### An FPTAS for #Knapsack and Related Counting Problems

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Polynomial-Time Approximation Schemes for Knapsack and Related Counting Problems using Branching Programs	<ul><li>PDF</li><li>PostScript</li><li>Other formats</li></ul>	A Deterministic Polynomial-time Approximation Scheme for Counting Knapsack Solutions	<ul><li>PDF</li><li>PostScript</li><li>Other formats</li></ul>
Parikshit Gopalan, Adam Klivans, Raghu Meka (Submitted on 18 Aug 2010) We give a deterministic, polynomial-time algorithm for approximately counting the number of {0,1}-solutions to any instance of the knapsack problem. On an instance of length n with total weight W and accuracy parameter eps, our algorithm produces a (1 + eps)-multiplicative approximation in time poly(n,log W,1/eps). We also give algorithms with identical guarantees for general integer knapsack, the multidimensional knapsack problem (with a constant number of constraints) and for contingency tables (with a constant number of rows). Previously, only randomized approximation schemes were known for these problems due to work by Morris and Sinclair and work by Oper. Our algorithms work by constructing small-width, read-once branching programs for approximating the underlying solution space under a carefully chosen distribution. As a byproduct of this approach, we obtain new query algorithms for learning functions of k halfspaces with respect to the uniform distribution on (0,1) <sup>a</sup> n. The running time of our algorithm is polynomial in the accuracy parameter eps. Previously even for the case	Current browse context: cs.DS < prev   next > new   recent   1008 Change to browse by: cs cs.CC cs.LG References & Citations • NASA ADS DBLP - CS Bibliography listing   bibtex Parikshit Gopalan	Daniel Stefankovic, Santosh Vempala, Eric Vigoda   (Submitted on 10 Aug 2010)   Given n elements with nonnegative integer weights w1, wn and an integer capacity C, we consider the counting version of the classic knapsack problem. find the number of distinct subsets whose weights add up to at most the given capacity. We give a deterministic algorithm that estimates the number of solutions to within relative error 1+-eps in time polynomial in n and 1/eps (till you/ponmial approximation scheme). More precisely, our algorithm takes time O(n*3 (1/eps) log (n/eps)). Our algorithm is based on dynamic programming. Previously, randomized polynomial time approximation schemes were known first by Morris and Sinclair via Markov chain Monte Carlo techniques, and subsequently by Dyer via dynamic programming and rejection sampling.   Comments: 11 pages   Subjects: Data Structures and Algorithms (cs.DS)   ACM classes: F. 2.: G. 2. 1	Current browse context: cs.DS < prev   next > new   recent   1008 Change to browse by: cs References & Citations • NASA ADS DBLP - CS Bibliography listing   bibtex Daniel Stefankovic Santosh Vempala Eric Vioda
of k=2, only algorithms with an exponential dependence on eps were known. Subjects: Data Structures and Algorithms (cs.DS); Computational Complexity (cs.CC); Learning (cs.LG) Cite as: arXiv:1008.3187v1 [cs.DS] Submission history From: Raghu Meka [view email] [v1] Wed, 18 Aug 2010 23:45:28 GMT (33kb,S) Which authors of this paper are endorsers?	Adam Klivans Raghu Meka Bookmark (watis this?) ■ ③ 米 № ఔ ㎡ 📲 😂 🛱	Cite as: arXiv:1008.1687v1 [cs.DS]   Submission history   From: Daniel Štefankovič (view email)   [v1] Tue, 10 Aug 2010 10:54:31 GMT (9kb)   Which authors of this paper are endorsers?   Link back to: arXiv, form interface, contact.	Bookmark (what is this?) ■ ۞ X III III III III IIII IIII ♀ X III III IIIIIIIIII

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### What can be counted? (in polynomial-time)

### exactly? very little...

number of spanning trees (using determinant), Kirchoff'1847.

perfect matchings in planar graphs (using Pfaffians), Kasteleyn'1960.

(rest: usually **#P-hard**)

### What can be counted? (in polynomial-time)

### approximately? a little more...

perfect matchings in bipartite graphs (permanent of non-negative matrices), Jerrum, Sinclair, Vigoda'2001.

Ferromagnetic Ising model, Jerrum, Sinclair'1989.

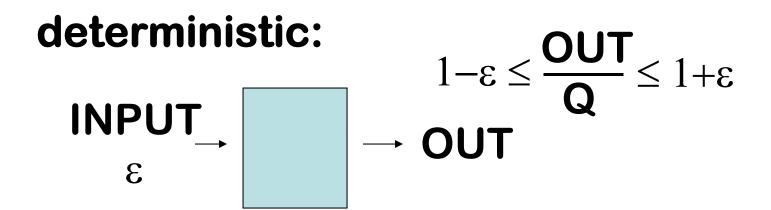
Independent sets ( $\Delta \leq$  5), Weitz'2004.

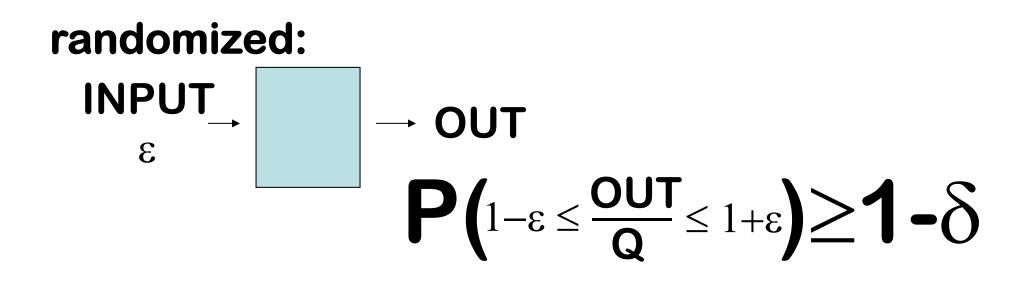
k-colorings (k  $\geq$  (11/6) $\triangle$ ), Vigoda'1999.

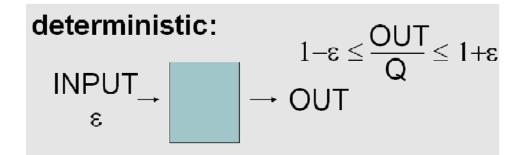
. . . .

(approximate counting ≈ random sampling, Jerrum, Valiant, Vazirani'1986)

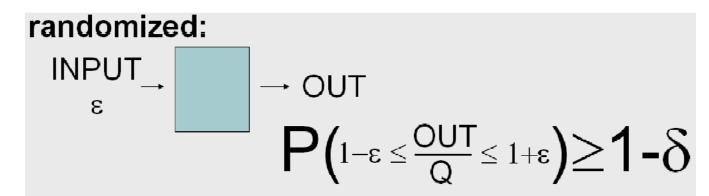
### Approximate counting (in polynomial-time)







not too many examples: independent sets in degree  $\leq$ 5 graphs (Weitz'2004), matchings in bounded degree graphs (Bayati, Gamarnik, Katz, Nair, Tetali'2007), satisfying assignments of DNF formulas with terms of size  $\leq$  C (Ajtai, Wigderson'1985)



more examples; Monte Carlo, usually using a Markov chain (dependence  $1/\epsilon^2$ )

$$\begin{array}{c} \text{deterministic:} \\ 1-\epsilon \leq \frac{\mathsf{OUT}}{\mathsf{Q}} \leq 1+\epsilon \\ \epsilon \end{array} \xrightarrow{\mathsf{OUT}} \to \mathsf{OUT} \end{array} \xrightarrow{\mathsf{OUT}} \mathsf{OUT} \xrightarrow{\mathsf{I}} \mathsf{OUT} \xrightarrow{\mathsf{OUT}} \mathsf{OUT} \\ \to \mathsf{OUT} \end{array} \xrightarrow{\mathsf{OUT}} \mathsf{OUT} \xrightarrow{\mathsf{OUT}} \operatorname{OUT} \xrightarrow{\mathsf{OUT}} \mathsf{OUT} \xrightarrow{\mathsf{OUT}} \xrightarrow{\mathsf{OUT}} \operatorname{OUT} \xrightarrow{\mathsf{OUT}} \xrightarrow{\mathsf{OUT}} \operatorname{OUT} \operatorname{OUT} \xrightarrow{\mathsf{OUT}} \operatorname{OUT} \xrightarrow{\mathsf{OUT}} \operatorname{OUT} \xrightarrow{\mathsf{OUT}} \operatorname{OUT} \xrightarrow{\mathsf{O$$

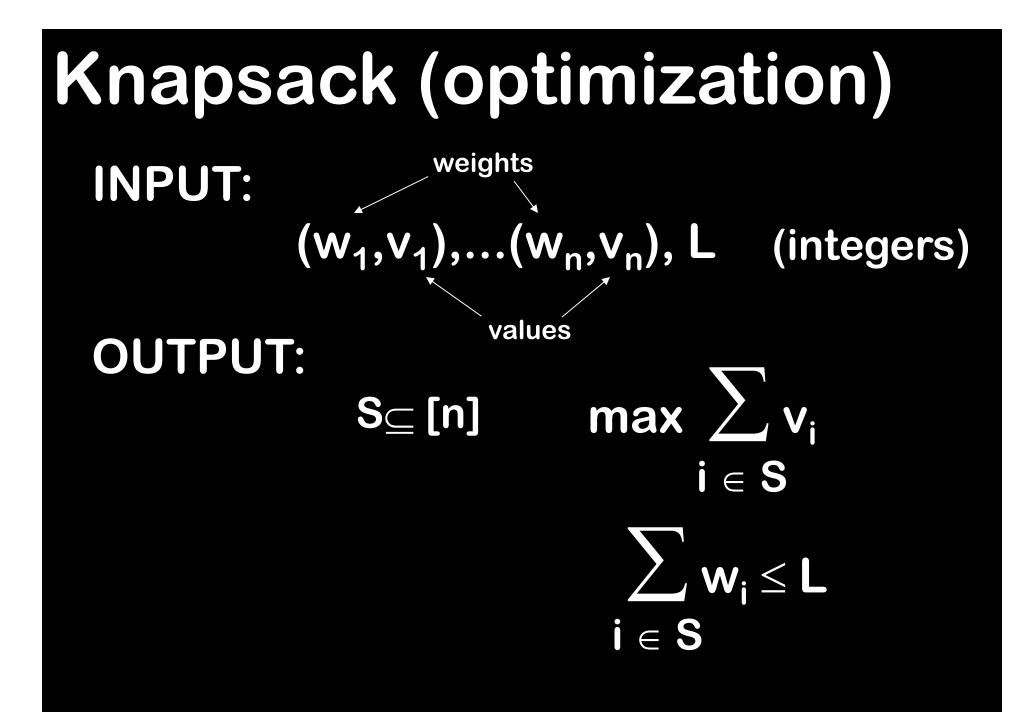
1) is randomness necessary ?

### Is P = BPP ?

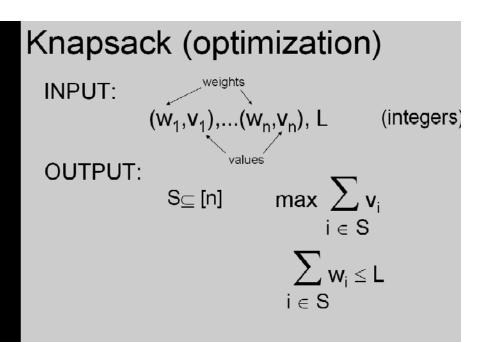
**Primes**  $\in$  **P** (Agarwal, Kayal, Saxena 2001)

2) dependence on  $\epsilon$  ?

Monte Carlo  $\rightarrow \Theta(1/\epsilon^2)$ 

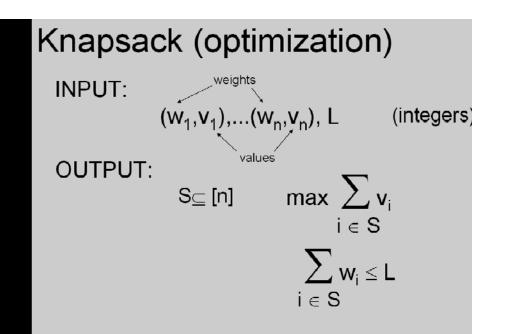


### Dynamic program #1 (L is small) T[i,M] (optimal solution with items 1,...,i and limit M)



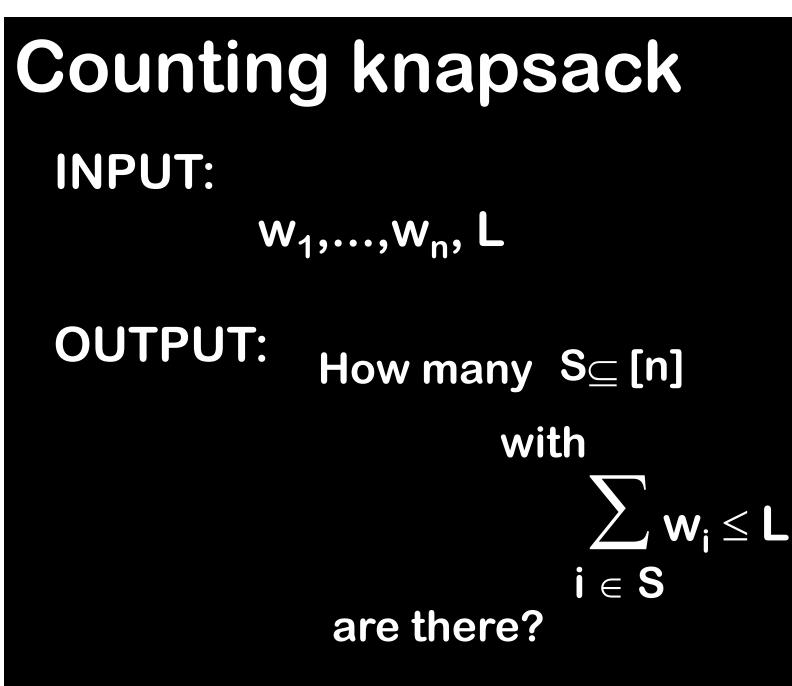
 $T[i,M] = \max \{ T[i-1,M] \\ T[i-1,M-w_i] + v_i \}$ 

### Dynamic program #2 (v<sub>i</sub>'s are small) T[i,V](smallest weight of a subset of 1,...,i, with value $\geq V$ )



## $T[i,V] = \min \left\{ \begin{array}{l} T[i-1,V] \\ T[i-1,V-v_i] + w_i \end{array} \right.$







**Counting knapsack** Dyer, Frieze, Kannan, Kapoor, Perkovic, Vazirani'1993 exp(O<sup>\*</sup>(n<sup>1/2</sup>)) / ε<sup>2</sup> randomized approximation algorithm

Morris, Sinclair'1999 O( n<sup>c</sup> / ε<sup>2</sup>) randomized approximation algorithm (MCMC, canonical paths)

Dyer'2003O(n<sup>2.5</sup> + n²/ε²)randomized approximation algorithm<br/>(dynamic programming)

**OURS: Ο**<sup>\*</sup>(**n**<sup>3</sup>/ε)

Dyer'2003:

T[i,M]

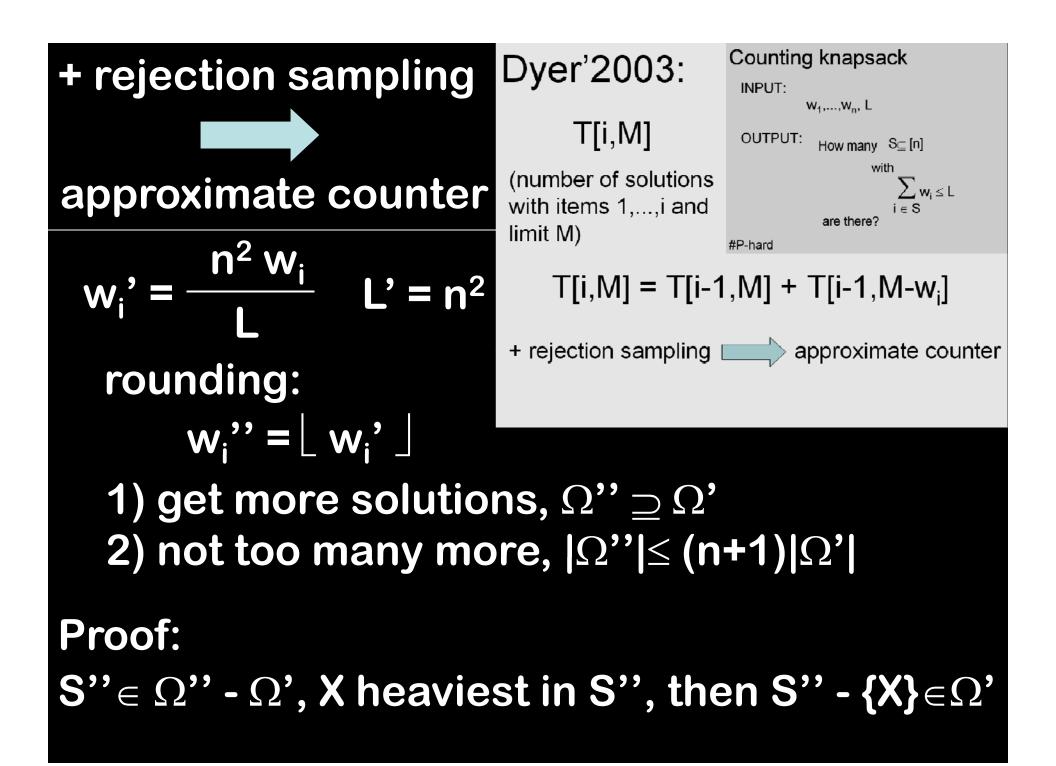
(number of solutions with items 1,...,i and limit M)

 $\label{eq:constraint} \begin{array}{l} \text{Counting knapsack} \\ \text{INPUT:} & & \\ & w_1, \dots, w_n, \ L \\ \text{OUTPUT:} & & \text{How many} \quad S \subseteq [n] \\ & & \text{with} \\ & & \sum_{i \ \in \ S} w_i \leq L \\ & \text{are there?} \end{array}$ 

### $T[i,M] = T[i-1,M] + T[i-1,M-w_i]$

#P-hard

+ rejection sampling ) approximate counter



### Our dynamic program

deterministic approximation algorithm

 $\tau(\mathbf{i},\mathbf{A}) = \begin{array}{l} \text{smallest M such that knapsack} \\ \text{with } w_1, \dots, w_i, \mathbf{M} \text{ has } \geq \mathbf{A} \text{ solutions} \end{array}$ 

 $\tau(\mathbf{i},\mathbf{A}) = \min_{\alpha \in [0,1]} \max \{ \begin{aligned} \tau(\mathbf{i}-\mathbf{1},\alpha \mathbf{A}) \\ \tau(\mathbf{i}-\mathbf{1},(\mathbf{1}-\alpha) \mathbf{A}) + w_{\mathbf{i}} \end{aligned}$ 

## $Q = 1 + \varepsilon/(n+1)$ $s = \lceil n \log_Q 2 \rceil$

### T[0..n,0..s]

#### Our dynamic program

deterministic approximation algorithm

 $\tau(i,A) = \begin{array}{ll} \text{smallest } M \text{ such that knapsack} \\ \text{with } w_1, \dots, w_i, M \text{ has } \geq A \text{ solutions} \end{array}$ 

$$\tau(i,A) = \min_{\alpha \in [0,1]} \max \left\{ \begin{array}{l} \tau(i-1,\alpha A) \\ \tau(i-1,(1-\alpha) A) + w_i \end{array} \right.$$

## $T(i,j)=\min \max \{ T(i-1,\lfloor j+\ln_Q \alpha \rfloor) \\ \alpha \in [0,1] \\ T(i-1,\lfloor j+\ln_Q (1-\alpha) \rfloor) + W_i \}$

### Lemma 1: $\tau(i,Q^{j-i}) \leq T[i,j] \leq \tau(i,Q^j)$

## $T(i,j)=\min \max \{ \begin{array}{c} T(i-1, \lfloor j+\ln_Q \alpha \rfloor) \\ \alpha \in [0,1] \end{array} : T(i-1, \lfloor j+\ln_Q (1-\alpha) \rfloor) + w_i \end{cases}$

Lemma 2: can compute recursion efficiently

only few values of α matter Q<sup>-j</sup>,....,Q<sup>0</sup>, 1-Q<sup>0</sup>, ...., 1-Q<sup>j</sup> can use binary search

TOTAL RUN TIME =  $O(\frac{n^3}{\epsilon}\log(n/\epsilon))$ 

## How to deal with more constraints ?

(e.g., contingency tables, multi-dimensional knapsack, ...)

### multi-dimensional knapsack:

### How many $S \subseteq [n]$ with $\overline{}$

 $W_{i,i} \leq$ 

 $i \in S$ 

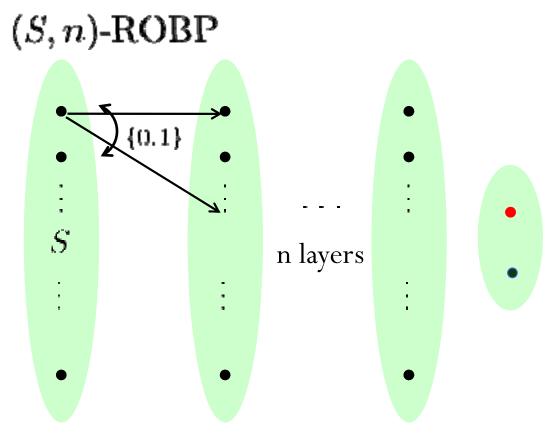


j∈{1,...,k}

### are there?

### O( $(n/\epsilon)^{O(k^2)} \log W$ ) algorithm

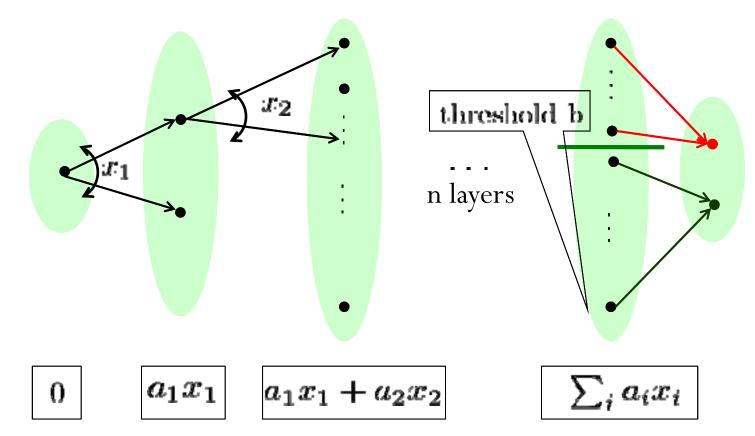
### Read once branching programs



- Layered directed graph
- S vertices per layer
- Edges between consecutive layers
- Edges labeled  $\{0,1\}$
- Input:  $(x_1, ..., x_n)$
- Output: Label of final vertex reached

Counting the number of accepting paths? dynamic programming, time = O(nS)

### **ROBP for knapsack**



### **Problem:** width too large

Solution: reduce width by approximating

### Monotone ROBPs

accepting paths from u

• U

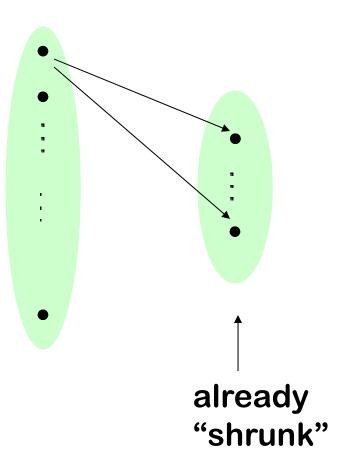
٩V

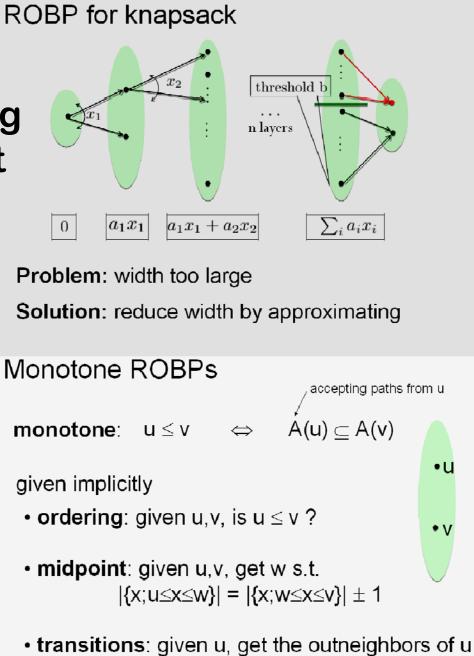
monotone:  $u \leq v \quad \Leftrightarrow \quad A(u) \subseteq A(v)$ 

given implicitly

- ordering: given u,v, is  $u \le v$ ?
- midpoint: given u,v, get w s.t.  $|\{x; u \le x \le w\}| = |\{x; w \le x \le v\}| \pm 1$
- transitions: given u, get the outneighbors of u

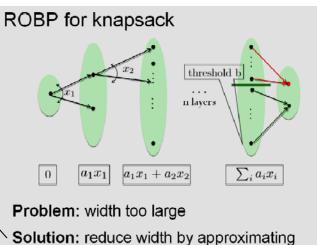
group the vertices in the R layers according to the rough number of accepting paths processing right-left



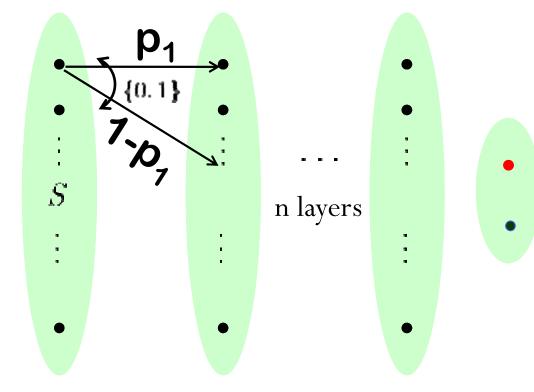


### More constraints?

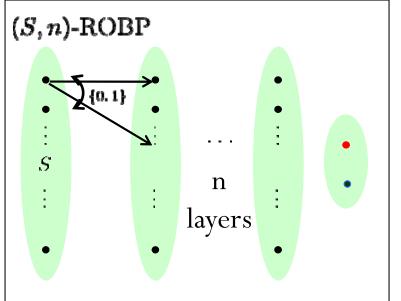
can be generalized to distribution given by small space sources.

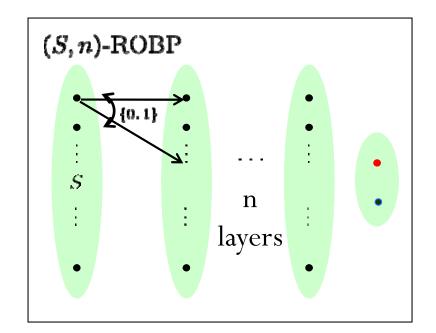


#### small space sources = ROBP + probability distributions on outgoing edges



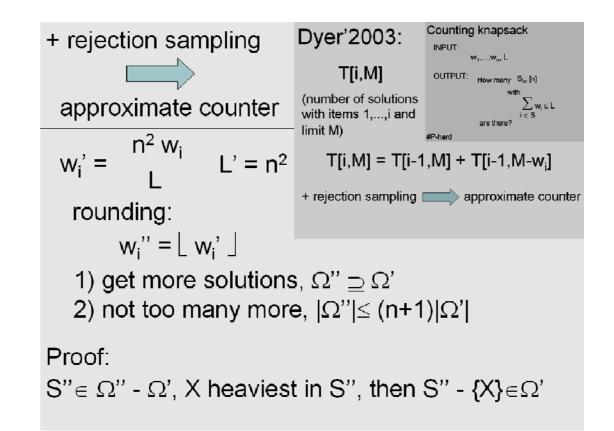




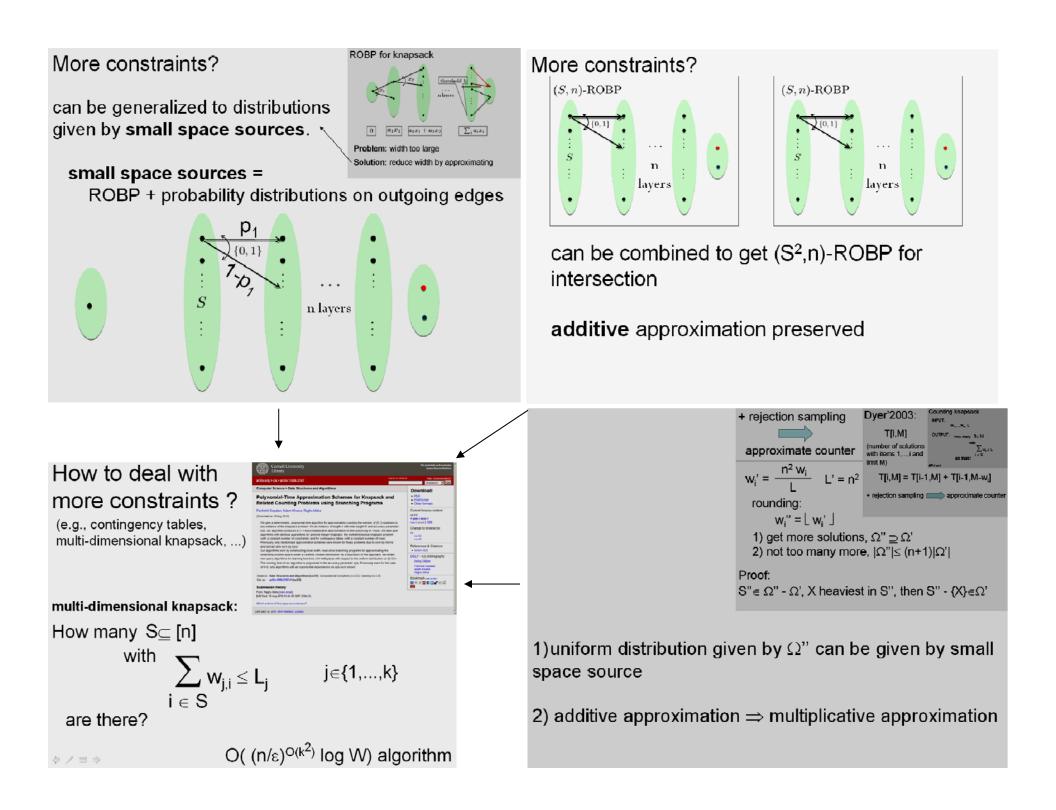


## can be combined to get (S<sup>2</sup>,n)-ROBP for intersection

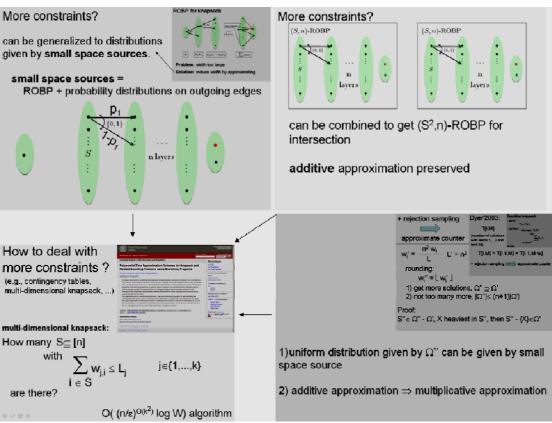
additive approximation preserved



- 1) uniform distribution given by  $\Omega$ ' can be given by small space source
- 2) additive approximation  $\Rightarrow$  multiplicative approximation



# Other problems:Contingency tables<br/>with constant<br/>number of rowsMore constraints?Contingency tables<br/>with constant<br/>number of rows



### What other problems are solvable using the technique?

### Thanks!