


# Demand Management

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- Principles of demand management
  - Airline yield management
  - Determining the booking limits
    - » A simple problem
    - » Stochastic gradients for general problems

# Principles of demand management

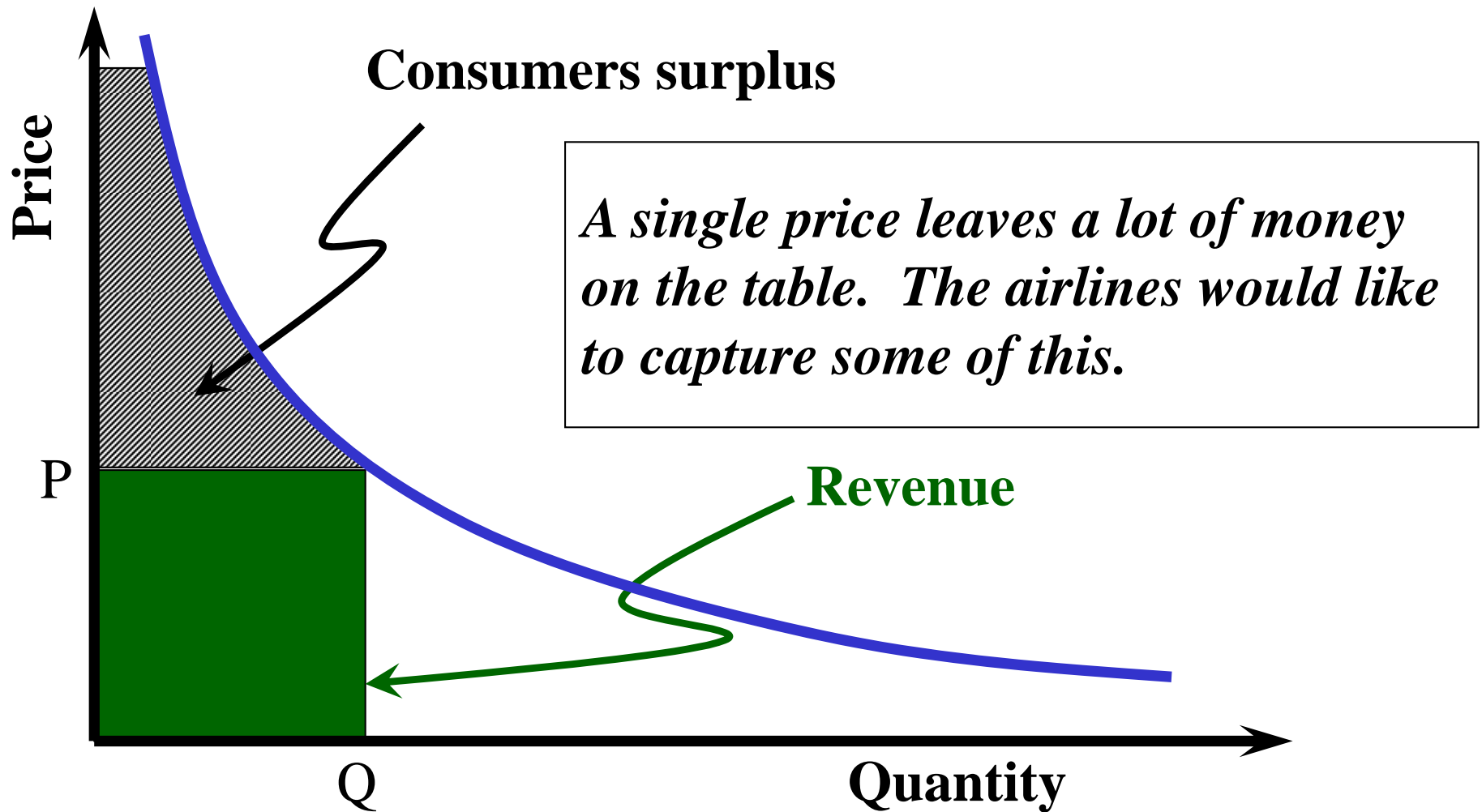
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## ■ Issues:

- » Customers are not all the same
- » The highest paying customers will not fill up available capacity.
- » The lowest paying customers will not cover the full costs of operations.
- » Marginal costs may be much lower than average costs.

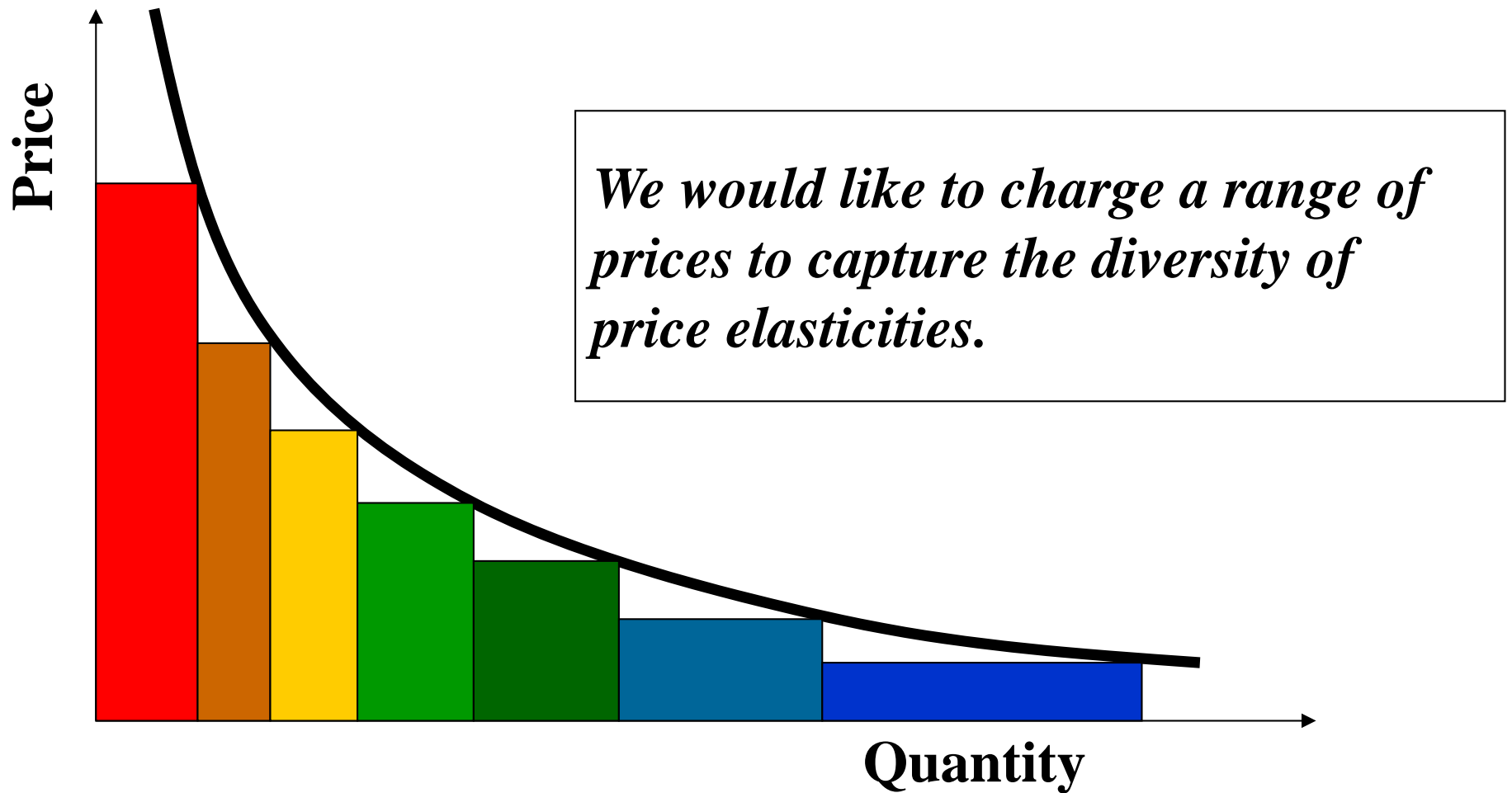
# Principles of demand management

## ■ From microeconomics:



# Principles of demand management

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# Principles of demand management

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## ■ Modes of control:

### » Primal

- Limit access to the system (now)
- Reservations (future)

### » Dual

- Pricing
- Service (waiting in line)

### » Informational

- Advertising
- Promotions

### » Combinations:

- Coupons - provide discounts (dual) with restricted services (primal) to customers who hold coupons (informational).

# Principles of demand management

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## ■ Market differentiating characteristics used in practice:


- » Advance notice (pre-booking, reservations)
- » Length of stay, stay over Saturday, etc.
- » Willingness to accept nonrefundable tickets
- » Service
  - Additional amenities (first class, business class)
  - Pre-board privileges (frequent flier status)

# Principles of demand management

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- Applications of demand management
  - » Airlines
  - » Hotels
  - » Universities (with rolling admissions)

# Lecture outline - Demand Management

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# Airline yield management

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## ■ Overview

### » Old system:

- Airlines establish a fare (or two, such as coach and first class)
- Sell seats at the fare until all seats are sold or the flight departs.
- Because of no-shows, airlines may sell too many seats to increase load factors.

### » Issue:

- Business travelers are willing to pay much more, but want the flexibility of making plans at the last minute.
- Different customers want different “products” measured in terms of price and service.

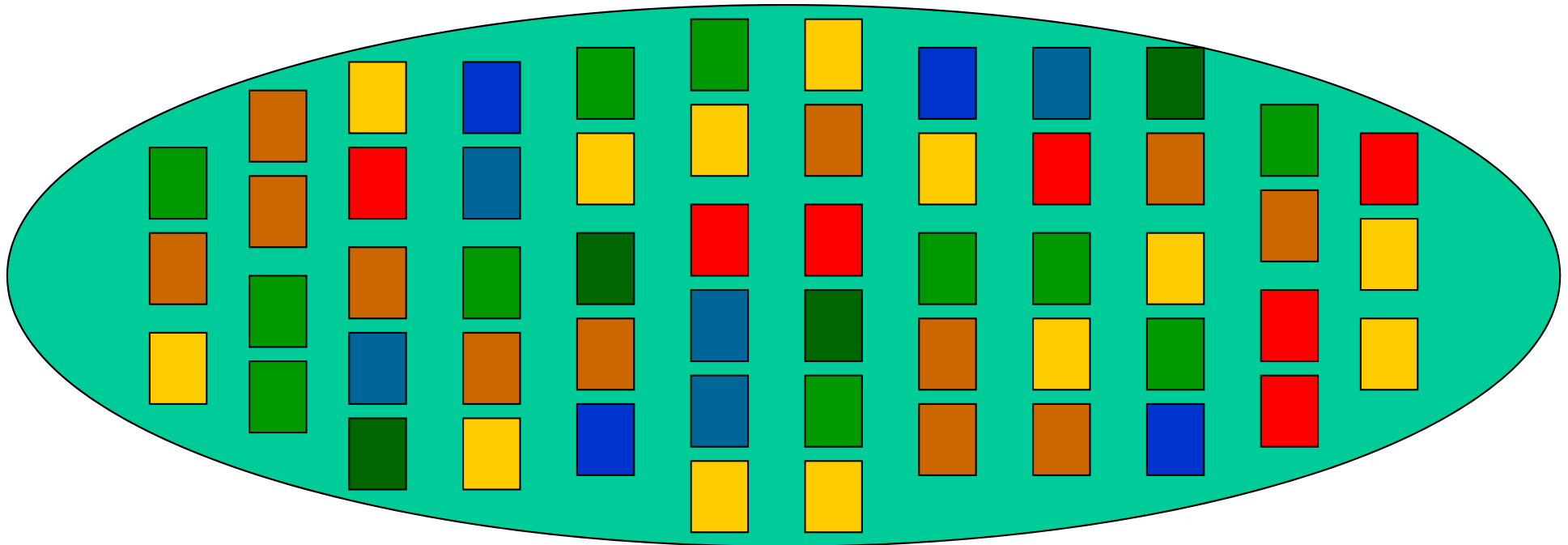
### » Solution:


- Offer blocks of seats at increasingly higher prices. The lower demand for higher priced seats has the effect of making seats available for last-minute business travelers.

# Airline yield management

## ■ Booking limits

» Blocks of seats are offered at different prices:

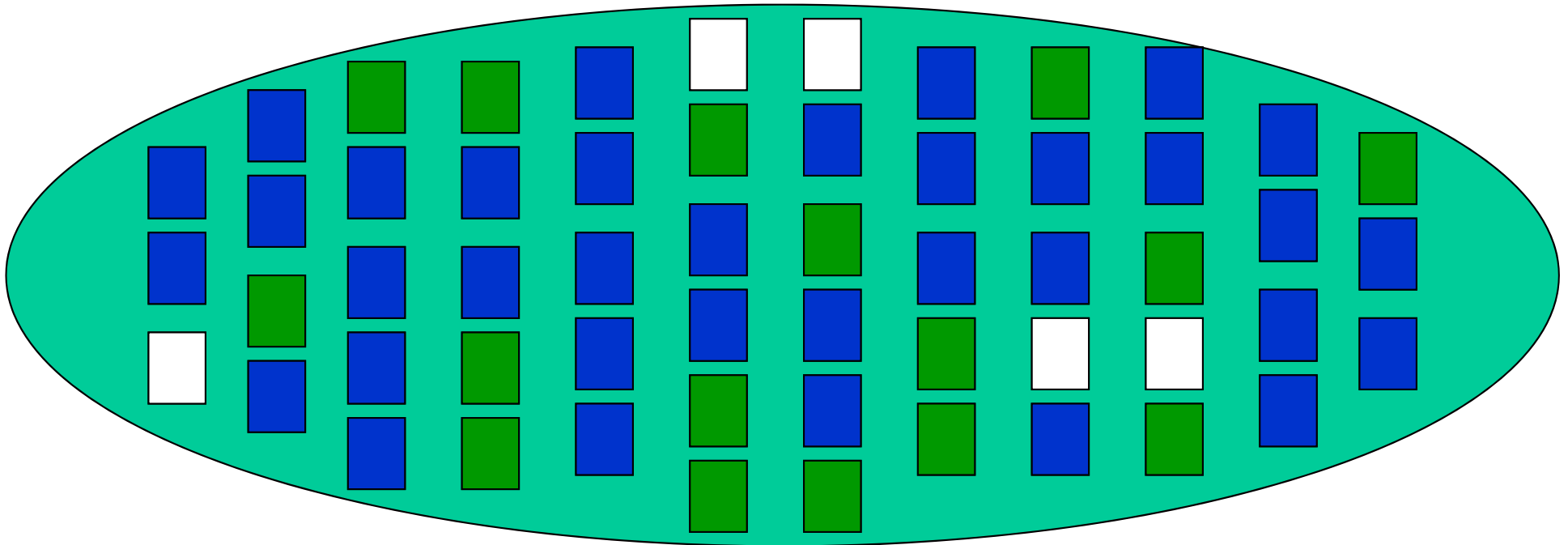


	<b>\$1700</b>		<b>\$910</b>		<b>\$525</b>
	<b>\$1425</b>		<b>\$850</b>		
	<b>\$1150</b>		<b>\$775</b>		

# Airline yield management

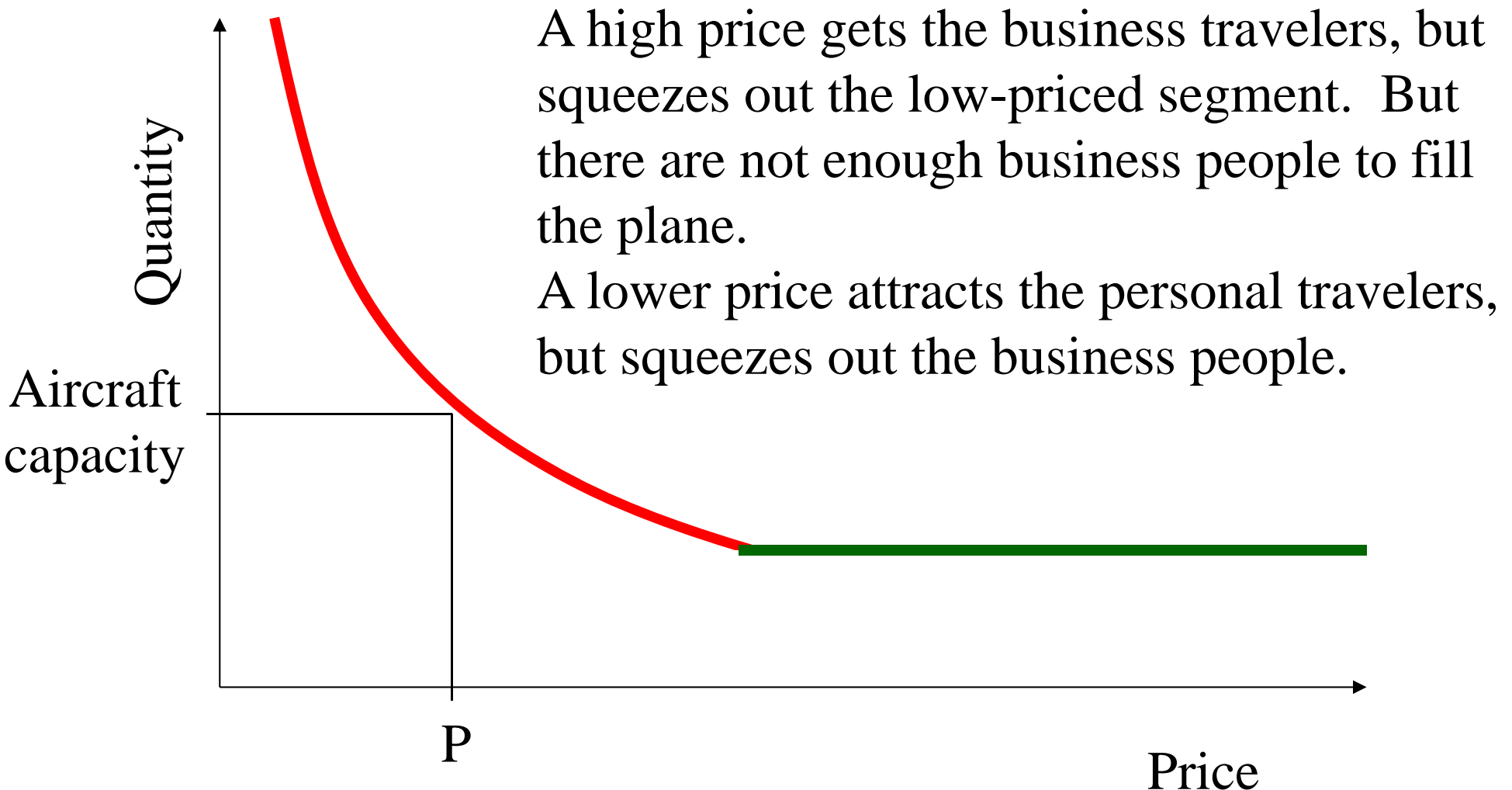
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- Consider a two-class problem:



# Airline yield management

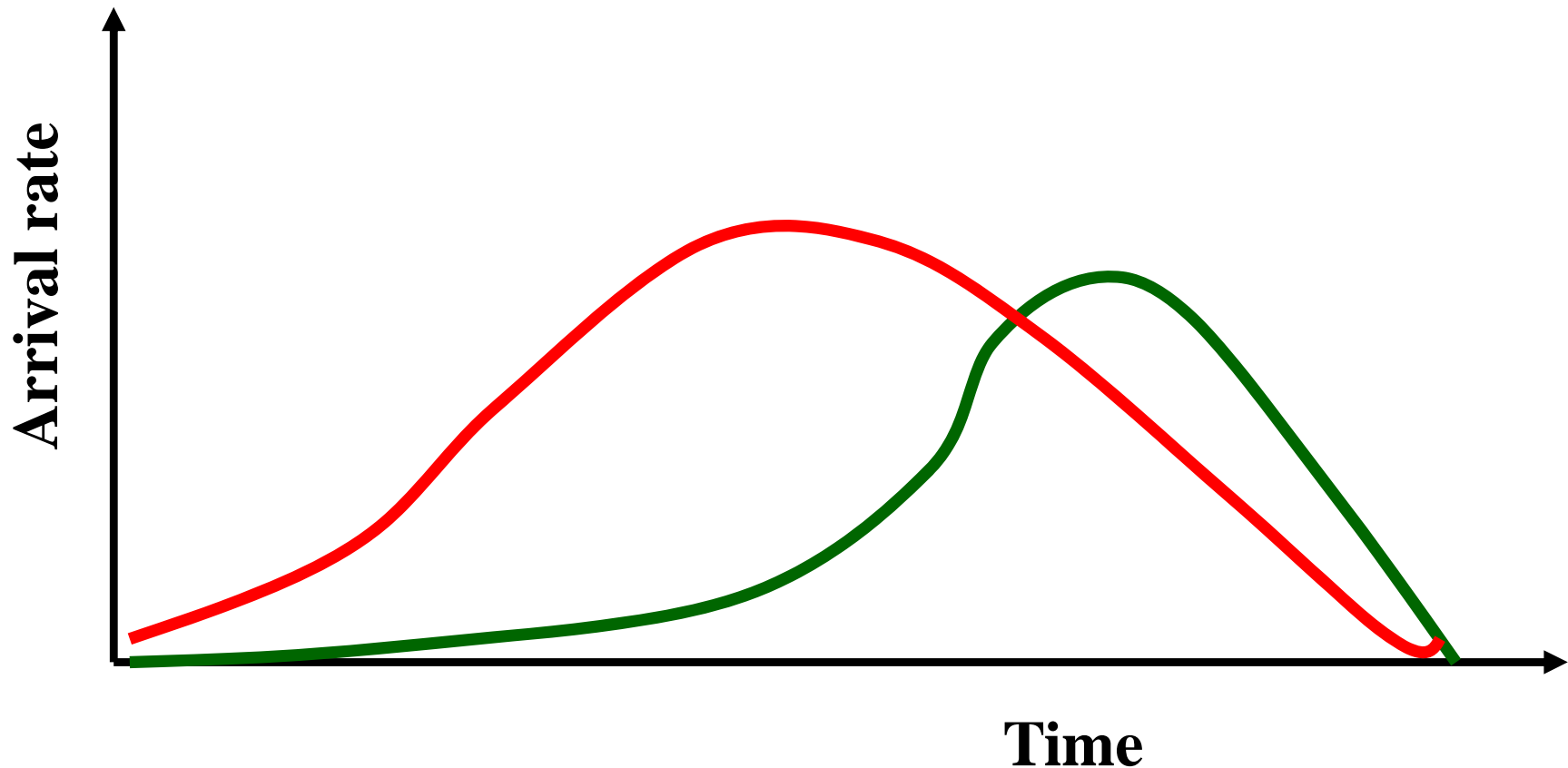
## ■ What price to charge?



# Airline yield management

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- Different customer classes vary in terms of their booking process.



# Airline yield management

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## ■ Demand management techniques

### » Informational

- Advertising services
- Discounts

### » Dual (pricing)

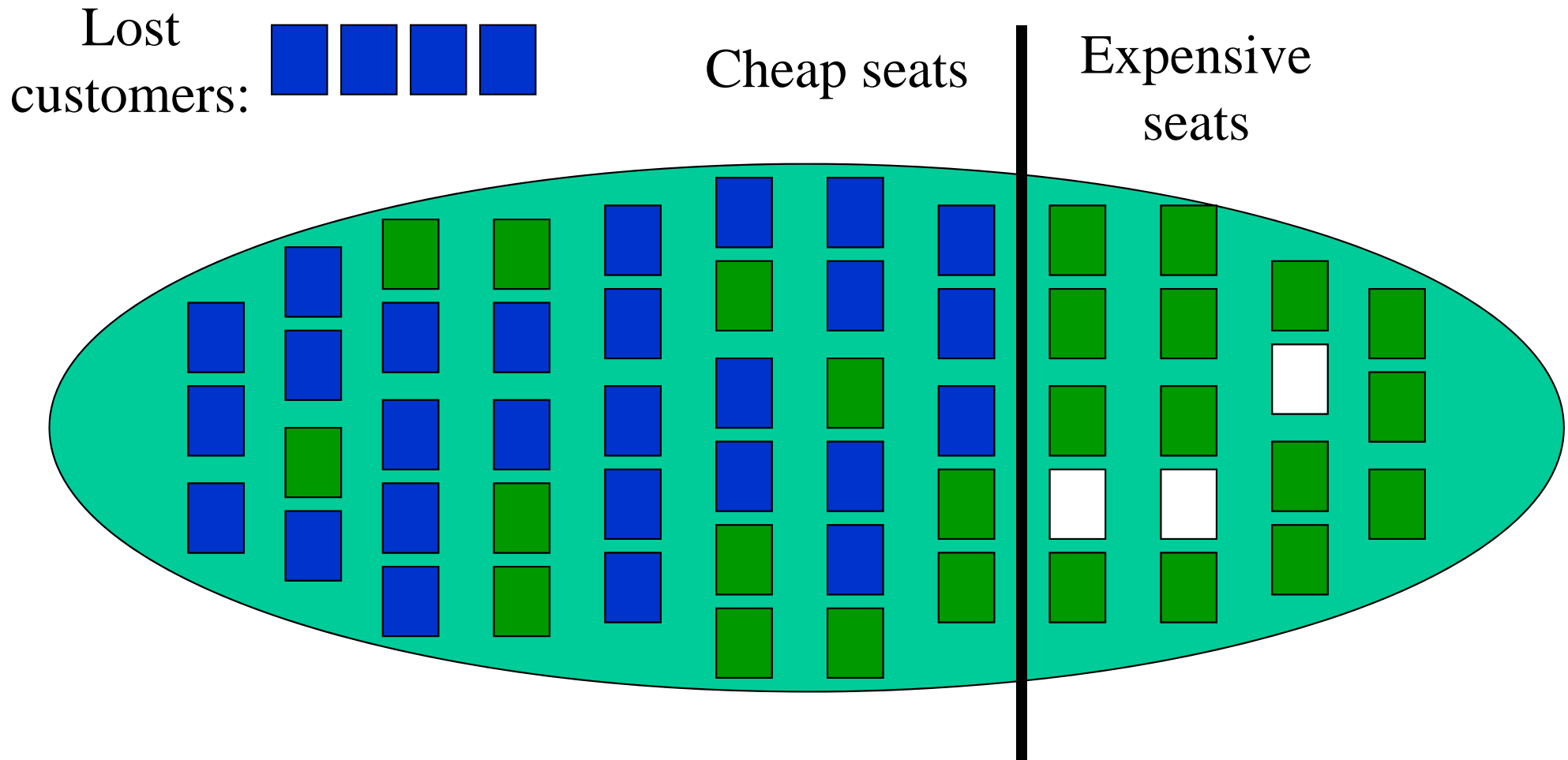
- Setting the price for seats
- Pricing differentiated services

### » Primal (booking limits)

- We have to limit the number of seats we sell at each fare.

# Airline yield management

■ Consider a two-class problem:



*How do we decide if we picked the right booking limit?*

# Airline yield management

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## ■ Challenge question:

- » What happens when the allotment of “cheap seats” is set too high? Too low?



# Airline yield management

## ■ Notation:

Activity variables :

$J$  = Set of fare classes.  $j = 1$  is the "highest" fare class.

$D_j(\omega)$  = (random) Demand for fare class  $j$

$f_j(y)$  = p.d.f. of demand for fare class  $j$ .

Decision variables :

$U_j$  = Limit (upper bound) on number of reservations allowed for fare class  $j$ .

$x_j(\omega)$  = Number of reservations made for fare class  $j$ .

Parameters :

$C$  = Capacity of the aircraft

$p_j$  = Fare (price) for fare class  $j$  ( $p_j \geq p_{j+1}$ )

# Airline yield management

## ■ Objective:

We would like to find the booking limits that solve :

$$\max_{U_1 \dots U_J} E \left\{ \sum_{j \in J} p_j x_j(\omega) \right\}$$

subject to limits on how much we can book.

We can think of  $\omega$  as a single flight, and the expectation is a summation over many flights. We cannot maximize the profits for a single flight, but we would like to maximize profits over many flights.

# Airline yield management

## ■ Booking limits:

- » Separable booking limits- Booking limit  $j$  limits reservations for fare class  $j$  alone:

$$x_j \leq U_j$$

$$\sum_{j \in J} U_j = C = \text{aircraft capacity}$$

- » Nested booking limits - Booking limit  $j$  limits fare class  $j$  and all lower fare classes.

$$\sum_{m \geq j} x_m \leq U_j$$

$$U_j \geq U_{j+1}$$

$$U_1 = C$$

# Airline yield management

A heuristic derivation for booking limits:

Start with a separable (non-nested) policy:

Let:

$$\begin{aligned} P_j(x) &= \text{Prob}[D_j \leq x] \\ &= \int_{y=0}^x f_j(y) dy \end{aligned}$$

This means that if we set a booking limit at  $U_j$ , the probability that we could have exceeded it is  $(1 - P_j(U_j))$ . Now let:

$$\begin{aligned} \text{EMSR}_j(U_j) &= \text{The expected marginal seat revenue for fare class } j \\ &= \text{The value an additional seat allocated for fare class } j. \\ &= p_j (1 - P_j(U_j)). \end{aligned}$$

# Airline yield management

We start by expressing the objective function :

$$F(U) = \max_{U_j} E\{F(U, \omega)\}$$

subject to : 
$$\sum_{j \in J} U_j = C$$

where :

$$F(U, \omega) = \sum_j p_j \min\{D_j(\omega), U_j\}$$

Now consider the derivative :

$$\frac{\partial F(U, \omega)}{\partial U_j} = \begin{cases} p_j & D_j(\omega) > U_j \\ 0 & \text{Otherwise} \end{cases}$$

This means that :

$$E\left\{\frac{\partial F(U, \omega)}{\partial U_j}\right\} = p_j \text{Prob}[D_j(\omega) > U_j] = p_j (1 - P_j(U_j))$$

# Airline yield management

We can formulate the problem as an unconstrained problem by relaxing the constraint:

$$F^L(U, \lambda) = \max_{U_j} E \{ F(U, \omega) \} + \lambda \left( C - \sum_{j \in J} U_j \right)$$

At optimality, we expect to find  $\frac{\partial F^L(U, \lambda)}{\partial U_j} = 0$ . This means that:

$$p_j (1 - P_j(U_j)) - \lambda = 0 \quad \text{for each } j.$$

Or:

$$p_j (1 - P_j(U_j)) = \lambda$$

**What is the economic interpretation of this equation?**

# Airline yield management

## ■ The EMSR algorithm:

For a given value of  $\lambda$ , find the booking limit  $U_j$  :

$$\begin{aligned} P_j(U_j) &= 1 - \frac{\lambda}{p_j} \\ &= \frac{p_j - \lambda}{p_j} \end{aligned}$$

Given  $\lambda$ , we can find  $U_j(\lambda)$  so that this is satisfied.

Now we have to pick  $\lambda$  so that

$$\sum_j U_j(\lambda) = C$$

Since  $\lambda$  is a scalar, just have to try different values.

# Lecture outline - Demand Management

- Principles of demand management
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# Stochastic gradient algorithms

## ■ Nested booking limits:

» A sample booking process:

Fare Class	Demand $D(\omega)$	Booking limit	Total reservations	Fare class reservations
5	42	15	15	15
4	38	40	40	25
3	27	80	67	27
2	59	120	120	53
1	19	150	139	19

# Stochastic gradient algorithms

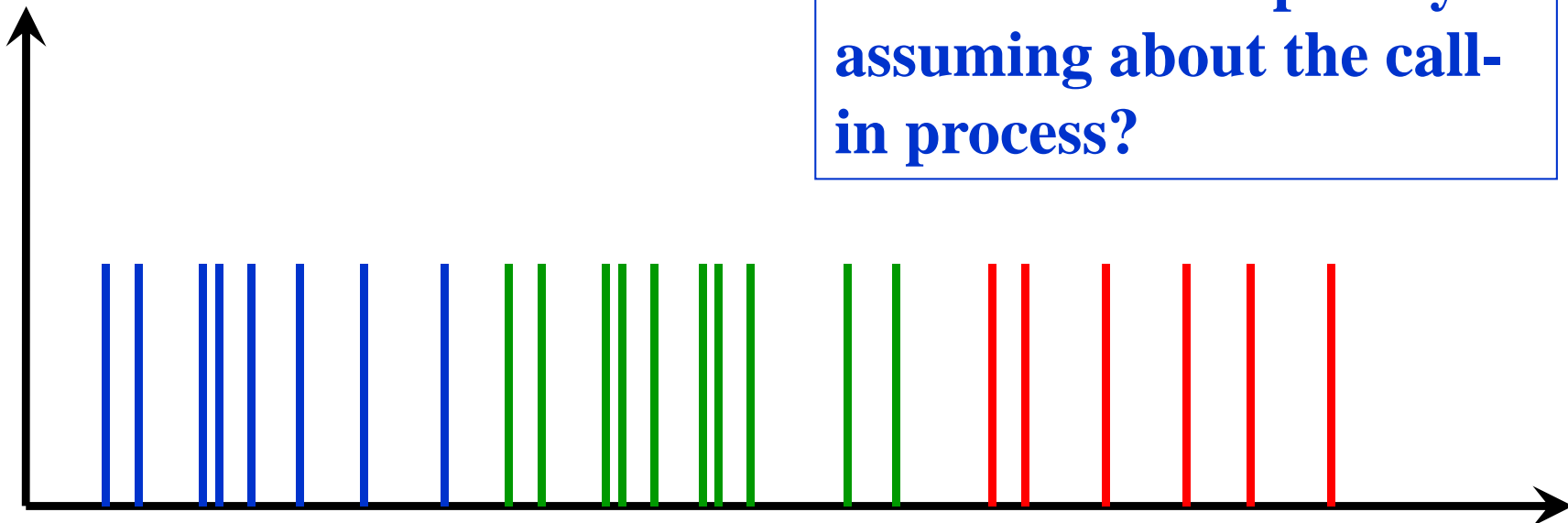
Nested booking limits:

Start by defining the cumulative demand:

$$\begin{aligned}\hat{D}_j(U, \omega) &= \text{Total reservations made of customer classes } j \text{ and "lower."} \\ &= \min \left\{ D_j + \hat{D}_{j+1}(\omega), U_j \right\}\end{aligned}$$

Note:  $\hat{D}_j(U, \omega) - \hat{D}_{j+1}(U, \omega)$  is the number of fare class  $j$  reservations that have been accepted.

**What are we implicitly assuming about the call-in process?**



# Stochastic gradient algorithms

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For this problem, we again want to solve :

$$F(U) = \max_{U_j} E\{F(U, \omega)\}$$

where :

$$F(U, \omega) = \sum_{j \in J} p_j \left( \hat{D}_j(U, \omega) - \hat{D}_{j+1}(U, \omega) \right)$$

But how do we find the  
gradient?

# Stochastic gradient algorithms

Consider a two class system...

We start with  $U_1 = C$ .

$$\begin{aligned} F(U, \omega) &= p_1 \left( \hat{D}_1(U, \omega) - \hat{D}_2(U, \omega) \right) + p_2 \left( \hat{D}_2(U, \omega) \right) \\ &= p_1 \left( \min \left\{ D_1 + \hat{D}_2, U_1 \right\} - \hat{D}_2(U, \omega) \right) + p_2 \hat{D}_2(U, \omega) \end{aligned}$$

Let's now assume that with our low "vacationers" fare  $p_2$ ,

$D_2 \gg U_2$ . In this case, we have:

$$\begin{aligned} F(U, \omega) &= p_1 \left( \min \left\{ D_1 + U_2, U_1 \right\} - U_2 \right) + p_2 U_2 \\ &= p_1 \left( \min \left\{ D_1 + U_2 - U_2, U_1 - U_2 \right\} \right) + p_2 U_2 \\ &= p_1 \min \left\{ D_1, U_1 - U_2 \right\} + p_2 U_2 \end{aligned}$$

# Stochastic gradient algorithms

Taking the derivative with respect to  $U_2$  gives:

$$\frac{\partial F(U, \omega)}{\partial U_2} = -p_1 I_{\{D_1 \geq (U_1 - U_2)\}} + p_2$$

(Remember:  $I_{\{X\}} = 1$  if event  $X$  is true, and 0 otherwise.  $E(I_{\{X\}})$  is then the probability that  $X$  is true.)

Taking expectations gives:

$$E\left[\frac{\partial F(U, \omega)}{\partial U_2}\right] = -p_1 \text{Prob}[D_1 \geq U_1 - U_2] + p_2$$

Recall that  $U_1 = C$ . Setting the derivative equal to zero, we get:

$$p_1 \text{Prob}[D_1 \geq C - U_2] = p_2$$

# Stochastic gradient algorithms

When the problem gets more complicated, we do not get such neat results. Instead, we can resort to our stochastic gradient algorithms.

$$\max E \{F(U, \omega)\} = \max E \left\{ \sum_{j \in J} p_j \left( \hat{D}_j(U, \omega) - \hat{D}_{j+1}(U, \omega) \right) \right\}$$

Assume we have an initial vector  $U^0$ . We can find the best value of  $U$  using the stochastic gradient iteration:

$$U^{k+1} = U^k + \alpha^k \nabla F(U^k, \omega^k)$$

# Stochastic gradient algorithms

■ Consider the change in a booking limit:

» Assume we change from  $U_4=40$  to  $U_4=41$ .

Fare Class	Demand $D(\omega)$	Booking limit	Total reservations	Fare class reservations	Change
5	42	15	15	15	
4	38	40(41)	40(41)	25(26)	+1
3	44	80(80)	80(80)	40(39)	-1
2	55	120(120)	120(120)	40(40)	
1	27	150(150)	147(147)	27(27)	

Change in profits is  $p_4 - p_3$ .

What if  $U_3$  was not binding?

# Stochastic gradient algorithms

■ Consider the change in a booking limit:

» What if  $U_3$  is not binding, but  $U_2$  is?

Fare Class	Demand $D(\omega)$	Booking limit	Total reservations	Fare class reservations	Change
5	42	15	15	15	
4	38	40(41)	40(41)	25(26)	+1
3	34	80(80)	74(75)	34(34)	
2	55	120(120)	120(120)	46(45)	-1
1	27	150(150)	147(147)	27(27)	

Now the change in profits is  $p_4 - p_2$ .



# Stochastic gradient algorithms

## ■ Consider the change in a booking limit:

» What if nothing else is binding?

Fare Class	Demand $D(\omega)$	Booking limit	Total reservations	Fare class reservations	Change
5	42	15	15	15	
4	38	40(41)	40(41)	25(26)	+1
3	34	80(80)	74(75)	34(34)	
2	40	120(120)	114(115)	40(40)	
1	27	150(150)	141(142)	27(27)	

Now the change in profits is  $p_4$ .

# Stochastic gradient algorithms

## ■ Consider the change in a booking limit:

» What if  $U_4$  is not binding, but  $U_2$  is?

Fare Class	Demand $D(\omega)$	Booking limit	Total reservations	Fare class reservations	Change
5	42	15	15	15	
4	20	40(41)	35(35)	20(20)	
3	34	80(80)	69(69)	34(34)	
2	55	120(120)	120(120)	51(51)	
1	27	150(150)	147(147)	27(27)	

Now there is no change in profits.

# Stochastic gradient algorithms

So we have a formula for the gradient when increasing  $U_i$ :

Let:

$$X^+(\omega) = \begin{cases} p_i & \text{If booking limit } U_i \text{ is binding.} \\ 0 & \text{Otherwise.} \end{cases}$$

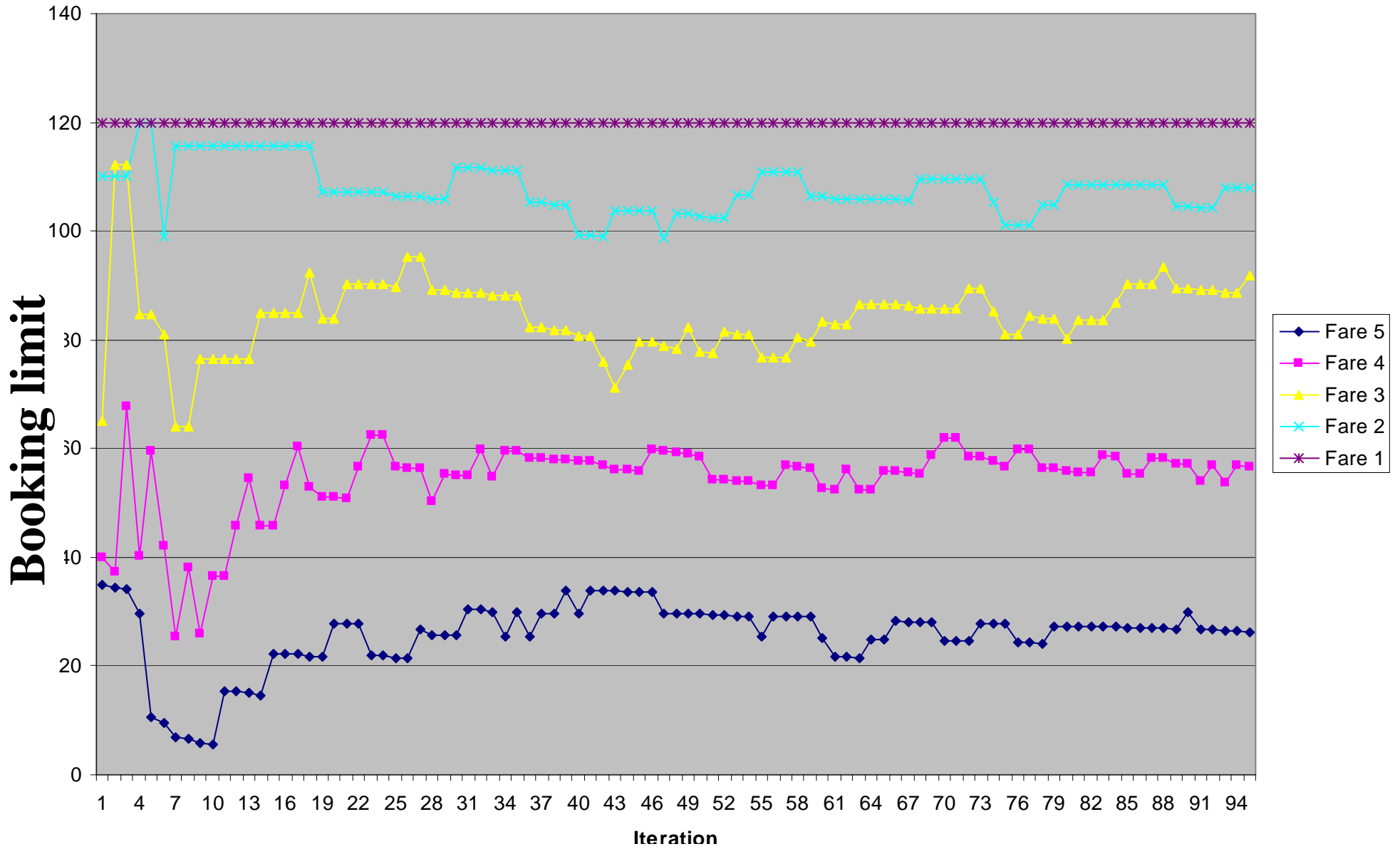
$$X^-(\omega) = \begin{cases} p_j & \text{If } U_i \text{ is binding, and } U_j \text{ is binding, and } j < i \text{ is the first class} \\ & \text{whose booking limit is binding.} \\ 0 & \text{If no other booking limit for a higher fare class is binding.} \end{cases}$$

We can now express the gradient using:

$$\frac{\partial F(U, \omega)}{\partial U_i} = X^+(\omega) - X^-(\omega).$$

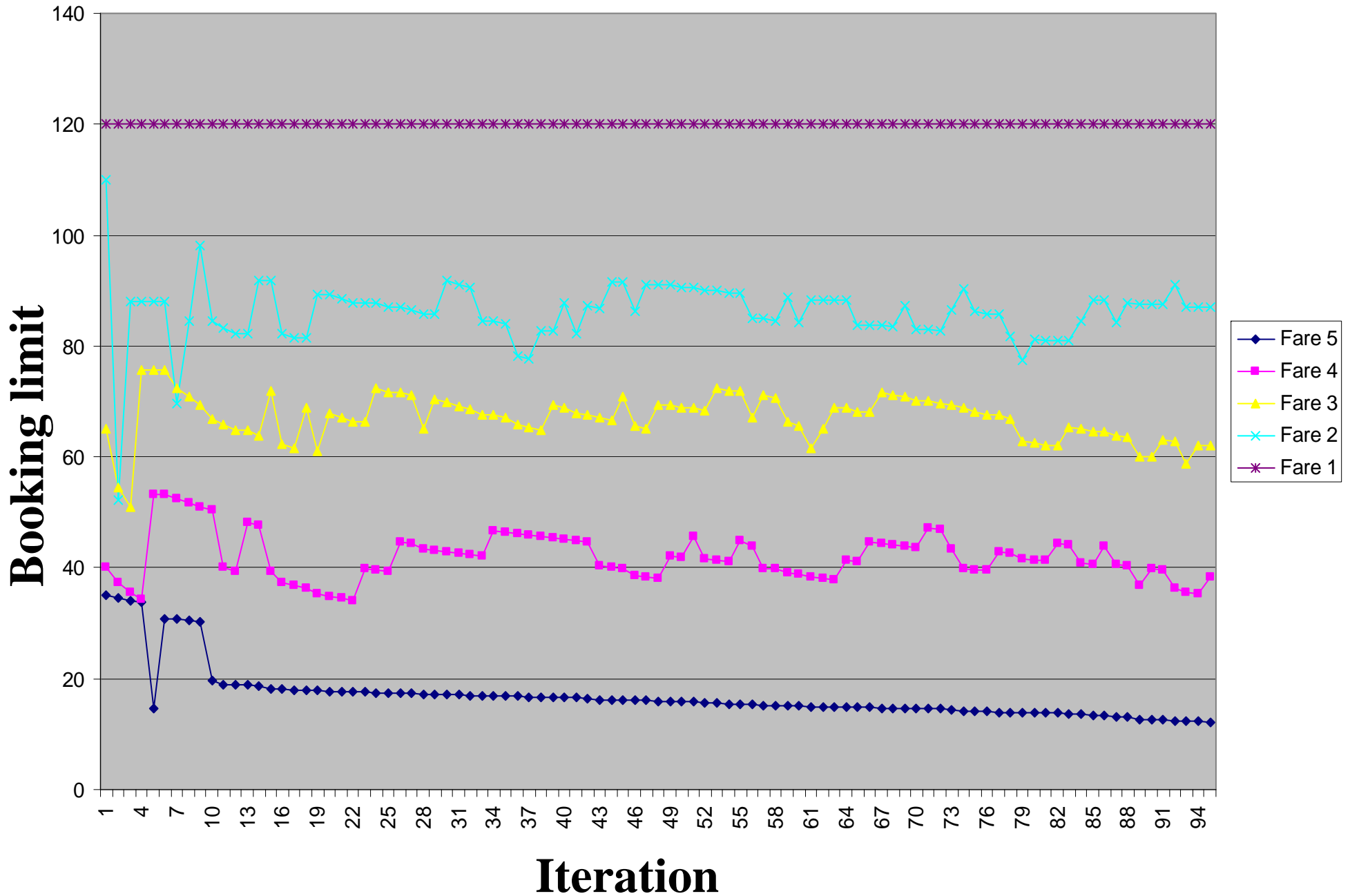
Since  $X^+$  and  $X^-$  depend on the sample realization of demands, they are random variables.

# Booking limits



Iteration

# Higher demand for fare class 1



# Lecture outline - Demand Management



- Issues in demand management

# Extensions

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## ■ The no-show problem

- » How can we manage this?

## ■ The fairness issue

- » Paying different amounts for the “same” seat
- » In what way are seats different?

## ■ Managing complexity

- » 1000 markets
- » 30,000 fares!