Approximation Algorithms for the Joint Replenishment Problem with Deadlines

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Approximation Algorithms for the Joint Replenishment Pro

## The problem, v1, acknowledgement aggregation



We will mostly consider hight 2 trees

## The problem, v2, joint replenishment



- given set of demands (retailer, time interval)
- compute a valid delivery schedule to quickly replenish used stock at retailers
- we may assume no storage at the warehouse
- minimize shipment costs (retailer orders + warehouse orders)
- pay per order, independent of the amount of items shipped

## The problem, v3, make to order production planning



- just like before, but time is reversed
- given set of demands (retailer, time interval)
- compute a valid delivery schedule to provide items in time
- we may assume no storage at the warehouse
- minimize shipment costs (retailer orders + warehouse orders)
- pay per order, independent of the amount of items shipped

#### The problem, we decide for notation v2:JRPD



- cost model: linear waiting cost vs. deadlines
- some algorithms work in both models
- today we concentrate on deadlines
- also consider the uniform deadline case

$$\begin{array}{lll} \text{minimize} & cost(\mathbf{x}) \ = \ \sum_{t=1}^{U} \left( C \, x_t + \sum_{\rho=1}^{m} c_{\rho} \, x_t^{\rho} \right) \\ \text{subject to} & x_t \ \ge \ x_t^{\rho} & \text{for all } t \in \mathcal{U}, \rho \in \{1, \dots, m\} \ \ (1) \\ & \sum_{t=r}^{d} x_t^{\rho} \ \ge 1 & \text{for all } (\rho, r, d) \in \mathcal{D} \\ & x_t, x_t^{\rho} \ \ge 0 & \text{for all } t \in \mathcal{U}, \rho \in \{1, \dots, m\}. \end{array}$$

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- NP-complete : Becchetti et al. '09
- APX-hard (even for 3 demands per retailer) : Nonner and Souza '09
- 2-apx. primal-dual algorithm: Levi, Roundy and Shmoys '06
- 1.8-apx. : Levi et. al '08
- $5/3 \approx 1.67$  -apx. : Nonner and Souza '09

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For general demands:

- e/(e-1) pprox 1.58 apx. (easier version of the analysis)
- 1.574-apx. (more refined analysis)
- 1.245-lower bound on the inegrality gap

For equal length intervals

- 1.5-apx.
- APX-hardness (even with up to 4 demands per retailer)
- 1.2-lower bound on integrality gap

- Cost naturally splits into warehouse and retailer orders
- We consider LP-rounding alg. which directly relate the cost of the algorith to the corresponding part of LP-cost
- We bound  $ALG = ALG_w + ALG_r \le \lambda_w LP_w + \lambda_r LP_r$
- We say ALG is a  $(\lambda_w, \lambda_r)$ -apx algorithm.
- We will next show a (1,2) and a (3, 1.5)-apx algorithm

# Easy algorithms: (1,2)-apx

- One level problem is easy
- Ignore retailer level cost to compute warehouse orders
- Compute optimal retailer orders given the fixed warehouse orders
- Show that retailer orders are now only twice more expensive than in OPT (or in LP)





Single retailer orders in OPT

- LP encodes density of shipments over time
- define "LP-time" between two events and the total LP-shipment between these events
- observe that "LP-time geos faster" on warehouse edge than on any retailers edge

Consider the following algorithm:

- plan warehouse shipment every 1/3 of "warehouse LP-time"
- plan retailer ho shipment every 2/3 of "retailer ho LP-time"

Note that we may combine (1,2) and (3, 1.5) to get (1.8, 1.8), which is essentially the work of Levi et al.

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Instead of scheduling warehouse orders every 1/3 of LP-time, we do it iteratively and the next order is selected to be a certain random distance from the previous one.

Fix  $\theta = 0.36455$  (slightly less than 1/e). Over the half-open interval [0, 1), the probability density function p is

$$p(y) = \begin{cases} 0 & \text{for } y \in [0, \theta) \\ 1/y & \text{for } y \in [\theta, 2\theta) \\ \frac{1 - \ln((y-\theta)/\theta)}{y} & \text{for } y \in [2\theta, 1). \end{cases}$$

The probability of choosing 1 is  $1 - \int_0^1 p(y) \, dy \approx 0.0821824$ .



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# Algorithm

#### **Algorithm** Round<sub>*p*</sub>( $C, c_{\rho}, D, x$ )

- 1: Draw independent random samples  $s_1, s_2, \ldots$  from p. Let  $g_i = \sum_{h \leq i} s_h$ . Set global cutoff sequence  $g = (g_1, g_2, \ldots, g_l)$ , where  $l = \min\{i \mid g_i \geq \hat{U} - 1\}$ .
- For each retailer ρ independently, choose ρ's local cutoff sequence ℓ<sup>ρ</sup> ⊆ g greedily to touch all intervals [a, b] with ω<sub>ρ</sub>(b) ω<sub>ρ</sub>(a) ≥ 1. That is, ℓ<sup>ρ</sup> = (ℓ<sup>ρ</sup><sub>1</sub>, ℓ<sup>ρ</sup><sub>2</sub>, ..., ℓ<sup>ρ</sup><sub>Jρ</sub>) where ℓ<sup>ρ</sup><sub>j</sub> is max{g ∈ g | ω<sub>ρ</sub>(g) - ω<sub>ρ</sub>(ℓ<sup>ρ</sup><sub>j-1</sub>) ≤ 1} (interpret ℓ<sup>ρ</sup><sub>0</sub> as 0), and J<sup>ρ</sup> is min{j | ω<sub>ρ</sub>(Û) - ω<sub>ρ</sub>(ℓ<sup>ρ</sup><sub>j</sub>) ≤ 1}.
  For each g: ∈ g define time t: ∈ [U] to be minimum such that
- 3: For each  $g_i \in g$ , define time  $t_i \in [U]$  to be minimum such that  $\sum_{t=1}^{t_i} x_t \geq g_i$ . Return the schedule  $\{(t_i, \{\rho \mid g_i \in \ell^{\rho}\}) \mid g_i \in g\}.$

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# Algorithm: intuition



- go forward one unit of retailer time (deadline)
- 2 go backward one unit of warehouse time
- go forward to the next warehouse order (there must be at least one)
- see how much time left before deadline (z on the picture)

#### Algorithm: fragment of easier variant of analysis

- take probability density function p(y) = 1/y for  $y \in [1/e, 1]$ and p(y) = 0 elsewhere.
- we bound the move on the warehouse time:  $\mathbf{E}[p] = \int_{1/e}^{1} y \ p(y) \ dy = \int_{1/e}^{1} 1 \ dy = 1 - 1/e.$
- we bound the waist on the retailer time:  $\Pr[s_1 > z]z + \Pr[s_1 \le z] \mathbb{E}[z - s_1 | s_1 \le z].$ This simplifies to  $z - \Pr[s_1 \le z] \mathbb{E}[s_1 | s_1 \le z]$ , which by calculation is

$$z - \int_{1/e}^{z} y \, p(y) \, dy = z - \int_{1/e}^{z} dy = z - (z - 1/e) = 1/e.$$

Let random index  $T \in \{0, 1, 2, ...\}$  be a stopping time for the sequence, that is, for any positive integer t, the event "T < t" is determined by state  $S_t$ .

#### Lemma (Wald's equation)

Suppose that (i)  $(\forall t < T) \mathbf{E}[\phi(S_{t+1}) | S_t] \ge \phi(S_t) + \xi$  for fixed  $\xi$ , and (ii) either  $(\forall t < T) \phi(S_{t+1}) - \phi(S_t) \ge F$  or  $(\forall t < T) \phi(S_{t+1}) - \phi(S_t) \le F$ , for some fixed finite F, and T has finite expectation. Then  $\xi \mathbf{E}[T] \le \mathbf{E}[\phi(S_T) - \phi(S_0)]$ .

In the applications here, we always have  $\xi = \mathcal{Z}(p) > 0$  and  $\phi(S_T) - \phi(S_0) \leq U$  for some fixed U. In this case Wald's equation implies  $\mathbf{E}[T] \leq U/\mathcal{Z}(p)$ .

# 1.5-apx for uniform length demands

- instances of length 3 are plynomial time solvable
- create a set of small instances that cover all requests
- show that there exists solution to the set of small instances with cost 1.5 OPT

## APX-hardness for uniform length demands, sketch

- reduce from degree 3 vertex cover
- synchronize 3 instances of a vertex
- for each edge: put two fresch copies of its endpoints nearby
- show that VC using K vertices corresponds to a solution of cost 10.5n + K + 6



Approximation Algorithms for the Joint Replenishment Pro

#### related work: implication for general penalty cost

- best known apx. so far: 1.8
- we can now improve it to 1.791 by a combination of 3 algorithms [\*]



[\*] Joint work with Bienkowski, B., Chrobak, Jez, and Sgall

- 3-competitive alg. for JRP [Buchbinder et al '08]
- 2.753-lower bound on comp. ratio for linear waiting costs [\*]
- 2-competitive algorithm for online JRPD [\*]
- matching lower bound of 2 on competitiveness fro JRPD [\*]
- [\*] Joint work with Bienkowski, B., Chrobak, Jez, and Sgall

Offline:

- polynomial time solvable on line networks [\*]
- constant factor apx. for general trees

Online:

- already on a single edge it encodes a rent-or-buy problem
- 5-competitive alg. for a line [\*]
- $2+\phi$  lower bound on a line [\*]
- open for general trees

[\*] Joint work with Bienkowski, Chrobak, Jez, Sgall, and Stachowiak

Than you for your attention!



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