

New Algorithmic Results for Scheduling and Bin Packing

Klaus Jansen¹

¹University of Kiel

Joint Work with Lin Chen, Kim-Manuel Klein, Lars Rohwedder, José Verschae and Gouchuan Zhang

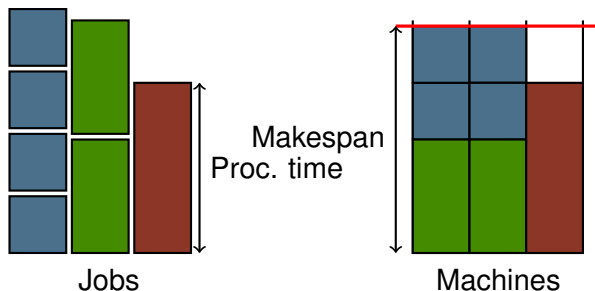
Overview:

Main Topics

- ▶ Scheduling on Identical Machines
- ▶ Integer Programming
- ▶ Bin Packing
- ▶ Open Problems

Scheduling on Identical Machines $P||C_{max}$:

- ▶ Given: n jobs with processing times p_j
- ▶ and m machines
- ▶ Objective: Minimize makespan (maximum machine load)



Literature:

Complexity

- ▶ Strongly NP-hard: If $P \neq NP$, then there is no FPTAS.

Known Algorithms

There is a PTAS with running time:

- ▶ $n^{O(\frac{1}{\epsilon^2})}$ [Hochbaum & Shmoys '87]
- ▶ $n^{O(\frac{1}{\epsilon} \log(\frac{1}{\epsilon}))}$ [Leung 97]

There is an EPTAS with running time:

- ▶ $2^{2^{\tilde{O}(\frac{1}{\epsilon})}} + O(n \log n)$ [Alon et al. '98 & H. & S. '96]
- ▶ $2^{\tilde{O}(\frac{1}{\epsilon^2})} + O(n \log n)$ [Jansen '10]

Closing the Gap:

Lower Bound [Chen, Jansen, Zhang '13]

- ▶ If the *Exponential Time Hypothesis* holds, there is no EPTAS with running time $2^{(\frac{1}{\varepsilon})^{1-\delta}} + \text{poly}(n)$.

Our Main Result:

Theorem [Jansen, Klein, Verschae '16]

Minimum makespan scheduling admits an EPTAS with running time

$$2^{\tilde{O}(\frac{1}{\varepsilon})} + O(n).$$

General Strategy:

General scheme for designing a PTAS:

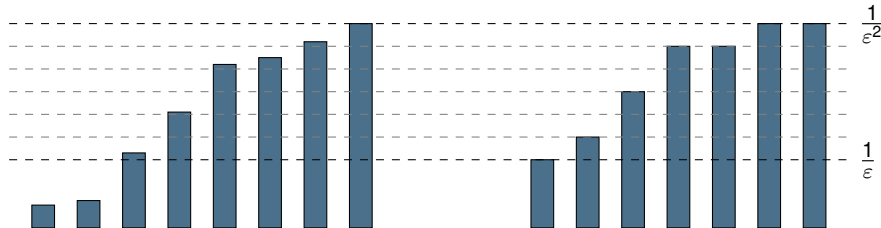
1. Guess the makespan T of the optimal solution.
2. Round instance $\rightsquigarrow (1 + \varepsilon)$ multiplicative loss in objective.
3. Solve the rounded instance using an ILP formulation.

Rounding:

Lemma (Rounding and scaling)

$T = 1/\varepsilon^2$ and jobs sizes belong to $\Pi = \{\pi_1, \dots, \pi_d\}$:

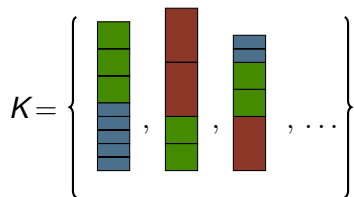
- ▶ $\Pi \subseteq \{\frac{1}{\varepsilon}, \frac{1}{\varepsilon} + 1, \dots, \frac{1}{\varepsilon^2}\}$ and, → integer numbers
- ▶ $|\Pi| = O(\frac{1}{\varepsilon} \log(\frac{1}{\varepsilon})) = \tilde{O}(\frac{1}{\varepsilon})$. → few sizes



Configurations:

A *configuration* represents one possibility of assigning jobs from Π to a single machine.

Example (The set of configurations)

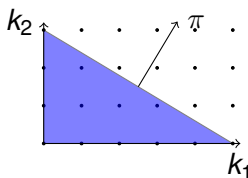


Configurations:

Knapsack polytope

$$P = \{k \in \mathbb{R}_{\geq 0}^{|\Pi|} : k^t \cdot \pi \leq T\}$$

Polyhedral view



Configurations:

Set of configurations

$$K := P \cap \mathbb{Z}_{\geq 0}^{|\Pi|}$$

Observation 1

$$|K| \leq (T + 1)^{|\Pi|} = 2^{O(\frac{1}{\varepsilon} \log^2(\frac{1}{\varepsilon}))} = 2^{\tilde{O}(\frac{1}{\varepsilon})}.$$

Integer Programming Formulation

Observation 2:

The vector $(x_k)_{k \in K}$ belongs to the system

$$\left. \begin{array}{l} \sum_{k \in K} x_k = m \\ \sum_{k \in K} k_i x_k = n_i \quad \text{for all } \pi_i \in \Pi \\ x \in \mathbb{Z}_{\geq 0}^K \end{array} \right\} \begin{array}{l} \# \text{ of constraints} = \tilde{O}\left(\frac{1}{\varepsilon}\right) \\ \# \text{ variables} = 2^{\tilde{O}\left(\frac{1}{\varepsilon}\right)} \end{array}$$

Solving the ILP, first Approach:

Method [Alon et al. '98] and [Hochbaum & Shmoys '97] uses

Theorem [Kannan '87 / Lenstra '83]

An integer program with N variables can be solved in time $2^{\tilde{O}(N)}$ (where s is the length of the input).

In our case $N = |K| = 2^{\tilde{O}(\frac{1}{\epsilon})}$ and thus the running time is

$$2^{\tilde{O}(N)} \log(n) = 2^{2^{\tilde{O}(\frac{1}{\epsilon})}} \log(n) \leftarrow \text{doubly exponential!}$$

Main Idea: Try to reduce the number of variables.

Solving the ILP, second Approach:

Guess the support [Jansen '10]

Theorem [Eisenbrand & Shmonin '06]

There is an optimum sol. x^* for $\{c^t x : Ax = b, x \geq 0, x \text{ integer}\}$
s.t. $|\text{support}(x^*)| \leq O(M(\log(M \cdot \Delta)))$ where

- ▶ M = number of constraints,
- ▶ Δ = largest coefficient in A, c .

In our case:

- ▶ $M = |\Pi| = \tilde{O}(\frac{1}{\varepsilon})$, and $\Delta = \frac{1}{\varepsilon}$
- ▶ $|\text{support}(x^*)| \leq \tilde{O}(\frac{1}{\varepsilon})$

Solving the ILP, second Approach:

Guess the support [Jansen '10]

Algorithm:

1. Try each possible support: there are $\tilde{O}(\frac{1}{\varepsilon}) \cdot \binom{|K|}{\tilde{O}(\frac{1}{\varepsilon})} = 2^{\tilde{O}(\frac{1}{\varepsilon^2})}$ many.
2. Solve ILP restricted to guessed variables with Kannan's algorithm (running time $2^{\tilde{O}(\frac{1}{\varepsilon}) \log(n)}$)
3. Total running time: $2^{\tilde{O}(\frac{1}{\varepsilon^2}) \log(n)}$.

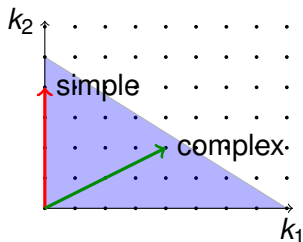
Solving the ILP, third Approach:

Understanding the Optimum

Definition

A configuration k is *complex* if it contains more than $\log(T + 1)$ different sizes; o.w. it is *simple*.

Example ($\log(T + 1) = 1$)



Example ($\log(T + 1) = 3$)



Simple



Complex

Solving the ILP, third Approach:

Understanding the Optimum

A “subconfiguration” $k' \leq k$ of configuration k is called *maximal* if it contains all possible jobs of each taken size.



Original
Configuration



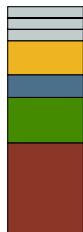
Maximal
Subconfiguration



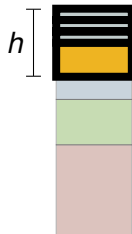
Non-Maximal
Subconfiguration

Lemma

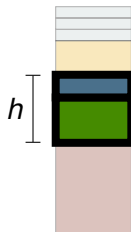
Every complex conf. $k \in K$ contains two maximal disjoint subconfigurations k_1, k_2 s.t. the total size of k_1 and k_2 coincide.



Complex
Configuration k



Subconfiguration
 k_1



Subconfiguration
 k_2

Lemma

Every complex conf. $k \in K$ contains two maximal disjoint subconfigurations k_1, k_2 s.t. $\pi \cdot k_1 = \pi \cdot k_2$.

Proof.

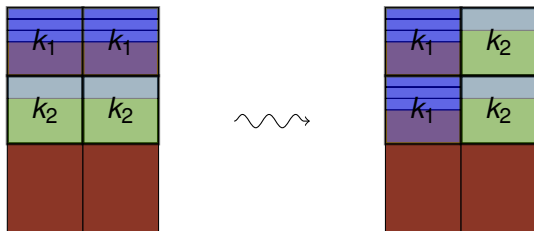
- ▶ Let $C > \log(T + 1)$ be the number of sizes (colors) in k .
- ▶ Number of maximal subconfigurations $= 2^C > T + 1$.
- ▶ Total size of each configuration is in $\{0, 1, 2, \dots, T\}$.
- ▶ Pigeonhole principle \Rightarrow there are two maximal subconfigurations of same total size.



Solving the ILP, third Approach:

Lemma (Sparsification Lemma (informal))

If a complex configuration is taken twice in a solution, then we can replace it by two other “less complex” configurations.

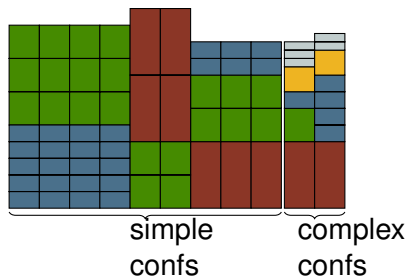


Solving the ILP, third Approach:

Theorem (Thin solutions)

If the ILP is feasible, then there is a solution x^ such that:*

- ▶ *At most $\tilde{O}(\frac{1}{\epsilon})$ machines get complex configurations.*
- ▶ *Each complex configuration is used at most once.*
- ▶ $|support(x^*)| \leq O(|\Pi| \log(|\Pi| T)) = \tilde{O}(\frac{1}{\epsilon})$.



Lemma

The number of simple configurations in K is $2^{O(\log^2(\frac{1}{\varepsilon}))} = 2^{\tilde{O}(1)}$.

Proof.

Let $D = \log(T + 1)$ and $T = 1/\varepsilon^2$.

$$\begin{aligned} \# \text{ simple conf} &\leq \sum_{i=0}^D \binom{|\Pi|}{i} \times (T + 1)^i \\ &\leq (D + 1) |\Pi|^D \times (T + 1)^D \\ &\leq \left(\frac{1}{\varepsilon} \log\left(\frac{1}{\varepsilon}\right)\right)^{O(\log(\frac{1}{\varepsilon}))} \\ &\leq 2^{O(\log^2(\frac{1}{\varepsilon}))} \leq 2^{\tilde{O}(1)}. \end{aligned}$$



Solving the ILP, third Approach:

Algorithm

Part 1: Complex Configurations.

1. Guess jobs assigned to complex configurations and number of complex machines.
2. Solve that subinstance optimally with a dynamic program.

Solving the ILP: Third Approach

Algorithm

Part 2: Remaining Instance.

1. Guess the (**simple!**) configurations in support:

$$\# \text{ possibilities} \leq \binom{2^{\tilde{O}(1)}}{\tilde{O}(\frac{1}{\varepsilon})} = 2^{\tilde{O}(\frac{1}{\varepsilon})}$$

2. For each possibility solve the ILP restricted to those variables with Kannan's algorithm.

Total running time: $2^{\tilde{O}(\frac{1}{\varepsilon})} \log(n)$

Main Result:

Algorithm

Theorem [Jansen, Klein, Verschae '16]

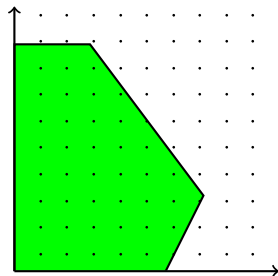
The minimum makespan problem on identical machines admits an EPTAS with running time

$$2^{O(\frac{1}{\varepsilon} \log^4(\frac{1}{\varepsilon}))} = 2^{\tilde{O}(\frac{1}{\varepsilon})} + O(n).$$

Integer Linear Programming

$$\begin{aligned} \max \quad & c^t x \\ \text{subject to} \quad & Ax = b \\ & x \in \mathbb{Z}_{\geq 0}^n \end{aligned}$$

where $A \in \mathbb{Z}^{M \times N}$, $b \in \mathbb{Z}^M$, $c \in \mathbb{Z}^N$.



Considered case

M (#constraints) is a constant, entries of A are small ($\leq \Delta$).

Pseudo-polynomial Algorithms

Known Algorithms

There is an algorithm with running time:

- ▶ $(M(\Delta + \|b\|_\infty))^{O(M^2)}$ [Papadimitrou '81]
- ▶ $N \cdot O(M\Delta)^{2M} \cdot \|b\|_\infty^2$. [Eisenbrand & Weismantel '18]

Theorem [Jansen & Rohwedder '19]

IP can be solved in time $O(M\Delta)^{2M} \cdot \log(\|b\|_\infty) + O(NM)$.

Moreover, improving the exponent to $2M - \delta$ is equivalent to finding a truly subquadratic algorithm for $(\min, +)$ -convolution.

Feasibility problem

Theorem [Jansen & Rohwedder '19]

Algorithm with running time:

$O(M\Delta)^M \cdot \log(\Delta) \cdot \log(\Delta + \|b\|_\infty) + O(NM)$. Improving exponent to $M - \delta$ would contradict the Strong Exponential Time Hypothesis (SETH).

Previous best result

$N \cdot O(M\Delta)^M \cdot \|b\|_\infty$.

[Eisenbrand & Weismantel '18]

Application $P||C_{max}$

Configuration IP

$$\begin{aligned}\sum_{k \in K} x_k &= m \\ \sum_{k \in K} k_i x_k &= n_i \quad \forall \pi_i \in \Pi \\ x_k &\in \mathbb{Z}_{\geq 0} \quad \forall k \in K\end{aligned}$$

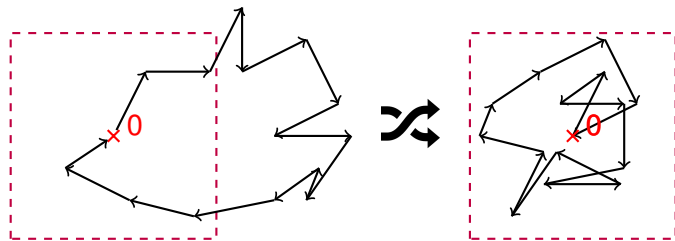
has $M + 1 = O(\frac{1}{\epsilon} \log(\frac{1}{\epsilon}))$ constraints and $N = |K| = 2^{O(\frac{1}{\epsilon})}$ many variables. The value $\Delta = \max_{k,i} k_i \leq \frac{1}{\epsilon}$ and $\|b\|_{\infty} \leq n$.

New result: Including preprocessing $O(n + \frac{1}{\epsilon} \log(\frac{1}{\epsilon}))$, we get:

$$\begin{aligned}O(M\Delta)^M \cdot \log(\Delta) \cdot \log(\Delta + \|b\|_{\infty}) + O(NM) + O(n + \frac{1}{\epsilon} \log(\frac{1}{\epsilon})) \\ \leq 2^{O(\frac{1}{\epsilon} \log^2(\frac{1}{\epsilon}))} \log(n) + O(n) \leq 2^{O(\frac{1}{\epsilon} \log^2(\frac{1}{\epsilon}))} + O(n).\end{aligned}$$

Steinitz Lemma

Let $\|\cdot\|$ be a norm in \mathbb{R}^M and $v^{(1)}, \dots, v^{(t)} \in \mathbb{R}^M$ with $\|v^{(i)}\| \leq 1$ $\forall i$ and $v^{(1)} + \dots + v^{(t)} = 0$. Then there is a permutation $\pi \in S_t$ with $\|\sum_{i=1}^j v^{(\pi(i))}\| \leq M$ for all $j = 1, \dots, t$.

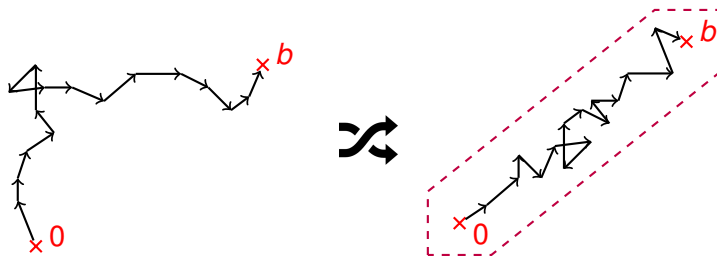


Consider an optimal solution x^* of (IP)
and the sequence of column vectors

$$\underbrace{A_1, \dots, A_1}_{x_1^* \text{ times}}, \underbrace{A_2, \dots, A_2}_{x_2^* \text{ times}}, \dots$$

$$\begin{aligned} \max \quad & c^t x \\ \text{subject to} \quad & Ax = b \\ & x \in \mathbb{Z}_{\geq 0}^N \end{aligned} \quad (\text{IP})$$

Recall that $\|A_i\|_{\infty} \leq \Delta$.



Steinitz for IP (formally)

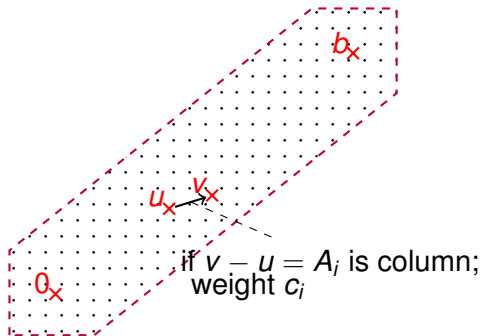
Corollary

Let $v^{(1)}, \dots, v^{(t)}$ denote columns of A with $\sum_{i=1}^t v^{(i)} = b$. Then there exists a permutation $\pi \in S_t$ such that for all $j \in \{1, \dots, t\}$

$$\left\| \sum_{i=1}^j v^{(\pi(i))} - j \cdot b/t \right\|_{\infty} \leq 2M\Delta.$$

This follows easily from the Steinitz Lemma: Insert vectors $\frac{v^{(i)} - b/t}{\Delta}$ in the Steinitz Lemma. Notice $\| \frac{v^{(i)} - b/t}{\Delta} \|_{\infty} \leq 2$.

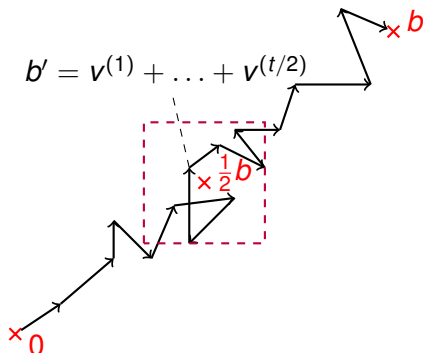
Eisenbrand & Weismantel



- ▶ Every $0 - b$ path gives a feasible solution
- ▶ Longest path is optimal
- ▶ $O(M\Delta)^M \cdot \|b\|_\infty$ vertices
- ▶ $N \cdot O(M\Delta)^M \cdot \|b\|_\infty$ edges
- ▶ Running time:
 $N \cdot O(M\Delta)^{2M} \cdot \|b\|_\infty^2$

Observation: There is an optimal solution of bounded norm,
i.e., $\|x\|_1 \leq O(M\Delta)^M \cdot \|b\|_\infty$.

Our Approach



Let $v^{(1)} + \dots + v^{(t)} = b$ be columns corresponding to an optimal solution of (IP).

Equivalent:

$v^{(1)} + \dots + v^{(t/2)}$ is optimal for

$$\{\max c^t x, Ax = b', x \in \mathbb{Z}_{\geq 0}^N\}$$

and $v^{(t/2+1)} + \dots + v^{(t)}$ is for

$$\{\max c^t x, Ax = b - b', x \in \mathbb{Z}_{\geq 0}^N\}.$$

If ordered via Steinitz Lemma, b' and $b - b'$ are not far from $\frac{1}{2}b$.

Assume w.l.o.g. there is an optimal solution x with $\|x\|_1 = 2^K$, where $K \in \log(O(M\Delta)^M \cdot \|b\|_\infty) = O(M \log(M\Delta) + \log(\|b\|_\infty))$

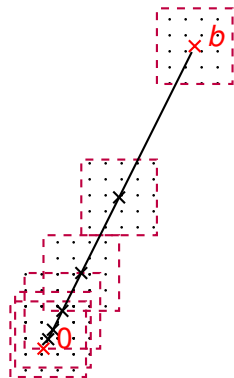
Solve for every $i = K, K-1, \dots, 0$ and every b' with

$$\left\| b' - \frac{1}{2^i} b \right\|_\infty \leq 4M\Delta$$

the problem

$$\begin{aligned} \max \quad & c^t x \\ \text{s.t.} \quad & Ax = b' \\ & \|x\|_1 = 2^{K-i} \\ & x \in \mathbb{Z}_{\geq 0}^N. \end{aligned}$$

Original problem for $i = 0$ and $b' = b$.



Consider iteration $i < K$ and b' with $\|b' - 1/2^i \cdot b\|_\infty \leq 4M\Delta$.

Let $v^{(1)}, \dots, v^{(2^{K-i})}$ be a solution of $\max\{c^t x, Ax = b', \|x\|_1 = 2^{K-i}, x \in \mathbb{Z}_{\geq 0}^N\}$ ordered via Steinitz. Set $b'' := v^{(1)} + \dots + v^{(2^{K-i-1})}$. Then we obtain

$$\left\| b'' - \frac{1}{2^{i+1}} b \right\|_\infty \leq \underbrace{\left\| b'' - \frac{1}{2} b' \right\|_\infty}_{\leq 2M\Delta} + \underbrace{\left\| \frac{1}{2} b' - \frac{1}{2^{i+1}} b \right\|_\infty}_{\leq 1/2 \cdot 4M\Delta} \leq 4M\Delta$$

Similarly, $\|(b' - b'') - \frac{1}{2^{i+1}} b\|_\infty \leq 4M\Delta$.

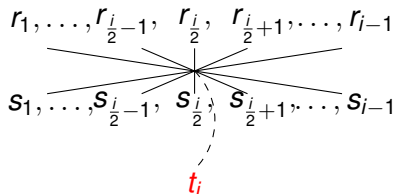
Our algorithm: Guess b'' ($O(M\Delta)^M$ candidates), look up solutions for $(i+1, b'')$ and $(i+1, b' - b'')$, and take the best.

Merging solutions

(MAX, +)-CONVOLUTION

Input: $r_1, \dots, r_n \in \mathbb{R}$,
 $s_1, \dots, s_n \in \mathbb{R}$

Output: $t_1, \dots, t_n \in \mathbb{R}$ with
 $t_i = \max_j [r_j + s_{i-j}]$



$T(n)$ time algorithm for (min, +)-convolution \Rightarrow
 $T(O(M\Delta)^M) \cdot O(M \log(M\Delta) + \log(\|b\|_\infty)) + O(NM)$ for IP.

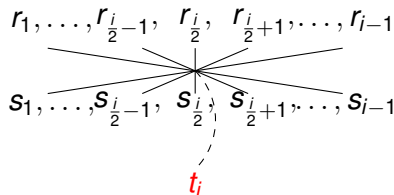
With $T(n) = O(n^2 / \log(n))$: $O(M\Delta)^{2M} \cdot \log(\|b\|_\infty) + O(NM)$.

Feasibility of IP

BOOLEAN-CONVOLUTION

Input: $r_1, \dots, r_n \in \{0, 1\}$,
 $s_1, \dots, s_n \in \{0, 1\}$

Output: $t_1, \dots, t_n \in \{0, 1\}$ s.t.
 $t_i = \bigvee_j [r_j \wedge s_{i-j}]$



Boolean Convolution can be computed in $T(n) = O(n \log n)$.

\Rightarrow Feasibility of IP in time

$$T(O(M\Delta)^M) \cdot (M \log(M\Delta) + \log(\|b\|_\infty)) + O(NM) \\
= O(M\Delta)^M \cdot \log(\Delta) \cdot \log(\Delta + \|b\|_\infty) + O(NM).$$

Bin Packing:

Problem Definition

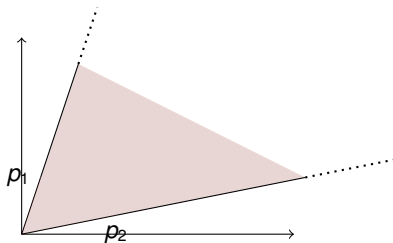
- ▶ d item sizes
- ▶ s_i : size of item
- ▶ b_i : multiplicity of item size s_i
- ▶ Objective: Find a packing into a minimum number of unit bins.



Cone:

Given a set of points $P \subset \mathbb{Z}^d$ then

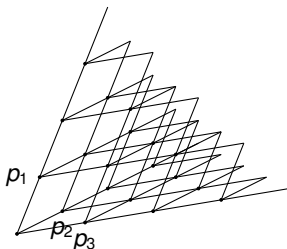
$$\text{Cone}(P) = \left\{ \sum_{p \in P} \lambda_p p \mid \lambda \in \mathbb{R}_{\geq 0}^P \right\}$$



Integer Cone:

Given a set of points $P \subset \mathbb{Z}^d$ then

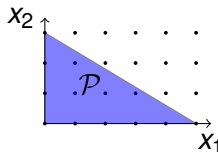
$$\text{int.cone}(P) = \left\{ \sum_{p \in P} \lambda_p p \mid \lambda \in \mathbb{Z}_{\geq 0}^P \right\}$$



Integer Cones of Polytopes:

Given Polytope $\mathcal{P} = \{x \in \mathbb{R}^d \mid Ax \leq c\}$ for some matrix $A \in \mathbb{Z}^{m \times d}$ and a vector $c \in \mathbb{Z}^d$.

Knapsack polytope $\mathcal{P} = \{x \in \mathbb{R}^d \mid s_1 x_1 + \dots s_d x_d \leq 1, x \geq 0\}$ for sizes s_1, \dots, s_d .



We consider $\text{int.cone}(\mathcal{P} \cap \mathbb{Z}^d)$.

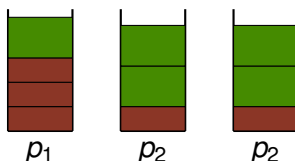
The Bin Packing Problem:

Given: a set of item sizes $s_1, \dots, s_d \in (0, 1]$ and multiplicities b_1, \dots, b_d of the corresponding item sizes.

Objective: Find a packing into a minimum number of unit bins.

Example:

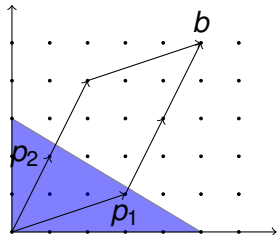
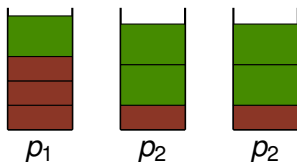
Item sizes: $s_1 = \frac{1}{5}$, $s_2 = \frac{1}{3}$ with multiplicities: $b_1 = 5$, $b_2 = 5$



The Bin Packing Problem:

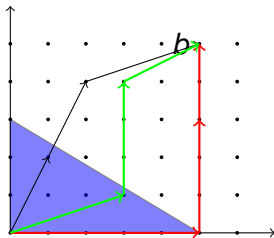
Each vector $\lambda \in \mathbb{Z}_{\geq 0}^{\mathcal{P} \cap \mathbb{Z}^d}$ with

$\sum_{p \in \mathcal{P} \cap \mathbb{Z}^d} \lambda_p p = b \in \text{int.cone}(\mathcal{P} \cap \mathbb{Z}^d)$ represents a possible solution of the bin packing problem.



Structural Properties:

Arguments about the set of possible solutions $\lambda \in \mathbb{Z}_{\geq 0}^{P \cap \mathbb{Z}^d}$.



The Structure of the Integer Cone:

Theorem [Eisenbrand, Shmonin '06]

For any integral point $b \in \text{int.cone}(\mathcal{P} \cap \mathbb{Z}^d)$, there exists an integral vector $\lambda \in \mathbb{Z}_{\geq 0}^{\mathcal{P} \cap \mathbb{Z}^d}$ such that $b = \sum_{p \in \mathcal{P} \cap \mathbb{Z}^d} \lambda_p p$ and $|\text{supp}(\lambda)| \leq 2^d$.

The Structure of the Integer Cone:

Theorem [Goemans & Rothvoß'14]

There exists a set $X \subseteq \mathcal{P} \cap \mathbb{Z}^d$ with $|X| \leq m^d d^{O(d)} (\log \Delta)^d$ such that for any point $b \in \text{int.cone}(\mathcal{P} \cap \mathbb{Z}^d)$, there exists an integral vector $\lambda \in \mathbb{Z}_{\geq 0}^{\mathcal{P} \cap \mathbb{Z}^d}$ such that $b = \sum_{p \in \mathcal{P} \cap \mathbb{Z}^d} \lambda_p p$ and

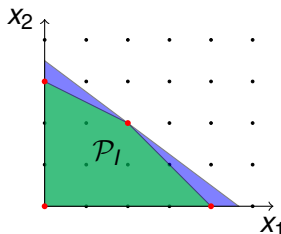
1. $\lambda_p \leq 1 \quad \forall p \in (\mathcal{P} \cap \mathbb{Z}^d) \setminus X$
2. $|\text{supp}(\lambda) \cap X| \leq 2^{2d}$
3. $|\text{supp}(\lambda) \setminus X| \leq 2^{2d}$

Theorem [Goemans & Rothvoß'14]

Bin packing with d different item sizes can be solved in time $(\log \Delta)^{2^{O(d)}}$, where Δ is the maximum over all multiplicities b and denominators in s .

Vertices of the Integer Polytope:

Integer Polytope $\mathcal{P}_I = \text{Conv}(\mathcal{P} \cap \mathbb{Z}^d)$



$$s_1 = \frac{3}{14}$$
$$s_2 = \frac{2}{7}$$

Theorem [Cook et al. '92]

For a polytope $\mathcal{P} = \{x \in \mathbb{R}^d \mid Ax \leq c\}$ the integer polytope \mathcal{P}_I has at most $m^d \cdot O((\log \Delta)^d)$ vertices.

Our Structure Theorem:

Theorem [Jansen & Klein '17]

Let $V_I \subseteq \mathcal{P} \cap \mathbb{Z}^d$ be the set of vertices of the integer polytope \mathcal{P}_I . Then for any vector $b \in \text{int.cone}(\mathcal{P} \cap \mathbb{Z}^d)$, there exists an integral vector $\lambda \in \mathbb{Z}_{\geq 0}^{\mathcal{P} \cap \mathbb{Z}^d}$ such that $b = \sum_{p \in \mathcal{P} \cap \mathbb{Z}^d} \lambda_p p$ and

1. $\lambda_p \leq 2^{2^{O(d)}} \quad \forall p \in (\mathcal{P} \cap \mathbb{Z}^d) \setminus V_I$
2. $|\text{supp}(\lambda) \cap V_I| \leq d \cdot 2^d$
3. $|\text{supp}(\lambda) \setminus V_I| \leq 2^{2^d}$

Theorem [Jansen & Klein '17]

The bin packing problem can be solved in time $|V_I|^{2^{O(d)}} \cdot (\log \Delta)^{O(1)}$ and hence in *FPT*-time, parameterized by the number of vertices V_I .

Main Open Questions:

- ▶ Is there an EPTAS for scheduling on identical machines with running time $2^{O(1/\epsilon)} + O(n)$?
- ▶ Is there an FPT-algorithm for bin packing parameterized by the number d of different sizes?

Thanks for your attention!