

# Detecting Misbehavior in Wireless Sensor Networks

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# Contents

- 1 Introduction and Motivation
  - Sensor Networks
  - Detecting misbehavior
- 2 Intrusion detection system
  - Architecture
  - Ruleset
  - Evaluation
- 3 En-route-filtering of injected false data
  - Overview
  - Key distribution
  - Report generation and filtering
  - Evaluation
- 4 Conclusion

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# Common challenges

- Energy constraints
  - Sensor running on battery
  - Not likely to get a new battery soon
- Resource constraints
  - Little main memory
  - Small processing unit
- Autonomy
  - User is not nearby

# What is misbehavior detection. . .

. . . and why is it important?

Even with

- encrypted communication and
- authenticated communication

**Attacker may have physical access to sensor nodes!**

- Extraction of cryptographic keys
- Wormhole, Blackhole, . . . attacks possible again

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# What is misbehavior detection. . .

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- Detect misbehaving nodes/compromised keys  
⇒ "Decentralized intrusion detection system for WSN"
- Handle intrusion when detected  
⇒ "Statistical enroute-filtering of injected false data"



# What is misbehavior detection...

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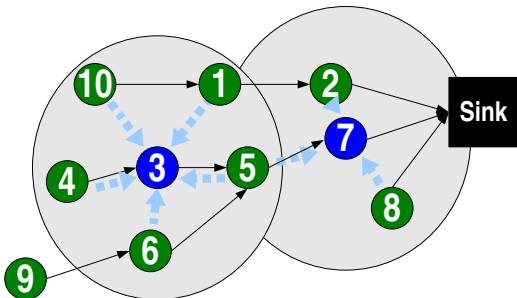
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# Global Architecture

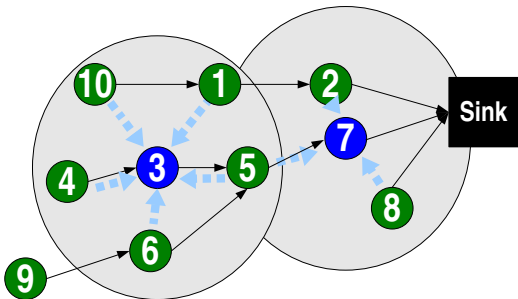
- Monitor nodes (3, 7) use promiscuous listening



- Nodes do not move
- Nodes can be identified
- Reliable connection from monitor to sink

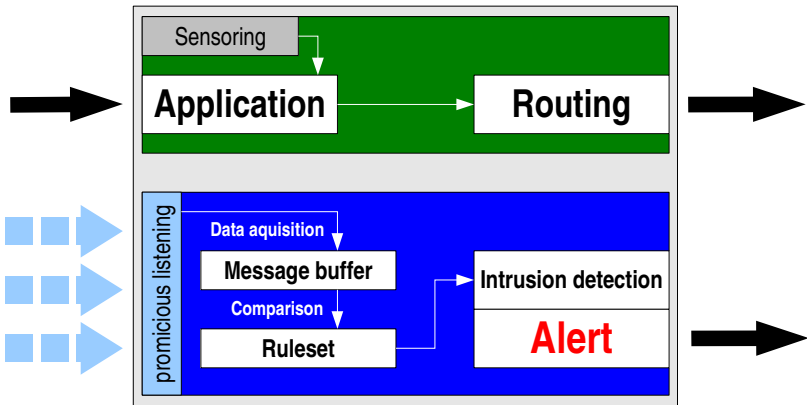
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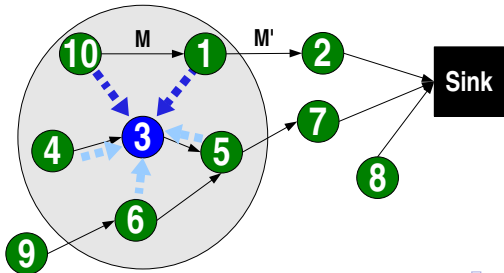
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# Monitor node



# Retransmission, delay and integrity rule

Retransmission rule	Does <b>1</b> forward the message? Blackhole attack or selective forwarding
Integrity rule	$M = M' ?$ Message alteration attack
Delay rule	$t(M') - t(M) < \text{treshold} ?$ Message delay attack



# Simulation setup

Size	Sensors	100 nodes
	Monitors	28 nodes
Procedure	Total duration	10000 iterations
	Learning phase	1000 iterations
	10 attack cycles with each	
	Idle time	700 iterations
	Attack duration	200 iterations
Simulated	One compromised node	
	One form of attack	
	Network failure rate	10% (20%)

# Detection effectiveness

## Simulation results

Attack	Small Message Buffer		Large Message Buffer	
	DR	FP	DR	FP
Message delay	bad	few	good	hardly any
Blackhole	good	too many	good	few
Selective forwarding	medium	too many	good	few
Wormhole	good	many	good	few
Message repetition	good	few	good	hardly any
Jamming	good	medium	good	few
Data alteration	good	too many	medium	few

**DR**=Detection Rate    **FP**=False Positives



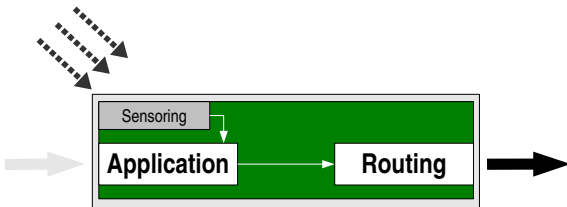
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# Motivation...

...of the statistical enroute-filtering

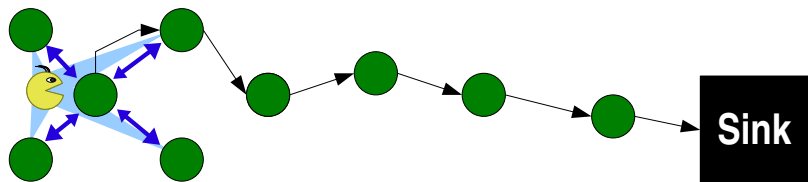
What if **initial report** is already fabricated?



Goal:

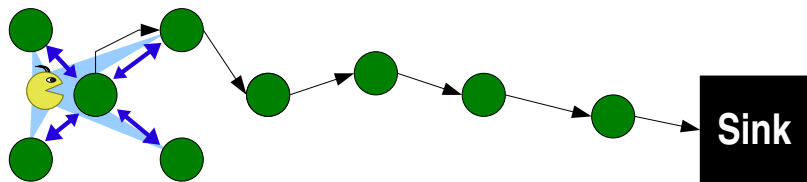
Recognition and early disposal of fabricated reports

# General approach



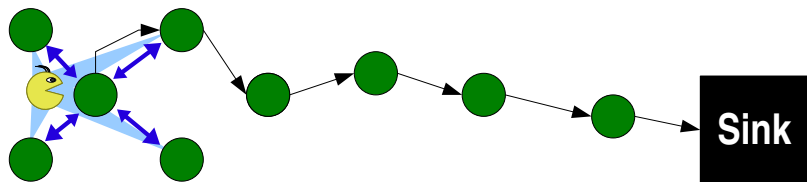
- Verification: **En-route** (to save energy) and **at the sink**
- How to distribute the keys?

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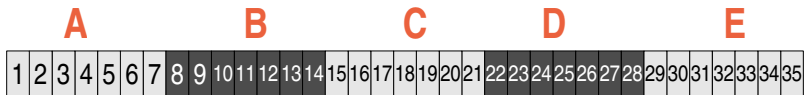
# Keys, categories, index numbers



- Global key pool
- Numbering, Partitioning

Key distribution

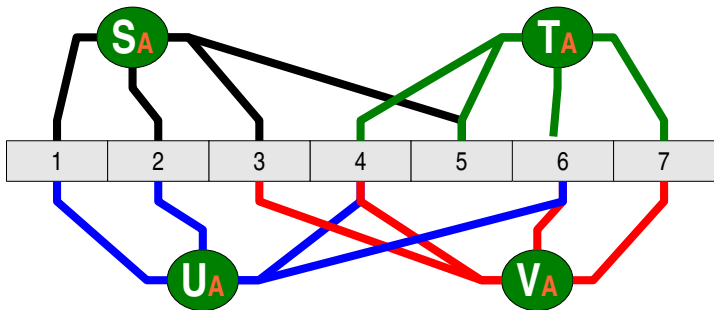
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## Key distribution

# A node stores 4 random keys from the same category

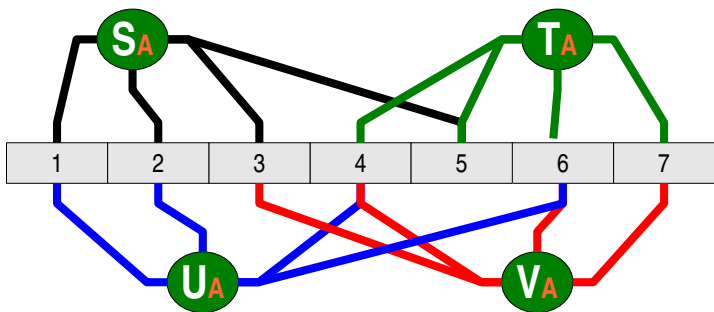


- Node **S** stores:  $\{(1, K_1), (2, K_2), (3, K_3), (5, K_5)\}$
- Node **T** stores:  $\{(4, K_4), (5, K_5), (6, K_6), (7, K_7)\}$
- Node **U** stores:  $\{(1, K_1), (2, K_2), (4, K_4), (6, K_6)\}$
- Node **V** stores:  $\{(3, K_3), (4, K_4), (6, K_6), (7, K_7)\}$



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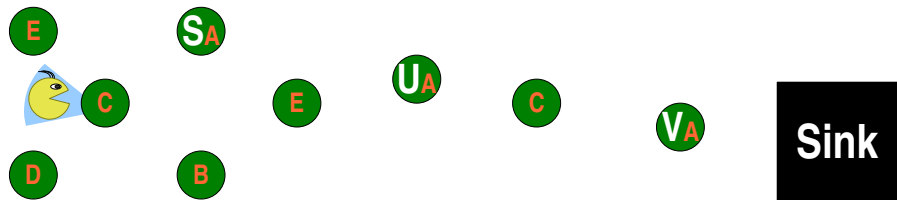
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# Report generation

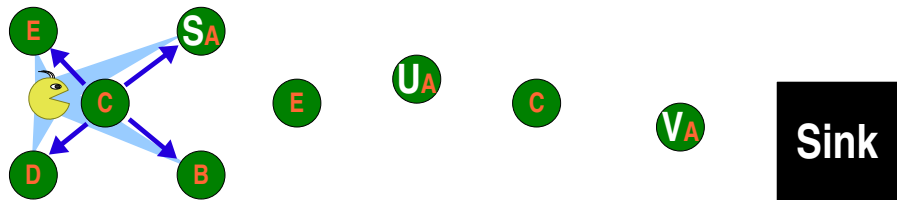
$(pos, timestamp, type), (2, MAC_2), (10, MAC_{10}), (17, MAC_{17})$



- C detects stimulus
- $report = (pos, timestamp, type)$  is verified
- Neighbors return  $(i, MAC(report, K_i))$

# Report generation

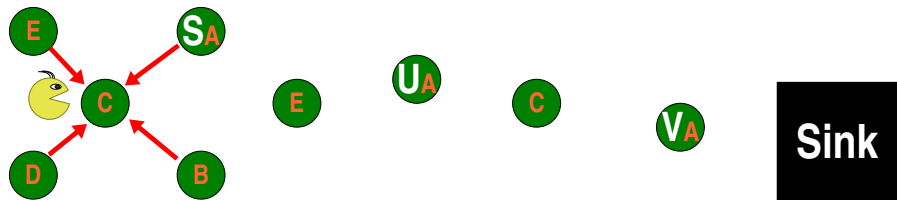
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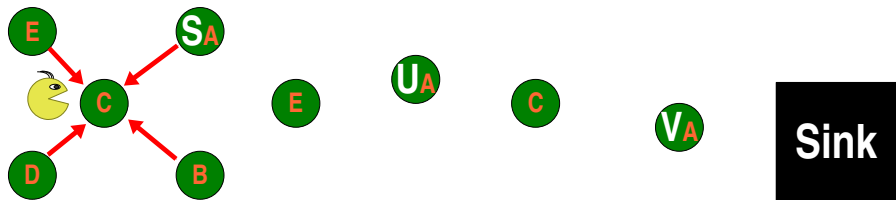
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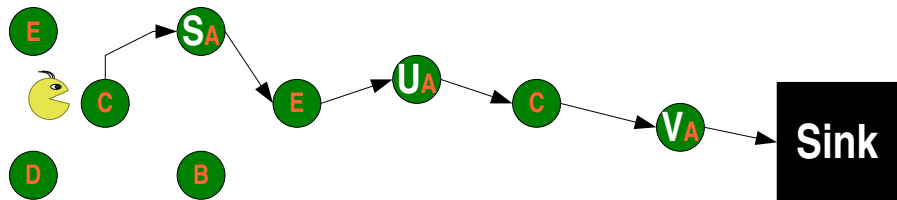
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- C selects 3 MACs each of a different category
- C sends report to sink, with MACs attached

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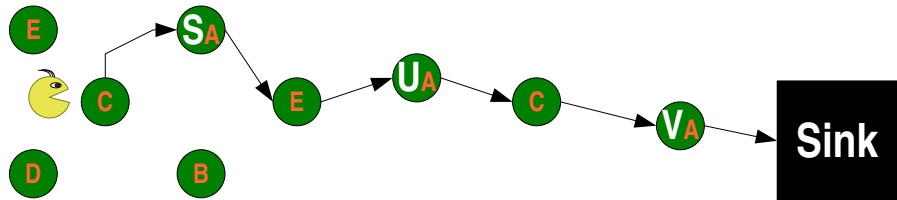
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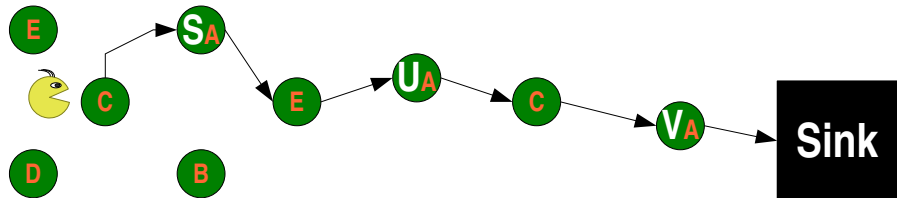
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- Invalid MAC found?  $\Rightarrow$  Drop
- MACs not verifiable or correct?  $\Rightarrow$  Forward

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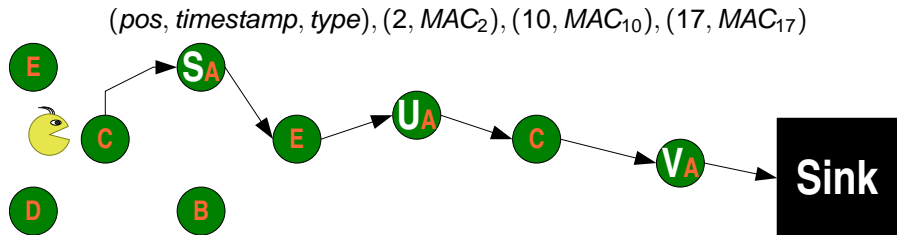
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# Filtering at the sink



- Sink knows all keys of every category
- Verification **all** MACs attached to the report

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# Theoretical efficiency estimate

Keys	Total number	1000 keys
	<b>10</b> categories	100 keys
	Each node	50 keys
	Each report	5 MACs

Assuming the attacker has compromised  $N_c < 5$  nodes.

- How likely that a node can identify a forged key?
- How likely that a forged key is identified after  $h$  hops?

# Theoretical efficiency estimate

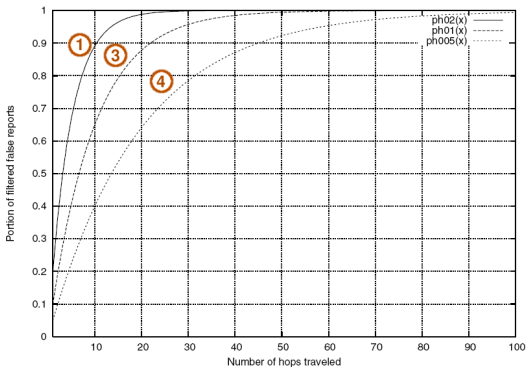
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# Packets dropped after $n$ hops...

... for 1,3 and 4 compromised categories



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- 1 Misbehavior detection in sensor networks is possible
  - Intrusion detection works for most attacks
  - False injection detection also works
- 2 Both systems have open issues
  - Intrusion detection and encrypted communication
  - Alerting the sink
  - En-route-filtering addresses only a single attack
- 3 Only systems for special aspects! Combination possible?
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# Quellen



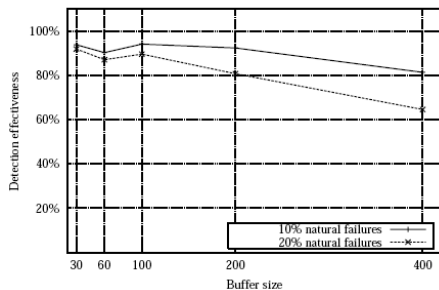
Decentralized Intrusion Detection in Wireless Sensor Networks, Ana Paula R. da Silva, Marcelo H. T. Martins, Bruno P. S. Rocha, Antonio A. F. Loureiro, Linnyer B. Ruiz, Hao Chi Wong, October 2005, Proceedings of the 1st ACM international workshop on Quality of service & security in wireless and mobile networks Q2SWinet '05



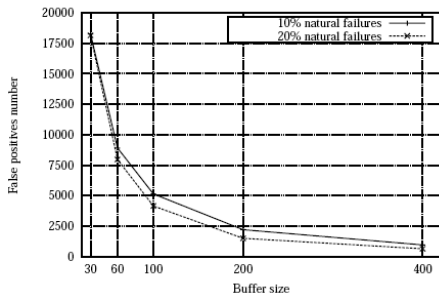
Statistical En-route Filtering of Injected False Data in Sensor Networks, Fan Ye, Haiyun Luo, Songwu Lu, Lixia Zhang, UCLA Computer Science Departement, Los Angeles

# Energy consumption

Data alteration



Data alteration



# Energy consumption

- Base values for energy consumption
  - Listening:  $0.01 \frac{mJ}{message}$
  - Receiving:  $0.15 \frac{mJ}{message}$
  - Sending:  $0.48 \frac{mJ}{message}$
- Results
  - Monitor nodes consume more energy
  - Nodes near the sink consume more energy

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