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**Scheduling interval planning with time-dependent no-shows and cancellations**



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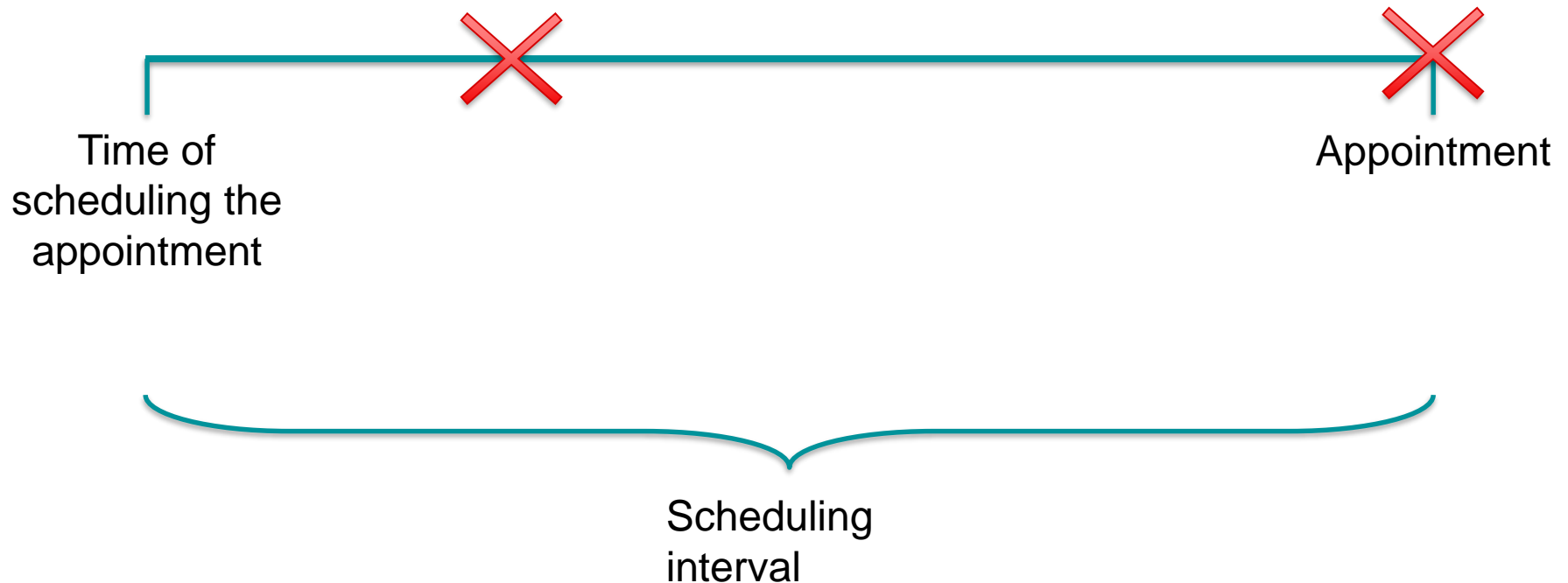
# Background

- Healthcare clinics face reduced quality of care and inefficient operations due to patient no-shows and cancellations
- Strategies:
  - Influencing patient behavior, e.g., by reminders
  - Overbooking
  - Open access scheduling
  - Panel sizing
  - Limiting the scheduling interval



# Data-analysis (1)

- No-show and cancellation rate depend on scheduling interval



# Data-analysis (2)

- No-show and cancellation rate depend on scheduling interval

## USA data

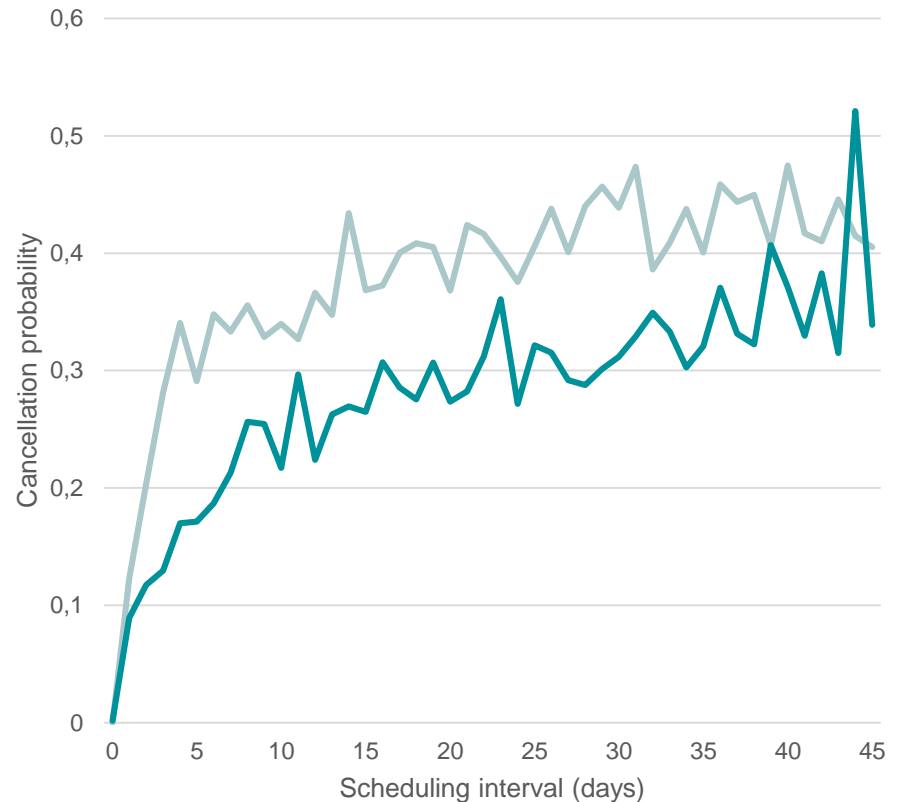
- No-show: weak positive monotonic correlation ( $\rho=0,344$ ,  $n=61$ ,  $p=0.007$ )
- Cancellation: strong positive monotonic correlation ( $\rho=0,741$ ,  $n=61$ ,  $p<0.001$ )

## EU data

- No-show: weak positive monotonic correlation ( $\rho=0,230$ ,  $n=61$ ,  $p=0.230$ )
- Cancellation: strong positive monotonic correlation ( $\rho=0,877$ ,  $n=61$ ,  $p<0.001$ )

## Data-analysis (3)

- Cancellation probability given the length of the scheduling interval for both institutions
- Data shows similar trend as described in the literature:

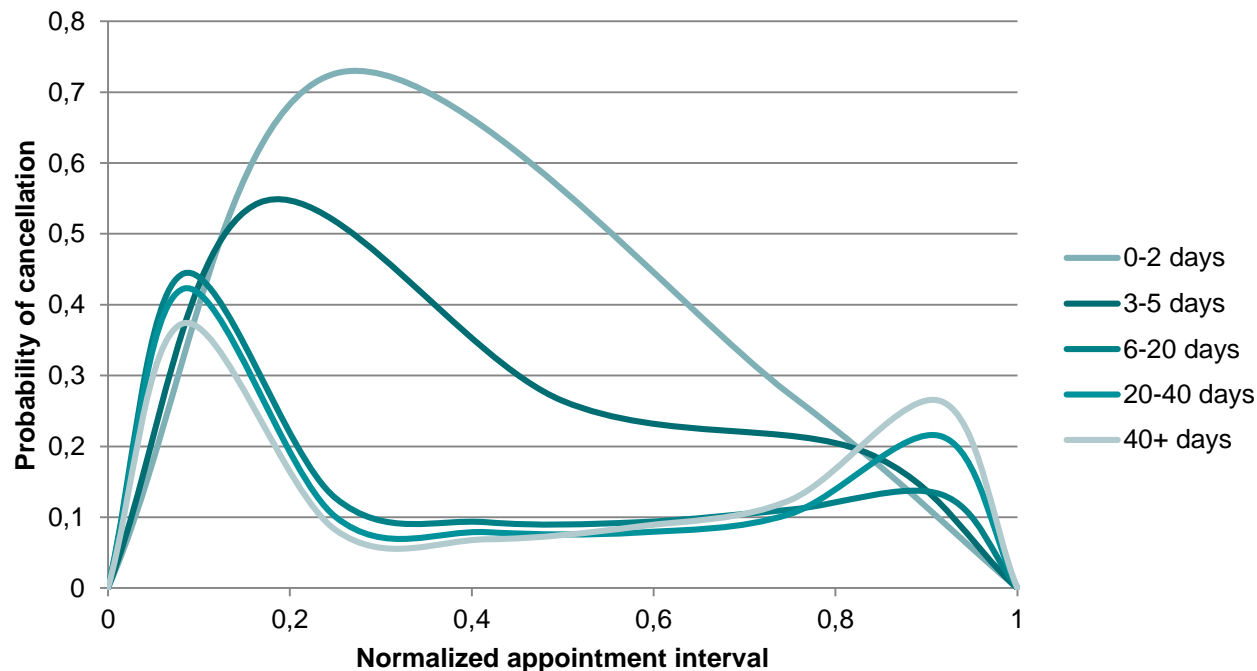


$$\chi_j = \chi_{max} - (\chi_{max} - \chi_0)e^{\lfloor -j/\mu \rfloor / c}$$

As derived from Green and Savin (2008)

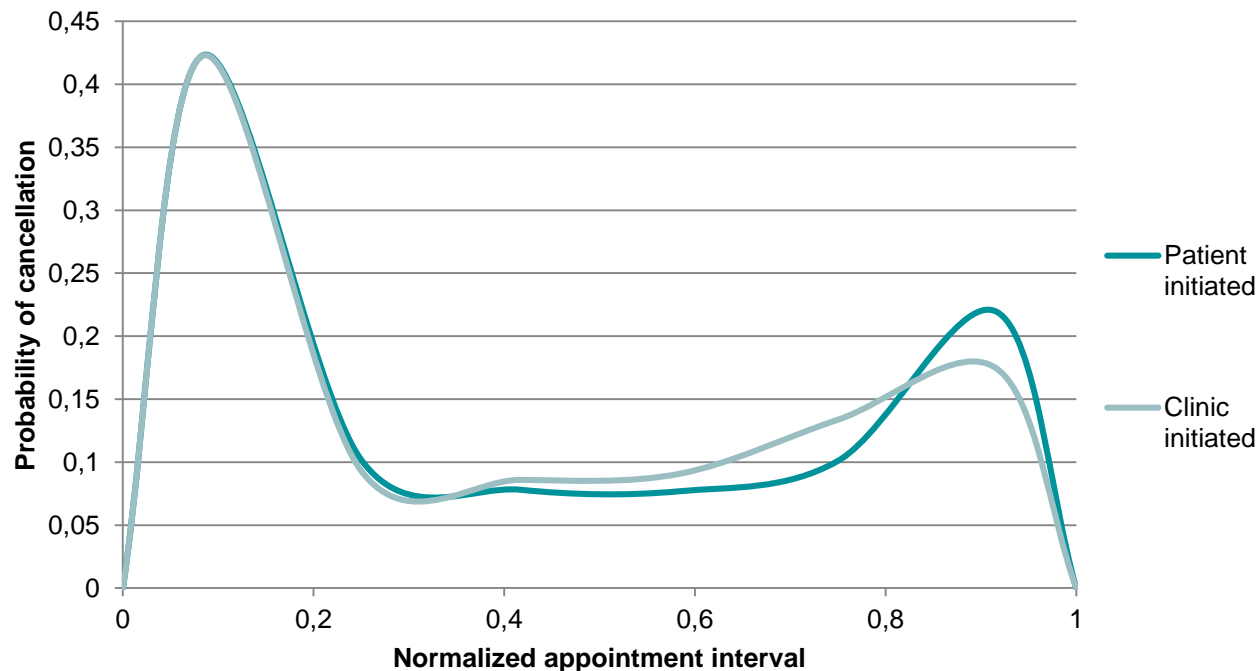
# Data-analysis (4)

- The timing of cancellations within the scheduling interval is bimodal distributed



## Data-analysis (5)

- Similar behavior in timing for patient and clinic initiated cancellations





# Length of scheduling interval?

## Open access

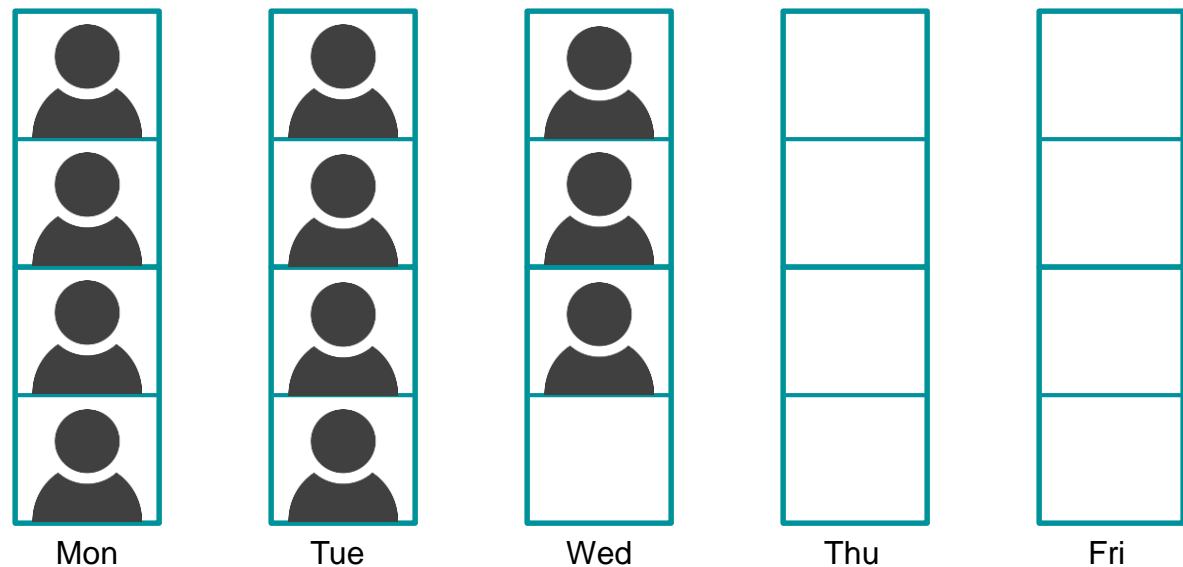
- Few no-shows and cancellations
- Increase of rejections, overtime, or need to maintain a waiting list

## Unlimited

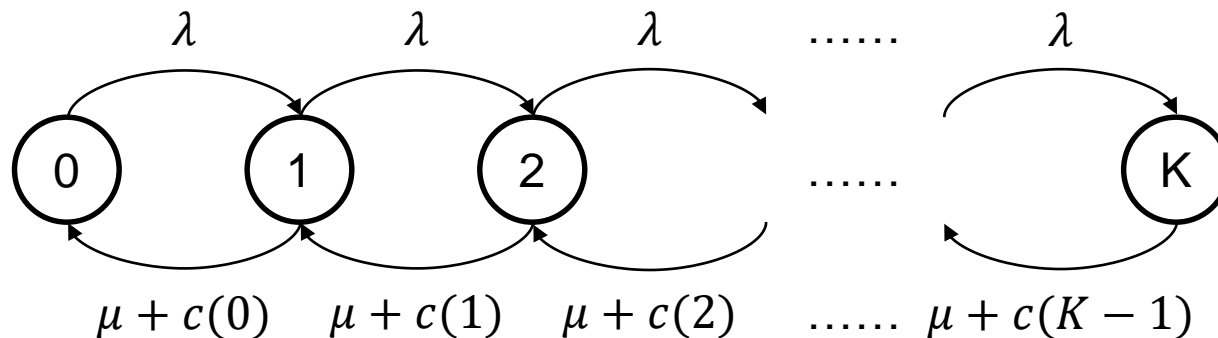
- Many no-shows and cancellations
- No rejections (but long lead times that may interrupt continuity of care)

# Scheduling interval optimization model (1)

- Single-server queueing system with no-shows, regening in the queue, and balking: M/M/1/K
  - FCFS service strategy
  - Example:  $K = 20$



## Scheduling interval optimization model (2)

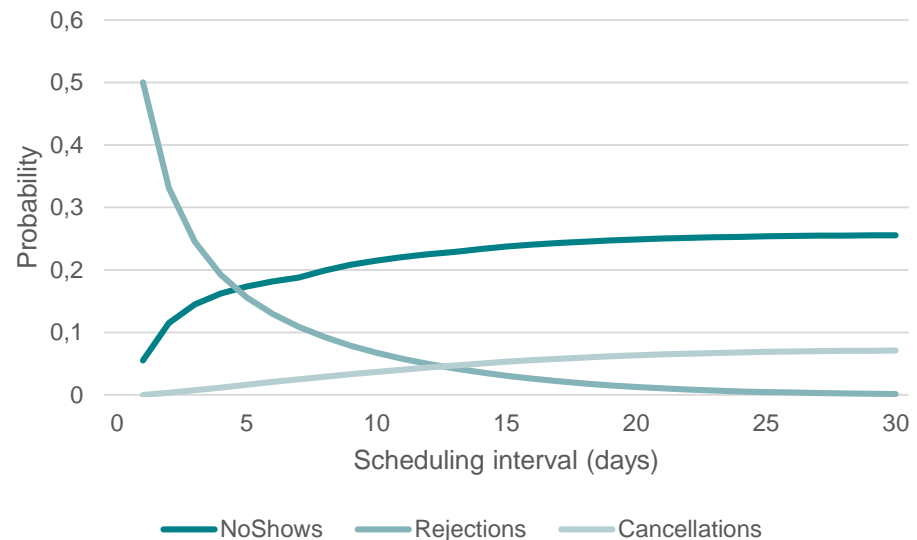


- Objective: find scheduling interval with maximal system revenue
  - Access (rejection cost)
  - Efficiency (service revenue)

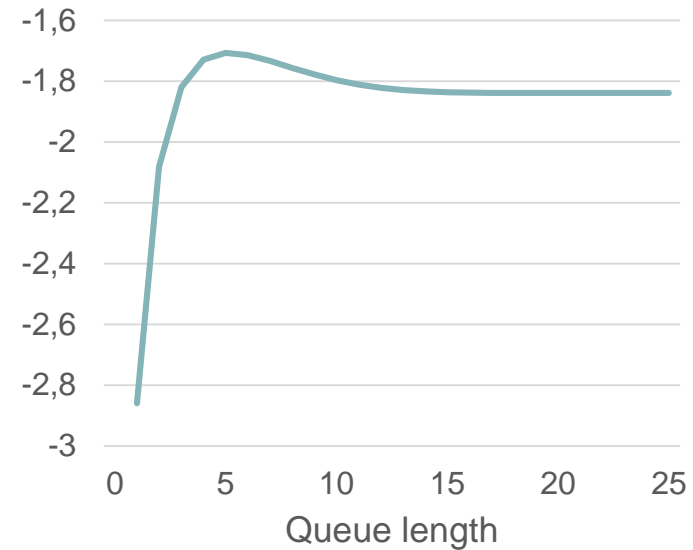
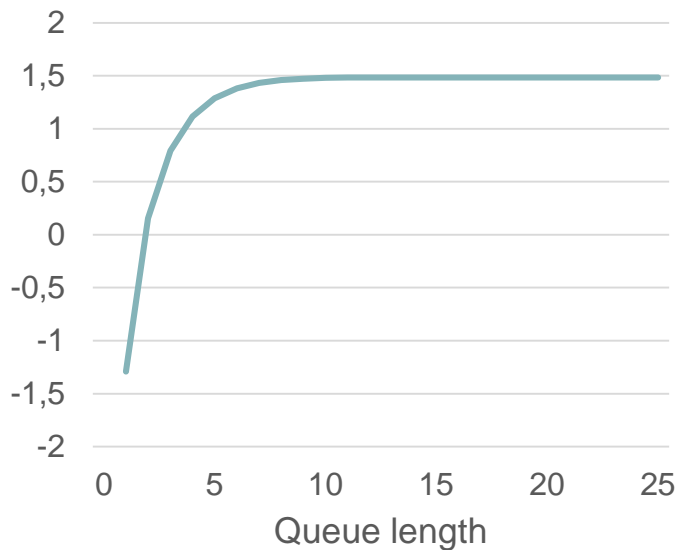
$$R(K) = \lambda P_S(K) + \mu P_N(K) \theta_N - \lambda P_B(K) \theta_B - \lambda P_C(K) \theta_C$$

# Experiments and results (1)

- Simulation model to assess assumption of time-dependent cancellation behavior and FCFS service strategy
- Experiments to assess impact of parameter variations
  - $\lambda$
  - No-show rate
  - Cancellation rate
  - Objective weights



## Experiments and results (2)



- Low demand & low cancellation
- Long scheduling interval

- High demand & high cancellation
- Short scheduling interval

# Conclusions

- No-show and cancellation rates are time-dependent
  - Timing of cancellations is bimodal distributed
  - No-show rate converges faster than cancellation rate
  - Analytical model to determine scheduling interval, which is applied to EU and USA institutions
- 
- |                                  |                                    |
|----------------------------------|------------------------------------|
| ➤ Low demand & low cancellation? | ➤ High demand & high cancellation? |
| ➤ Long scheduling interval       | ➤ Short scheduling interval        |





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**Thank you!**

