

Inria

Lightweight network
interface selection for
reliable communications
in multi-technologies
wireless sensor networks

Brandon Foubert
Nathalie Mitton

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Wireless sensor networks: a tool to help farmers

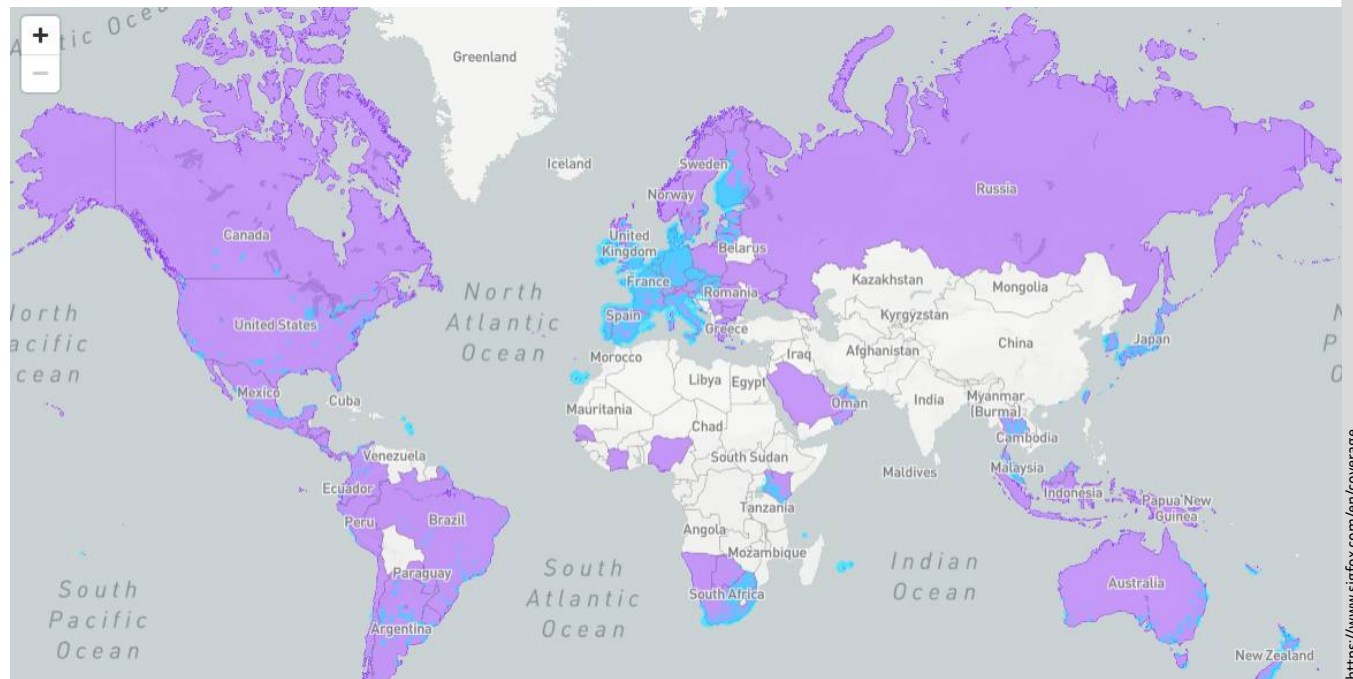
- Automate the collection of climate data
- Prevent risks (e.g. frost)
- Decision making (e.g. better use of pesticides)



<https://www.terre-net.fr/actualite-agricole/economie-sociale/article/revenus-agricoles-2017-en-hausse-202-133225.html>

Each Radio Access Technology (RAT) have limitations, *e.g.* Sigfox

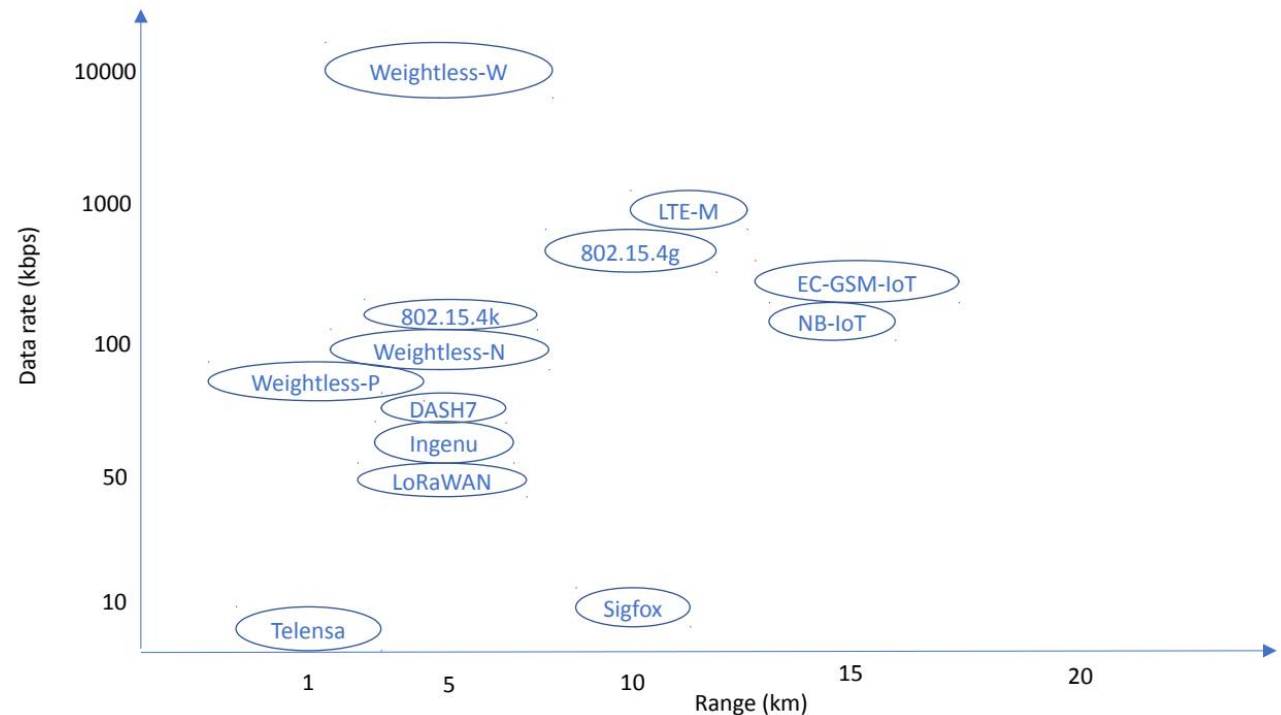
- Max 12 bytes per message
- Max 140 messages per day
- Limited worldwide coverage



<https://www.sigfox.com/en/coverage>

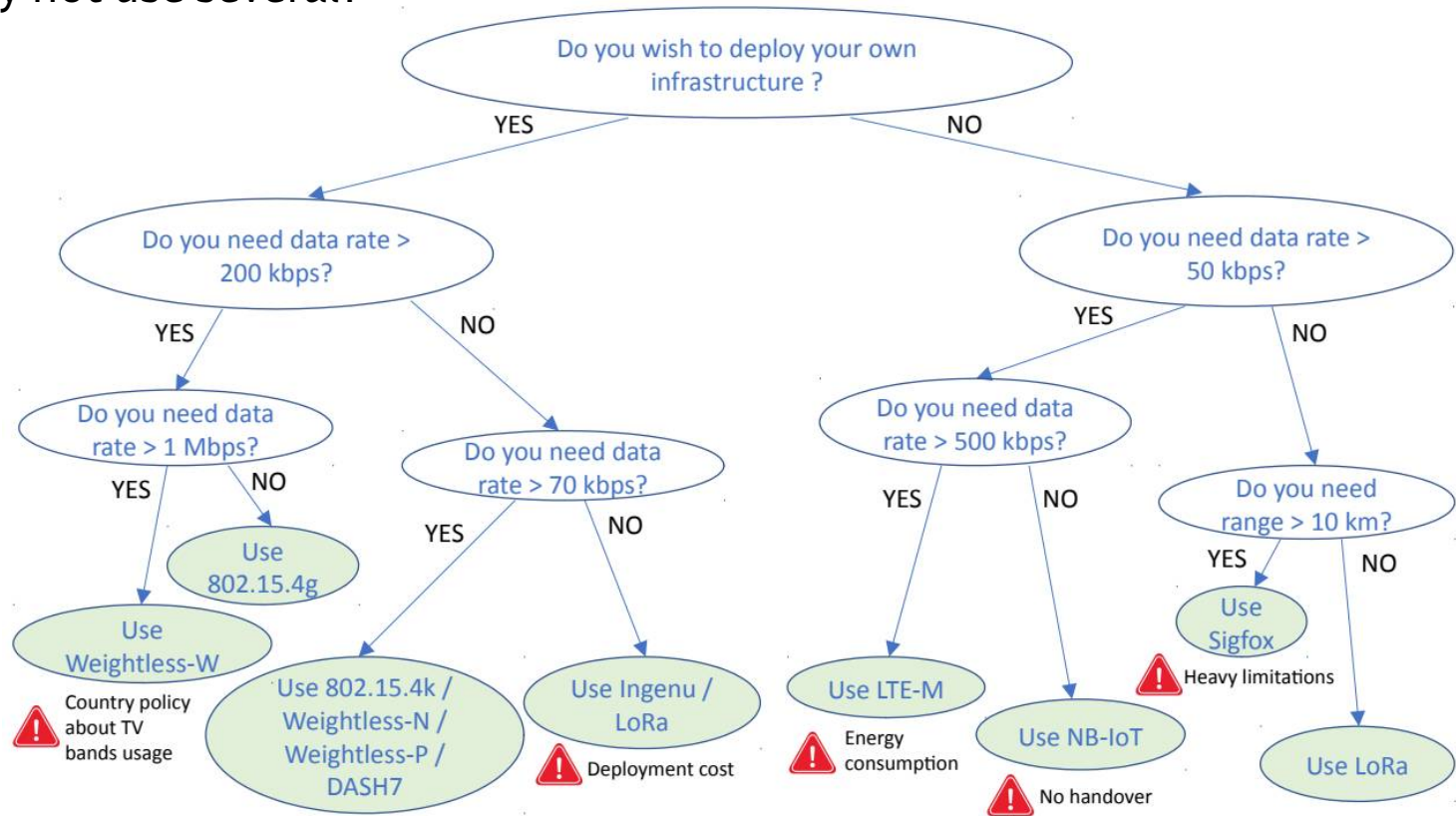
Many RAT, always a trade-off

- Spectrum congestion
- Energy consumption
- Financial cost
- Coverage
- Range
- ...



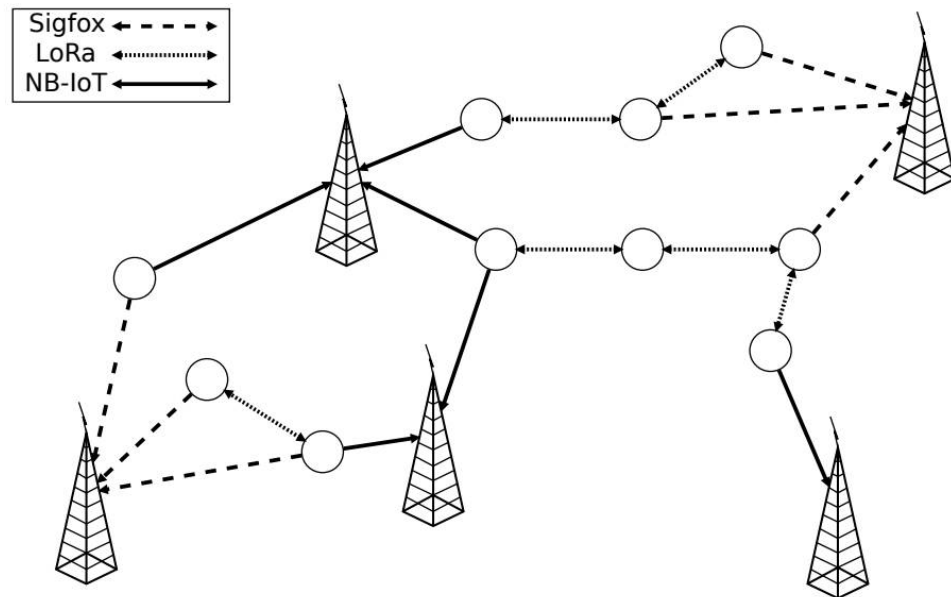
Hard to find the best fitted RAT

- So why not use several?



Introducing multi-technologies networks, with multi-RAT nodes

- Multi-RAT multi-hop networks
- Several radio links between two neighbors
- Several use cases (e.g. monitoring, video)
- Need a reliable RAT selection scheme



Route selection problem: multi attribute decision making

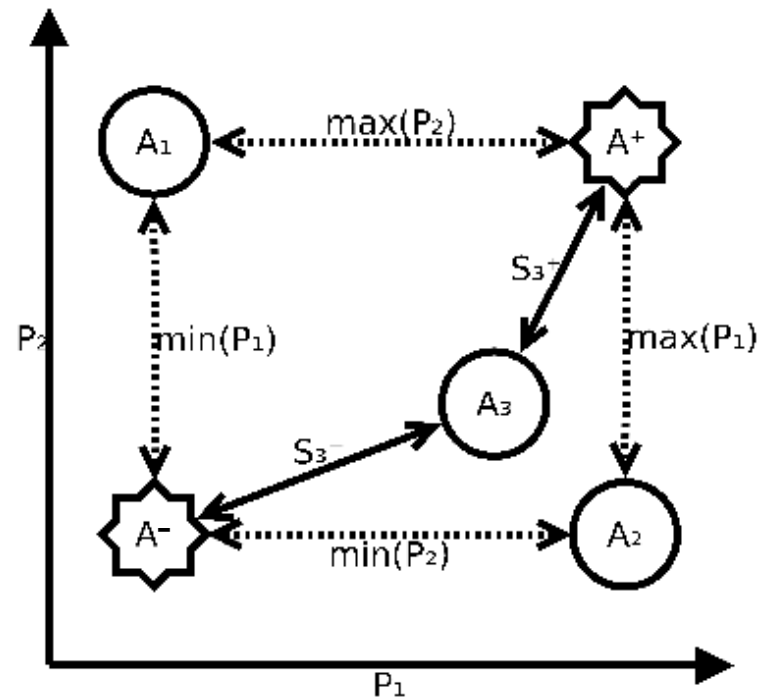
- $A = \{A_i, \text{ for } i=1,2,\dots,n\}$ the set of candidates
- $P = \{P_j, \text{ for } j=1,2,\dots,m\}$ the set of attributes
- $W = \{w_j, \text{ for } j=1,2,\dots,m\}$ the weights of each attribute

MADM DECISION MATRIX

	P_1 w_1	P_2 w_2	...	P_m w_m
A_1	x_{11}	x_{12}	...	x_{1m}
A_2	x_{21}	x_{22}	...	x_{2m}
...
A_n	x_{n1}	x_{n2}	...	x_{nm}

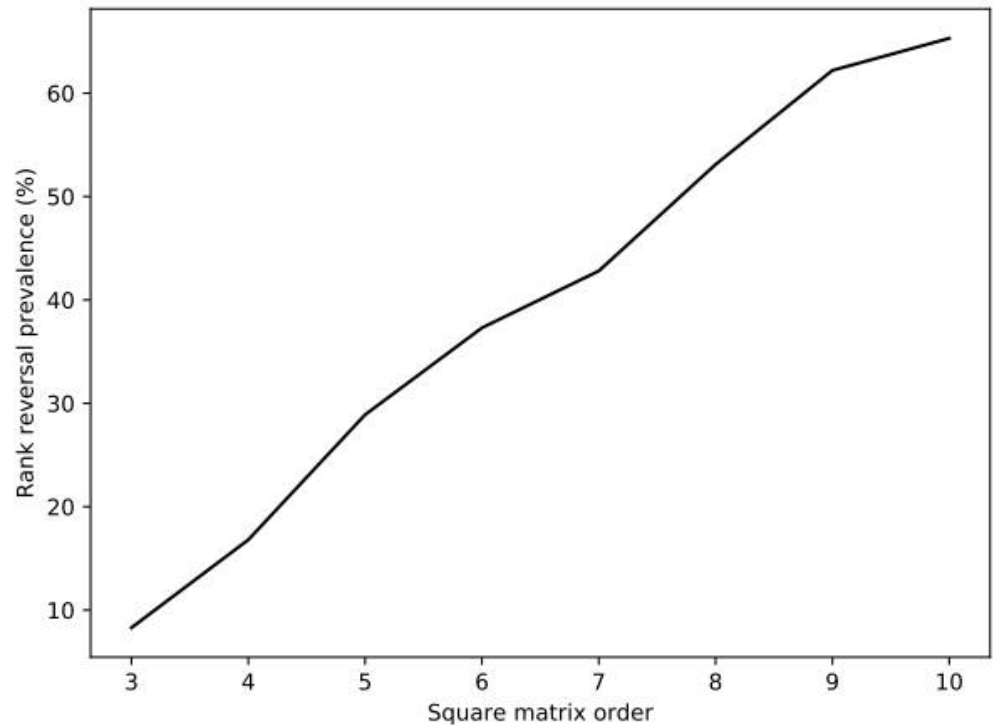
Technique for Order Preference by Similarity to Ideal Solution (TOPSIS)

- Ranks each candidate from best to worst
- Ranks are determined based on:
 - closeness to best solution
 - farness to worst solution



Complexity and ranking abnormalities

- Complex calculations for WSN hardware
- Ranking alteration when removing candidates (e.g. route loss)
- Caused by the euclidean normalization of values



Lightweight TOPSIS for WSN

- Simple and stable normalization algorithm
- Use fixed bounds for each attribute

Algorithm 1 Lightweight normalization

Require: x_{ij} the raw value of each attribute j for each candidate i

for each attribute P_j **do**

if P_j is an upward attribute **then**

B_j^+ is the upper bound of P_j

$$r_{ij} = \frac{x_{ij}}{B_j^+}$$

else if P_j is a downward attribute **then**

B_j^- is the lower bound of P_j

$$r_{ij} = \frac{B_j^-}{x_{ij}}$$

end if

end for

return r_{ij} the normalized value of x_{ij}

TOPSIS complexity lowered

- Spares $5mn-2$ operations (with mn the dimensions of the decision matrix)

$$r_{ij} = \frac{x_{ij}}{\sqrt{\sum_{i=1}^n x_{ij}^2}}$$



$$r_{ij} = \frac{B_j^-}{x_{ij}}$$

$$A^+ = [v_1^+ \dots v_m^+]$$

$$A^- = [v_1^- \dots v_m^-]$$

$$v_j^+ = \max\{v_{ij}, i = 1, \dots, n\}$$

$$v_j^- = \min\{v_{ij}, i = 1, \dots, n\}$$

$$v_j^+ = \min\{v_{ij}, i = 1, \dots, n\}$$

$$v_j^- = \max\{v_{ij}, i = 1, \dots, n\}$$



$$A^+ = [1 \dots 1]$$

$$A^- = [0 \dots 0]$$

$$v_j^+ = 1$$

$$v_j^- = 0$$

$$S_i^+ = \sqrt{\sum_{j=1}^m (v_j^+ - v_{ij})^2}$$

$$S_i^- = \sqrt{\sum_{j=1}^m (v_j^- - v_{ij})^2}$$



$$S_i^+ = \sqrt{\sum_{j=1}^m (1 - v_{ij})^2}$$

$$S_i^- = \sqrt{\sum_{j=1}^m v_{ij}^2}$$

Hardware for experiments: Pycom Fipy

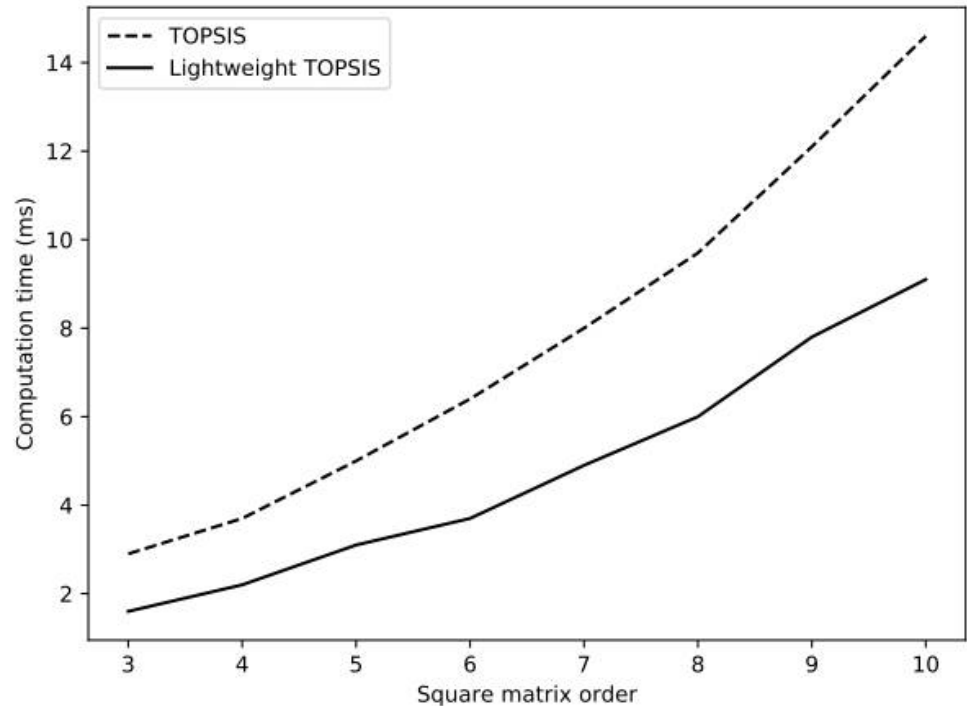
- Allows Python implementation (port of MicroPython)
- 5 RATs available



<https://pycom.io/product/fipy/>

Lightweight TOPSIS results

- Mean speed up of 39%
- 82% NIS similarity with vanilla TOPSIS
- Saves 448 μ J per TOPSIS run (based on FiPy data-sheet)



Recap & future works

- TOPSIS method for multi-criteria selection in WSN
- Custom based method with simpler normalization algorithm
- Quicker execution time for almost identical results, thus energy savings

- Plan to use for route selection in multi-technologies routing

Thank you for your attention!
Any questions?

brandon.foubert@inria.fr