

# Revenue Management for Small Hotels<sup>1</sup>

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## Small Independent Hotels vs. Big Chain Hotels



- Less rooms (25-50)
- No (centralised) revenue management system
- Small management team (typically of size 1)

## Our Research

- Collaboration with 5 small independent hotels in the Netherlands
- Research motivated by real hotel data



## Important Properties of Revenue Management System

- Group bookings
- Networks (multiple night stays)
- Cancellations

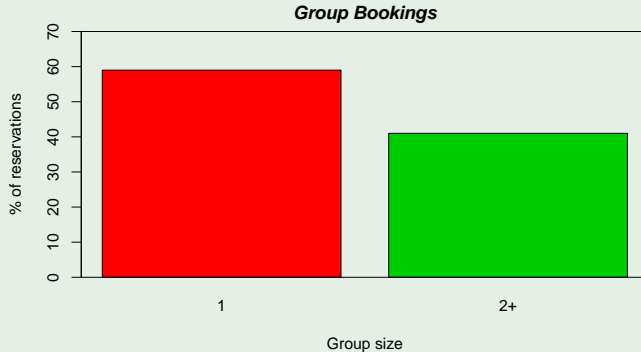


## Example: Data from One Small Dutch Hotel

- The hotel has **34 hotel rooms** with **5 room types**
- Located in the **countryside**
- Attracts **business** as well as **leisure** clients
- **Competition** mainly consists of a big chain hotel



## Group Bookings

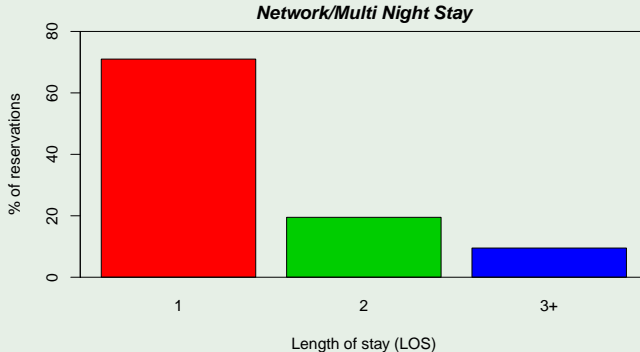


### Observation

Large part (41%) of all bookings are group bookings



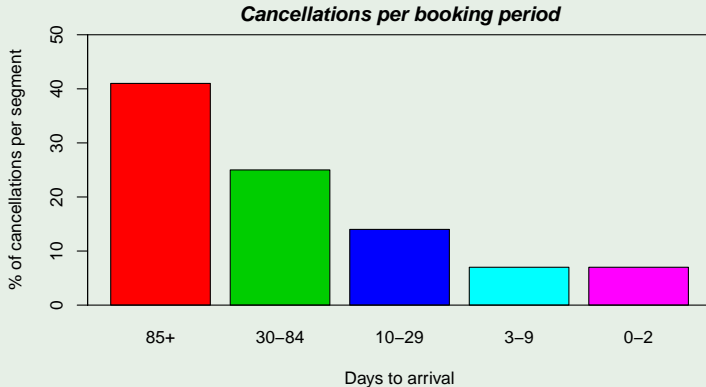
## Networks/Multiple Night Stays



### Observation

Big part (29%) stays more than one night

## Cancellations



## Observations

- 22% of all bookings are cancelled
- Early booking  $\implies$  high cancellation probability



## Customer Choice Cancellation Model

### *Properties:*

- Customer choice behaviour
- Cancellation
- Overbooking

### *Related work:*

- Subramanian et alii (1999): Overbooking and cancellations
- Talluri and Van Ryzin (2004): Customer choice behaviour
- Ge and Pan (2010): Customer choice behaviour and overbooking

## Example (Talluri & van Ryzin, 2004)

Hotel with

- $C = 20$  rooms
- $n = 3$  products with prices

$$r_1 = 800$$

$$r_2 = 500$$

$$r_3 = 450$$

- $T$  days before arrival
- $\lambda = 1/4$  probability that a customer arrives
- $x_j$  number of reservations for product  $j$  ( $x = (x_1, x_2, x_3)$ )
- $\gamma(x_j) = x_j/100$  probability that product  $j$  is cancelled
- $L_j = r_j$  loss if product  $j$  is cancelled

## Example (continued)

- $P(S, j)$  probability that customer buys product  $j$  if  $S \subset \{1, 2, 3\}$  is offered
- $P(S, 0) = 0$  probability that customer buys nothing
- E.g.  $S = \{1, 2\}$  and

$$P(S, 1) = 0.1$$

$$P(S, 2) = 0.6$$

$$P(S, 3) = 0$$

$$P(S, 0) = 0.3$$

## Objective

Which products  $S \subset \{1, 2, 3\}$  do we offer  $t$  days before arrival in state  $x$

## Solution

Model as Markov decision process and solve with dynamic programming



## Problems with exact model

- State space too large to solve ( $x = (x_1, \dots, x_n)$ )
- Action space too large ( $|\mathcal{P}(\{1, \dots, n\})| = 2^n$ )

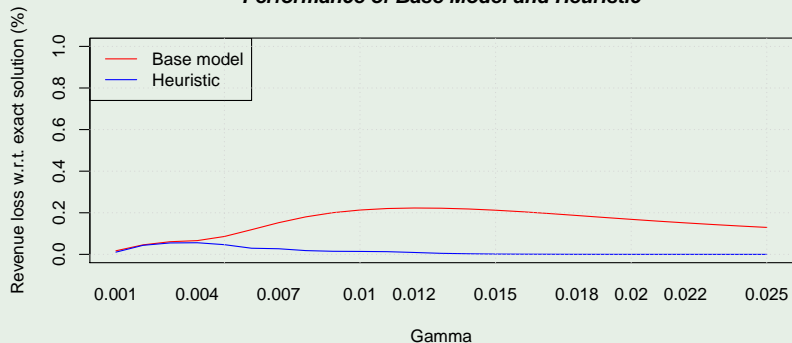
## Solution: heuristic

Outline algorithm:

- One-dimensional state space ( $x \in \mathbb{N}$  number of reservations)
- Procedure:
  - 1 Fix strategy and approximate loss  $L_j$
  - 2 Solve one-dimensional problem to find next strategy
  - 3 Use strategy to approximate loss  $L_j$  and go to step 2
- *Upside:*
  - 1 Smaller state space (one-dimensional)
  - 2 Smaller action space (most sets  $S$  can be ruled out)
- *Downside:* Not always convergent

## Numerical Results

*Performance of Base Model and Heuristic*



## Observations

- Base model performs bad
- Heuristic performs well



## Conclusions

- Cancellations have a big impact on revenue
- The heuristic approximates solution well

## Further Research

- Bounds solution of heuristic
- Apply cancellation model to real data
- Group bookings and networks