

New Jersey Institute of Technology

Quick Minimization of Tardy Processing Time on a Single Machine

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Joint work with

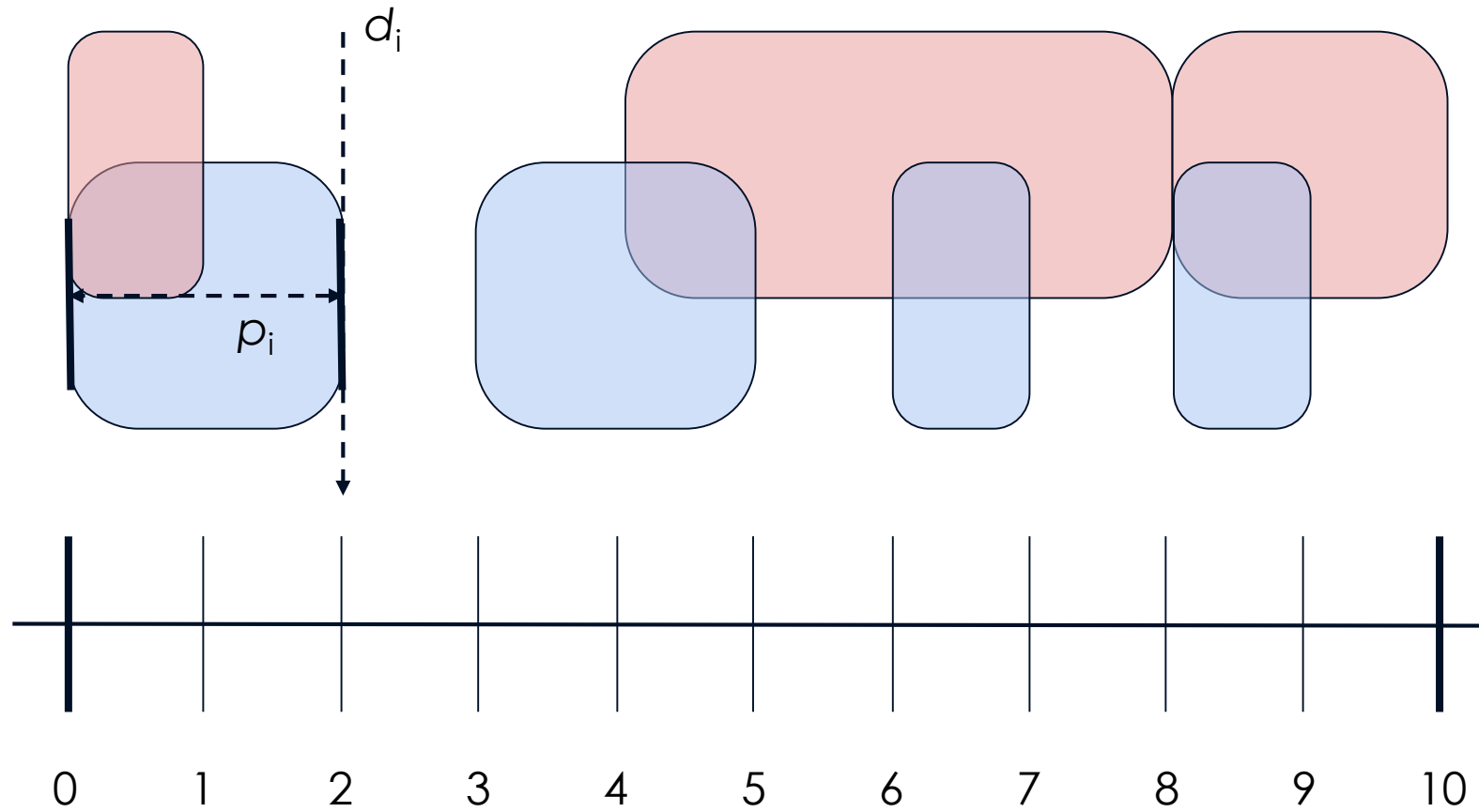
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Minimum Tardy Processing Time (MTPT) Problem

- n jobs j_1, j_2, \dots, j_n
- Each job j_i is associated with
 - processing time p_i
 - due date d_i
- A job is **tardy** if it terminates after its due date.
- **Goal:** Find a feasible schedule of the jobs that minimizes the **total processing time** of tardy jobs.
- Assume that the jobs are ordered by due dates $d_1 < d_2 < \dots < d_{D\#}$

Example



MTPT when all jobs have the same due date

- Let d be the due date and P be the sum of all processing times
- $P-d$ is a lower bound on the tardy processing time
- This lower bound can be achieved iff there is a subset of jobs whose total processing time is exactly d
- The Subset Sum problem (problem Weakly NP-Hard)

A pseudo polynomial time algorithm for MTPPT

- [Lawler Moore 1969]
 - Any instance of the problem has an optimal Earliest Due Date (EDD) schedule.
 - In such a schedule:
 1. any early job precedes all late jobs
 2. any early job precedes all early jobs with later due dates
 - **Dynamic programming**: scan the jobs in order of due dates and at each stage maintain all the feasible “prefixes” of the EDD schedules
 - $O(nP)$ time

Is there a faster algorithm for MTPT?

- [Bringmann et al. 2020]
 - Defined (max,min)-skewed-convolution
 - Showed an $\tilde{O}(P^\alpha)$ time algorithm for MTPT, where $\tilde{O}(P^\alpha)$ is the running time of a (max,min)-skewed-convolution of 2 vectors of size P
 - gave an $\tilde{O}(P^{7/4})$ time algorithm for a (max,min)-skewed convolution and thus also for MTPT
- [Klein et al. 2022]
 - gave an $\tilde{O}(P^{5/3})$ time algorithm for a generalized (max,min)-skewed convolution and thus for the MTPT problem

Our result

- An $\tilde{O}(P^{2-1/\alpha})$ time algorithm for MTPT, where $\tilde{O}(P^\alpha)$ is the running time of a **(max,min)-skewed-convolution** of vectors of size P
- Results in an $\tilde{O}(P^{7/5})$ time algorithm for MTPT
- Breaks the $\tilde{O}(P^{3/2})$ time barrier of the previous approach
 - this is the running time of the best known and decades old algorithm of a **(max,min)-convolution** [Kosaraju 1989]
- Faster than [Lawler Moore 1969] when $n = \tilde{\omega}(P^{2/5})$

(max,min)-skewed-convolution

- Given two vectors with $n+1$ entries:
 - $A[0], \dots, A[n]$ and $B[0], \dots, B[n]$
- The (max,min)-skewed-convolution of A and B is a the $2n+1$ vector $C[0], \dots, A[2n]$ defined as
 - $C[k] = \max_{i+j=k} \min\{A[i], B[j] + k\}$
- We use a slight (equivalent) variation in which
 - $C[k] = \max_{i+j=k} \min\{A[i], B[j] - i\}$

Sumsets and set of subset sums

- Let A and B be two vectors of integers in the range $[0..P]$
- The **sumset** $A \oplus B = \{a + b \mid a \in A, b \in B\}$
 - Can be computed in $\tilde{O}(P)$ time via **(+,x)-convolution** of vector of size P
- The set of subset sums $\mathbf{S}(A) = \{\sum_{a \in Z} a \mid Z \subseteq A\}$
 - Can be computed in $\tilde{O}(\sum_{a \in A} a)$ time by successive sumset computations

An $\tilde{O}(P \bullet D\#)$ algorithm [Bringmann et al. 2020]

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- 1: Let $d_1 < \dots < d_{D\#}$ denote the different due dates of jobs in \mathcal{J} .
 - 2: **for** $i = 1, \dots, D\#$ **do**
 - 3: Compute $X_i = \{p_j : J_j \in \mathcal{J}_i\}$
 - 4: Compute $\mathcal{S}(X_i)$
 - 5: Let $S_0 = \emptyset$.
 - 6: **for** $i = 1, \dots, D\#$ **do** *▷ compute the sumsets and exclude infeasible sums*
 - 7: Compute $S_i = S_{i-1} \oplus \mathcal{S}(X_i)$.
 - 8: Remove any $x \in S_i$ with $x > d_i$.
 - 9: Return $P - x$, where x is the maximum value in $S_{D\#}$.
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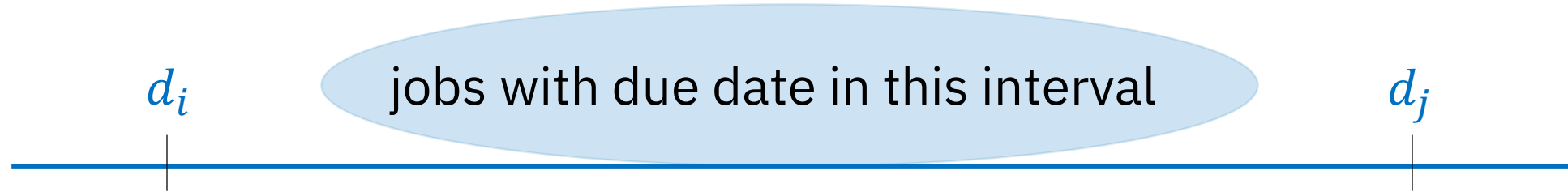
Job bundles

- Parameter $\delta \in (0,1)$
- **Red due dates** d_i all due dates such that the sum of processing time of jobs with this due date $> P^{1-\delta}$
- Group the jobs with the rest of the due dates into **bundles**
 - **bundle**: defined by maximal consecutive subsequences of due dates, none of which are red, such that the sum of processing time of jobs with these due dates $\leq P^{1-\delta}$
- The number of **red due dates** and the number of **bundles** is $O(P^\delta)$

The improved algorithm: outline

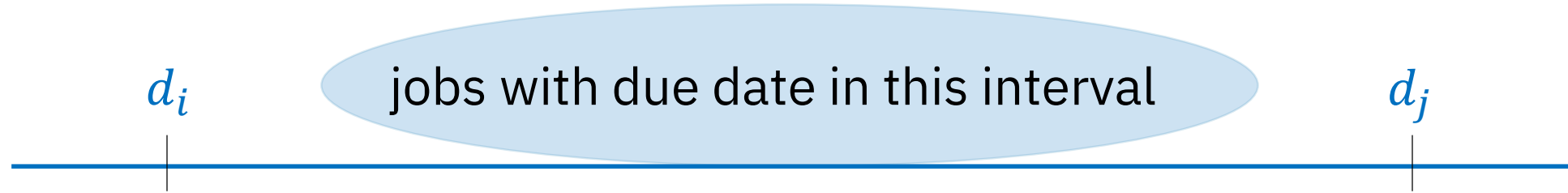
- Follows the structure of Algorithm 1 with additional processing of entire bundles that avoids processing each due date in a bundle individually
- The bundles are processed using a **(max, min)-skewed-convolution**
- Computing the bundles and red due dates takes $\tilde{O}(P)$ time
- Processing each bundle takes $\tilde{O}(P^{(1-\delta)\alpha} + P)$ time, where $\tilde{O}(P^\alpha)$ is the running time of a **(max,min)-skewed-convolution** of vectors of size P
- Processing each **red** due date takes $\tilde{O}(P)$ time
- Total running time $\tilde{O}(P^\delta P^{(1-\delta)\alpha} + P^\delta P)$
- Substituting $1-\delta=1/\alpha$ we get $\tilde{O}(P^{2-1/\alpha})$ time

Processing job bundles



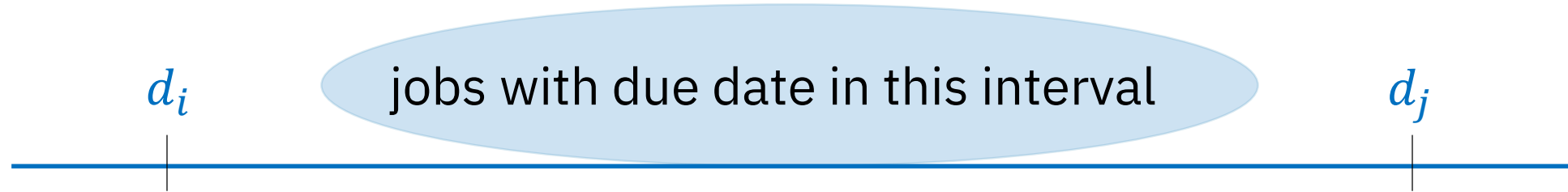
- J - the set of jobs in the bundle
- P_J - the total processing time of the jobs in J
- Compute the vector M
 - $M[x]$: the latest time t starting at which a subset of jobs in J with total processing time x can be scheduled feasibly , $-\infty$ otherwise
 - Can be done in $\tilde{O}(P_J)$ time via a (max,min)-skewed-convolution

Processing job bundles (cont.)



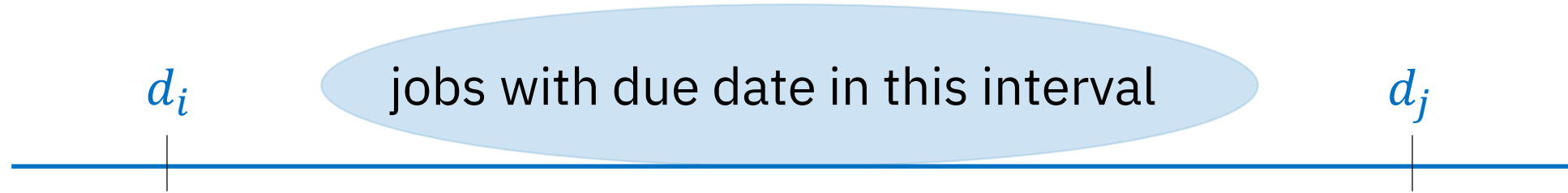
- **Input:** S_i - the processing times of all feasible schedules of jobs with due date up to (and including) d_i
- **Output:** S_j - the processing times of all feasible schedules of jobs with due date up to (and including) d_j
- Initially, $S_j \leftarrow S_i$

Computing S_i



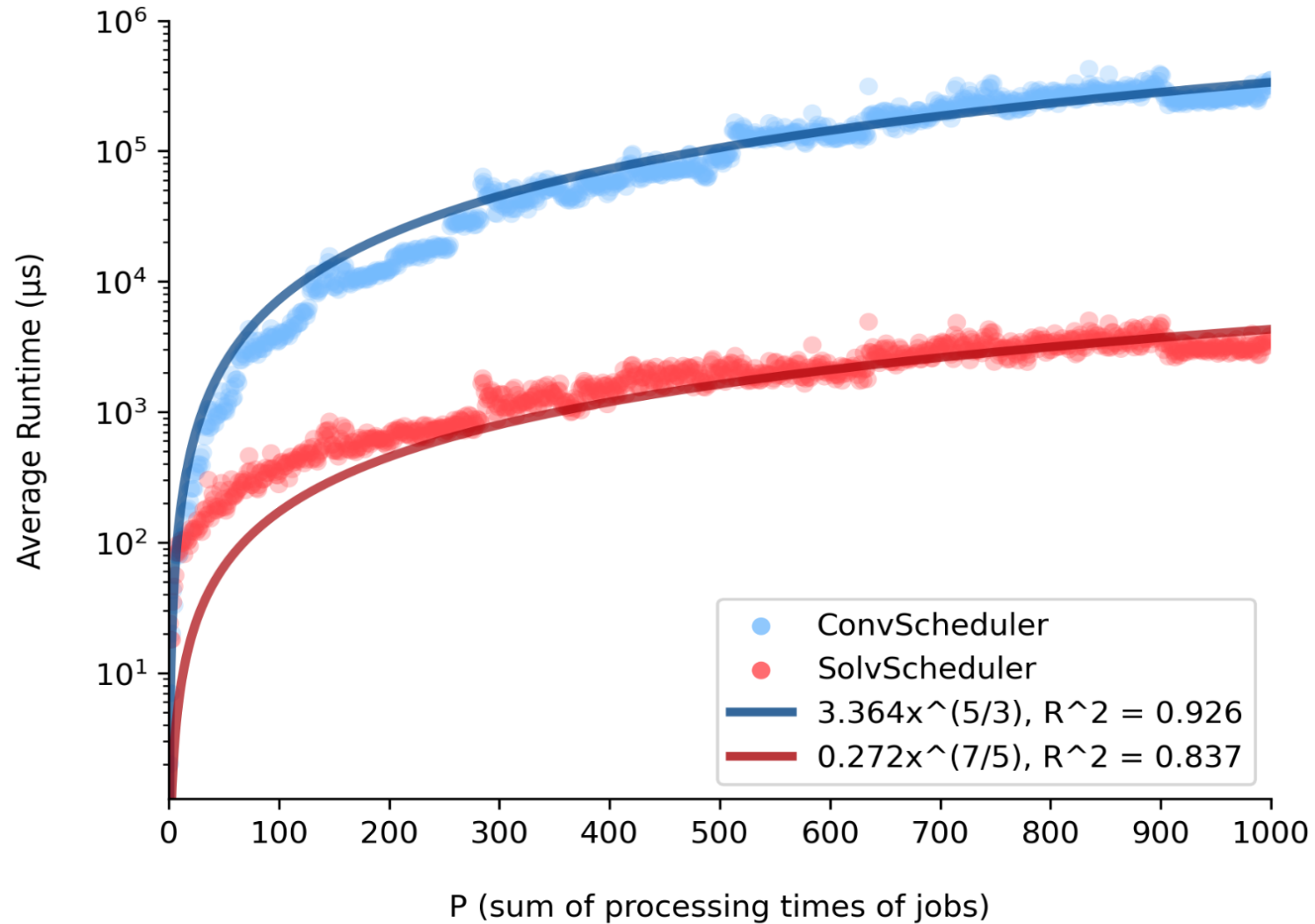
- $T_2 = \{x | M[x] > -\infty\}$ and $T_1 = S_i \cap [0.. d_i - P_J]$
- $S_j \leftarrow S_j \cup (T_1 \oplus T_2)$
- We are left with feasible schedules in which jobs with due date up to (and including) d_i end in $(d_i - P_J, d_i]$
- To find these schedules we compute a **(max,min)-skewed-convolution**

Computing the rest of the feasible schedules



- Two vectors of size P_J :
 - $A[x] = 0$ if $x + d_i - P_J + 1 \in S_i$, $-\infty$ otherwise
 - $B[x] = M[x] - d_i + P_J - 1$
- $C[k] = \max_{x+y=k} \min\{A[x], B[y] - x\}$
- $C[k] = 0 \Rightarrow k + d_i - P_J + 1 \in S_j$

Runtime



Open problems

- Since it is reasonable to assume that computing a **(max, min)-skewed-convolution** requires $\tilde{\omega}(P^{3/2})$ time our technique is unlikely to yield a $\tilde{o}(P^{4/3})$ running time
- It will be interesting to see whether this running time barrier can be broken, and whether the MTPT problem can be solved without computing a **(max, min)-skewed-convolution**
- Faster algorithms for **(max, min)-skewed-convolution**

