

# Optimal pricing in a Single Server Queue

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Joint work with

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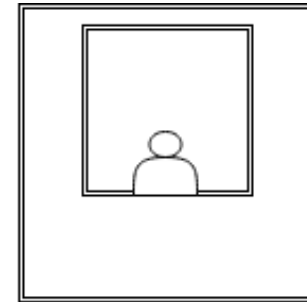
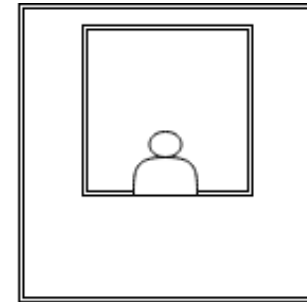
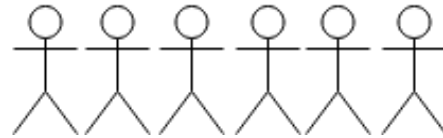
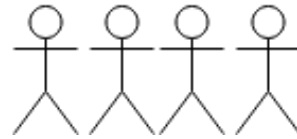
<sup>1</sup> IISc Bangalore <sup>2</sup> Georgia Tech



Source:  
Alexandre Duret-Lutz  
[https://www.flickr.com/p  
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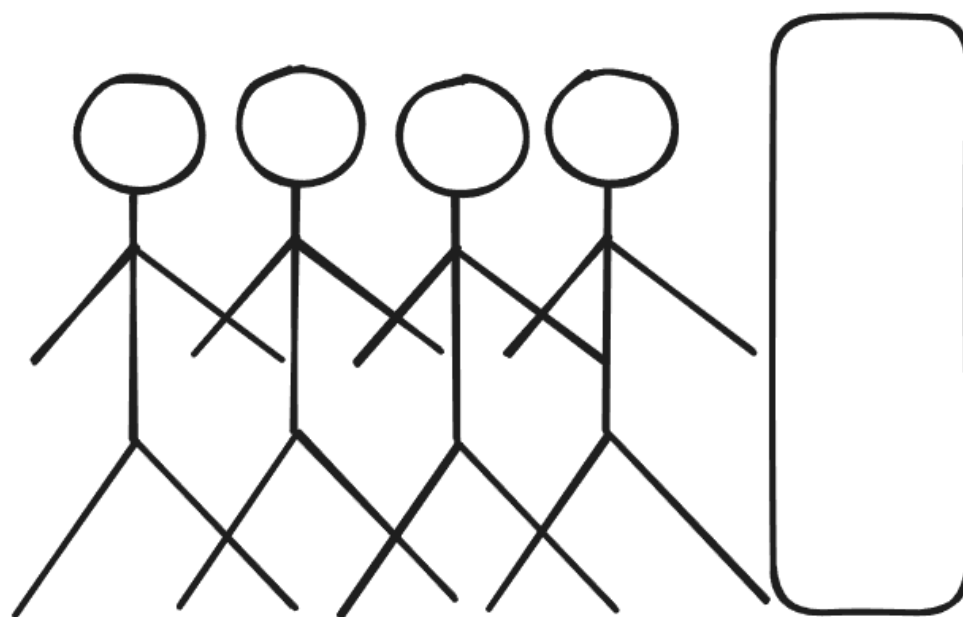
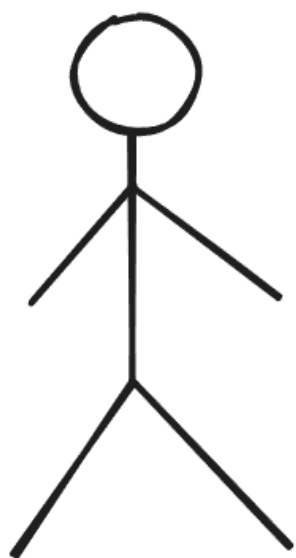
# To Join or Not to Join

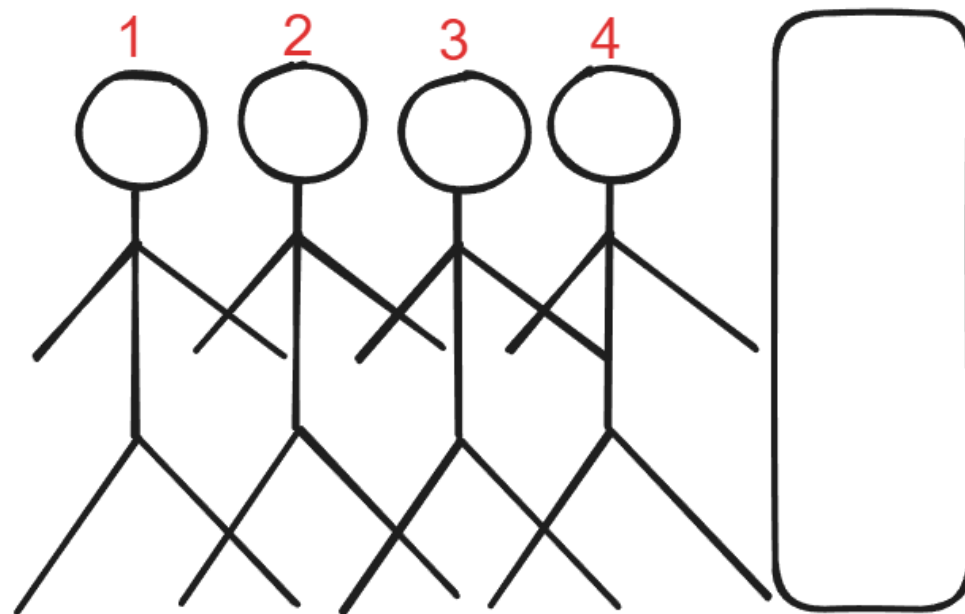
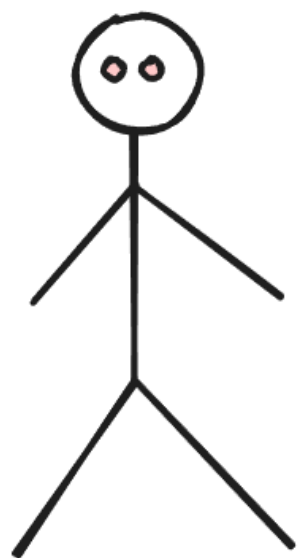
- Example

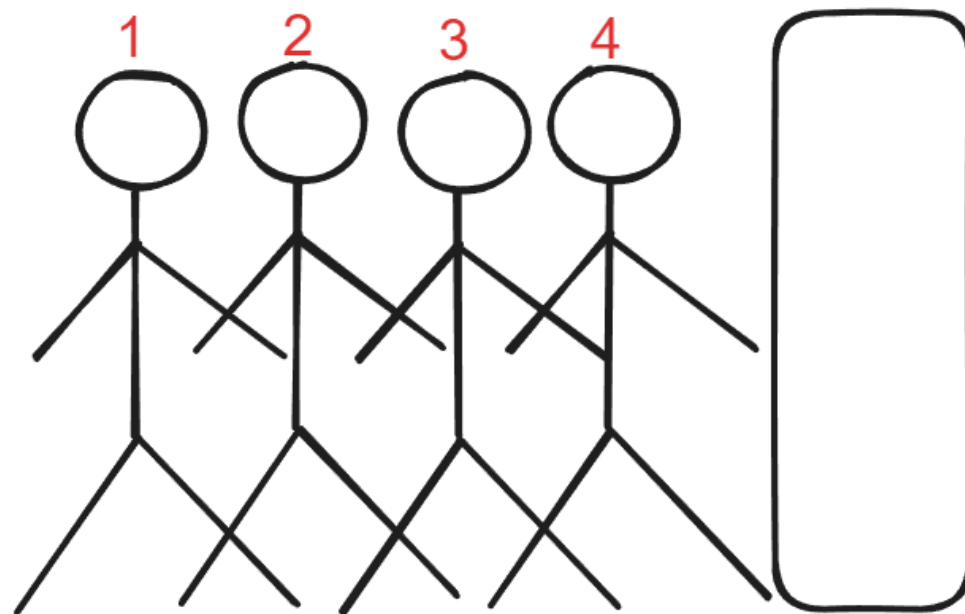
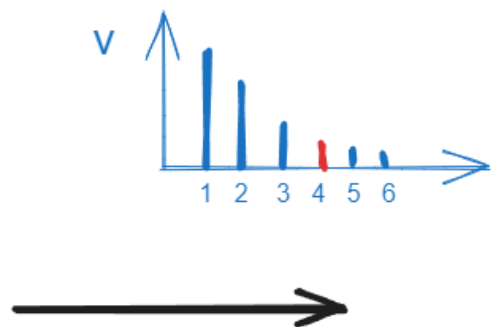
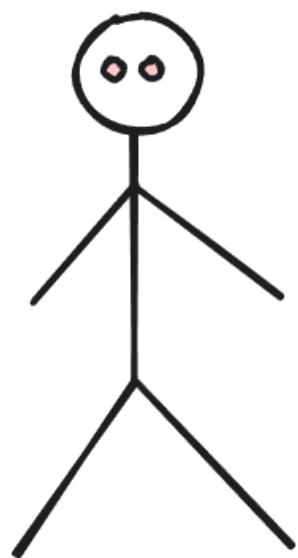


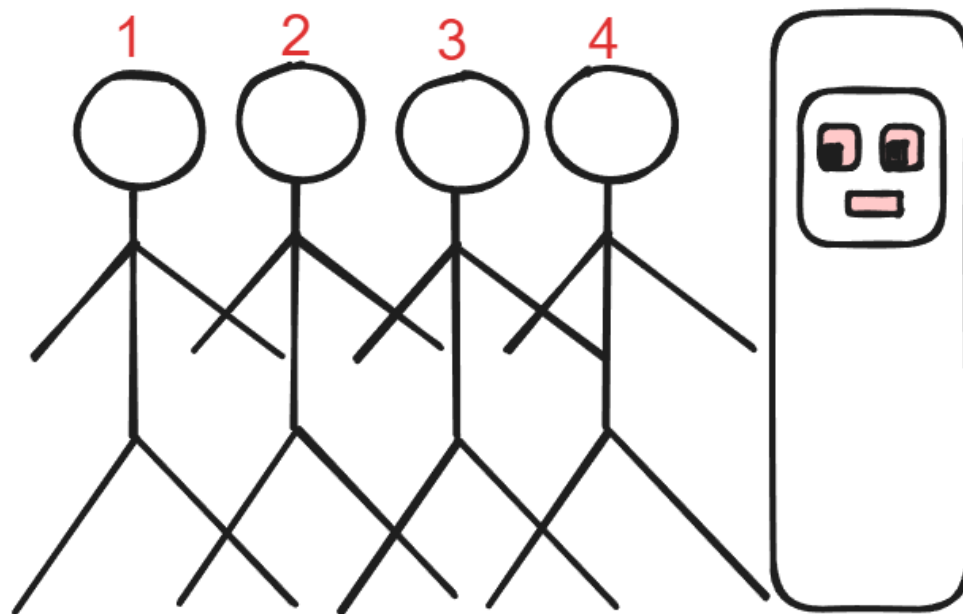
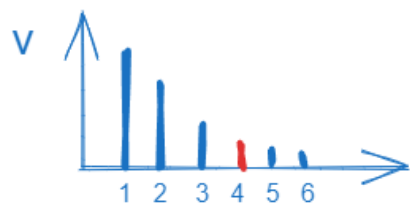
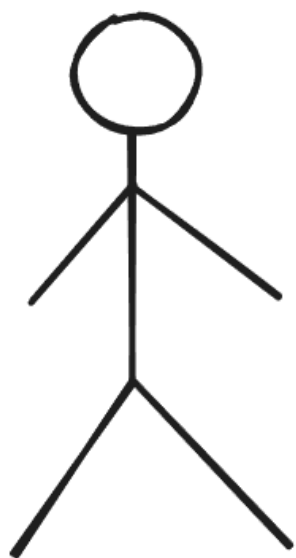
# Examples

- Cloud Servers
- Uber
- Movies
- Restaurants
- And more...

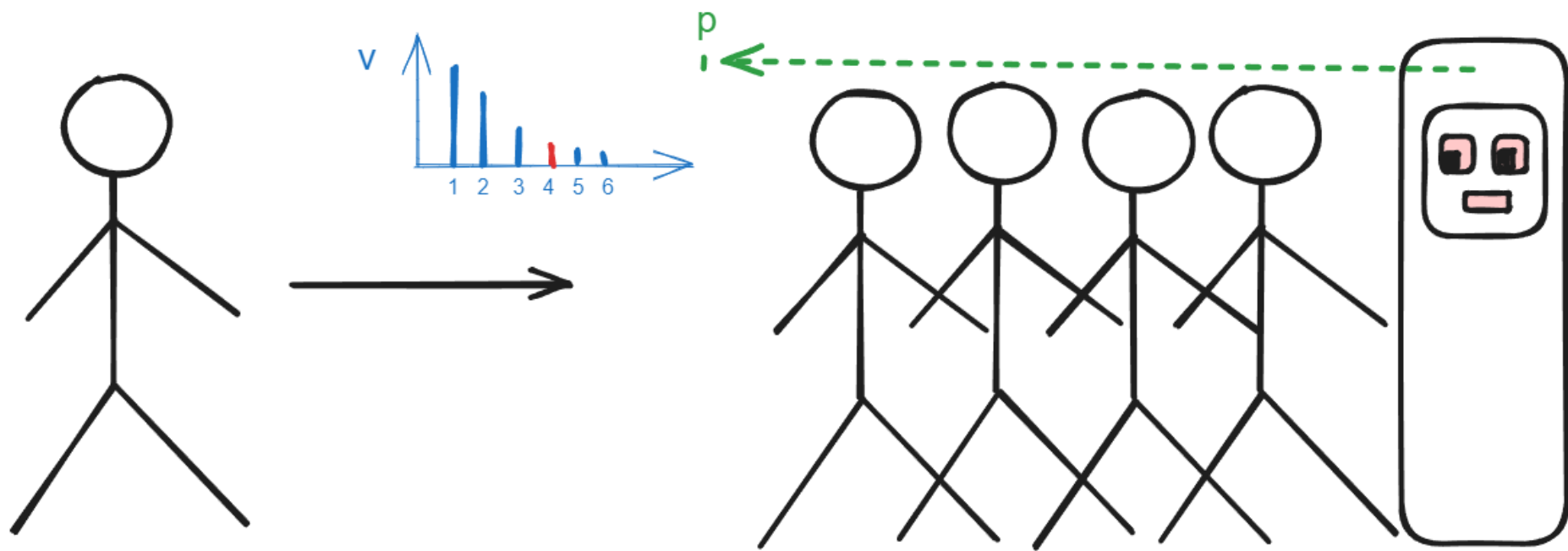


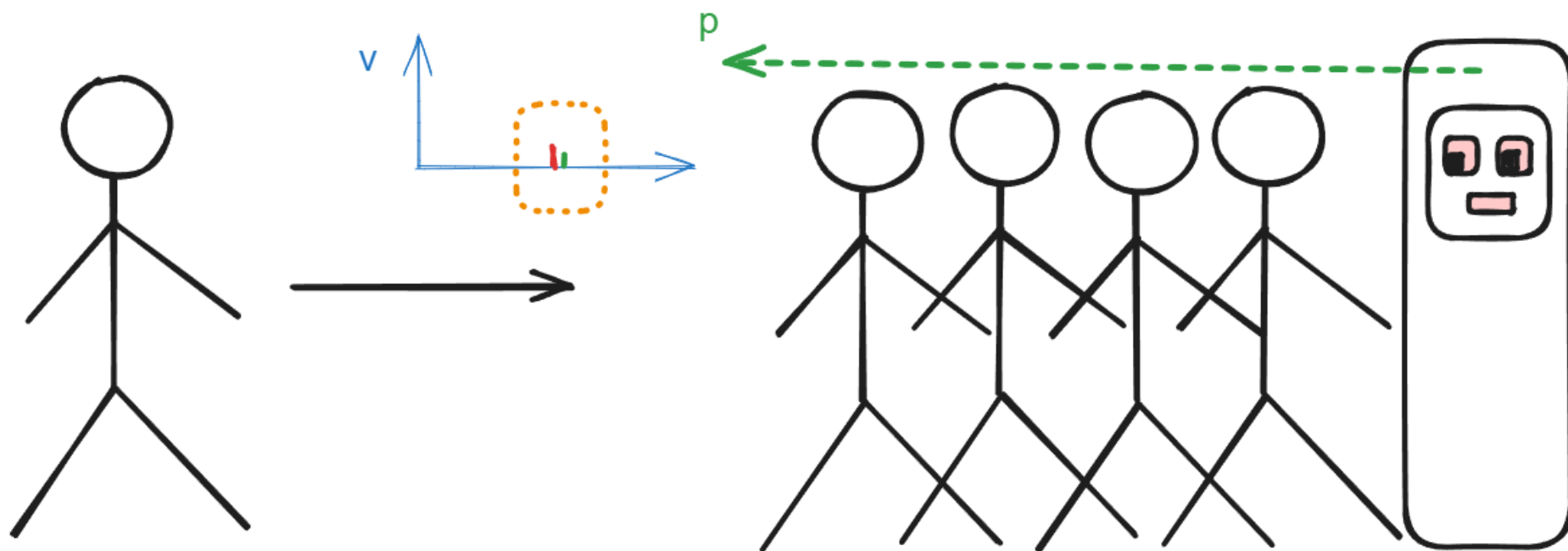


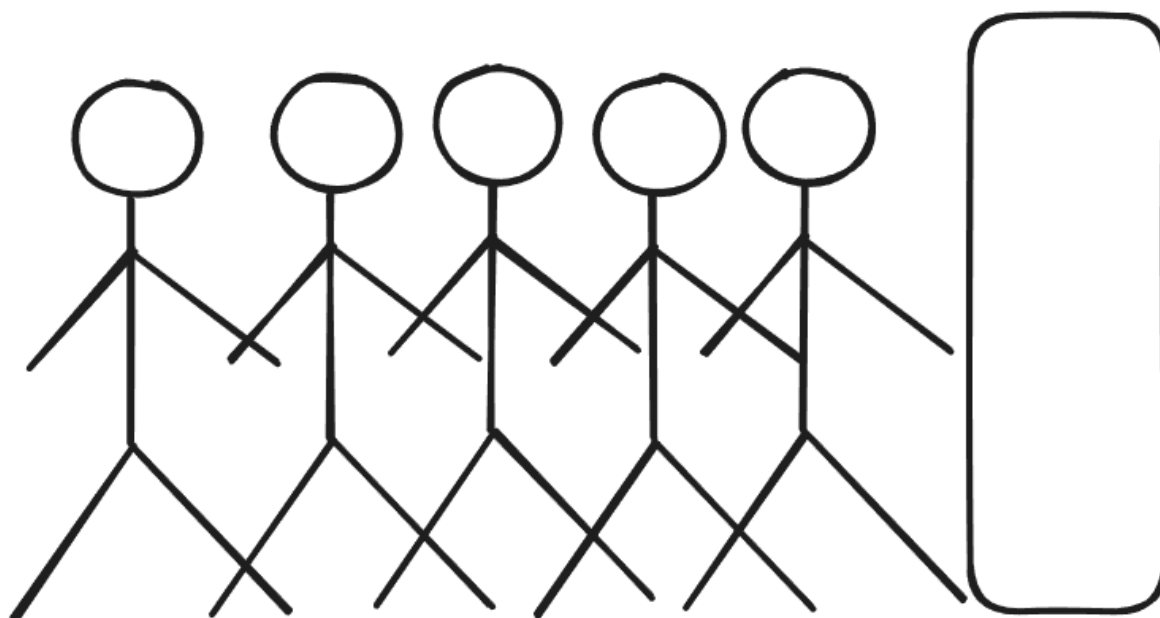
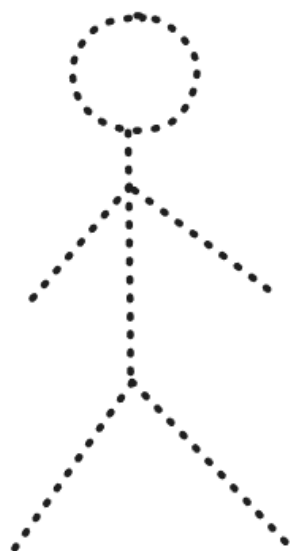




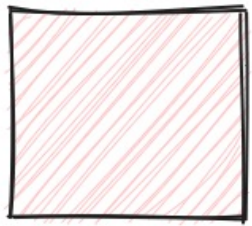




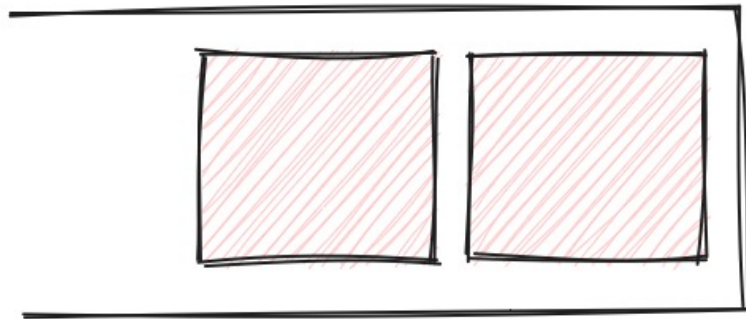




# Model



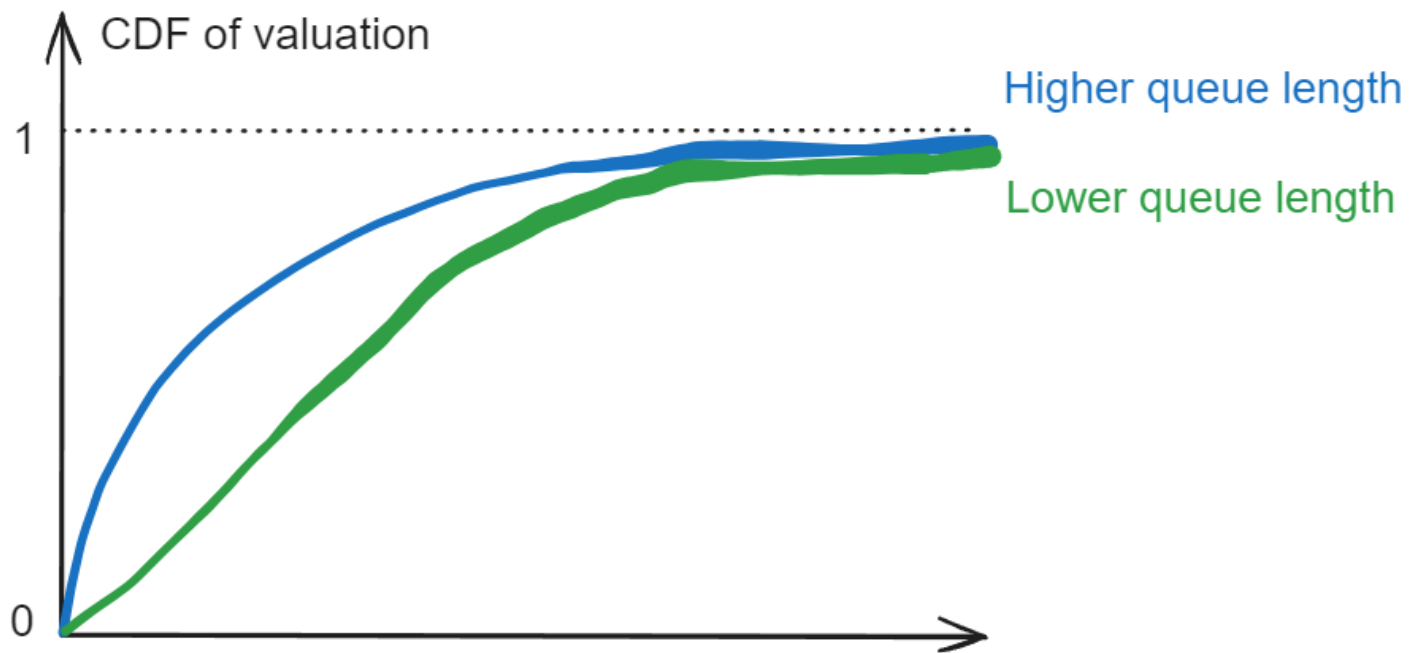
Poisson  
 $\lambda$

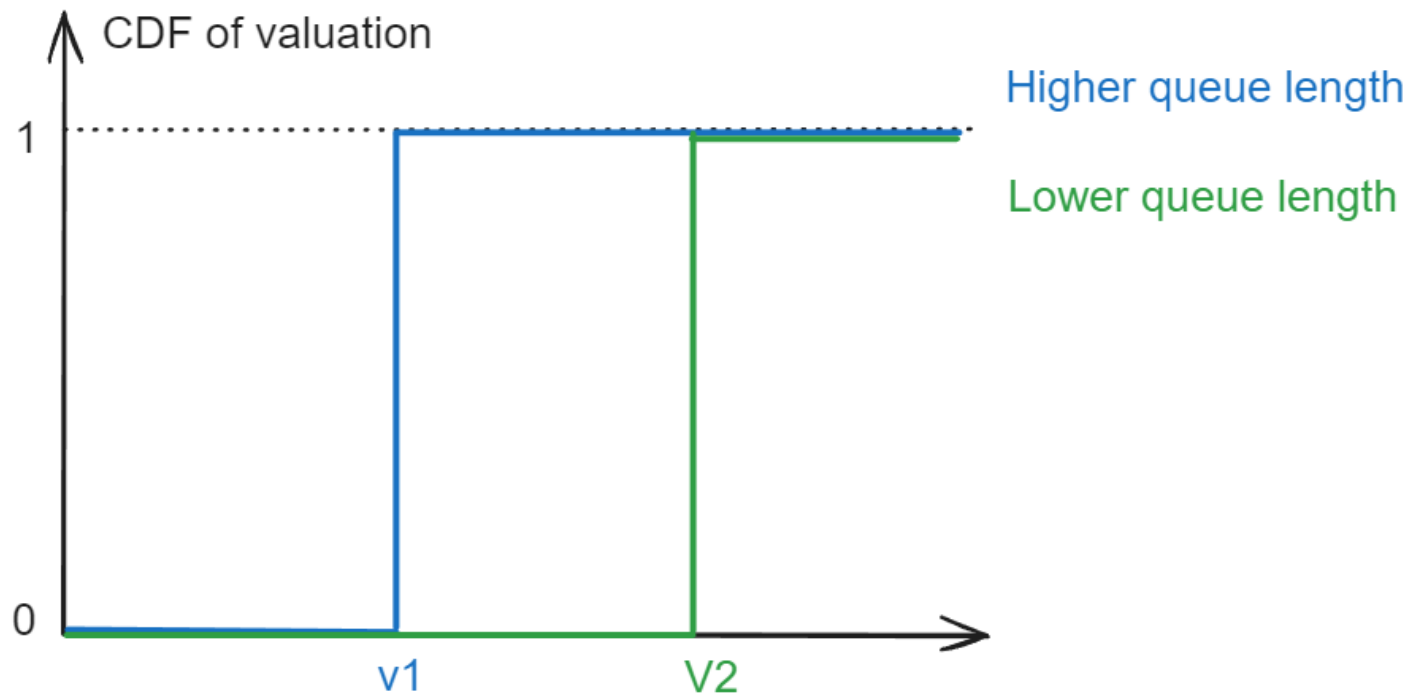


Exponential  
 $\mu$

# Model

- Random valuation for each user
- Valuation depends on queue length
- Valuation decreases stochastically as queue length increases





# Optimization

- Choose admission prices at each queue length
- Maximize long run average revenue
- A simple solution: Myopic policy

*Myopic Policy: For each queue length, charge the price that is optimal for its valuation.*

## Theorem[KSMP]

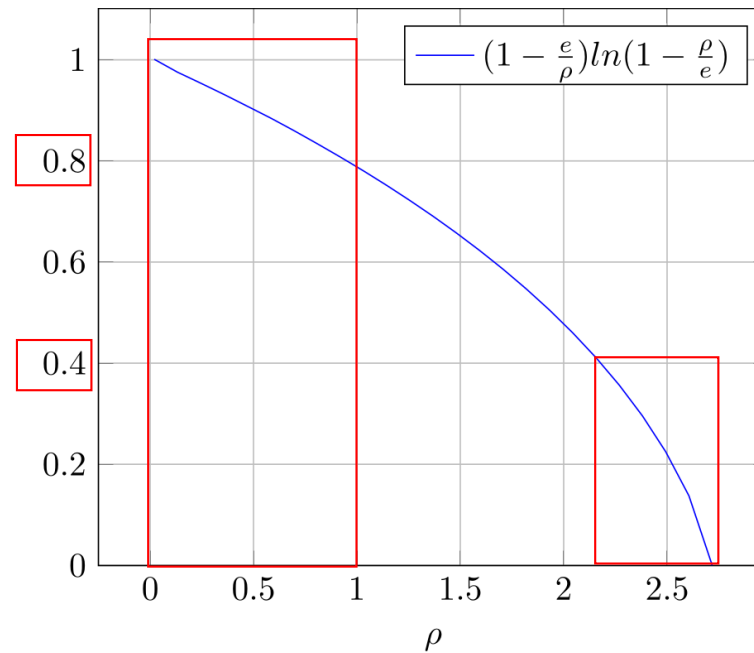
The myopic policy is guaranteed to provide at least a fixed fraction of the optimal revenue.



# Myopic Policy

- Can perform well when variation of valuation with load is small
- Performs badly at high loads in some cases

Ratio of  
Myopic revenue  
To Optimal  
Revenue for  
Exponential  
valuation



## Queue without waiting cost



- All packets have same valuation distribution  $G$
- Price  $p$  independent of queue length
- Arrival joins if valuation greater than price, i.e, with probability  $1 - G(p)$
- Revenue equals  $\max_p \lambda p(1 - G(p))$

## Queue **with** waiting cost



- Valuation  $G_i$  will depend on queue length  $i$
- Modelled as MDP
- Obtain Bellman's equations

$$m_i(\Delta_i) = \frac{\theta - \mu\Delta_{i-1}}{\lambda}, \quad i = 0, 1, 2, 3, \dots$$

$$\Delta_{-1} = 0$$

- Congestion loss  $\Delta_i$  increasing in queue length  $i$
- Optimal prices correspond to  $\max_p \lambda(p - \Delta(i))(1 - G(p))$

# Solving the equations: Outline

- Truncate to  $K$  equations
- Solve to get optimal price vector  $u(K)$

Theorem[KSMP]

As  $K$  becomes larger,  $\text{Revenue}(u(K)) \rightarrow \text{optimal Revenue}$

Theorem[KSMP]

As  $K$  becomes larger,  $u(K) \rightarrow \text{optimal price vector}^*$

\*conditions apply

## Solving the equations: Outline

- The  $K$  equations yield a scalar fixed point equation in the optimal revenue rate  $\theta$

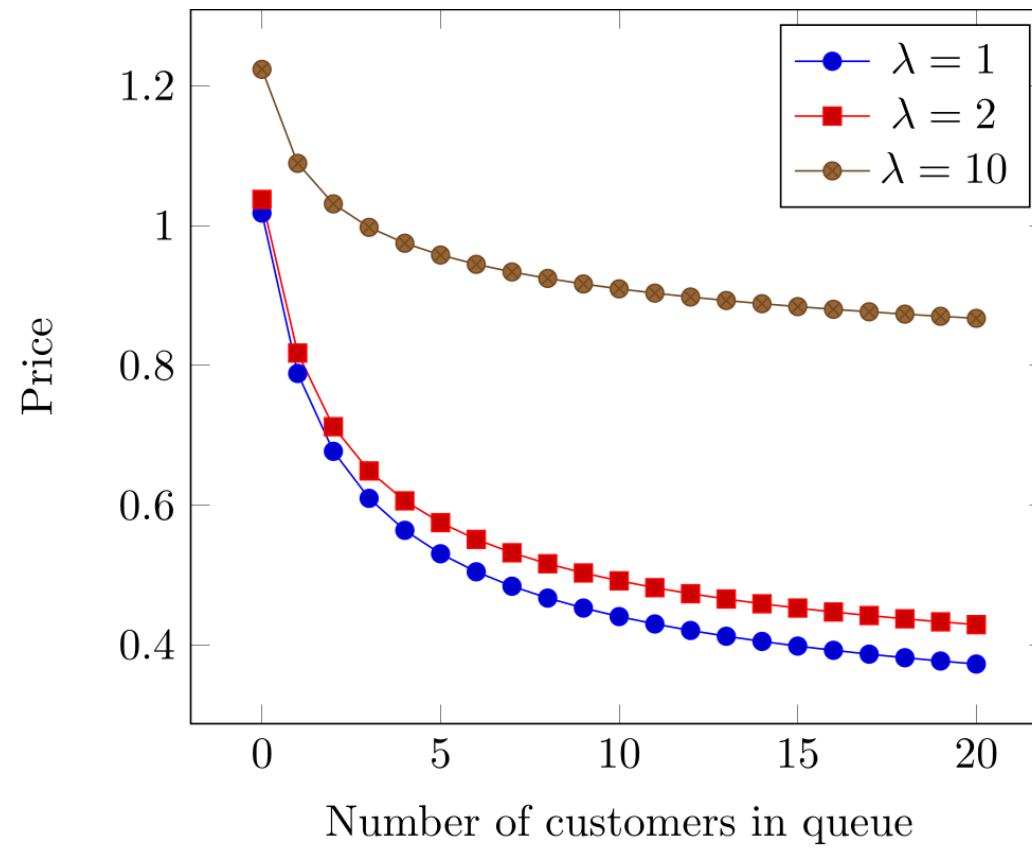
$$\theta = \psi(\theta)$$

- Solve this iteratively and obtain the revenue rate  $\theta$
- Plug this back in the equations to obtain the optimal prices
- Iterative algorithm gives optimal revenues and prices

## Special Case: Non Random valuations

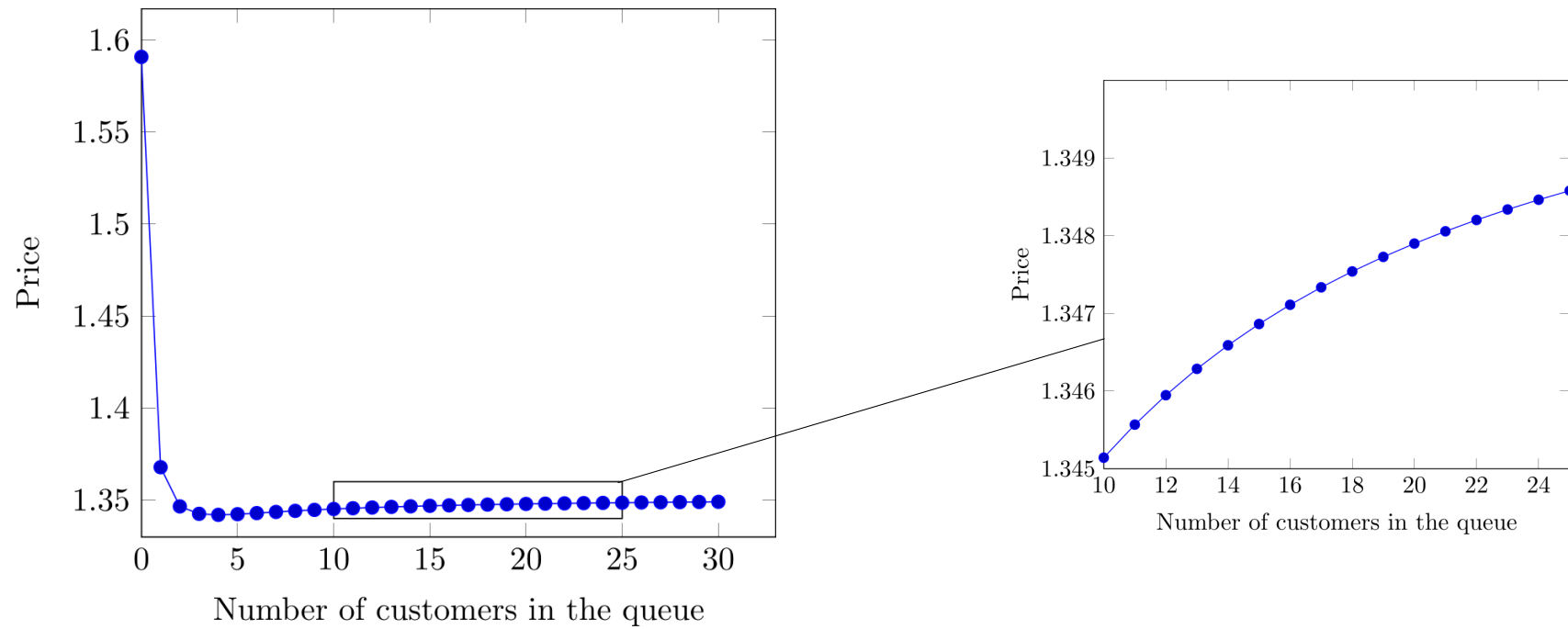
- Valuation is non random: when queue length is  $i$ , valuation is  $v_i$
- $v_i \rightarrow v$
- Case 1: There some queue length beyond which no one is accepted
  - Finite system of equations
- Case 2: All are accepted always
  - Infinite system of equations

# Numerical Results



Valuation  $G(u) = 1 - \exp(-\log(i + e) u)$

# Numerical Results

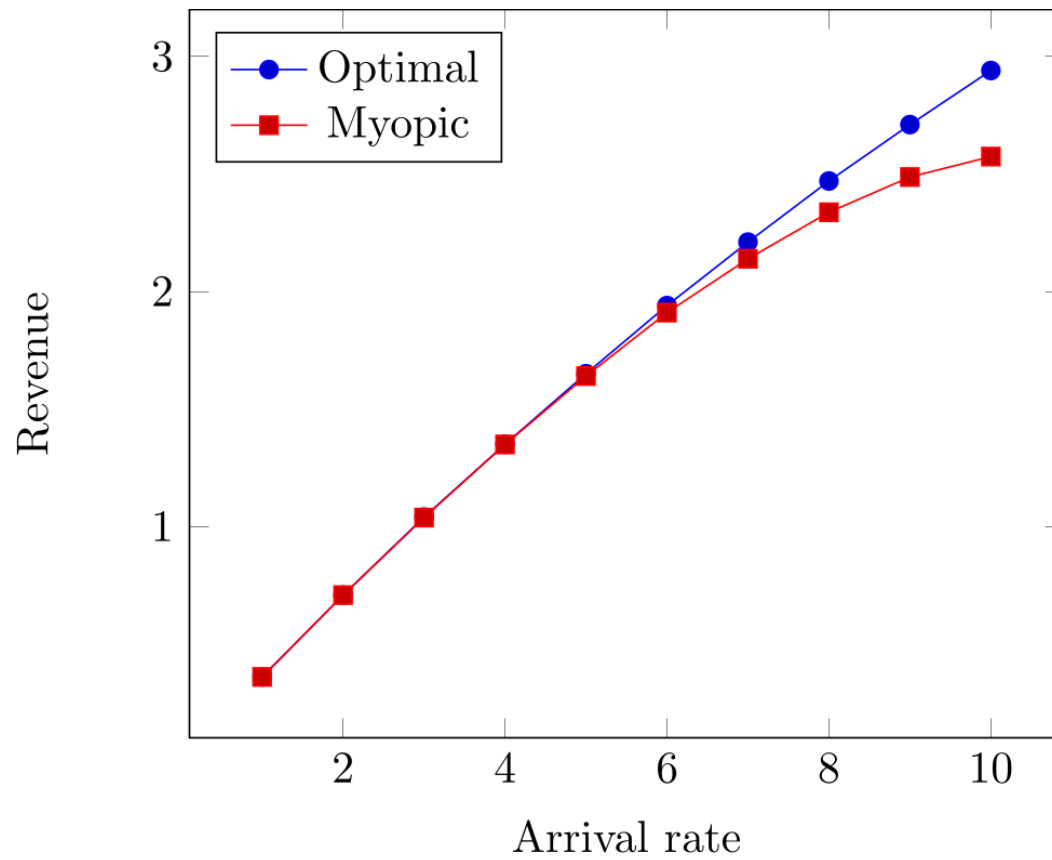


Not decreasing!

$$\text{Valuation } G(u) = 1 - \exp\left(-\left(2 - \frac{1}{1+i}\right)u\right)$$



# Numerical Results



Valuation  $G(u) = 1 - \exp(-\log(i + e) u)$