

# ORIE 4154 - Pricing and Market Design

## Module 2: Network RM and Approximate DP (The Network RM Dynamic Program)

Instructor: Sid Banerjee, ORIE



Cornell University

# From Last Class: The Network RM Problem

Pricing/admission control for **products** using **interlinked resources**.

Connecting flights, multi-day hotel bookings, project teams, etc.

- **Resources**
  - Perishable units of capacity managed by firm  
E.g. Seats on a flight, hotel room nights, employee hours
  - Constrained ( $C_i$  units of resource  $i$ )
  - Perishable (each resource expires at a certain time)
- **Product**
  - Bundle of resources for selling to customer  
E.g. multi-leg flight, multiple days stay at hotel
  - Each product has a specified set of resources and price

## What we need

- **Compact representation** for DP formulation

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- **Compact representation** for DP formulation
- **Good approximations** for solving the DP

# Formal Model for Network RM

## Basic Setting

- Time periods  $\{1, 2, \dots, T\}$
- $m$  resources  $\{1, 2, \dots, m\}$
- Resource  $i$  has initial capacity  $c_i$ ; expires at  $T$

## Products

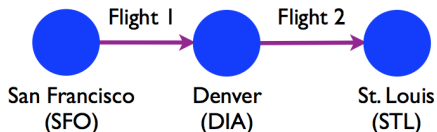
- $n$  unique products  $\{0, 1, \dots, n\}$
- Product  $j \equiv$  (price  $p_j$ , resource reqt  $A_j$ )
- **Incidence vector**  $A_j = \{a_{ij}\}$ , where  $a_{ij} = \mathbb{1}_{\{j \text{ uses resource } i\}}$

More generally,  $a_{ij} = \#$  of units of resource  $i$  used by product  $j$

**Incidence Matrix:**  $A = [A_1, A_2, \dots, A_n]$  ( $m \times n$  matrix)

# Incidence Matrix: Example 1

- 2 resources
  - Flight 1 from SFO to DIA
  - Flight 2 from DIA to STL
- Suppose we have 3 products and two fare classes: 6 ODFs
  - SFO to DIA full fare
  - SFO to DIA discount fare
  - DIA to STL full fare
  - DIA to STL discount fare
  - SFO to STL full fare
  - SFO to STL discount fare



## Incidence Matrix

$$A = \begin{bmatrix} 1 & 1 & 0 & 0 & 1 & 1 \\ 0 & 0 & 1 & 1 & 1 & 1 \end{bmatrix}$$

Courtesy: Paat Rusmevichentong

## Incidence Matrix: Example 2

- Suppose a hotel offers 1-night, 2-night, and 3-night stays only.
- Resources: Room nights
- Products: Combinations of arrival night and length of stay.
  - Assume one fare class for each product, that is, number of ODFs is the same as the number of products.
- Incidence matrix for 1 week of resources.

Resources/ Length of Stay	Arrival Date														
	Sunday			Monday			Tuesday			Wednesday			Thursday		
	1	2	3	1	2	3	1	2	3	1	2	3	1	2	3
Sunday	1	1	1	0	0	0	0	0	0	0	0	0	0	0	0
Monday	0	1	1	1	1	1	0	0	0	0	0	0	0	0	0
Tuesday	0	0	1	0	1	1	1	1	1	0	0	0	0	0	0
Wednesday	0	0	0	0	0	1	0	1	1	1	1	1	0	0	0
Thursday	0	0	0	0	0	0	0	0	1	0	1	1	1	1	1
Friday	0	0	0	0	0	0	0	0	0	0	0	1	0	1	1
Saturday	0	0	0	0	0	0	0	0	0	0	0	0	0	0	1

## Incidence Matrix: Example 2 (some caveats)

- In theory, the network management problem for a hotel stretches out indefinitely into the future.
- In practice, a hotel will only accept booking for some limited period into the future (often a year), limiting the number of resources and products that need to be managed.
- Moreover, in theory a hotel offers an infinite number of products, since customers can buy any length of stay.
- In practice, lengths of stay longer than 14 nights are extremely rare at most hotels, and hotels usually manage them as a single product.

Courtesy: Paat Rusmevichentong

# The Network RM Dynamic Program

- $m$  resources,  $n$  products, time periods  $\{1, 2, \dots, T\}$
- Resource  $i$  has initial capacity  $c_i$
- Product  $j \equiv (\text{price } p_j, \text{ resource reqt } A_j)$ , where  
 $a_{ij} = \mathbb{1}_{\{j \text{ uses resource } i\}}$

## Dynamics and actions

- In period  $t$ , state of system is  $\underline{\mathbf{x}} = [x_1, x_2, \dots, x_n]$
- *At most one* request arrives in each period:
  - No request arrives with probability  $\lambda_0$
  - Request for product  $j$  arrives with probability  $\lambda_j$
  - $\sum_{j=0}^n \lambda_j = 1$
- **Action/policy vector:**  $\underline{\mathbf{u}}(t, \underline{\mathbf{x}}) = [u_1(t, \underline{\mathbf{x}}), u_2(t, \underline{\mathbf{x}}), \dots, u_n(t, \underline{\mathbf{x}})]$ ,

$$u_j(t, \underline{\mathbf{x}}) = \mathbb{1}_{\{\text{Sell product } j \text{ in period } t \text{ with initial state } \underline{\mathbf{x}}\}}$$



# The Network RM Dynamic Program

- $m$  resources,  $n$  products, time periods  $\{1, 2, \dots, T\}$
- Resource  $i$  has initial capacity  $c_i$
- Product  $j \equiv (p_j, A_j)$ , where  $a_{ij} = \mathbb{1}_{\{j \text{ uses resource } i\}}$
- Request for product  $j$  arrives w.p.  $\lambda_j$ ; no request w.p.  $\lambda_0$
- $u_j(t, \underline{\mathbf{x}}) = \mathbb{1}_{\{\text{Sell product } j \text{ in period } t \text{ with initial state } \underline{\mathbf{x}}\}}$

For any  $t, \underline{\mathbf{x}}$ , **value function**  $V_t(\underline{\mathbf{x}})$  denotes the maximum expected revenue that can be obtained from period  $t$  until  $T$  given a remaining capacity  $\underline{\mathbf{x}}$  among the  $m$  resources

## The Bellman Equation

$$V_t(\underline{\mathbf{x}}) = \lambda_0 V_{t+1}(\underline{\mathbf{x}}) + \sum_{j=1}^n \lambda_j \max_{u \in \{0,1\}: u \cdot A_j \leq \underline{\mathbf{x}}} \{u \cdot p_j + V_{t+1}(\underline{\mathbf{x}} - u \cdot A_j)\}$$

where  $u \cdot A_j \leq \underline{\mathbf{x}}$  is equivalent to  $u \cdot a_{ij} \leq x_i \forall i$

# The Network RM Dynamic Program: Optimal Policy

## The Bellman Equation

$$V_t(\underline{\mathbf{x}}) = \lambda_0 V_{t+1}(\underline{\mathbf{x}}) + \sum_{j=1}^n \lambda_j \max_{u \in \{0,1\}: u \cdot A_j \leq \underline{\mathbf{x}}} \{u \cdot p_j + V_{t+1}(\underline{\mathbf{x}} - u \cdot A_j)\}$$

Suppose we are given  $V_{t+1}(\underline{\mathbf{x}})$  for all states  $\underline{\mathbf{x}}$ . Then we can solve for the optimal policy  $\underline{\mathbf{u}}^*(t, \underline{\mathbf{x}})$  to get for each product  $j$ :

$$u_j^*(t, \underline{\mathbf{x}}) = \begin{cases} 1 & \text{if } p_j > V_{t+1}(\underline{\mathbf{x}}) - V_{t+1}(\underline{\mathbf{x}} - A_j), \text{ AND } A_j \leq \underline{\mathbf{x}} \\ 0 & \text{otherwise} \end{cases}$$

Intuition: Sell a product only if its price exceeds the **opportunity cost** of losing future sales due to reduction in resource capacities.

# The Curse of Dimensionality

Thus, we have the optimal policy  $\underline{\mathbf{u}}^*(t, \underline{\mathbf{x}})$

$$u_j^*(t, \underline{\mathbf{x}}) = \begin{cases} 1 & \text{if } p_j > V_{t+1}(\underline{\mathbf{x}}) - V_{t+1}(x - A_j), \text{ AND } A_j \leq \underline{\mathbf{x}} \\ 0 & \text{otherwise} \end{cases}$$

Thus, given  $V_t(\underline{\mathbf{x}})$  for all  $t, \underline{\mathbf{x}}$ , we can find the optimal policy.

- However,  $\underline{\mathbf{x}}$  can take  $\prod_{i=1}^n (c_i + 1) \approx (c + 1)^m$  states!
- This is the **curse of dimensionality**: infeasible to store  $V_t(\underline{\mathbf{x}})$  exactly for moderately sized problems

To get around this, we want **good approximations of  $V_t(\underline{\mathbf{x}})$**