(n^{\frac{1}{4}})}$$

The background consists of several overlapping triangles in various shades of purple and blue. The central area is a dark purple triangle, while the surrounding areas are lighter shades of purple and blue, creating a layered, geometric effect.

Thank You