

Dribble: A learn-based timer scheme selector for mobility management in IoT

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Agenda

1. Introduction
2. Background
3. Dribble
4. Evaluation
5. Conclusion and Future work

Introduction

Contextualization

- Smart devices are part of our daily routine



Mobility is a key challenge!

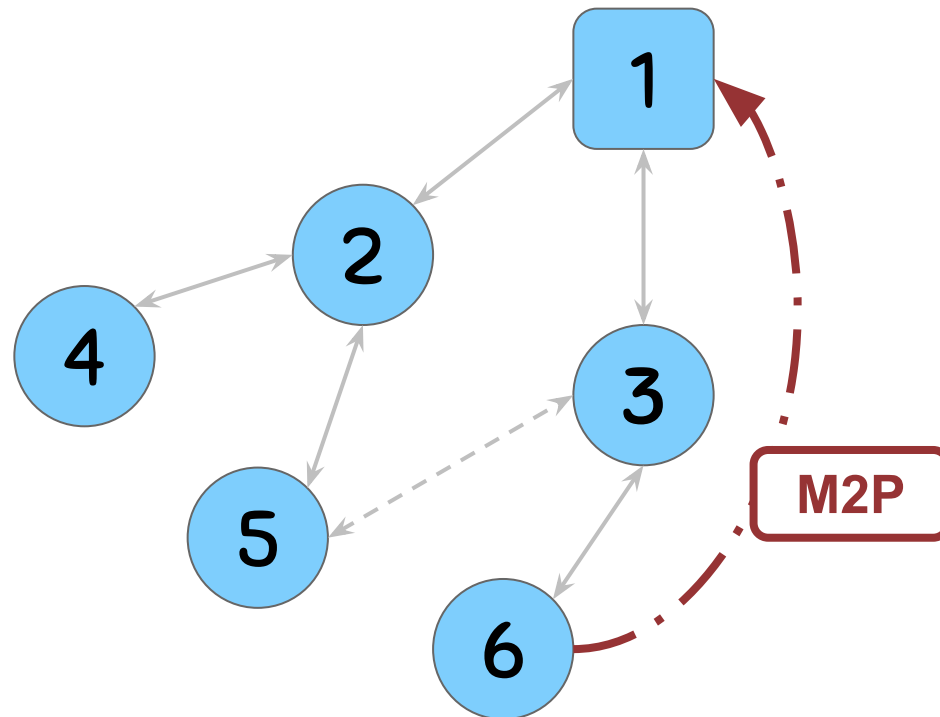
- IoT - Challenges
 - Internet adaptations
 - Heterogeneous devices
 - Constrained resources (Energy, CPU, Memory...)
 - Mobility

Mobility is a key challenge!

- IoT - Challenges
 - Internet adaptations
 - Heterogeneous devices
 - Constrained resources (Energy, CPU, Memory...)
 - **Mobility**
- We are interested in handle Mobility
 - Key aspect for mobile and wireless environment
 - Mobility from routing protocol lens

Background

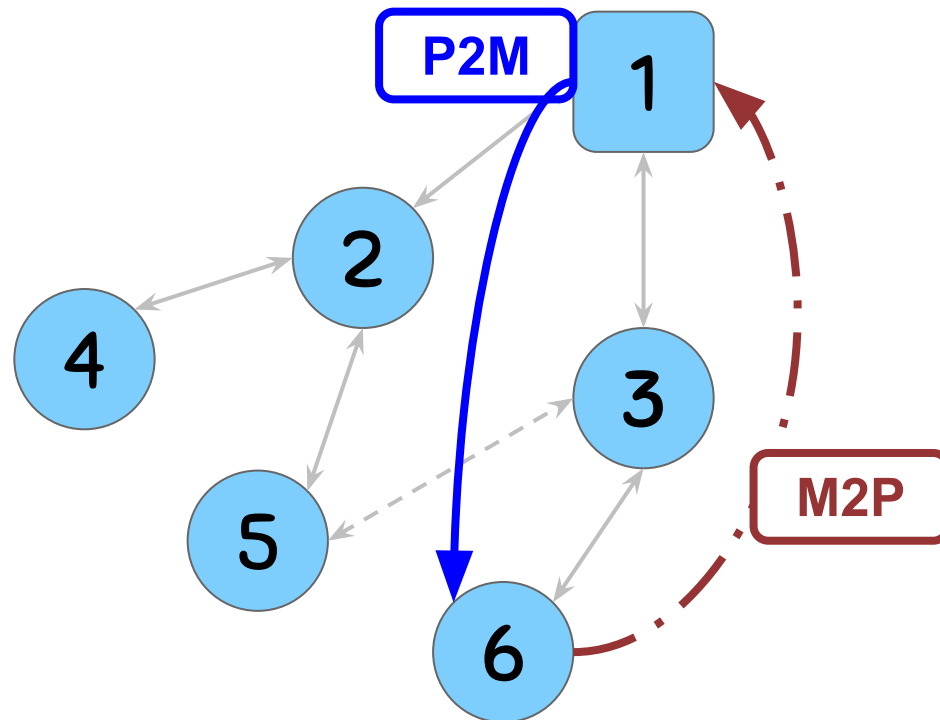
IoT routing in a nutshell



Data traffic patterns
over routing structures

Background

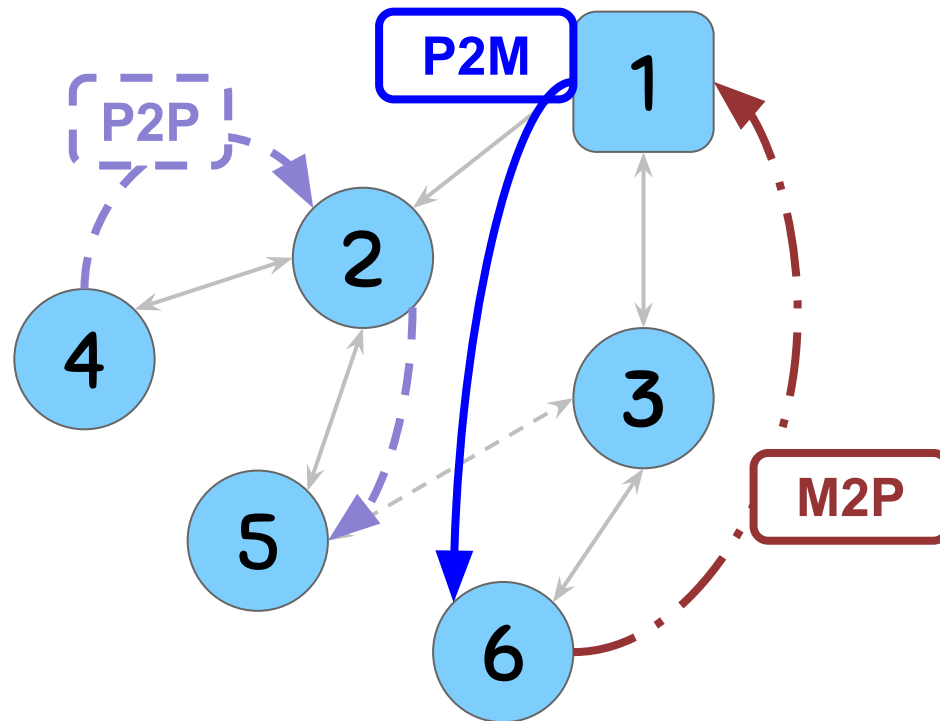
IoT routing in a nutshell



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Background

IoT routing in a nutshell



Data traffic patterns
over routing structures

IoT routing in a nutshell

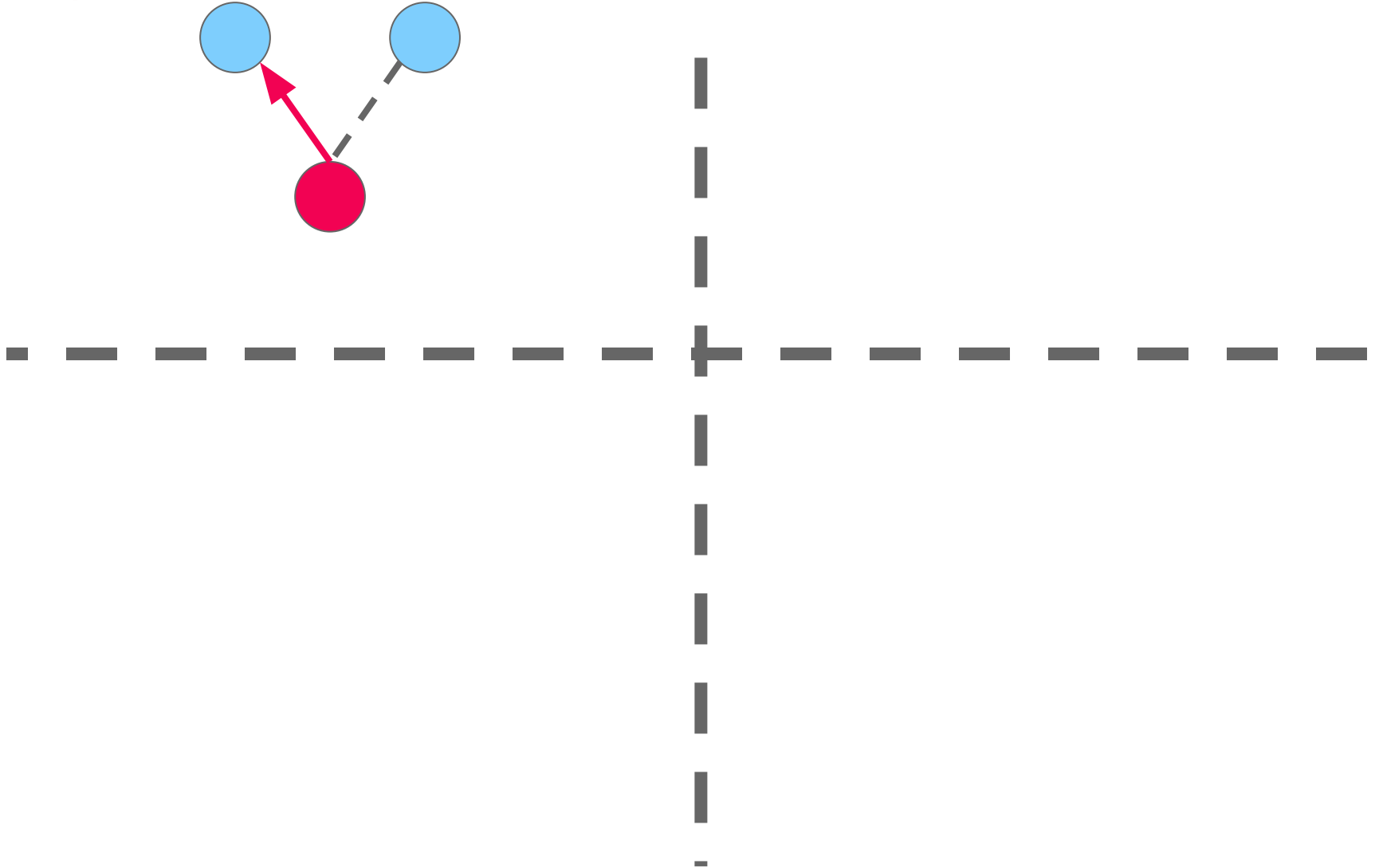
- Literature routing protocols
 - RPL (*de facto* the state-of-the-art)
 - Several RPL adaptations for mobile scenarios
 - Co-RPL, MRPL, MMRPL, ERPL...
 - Mobile Matrix
 - Hydro
 - XCTP
 - ...

Routing under mobility events

- Mostly of routing protocols for mobile IoT have one timer scheme
 - It governs the communication structure construction and maintenance

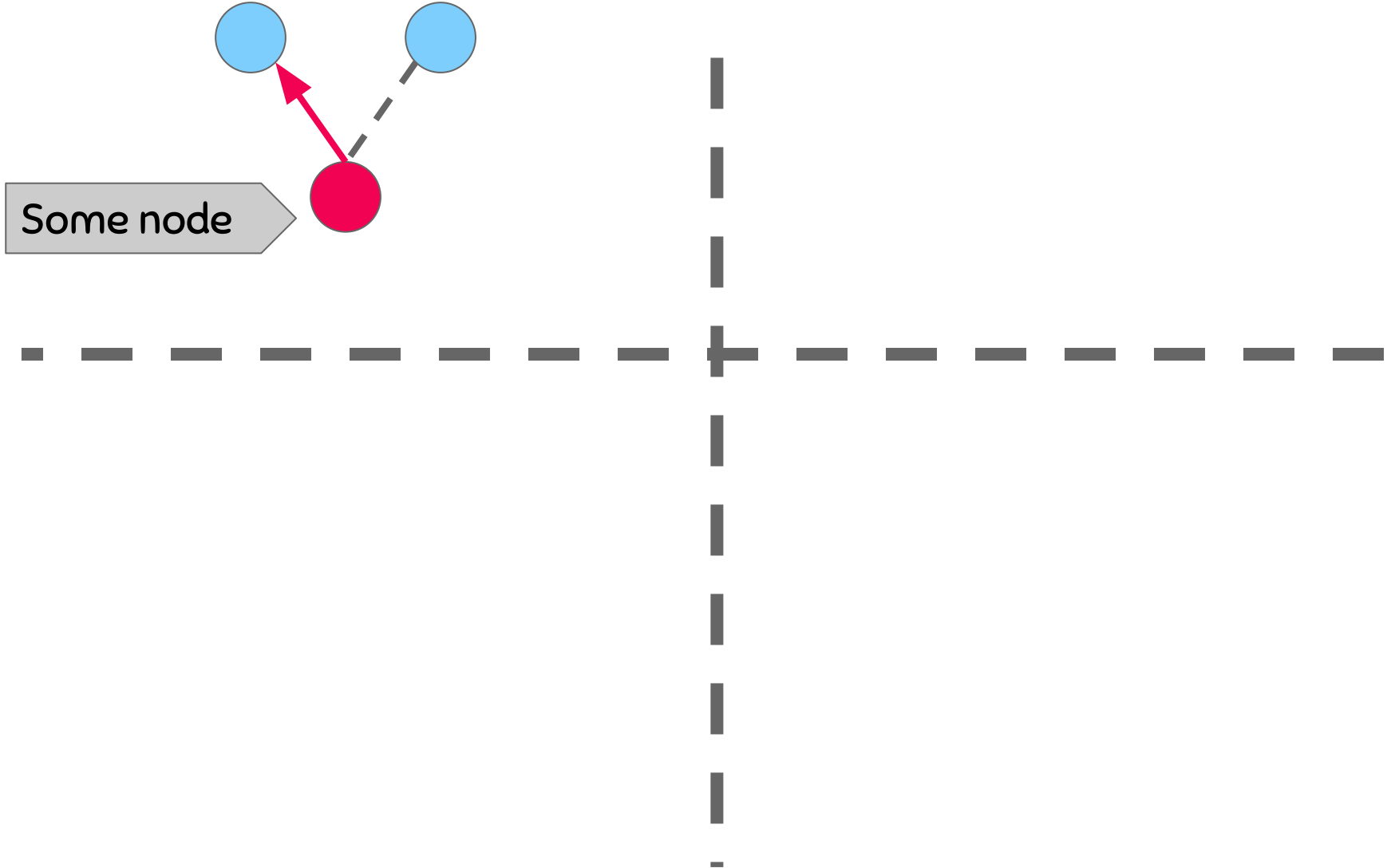
Routing under mobility events

Example. (note there are other solutions)



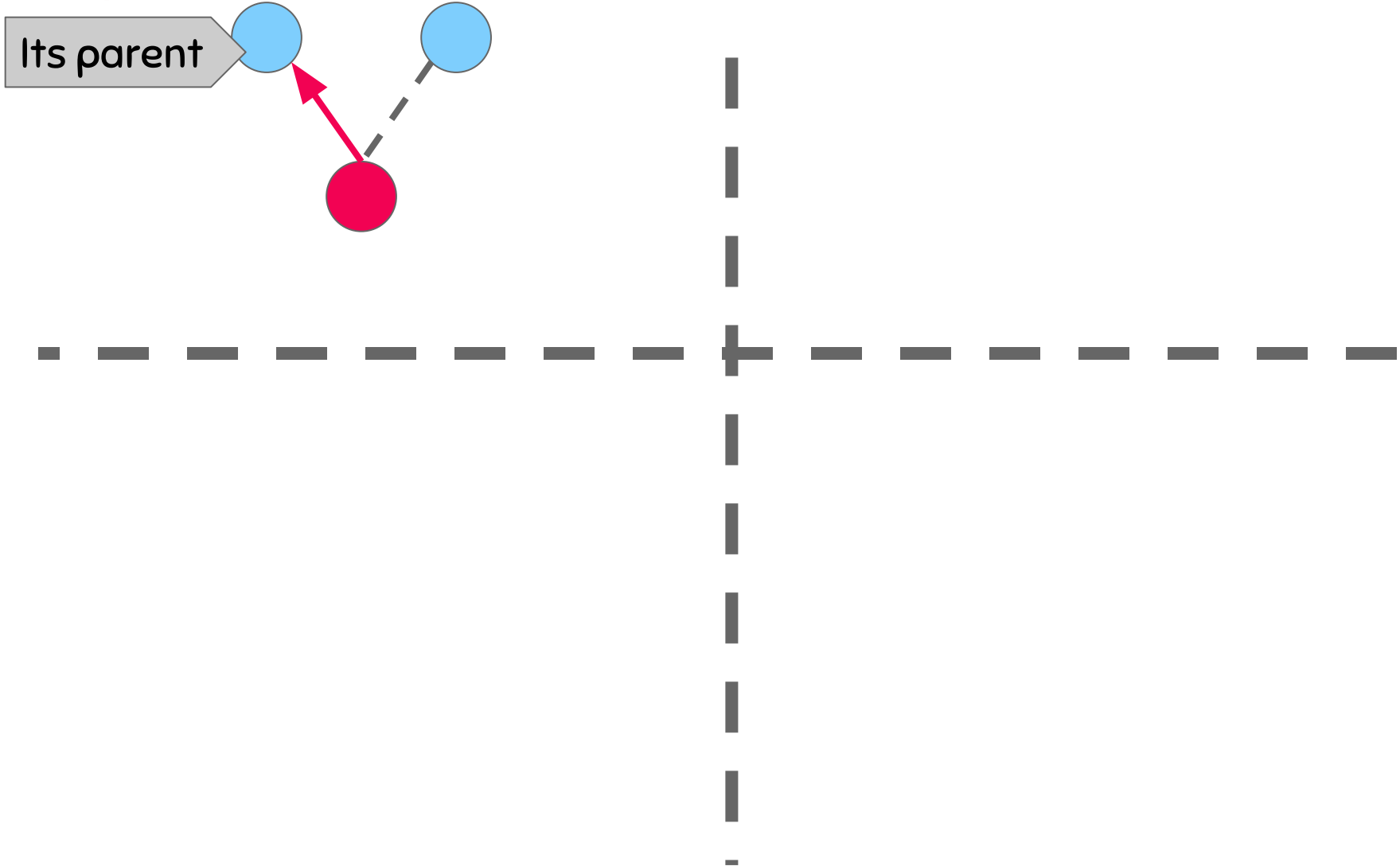
Routing under mobility events

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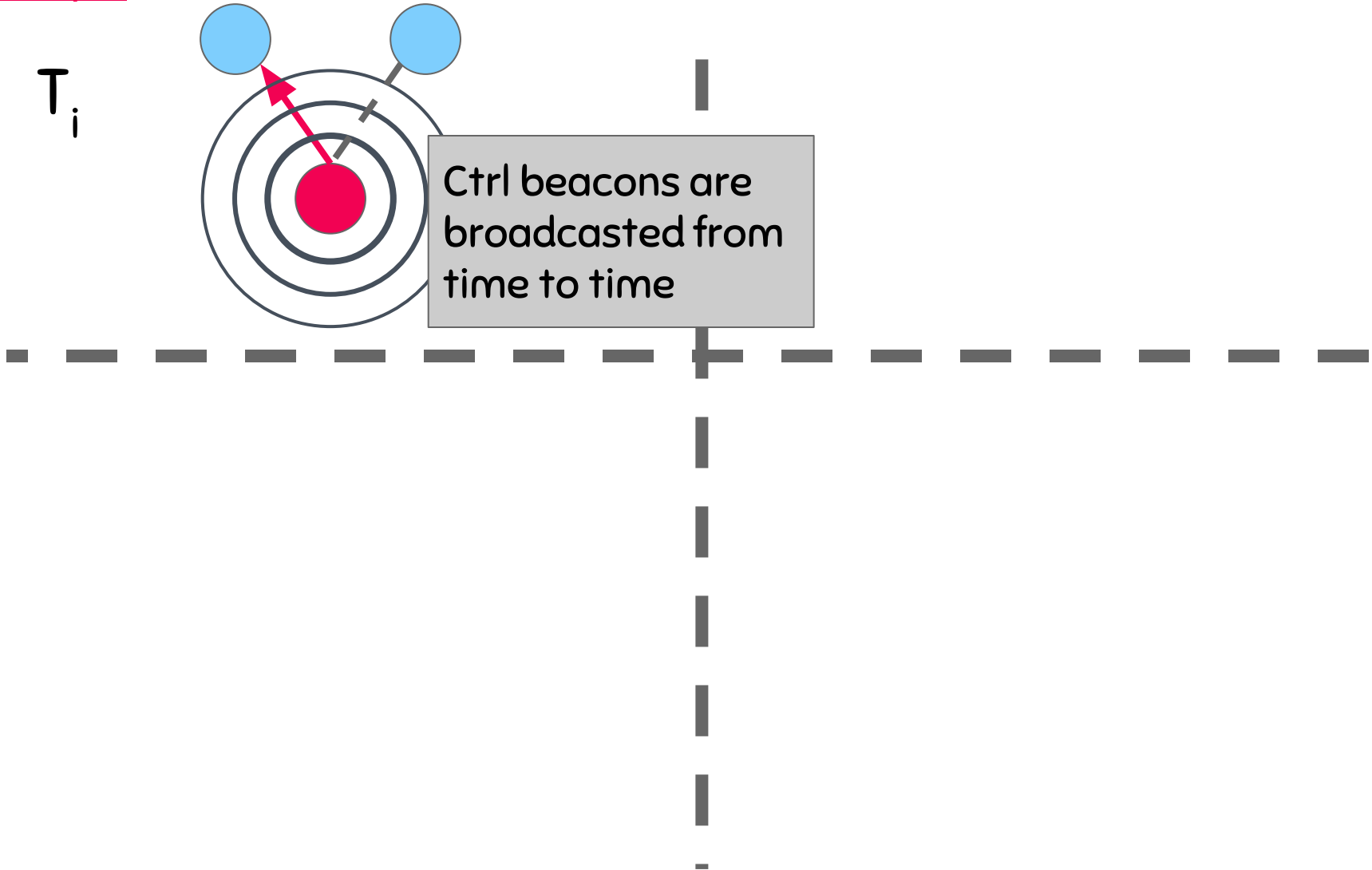
Routing under mobility events

Example. (note there are other solutions)



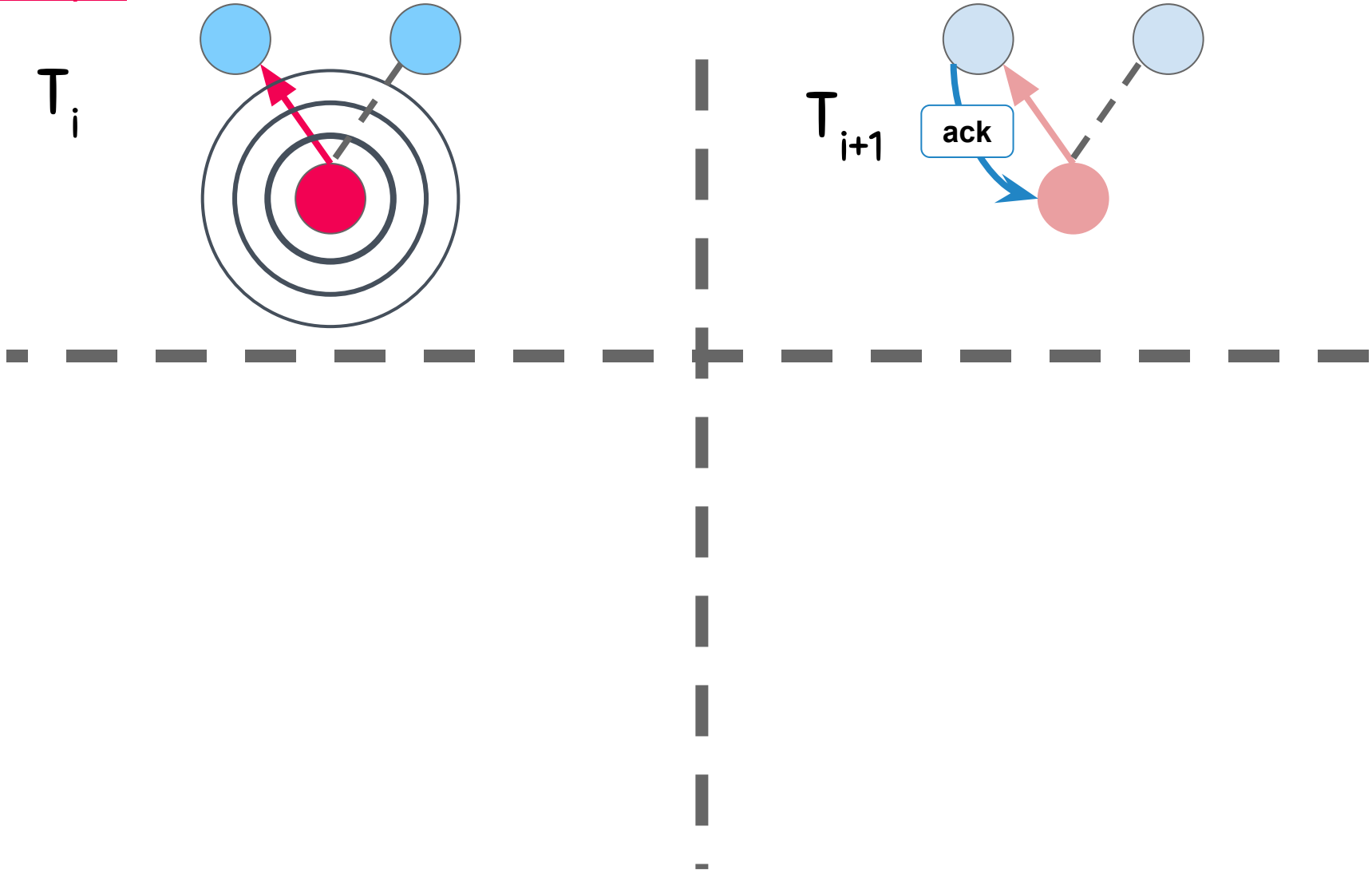
Routing under mobility events

Example. (note there are other solutions)



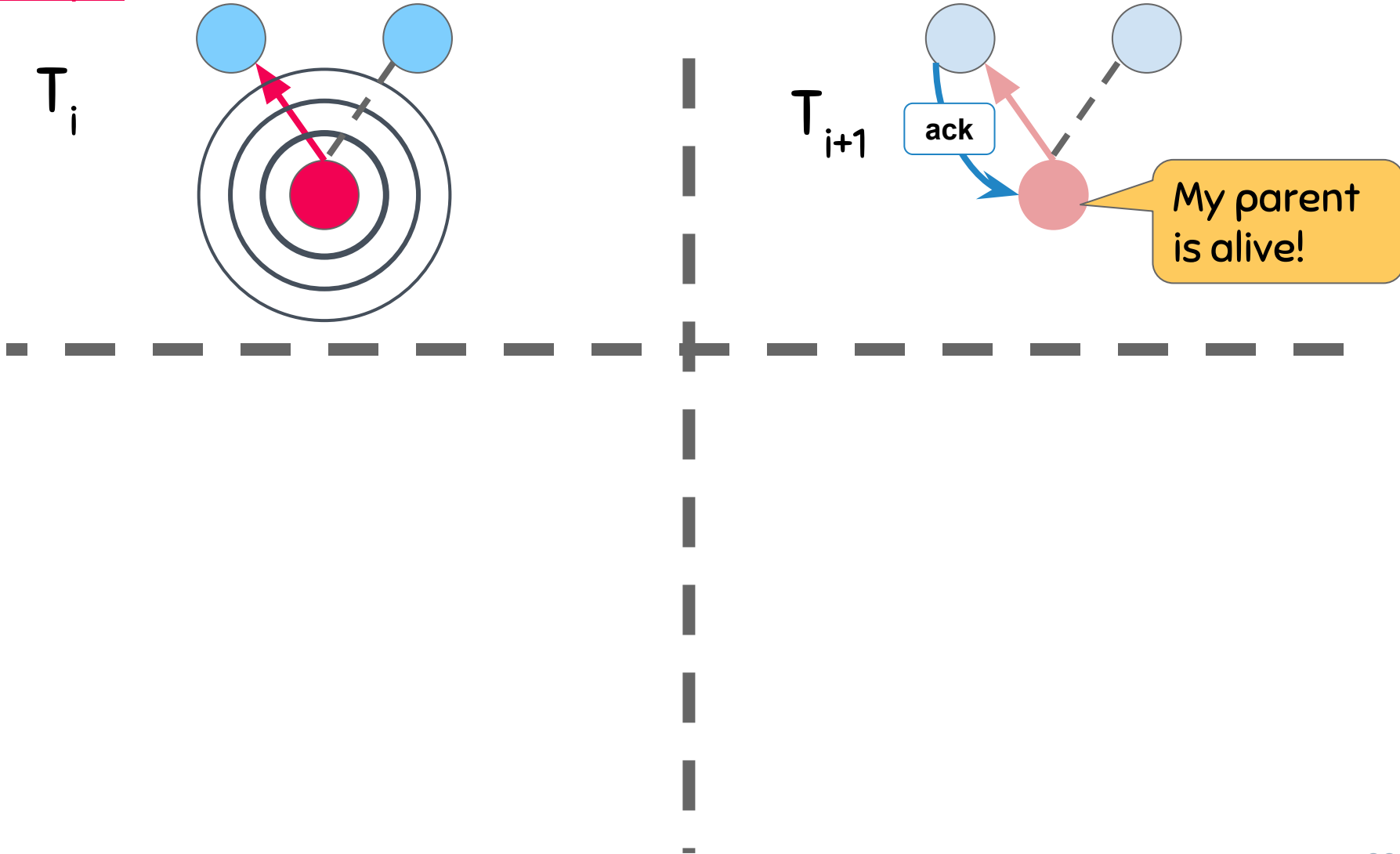
Routing under mobility events

Example. (note there are other solutions)



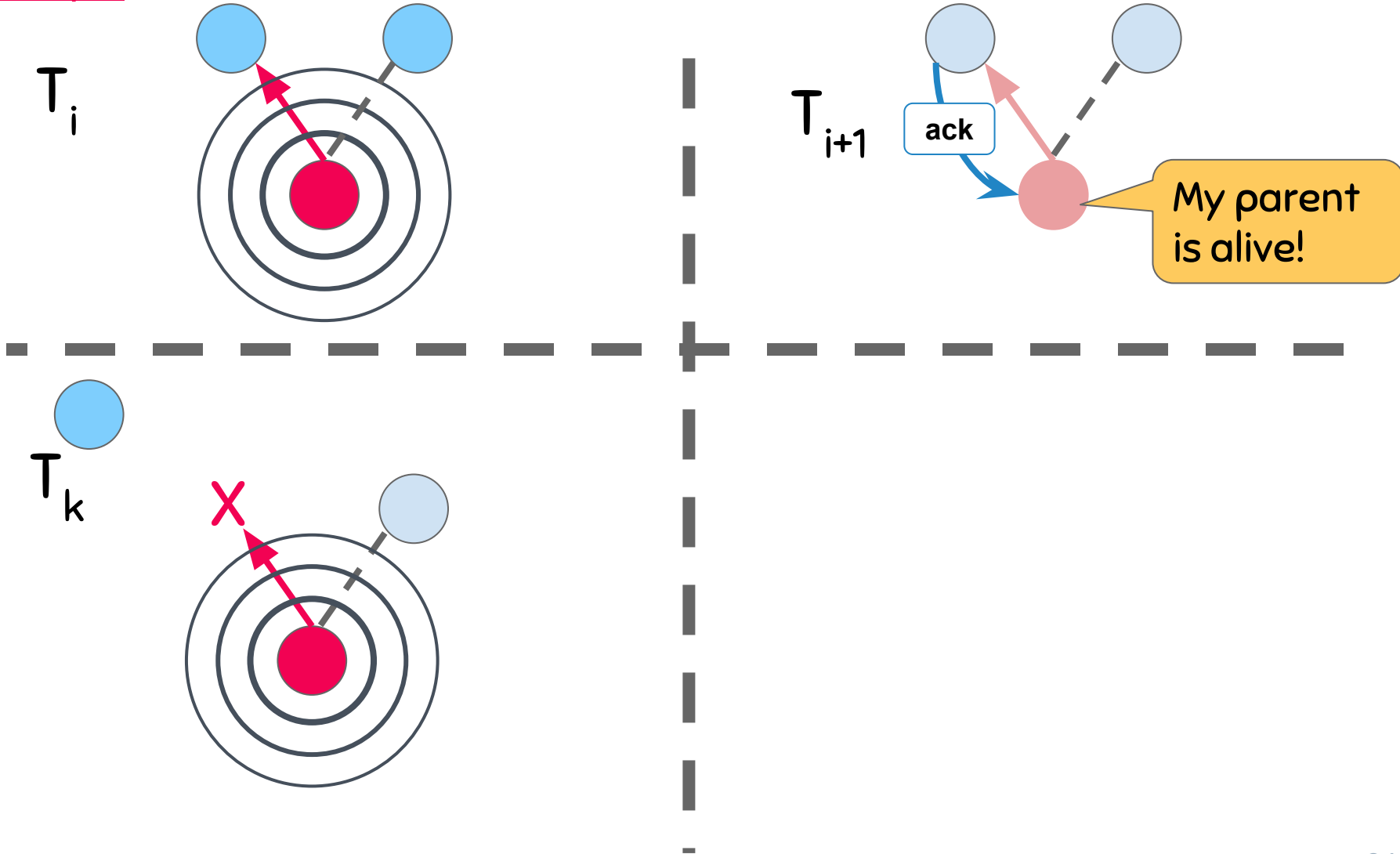
Routing under mobility events

Example. (note there are other solutions)



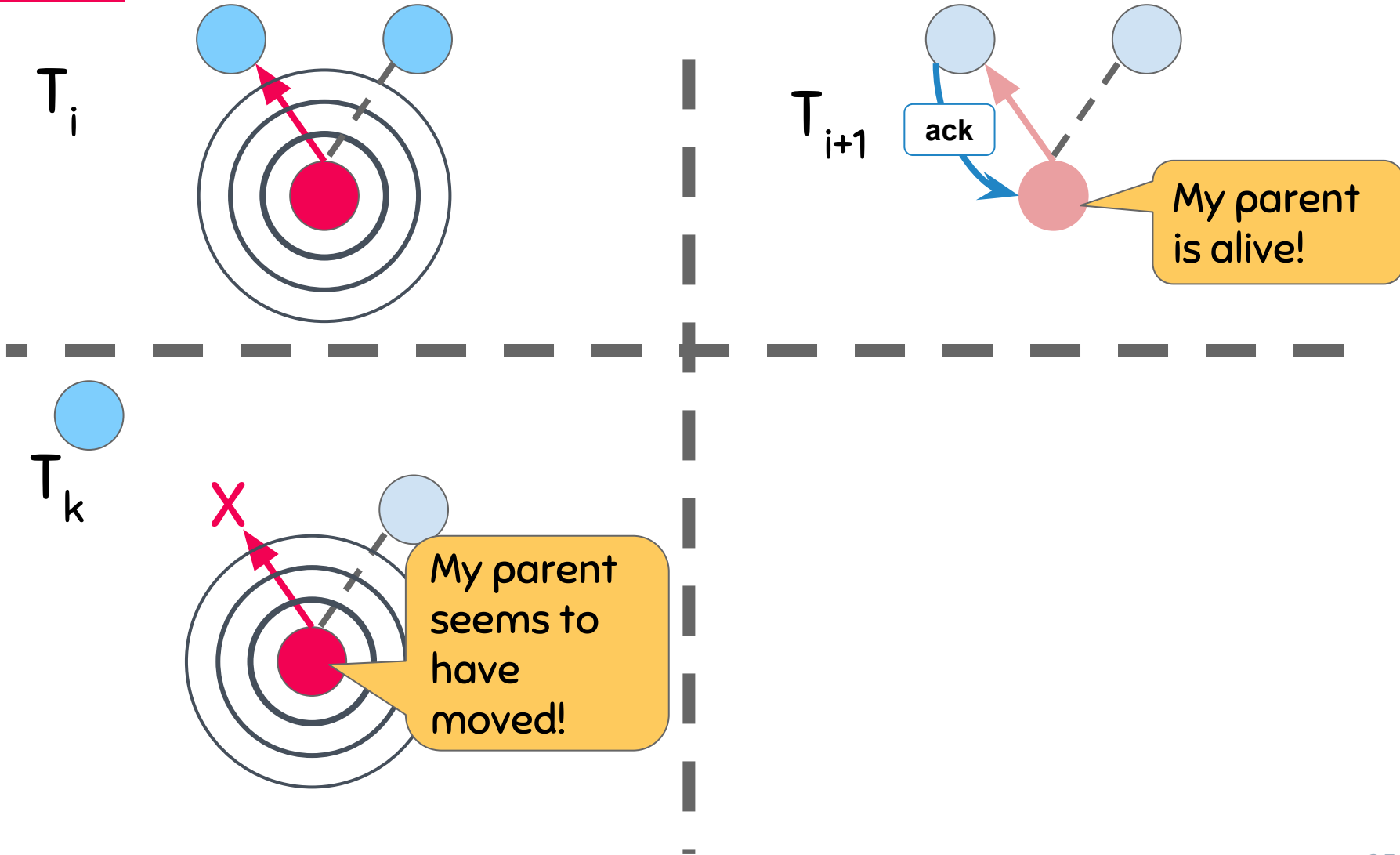
Routing under mobility events

Example. (note there are other solutions)



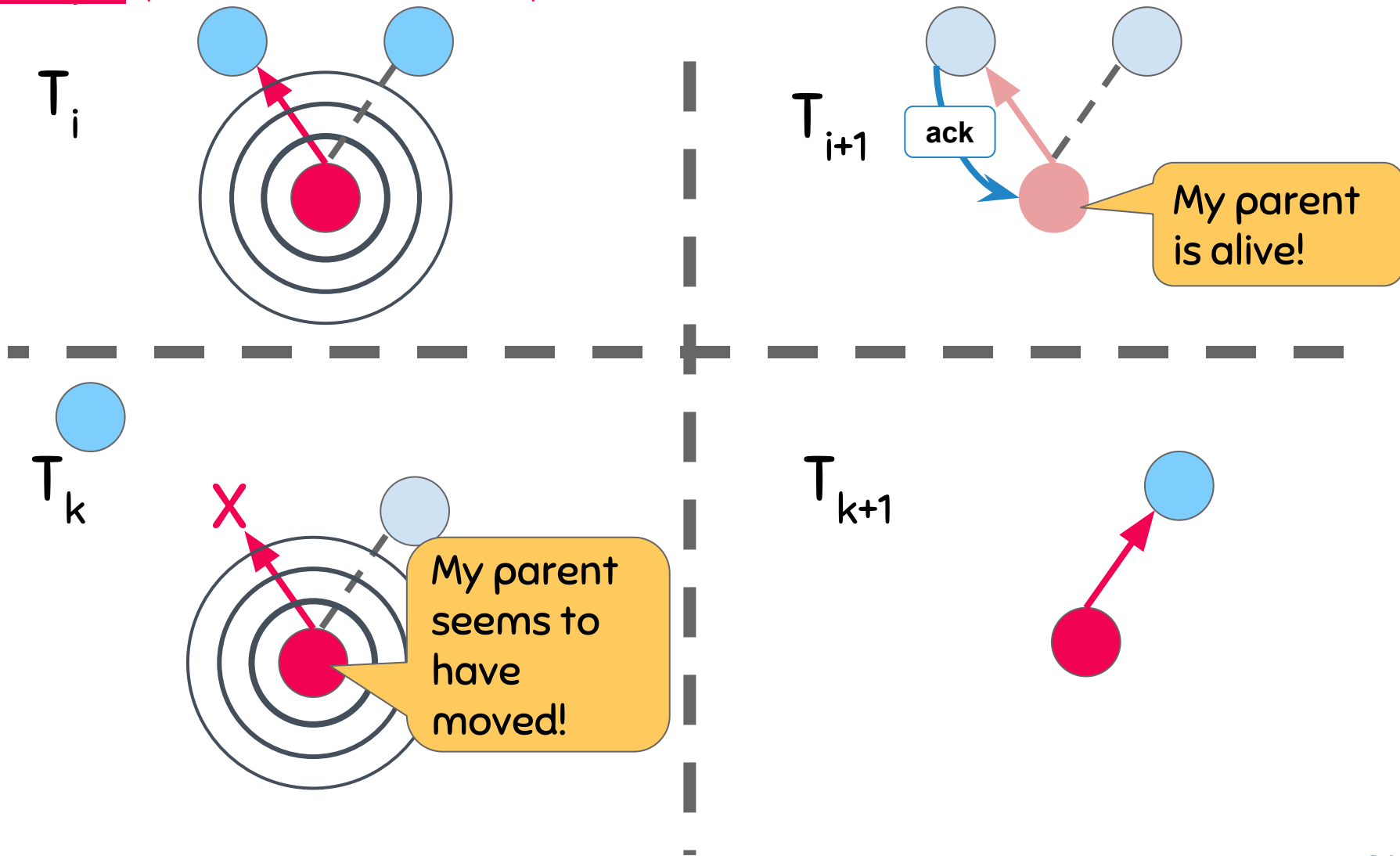
Routing under mobility events

Example. (note there are other solutions)



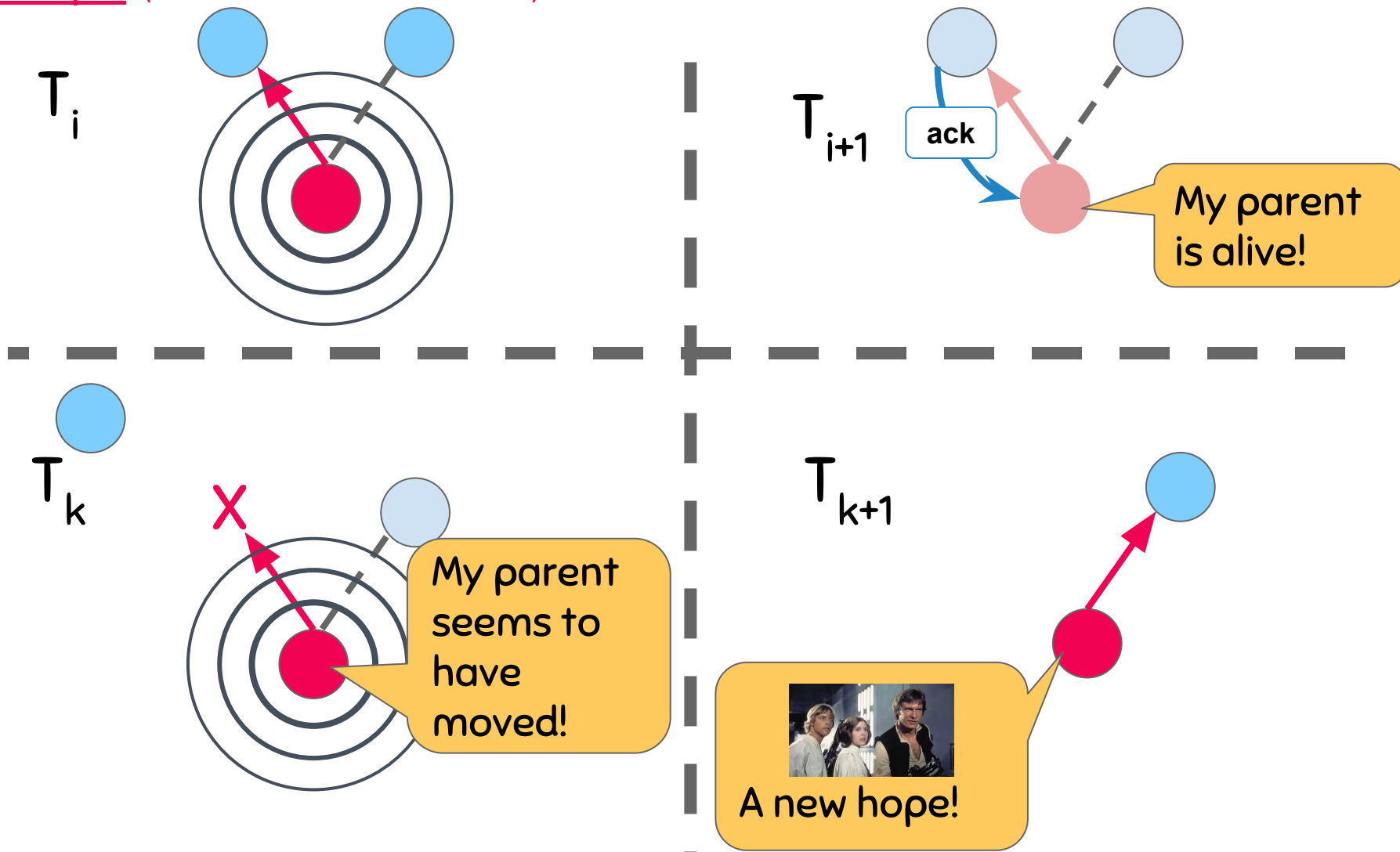
Routing under mobility events

Example. (note there are other solutions)

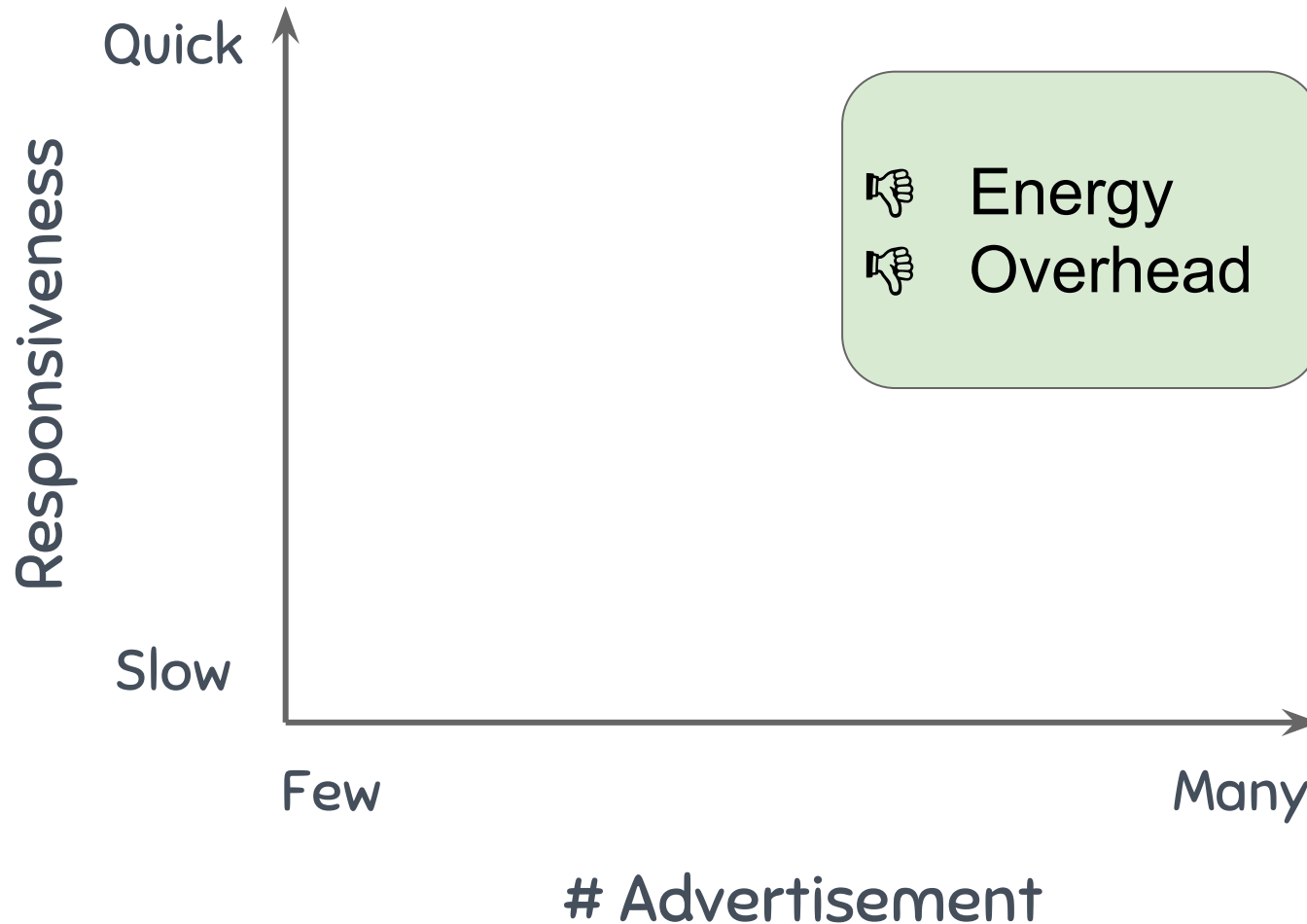


Routing under mobility events

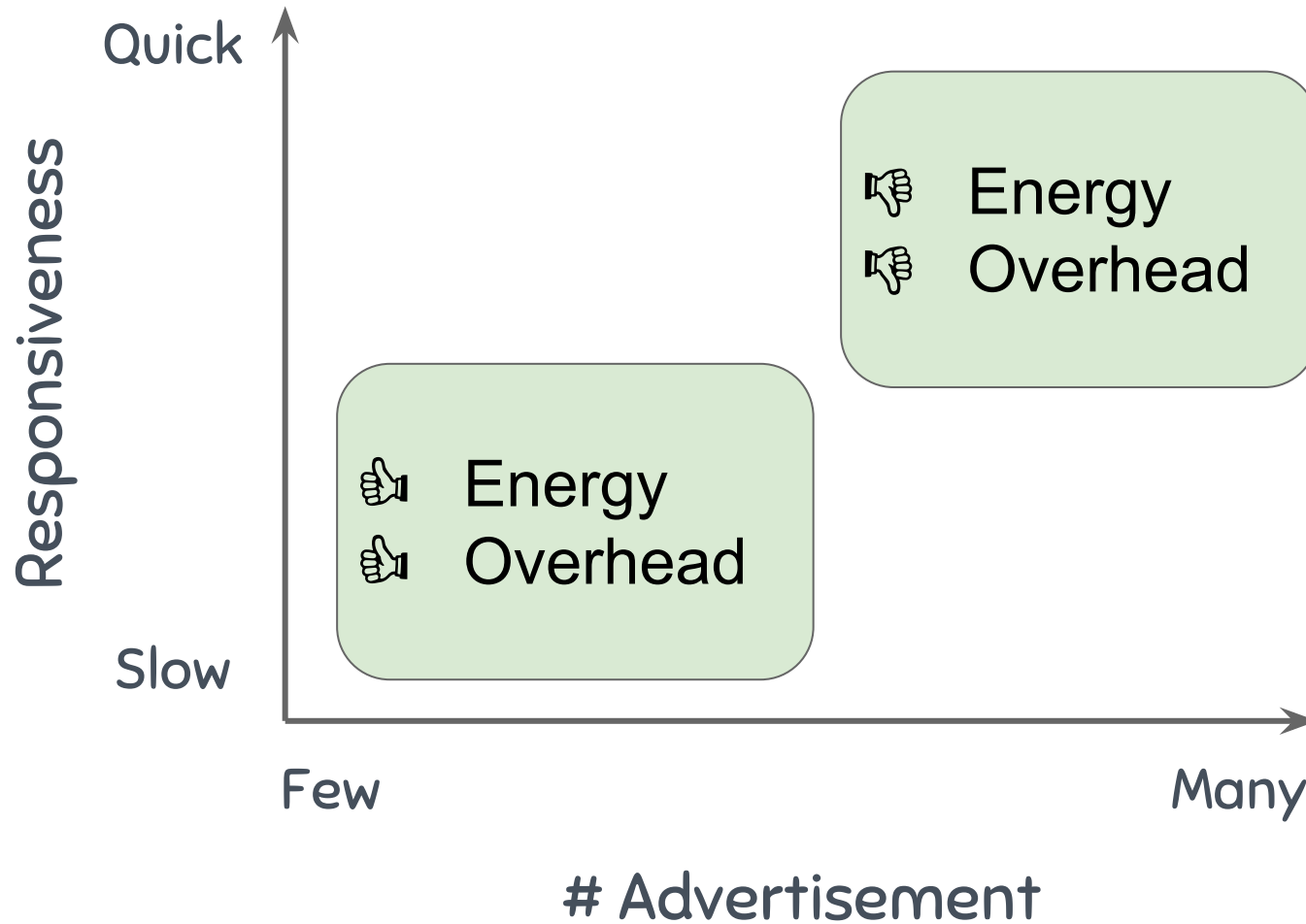
Example. (note there are other solutions)



Timer scheme trade-off



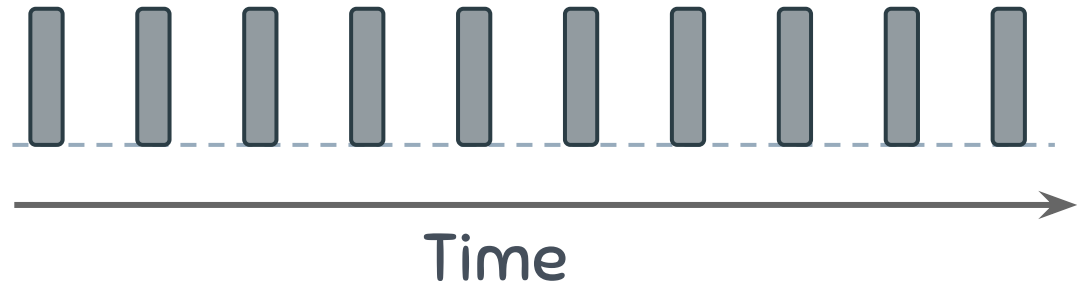
Timer scheme trade-off



Dealing with mobility and link dynamics

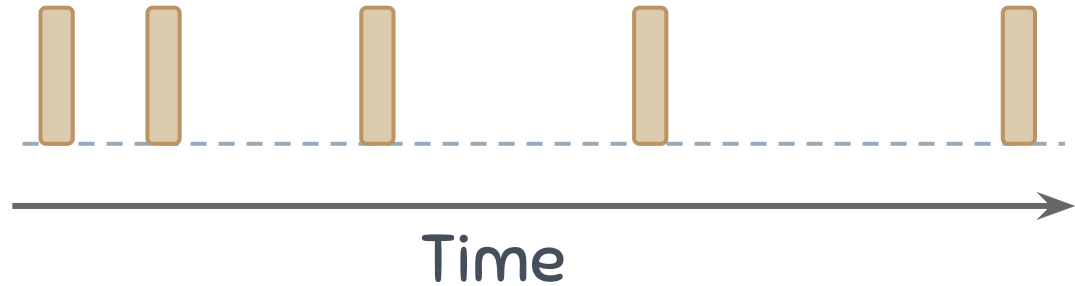
- Timer schemes
 - Control advertisements
 - Govern the communication structure construction and maintenance
- What timer schemes are most commonly used?

Background

Dealing with mobility and link dynamics**● Periodic**

- Large interval
 - 👍 Low channel and energy usage
 - 👎 Slow responsivity
- Small interval
 - 👎 High channel and energy usage
 - 👍 Quick responsivity

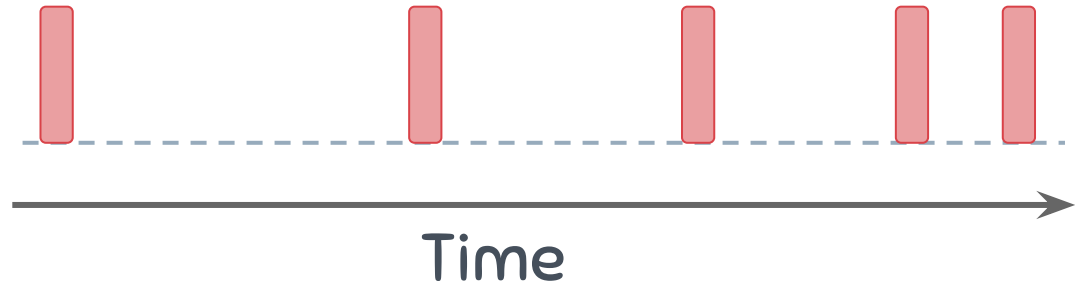
Background

Dealing with mobility and link dynamics

- **Trickle Timer**

- Assumes that network will be stable (few link changes)
- Fires bursts of advertisements when some inconsistency is detected
- Decrease advertisement rate exponentially
- Maximum interval **~2.3 h** (RFC 6550) or **~20 min** (ContikiOS)

Background

Dealing with mobility and link dynamics

- **Reverse Trickle Timer**

- The “opposite” of Trickle Timer
- Assumes that as long as a node remains connected to a parent, it is likely that node will move away
- Increase advertisement rate exponentially
- Authors use **~20 min** in their experiments

Dealing with mobility and link dynamics

1. Reverse Trickle Timer,
 2. Trickle Timer,
 3. Periodic.
- Such schemes assume:
 - 👎 Only one scheme governs the entire network
 - 👎 All devices follow the same mobility pattern

Dribble

A learn-based timer scheme selector for mobility management in IoT

- It learns the IoT device mobility pattern
- Automatically assign a proper timer scheme
 - Better balance the timer scheme trade-off

Dribble

How it works...

Start with a default
timer scheme

Ex:
Trickle Timer

Dribble

How it works...

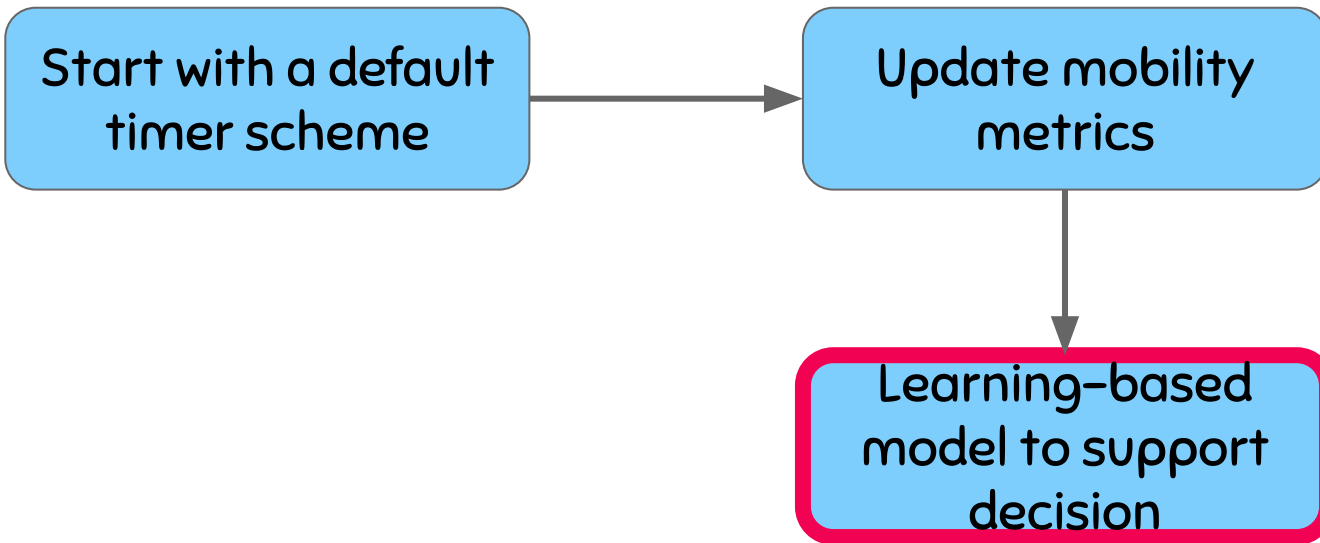


Ex:

- Speed,
- GPS,
- Travel Distance,
- Visit Time,
- Interconnection Time

Dribble

How it works...



- We've tested
 - Supervised and unsupervised models
 - But we have labeled data
- Multi-Layer Perceptron classifier as learning algorithm

$$f : R^m \rightarrow R^p$$

m is the mobility metrics
 p is the mobility patterns

Dribble

How it works...

Start with a default timer scheme

Update mobility metrics

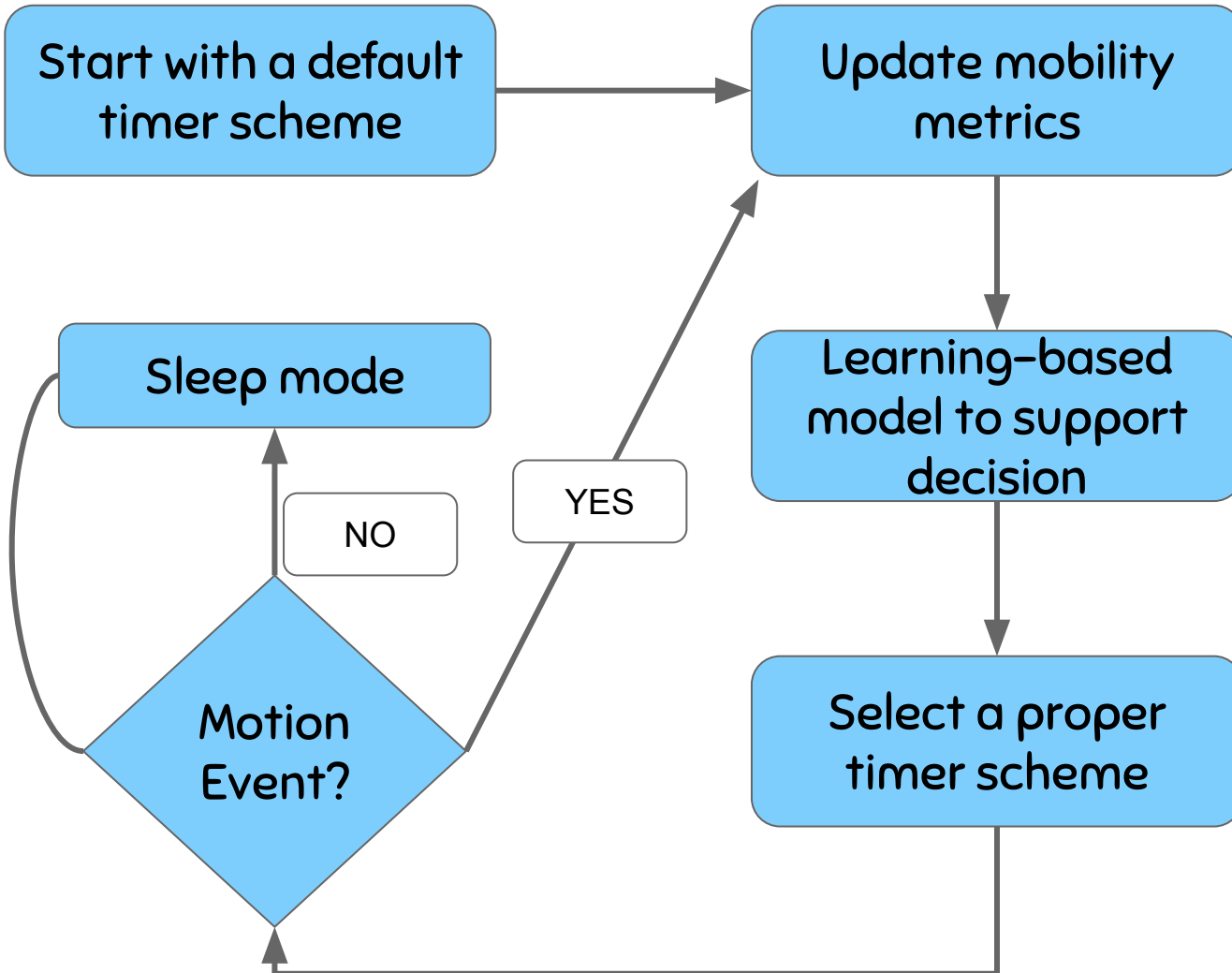
Learning-based model to support decision

Select a proper timer scheme

- Infrastructure (or almost static nodes)
 - ✓ Trickle timer
- Human behavior (assumes long stay position)
 - ✓ Reverse TT
- Non-human (high mobility)
 - ✓ Periodic

Dribble

How it works...



Experimental environment

- Sinalgo simulator
- RPL as routing protocol
 - Tree data traffic enabled: M2P, P2M, and P2P
 - Storing mode
 - ETX as Objective function

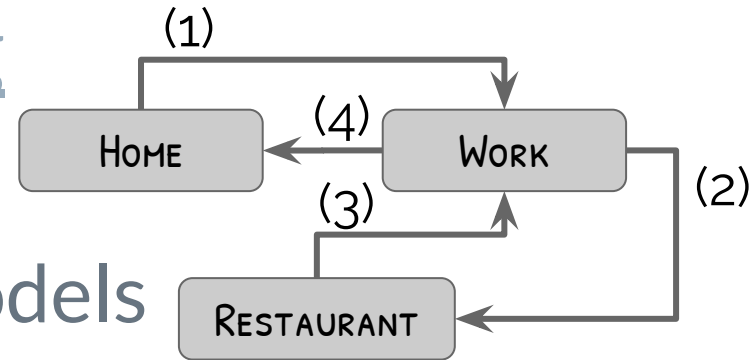
Simulation setup	
Duration	15 days
# nodes	200
Base station	1 (center)
Distribution	Random

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Simulation setup	
Duration	15 days
# nodes	200
Base station	1 (center)
Distribution	Random
DIM	1500m x 1500m (campus)
Radio Range	100 (m)
Transmission Model	CC2420-like
# random topologies	15
Timer schemes	
Trickle and Reverse Trickle timers	Min = 1s, Max = ~20 min
Periodic	60s

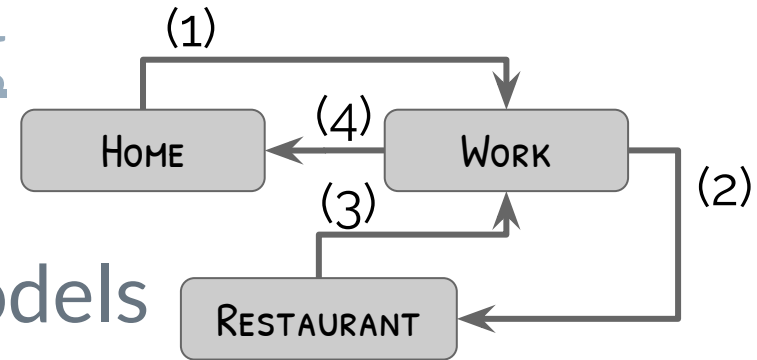
Evaluation

Device mobility modelling

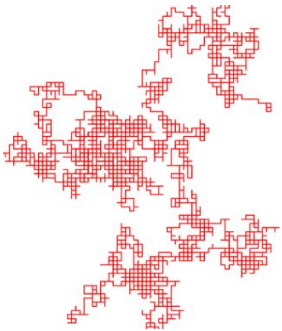


- We use two mobility models
 - Group Regularity Mobility model (GRM)
 - Human-like

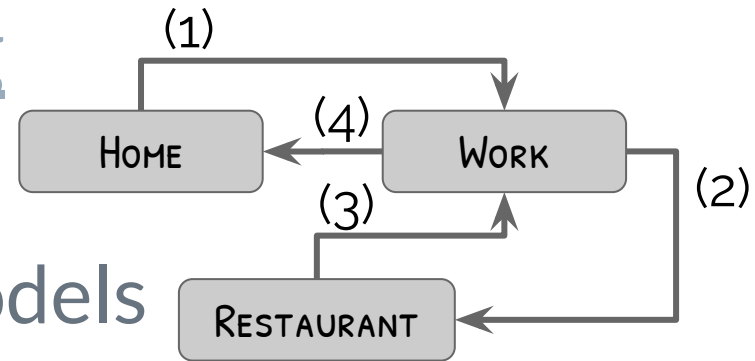
Evaluation

Device mobility modelling

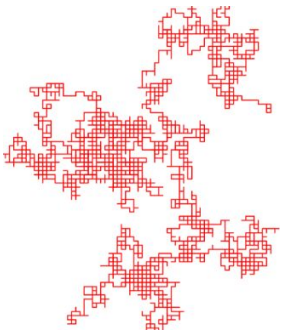
- We use two mobility models
 - Group Regularity Mobility model (GRM)
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 - Cyclical Random Waypoint Mobility Model (CRWP)
 - Non-human

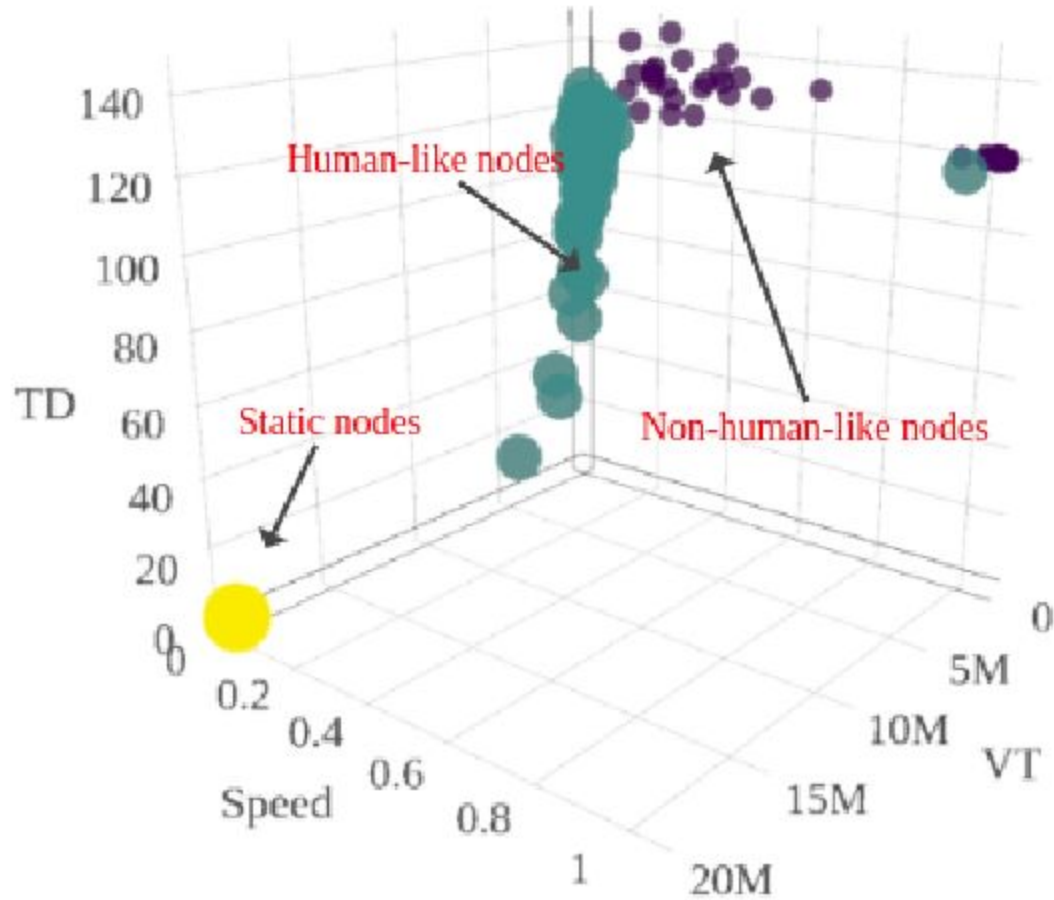


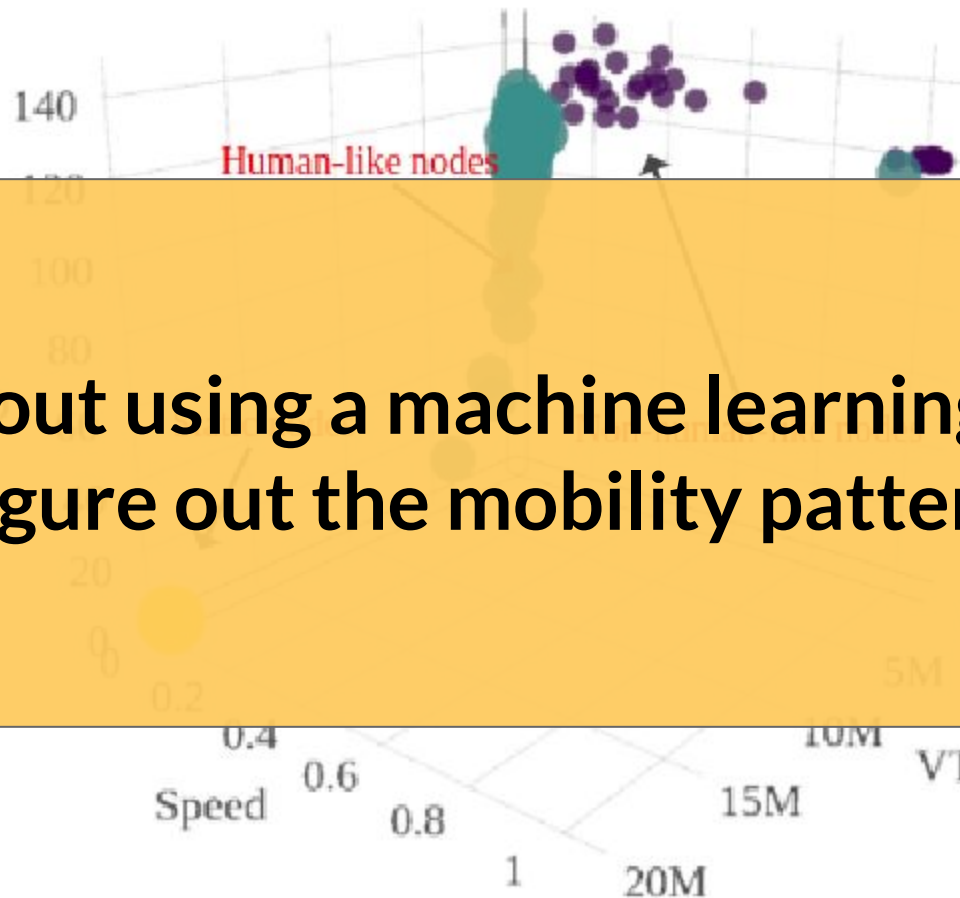
Evaluation

Device mobility modelling

- We use two mobility models
 - Group Regularity Mobility model (GRM)
 - Human-like
 - Cyclical Random Waypoint Mobility Model (CRWP)
 - Non-human
- Some static nodes to represent the infrastructure







What about using a machine learning model to figure out the mobility pattern?

Neural Network (Multi-Layer Perceptron) Architecture and parameters

Architecture	1 Hidden layer with 100 neurons

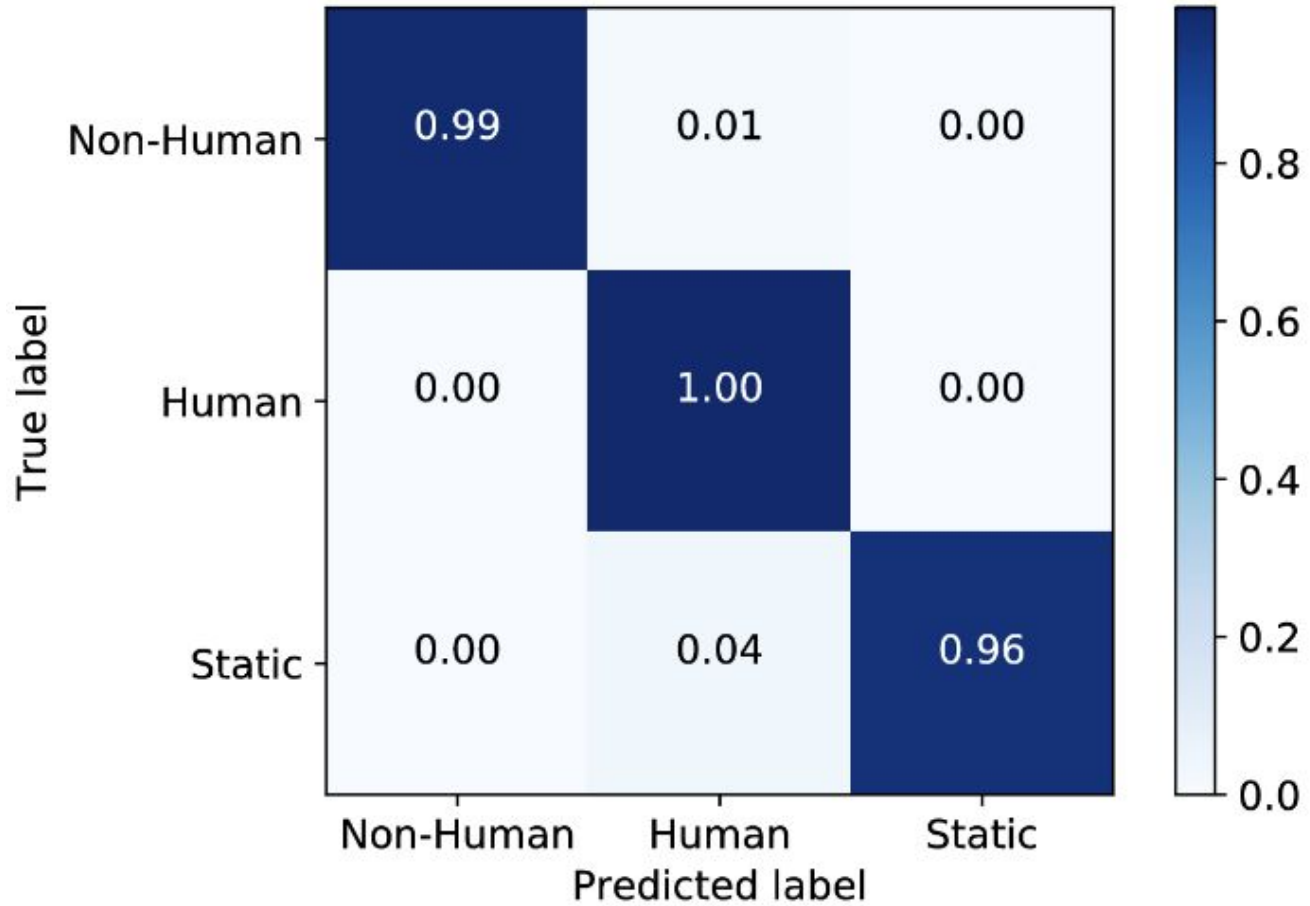
Neural Network (Multi-Layer Perceptron) Architecture and parameters

Architecture	1 Hidden layer with 100 neurons
Activation	Rectified linear unit function
Learning rate	Constant
# epochs	500
Weight optimization	Adam
Train dataset	10 random topologies
Validation model	10-fold cross-validation

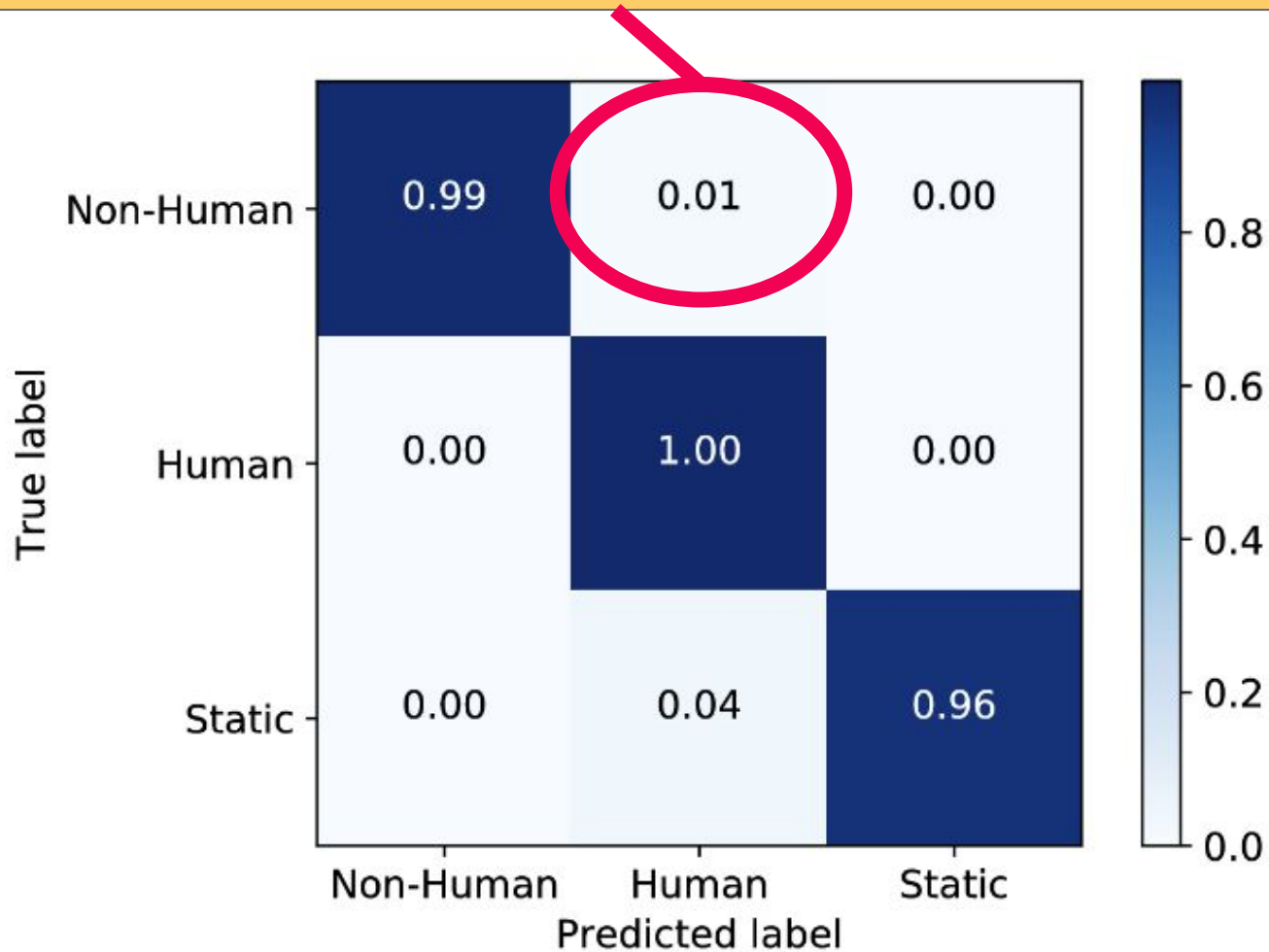
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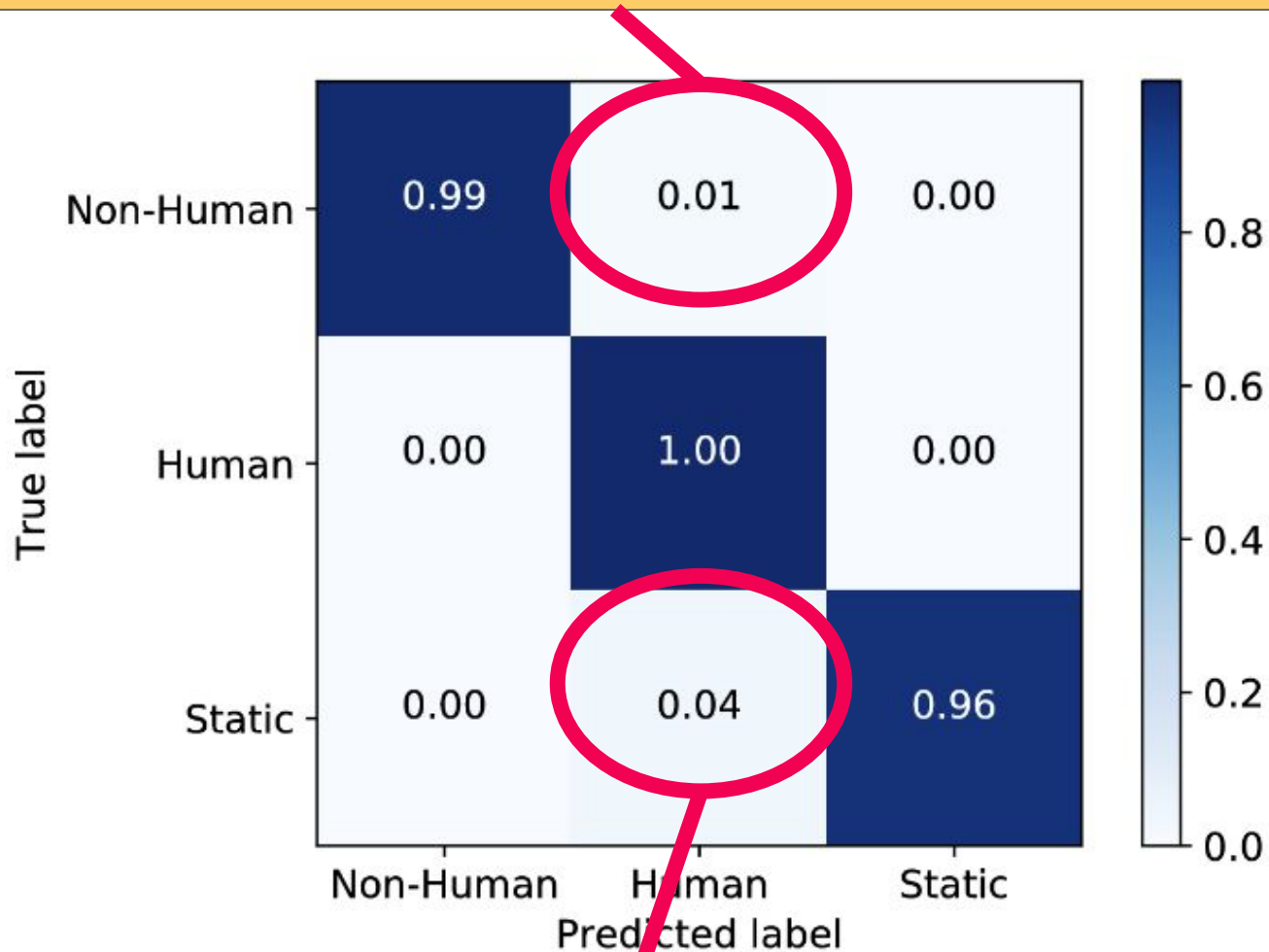
	Precision	Recall
Non-Human	1	0.99
Human	0.98	1
Static	1	0.96
Avg/Total	0.99	0.99



1% of Non-human were misclassified as human

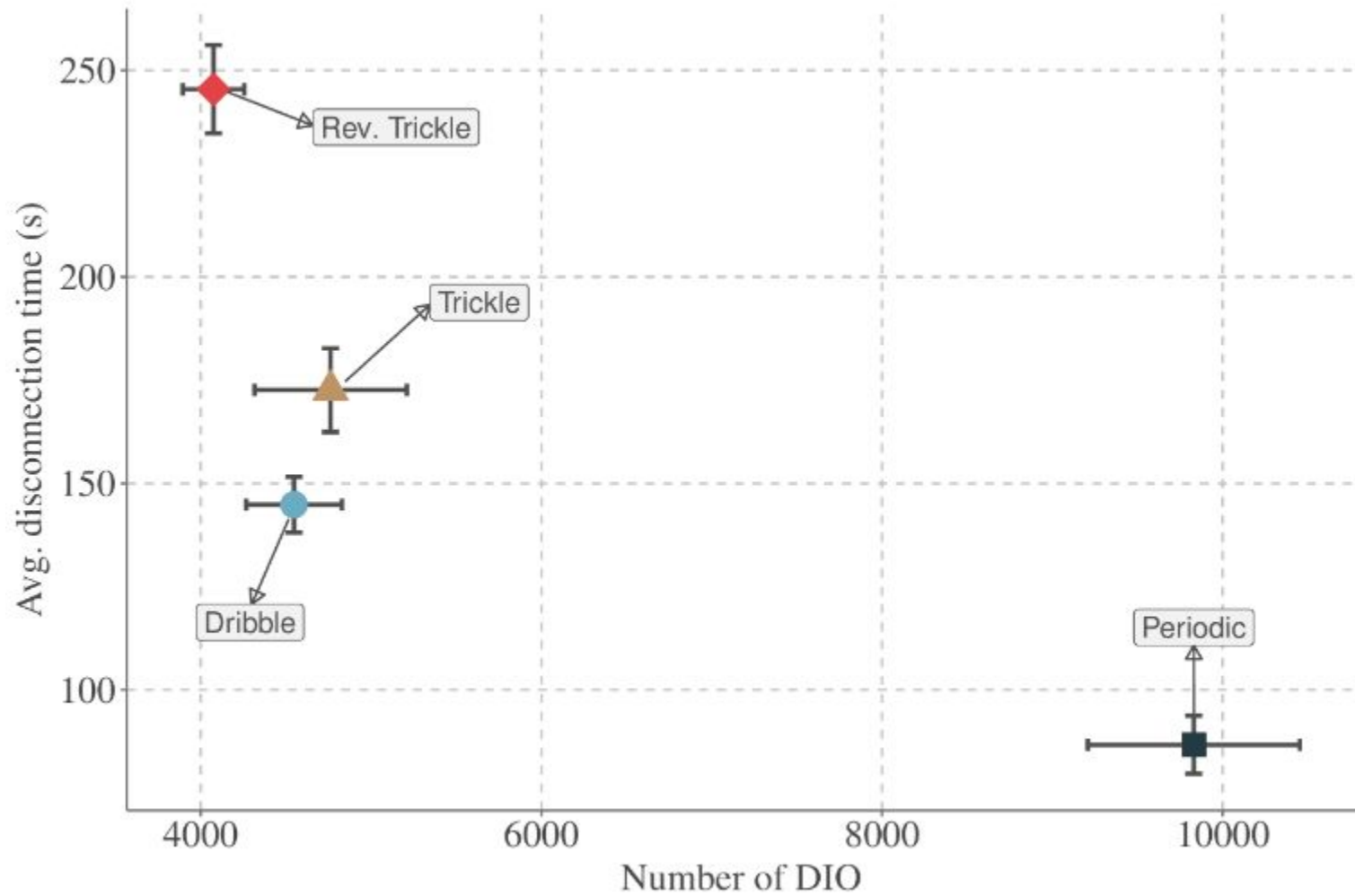


1% of Non-human were misclassified as human

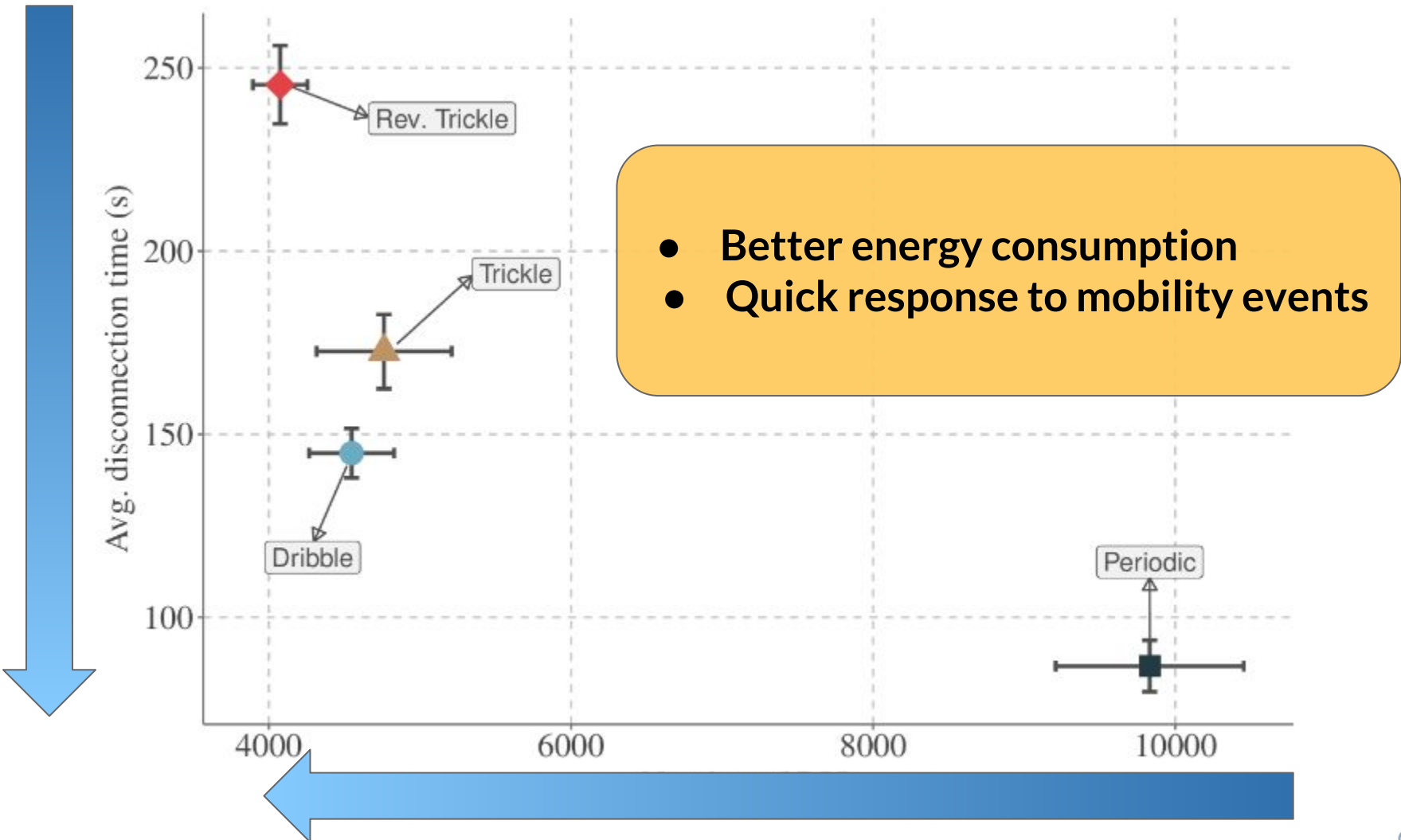


4% of static were misclassified as human

Trade-off balance

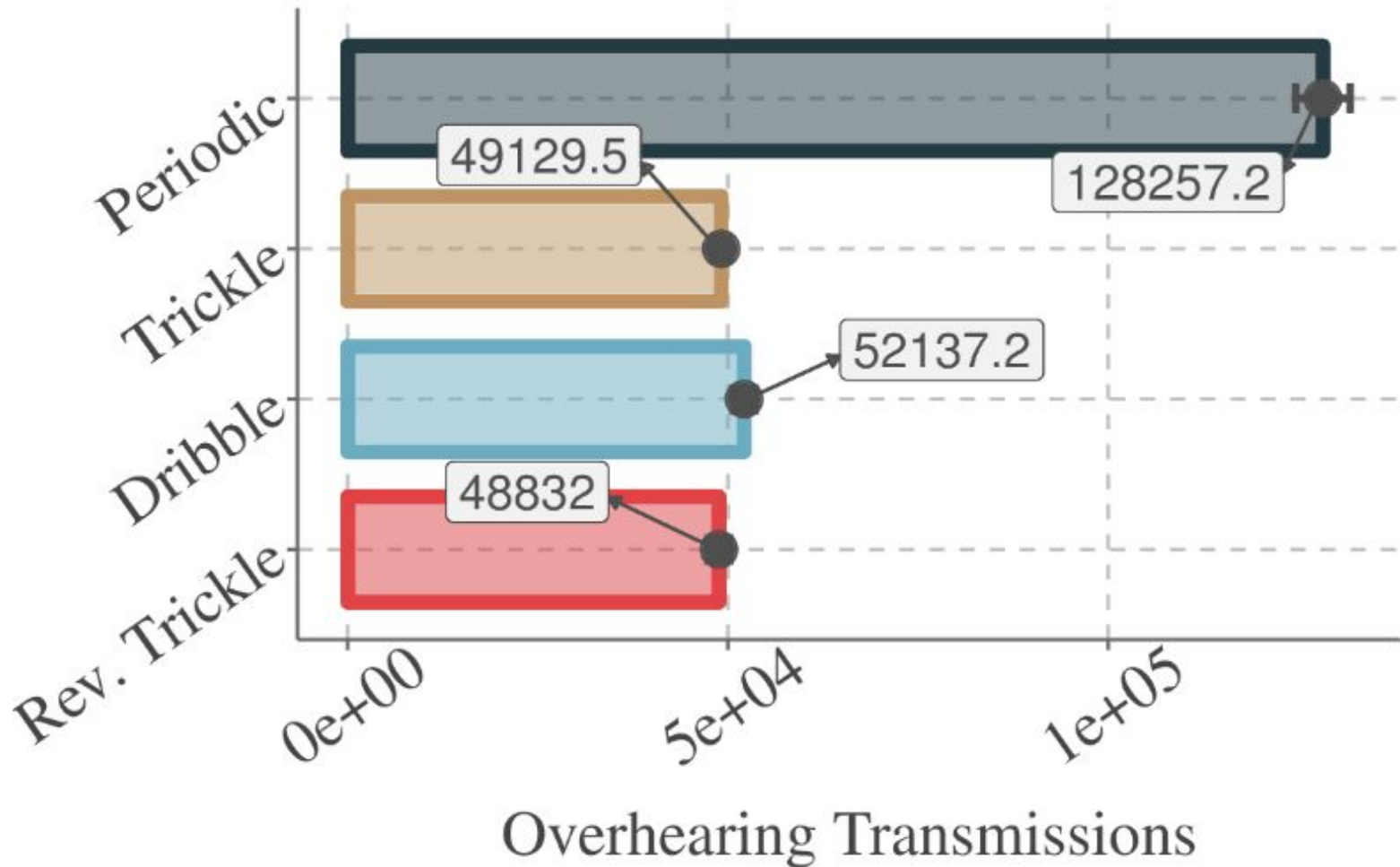


Trade-off balance



Evaluation

Overhead



Conclusion and future work

- We have proposed Dribble
 - A learn-based time scheme selector
 - It sets a custom timer scheme given the mobility pattern of a IoT device
 - Also, Dribble presented a better timer scheme trade-off balance

Conclusion and future work

- We intent to extend Dribble to support:
 - Automatic parametrization of timer schemes
 - Automatic way to associate mobility patterns to timer schemes.

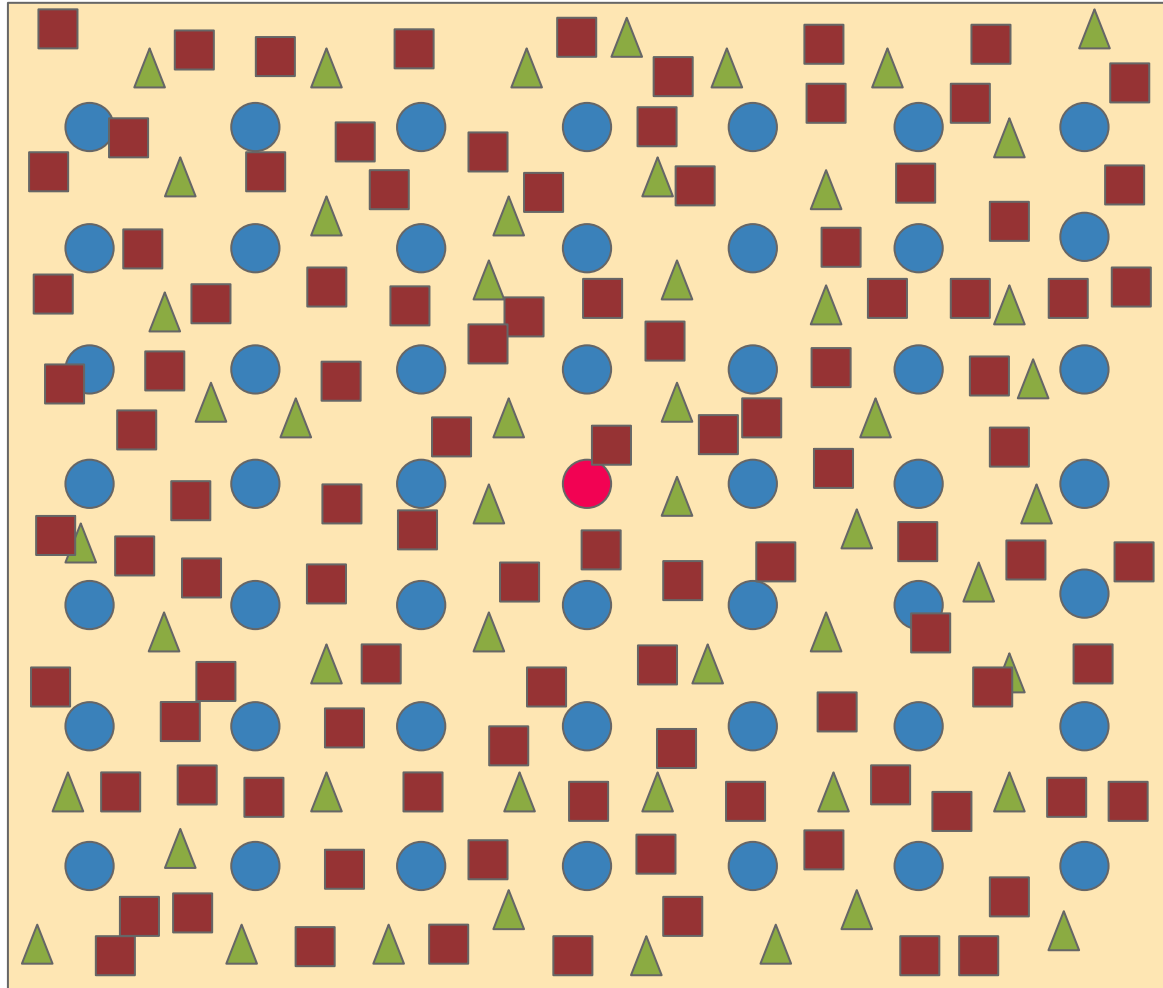
Thanks!

Any questions?

You can find us at:

- bruno.ps@dcc.ufmg.br
- rettore@dcc.ufmg.br
- lfvieira@dcc.ufmg.br
- loureiro@dcc.ufmg.br

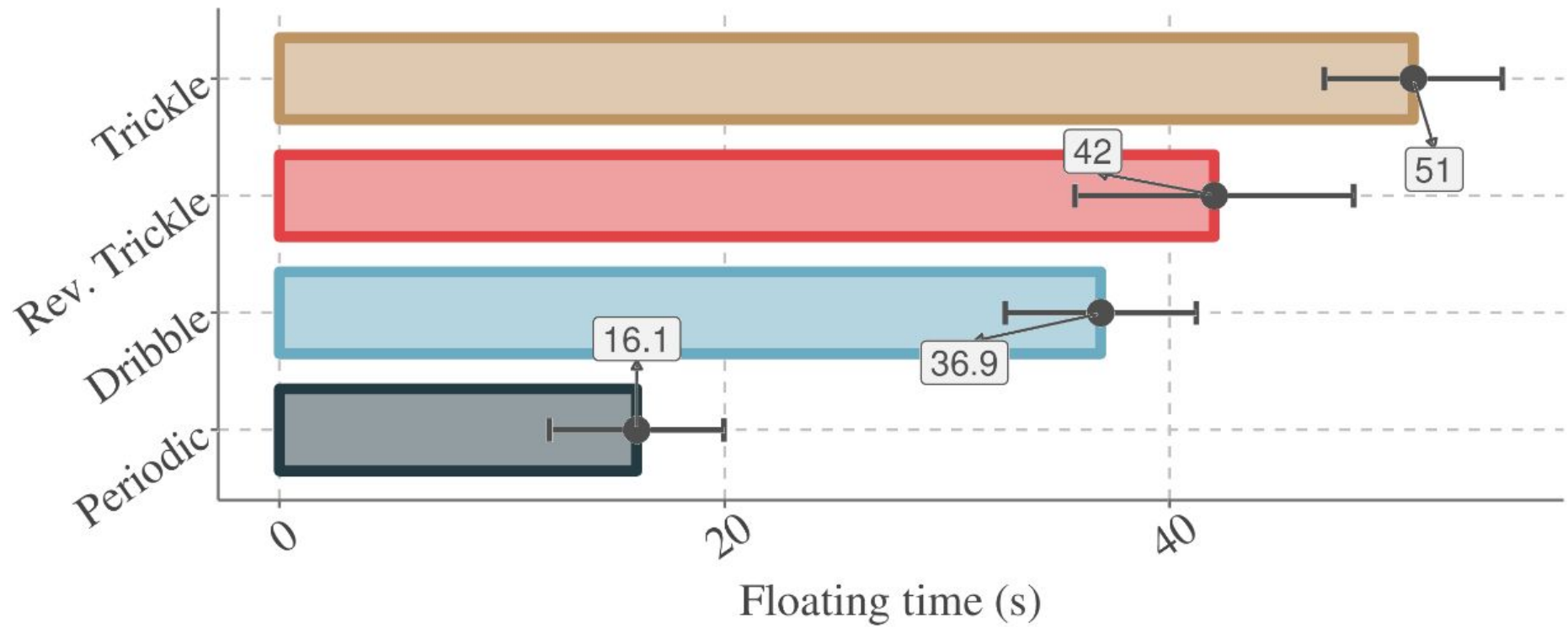
- 49 Static
- 1 BR
- ▲ 50 CRWP
- 100 GRM



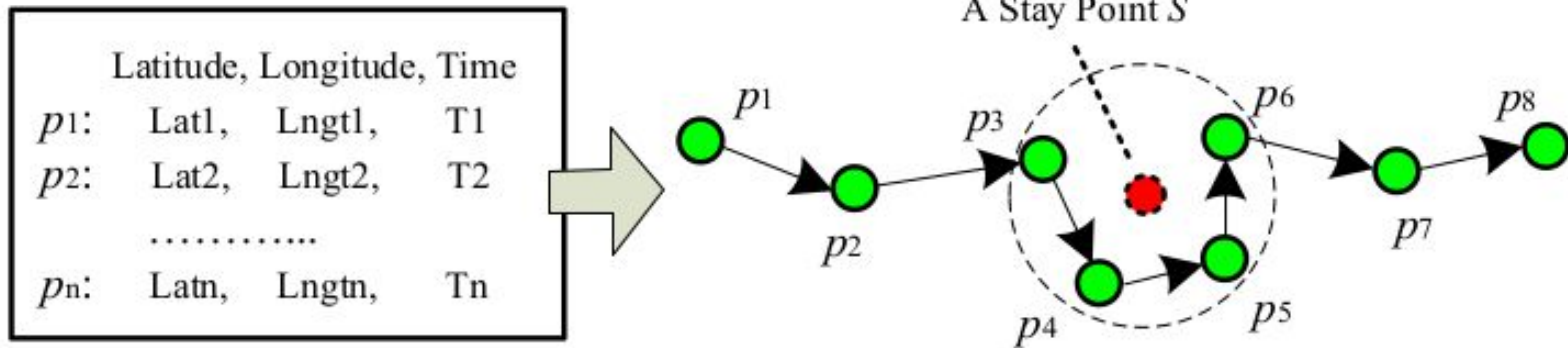
1500 m

Evaluation

Time of disconnection

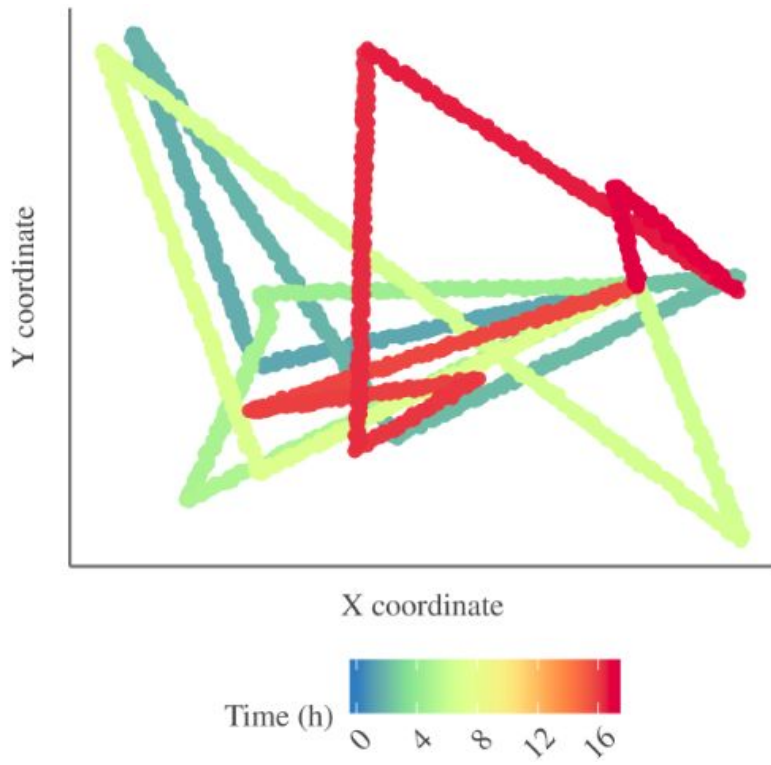


When do updates?

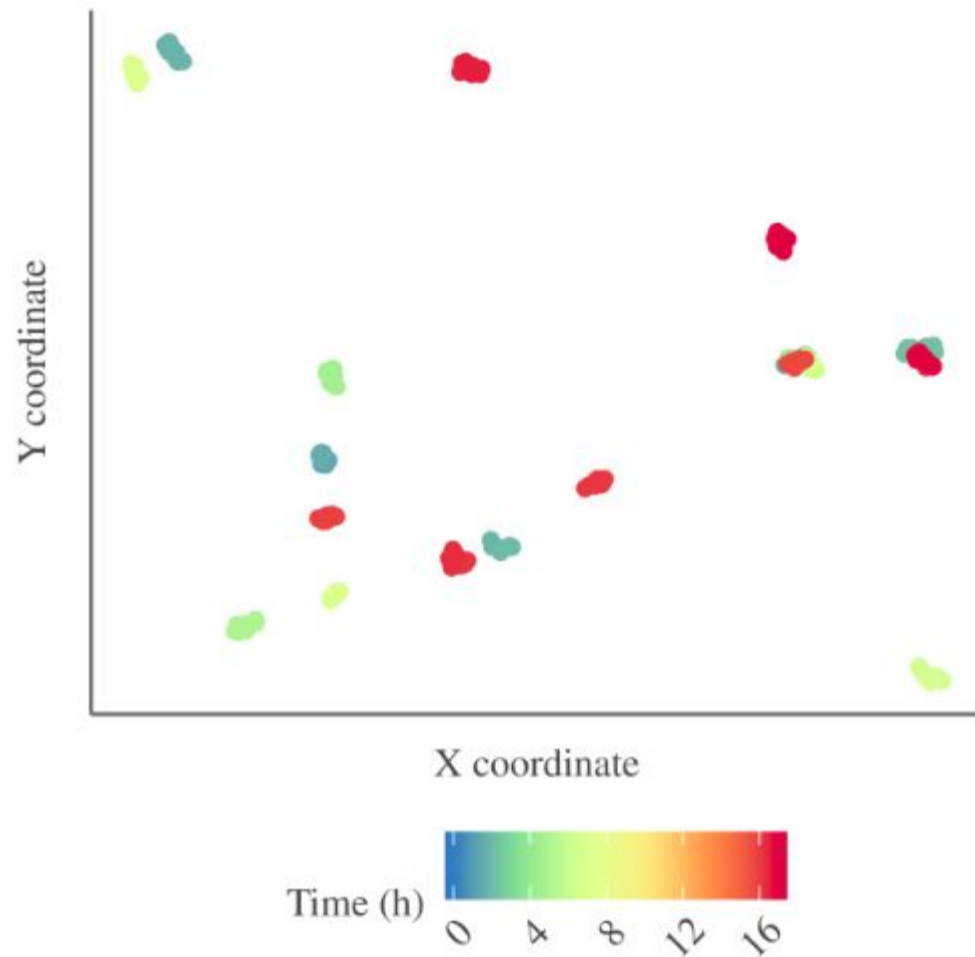
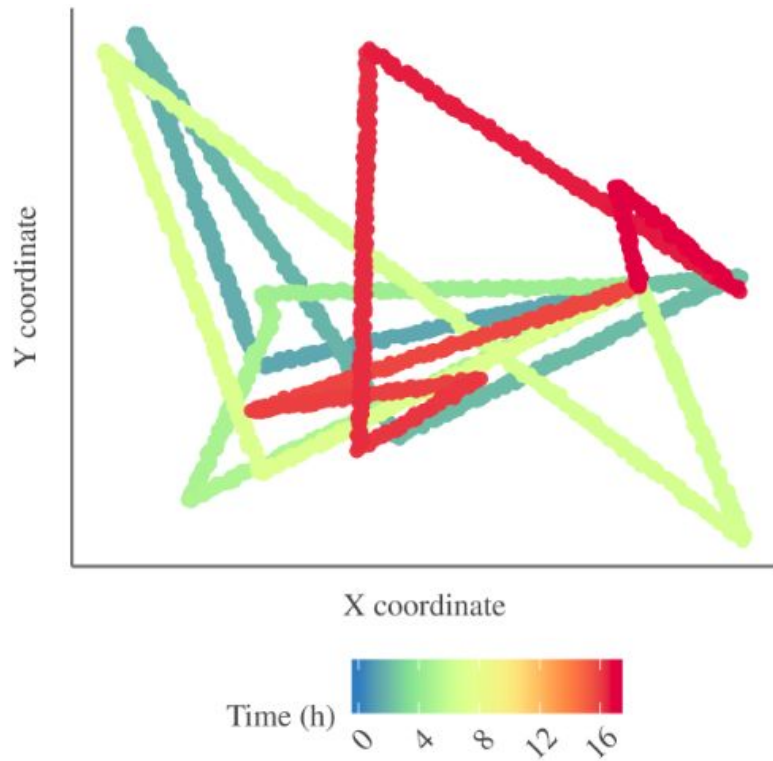


GPS points: $P = \{p_m, p_{m+1}, \dots, p_n\} \forall m < i \leq n$

- $\text{Distance}(p_m, p_n) \leq D_{\text{threshold}}$ and
- $|p_n.T - p_m.T| \geq T_{\text{threshold}}$



Raw data



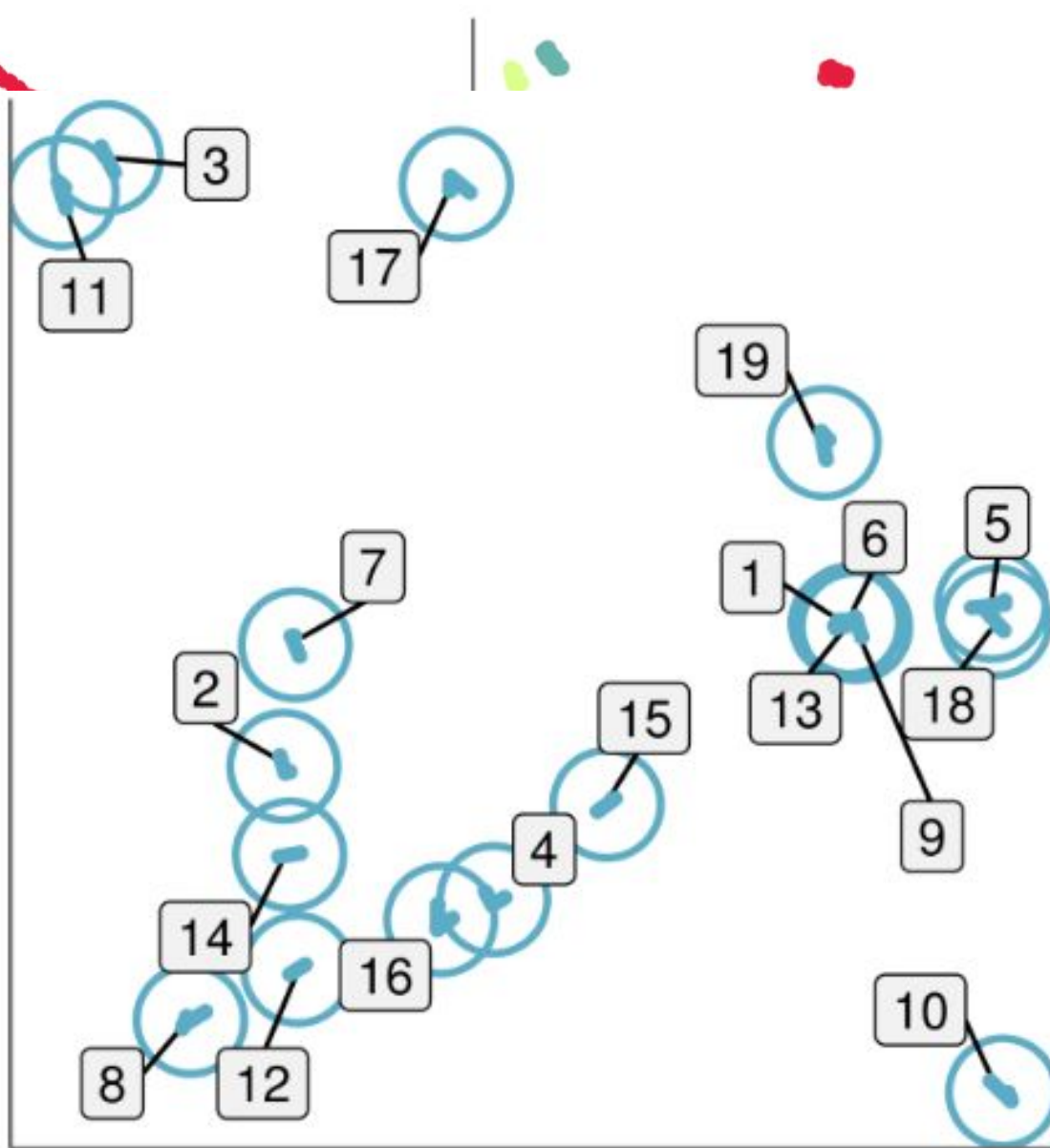
Stay locations

Y coordinate



Time (t)

Y coordinate



X coordinate

