

Modelling ambulance deployment with CARMA

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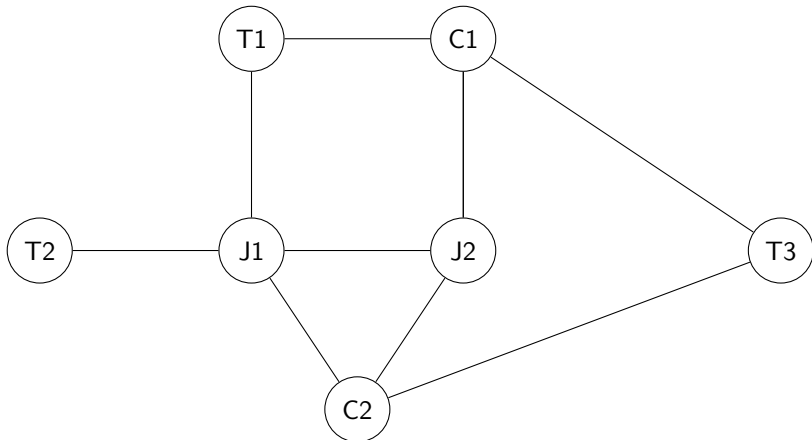
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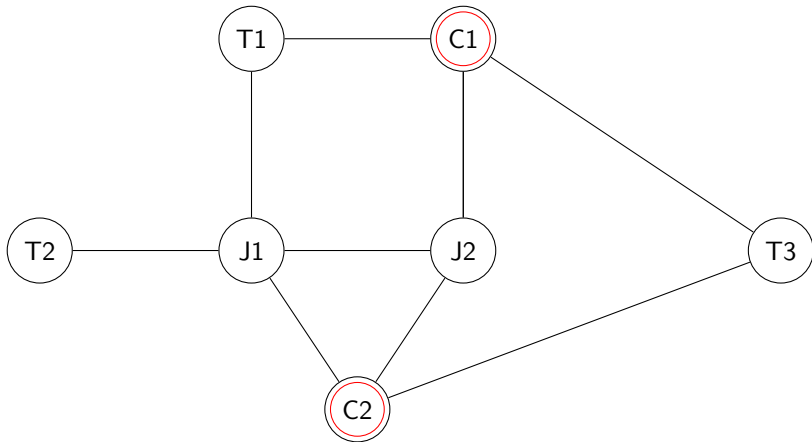
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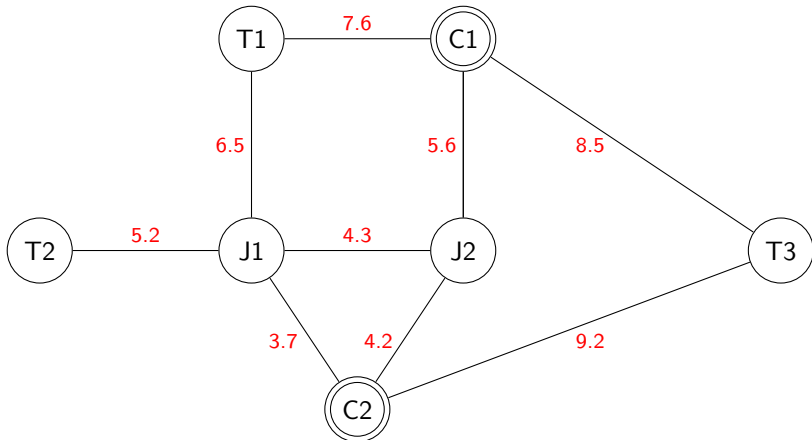
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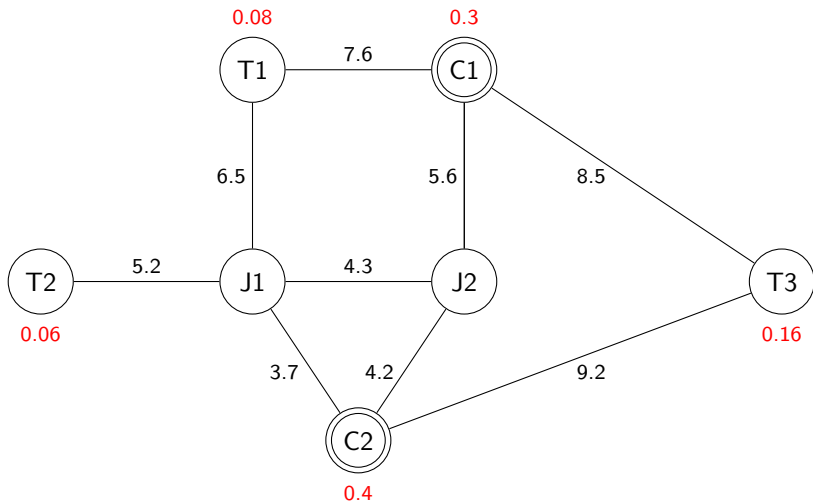
- 1 Introduction and motivation
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- 3 CARMA modelling language
- 4 Ambulance model
- 5 Experiments and results
- 6 Conclusion

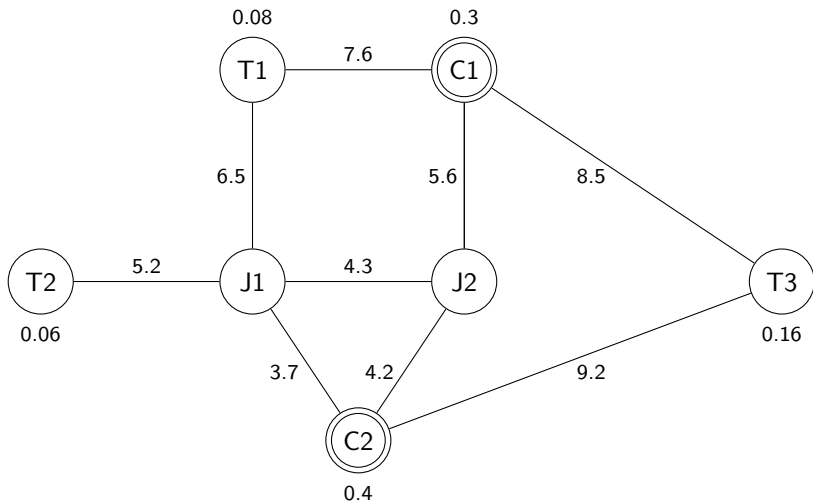
- modelling of collective adaptive systems
 - collective: multiple agents interacting to achieve goals
 - adaptive: change behaviour in response to changing environment
- CARMA: language developed for modelling these systems
- ambulance deployment
 - case study
 - network of places and roads
 - heuristic to decide waiting place of idle ambulance
- experimentation through modelling
- exploration of CARMA and heuristic











1. An incident occurs and its level of severity is determined.
2. An idle ambulance is identified based on nearness to the incident.
3. The ambulance takes the shortest route using sirens and lights.
4. The ambulance treats the patient at the scene and then proceeds with item 7, or the ambulance uploads the patient.
5. The ambulance uses the shortest route to the hospital with sirens and lights.
6. The ambulance drops the patient off at the hospital.
7. The ambulance uses the shortest route without sirens and lights to return to a base location.

- static: each ambulance has a specific base
 - decided in advance of the operation of the system
 - based on simulation or other measures
- dynamic: when an ambulance has finished dealing with an incident, its base is determined by the locations of other idle ambulances
 - table-based approach: requires dispatchers to steer system to optimal states
 - using approximate dynamic programming and post-decision state is time-consuming, requires expert to implement
 - lightweight on-the-fly heuristic (Jagtenberg, Bhulai and van der Mei, 2015)

- assume set of ambulances A following protocol
- target of T minutes for an ambulance to reach incident from time of notification of incident
- classification of arrivals at incidents
 - **on-time**: T or fewer minutes to reach incident
 - **late**: more than T minutes to reach incident
- performance evaluated by proportion of late arrivals determined with respect to T

$$\text{late rate} = \frac{\text{late arrivals}}{\text{late arrivals} + \text{on-time arrivals}}$$

- $N = (V, E)$, $E \subseteq V \times V$: graph of places and roads
- $d(v) \in [0, 1]$: distribution of incidents
- $\rho(v, u) \in \mathbb{R}_{>0}$: deterministic time to get from vertex v to vertex u (with sirens and lights)
- $n_u(t)$: idle ambulances at or moving towards base $u \in B$

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- marginal coverage at vertex v for k ambulances that can arrive within T minutes: $E_k(v) - E_{k-1}(v) = d(v)(1 - q)q^{k-1}$
- the function returns a base giving maximum marginal coverage

$$p(N(t)) = \arg \max_{w \in B} \sum_{v \in V} \mathbf{1}(\rho(w, v) \leq T) \cdot d(v)(1 - q)q^{c(v, N(t))}$$

$$c(v, N(t)) = \sum_{u \in B} n_u(t) \cdot \mathbf{1}(\rho(u, v) \leq T)$$

- quantitative modelling of collective adaptive system
- focus on smart transport and smart grid
- define collectives of components that interact
- consider their behaviour within a specific environment defined separately
- CARMA features
 - local store in each component
 - unicast and broadcast communication
 - flexible modelling of discrete space
 - model behaviours described as time-inhomogeneous continuous-time Markov chains
- simulation using the CARMA Eclipse Plug-in

- set of CARMA *systems* SYS defined by

$$S ::= N \text{ in } \mathcal{E}$$

where N is a *collective* and \mathcal{E} is an *environment*.

- set of collectives COL is defined by

$$N ::= C \mid N \parallel N$$

- set of components COMP is defined by

$$C ::= \mathbf{0} \mid (P, \gamma)$$

where $\mathbf{0}$ is the null component, P is a process and γ is its store

$ \begin{array}{l} P, Q ::= \mathbf{nil} \\ \mathbf{kill} \\ act.P \\ P + Q \\ P \mid Q \\ [\pi]P \\ A \quad (A \stackrel{\text{def}}{=} P) \end{array} $	$ \begin{array}{l} act ::= \alpha^*[\pi]\langle \vec{e} \rangle \sigma \\ \alpha[\pi]\langle \vec{e} \rangle \sigma \\ \alpha^*[\pi](\vec{x})\sigma \\ \alpha[\pi](\vec{x})\sigma \end{array} $
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$$e ::= a \mid \text{my}.a \mid x \mid v \mid \text{now} \mid \dots$$

$$\pi ::= \top \mid \perp \mid e_1 \bowtie e_2 \mid \neg\pi \mid \pi \wedge \pi \mid \dots$$

four different action prefixes

- *broadcast output*: $\alpha^*[\pi]\langle \vec{e} \rangle \sigma$,
- *broadcast input*: $\alpha^*[\pi](\vec{x})\sigma$,
- *output*: $\alpha[\pi]\langle \vec{e} \rangle \sigma$,
- *input*: $\alpha[\pi](\vec{x})\sigma$,

where

- α : action type, π : predicate, x : variable, e : expression
- $\vec{\cdot}$ indicates a sequence of elements
- σ is an *update* mapping the current store to a distribution over stores
- $\alpha^*[\perp]\langle \rangle$ is an action without communication

- γ_g : global store for global attributes
- $\langle \mu_p, \mu_r, \mu_u \rangle$: tuple of functions known as the *evaluation context*
 - $\mu_p(\gamma_s, \gamma_r, \alpha) \in [0, 1]$ determines the probability that a component with store γ_r can receive a message from a component with store γ_s when α is executed
 - $\mu_r(\gamma, \alpha) \in \mathbb{R}_{\geq 0}$ determines the execution rate of action α executed at a component with store γ
 - $\mu_u(\gamma, \alpha)$ determines the updates on the environment (global store and collective) induced by the execution of action α at a component with store γ

- $|A|$ Ambulance components
- *Incident_Queue* component, *Closest_Idle_Ambulance* component
- other components are created when necessary:
Incident_Queue_Item, *Incident_Handler*, *Route*, *Return_Handler*

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- functions embody knowledge of network
 - **IncidentLocation()**, **IncidentType()**
 - **HospitalLocation**(*loc*)
 - **MoveTime**(*nexts*, *nexte*, *dest*)
 - **NextHop**(*i*, *aloc*, *loc*), **RouteLength**(*aloc*, *loc*)
 - **GetBase**(*anum*, *aloc*, *N*)
 - **ClosestIdleAmbulance**(*iloc*, *N*)

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 - **ClosestIdleAmbulance**(*iloc*, *N*)
- timing and probabilities
 - deterministic: time to move between vertices
 - stochastic: time for treatment, pick up, drop off
 - probabilistic: distribution of incidents and type of incident

Constants:

T limit for response time
 $timeout$ time to wait for a response for a request

Measures:

N set containing the number of idle ambulances
at each possible base

Initial collective:

$EMS \stackrel{\text{def}}{=} (Incident_Queue, \{inum \mapsto 0, rnum \mapsto 1\}) \parallel$
 $(ClosestIdleAmbulance, \{dloc \mapsto nullLoc, iloc \mapsto nullLoc, t \mapsto 0\}) \parallel$
 $(Ambulance, \{anum \mapsto 1, aloc \mapsto l_1, abase \mapsto l_1, idle \mapsto \top\}) \parallel \dots \parallel$
 $(Ambulance, \{anum \mapsto n, aloc \mapsto l_n, abase \mapsto l_n, idle \mapsto \top\})$

Global store:

ontime number of ontime ambulances
late number of late ambulances

Evolution rule functions:

$$\mu_p(\gamma_s, \gamma_r, \alpha) = 1$$

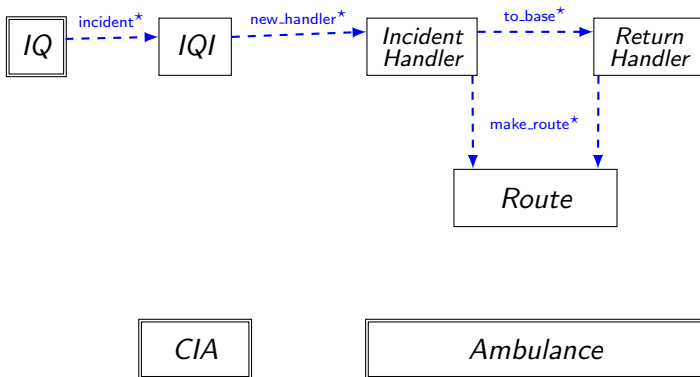
$$\mu_r(\gamma_s, \alpha) = \begin{cases} 1/r & \alpha = \text{incident}^* \quad (r: \text{mean time between incidents}) \\ \lambda_p & \alpha = \text{pickup}^* \\ \lambda_t & \alpha = \text{treat}^* \\ \lambda_d & \alpha = \text{dropoff}^* \\ 0 & \alpha = \text{pause}^* \wedge \text{now} < \gamma_s(t) + \text{timeout} \\ \lambda_{fast} & \text{otherwise} \end{cases}$$

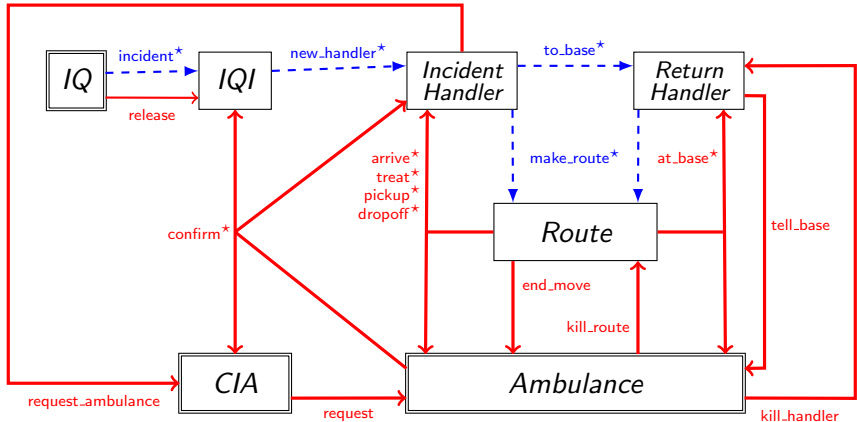
$$\mu_u(\gamma_s, \alpha) = \begin{cases} \{ontime \leftarrow ontime + 1\}, 0 \\ \quad \alpha = \text{timecheck}^* \wedge \text{now} \leq \gamma_s(\text{itime}) + T \\ \{late \leftarrow late + 1\}, 0 \\ \quad \alpha = \text{timecheck}^* \wedge \text{now} > \gamma_s(\text{itime}) + T \\ \vdots \end{cases}$$

$$\mu_u(\gamma_s, \alpha) = \begin{cases} \vdots \\ \{\}, (Incident_Queue_Item, \{qnum \leftarrow \gamma_s(num), itime \leftarrow now\}) \\ \alpha = incident^* \\ \{\}, (Incident_Handler, \{loc \leftarrow IncidentLocation(), \\ dest \leftarrow IncidentType(), \\ itime \leftarrow \gamma_s(itime)\}) \\ \alpha = new_handler^* \\ \vdots \end{cases}$$

$$\mu_u(\gamma_s, \alpha) = \left\{ \begin{array}{l} \vdots \\ \{\}, (Route, \{ anum \leftarrow \gamma_s(anum), \\ \quad dest \leftarrow \gamma_s(dest), \\ \quad start \leftarrow \gamma_s(alloc), \\ \quad end \leftarrow \gamma_s(loc), \\ \quad nexts \leftarrow \gamma_s(alloc), \\ \quad nexte \leftarrow \mathbf{NextHop}(1, \gamma_s(alloc), \gamma_s(loc)), \\ \quad h \leftarrow \mathbf{RouteLength}(\gamma_s(alloc), \gamma_s(loc)) \}) \\ \alpha = \mathbf{makeroute}^* \\ \vdots \end{array} \right.$$

$$\mu_u(\gamma_s, \alpha) = \left\{ \begin{array}{l} \vdots \\ \{\}, (Return_Handler, \{ anum \leftarrow \gamma_s(anum), \\ \quad aloc \leftarrow \gamma_s(aloc), \\ \quad dest \leftarrow \gamma_s(dest), \\ \quad loc \leftarrow \mathbf{GetBase}(\gamma_s(anum), \gamma_s(aloc), N) \}) \\ \alpha = \text{tobase}^* \\ \{\}, 0 \quad \text{otherwise} \end{array} \right.$$





Store of *Ambulance* component:

<i>anum</i>	ambulance id
<i>aloc</i>	current location of ambulance
<i>abase</i>	current base of ambulance
<i>idle</i>	whether ambulance is idle or not

Initial state of *Ambulance* component: *Idle*

Behaviour of Ambulance component:

<i>Idle</i>	$\stackrel{\text{def}}{=}$	$\text{request}[al == \text{my.}aloc](al)\{\text{idle} \leftarrow \perp\}.\text{Respond}$
<i>Respond</i>	$\stackrel{\text{def}}{=}$	$\text{confirm}^*[\top]\langle anum, aloc \rangle.\text{Busy}$
<i>Busy</i>	$\stackrel{\text{def}}{=}$	$\text{end_move}[an == \text{my.}anum \wedge \text{now} \geq t + d](an, al, t, d)$ $\{\text{aloc} \leftarrow al\}.\text{Busy} +$ $\text{arrive}^*[an == \text{my.}anum](an).\text{AtScene} +$ $\text{drop_off}^*[an == \text{my.}anum](an)\{\text{idle} \leftarrow \top\}.\text{AskBase}$
<i>AtScene</i>	$\stackrel{\text{def}}{=}$	$\text{pickup}^*[an == \text{my.}anum](an).\text{Busy} +$ $\text{treat}^*[an == \text{my.}anum](an)\{\text{idle} \leftarrow \top\}.\text{AskBase}$
<i>AskBase</i>	$\stackrel{\text{def}}{=}$	$\text{tell_base}[an == \text{my.}anum](an, ab)\{\text{abase} \leftarrow ab\}.\text{GoToBase}$
<i>GoToBase</i>	$\stackrel{\text{def}}{=}$	$\text{end_move}[an == \text{my.}anum \wedge \text{now} \geq t + d](an, al, t, d)$ $\{\text{aloc} \leftarrow al\}.\text{GoToBase} +$ $\text{atbase}^*[an == \text{my.}anum](an).\text{Idle} +$ $\text{request}[an == \text{my.}anum](an).\text{CleanUp1}$
<i>CleanUp1</i>	$\stackrel{\text{def}}{=}$	$\text{kill_handler}^*[\top]\langle anum \rangle.\text{CleanUp2}$
<i>CleanUp2</i>	$\stackrel{\text{def}}{=}$	$\text{kill_route}^*[\top]\langle anum \rangle.\text{Respond}$

Store of *Route* component:

<i>anum</i>	number of ambulance
<i>dest</i>	current destination type and incident type
<i>start</i>	start of route
<i>end</i>	end of route
<i>nexts</i>	start of next hop
<i>nexte</i>	end of next hop
<i>h</i>	number of hops in route
<i>i</i>	hop counter
<i>t</i>	timer variable for deterministic movement

Initial state of *Route* component: $R \mid KR$

Behaviour of *Route* component:

$$\begin{aligned}
 R \stackrel{\text{def}}{=} & [i < h] \text{start_move}^*[\perp](\langle \rangle) \{i \leftarrow i + 1, t \leftarrow \text{now}\}.RC + \\
 & [i = h \wedge (\text{my.dest} == \text{ToSevere} \vee \text{my.dest} == \text{ToMinor})] \\
 & \quad \text{arrive}^*[\top](\langle \text{anum} \rangle).RS + \\
 & [i = h \wedge \text{my.dest} == \text{ToHosp}] \text{dropoff}^*[\top](\langle \text{anum} \rangle).\text{kill} + \\
 & [i = h \wedge \text{my.dest} == \text{ToBase}] \text{atbase}^*[\top](\langle \text{anum} \rangle).\text{kill}
 \end{aligned}$$

$$\begin{aligned}
 RS \stackrel{\text{def}}{=} & [i = h \wedge \text{my.dest} == \text{ToSevere}] \text{pickup}^*[\top](\langle \text{anum} \rangle).\text{kill} + \\
 & [i = h \wedge \text{my.dest} == \text{ToMinor}] \text{treat}^*[\top](\langle \text{anum} \rangle).\text{kill}
 \end{aligned}$$

$$\begin{aligned}
 RC \stackrel{\text{def}}{=} & \text{end_move}[\top](\langle \text{anum}, \text{nexte}, t, \mathbf{MoveTime}(\text{nexts}, \text{nexte}, \text{dest}) \rangle) \\
 & \quad \{ \text{nexts} \leftarrow \text{my.nexte}, \text{nexte} \leftarrow \mathbf{NextHop}(i, \text{my.start}, \text{my.end}) \}.R
 \end{aligned}$$

$$KR = \text{kill_route}^*[\text{my.anum} == \text{an}](\text{an}).\text{kill}$$

Store of *Incident_Handler* component:

<i>loc</i>	location of incident or hospital
<i>anum</i>	id of ambulance assigned to incident
<i>aloc</i>	current location of ambulance assigned to incident
<i>dest</i>	current destination type and incident type
<i>itime</i>	time of incident
<i>atime</i>	arrival time at incident

Initial state of *Incident_Handler* component: *IH*

Behaviour of *Incident_Handler* component:

$IH \stackrel{\text{def}}{=} \text{request_ambulance}^*[\top]\langle loc \rangle.IH_C$
 $IH_C \stackrel{\text{def}}{=} \text{confirm}^*[an, al](\{anum \leftarrow an, aloc \leftarrow al\}).IH_I$
 $IH_I \stackrel{\text{def}}{=} \text{makeroute}^*[\perp]\langle \rangle.IH_S$
 $IH_S \stackrel{\text{def}}{=} \text{arrive}^*[an == \text{my.anum}](an).IH_T$
 $IH_T \stackrel{\text{def}}{=} \text{timecheck}^*[\perp]\langle \rangle\{atime \leftarrow \text{now}\}.IH_P$
 $IH_P \stackrel{\text{def}}{=} \text{pickup}^*[an == \text{my.anum}](an)\{aloc \leftarrow \text{my.loc},$
 $\quad loc \leftarrow \mathbf{HospitalLocation}(\text{my.loc}), dest \leftarrow \text{ToHosp}\}.IH_H +$
 $\text{treat}^*[an == \text{my.anum}](an)\{aloc \leftarrow \text{my.loc}, dest \leftarrow \text{ToBase}\}.IH_F$
 $IH_H \stackrel{\text{def}}{=} \text{makeroute}^*[\perp]\langle \rangle.IH_D$
 $IH_D \stackrel{\text{def}}{=} \text{dropoff}^*[an == \text{my.anum}](an)\{aloc \leftarrow \text{my.loc}, dest \leftarrow \text{ToBase}\}.IH_F$
 $IH_F \stackrel{\text{def}}{=} \text{tobase}^*[\perp]\langle \rangle.\mathbf{kill}$

Store of *Return_Handler* component:

<i>anum</i>	id of ambulance assigned
<i>aloc</i>	current location of ambulance
<i>dest</i>	current destination and incident type
<i>loc</i>	location of base of ambulance

Initial state of *Return_Handler* component: RH

Behaviour of *Return_Handler* component:

RH	$\stackrel{\text{def}}{=} \text{tell_base}[\top]\langle loc \rangle.RH'$
RH'	$\stackrel{\text{def}}{=} \text{makeroute}^*[\perp]\langle \rangle.RH'' + \text{kill_handler}^*[\text{my.anum} == an](an).\text{kill}$
RH''	$\stackrel{\text{def}}{=} \text{atbase}^*[\text{my.anum} == an](an).\text{kill} + \text{kill_handler}^*[\text{my.anum} == an](an).\text{kill}$

Store of *Incident_Queue* component:

inum number of incidents generated

rnum next incident to be dealt with

Initial state of *Incident_Queue* component: $IG \mid RN$

Behaviour of *Incident_Queue* component:

$IG \stackrel{\text{def}}{=} \text{incident}^*[\perp]\langle \rangle \{inum \leftarrow inum + 1\}.IG$

$RN \stackrel{\text{def}}{=} \text{release}[\top]\langle rnum \rangle.RN'$

$RN' \stackrel{\text{def}}{=} \text{confirm}^*[\top](an, al) \{rnum \leftarrow rnum + 1\}.RN$

Store of *Incident_Queue_Item* component:

qnum number of incident

itime time of incident

Initial state of *Incident_Queue_Item* component: IQI

Behaviour of *Incident_Queue_Item* component:

$IQI \stackrel{\text{def}}{=} \text{release}[\text{my}.qnum == n](n).IQI'$

$IQI' \stackrel{\text{def}}{=} \text{new_handler}^*[\perp]\langle \rangle.\text{kill}$

Store of *Closest_Idle_Ambulance* component:

iloc location of incident
dloc location of idle ambulances
t timer variable for timeout

Initial state of *Closest_Idle_Ambulance* component: *CIA*

Behaviour of *Closest_Idle_Ambulance* component:

$CIA \stackrel{\text{def}}{=} \text{request_ambulance}[\top](l) \{ iloc \leftarrow l, \quad dloc \leftarrow \mathbf{ClosestIdleLoc}(iloc, N), t \leftarrow \text{now} \}. CIA'$

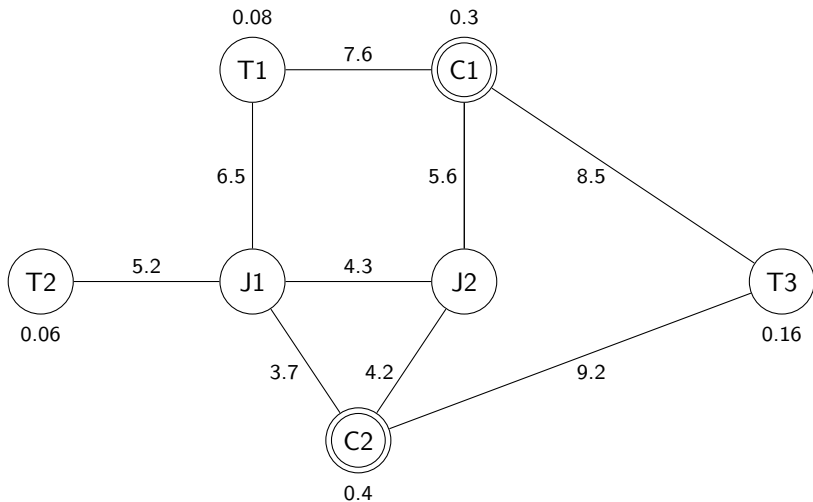
$CIA' \stackrel{\text{def}}{=} \text{request}[\top](dloc). CIA'' + \text{pause}^*[\perp](\rangle) \{ dloc \leftarrow \mathbf{ClosestIdleLoc}(iloc, N), t \leftarrow \text{now} \}. CIA'$

$CIA'' \stackrel{\text{def}}{=} \text{confirm}^*[\top](an, al). CIA$

- 3 ambulances
- parameters

λ_p	1/12	rate of pickup* action
λ_t	1/12	rate of treat* action
λ_d	1/15	rate of dropoff* action
λ_{fast}	100	all other actions
q	0.65	busy fraction
m	1	proportion of serious incidents
r	25	mean time between incidents

- 500 simulation runs of 20 hours of simulated time
- investigation of hospital location and availability
- investigation of behaviour of heuristic



late rate

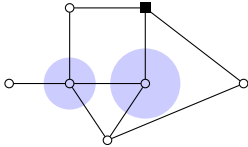
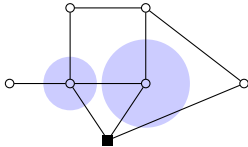
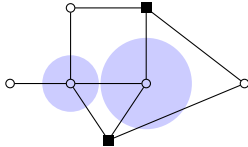
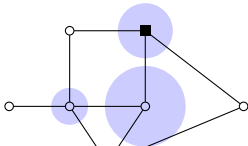
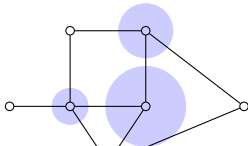
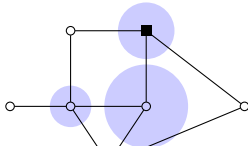
hospital locations

base locations

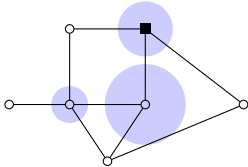
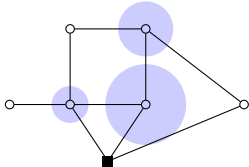
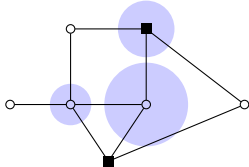
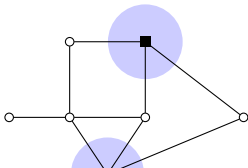
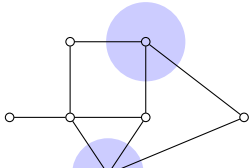
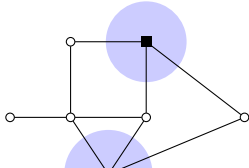
C1 C2 both

7 min	0.53	0.46	0.46	C1 C2 J1 J2 T1 T2 T3
9 min	0.42	0.41	0.37	C1 C2 J1 J2 T1 T2 T3
10 min	0.30	0.30	0.26	C1 C2 J1 J2 T1 T2 T3
11 min	0.35	0.20	0.22	C1 C2 J1 J2 T1 T2 T3
13 min	0.44	0.35	0.35	C1 C2 J1 J2 T1 T2 T3

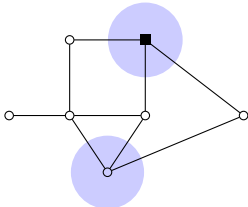
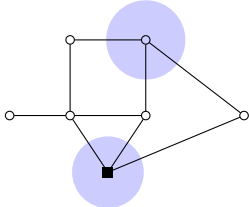
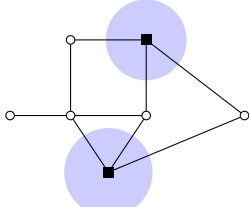
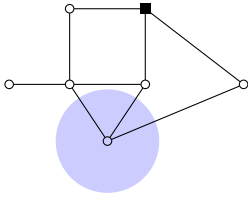
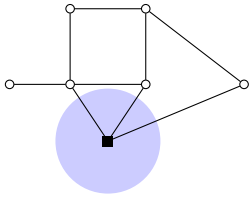
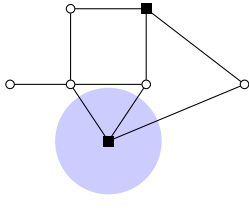
Results: 7 minutes and 9 minutes

Hospital at C1	Hospital at C2	Hospital at C1 and C2
 <p>A network diagram with 6 nodes and 7 edges. The top node is a black square (hospital). The other nodes are white circles. Three nodes are highlighted with blue circles of varying sizes.</p>	 <p>A network diagram with 6 nodes and 7 edges. The bottom node is a black square (hospital). The other nodes are white circles. Three nodes are highlighted with blue circles of varying sizes.</p>	 <p>A network diagram with 6 nodes and 7 edges. Both the top and bottom nodes are black squares (hospitals). The other nodes are white circles. Three nodes are highlighted with blue circles of varying sizes.</p>
$T = 7 \text{ min}$ $lr = 0.53$	$T = 7 \text{ min}$ $lr = 0.46$	$T = 7 \text{ min}$ $lr = 0.46$
 <p>A network diagram with 6 nodes and 7 edges. The top node is a black square (hospital). The other nodes are white circles. Three nodes are highlighted with blue circles of varying sizes.</p>	 <p>A network diagram with 6 nodes and 7 edges. The bottom node is a black square (hospital). The other nodes are white circles. Three nodes are highlighted with blue circles of varying sizes.</p>	 <p>A network diagram with 6 nodes and 7 edges. Both the top and bottom nodes are black squares (hospitals). The other nodes are white circles. Three nodes are highlighted with blue circles of varying sizes.</p>
$T = 9 \text{ min}$ $lr = 0.42$	$T = 9 \text{ min}$ $lr = 0.41$	$T = 9 \text{ min}$ $lr = 0.37$

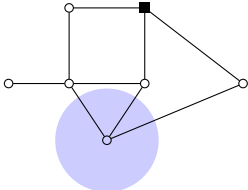
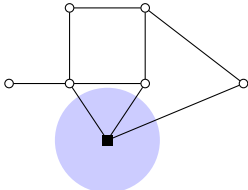
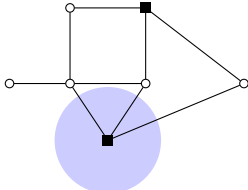
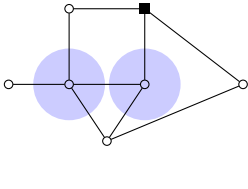
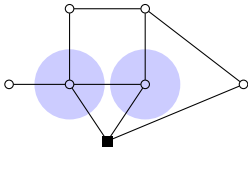
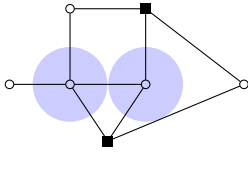
Results: 9 minutes and 10 minutes

Hospital at C1	Hospital at C2	Hospital at C1 and C2
 <p>$T = 9 \text{ min}$ $lr = 0.42$</p>	 <p>$T = 9 \text{ min}$ $lr = 0.41$</p>	 <p>$T = 9 \text{ min}$ $lr = 0.37$</p>
 <p>$T = 10 \text{ min}$ $lr = 0.30$</p>	 <p>$T = 10 \text{ min}$ $lr = 0.30$</p>	 <p>$T = 10 \text{ min}$ $lr = 0.26$</p>

Results: 10 minutes and 11 minutes

Hospital at C1	Hospital at C2	Hospital at C1 and C2
 <p>$T = 10 \text{ min}$ $lr = 0.30$</p>	 <p>$T = 10 \text{ min}$ $lr = 0.30$</p>	 <p>$T = 10 \text{ min}$ $lr = 0.26$</p>
 <p>$T = 11 \text{ min}$ $lr = 0.35$</p>	 <p>$T = 11 \text{ min}$ $lr = 0.20$</p>	 <p>$T = 11 \text{ min}$ $lr = 0.22$</p>

Results: 11 minutes and 13 minutes

Hospital at C1	Hospital at C2	Hospital at C1 and C2
 <p>A network diagram with 6 nodes and 7 edges. A blue circle is centered at node C1. A black square is at node C2.</p>	 <p>A network diagram with 6 nodes and 7 edges. A blue circle is centered at node C2. A black square is at node C2.</p>	 <p>A network diagram with 6 nodes and 7 edges. A blue circle is centered at node C1. A black square is at node C2.</p>
$T = 11 \text{ min}$ $lr = 0.35$	$T = 11 \text{ min}$ $lr = 0.20$	$T = 11 \text{ min}$ $lr = 0.22$
 <p>A network diagram with 6 nodes and 7 edges. Two blue circles are centered at nodes C1 and C2. A black square is at node C2.</p>	 <p>A network diagram with 6 nodes and 7 edges. Two blue circles are centered at nodes C1 and C2. A black square is at node C2.</p>	 <p>A network diagram with 6 nodes and 7 edges. Two blue circles are centered at nodes C1 and C2. A black square is at node C2.</p>
$T = 13 \text{ min}$ $lr = 0.44$	$T = 13 \text{ min}$ $lr = 0.35$	$T = 13 \text{ min}$ $lr = 0.35$

- exploration of parameter space
- exploration of different road network graphs
- calculation of busy fraction during simulation
 - longer simulations required to achieve stable busy fraction
- variants of heuristic
 - include hospital locations explicitly
 - different targets for different levels of severity
- variants of protocol
 - switch to ambulance closer to incident if one becomes available
 - move ambulances between bases when an incident occurs to improve coverage

- quantitative modelling of an ambulance scenario
- CARMA: a language for modelling collective adaptive systems
- CARMA supports separation of concerns
 - environment distinct from collective
 - functions to capture network specifics
- exploration of heuristic which determines next base
- heuristic behaviour is not monotonic
- various options for further exploration

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Thank you for your attention. Any questions?

- the function returns a base using maximum marginal coverage

$$\rho(N(t)) = \arg \max_{w \in B} \sum_{v \in V} \mathbf{1}(\rho(w, v) \leq T) \cdot d(v)(1 - q)q^{k(v, w, N(t)) - 1}$$

$$k(v, w, N(t)) = \mathbf{1}(\rho(w, v) \leq T) + \sum_{u \in B} n_u(t) \cdot \mathbf{1}(\rho(u, v) \leq T)$$

$$\text{where } \mathbf{1}(\phi) = \begin{cases} 1 & \text{if } \phi \text{ is true} \\ 0 & \text{if } \phi \text{ is false} \end{cases}$$

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$$p(N(t)) = \arg \max_{w \in B} \sum_{v \in V} \mathbf{1}(\rho(w, v) \leq T) \cdot d(v) (1 - q) q^{k(v, w, N(t)) - 1}$$

$$k(v, w, N(t)) = \mathbf{1}(\rho(w, v) \leq T) + \sum_{u \in B} n_u(t) \cdot \mathbf{1}(\rho(u, v) \leq T)$$

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$$p(N(t)) = \arg \max_{w \in B} \sum_{v \in V} \mathbf{1}(\rho(w, v) \leq T) \cdot d(v) (1 - q) q^{k(v, w, N(t)) - 1}$$

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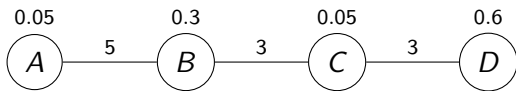
$$k(v, w, N(t)) = \mathbf{1}(\rho(w, v) \leq T) + \sum_{u \in B} n_u(t) \cdot \mathbf{1}(\rho(u, v) \leq T)$$

$$\text{where } \mathbf{1}(\phi) = \begin{cases} 1 & \text{if } \phi \text{ is true} \\ 0 & \text{if } \phi \text{ is false} \end{cases}$$

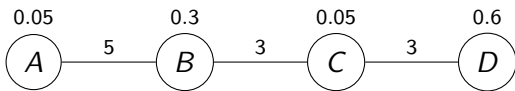
- it can be simplified to

$$\rho(N(t)) = \arg \max_{w \in B} \sum_{v \in V} \mathbf{1}(\rho(w, v) \leq T) \cdot d(v)(1 - q)q^{c(v, N(t))}$$

$$c(v, N(t)) = \sum_{u \in B} n_u(t) \cdot \mathbf{1}(\rho(u, v) \leq T)$$

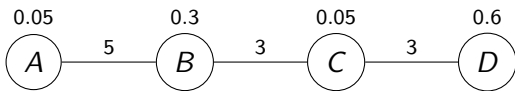


$$T = 11$$

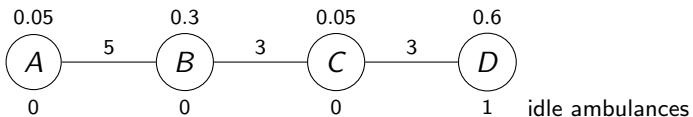


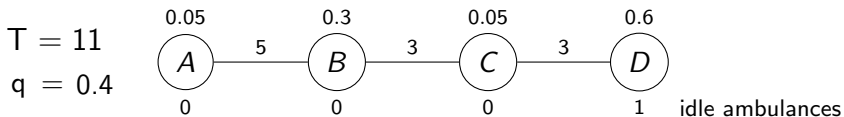
$$T = 11$$

$$q = 0.4$$



$$T = 11$$
$$q = 0.4$$

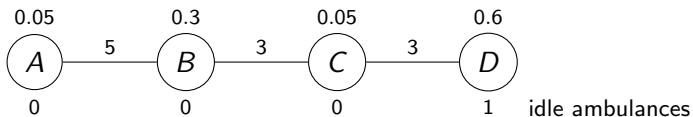




Base for return	A	B	C	D	
	Total idle ambulances in range when coverage increases				
A	2	2	2	2	
B	2	2	2	2	
C	2	2	2	2	
D	2	2	2	2	
	Increase in coverage				Total
A	0.01	0.07	0.01	0.14	0.24
B	0.01	0.07	0.01	0.14	0.24
C	0.01	0.07	0.01	0.14	0.24
D	0.01	0.07	0.01	0.14	0.24

$$T = 11$$

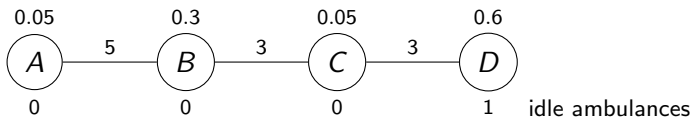
$$q = 0.8$$



Base for return	A	B	C	D	
	Total idle ambulances in range when coverage increases				
A	2	2	2	2	
B	2	2	2	2	
C	2	2	2	2	
D	2	2	2	2	
	Increase in coverage				Total
A	0.01	0.05	0.01	0.10	0.16
B	0.01	0.05	0.01	0.10	0.16
C	0.01	0.05	0.01	0.10	0.16
D	0.01	0.05	0.01	0.10	0.16

$$T = 11$$

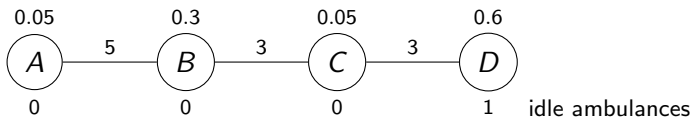
$$q = 0.4$$



Base for return	A	B	C	D	
	Total idle ambulances in range when coverage increases				
A	2	2	2	2	
B	2	2	2	2	
C	2	2	2	2	
D	2	2	2	2	
	Increase in coverage				Total
A	0.01	0.07	0.01	0.14	0.24
B	0.01	0.07	0.01	0.14	0.24
C	0.01	0.07	0.01	0.14	0.24
D	0.01	0.07	0.01	0.14	0.24

$$T = 7$$

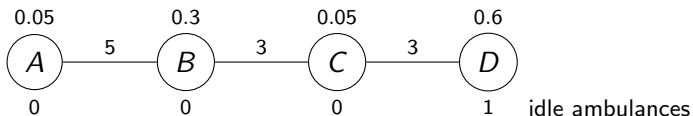
$$q = 0.4$$



Base for return	A	B	C	D	
	Total idle ambulances in range when coverage increases				
A	1	2	-	-	
B	1	2	2	2	
C	-	2	2	2	
D	-	2	2	2	
	Increase in coverage				Total
A	0.03	0.07			0.10
B	0.03	0.07	0.01	0.14	0.26
C		0.07	0.01	0.14	0.23
D		0.07	0.01	0.14	0.23

$$T = 5$$

$$q = 0.4$$



Base for return	A	B	C	D	Total
	Total idle ambulances in range when coverage increases				
A	1	1	-	-	
B	1	1	2	-	
C	-	1	2	2	
D	-	-	2	2	
	Increase in coverage				Total
A	0.03	0.18			0.21
B	0.03	0.18	0.01		0.22
C		0.18	0.01	0.14	0.34
D			0.01	0.14	0.16