

# Helping the Captive Audience: Advance Notice of Diagnostic Service for Hospital Inpatients

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# Hospital Diagnostic Service is Essential to Patient Care

- Canada expenditures

- Total hospital expenditures exceed \$80 billion in 2022
- Medical imaging alone accounts for 4.4%, i.e., \$3.52 billion, in 2022
- Average annual growth rate in total hospital expenditures since 2005 is about 4% (Canadian Institute for Health Information, 2024)

- US expenditures

- Medicare Part B expenditures on hospital imaging services reached \$3.53 billion in 2006 (*U.S. GAO 2008*).
- 2006: 51.0 million CT scans and 14.6 million MRI scans in U.S. hospitals.
- 2019: 71.3 million CT scans and 29.4 million MRI scans (*OECD 2021*).



# Hospital Diagnostic Service Management

- Efforts in improving patient scheduling
  - Improving timeliness for outpatients (**OPs**) and emergency patients (**EPs**).
- However, hospital inpatients (**IPs**) may not be the focus of these efforts.

IP scheduling is a “*pro forma scheduling*” ... “*with schedulers entering incorrect or imaginary appointments as placeholders*”

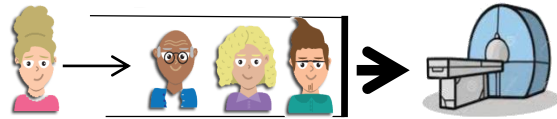
--- *Efficient Radiology*, Daniel Rosenthal and Oleg Pianykh (2021)

# Current Practice

- IPs are scheduled arbitrary appointment times, but are **not** honored.
  - Service provider calls for IPs **whenever she has time** and the diagnostic equipment becomes available.
- IPs are perceived as always on standby and readily available.
  - IPs are “**captive audience**” (*Rosenthal and Pianykh 2021*).
- This approach “has little basis” *and* “creates numerous problems” (*Rosenthal and Pianykh 2021*).
  - Conflict with other IP care activities, or nursing/supporting staff availability.
  - May lead to distrust among these providers and patients.

# Classic Scheduling Paradigms

- Allocation scheduling



- IPs added to a waitlist with no specific service time.
- IPs have to stand by until their services.

- *Gerchak et al. 1996, Huh et al. 2013, Wang and Truong 2017, Liu et al. 2019 ...*

Min patient convenience

Max provider flexibility

- Advance scheduling



- Patients are provided specific service times.
- Providers must honor scheduled times.

- *Liu et al. 2010, Truong 2015, Samiedaluie et al. 2017, Wang et al. 2018, Diamant et al. 2018, Keyvanshokoo et al. 2021, Zacharias et al. 2022 ...*

Max patient convenience

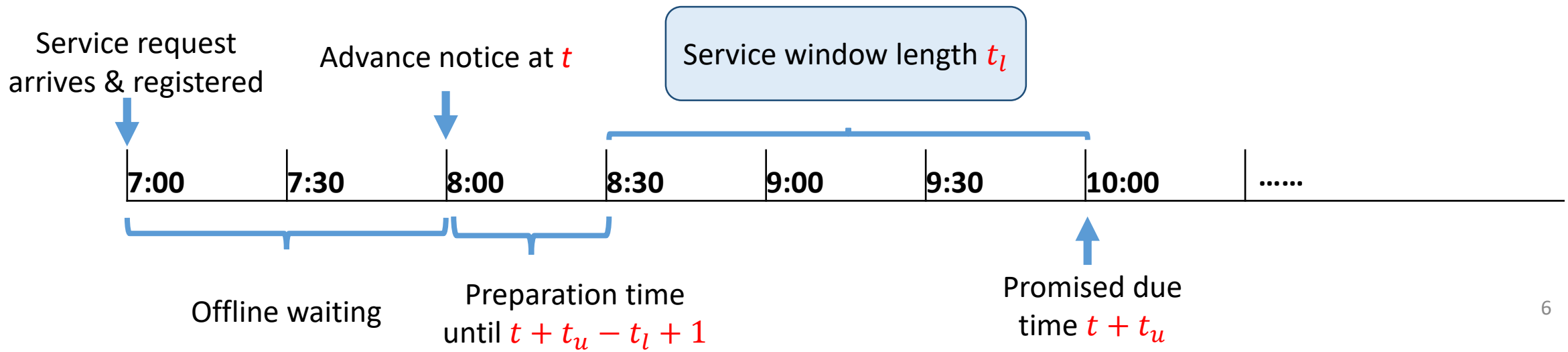
Min provider flexibility

Can we achieve a better tradeoff?



# A New Scheduling Paradigm – Advance Notice

- We propose **advance notice** for IPs.
  - Guarantee that IP exams are completed within  $t_u$  intervals after the notice
  - Not starting their exams until  $(t_u - t_l + 1)$  intervals after the notice
  - The length of service time window is  $t_l$
- It reduces IP standby time (no last-minute notice) and also grants provider service flexibility (no specific service time).



# Feedback from Practitioners on Advance Notice

- Mayo Clinic is fairly positive because they tried something similar.
  - The radiology staff will leave a note to inpatient nurses in advance, notifying that they plan to pick up the inpatient for exams in the next 15 minutes to 1 hour.
- Dr. Daniel Rosenthal shared that they have also implemented “something somewhat similar” (at Mass General), but in a “less structured” way, by “[providing] the inpatient floors with information about the usual time that a call should be expected”.
  - Dr. Rosenthal is Senior Vice Chairman, Department of Radiology & Professor of Radiology, Harvard Medical School. He coauthored the book “Efficient Radiology: How to Optimize Radiology Operations”.
- Largely based on staff experience and intuition, without scientific or data support.

# (Brief) Literature Review

- Dynamic scheduling
  - Advance scheduling: *Patrick et al. 2008, Liu et al. 2010, Truong 2015, Diamant et al. 2018, Keyvanshokoo et al. 2021, Zacharias et al. 2022, Chambers et al. 2023*
  - Allocation scheduling: *Gerchak et al. 1996, Huh et al. 2013, Wang and Truong 2017, Liu et al. 2019*
- Diagnostic service management
  - Follow either the advance or allocation scheduling frameworks: *Green et al. 2006, Kolisch and Sickinger 2008, Patrick and Puterman 2007, Saure et al. 2020*
- Other OM literature
  - Dynamic inventory control for perishable product: *Nahmias 1975, Chen et al. 2014, Li and Yu 2014, Li et al. 2016, Sarhangian et al. 2018*
  - Curbside pickup: *Cavdar and Isik 2020, Farahani et al. 2020*

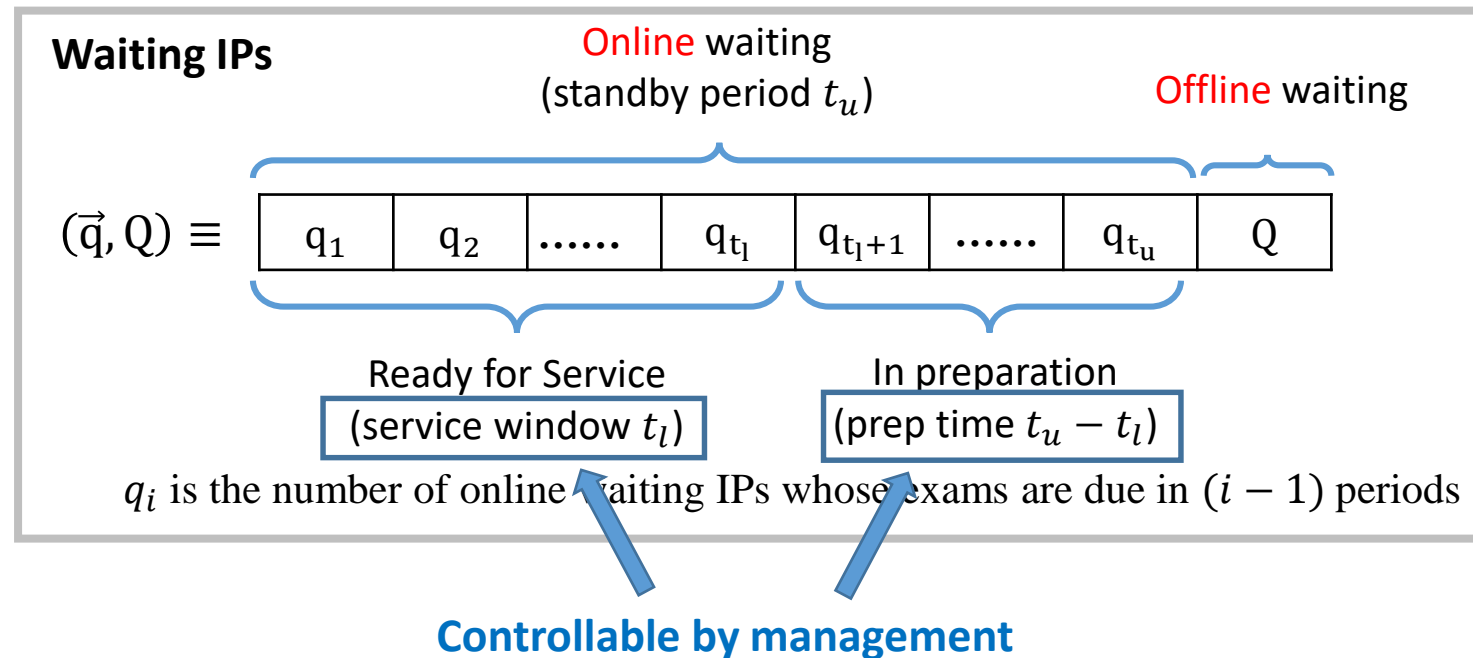


# Problem Formulation as an MDP

- Diagnostic service
  - Operate  $M$  servers within a time horizon of  $T$  periods
  - Each server can serve one patient in a period.
- Stochastic demand from OPs, IPs, and EPs at each interval
  - OPs **random**: scheduled ahead but with uncertainty in arrival time and duration of check-in and preparation procedures
  - IPs and EPs **random**: random arrivals
- Priorities
  - EPs and OPs served in the same interval as they arrive, but IPs can wait.  
(*Kolisch and Sickinger 2008, Green et al. 2006*)

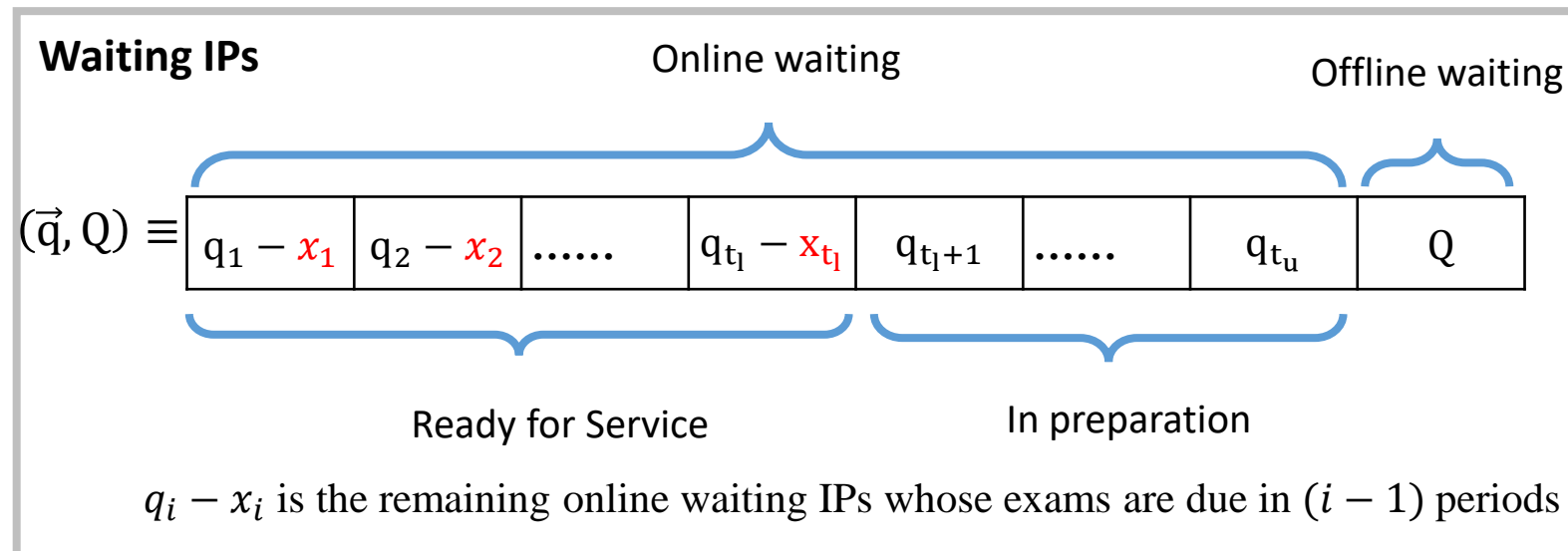
# As Each Interval Starts, the Provider Observes:

- **Offline** waiting IPs ( $Q$ ) who have not received advance notice
- **Online** waiting IPs ( $\vec{q}$ ) who have received advance notice



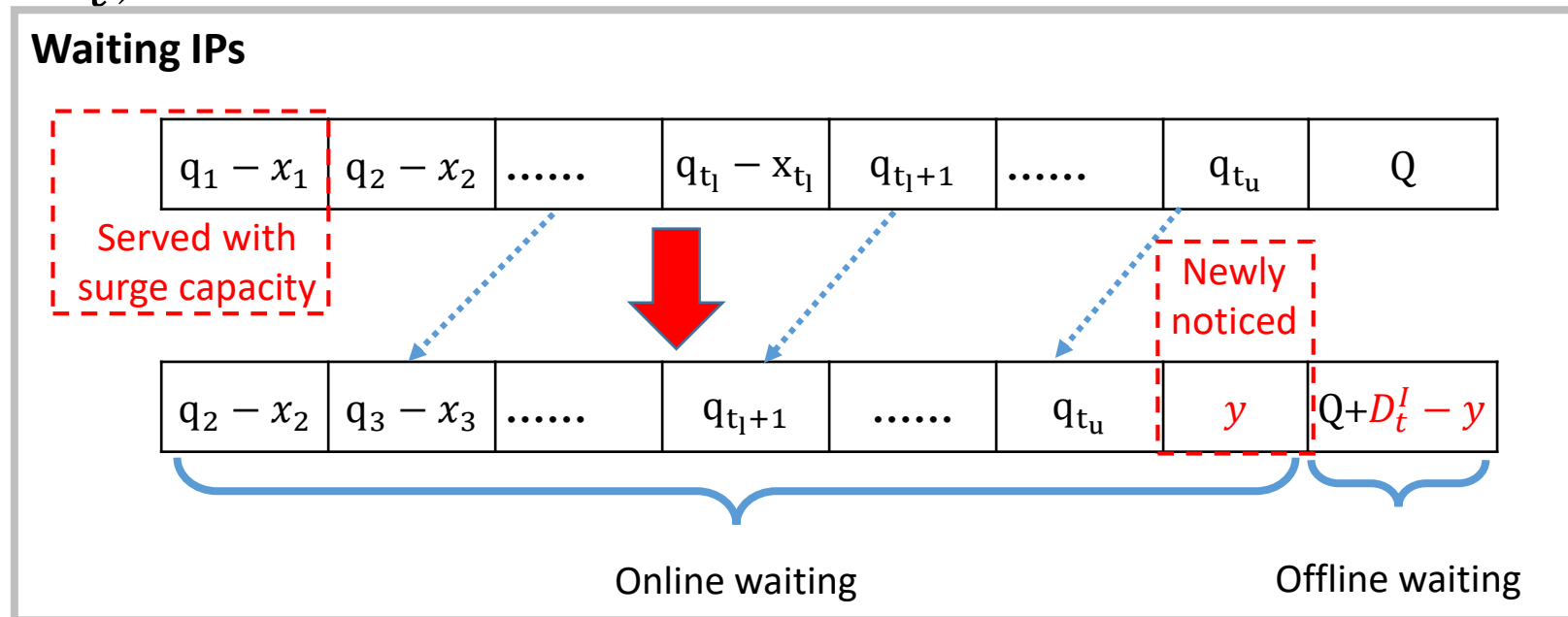
# First Decision $\vec{x}$ : Who to Serve

- New demands from OPs ( $D_t^O$ ) and EPs ( $D_t^E$ ) realized and served
- Use the remaining capacity to serve online waiting IPs
  - Remaining capacity  $S_t \equiv (M - D_t^O - D_t^E)^+$
  - IP exams  $\vec{x} \equiv (x_1, x_2, \dots, x_{t_l})$ ,  $x_i \leq q_i, i = 1, 2, \dots, t_l$ ;  $\sum_{i=1}^{t_l} x_i \leq S_t$



# Second Decision $y$ : How Many Advance Notices to Send

- The remaining IP exams due in the current period ( $q_1 - x_1$ ), if any, are served by using surge capacity
- Observe  $D_t^I$  new IP requests
- At the end of the period, send out  $y$  advance notices to offline waiting IPs ( $y \leq Q + D_t^I$ )



# Cost Structure, Objective, and Formulation

- Maximize the total discounted “rewards” of serving IPs
  - Revenue ( $r$ ) of serving each IP
  - Cost ( $w_{on}$ ) of online IP waiting per interval
  - Cost ( $w_{off}$ ) of offline IP waiting per interval
  - Cost ( $p$ ) of using surge capacity to serve each “overdue” IP and to serve remaining IPs at the end of the horizon
- The problem can be formulated as a Markov Decision Process model
  - First decision: who to serve, i.e.,  $\vec{x}$
  - Second decision: how many advance notices to send, i.e.,  $y$

# Structural Results - Service Strategy $\vec{x}$

- Cost of online waiting  $>$  offline waiting

**Assumption 1**

$$w_{on} > w_{off} \geq 0.$$

- Service provider does not use surge capacity unless it is absolutely necessary.

**Assumption 2**

$$\frac{w_{on}}{1-\gamma} \leq p$$

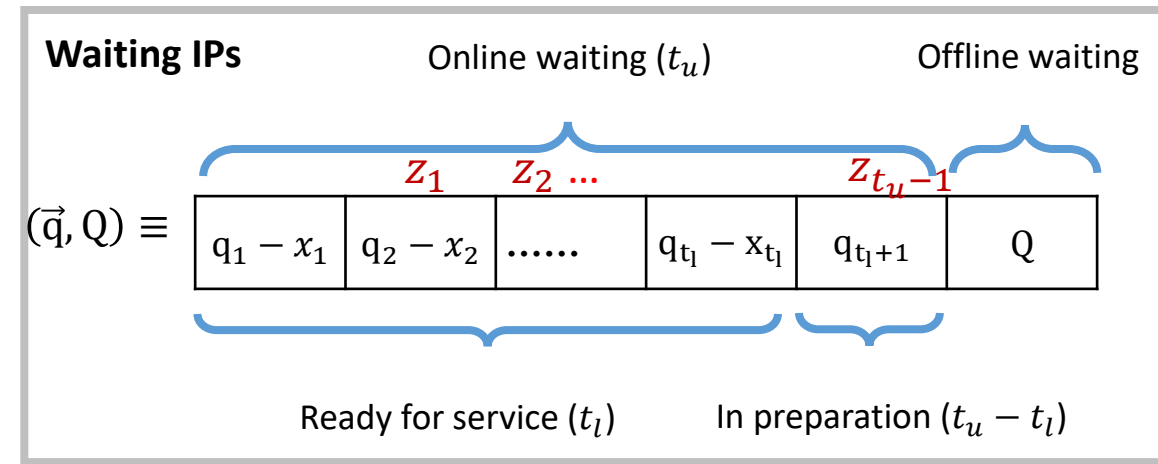
## **Optimality of the FCFS service strategy**

The optimal service strategy is to serve online waiting IPs on a FCFS basis, i.e., use up all available capacity and serve those with most imminent promised due time first.

# Structural Results - Advance Notice $y$

- Structural analysis on advance notice decisions
  - Via a simple and yet novel variable transformation
  - Reveal the hidden *antimultimodular* structure of the problem

# Monotonicity and Bounded Sensitivity of $y^*$



## □ The optimal number of advance notices ( $y^*$ )

- $y^*$  increases as offline waiting IPs ( $Q$ ) increase
- $y^*$  decreases as online waiting IPs ( $z_i$ ) increase
- Online waiting IPs who have imminent promised due time (*smaller  $i$* ) have **smaller impacts** on  $y^*$

## □ Bounded sensitivity

- If  $Q$  or  $z_i$  changes by 1,  $y^*$  **changes no more than 1**
- we only need to check two possible neighboring solutions around a known solution  $\rightarrow$  efficient solutions!



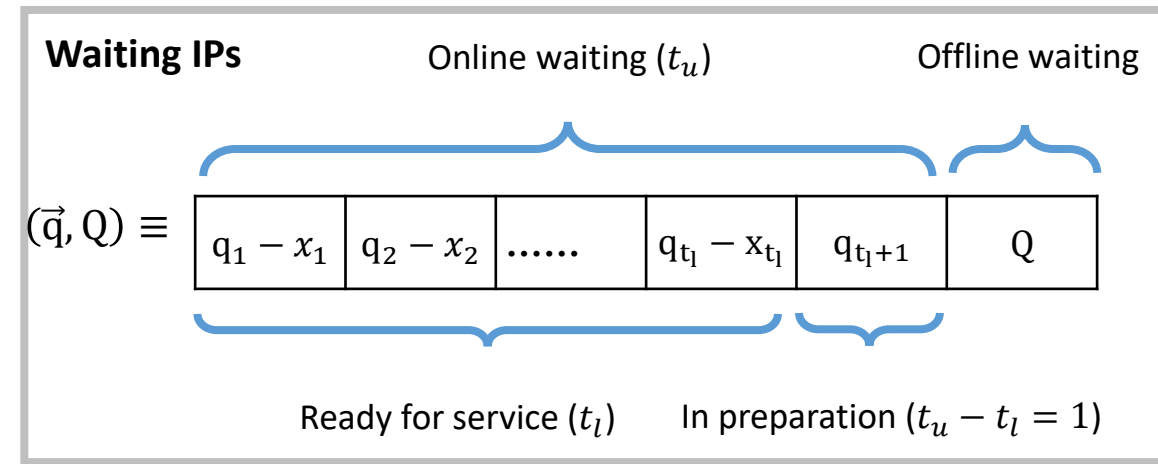
# Case Study

- 9-month data from a computed tomography (CT) practice in Mayo Clinic
  - Operates  $M=4$  scanners and serves IPs, OPs and EPs.
  - Day shift consists of  $T=18$  service intervals (each 30 minutes based on median exam duration)
  - We derive empirical demand distribution based on historical data (non-stationary arrivals)
- Cost parameters based on empirical studies in literature (*Sistrom and McKay 2005, Liu et al. 2018, Hathaway et al. 2021*)
  - Revenue per patient  $r = \$416$
  - Cost of surge capacity usage  $p = \$76.5$
  - Cost of online waiting per interval  $w_{on} = \$3.18$
  - Cost of offline waiting per interval  $w_{off} = \$0.57$

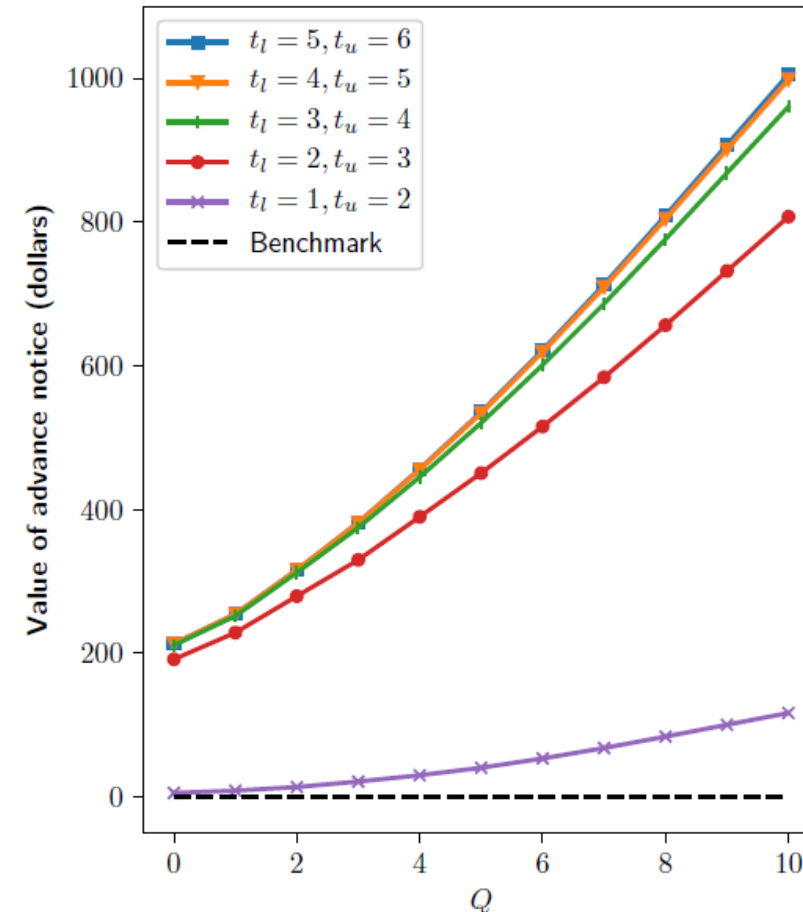
# Value of Advance Notice

- In benchmark model, we set  $t_l = 1$ ,  $t_u=2$ , and treat all waiting as online waiting.
  - Everyone is standing by, send notice 1 period ahead, and allow 1 period transportation time
  - Likely provides a performance upper bound of current practice.
- In our advance notice model, we set the prep time, i.e.,  $t_u - t_l$ , as one period (i.e., 30 mins) based historical data.
- Compare the cumulative rewards for different starting  $Q$ 's, while fixing  $\mathbf{q} = \mathbf{0}$ .
  - There may be physician orders sent over night.

# Impact of Service Window $t_l$

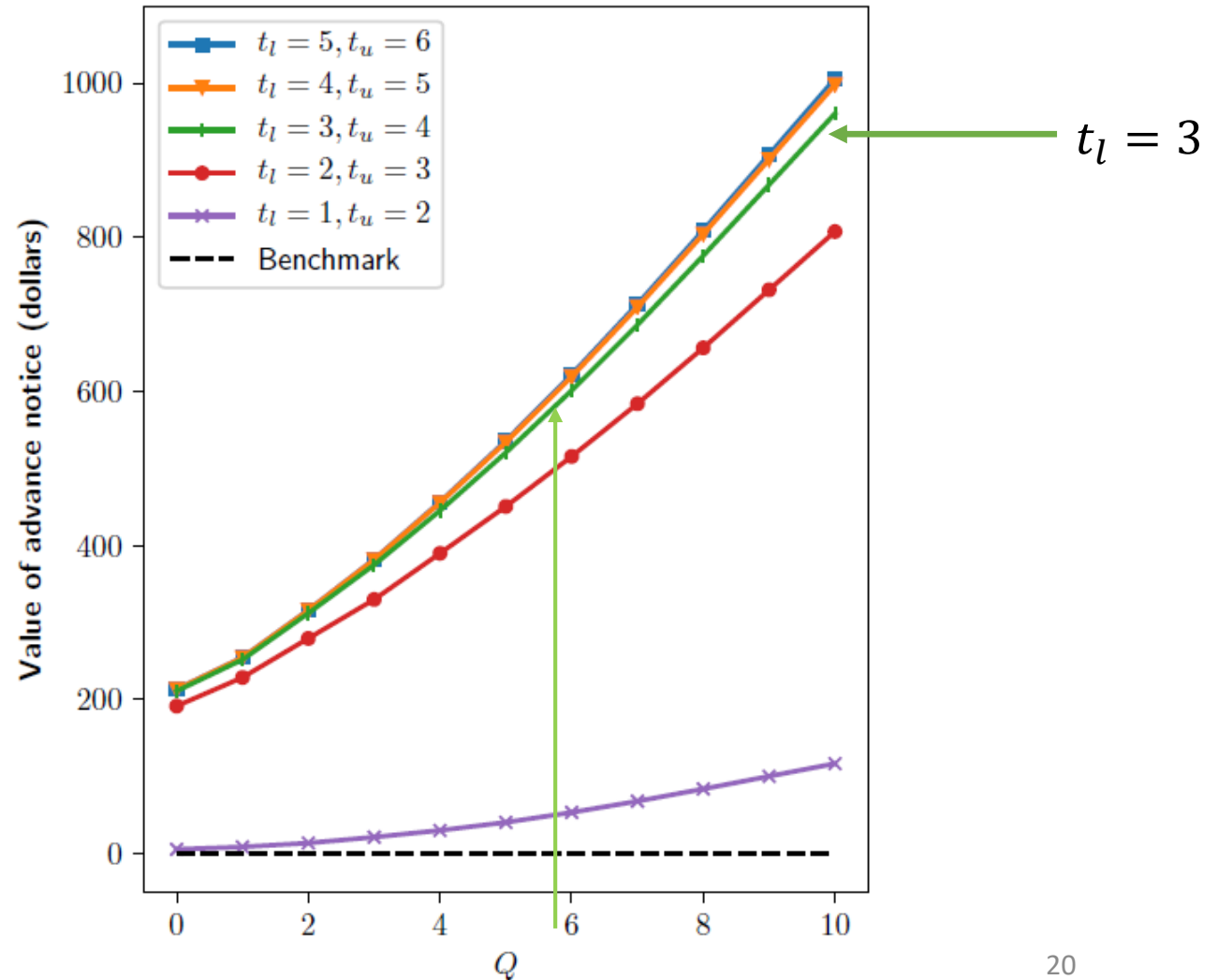


- As  $t_l$  increases, benefits turn more significant.
  - More service flexibility for provider with larger  $t_l$ .
  - Provable result.
- Marginal benefit of increasing  $t_l$  is diminishing.
  - Larger  $t_l$  requires longer IP standby.
  - Enables providers to use a small  $t_l$  to achieve good improvement.



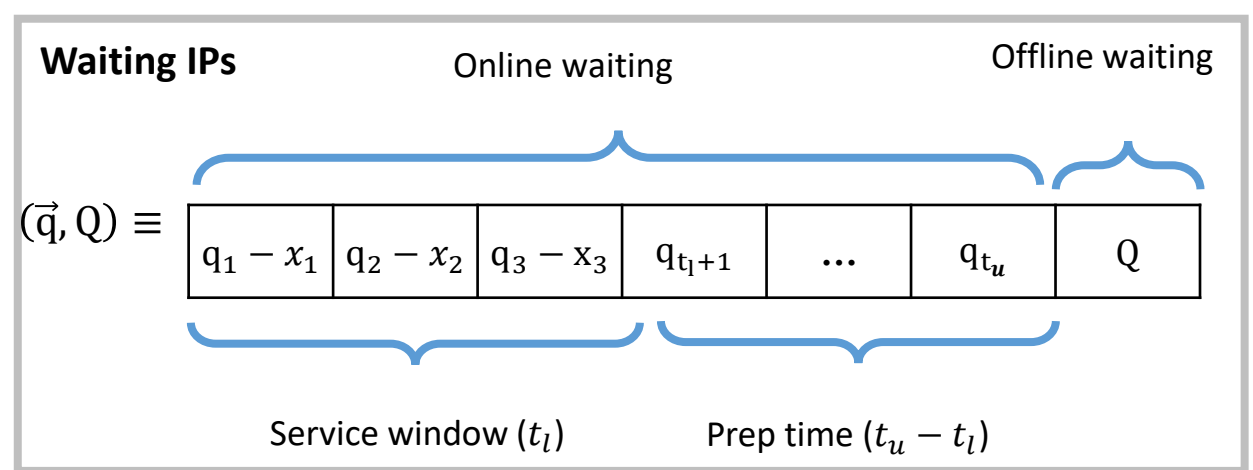
# How much improvement?

- With a little flexibility in service time window (e.g.,  $t_l = 3$  or equivalently 1.5 hours), the financial improvement can be **\$211-961** per day or equivalently **\$55,000-\$250,000** per year for  $M=4$  scanners.
- With  $Q = 5 \sim 6$ , annual cost savings amount to **\$135,000-\$156,000**.
- The annual total expected discounted reward under the current benchmark policy amounts to **\$1.05M-\$1.12M**.
- Improvement @ **13-14%**.

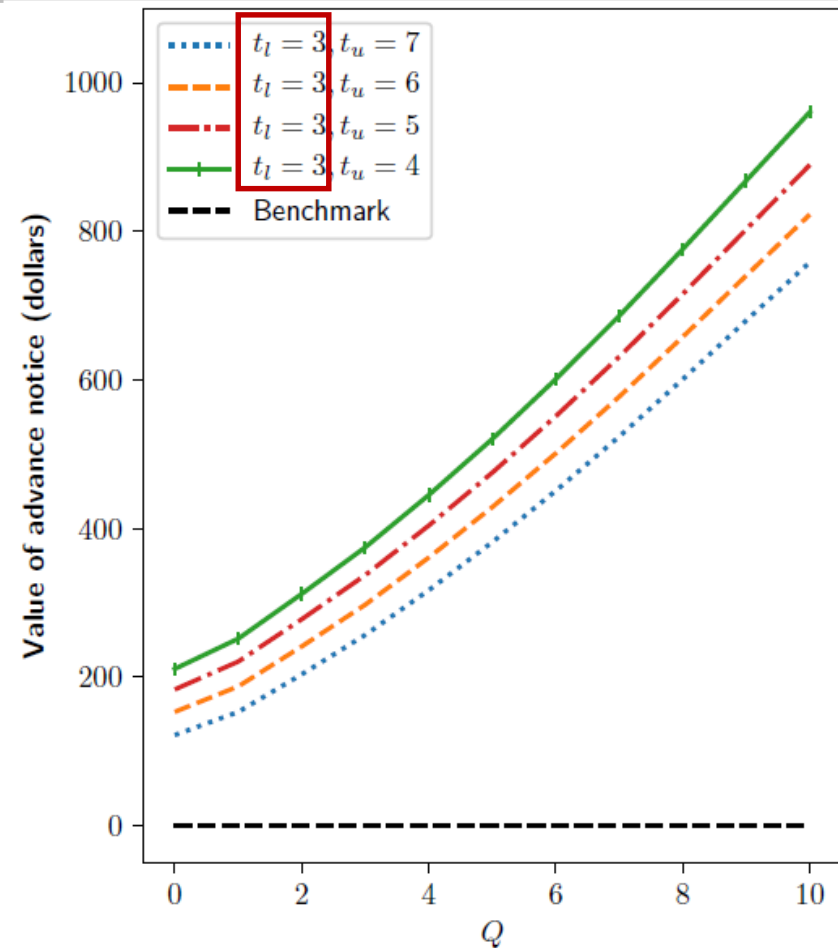


# Impact of Prep Time

$$t_u - t_l$$



- Fixing the service window  $t_l$ , the value function drops as lead time  $t_u - t_l$  increases.
  - Neither patients or providers benefit from long notice.
- Should not rush to send advance notice.
- Reducing lead time (e.g., transport time) can be an improvement lever.



# Takeaways

- We propose a new scheduling paradigm: *advance notice*.
  - Strike a fine balance between advance and allocation scheduling.
- Formulate an MDP model and derive structural properties.
  - Reveal its hidden antimultimodular structure via variable transformations.
  - Prove the monotonicity and bounded sensitivity of optimal decisions.
- Demonstrate significant values of advance notice in a Mayo Clinic case.
  - A little flexibility (in the service window) can go a long way.
  - Reduce prep time is a useful improvement lever.
- Advance notice offers a promising approach to improving general appointment-based services, e.g., maintenance, tech support, and online delivery.

Thank you!

Q & A

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Working paper available @

