

Complexity of Combinatorial Optimization in Power Law Graphs

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Algorithmic Complexity of Real-World Graphs

- Is combinatorial optimization easy or hard in real-world graphs?
- Real-world graphs possess special properties which may very well make the problems tractable.

Power Law Graphs

Many large-scale real-world graphs exhibit a **power law** distribution.

The number of nodes y_i of a given degree i is proportional to $i^{-\beta}$ where $\beta > 0$ is a constant.

Power Law in Real-World Graphs

Power-law degree distribution has been observed in:

- Internet — 2.1
- World Wide Web — 2.1
- Social networks: movie actors graph — 2.3;
citation graph — 3
- Biological networks: protein-protein interaction
graphs — 2.5.

In most real-world graphs, β ranges between **1** and **4**.

Power-law graphs have emerged as a partial answer to the perennial search for representative real-world graphs in combinatorial optimization.

Combinatorial Optimization in Real-World Power Law Graphs

Practical evidence suggests that combinatorial optimization in real-world power law graphs is **easier** than in general graphs.

Internet

Experiments in Internet measurement graphs show that a simple greedy algorithm that exploits the power law property yields a very good approximation to the **MINIMUM VERTEX COVER** problem. [Park and Lee, 2001]

Social Networks

Experiments show that a simple greedy algorithm that exploits the power law property gives almost optimal solution to the **DOMINATING SET** problem. [Eubank et al. 2004]

Algorithmic Theory of Power Law Graphs

Above results on disparate problems on various real-world power law graphs motivate a coherent and systematic algorithmic theory of optimization in power law graphs.

Power Law Graph Models

Power law graph models have been proposed to capture/explain the empirically observed power-law degree distribution in real-world graphs.

Can be classified into two types:

1. Takes a power-law degree sequence and generates graph instances with this distribution.
2. Arises from attempts to explain the power-law starting from basic assumptions about a growth evolution.

Aiello, Chung, and Lu (ACL) Model

A robust and general model for (undirected) power-law graphs:

the number of vertices y_i with degree i is (roughly) given by $y_i = e^\alpha / i^\beta$,

where e^α is a normalization constant.

The structural properties that is true in this model will be true for **all** graphs with the given degree sequence.

We investigate the complexity of **simple** power-law graphs (no multi-edges or self-loops).

Algorithmic Complexity of Power Law Graphs

Implications of power-law degree distribution to the algorithmic complexity of NP-hard optimization problems.

- Can power-law property alone be sufficient to design polynomial-time algorithms for NP-hard problems on power-law graphs?
- Does the answer depend on the power-law exponent β ?

Worst-Case Complexity

All NP-hard graph-theoretic optimization problems that satisfy an **optimal substructure** property remain **NP-Hard** on simple power law graphs for all $\beta > 0$.

Examples: MINIMUM VERTEX COVER, MAXIMUM INDEPENDENT SET, and MINIMUM DOMINATING SET.

CLIQUE and COLORING (that do not satisfy the optimal substructure property) are also **NP-hard** on simple power law graphs for all $\beta \geq 1$.

Average-Case Complexity

Problems such as VERTEX COVER, DOMINATING SET, can be solved asymptotically **optimally** in dense **random** power law graphs.

For random power law graphs with $\beta < 2$, there exists a polynomial time algorithm that with probability at least $1 - o(1)$ outputs within $o(1)$ of the optimal.

Talk Outline

1. Hardness Results.
2. Average-Case Results.
3. Conclusion.

Optimal Substructure Property

Optimal solution for a problem on given graph is the **union** of the optimal (sub-)solutions on its **maximal connected components**.

Let P be an optimization problem which takes a graph as input.

For **every** input G : **every** optimum solution of P on G should contain an optimum solution of P on **each** of G 's maximal connected components.

MINIMUM VERTEX COVER **satisfies** this property.

MINIMUM COLORING and MAXIMUM CLIQUE **do not** satisfy the property.

Preliminaries: y and d -degree Sequence

Given a graph, we will refer to two types of sequence of integers:

- Y^G — y -degree sequence of G : lists the number of vertices of G with a certain degree i.e., the degree distribution.
- D^G — d -degree sequence of G : lists the degrees of the vertices of G in non-increasing order i.e., the degree sequence of the graph in non-increasing order.

We use expansion of Y , $EXP(Y^G) = D^G$.

β -Graphs

The ACL model of power-law graphs have a particular kind of y -degree sequence which we call as

(β, α) -degree sequence:

Definition 1. [(β, α) -degree sequence] Given $\alpha, \beta \in \mathbb{R}^+$, the y -degree sequence of a graph $G = (V, E)$ is a (β, α) -degree sequence denoted by $Y^{(\beta, \alpha)} = \langle y_1^{(\beta, \alpha)}, \dots, y_m^{(\beta, \alpha)} \rangle$, if $m = \lfloor e^{\alpha/\beta} \rfloor$ and, for $i \in [1, m]$

$$y_i = \begin{cases} \lfloor \frac{e^\alpha}{i^\beta} \rfloor & \text{if } i > 1 \text{ or } \sum_{k=1}^m \lfloor \frac{e^\alpha}{k^\beta} \rfloor \text{ is even} \\ \lfloor e^\alpha \rfloor + 1 & \text{otherwise.} \end{cases}$$

Definition 2. [β -graph] Given $\beta \in \mathbb{R}^+$, a graph $G = (V, E)$ is a β -graph if it is simple and there exists $\alpha \in \mathbb{R}^+$ such that the y -degree sequence of G is a (β, α) -degree sequence.

Embedding Lemma

A key technical lemma that is needed for hardness results.

Shows the existence of a β -graph with a certain property.

Lemma 1. [Embedding Lemma] *Let $G = (V, E)$ be a simple undirected graph and $\beta \in \mathbb{R}^+$. Then there exists a simple undirected graph $G_1 = (V_1, E_1)$ such that G is a set of maximal connected components of G_1 , $|V_1| = \text{poly}(|V|)$ and G_1 is a β -graph. Furthermore, given G , we can construct G_1 in time polynomial in the size of G .*

The difficulty is in proving that such a degree sequence exists.

NP-Hardness

Theorem 1. *Let P be an optimization problem on graphs with the optimal substructure property. If P is NP-hard on (simple) general graphs, then it is also NP-hard on β -graphs for all $\beta > 0$.*

Erdos-Gallai Condition

A degree sequence is called **graphic** if it can be realized as simple graph (no self-loops or multi-edges).

Definition 3. [Eligible Sequence] A sequence of integers $S = \langle s_1, \dots, s_n \rangle$ is eligible if $s_1 \geq \dots \geq s_n$ and, for all $k \in [1, n]$, $f_S(k) \geq 0$, where

$$f_S(k) = k(k-1) + \sum_{i=k+1}^n \min\{k, s_i\} - \sum_{i=1}^k s_i.$$

Lemma 2. [Erdos and Gallai] A sequence of integers S is graphic if and only if it is non-increasing, $\text{tot}(S) = \sum_{i=1}^n s_i$ is even and S is eligible.

The weaker condition $\max_i d_i^2 < \sum_i d_i$ and $\sum_i d_i$ is even only ensures that the degree sequence can be realized as a multi-graph.

Havel and Hakimi Condition

Lemma 3. [Havel and Hakimi] *A sequence of integers $D = \langle d_1, \dots, d_n \rangle$ is graphic if and only if it is non-increasing, and the sequence of values $D' = \langle d_2 - 1, d_3 - 1, \dots, d_{d_1+1} - 1, d_{d_1+2}, \dots, d_n \rangle$ when sorted in non-increasing order is graphic.*

Gives a straightforward algorithm to construct a simple graph from a graphic degree sequence.

Embedding Lemma

Given two degree sequences $S = \langle s_1, \dots, s_n \rangle$ and $T = \langle t_1, \dots, t_m \rangle$ with $n \geq m$, we will denote $S - T = \langle x_1, \dots, x_n \rangle$ with $x_i = s_i - t_i$ if $i \in [1, m]$ and $x_i = s_i$ otherwise.

To prove the existence of the β -graph G_1 :

Given $\beta \in \mathbb{R}^+$, there exists $\alpha_0 \in \mathbb{R}^+$ such that, for all $\alpha \geq \alpha_0$, the sequence $Y = Y^{(\beta, \alpha)} - Y^G$ is the y -degree sequence of a simple graph.

To show this, we will show that the expansion of the prefix $\langle y_1, \dots, y_m \rangle$ of Y is eligible for $\alpha \geq \alpha_1$, and that the expansion of the suffix $\langle y_{m+1}, \dots, y_{m_1} \rangle$ is eligible for $\alpha \geq \alpha_2$, where m is the maximum degree of G and m_1 is the number of elements in $Y^{(\beta, \alpha)}$.

These give a sufficient condition for $EXP(Y)$ to be graphic.

CLIQUE and COLORING

- Given an arbitrary graph G , construct a simple β -graph G_1 containing G as a maximal connected component.

Let $G_2 = G_1 \setminus G$ be the remaining graph.

- Show that problem can be solved optimally in G_2 .
- Construct G_1 (and hence G_2) in such a way that it admits an efficient algorithm.
- Make G_2 to be a simple **bipartite** graph.

Embedding Lemma for $\beta \geq 1$

Lemma 4. *Let $G = (V, E)$ be a simple graph. Then for all $\beta \geq 1$, there exists a β -graph G_1 such that $G_2 = G_1 - G$ is a bipartite graph. The size of G_1 is at most polynomial in the size of G .*

Corollary 1. *CLIQUE, and COLORING are NP-Complete in β -graphs for all $\beta \geq 1$.*

Bipartite Graph Construction

A degree sequence $D = \langle d_1, \dots, d_n \rangle$ with maximum value m is **contiguous** if $y_i > 0$ for all $i \in [1, m]$.

Lemma 5. *Let $D = \langle d_1, \dots, d_n \rangle$ be a degree sequence. If D is contiguous and $d_1 \leq \lfloor n/2 \rfloor$ then there exist a simple bipartite graph $G = (V, E)$ such that $D^G = D$. (*Bipartite-eligible*)*

Proof by Algorithm

Algorithm outputs the graph $G = (S \cup T, E)$:

1. **Assign vertices** to S and T :

Starting from the highest vertex, place vertices successively in S and T whichever set has the lower total degree currently.

2. **Assign edges** between S and T :

Residual degree of a vertex: final degree - “current” degree.

Initially all the vertices have degree 0.

3. **repeat** till possible:

- 3.1 Assign all edges between the top (highest residual degree) vertex of S and the top vertices of T .

- 3.2 Assign all edges between the top vertex of T and the top vertices of S .

Correctness of Algorithm

The residual degree of the set S (T respectively) at the beginning of round i is denoted by $R_i(S)$ ($R_i(T)$ respectively). The number of vertices with positive residual degree (*non full* vertices) in S (T) is denoted by $N_i(S)$ ($N_i(T)$).

The residual degrees of set S at round i in decreasing order is s_1^i, \dots, s_h^i and those of set T is t_1^i, \dots, t_k^i .

The proof is by induction on the round i . The following invariant holds:

1. $R_i(S) = R_i(T)$ and
2. $N_i(T) \geq d(s_1^i)$ and $N_i(S) \geq d(t_1^i)$.

Choosing a suitable β -graph

Lemma 6. *Let $G = (V, E)$ be a simple graph with n_1 vertices and $\beta \geq 1$. Let $\alpha_0 = \max\{4\beta, \beta \ln n_1 + \ln(n_1 + 1)\}$. Then, for all $\alpha \geq \alpha_0$ the sequence $D = EXP(Y^{(\beta, \alpha)} - Y^G)$ is bipartite-eligible.*

Proof

Let n_2 be the number of elements in D and $\alpha \geq \alpha_0$. We have

$$\begin{aligned} n_2 &\geq \sum_{i=1}^{\lfloor e^{\alpha/\beta} \rfloor} \left\lfloor \frac{e^\alpha}{i^\beta} \right\rfloor - n_1 > e^\alpha \sum_{i=1}^{\lfloor e^{\alpha/\beta} \rfloor} \frac{1}{i^\beta} - \lfloor e^{\alpha/\beta} \rfloor - n_1 \\ &\geq e^\alpha \int_{i=1}^{\lfloor e^{\alpha/\beta} \rfloor + 1} \frac{1}{i^\beta} - \lfloor e^{\alpha/\beta} \rfloor - n_1. \end{aligned}$$

If $\beta = 1$, then we have

$$n_2 \geq \alpha e^\alpha - e^\alpha - n_1 \geq 4e^\alpha - 2e^\alpha + 1 \geq 2m + 1$$

and if $\beta > 1$ we have

$$n_2 \geq \frac{e^\alpha}{\beta - 1} - e^{\alpha/\beta} - n_1 \geq 4e^{\alpha/\beta} - 2e^{\alpha/\beta} + 1 \geq 2m + 1,$$

$$y_{n_1}^{(\beta, \alpha)} \geq \left\lfloor \frac{e^\alpha}{n_1^\beta} \right\rfloor > \frac{e^\alpha}{n_1^\beta} - 1 \geq \frac{n_1^{\beta+1} + n_1^\beta}{n_1^\beta} - 1 = n_1,$$

A Greedy Algorithm for VERTEX-COVER

1. Choose the top degree vertices that covers at least $(1 - \delta(n))$ fraction of the total number of edges for an appropriately chosen function $\delta(n) = o(1)$.
2. Include vertices that covers the remaining uncovered edges.

Average-Case Analysis

A random power law graph model: A random graph model that generates graphs whose **expected** degree-sequence follows a given power law degree sequence. [Lu and Chung, 2002]

Given a power-law degree sequence d_1, \dots, d_n , an edge occurs between vertices i and j with probability

$$\frac{d_i d_j}{\sum_{k=1}^n d_k}.$$

A generalization of the classical Erdos-Renyi model.

Theorem 2. *For a random power law graph with $\beta \leq 2$, there exists a polynomial time algorithm that with probability at least $1 - o(1)$ outputs a $1 + o(1)$ approximation to the VERTEX COVER problem.*

Conclusion

- Real-world power law graphs are not “worst-case” instances of power law graphs, but rather typical instances, and may be well-modeled by **power law random graph models**.
- A general technique for establishing NP-hardness of a large class of problems in power-law graphs.
- Embedding Lemma can have other applications, e.g., in showing hardness of **approximation** in power-law graphs
- On the positive side, investigate approximation algorithms for sparse power law graphs.