Approximation Algorithms for the Joint Replenishment Problem with Deadlines

Marcin Bienkowski, Jaroslaw Byrka, Marek Chrobak, Neil Dobbs, Tomasz Nowicki, Maxim Sviridenko, Grzegorz Świrszcz, Neal E. Young

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

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

æ

- 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

伺 ト イヨト イヨト

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 (λ_w, λ_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.

・ 同 ト ・ ヨ ト ・ ヨ ト

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 $1 - \int_0^1 p(y) \, dy \approx 0.0821824$.



・ロト ・聞 と ・ ほ と ・ ほ と … Approximation Algorithms for the Joint Replenishment Pro

æ

Algorithm

Algorithm Round_{*p*}(C, c_{ρ}, 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^ρ ⊆ g greedily to touch all intervals [a, b] with ω_ρ(b) ω_ρ(a) ≥ 1. That is, ℓ^ρ = (ℓ^ρ₁, ℓ^ρ₂, ..., ℓ^ρ_{Jρ}) where ℓ^ρ_j is max{g ∈ g | ω_ρ(g) - ω_ρ(ℓ^ρ_{j-1}) ≤ 1} (interpret ℓ^ρ₀ as 0), and J^ρ is min{j | ω_ρ(Û) - ω_ρ(ℓ^ρ_j) ≤ 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 $\{(t_i, \{\rho \mid g_i \in \ell^{\rho}\}) \mid g_i \in g\}.$

・ 同 ト ・ ヨ ト ・ ヨ ト ・ ヨ

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 ξ , 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!



æ