# A 3-player game theoretic model of a choice between two queueing systems with strategic managerial decision making

Michalis Panayides



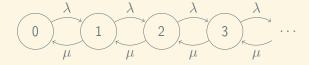
THIS.

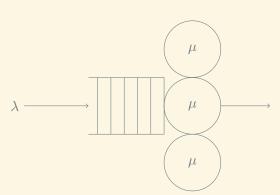
#### Supervisors:

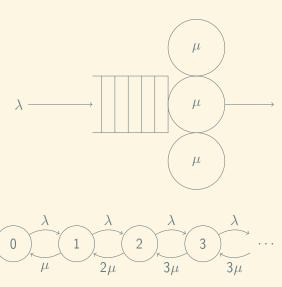
Dr. Vince Knight, Prof. Paul Harper







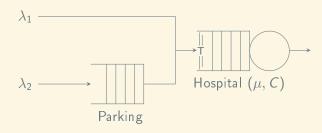




# Queues - Custom network of queues



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#### Parameters:

 $\blacktriangleright$   $\lambda_1$ : Arrival rate of type 1 individuals

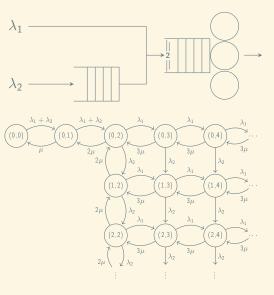
 $\triangleright$   $\lambda_2$ : Arrival rate of type 2 individuals

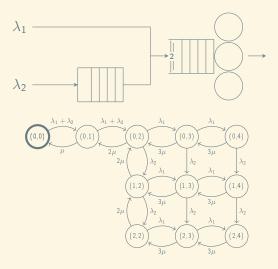
 $\blacktriangleright$   $\mu$ : Service rate

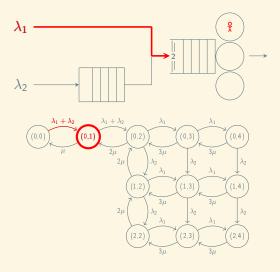
C: Number of servers

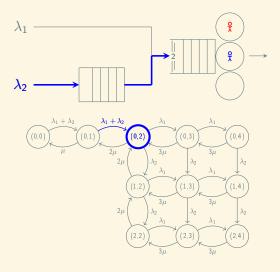
► T: Threshold

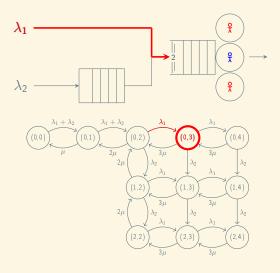
#### Markov Chain - Custom network

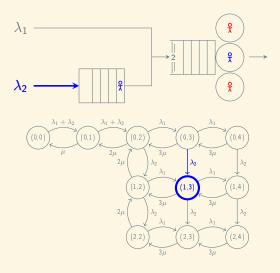


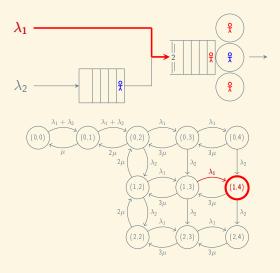


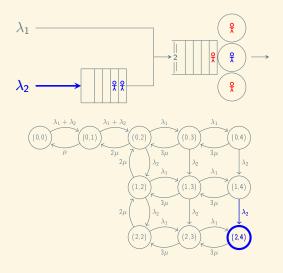


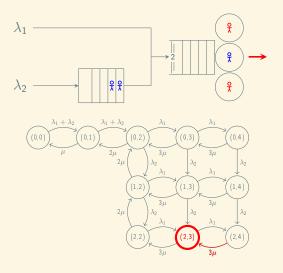


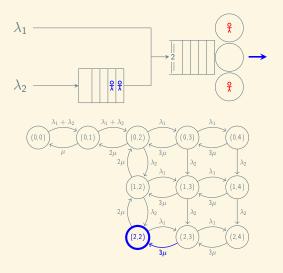


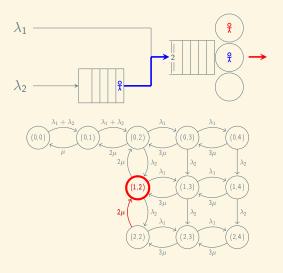




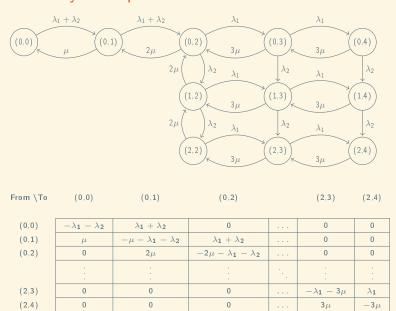








# Steady state probabilities - Generator matrix



# Steady state probabilities - Generator matrix (Q)

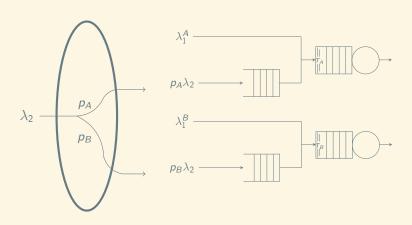
$$\pi = \begin{bmatrix} \pi_{(0,0)} & \pi_{(0,1)} & \pi_{(0,2)} & \dots & \pi_{(2,3)} & \pi_{(2,4)} \end{bmatrix}, \qquad \sum \pi_{(u,v)} = 1$$

 $\frac{d\pi}{dt} = \pi Q = 0$ 

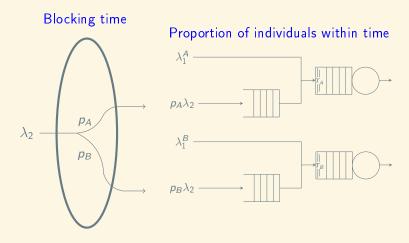
# Game - Definition



# Game - Players and objectives



# Game - Players and objectives



# Performance Measures - Blocking time

$$B = \frac{\sum_{(u,v) \in S_A^{(2)}} \pi_{(u,v)} \ b(u,v)}{\sum_{(u,v) \in S_A^{(2)}} \pi_{(u,v)}}$$

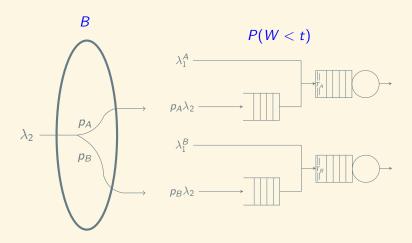
# Performance Measures - Proportion within time

$$P(W < t) = \frac{\lambda_1 P_{L_1'}}{\lambda_2 P_{L_2'} + \lambda_1 P_{L_1'}} P(W^{(1)} < t) + \frac{\lambda_2 P_{L_2'}}{\lambda_2 P_{L_2'} + \lambda_1 P_{L_1'}} P(W^{(2)} < t)$$

$$P(W^{(1)} < t) = \frac{\sum_{(u,v) \in S_A^{(1)}} P(W_{(u,v)}^{(1)} < t) \pi_{u,v}}{\sum_{(u,v) \in S_A^{(1)}} \pi_{u,v}}$$

$$P(W^{(2)} < t) = \frac{\sum_{(u,v) \in S_A^{(2)}} P(W_{(u,v)}^{(2)} < t) \pi_{u,v}}{\sum_{(u,v) \in S_A^{(2)}} \pi_{u,v}}$$

# Game - Players and objectives



# Game - Strategies











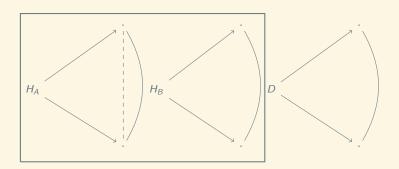


 $p_A, p_B \in [0, 1]$  $p_A + p_B = 1$ 

$$T_A \in [1, N_A]$$

 $T_B \in [1, N_B]$ 

# Game - Formulation



# Game - Payoff matrices

$$A = \begin{pmatrix} U_{1,1}^A & U_{1,2}^A & \dots & U_{1,N_B}^A \\ U_{2,1}^A & U_{2,2}^A & \dots & U_{2,N_B}^A \\ \vdots & \vdots & \ddots & \vdots \\ U_{N_A,1}^A & U_{N_A,2}^A & \dots & U_{N_A,N_B}^A \end{pmatrix}, \quad B = \begin{pmatrix} U_{1,1}^B & U_{1,2}^B & \dots & U_{1,N_B}^B \\ U_{2,1}^B & U_{2,2}^B & \dots & U_{2,N_B}^B \\ \vdots & \vdots & \ddots & \vdots \\ U_{N_A,1}^B & U_{N_A,2}^B & \dots & U_{N_A,N_B}^B \end{pmatrix}$$

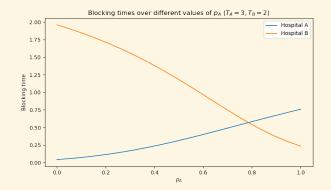
$$R = \begin{pmatrix} p_{1,1} & p_{1,2} & \dots & p_{1,N_B} \\ p_{2,1} & p_{2,2} & \dots & p_{2,N_B} \\ \vdots & \vdots & \ddots & \vdots \\ p_{N_A,1} & p_{N_A,2} & \dots & p_{N_A,N_B} \end{pmatrix}$$

#### Ambulance's Decision

$$R = \begin{pmatrix} p_{1,1} & p_{1,2} & \dots & p_{1,N_B} \\ p_{2,1} & p_{2,2} & \dots & p_{2,N_B} \\ \vdots & \vdots & \ddots & \vdots \\ p_{N_A,1} & p_{N_A,2} & \dots & p_{N_A,N_B} \end{pmatrix}$$

#### Ambulance's Decision

$$T_A = 3$$
 $T_B = 2$ 
 $\rightarrow \begin{pmatrix} - & - & - & - \\ - & - & - & - \\ - & x & - & - \\ - & - & - & - \end{pmatrix}$ 
 $\rightarrow B_A(p_A) = B_B(1 - p_A)$ 



# Hospitals' Decision

$$A = \begin{pmatrix} U_{1,1}^A & U_{1,2}^A & \dots & U_{1,N_B}^A \\ U_{2,1}^A & U_{2,2}^A & \dots & U_{2,N_B}^A \\ \vdots & \vdots & \ddots & \vdots \\ U_{N_A,1}^A & U_{N_A,2}^A & \dots & U_{N_A,N_B}^A \end{pmatrix}, \quad B = \begin{pmatrix} U_{1,1}^B & U_{1,2}^B & \dots & U_{1,N_B}^B \\ U_{2,1}^B & U_{2,2}^B & \dots & U_{2,N_B}^B \\ \vdots & \vdots & \ddots & \vdots \\ U_{N_A,1}^B & U_{N_A,2}^B & \dots & U_{N_A,N_B}^B \end{pmatrix}$$

$$U_{T_A, T_B}^{(i)} = 1 - \left[ (P(X^{(i)} < t) - 0.95)^2 \right]$$

#### Nash Equilibrium

$$R = \begin{pmatrix} 0.5 & 0.1 & 0 & 0 \\ 0.9 & 0.5 & 0.2 & 0 \\ 1 & 0.8 & 0.5 & 0.3 \\ 1 & 1 & 0.7 & 0.5 \end{pmatrix}$$

$$A = \begin{pmatrix} 8.39 & 8.39 & 8.39 & 8.39 \\ 8.96 & 8.85 & 8.65 & 8.45 \\ 9.95 & 9.87 & 9.6 & 9.2 \\ 4.37 & 5.11 & 8.6 & 9.91 \end{pmatrix}$$

$$B = \begin{pmatrix} 8.39 & 8.96 & 9.95 & 4.37 \\ 8.39 & 8.85 & 9.87 & 5.11 \\ 8.39 & 8.65 & 9.6 & 8.6 \\ 8.39 & 8.45 & 9.2 & 9.91 \end{pmatrix}$$

Nash Equilibria: 
$$(0, 0, 0.4, 0.6)$$
  $(0, 0, 0.4, 0.6)$ 

# Asymmetric Replicator Dynamics

$$\frac{dx}{dt_i} = x_i((f_x)_i - \phi_x), \quad \text{for all } i$$

$$\frac{dy}{dt_i} = y_i((f_y)_i - \phi_y), \quad \text{for all } i$$

# Inefficiency measure

$$PoA = \frac{\max_{s \in E} Cost(s)}{\min_{s \in S} Cost(S)}$$

# Inefficiency measure

$$PoA = \frac{\max_{s \in E} Cost(s)}{\min_{s \in S} Cost(S)}$$

$$PoA_A(s_r) = \frac{Cost(s_r)}{\min_{s \in S} Cost(S)}, \qquad PoA_B(s_c) = \frac{Cost(s_c)}{\min_{s \in S} Cost(S)}$$

# Learning algorithms - Asymmetric replicator dynamics



Inefficiencies can be learned and emerge naturally

# Learning algorithms - Asymmetric replicator dynamics



# Targeted incontinication of

Targeted incentivisation of behaviours can help escape

learned inefficiencies

#### Thank you!

"Inefficiencies can be learned and emerge naturally"

"Targeted incentivisation of behaviours can help escape learned inefficiencies"

\$ pip install ambulance\_game
https://github.com/11michalis11/AmbulanceDecisionGame

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