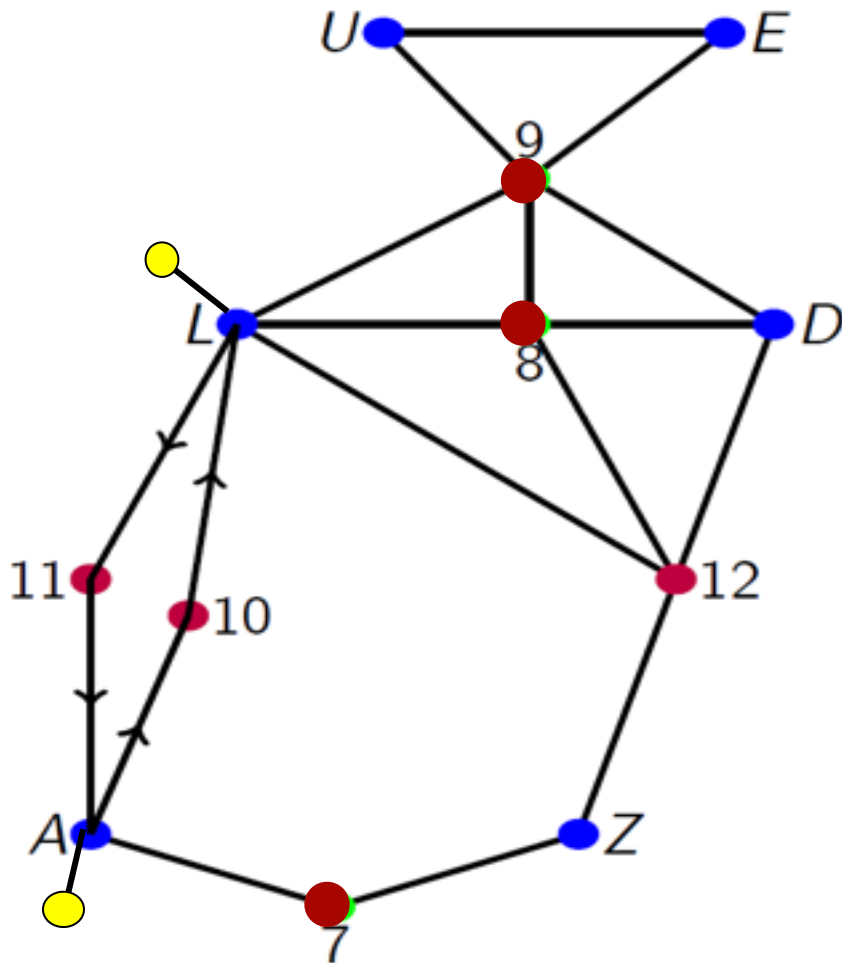


# **Minimizing Average Response Times in a Dynamic Ambulance Management Model**

**Thije van Barneveld, CWI, Amsterdam**

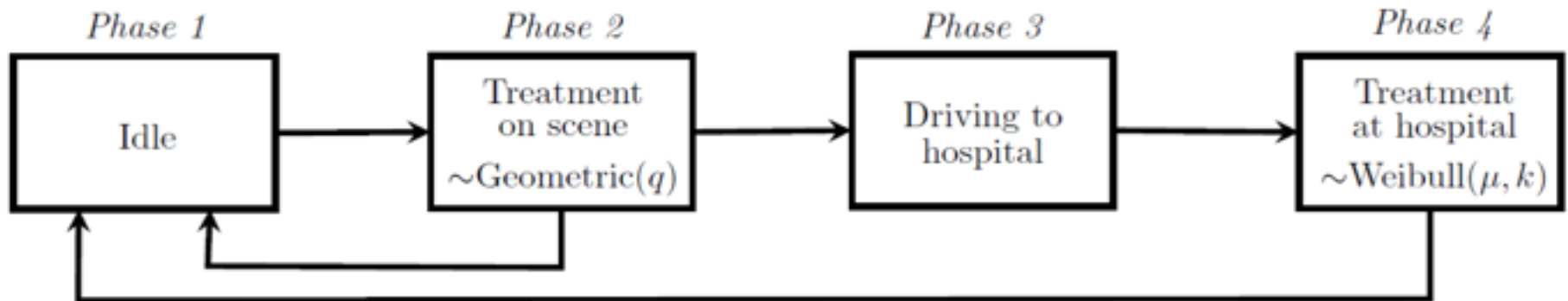


# Region



- **Equidistant graph**
- **Blue: Demand locations**
- **Yellow: Hospitals**
- **Red: Additional nodes**

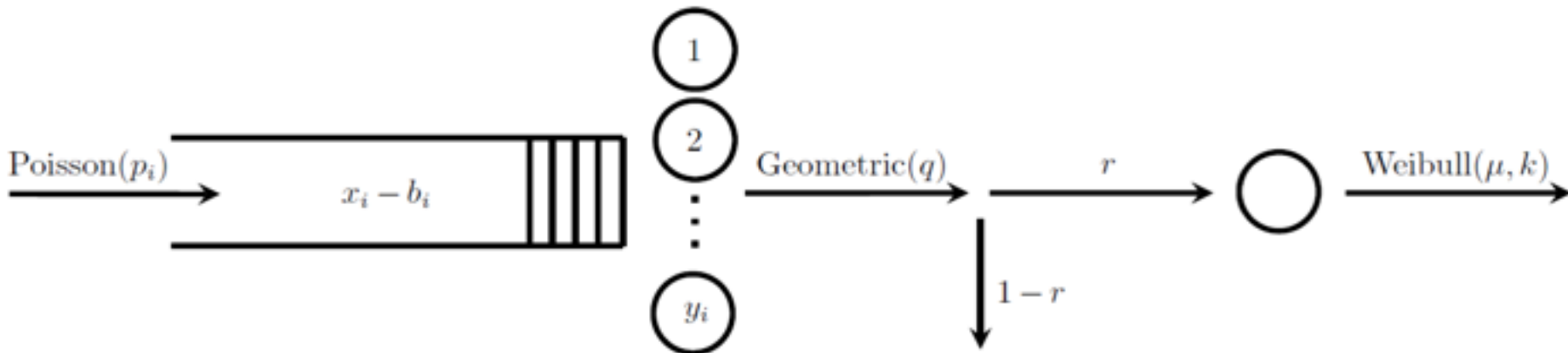
# Ambulance Phases



Different Stages of Ambulances.

- $q$ : Parameter Geometric distribution
- $\mu$ : Scale parameter Weibull distribution
- $k$ : Shape parameter Weibull distribution

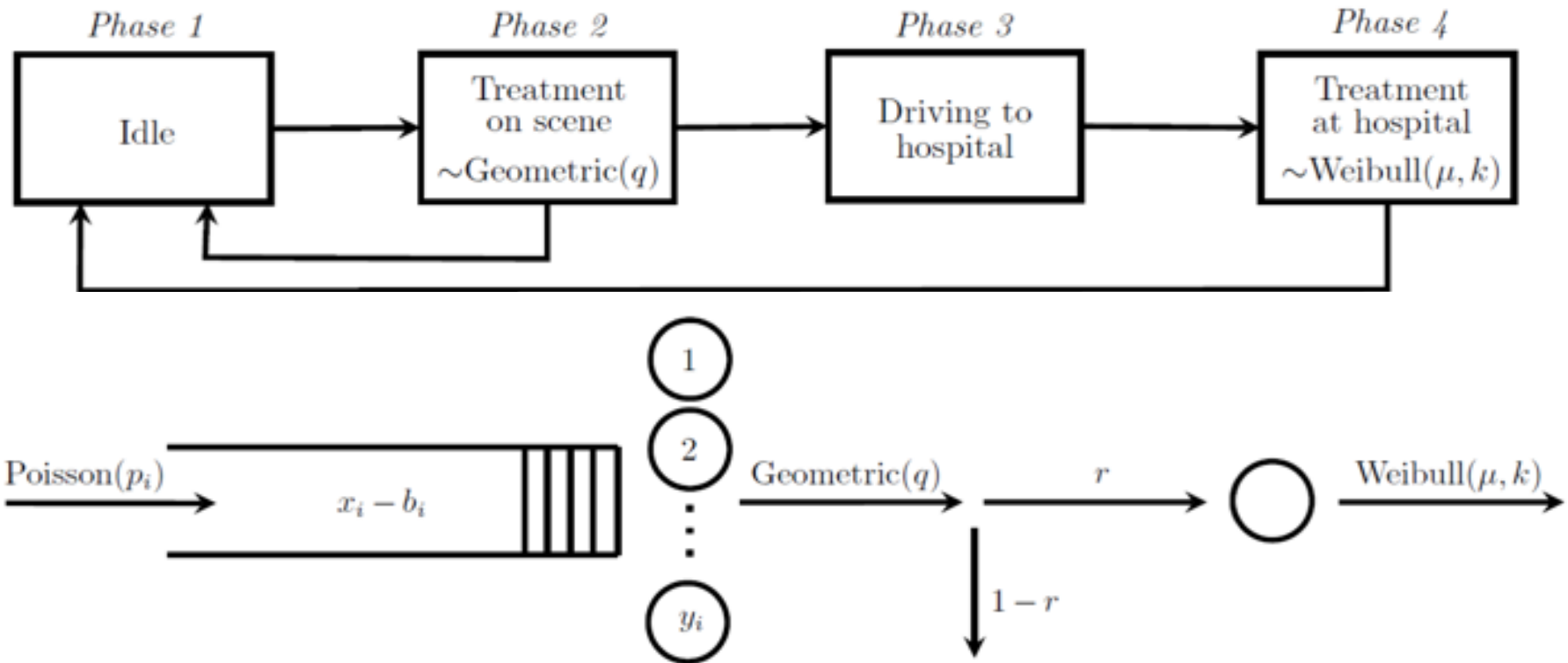
# “Life Cycle” of a Request



“Life Cycle” of a Request arriving at Location  $i$ .

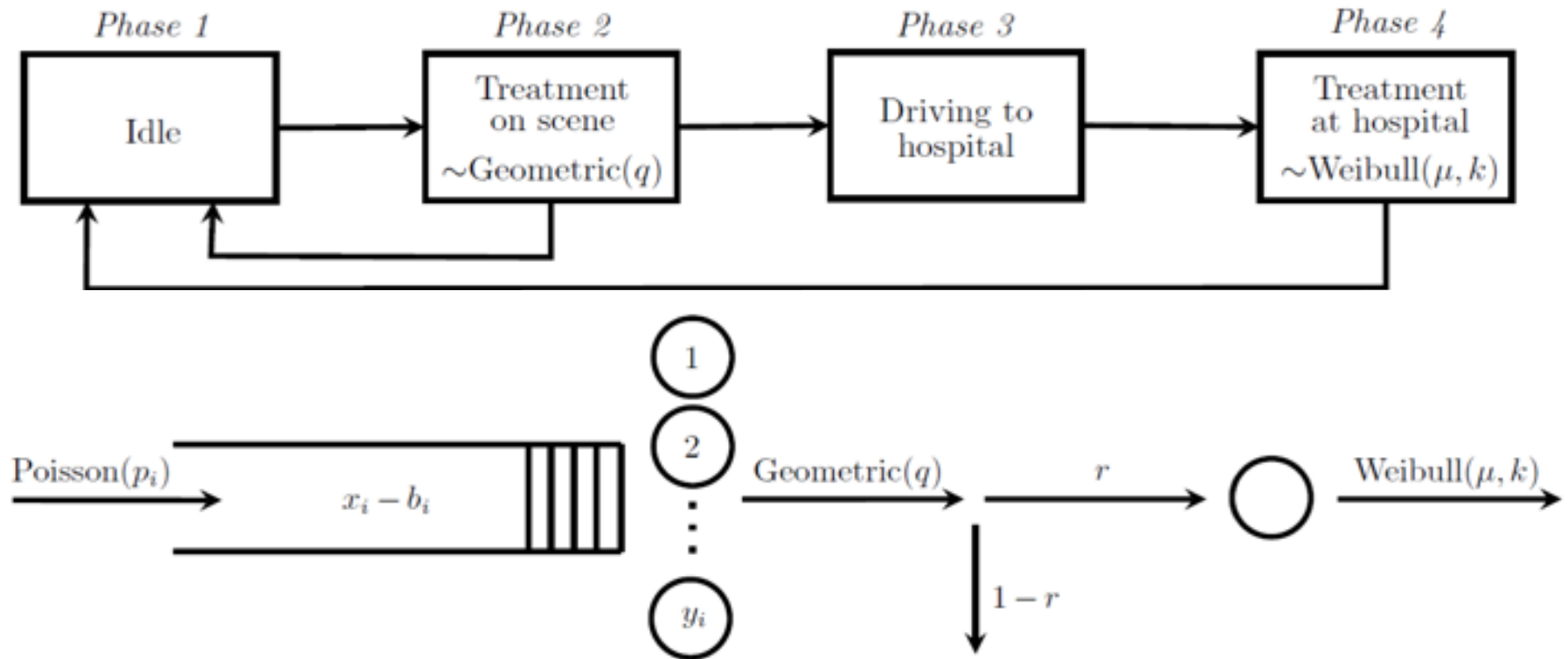
- $p_i$ : Parameter Poisson distribution location  $i$
- $x_i$ : Total number of patients at demand location  $i$
- $b_i$ : Number of busy ambulances at location  $i$
- $y_i$ : Total number of ambulances at location  $i$
- $r$ : Probability to hospital

# State components



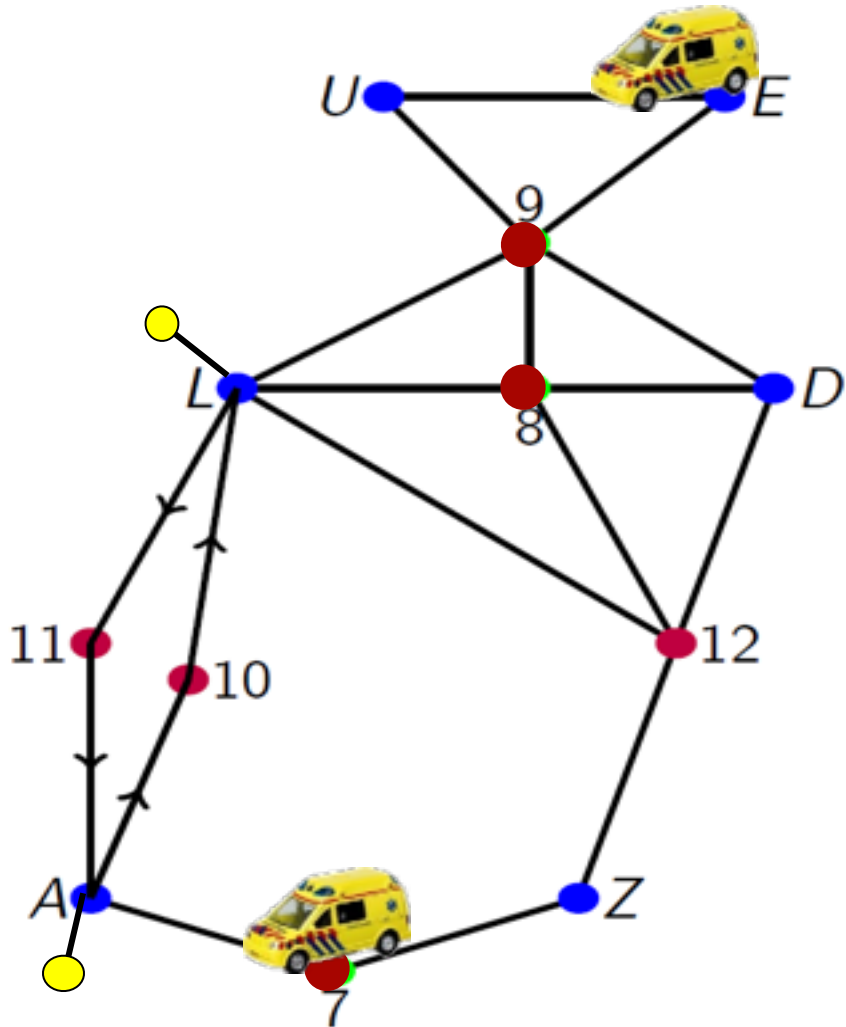
- 1. Number of patients per demand location ( $x_i$ )**
- 2. Number of ambulances either in phase 1, 2, 4 per demand location**
- 3. Number of patients that needs transportation to hospital per demand location**

# State components



4. Elapsed service time of ambulances in phase 4
5. Destinations and remaining driving times of phase 3 ambulances

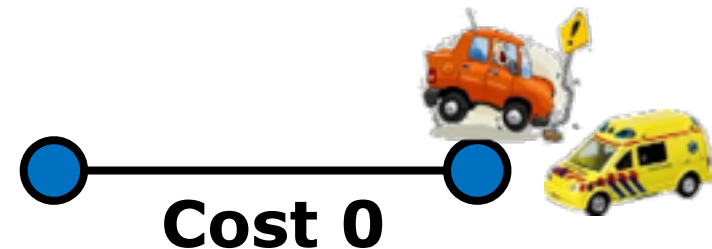
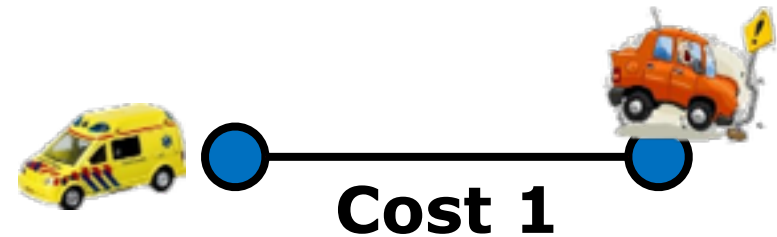
# Actions



- Dispatch nearest ambulance
- Change in ambulance configuration

# Objective

- **Cost in a state: number of patients waiting**
- **Minimize costs: minimize the average number of patients waiting**
- **Minimize the average response time**





# Heuristic Solution - Idea

- **Observe state**
- **Consider all actions**
- **Consider possible scenarios**
- **Combine each action with each scenario**
- **Classify each action and optimize**

# Scenarios

- Possible next state
- One new request
- Ambulances that finish service



OCCUPIED



# Scenarios

- Possible next state
- One new request
- Ambulances that finish service



OCCUPIED



# Scenarios

- Possible next state
- One new request
- Ambulances that finish service



OCCUPIED



# Scenarios

- Possible next state
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# Scenarios

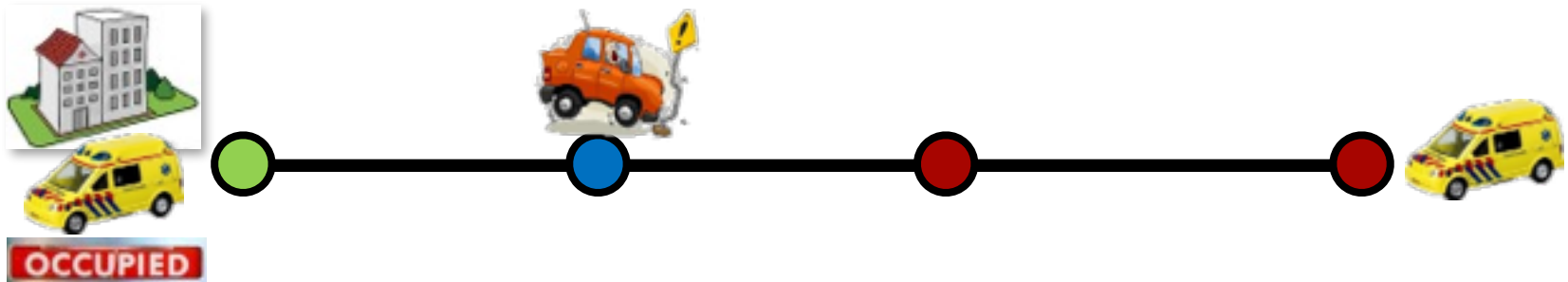
- Possible next state
- One new request
- Ambulances that finish service



# Eligible ambulances

Eligible for responding to new request:

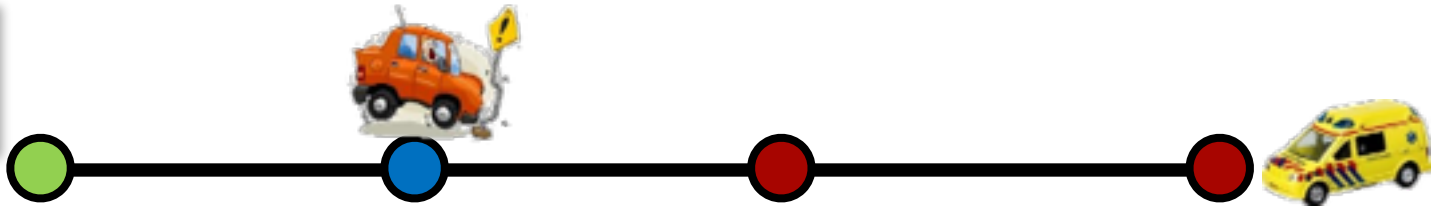
- 1. Nearest idle unassigned ambulance**
- 2. Nearest busy ambulance at hospital**
- 3. Nearest busy ambulance on scene, not required to transport**



**Expected shortest response time**

# CWI

## Example

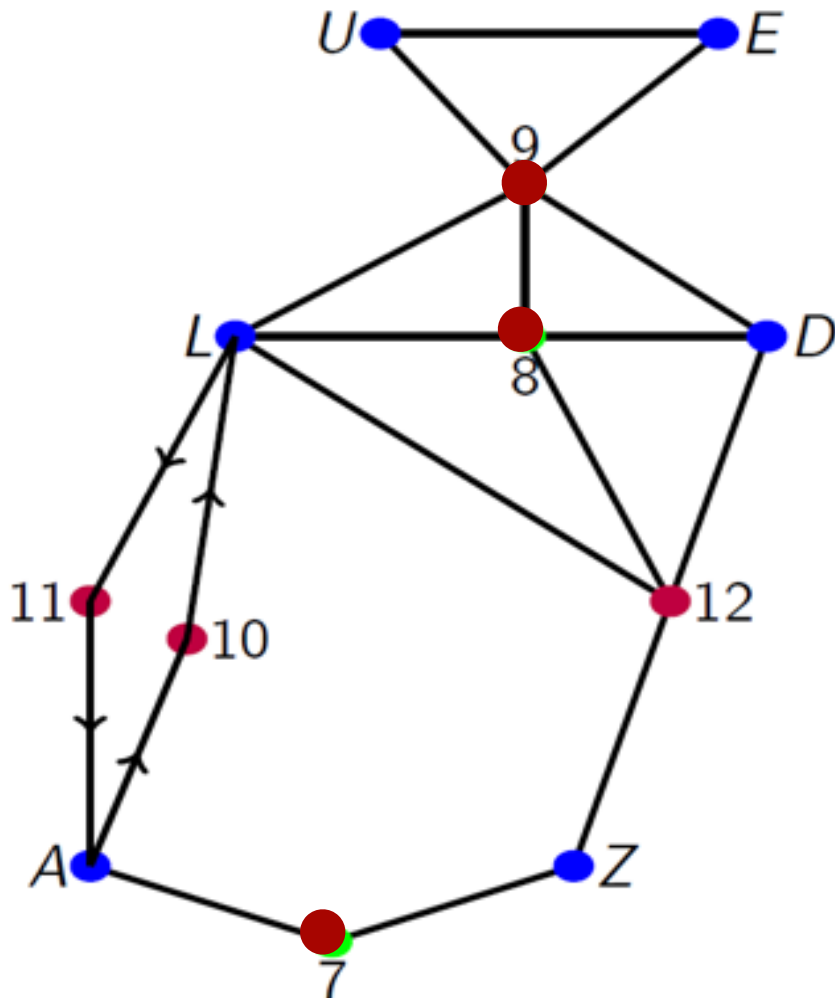


- **Classify action:**
  - **Scenario probability** × **Expected shortest response time to request**
  - **Sum over scenarios**
  - **Take best classified action**

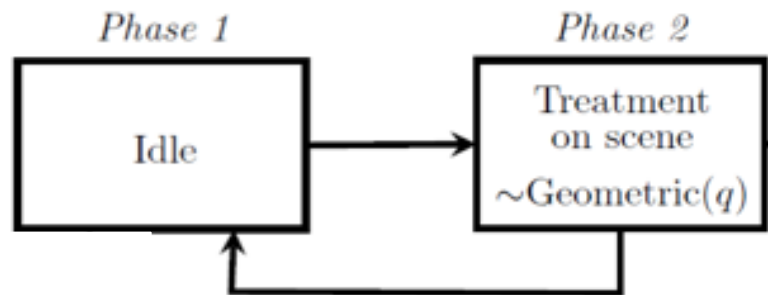




# Results

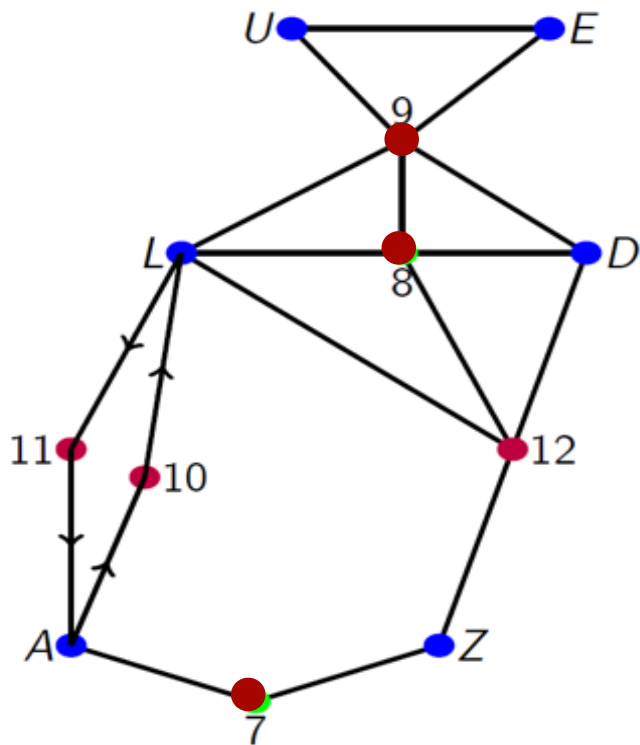


- $p_i$ 's estimated using historical data
- No hospitals
- $q = 0.3$
- 4 ambulances



# Results

	Mean Response Time	95% Confidence Interval	Compared to Optimal
Heuristic	0.6901	[0.6818, 0.6984]	4.01%
Optimal	0.6624	[0.6547, 0.6701]	-
Compliance Table	0.7521	[0.7437, 0.7605]	11.93%



$$p_A = 0.1568$$

$$p_L = 0.0743$$

$$p_Z = 0.0167$$

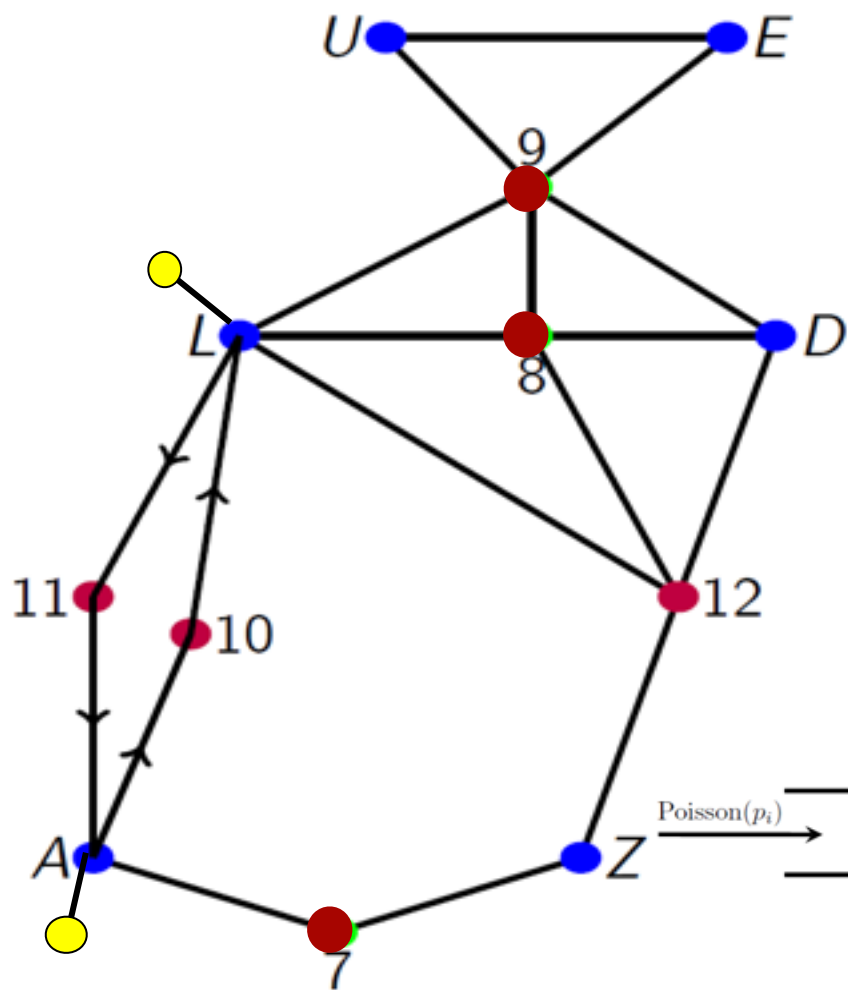
$$p_D = 0.0223$$

$$p_U = 0.0120$$

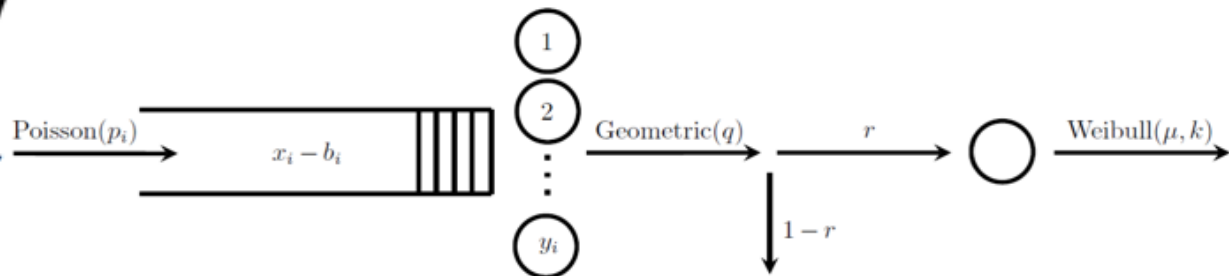
$$p_E = 0.0162$$

#	Compliance Table
1	A
2	A, L
3	A, L, 9
4	A, L, D, E

# Results

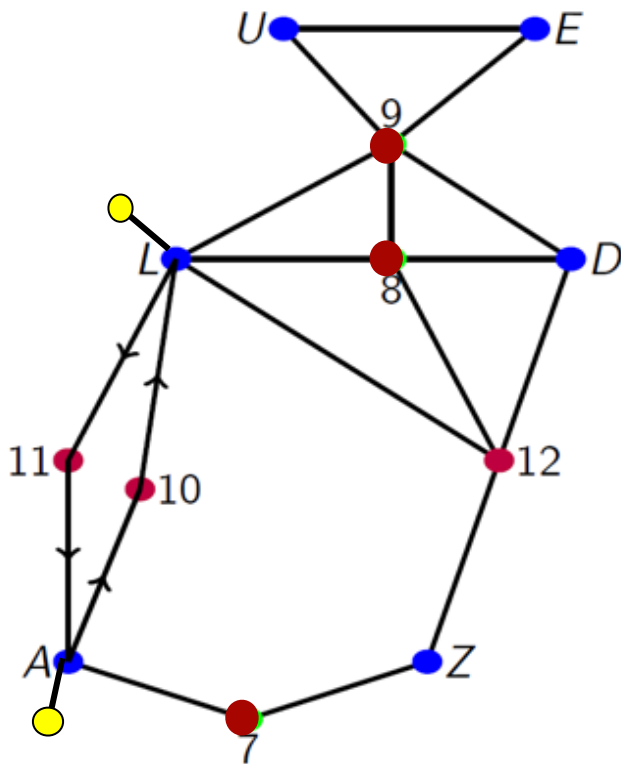


- $p_i$ 's as before
- 2 hospitals: ●
- $q = r = 0.75$
- $\mu = 3.1, k = 2.5$
- 6 ambulances



# Results

	Mean Response Time	95% Confidence Interval
Heuristic	0.3110	[0.3076, 0.3144]
Compliance Table	0.3448	[0.3406, 0.3489]



- $p_A = 0.1568$
- $p_L = 0.0743$
- $p_Z = 0.0167$
- $p_D = 0.0223$
- $p_U = 0.0120$
- $p_E = 0.0162$

#	Compliance Table
1	A
2	A, L
3	A, L, 9
4	A, L, D, E

CWI

# Questions?

