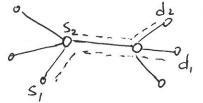
- · Capacity control across multiple nesources linked by demands for sets of resources
- · Eg Hub and spoke networks

 o de (ainlines)

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Linear networks (hotels, rental demond 2 0-0-0-0domand 1 demand 3

· No natural ordering of fare-classes => Use time for DP formulation

Problem setting

- products (or ODFs = Drigin Dest Fare) - M nesources, n
- $-C=(C_1,C_2,\ldots,C_m)\in\mathbb{Z}^m$
- t = time to go E[T, 0]
- demand for product i ~ PP(J:(t)) [P=(P. P. P. P. P.)
- $A = \begin{bmatrix} 1 & 1 & 1 & 0 \\ 0 & 0 & 1 & 1 \end{bmatrix}$
- nequitements a: E {0,13", price P; - Product i =
- State (t, 2)
- Bellman Eqn. V(t,0)= V(0,≥c)=0

$$\frac{\partial V(t,x)}{\partial t} = \sum_{i=1}^{n} \lambda_i(t) \max_{u_i \in \{0,1\}} \left[P_i u_i - \Delta_{i} V(t,x) \right]^{+}$$

$$= V(t,x) - V(t,x-4)$$

$$\leq t$$

Let
$$R(\underline{\theta}) = \max_{\underline{u} \in \{0,1\}^n} \sum_{i=1}^n \lambda_i(t) \left[P_i u_i - \theta_i\right]^{+}$$

When $\underline{u}^* = \left\{u_i = 1 \text{ if } P_i \geqslant \theta_i\right\} \Rightarrow R(\underline{\theta}) = \sum_{i=1}^n \lambda_i(t) \left[P_i - \theta_i\right]^{+}$

And $\underline{u}^*(t,\underline{x}) = \sum_{i=1}^n \lambda_i(t) \left[R(\Delta V(t,\underline{x}))\right]$

and $\underline{u}^*(t,\underline{x}) = \left\{u_i = 1 \text{ iff } \underline{x} - \underline{a}_i \geqslant 0 \text{ and } P_i \geqslant \Delta_i V(t,\underline{x})\right\}$

Difficulty

i) $V(t,\underline{x})$ may be difficult - $|\theta|$ compute

ii) $\underline{u}^*(t,\underline{x})$ needs too much stooge / is complicated the want simples policies with good properties

We want simples policies with good properties bid-Price controls/Probabilistic admission conhol

Suppose
$$\triangle_{k}V(t,x) = V(t,x)-V(t,x-a)$$
 $\approx \frac{\partial V(t,x)}{\partial x_{i}}$
 $\Rightarrow \frac{\partial V(t,x)}{\partial x_{i}}$
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And $\Rightarrow \frac{\partial V(t,x)}{\partial x_{i}}$

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Now condition for admission $\Rightarrow \frac{\partial V(t,x)}{\partial x_{i}}$

(and $\Rightarrow \frac{\partial V(t,x)}{\partial x_{i}}$

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· First, we rescale time to get a discrete time equivalent to the Bellman eqn - Given a>1, we set T→aT, \(\begin{array}{c} \gamma_i(t) \rightarrow \frac{\gamma_i(at)}{a} \end{array} and consider disorde times {Ta, aT-1, ..., 0} (henceforth, we use T and 2;(t) to refer to scaled values) -a is chosen s.t $2i(at)/a \ll 1 \forall t$ Now we have $V(t,x)-V(t,x)=\mathbb{R}\left(\Delta V(t-1,x)\right)$ (where again $R(\theta) = \sum_{i=1}^{n} \lambda_i(t) \left[P_i - \theta_i\right] + \frac{3}{2} V(t,0) = V(\theta_i x)$)
and opt control $u_i^*(t,x) = \frac{1}{2} \left[a_i \le x \text{ AND } P_i \geqslant \Delta_i V(t,x)\right]$ - To implement bid-prices, we need linear approx of V(t,z)

(Fluid) upper bound on V(t,x)

- Oracle based bound = Suppose we know sealization of domad - Let $D_i \sim P_{0i}$ ($\int \lambda_i(s) ds$) be total arrivals to class j(i.e., total demand for product j). Define $\Lambda_j = \int \lambda_i(s) ds$ $V^{U}(T,C|D) \equiv \max \sum_{i=1}^{n} P_i y_i$

Claim - V (T, C/D) is concave in D Pf - If you and you solves to V'(T, SID) and V(T, SID) then dy + (1-d) y' is feasible for aD + D(1-d) $) \qquad \bigvee \left(T, \subseteq |D \times + D'(1-\alpha) \right) > X \bigvee \left(T, \subseteq |D \right) + \left(1-\alpha \right) \bigvee \left(T, \subseteq |D \right)$ · Thus, if we seplace D; with E[D;] = A; in (*), we get V fluid (T, C), which by Jensen's satisfies $V^{\text{fluid}}(T, C) > E[V'(T, C | D)]$ · Primal-Dual forms for fluid Problem V'(T,E)

Primal max $\sum_{i=1}^{n} P_{i} y_{i}$ S.t $\sum_{i=1}^{n} a_{i}(j) y_{i} \leq C_{j} \forall j \in [M]$ $0 \leq y_{i} \leq \Lambda_{i} \quad \forall i \in [N]$ $0 \leq y_{i} \leq \Lambda_{i} \quad \forall i \in [N]$ Dual min $\sum_{i=1}^{n} \Lambda_{i} \beta_{i} + \sum_{j=1}^{m} G_{j} Z_{j}$ S.t $\sum_{i=1}^{m} a_{i}(j) Z_{j} + \beta_{i} \gg P_{i} \quad \forall i \in [N]$

Z;>0, B;>0

$$\beta_i^* > 0 \Rightarrow 90$$
 $y_i = \Lambda_i$

$$\mathcal{B}_{i}^{*} = \left(P_{i} - \sum_{j=1}^{n} a_{i}(j) z_{j}^{*}\right)^{+} \quad \forall i$$

$$=) \sum_{i=1}^{n} \Lambda_{i} \beta_{i}^{*} = \sum_{i=1}^{n} \Lambda_{i} \left(\beta_{i} - \sum_{i=1}^{n} a_{i}(i) z_{i}^{*} \right)^{+} = \mathcal{R} \left(A^{T} z^{*} \right)$$

$$=) \qquad \forall \text{fluit}(T,c) = \min_{z > 0} \left\{ \mathcal{R}(A^T z) + c^T z \right\}$$

(Equivalently - we are approximating
$$\Delta_i V(T,c) \approx \sum_{i=1}^m a_i(i)z_i^*$$
)

· Now given yit formal gropt Froms) or zit (disal opt vars) we have 2 simple hemistics

we have
$$Z$$
 simple realisines
i) Bid price = Admit i if $\sum_{j=1}^{m} Q_i(j) Z_j^* \leq P_i$ ad $z_i - q_i \geqslant 0$

. The notion of a bid price is more general. Given any linear approx" of AV (t, z) & Su; a:(j), u; are bid prices (Similarly for PAC policies)

Agre bid-prices optimal? No $P_1 = P_2 = 250$ Eq. $P_3 \longrightarrow P_3$ $P_4 = P_2 = 250$ $P_5 \longrightarrow P_2 \longrightarrow P_3$ $P_7 = P_2 = 250$ Suppose 2(2) = (0.3, 0.3, 0.4) $2(1) = (0, 0, 0.8) \left(\begin{array}{c} \text{and no} \\ \text{anival wp } 0.2 \end{array}\right)$ Claim - OPT = Accept only & customen for product 3 =) $R^{+} = (0.4 + 0.6 \times 0.8) \times 450 = 396$ However, to implement this via bid prior Mi, we need $M_1 + M_2 \le 450$, $M_1 > 250$, $M_2 \ge 250 = 200$ · Can we show bid-prices perform we'll? Yes!

Can we show bid-Prices perform we'll? Yes!

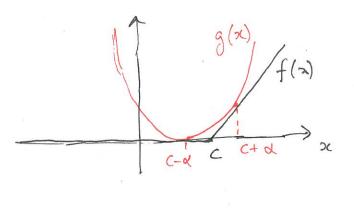
Idea - (onsider the problem under a 'knge manket' scaling $C \rightarrow \Theta C$, $\lambda_i(t) \rightarrow \Theta \lambda_i(t) \forall i, t$ for some $\Theta > 1$ - Note $\{Z_j^*\}$ the mains the same $\left(\bigvee_{j \in I} f(x_j) = \max_{j \in I} \Theta \left(\sum_{i=1}^{n} \lambda_i \beta_i + \sum_{j=1}^{n} C_i Z_i \right) \right) + \sum_{j \in I} \sum_{j \in I} C_j Z_j \beta_j = 0$ - Let $\bigvee_{j \in I} V_j = \sum_{j \in I} V_j = \sum_{j$

We first need an additional Lemma

Lamma - For any
$$90 \times \text{with } \mathbb{E}[x] = \mu$$
, $Var(x) = \sigma^2$, $\forall c$

$$\mathbb{E}\left[(x-c)^+\right] \leq \frac{1}{2}\left(\sqrt{\sigma^2+(c-\mu)^2} - (c-\mu)\right)$$

Pf Consider $f(x) = (X-c)^+$ Moreover, for any d>0, define $g(x) = (x - (c-x))^2$



 $g(x) \gg f(x) \forall x$ Then (from figure) we have

$$=) \mathbb{E}\left[f(x)\right] \leq \mathbb{E}\left[g(x)\right] = \mathbb{I}\mathbb{E}\left[x^2 - 2x(c-d) + (c-d)^2\right]$$

$$=) E[(x-c)^{\dagger}] < min \left(\frac{\mu^2 + \sigma^2 - 2\mu(c-d) + (c-d)^2}{4d}\right)$$

Setting $d = \sqrt{\sigma^2 + (c-\mu)^2}$, we get

$$\mathbb{E}\left[\left(\mathbf{x}-\mathbf{c}\right)^{+}\right]\leq\sqrt{\sigma^{2}+\left(\mathbf{c}-\mathbf{\mu}\right)^{2}-\left(\mathbf{c}-\mathbf{\mu}\right)}$$

[[(x-c)+] ≤ 0.5 + 0.5 (|c-μ|-(c-μ)) = = 0.50 if cm

The (Talluni & Van Ryzin '98) - Let Bo be the total expected revenue under the bid price heuristic using bid price $\{Z_i^*\}$ from the fluid LP. Then

Bo > 1- O(1)

VO(T,C)

(Strictly speaking - this requires a small modification to the policy-see below)

(Strictly speaking - this requires a small modification to the policy - see below)

Pf - We consider a small modification of the basic bid-price

heuristic, as follows (based on Reiman & Wang '07)

Recall for V Pofluid (T,c), the primal solve is { yi } d }, and

dual solve is { Zi } jelm . Now consider the following policy

- If Pi > \(\sum_{j=1}^{m} \alpha_{i}(j) \) \(z_{j}^{*} \) (and \(z_{-}a_{i} \ge o_{i} \)): Admit i

- If P: = \(\sum_{j=1}^{m} \alpha_{i}(j) \) \(z_{j}^{*} \) (and \(z_{-}a_{i} \ge o_{i} \)): Admit i wp \(\frac{3y}{\Lambda_{i}} \)

- Else reject i

Now we show that under this policy, revenue \overline{B}^{θ} satisfies $\overline{B}^{\theta} \gg \left(1 - O\left(\frac{1}{\sqrt{\theta}}\right)\right) V^{\theta}(\overline{\tau}, c)$

Now we have the following.

$$\frac{\bar{B}^{\theta}}{V^{\theta}(\bar{t},c)} > \frac{\bar{B}^{\theta}}{\bar{\theta}V^{\theta}(\bar{t},c)}$$
(Note: by Jensen's $\bar{B}^{\theta} \leq \bar{\theta}V^{\theta}(\bar{t},c)$)

2) We can write
$$y_i^* = \Lambda_i \cdot \left(\frac{y_i^*}{\Lambda_i}\right) \forall i$$

and also
$$\sum_{i \in [n]} \Lambda_i \left(\frac{y_i^*}{\Lambda_i} \right)$$
. $q_i(j) \leq C_j \forall j$ (by feasibility)

3) Now consider an alternate admission policy, where we admit all armiving customers in class i w.p. 2 yi,

Ignoring capacity constraints, however, we incur a cost of Pmax = max {Pi} for each additional unit of capacity used on any leg

Let
$$L^{\partial} = \lambda \text{ evenue under this new policy}$$

$$\Rightarrow i) L^{\partial} = \sum_{i \in [n]} \partial \Lambda_i \left(\frac{y_i}{\Lambda_i} \right) P_i - \sum_{j=1}^{m} P_{\text{max}} \left(\sum_{i \in [n]} \Delta_i \frac{y_i}{\Lambda_i} a_i | j \right) - \partial a_j$$

$$\Rightarrow i) L^{\partial} = \sum_{i \in [n]} \partial \Lambda_i \left(\frac{y_i}{\Lambda_i} \right) P_i - \sum_{j=1}^{m} P_{\text{max}} \left(\sum_{i \in [n]} \Delta_i \frac{y_i}{\Lambda_i} a_i | j \right) - \partial a_j$$

$$\Rightarrow i) L^{\partial} = \sum_{i \in [n]} \Delta_i \left(\frac{y_i}{\Lambda_i} \right) P_i - \sum_{i \in [n]} P_{\text{max}} \left(\sum_{i \in [n]} \Delta_i \frac{y_i}{\Lambda_i} \right) P_i - \sum_{i \in [n]} P_i \left(\sum_{i \in [n]} \Delta_i \frac{y_i}{\Lambda_i} \right) P_i - \sum_{i \in [n]} P_i \left(\sum_{i \in [n]} \Delta_i \frac{y_i}{\Lambda_i} \right) P_i - \sum_{i \in [n]} P_i \left(\sum_{i \in [n]} \Delta_i \frac{y_i}{\Lambda_i} \right) P_i - \sum_{i \in [n]} P_i \left(\sum_{i \in [n]} \Delta_i \frac{y_i}{\Lambda_i} \right) P_i - \sum_{i \in [n]} P_i \left(\sum_{i \in [n]} \Delta_i \frac{y_i}{\Lambda_i} \right) P_i - \sum_{i \in [n]} P_i \left(\sum_{i \in [n]} \Delta_i \frac{y_i}{\Lambda_i} \right) P_i - \sum_{i \in [n]} P_i \left(\sum_{i \in [n]} \Delta_i \frac{y_i}{\Lambda_i} \right) P_i - \sum_{i \in [n]} P_i \left(\sum_{i \in [n]} \Delta_i \frac{y_i}{\Lambda_i} \right) P_i - \sum_{i \in [n]} P_i \left(\sum_{i \in [n]} \Delta_i \frac{y_i}{\Lambda_i} \right) P_i - \sum_{i \in [n]} P_i \left(\sum_{i \in [n]} \Delta_i \frac{y_i}{\Lambda_i} \right) P_i - \sum_{i \in [n]} P_i \left(\sum_{i \in [n]} \Delta_i \frac{y_i}{\Lambda_i} \right) P_i - \sum_{i \in [n]} P_i \left(\sum_{i \in [n]} \Delta_i \frac{y_i}{\Lambda_i} \right) P_i - \sum_{i \in [n]} P_i \left(\sum_{i \in [n]} \Delta_i \frac{y_i}{\Lambda_i} \right) P_i - \sum_{i \in [n]} P_i \left(\sum_{i \in [n]} \Delta_i \frac{y_i}{\Lambda_i} \right) P_i - \sum_{i \in [n]} P_i \left(\sum_{i \in [n]} \Delta_i \frac{y_i}{\Lambda_i} \right) P_i - \sum_{i \in [n]} P_i \left(\sum_{i \in [n]} \Delta_i \frac{y_i}{\Lambda_i} \right) P_i - \sum_{i \in [n]} P_i \left(\sum_{i \in [n]} \Delta_i \frac{y_i}{\Lambda_i} \right) P_i - \sum_{i \in [n]} P_i \left(\sum_{i \in [n]} \Delta_i \frac{y_i}{\Lambda_i} \right) P_i - \sum_{i \in [n]} P_i \left(\sum_{i \in [n]} \Delta_i \frac{y_i}{\Lambda_i} \right) P_i - \sum_{i \in [n]} P_i \left(\sum_{i \in [n]} \Delta_i \frac{y_i}{\Lambda_i} \right) P_i - \sum_{i \in [n]} P_i \left(\sum_{i \in [n]} \Delta_i \frac{y_i}{\Lambda_i} \right) P_i - \sum_{i \in [n]} P_i \left(\sum_{i \in [n]} \Delta_i \frac{y_i}{\Lambda_i} \right) P_i - \sum_{i \in [n]} P_i \left(\sum_{i \in [n]} \Delta_i \frac{y_i}{\Lambda_i} \right) P_i - \sum_{i \in [n]} P_i \left(\sum_{i \in [n]} \Delta_i \frac{y_i}{\Lambda_i} \right) P_i - \sum_{i \in [n]} P_i \left(\sum_{i \in [n]} \Delta_i \frac{y_i}{\Lambda_i} \right) P_i - \sum_{i \in [n]} P_i \left(\sum_{i \in [n]} \Delta_i \frac{y_i}{\Lambda_i} \right) P_i - \sum_{i \in [n]} P_i \left(\sum_{i \in [n]} \Delta_i \frac{y_i}{\Lambda_i} \right) P_i - \sum_{i \in [n]} P_i \left(\sum_{i \in [n]} \Delta_i \frac{y_i}{\Lambda_i} \right) P_i - \sum_{i \in [n]} P_i \left(\sum_{i \in [n]} \Delta_i \frac{y_i}{\Lambda_i} \right$$

(II

$$-\theta C_{j} > \theta \sum_{i \in (n)} y_{i}^{*} q_{i}(i)$$

=) We can use
$$\mathbb{E}\left[(x-c)^{\dagger}\right] \leq 0.5 \, \sigma$$

$$-\sqrt{Var\left(\sum_{i\in [n]}\widehat{Q_i}\left(y_i^*\right)\alpha_i(i)\right)} \leq \sqrt{\theta\left(\sum_{i\in [n]}\widehat{Q_i}(i)y_i^*\right)}$$

$$\frac{1}{\sqrt{\theta(\tau,c)}} > 1 - \frac{P_{\text{max}} \sqrt{\theta} \sqrt{\sum_{i \in [n]} S_{i}(i)} y_{i}^{*}}{2 \left(\sum_{i \in [n]} y_{i}^{*} P_{i}\right) \theta}$$

$$= 1 - \Theta \left(\frac{1}{\sqrt{\theta}}\right)$$