

Joint Learning & Dynamic Pricing

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- Dynamic pricing without knowing the demand function - Besbes & Zeevi
- Dynamic pricing with limited supply - Babaioff, Dughmi, Kleinberg, Slivkins

Model

- k items, n agents, sequential arrival ($k < n$)
(Alt: time horizon T , arrival rate λ Poisson process)
- Posted price P_t for agent t , $P \in [0, 1]$
- Agent t has value $V_t \sim F$. F unknown, $F \in [0, 1]$
- $S(p) = 1 - F(p)$ - Sales rate / quantile. $S(p)$ strictly dec
- $R(p) = p S(p)$ - Revenue function

$$F \text{ regular} \Rightarrow R''(p) \leq 0 \quad \forall p \in [0, 1]$$

$$F \text{ strictly regular} \Rightarrow R''(p) < 0$$

- P_{MP}^* \equiv Monopolist price ($\arg \max_p R(p)$)
- Fixed price benchmark - $A_k^n(p)$. If $k = \alpha n$, $A_k^n(p) = R(p)$

Lemma let $\lambda(p) = p \min(k, nS(p)) = \min(kp, nR(p))$

$$\text{then } \lambda(p) - O(p \sqrt{k \log k}) \leq \text{Rev}(A_k^n(p)) \leq \lambda(p)$$

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thus the best choice of fixed price is $p^* = \underset{p}{\operatorname{argmax}} \mathcal{D}(p)$ ②
 $= \max [P_{MP}, S^{-1}(\frac{k}{n})]$

Pf - The upper bound is obvious. For the lower bound, let

$$X_t \equiv \mathbb{1}\{\text{Sale to } t^{\text{th}} \text{ agent}\}, X = \sum_{t=1}^n X_t, \mu = \mathbb{E}[X] (= n S(p))$$

$$\text{By Chernoff bounds, } \mathbb{P}[X - \mu \leq -O(\sqrt{\mu \log k})] \leq \frac{1}{k}$$

$$\begin{aligned} \Rightarrow \# \text{ of sales} = \mathbb{E}[\min[k, X]] &\geq \min(k, \mu - O(\sqrt{\mu \log k})) \\ &\geq \min(k, \mu) - O(\sqrt{k \log k}) \end{aligned}$$

Capped UCB (n, k)

$$\text{Regret} = \mathcal{D}(p^*) - \mathbb{E}[\text{Rev}(A)]$$

- Choose parameter $\delta \in (0, 1)$
- Set 'active prices' set $\mathcal{P} \equiv \left\{ \underset{\in [0, 1]}{S(1+\delta)^i}; i \in \mathbb{N} \right\}$
- While ~~there~~ \exists unsold item
 - Pick $p \in \underset{p \in \mathcal{P}}{\operatorname{argmax}} \mathbb{I}_t(p)$
index for price p
- Else set $p = \infty$ (close shop!)

Thm - With $\delta = k^{-1/3} (\log n)^{2/3}$, Capped UCB has regret $= O((k \log n)^{2/3})$
 for any distribution (regular)

- For any distrib (regular), \exists SF, CF s.t Capped UCB with $\delta = \sqrt{k \log n}$ achieves regret $O(CF \sqrt{k \log n})$ when $\frac{k}{n} \leq SF$. For MMR

Notes

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- The choice of index I_t depends on (n, k) , not instantaneous state
- Benchmark is $\mathcal{V}(P)$, not $E[\text{Rev}(A_n^k(P))]$ - the previous lemma shows this is justified
- Need a refined UCB to handle 'rare' prices
- Index should incorporate 'stock-out price' $S^{-1}(\frac{k}{n})$ in addition to P_{MP}
- More details for Capped UCB (n, k)

$$I_t(P) \triangleq P \cdot \min \left[k, \underbrace{n S_t^{UB}(P)}_{\text{UCB}} \right]$$

$$S_t^{UB}(P) \triangleq \underbrace{\hat{S}_t(P)}_{\text{empirical rate}} + \underbrace{\mathcal{R}_t(P)}_{\text{confidence radius}}$$

$$\hat{S}_t(P) = \min \left\{ \frac{k_t(P)}{N_t(P)}, 1 \right\}, \quad \begin{array}{l} k_t(P) = \# \text{ of sales at price } P \\ N_t(P) = \# \text{ of agents offered price } P \\ \text{when } N_t(P) = 0 \end{array}$$

$$\mathcal{R}_t(P) = \frac{\alpha \log n}{N_t(P) + 1} + \sqrt{\frac{c \log n \hat{S}_t(P)}{N_t(P) + 1}}$$

Henceforth, we define $\alpha = c \log n$

Pf of $O((k \log n)^{2/3})$ (worst-case) regret

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• $X_t = \mathbb{1}_{\{\text{Sale to } t^{\text{th}} \text{ agent}\}} \sim \text{Bin}(S(p_t))$

$X = \sum_{t=1}^n X_t$, $S \triangleq E[X] = \sum_{t=1}^n \hat{S}(p_t)$

Suppose we ignore the capacity k to define X (i.e., we continue to sell after running out of items)

Then $\text{Rev} = \sum_{t=1}^N p_t X_t$, where $N = \max\{N \leq n \mid \sum_{t=1}^N X_t \leq k\}$

• Pf outline - First we define a set of 'good events' and analyze a deterministic algo under these. Then we bound the probability of 'not good events' to show small loss in regret.

• **Lemma 2** With probability at least $1 - n^{-2}$ the following are true: for all $t \in \{1, 2, \dots, T\}$, $p \in \mathcal{P}$

i) $|S(p) - \hat{S}_t(p)| \leq \eta_t(p) \leq 3 \left(\frac{\alpha}{N_t(p)+1} + \sqrt{\frac{\alpha S_t(p)}{N_t(p)+1}} \right)$

And
ii) $|X - S| < O(\sqrt{S \log n} + \log n)$

iii) $|\sum_{t=1}^n p_t (X_t - S(p_t))| < O(\sqrt{S \log n} + \log n)$

Note - the last two bounds depend on S , not n

- Henceforth, assume i, ii and iii are TRUE

• Define $P_{act}^* \equiv \arg \max_{P \in \mathcal{P}} [v(P)]$ (best active price)

$$\Delta(P) \equiv \max \left(0, \frac{1}{n} v(P_{act}^*) - P \cdot \frac{1}{n} S(P) \right)$$

similar to $\mu^* - \mu_i$ in DCB

$$N(P) \equiv N_{tot}(P) = \# \text{ of agents offered } P$$

Lemma 3 - $\forall P \in \mathcal{P}, N(P) \Delta(P) \leq O(\log n) \left[1 + \frac{k \cdot 1}{n \Delta(P)} \right]$

Pf - by defⁿ, $\forall t, P \in \mathcal{P} - |S(P) - \hat{S}_t(P)| \leq \eta_t(P)$

$$\Rightarrow v(P) \leq I_t(P) \leq P \cdot \min \left[k, n(S(P) + 2\eta_t(P)) \right]$$

Thus - $\begin{cases} I_t(P_t) \geq v(P_{act}^*) \\ I_t(P_t) \leq P_t \cdot \min [k, n(S(P_t) + 2\eta_t(P_t))] \end{cases}$ (Choose highest $I_t(P)$!)

$$\Rightarrow \frac{v(P_{act}^*)}{n} \leq P_t \min \left[\frac{k}{n}, S(P_t) + 2\eta_t(P_t) \right]$$

$$\Rightarrow \text{i) } P_t \geq \frac{v_{act}^*}{k}, \text{ ii) } \Delta(P_t) \geq 2P_t \eta_t(P_t), \text{ iii) } \Delta(P_t) > 0 \Rightarrow S(P_t) < \frac{k}{n}$$

Now we use the form of $\eta_t(\cdot)$

Consider, the last time price p was chosen (for any $p \in P$) ⑥

$$\Rightarrow \Delta(p) - N(p) = N_t(p) + 1 \quad] \text{ by defn}$$

$$\begin{aligned} & - \Delta(p) \leq 2p \pi_t(p) \\ & - \Delta(p) > 0 \Rightarrow S_t(p) < \frac{k}{n} \end{aligned} \quad] \text{ from above}$$

$$- \pi_t(p) \leq 3 \left(\frac{O(\log n)}{N(p)} + \sqrt{\frac{S_t(p)}{N(p)}} \right)$$

Combining, we get $\Delta(p) \leq O(p) \cdot \max \left[\frac{\log n}{N(p)}, \sqrt{\frac{k \log n}{n N(p)}} \right]$

• Next, instead of Rev , we want to analyze $\sum_{t=1}^n P_t S(p_t)$
 (i.e., ignore capacity constraints)

Lemma - Let $\beta(s) = O(\sqrt{s \log n} + \log n)$

then $Rev \geq \min \left(D(p_{act}^*), \sum_{t=1}^n P_t S(p_t) \right) - \beta(k)$

Pf - $\because P_t \geq \frac{D(p_{act}^*)}{k} \forall t \Rightarrow Rev \geq D(p_{act}^*)$ if $\sum_{t=1}^n x_t > k$

• If $\sum_{t=1}^n x_t < k$, then $Rev = \sum_{t=1}^n P_t x_t \geq \sum_{t=1}^n P_t S(p_t) - \beta(s)$

from the assumed high prob events

• Now let's consider $\sum_{P \in \mathcal{P}} S(P)$

$$\begin{aligned} \sum_{t=1}^n P_t S(P_t) &\geq \sum_{t=1}^n \left[\frac{v(P_{\text{act}}^*)}{n} - \Delta(P_t) \right] \\ &= v(P_{\text{act}}^*) - \sum_{P \in \mathcal{P}} \Delta(P) N(P) \end{aligned}$$

Define $\mathcal{P}_{\text{sel}} = \{P \in \mathcal{P} \mid N(P) \geq 1\}$ (prices selected at least once)

$\mathcal{P}_\epsilon = \{P \in \mathcal{P}_{\text{sel}} \mid \Delta(P) > \epsilon\}$ (will choose ϵ later)

$$\begin{aligned} \Rightarrow \sum_{P \in \mathcal{P}} N(P) \Delta(P) &= \sum_{P \in \mathcal{P}(\epsilon)} \Delta(P) N(P) + \sum_{P \notin \mathcal{P}(\epsilon)} \Delta(P) N(P) \\ &\leq O(\log n) \left[|\mathcal{P}_{\text{sel}}| + \frac{k}{n} \sum_{P \in \mathcal{P}} \frac{1}{\Delta(P)} \right] + \epsilon n \end{aligned}$$

Thus, for any set \mathcal{P} , any $\epsilon > 0$

$$v(P_{\text{act}}^*) - \mathbb{E}[\text{Rev}] \leq \epsilon n + O(\log n) \left[|\mathcal{P}_{\text{sel}}| + \frac{k}{n} \sum_{P \in \mathcal{P}} \frac{1}{\Delta(P)} \right] + \beta(k)$$

• For $\mathcal{P} = \{\delta(1+\delta)^i\}$, $v(P_{\text{act}}^*) - v(P^*) \geq -\delta k$

- If $P^* < \delta$, $v(P^*) \leq \delta k$. Else let $P_0 = \max\{P \in \mathcal{P}, P \leq P^*\}$

$$\Rightarrow \frac{P_0}{P^*} \geq \frac{1}{1+\delta} \geq 1-\delta \Rightarrow v(P_{\text{act}}^*) \geq v(P_0) \geq \frac{P_0}{P^*} v(P^*) \geq v(P^*) - \delta k$$

Putting everything together, we get

$$\text{Regret} = \sum_{t=1}^n \ell_t(\mathbf{P}_{e_t})$$

$$\leq \epsilon n + O(\log n) \left[|\mathbf{P}_{e_t}| + \frac{k}{n} \sum_{\mathbf{P} \in \mathcal{P}_e} \Delta(\mathbf{P})^{-1} \right] + \beta(k) + \delta k$$

$$\leq O(\log n) \left[|\mathbf{P}_{e_t}| \left(1 + \frac{k}{\epsilon n} \right) \right] + O(\sqrt{k \log n} + \log n) + \delta k + \epsilon n$$

Also $|\mathbf{P}_{e_t}| \leq \frac{1}{\delta} \log n$. assume $\delta \geq \frac{1}{n}$, $\epsilon = \delta \frac{k}{n}$

$$\Rightarrow \text{Regret} \leq O\left(\delta k + \frac{1}{\delta^2} (\log n)^2 + \sqrt{k \log n}\right)$$

Choosing $\delta = k^{-1/3} (\log n)^{2/3} \Rightarrow \text{Regret} = O((k \log n)^{2/3})$

Concentration Bounds in Babaioff et al.

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* Standard Chernoff bounds - For $X_1, X_2, \dots, X_n \in [0, 1]$,
i.i.d, $X = \frac{1}{n} \sum_{i=1}^n X_i$, $\mu = E[X]$. Then

$$i) \mathbb{P}[|X - \mu| \geq \epsilon \mu] \leq 2e^{-n\mu \epsilon^2/3} \quad \forall 0 < \epsilon < 1$$

$$ii) \mathbb{P}[X > a] < 2^{-an} \quad \text{for } a > 6\mu$$

* Lemma - In the above setting, let $\eta(\alpha, x) = \frac{\alpha}{n} + \sqrt{\frac{\alpha x}{n}}$.

$$\text{Then } \mathbb{P}[|X - \mu| < \eta(\alpha, X) < 3\eta(\alpha, \mu)] > 1 - e^{-\Omega(\alpha)}$$

Pf - The main idea is to separately deal with small and large μ .

i) Consider $\mu > \alpha/6n$. Let $\epsilon = \frac{1}{2} \sqrt{\frac{\alpha}{6\mu n}}$. Now by (i)

$$\mathbb{P}[|X - \mu| \geq \epsilon n] \leq 2 \exp\left(\frac{-n\mu \epsilon^2}{72\mu n}\right) = 2e^{-\alpha}$$

$$\Rightarrow |X - \mu| < \mu \epsilon < \mu/2 \quad \text{w.p. } 1 - e^{-\Omega(\alpha)}$$

Also by choice of ϵ , we have w.p. $1 - e^{-\Omega(\alpha)}$

$$|X - \mu| < \frac{\mu}{2} \sqrt{\frac{\alpha}{6\mu n}} \leq \sqrt{\frac{\alpha X}{n}} \leq \eta(\alpha, X) \leq 1.5\eta(\alpha, \mu)$$

ii) Consider $\mu < \alpha/n$. Let $a = \alpha/n$; by (ii) (10)

$$X < \alpha/n \quad \text{w.p.} \geq 1 - e^{-\Omega(\alpha)}$$

$$\Rightarrow |X - \mu| < \alpha/n < \eta(\alpha, X)$$

$$\text{Finally } \eta(\alpha, X) = \frac{\alpha}{n} + \sqrt{\frac{\alpha X}{n}} < (1 + \sqrt{2}) \frac{\alpha}{n} < 3 \eta(\alpha, \mu)$$

Now we return to our 'bad event' bound

$$i) |S(p) - \hat{S}_t(p)| \leq \eta_t(p) \leq 3 \left(\frac{\alpha}{N_t(p)+1} + \sqrt{\frac{\alpha S_t(p)}{N_t(p)+1}} \right)$$

Pf- For any $p \in \mathcal{P}$, let $\{Z_{i,p}\}_{i \leq n} \equiv \mathbb{1}_{\text{Bad}}(S(p))$ over

\Rightarrow Sale to $\equiv \{Z_{i,p} = 1\}$ Now we can use our lemma
 i^{th} agent who sees p

$$\Rightarrow \mathbb{P} \left[|S(p) - \hat{S}_t(p)| \leq \eta_t(p) \leq 3 \left(\frac{\alpha}{N_t(p)+1} + \sqrt{\frac{\alpha S_t(p)}{N_t(p)+1}} \right) \right] \geq 1 - n^{-4}$$

Also $|\mathcal{P}| \leq n$ (if $d = c \log n$)

\Rightarrow By union bound over $t \in \{1, \dots, n\}$, $p \in \mathcal{P}$, we get the result.