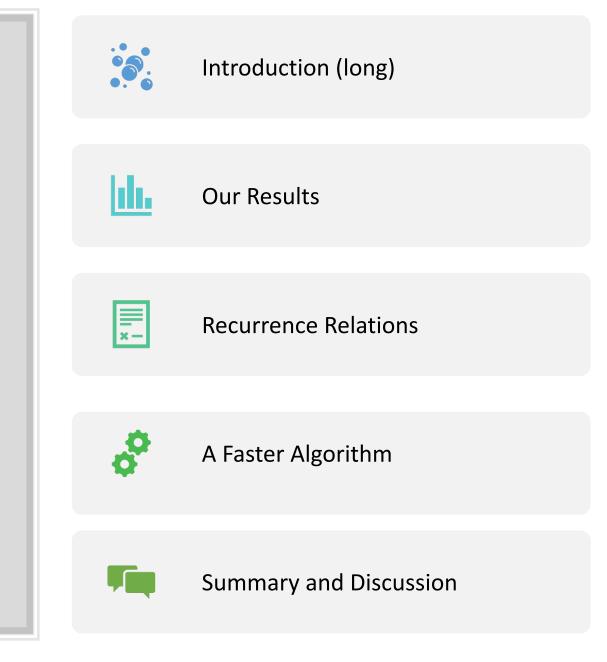
Analysis of Two-variable Recurrence Relations with Application to Parameterized Approximations

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Technion Theory Seminar, Nov 2020

Talk Outline



Introduction

Vertex Cover (VC)

- An undirected graph G = (V, E) $S \subseteq V$ is a cover if: $\forall (u, v) \in E \rightarrow S \cap \{u, v\} \neq \emptyset$
- Decision problem: k -Vertex Cover Decide if G has a cover S with $|S| \le k$
- Optimization:
 find a cover S of G such that |S| is minimal
- NP-hard

Approximation

- Find a non-optimal solution in polynomial time.
- A simple polynomial time 2-approximation
 - If G has a cover of size k
 - Finds a vertex cover S with $|S| \le 2k$
- No (2ε) -approximation under UGC

Parameterization

- $\ensuremath{\bullet}$ Associate an instance with a parameter k
- Language $L \subseteq \Sigma^*$
- Paramterization $k = \kappa(I)$ for all $I \in \Sigma^*$
- Parameterized Algorithm:
 - Decide if $I \in L$
 - In time $O(f(k) \cdot poly(n)) = O^*(f(k))$
 - |I| = n, $k = \kappa(I)$
- Special class of non-polynomial algorithms

Parameterized Vertex Cover

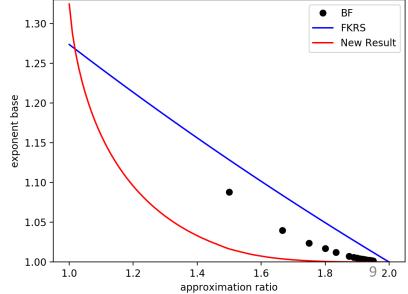
- k -Vertex Cover:
 - Input G and k
 - Decide if G has a cover S with $|S| \le k$
- Standard parameterization $\kappa(G, k) = k$
- A paramaterized $O^*(1.273^k)$ algorithm
- No 2^{o(k)} algorithm under ETH

Parameterized Approximation

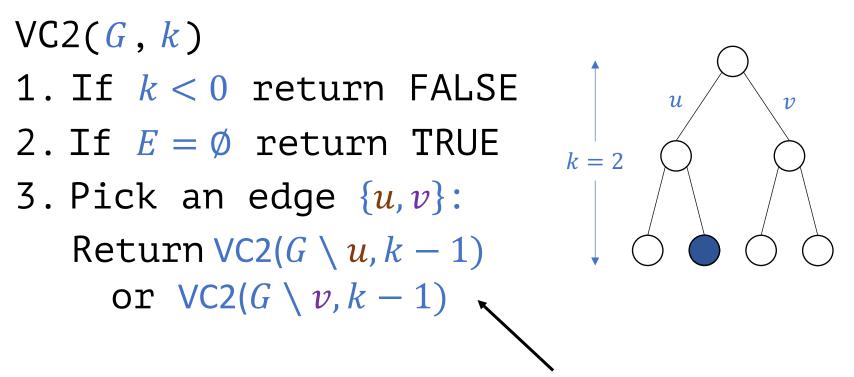
- *α* ∈ [1,2].
- An algorithm is a parameterized *α*-approximation for Vertex Cover:
 - If G has a vertex cover of size k
 → Returns a vertex cover of size αk
 - Running time $O(f(k) \cdot poly(n))$
- Goal: Tradeoff Increase $\alpha \rightarrow$ Reduce the running time

Previous Results

- A parameterized α-approximation with running time O^{*}((1.2378^{2-α})^k) [Fellows, K, Rosamond, Shachnai] [Bourgeois, Escoffier, Paschos]
- A parameterized 1.5-approximation in $O^*(1.0883^k)$ [Brankovic, Fernau]



A simple Algorithm for VC

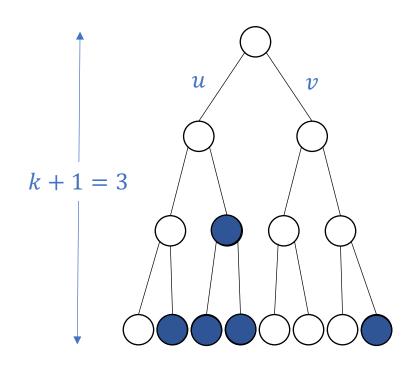


• An $O^*(2^k)$ algorithm

Branching

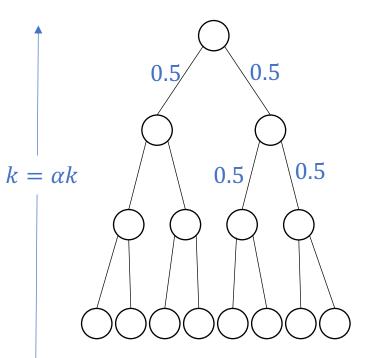
Approximate Solutions

- Pick an edge {*u*, *v*}
- Branch over:
 - •u is in the cover
 - v is in the cover
- Repeat k + 1 times
- Doubled the number of leaves
- k + 2 "good" leaves



Randomization

- Pick an edge {*u*, *v*}
- With Prob. 0.5 Add u to the cover
 - Else
 - Add v to the cover
- Repeat $\alpha \cdot k$ times
- What is the probability a cover is found?



Analysis

• Let X_n be a random indicator:

Pick an edge $\{u, v\}$ With Prob. 0.5 Add u to the cover Else Add v to the cover Repeat $\alpha \cdot k$ times

- X_n = 1 the n-th selected vertex reduced the minimal cover size by 1 (or, a cover was already found)
- Otherwise $X_n = 0$
- $\Pr(X_n = 1 \mid X_1, \dots, X_{n-1}) \ge \frac{1}{2}$ $\Pr(\text{Found a cover}) \ge \Pr(X_1 + \dots + X_{\alpha k} \ge k)$

$$\geq \Pr\left(\text{Binomial}\left(\alpha k, \frac{1}{2}\right) \geq k\right)$$

Analysis- cont'd

Pr(Found a cover) $\geq \Pr\left(\text{Binomial } \left(\alpha k, \frac{1}{2}\right) \geq k\right)$

$$\geq \frac{1}{(\alpha k + 1)^2} \exp\left(\alpha D\left(\frac{1}{\alpha} \parallel \frac{1}{2}\right)\right)$$
$$= \frac{1}{(\alpha k + 1)^2} (c_{\alpha})^{-k}$$

D is the Kullback-Leibler Divergence $D(a \parallel b) = a \log\left(\frac{a}{b}\right) + (1 - b)$ Standard tail bound. Commonly dervied using the Method of Types.

Parameterized Approximation

Run the algorithm:

```
Pick an edge \{u, v\}
With Prob. 0.5
Add u to the cover
Else
Add v to the cover
Repeat \alpha \cdot k times
```

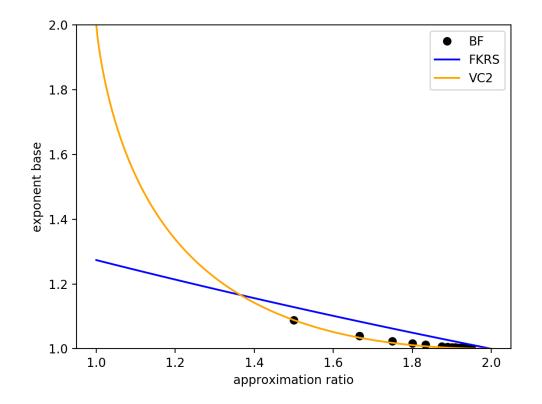
 $(\alpha k + 1)^2 (c_\alpha)^k$ times

The graph has a cover of size k

• With constant probability the algorithm finds a cover of size αk

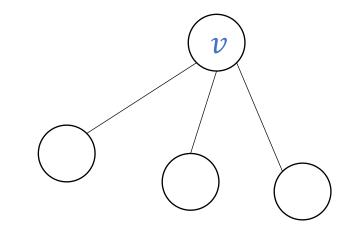
Running Times

- For $\alpha = 1.5$ the running time is $O^*(1.0887^k)$
- Nearly matches the current best result



Faster Algorithm

• An $O^*(1.46^k)$ algorithm for VC:



- If there is $v \in V$, $deg(v) \ge 3$ branch over:
 - v is in the cover
 - N(v) is in the cover

Otherwise, $\max \deg(v) \le 2$, find a minimum vertex cover

Running time: p(k) = p(k-1) + p(k-3)

Faster Randomized Algorithm Fix $\gamma \in (0,1)$.

```
VC3_{\nu}(G)
If there is v \in V, deg(v) \ge 3:
   With probability \gamma
      Add v to the cover
   With probability 1-\gamma
      Select S \subseteq N(v), |S|=3
      Add S to the cover
Otherwise, \max \deg(v) \leq 2, find a minimum
vertex cover
```

Success Probability

If there is $v \in V$, $deg(v) \ge 3$: With probability γ : Add v to the cover. With probability $1 - \gamma$: Select $S \subseteq N(v)$, |S|=3Add S to the cover.

What is the probabiliy the algorithm returns a cover of size αk ?

- P(b, k)- the minimal probabiliy:
 - The algorithm returns a cover of size *b* budget
 - Given a graph with cover of size *k* parameter

$$p(b,k) = \min \begin{cases} \frac{\gamma \cdot p(b-1,k-1) + (1-\gamma) \cdot p(b-3,k)}{\gamma \cdot p(b-1,k) + (1-\gamma) \cdot p(b-3,k-3)} \\ p(b,k) = 0 \text{ for } b < 0 \text{ and } p(b,0) = 1 \text{ for } b \ge 0. \end{cases}$$

Then $P(b,k) \ge p(b,k)$

Success Probability- cont'd $p(b,k) = \min \begin{cases} \gamma \cdot p(b-1,k-1) + (1-\gamma) \cdot p(b-3,k) \\ \gamma \cdot p(b-1,k) + (1-\gamma) \cdot p(b-3,k-3) \end{cases}$ $p(b,k) = 0 \text{ for } b < 0 \text{ and } p(b,0) = 1 \text{ for } b \ge 0.$

- *p*(*b*, *k*) can be computed using dynamic programing
 → Run the algorithm ¹/_{p(αk,k)} times
 → α-approximation
- We want to find c such that $p(\alpha k, k) \approx c^{-k}$

$$\lim_{k\to\infty}\frac{1}{k}\log(p(\alpha k,k)) = ?$$

Our Results

Our results- Highlights

- A solution for a wide familiy of recurrence relations, generalizing: $p(b,k) = \min \begin{cases} \gamma \cdot p(b-1,k-1) + (1-\gamma) \cdot p(b-3,k) \\ \gamma \cdot p(b-1,k) + (1-\gamma) \cdot p(b-3,k-3) \end{cases}$ $p(b,k) = 0 \text{ for } b < 0 \text{ and } p(b,0) = 1 \text{ for } b \ge 0.$
- Parameterized approximation algorithms for:
 - Vertex Cover
 - 3-Hitting Set

Significant improvement of the running times.

Generalizing The Reccurence

 $p(b,k) = \min \begin{cases} \gamma \cdot p(b-1,k-1) + (1-\gamma) \cdot p(b-3,k) \\ \gamma \cdot p(b-1,k) + (1-\gamma) \cdot p(b-3,k-3) \end{cases}$

$$= \min \begin{cases} \bar{\gamma}_{1}^{1} \cdot p(b - \bar{b}_{1}^{1}, k - \bar{k}_{1}^{1}) + \bar{\gamma}_{2}^{1} \cdot p(b - \bar{b}_{2}^{1}, k - \bar{k}_{2}^{1}) \\ \bar{\gamma}_{1}^{2} \cdot p(b - \bar{b}_{1}^{2}, k - \bar{k}_{1}^{2}) + \bar{\gamma}_{2}^{2} \cdot p(b - \bar{b}_{2}^{2}, k - \bar{k}_{2}^{2}) \end{cases}$$

 $= \min\left\{\sum_{i=1}^{r_{1}} \bar{\gamma}_{i}^{1} \cdot p(b - \bar{b}_{i}^{1}, k - \bar{k}_{i}^{1}), \sum_{i=1}^{r_{2}} \bar{\gamma}_{i}^{2} \cdot p(b - \bar{b}_{i}^{2}, k - \bar{k}_{i}^{2})\right\}$ $= \min_{1 \le j \le N} \sum_{i=1}^{r_{j}} \bar{\gamma}_{i}^{j} \cdot p(b - \bar{b}_{i}^{j}, k - \bar{k}_{i}^{j})$ $\bar{\gamma}^{1} = \bar{\gamma}^{2} = (\gamma, 1 - \gamma), \bar{b}^{1} = \bar{b}^{2} = (1, 3), \bar{k}^{1} = (1, 0), \bar{k}^{2} = (0, 3)$ $N = 2, \qquad r_{1} = r_{2} = 2$

Recurrence Relations

Consider a function $p: \mathbb{Z} \times \mathbb{N} \rightarrow [0,1]$ satisfying*:

$$p(b,k) = \min_{1 \le j \le N} \sum_{i=1}^{r_j} \bar{\gamma}_i^j \cdot p(b - \bar{b}_i^j, k - \bar{k}_i^j)$$

And $p(b,k) = 0$ for $b < 0$, $p(b,0) = 1$ for $b \ge 0$

Where $N \in \mathbb{N}$, for every $1 \leq j \leq N$: • $r_j \in \mathbb{N}$ • $\bar{\gamma}^j \in \mathbb{R}_{>0}^{r_j}$ and $\sum_{i=1}^{r_j} \bar{\gamma}_i^j = 1$ • $\bar{k}^j \in \mathbb{N}^{r_j}, \bar{b}^j \in \mathbb{N}_+^{r_j}$ Defines the "term" $\sum_{i=1}^{r_j} \bar{\gamma}_i^j \cdot p(b - \bar{b}_i^j, k - \bar{k}_i^j)$

Main Theorem

For $\alpha > 0$, the α -Branching number of $(\overline{\gamma}^{j}, \overline{b}^{j}, \overline{k}^{j})$ is

$$M_j = f(\alpha, \bar{\gamma}^j, \bar{b}^j, \bar{k}^j)$$

f is defined by quasiconvex minimzation problem.
 → Can be computed.

Then,

$$\lim_{k\to\infty}\frac{1}{k}\log(p(\alpha k,k)) = -\max_{1\leq j\leq N}M_j$$

$$p(\alpha k, k) \approx \left(\exp\left(\max_{1 \le j \le N} M_j\right)\right)^{-k}$$

Main Theorem - Example

$$p(b,k) = \min \begin{cases} \gamma \cdot p(b-1,k-1) + (1-\gamma) \cdot p(b-3,k) \\ \gamma \cdot p(b-1) \end{cases}$$
Can be enhanced using quasiconvex optimization
Select $\gamma = 0.7463$. By the theorem,
A 1.5-approximation in time $\frac{1}{p(1.5k,k)} \approx 1.04364^k$

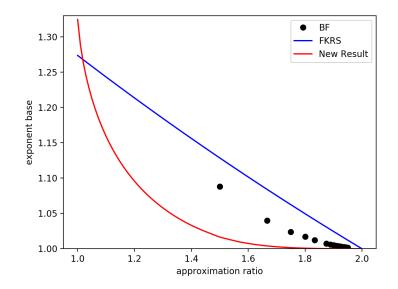
Thus, $p(1.5k, k) \approx \exp(-0.04271)^k \approx 1.04364^{-k}$

$$\bar{\gamma}^1 = \bar{\gamma}^2 = (\gamma, 1 - \gamma), \bar{b}^1 = \bar{b}^2 = (1,3), \bar{k}^1 = (1,0), \bar{k}^2 = (0,3)$$

Vertex Cover

Faster parameterized approximation algorithms for vertex cover.

• $\alpha > 1.4$ - a variant of $VC3_{\gamma}$.



- → 1.5-approximation in $O^*(1.017^k)$.
- $\alpha < 1.4$ a variant of a textbook algorithm.
 - → 1.1-approximation in $O^*(1.16^k)$.

Additional resuls for 3-Hitting Set

Related Work

Related Work – Recurrences

• Single variable recurrence are studied in introductory courses.

$$p(k) = p(k-1) + p(k-4)$$

- Eppstein, 2006. A different class of multivariate relations.
 - Commonly used in Measure and Conquer.

Recurrences and the Method of Types

Back to Our Algorithm Fix $\gamma \in (0,1)$.

```
VC3_{\nu}(G)
If there is v \in V, \deg(v) \ge 3:
   With probability \gamma
      Add v to the cover
   With probability 1 - \gamma:
      Select S \subseteq N(v), |S|=3
      Add S to the cover
Otherwise, \max \deg(v) \leq 2, find a minimum
vertex cover
```

Assumption: v is always in a minimal vertex cover

Random Process

- P(b, k)- the probability of finding:
 - A cover of size *b* or less.
 - Assuming the graph has a cover of size k or less.
- Define $(Y_n)_{n=1}^{\infty}$ by:
 - $Y_n = 1$ if v was selected at the n-th step.
 - Otherwise $Y_n = 2$.
- $\Pr(Y_n = 1) = \gamma$, $\Pr(Y_n = 2) = 1 \gamma$

Random Process – Cont'd

- $\bar{k} = (1,0)$.
- $\bar{b} = (1,3)$

$$P(1.5k,k) = \Pr\left(\exists n: \sum_{\ell=1}^{n} \overline{b}_{Y_{\ell}} \le 1.5k \text{ and } \sum_{\ell=1}^{n} \overline{k}_{Y_{\ell}} \ge k\right)$$

The Method of Types

- $(a_1, \dots, a_n) \in \{1, 2\}^n$.
- The Type of $(a_1, ..., a_n)$ is $T(a_1, ..., a_n) = (T_1, T_2)$,

$$T_i = \frac{|\{\ell \mid a_\ell = i\}|}{n}$$
 - the frequency of *i*.

• For example,

$$\mathbf{T}(1,1,1,2,2,1) = \left(\frac{4}{6},\frac{2}{6}\right)$$

Sanov's Theorem [Sanov, 1961] $\boldsymbol{Q} \subseteq \{ \bar{\boldsymbol{q}} \in \mathbb{R}^2_{\geq 0} \mid \sum_{i=1}^2 \bar{\boldsymbol{q}}_i = 1 \}. \ \bar{\boldsymbol{\gamma}} = (\boldsymbol{\gamma}, 1 - \boldsymbol{\gamma})$ $\Pr(\mathbf{T}(Y_1, \dots, Y_n) \in \mathbf{Q}) \approx \exp(-nc), \ \mathbf{c} = \min_{\bar{q} \in \mathbf{Q}} D(\bar{q} \parallel \bar{\gamma})$ $\lim_{n \to \infty} \frac{1}{n} \log \Pr(\mathbf{T}(Y_1, \dots, Y_n) \in \mathbf{Q}) = -\mathbf{c}$ $n \rightarrow \infty n$

D is the Kullback-Leibler Divergence: $D(\bar{q} \parallel \bar{\gamma}) = \sum_{i=1}^{2} \bar{q}_{i} \log \frac{\bar{q}_{i}}{\bar{\gamma}_{i}}$

The point in Qnearest to $\overline{\gamma}$

$\bar{k} = (1,0), \qquad \bar{b} = (1,3).$

Using Types $P(1.5k,k) = \Pr\left(\exists n: \sum_{\ell=1}^{n} \overline{b}_{Y_{\ell}} \le 1.5k \text{ and } \sum_{\ell=1}^{n} \overline{k}_{Y_{\ell}} \ge k\right)$

$$= \Pr\left(\exists n: T = \mathbf{T}(Y_1, \dots, Y_n), \frac{n \cdot T_1 \cdot \overline{b}_1 + n \cdot T_2 \cdot \overline{b}_2 \leq 1.5k \text{ and}}{n \cdot T_1 \cdot \overline{k}_1 + n \cdot T_2 \cdot \overline{k}_2 \geq k}\right)$$

$$= \Pr\left(\exists n: T = \mathbf{T}(Y_1, \dots, Y_n), \sum_{i=1}^2 T_i \cdot \overline{b}_i \le 1.5 \frac{k}{n} \text{ and } \sum_{i=1}^2 T_i \cdot \overline{k}_i \ge \frac{k}{n}\right)$$

$$= \Pr\left(\exists n: \mathbf{T}(Y_1, \dots, Y_n) \in Q_{\frac{k}{n}}\right) \text{ where } Q_{\beta} = \begin{cases} \overline{q} \in \mathbb{R}^2_{\geq 0} & \sum_{i=1}^2 \overline{q}_i = 1\\ \sum_{i=1}^2 \overline{q}_i \cdot \overline{b}_i \leq 1.5\beta\\ \sum_{i=1}^2 \overline{q}_i \cdot \overline{k}_i \geq \beta \end{cases}$$

Using Types (cont'd)

n =

$$P(1.5k,k) = \Pr\left(\exists n: \mathbf{T}(Y_1, \dots, Y_n) \in Q_{\frac{k}{n}}\right)$$
$$\geq \Pr\left(\mathbf{T}\left(Y_1, \dots, Y_{\frac{k}{\beta^*}}\right) \in Q_{\beta^*}\right)$$
$$\approx \exp\left(-\frac{k}{\beta^*}D(\bar{q}^* \parallel \bar{\gamma})\right)$$

matching upper bound.

$$\beta^*, \bar{q}^* = \operatorname{argmin}_{\beta, \bar{q} \in Q_{\beta}} \left(\frac{1}{\beta} D(\bar{q} \parallel \bar{\gamma}) \right)$$

The General Solution

- A more complicated random process
 - \rightarrow Involves all the terms
 - → Cannot apply Sanov's theorem directly
- Properties of types carry over to the new process
- A variant of Sanov's theorem

A Faster Algorithm

Recap

- We showed a 1.5-approximation in $O^*(1.0437^k)$
 - Prob. γ : select v to the cover
 - Prob. 1γ : select 3 neighbors of ν to the cover

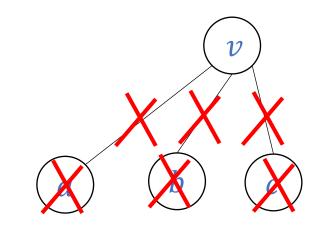
 $p(b,k) = \min \begin{cases} \gamma \cdot p(b-1,k-1) + (1-\gamma) \cdot p(b-3,k) \\ \gamma \cdot p(b-1,k) + (1-\gamma) \cdot p(b-3,k-3) \end{cases}$

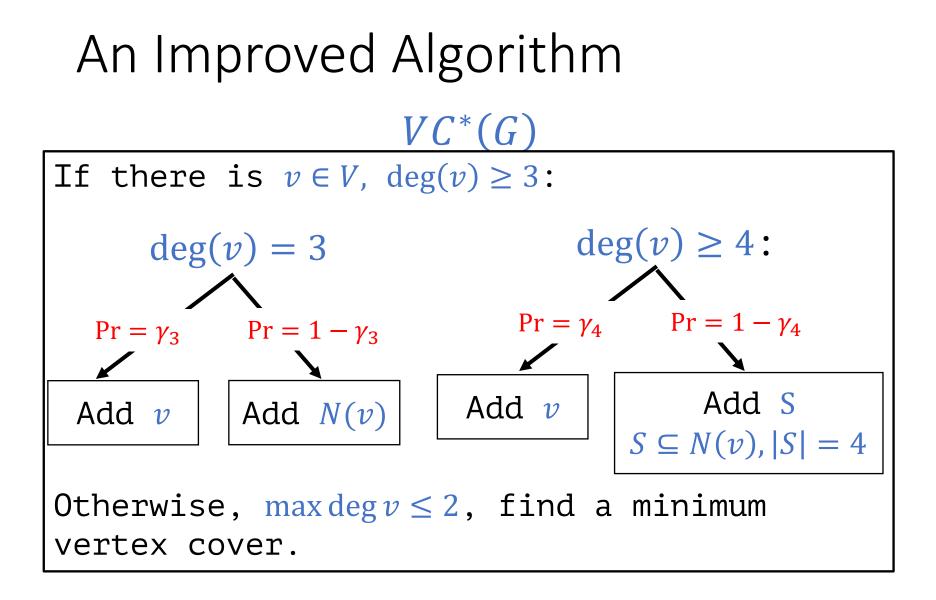
p(b,k) = 0 for b < 0 and p(b,0) = 1 for $b \ge 0$

12

Refined Analysis

- Focus on:
 - v is in a minimal cover S
 - $N(v) = \{a, b, c\}$
- Two scenarios:
 - v is selected to the cover $\rightarrow k$ decreases by 1 $S \setminus \{v\}$ is a cover of $G \setminus \{v\}$
 - N(v) is selected to the cover → k decreases by 1
 S \ {v} is a cover of G \ N(v)





Analysis

probability of finding a cover of size b or less

$$p(b,k) = \min \begin{cases} \frac{(\gamma_2 \cdot p(b-1,k-1) + (1-\gamma_2) \cdot p(b-3,k-1))}{\gamma_3 \cdot p(b-1,k) + (1-\gamma_3) \cdot p(b-3,k-3)} \\ \frac{\gamma_4 \cdot p(b-1,k-1) + (1-\gamma_4) \cdot p(b-4,k)}{\gamma_4 \cdot p(b-1,k) + \gamma_4 \cdot p(b-4,k-4)} \end{cases}$$

p(b,k) = 0 for b < 0 and p(b,0) = 1 for $b \ge 0$

 v_i is is not an appripriate degree $g(w) \neq 3$.

Analysis -cont'd

Using the main theorem we can optimize γ_3, γ_4 and get $p(1.5k, k) \approx 1.035^{-k}$

By repeating the algorithm $p(1.5k, k)^{-1}$ times we get:

- 1.5 -approximation with constant probability
- The running time is $O^*(1.035^k)$
- → Can be generalized

➔ "Incorrect" branching is important

Summary and Discussion

Summary

- Analysis of two variable recurrence relations
 - ➔ Method of Types
 - → Simple formula
 - ➔ In-depth understanding
- Faster parameterized approximation algorithm
 - ➔ Randomized Branching
 - ➔ Significant improvement of running time
 - → Simple algorithms

Thank You