

# Simulation of a Novel MAC Protocol on the NS-3 Platform

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- Introduction
- Objectives
- NS-3 - An Introduction
- Governing Algorithm
- Tag-Augmented Nodes
- Simulation of Readers
- Results
- Conclusions and Future Work
- References



# Introduction

## Smart Home

- Home appliances controlled remotely
- Fuelled by IoT technologies
- Battery usage increased in IoT era
- Cons of excessive battery usage
  - environmental degradation
  - stress in changing dead batteries



**Figure:** 1. Smart Home Visual (source: [www.bemi.fi](http://www.bemi.fi)).



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# Introduction (contd...)

## Battery-free Smart Home

- Home electronics still dependent on batteries ?

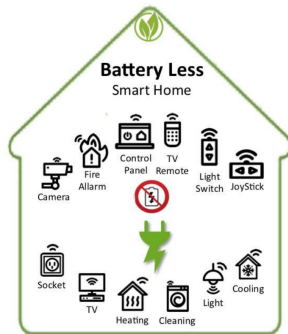


Figure: 2. Battery-free Smart Home



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# Objectives

- Introduce the NS-3 library.
- Understanding the Novel MAC protocol.
- Simulate the various battery-free home electronic devices.
- Simulate the Novel MAC protocol on the reader.
- Performance analysis of the MAC protocol.



# NS-3 - An Introduction<sup>1</sup>

## The NS-3 Library

- A discrete event network simulator
- C++ and Python
- *Waf* build system
- Abstractions :
  - Node
  - Application
  - Channel
  - Transport
  - Network
  - Data Link
- Helpers

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<sup>1</sup>NS-3 Project. *NS-3: Conceptual Overview*. URL:  
<https://www.nsnam.org/docs/release/3.30/tutorial/html/conceptual-overview.html>.



# Governing Algorithm

## Multi-Arm Bandit Problem<sup>2</sup>

- Multiple actions with unknown reward to choose from
- Goal is to maximize reward margin through series of actions
- Example: Website advertisement to visitors
- Exploitation and Exploration
- Action-Reward function:




$$\theta^* = Q(a^*) = \max_{a \in A} Q(a) \quad (1)$$

Where:

$$Q(a_t) = \mathbb{E}[R|A = a] = \theta$$

$\theta^*$  optimal probability

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<sup>2</sup>Richard S. Sutton and Andrew G. Barto. *Reinforcement Learning, An Introduction*. Second. Cambridge, Massachusetts London, England: The MIT Press, 2018.    7/29



# Governing Algorithm (contd...)

## APT-MAC Protocol<sup>3</sup>

- Translates into Multi-Arm Bandit as:
  - Reader: agent performing actions
  - Set of Actions: query tag<sub>i</sub> , query tag<sub>j</sub> ...
  - State set: ready to perform new query
  - Computing expected reward of each action
  - Keep record of the reward of each action

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<sup>3</sup>Gaia Maselli, Mauro Piva, and John A. Stankovic. "Adaptive Communication for Battery-Free Devices in Smart Homes". In: *IEEE Internet of Things (2019)*.  8/29





# Governing Algorithm (contd...) I

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## Algorithm 1: APT-MAC pseudocode

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```
1: Master M                                ▷ Reader
2: Set D                                    ▷ Devices
3: Map R:( d∈D) →float                    ▷ Reward Devices Map
4: Set MinQD                               ▷ Minimum Query Delay
5: Maximum Query Delay MaxQD
6: Set DLTH                                ▷ Data Loss Threshold
7: for d ∈ D do
8:   R[d] = 1.0
9:   Set MaxQD
10: end for
11: R = softmax( R)
12: while true do
```



# Governing Algorithm (contd...) II

```
13:   Device next = chooseNext( R)
14:   Bool goodQuery = M.query( next)
15:   if goodQuery then
16:       R[next] =updateReward( next,bonus)
17:       if getDataLoss < DLTH then
18:           updateMaxQD
19:       end if
20:   else
21:       R[next] =updateReward( next,malus)
22:   end if
23:   R = softmax( R)
24: end while
```



# Governing Algorithm (contd...)

## APT-MAC Protocol

- Reward is updated with function:

$$Q(a_i)(n+1) = Q(a_i)(n) + \alpha(Reward - Q(a_i)(n)) \quad (2)$$

Where:

$\alpha$  = learning rate



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# Tag-Augmented Nodes

## Markov Representation

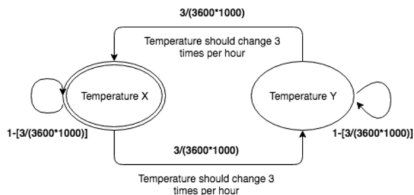


Figure: Temperature Sensor model

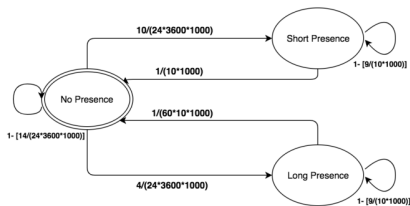


Figure: Presence Sensor model



# Tag-Augmented Nodes (contd...)

## Markov Representation

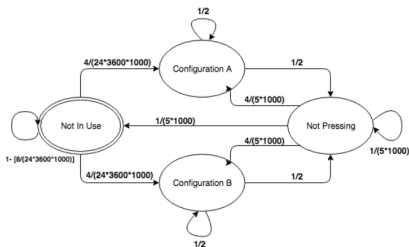


Figure: TV Remote model

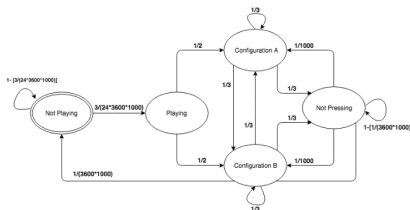


Figure: Joystick model



# Tag-Augmented Nodes (contd...)

## Packet Header

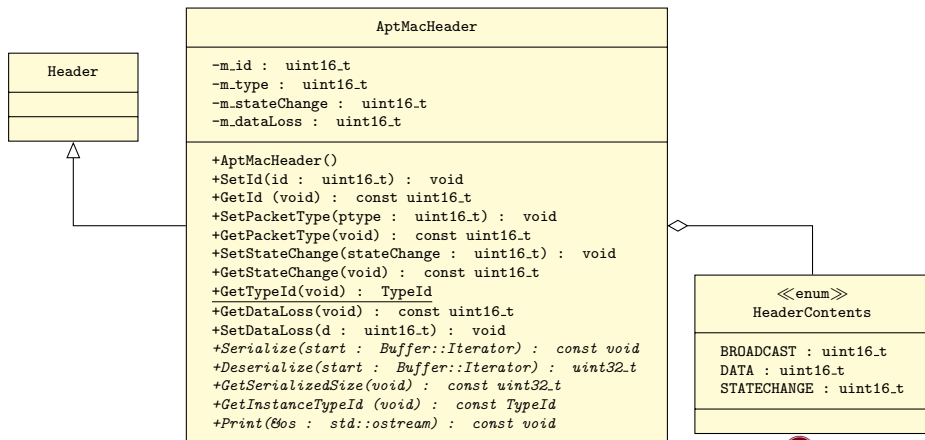


Figure: Class Diagram - APT-MAC Header



# Tag-Augmented Nodes (contd...)

## Devices Simulation



Figure: Class Diagram - Sensor Nodes

# Tag-Augmented Nodes (contd...)

## Devices Simulation

- Enumeration of some functions :
  - StartApplication
  - StopApplication
  - HandleBroadcast
  - HandleDataQuery
  - GeneratePacket
  - SchedulePacketGenerate
  - NextState
  - Send
  - UpdateNewDataSent
  - UpdateNewDataProduced
  - GetQueryCounter





# Simulation of Readers

## APT-MAC Reader



# Simulation of Readers (contd. . .)

## TDMA Reader



# Simulation of Readers (contd. . . )

## Run Configuration.

- Data rate of the net-device and channel set to *640kbps*.
- Tag memory 256 bits



## Performance Metrics

- Packet Delay :  
the difference in time between generating a packet data and sending to the reader.
- Data Loss :  
this is the percentage of generated data that is not delivered to the reader.



**Table:** WORKLOAD SCENARIOS DESCRIPTION

No. of Sensors	Scenario	Joystick	Remote	Env. Sensors
20	Case 1	1	2	17
	Case 2	2	3	15
	Case 3	3	3	14
	Case 4	4	4	12
30	Case 1	1	2	27
	Case 2	2	3	25
	Case 3	3	3	24
	Case 4	4	4	22
40	Case 1	1	2	37
	Case 2	2	3	35
	Case 3	3	3	34
	Case 4	4	4	32



# Results (contd...)

Transient time :

duration for APT-MAC device to get into optimum performance of reduced data loss.

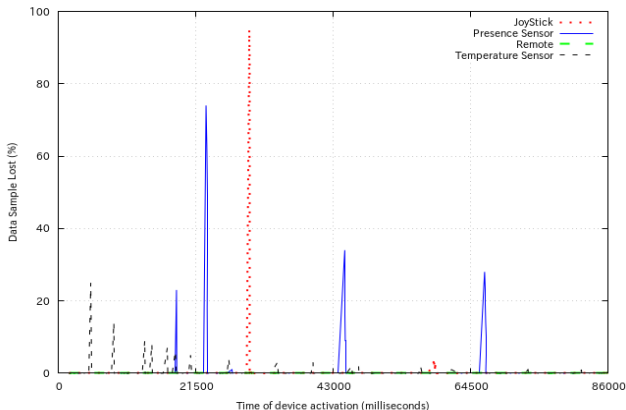


Figure: Transient Time



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# Results (contd...)

## Transient time

- approximately 50ms to get the data loss of a joystick to less than 15%
- 0.283s for presence sensor
- temperature sensor took approximately a milli of a second.



# Results (contd...)

## Performance based on uptime - 20 Devices

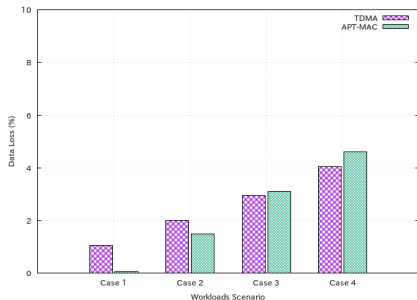


Figure: Data Loss Short Run

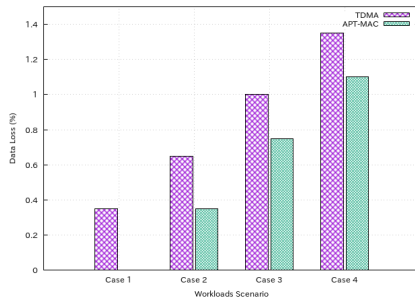


Figure: Data Loss





# Results (contd...)

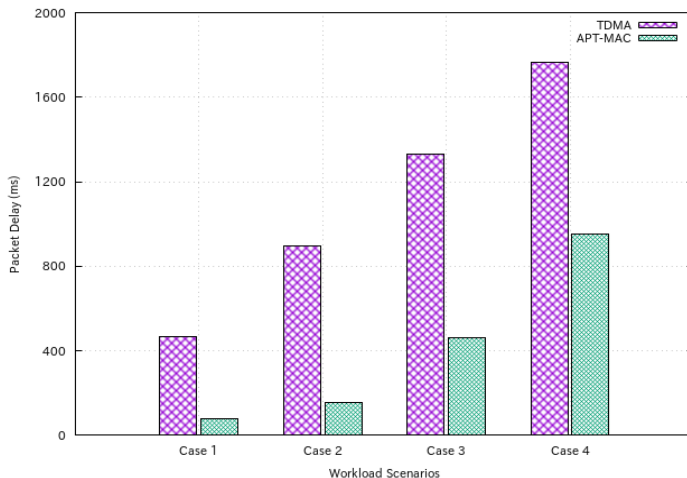


Figure: 20 Devices : Packet Delay



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# Results (contd...)

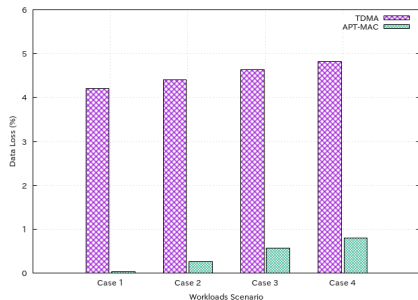


Figure: 30 Devices : Data Loss

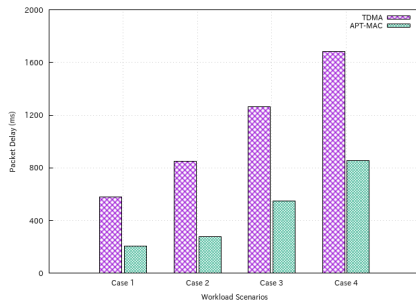


Figure: 30 Devices : Packet Delay



# Results (contd...)

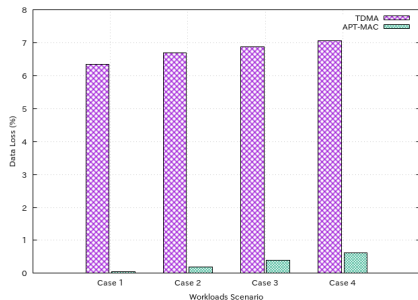


Figure: 40 Devices : Data Loss

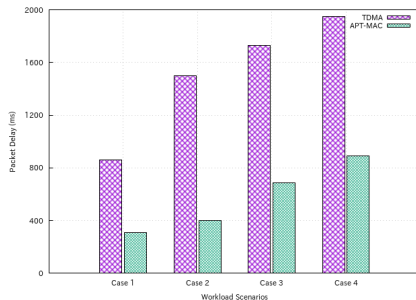


Figure: 40 Devices : Packet Delay



# Conclusions and Future Work

## Conclusion

- Objectives met
- APT-MAC outperforms TDMA

## Future Work

- Mobility of the devices with respect to APT-MAC
- other variants of reinforcement algorithms



# References

- Maselli, Gaia, Mauro Piva, and John A. Stankovic. "Adaptive Communication for Battery-Free Devices in Smart Homes". In: *IEEE Internet of Things* (2019).
- Project, NS-3. *NS-3: Conceptual Overview*. URL: <https://www.nsnam.org/docs/release/3.30/tutorial/html/conceptual-overview.html>.
- Sutton, Richard S. and Andrew G. Barto. *Reinforcement Learning, An Introduction*. Second. Cambridge, Massachusetts London, England: The MIT Press, 2018.

