

# **PICO: Privacy-Preserving Access Control in IoT Scenarios through Incomplete Information**

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May 2022

# Problem Statement

- ❑ Fine-grained info are required for services and resources to be granted in IoT
- ❑ The user might have privacy preserving concerns regarding the granularity of sensitive information disclosure

## PICO

- ❑ Provides an evaluation platform for users to assess its disclosure risk and to act based on privacy exchange and energy consumption tradeoff

- Attribute Provider (AP)

*Collects and certifies the client's attribute values*

*Provides cryptography materials for server's attribute validation*

- Client

*Attribute Handler (AH)*

*Local storage and management of attributes*

*Decision Risk Assessment (DRA)*

*Protocol Handler (PH)*

*Supports interactions with the server*

- Server

*Access Control Manager (ACM)*

*Protection of sensitive resources and services*

*Probabilistic Policy Decision Point (PPDP)*

*Policy Administration Point (PAP)*

*Risk Resolution (RR)*

*Protocol Handler (PH)*

*Supports interactions with the server*

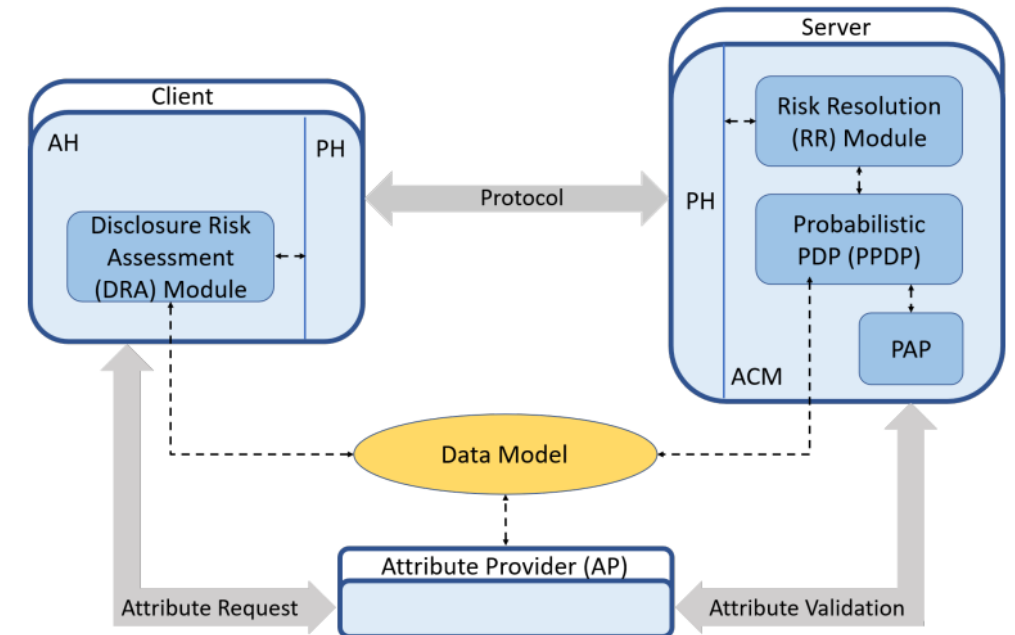
- Data Model

*Shared*

*Tree-based structure*

*Clients use it to determine the level of granularity*

*Servers use it for incorrect risk assessment*

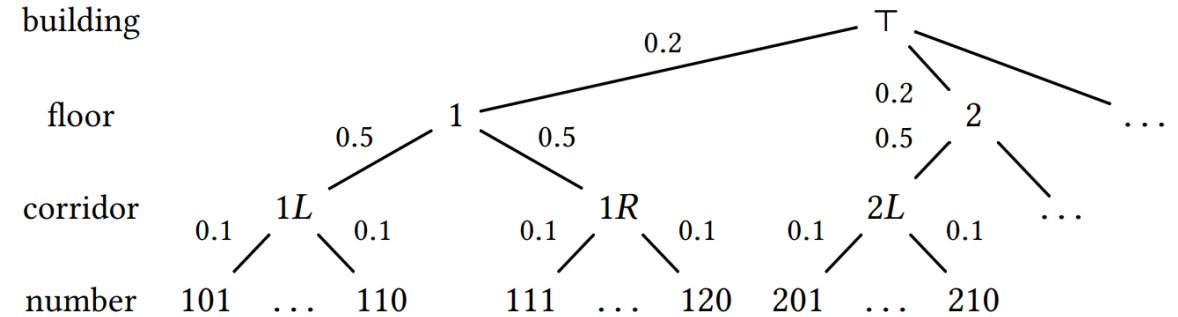


PICO Architecture

# Notation

- $A = \{a_1, \dots, a_n\}$ ,  $a_i$ : attribute  $i$
- $V_{a_i}$ : set of possible values for  $a_i$
- $H_a = (N, S, \lambda)$ : attribute hierarchy
- $N$ : set of nodes
- $S \subseteq N \times N$ : specialization relation on  $N$
- $(n_i, n_j)$ : path from  $i$  to  $j$
- $\lambda : S \rightarrow [0,1]$ : labeling function, shows closeness between two attribute values
- $(a, v_i) \subseteq (a, v_j)$ : In attribute  $a$ ,  $v_i$  is a specialization of  $v_j$
- Specialization is:
  - Reflexive
  - Antisymmetric
  - transitive

## Hierarchy Tree



## Degree of Similarity

- Base case: for adjacent  $n_i$  and  $n_j$

$$\sigma(n_i, n_j) = \begin{cases} 1 & \text{if } n_i = n_j \\ 1 & \text{if } (n_i, n_j) \in S \\ \lambda(n_i, n_j) & \text{if } (n_j, n_i) \in S \end{cases}$$

Probability of inferring  $j$   
when we know  $i$

- Path: for a path from  $n_i$  to  $n_j$

$$\sigma(n_i, n_j) = \prod_{k=1}^{m-1} \sigma(n_k, n_{k+1}) \quad (\text{with } n_i = n_1 \text{ and } n_j = n_m)$$

# Disclosure Risk Assessment (DRA) - Client Side

- Assesses the risk associated to the disclosure of attribute values and decides at which level of granularity information can be disclosed
- For an attribute value, the disclosure risk is the probability of inferring the most specialized value of a user

$$\delta_{H_a}(n') = \sigma(n', n)$$

$n$  : exact attribute value

$n'$  : generalization of  $n$

i.e.  $(a, n) \subset (a, n')$

- Attribute risk tolerance:
  - sensitive
  - non-sensitive

$$\delta_{H_a}(n) \geq \tau_a$$

$$\delta_{H_a}(v) < \tau_a$$

- All generalizations of a non-sensitive value are also non-sensitive
- Risk disclosure of a set of attribute values is equal to the highest disclosure risk.

$$\delta(V) = \max\{\delta_{H_{a_i}}(v_i) \mid v_i \in V\}$$

# Probabilistic PDP – Server Side

- Represents the risks of incorrectly granting/denying access due to incomplete information
- Query : a set of attribute name-value pairs  $(a_i, v_i)$
- Set of queries over  $\mathcal{A}$
- Using XACML policy language syntax
- $t$  : policy targets
  1. Atomic :  $(a, v)$
  2. Composite : Boolean expression  $(\neg, \wedge, \vee)$  over atomic targets
- $p$  : policy
  1. Atomic : single decision, 1 or 0
  2. Target policy :  $(t, p)$
  3. Composite : deny-override ( $\Delta$ ) or permit-override ( $\nabla$ ) over  $\{1, 0, \perp\}$
- Deny-override algorithm ( $\Delta$ ) : Deny decision has priority over a permit decision
- Permit-override algorithm ( $\nabla$ ) : Permit decision has priority over a deny decision

$$q = \{(a_1, v_1), \dots, (a_k, v_k)\}$$

$$Q_{\mathcal{A}} = \wp(\bigcup_{a_i \in \mathcal{A}} \bigcup_{v_j \in \mathcal{V}_{a_i}} (a_i, v_j))$$

Deny-override & Permit-override look-up tables

$P_1$	$P_2$	$P_1 \Delta P_2$
0	0	0
0	1	0
0	$\perp$	0
$\perp$	0	0
1	0	0
1	1	1
1	$\perp$	1
$\perp$	1	1
$\perp$	$\perp$	$\perp$

$P_1$	$P_2$	$P_1 \nabla P_2$
1	1	1
1	0	1
1	$\perp$	1
$\perp$	1	1
0	1	1
0	0	0
0	$\perp$	0
$\perp$	0	0
$\perp$	$\perp$	$\perp$

# Probabilistic PDP – Server Side

- Set of targets over A :  $T_A$
- Set of policies over A :  $P_A$

## □ Evaluation Function $\llbracket . \rrbracket_T$

- $T_A \times Q_A \rightarrow [0, 1]$
- $\llbracket t \rrbracket_T(q)$  : likelihood that  $q$  matches  $t$  based on the degree of similarity between the attribute values in  $t$  &  $q$ 
  - 1 : target certainly matches the query
  - 0 : target certainly doesn't match the query

## □ Evaluation Function $\llbracket . \rrbracket_P$

- $P_A \times Q_A \rightarrow [0, 1]^3$
- $\llbracket p \rrbracket_P(q)$  : likelihood that a certain decision is returned based on the degree of similarity between attributes
- $\llbracket p \rrbracket_P(q) = (\ell^1, \ell^0, \ell^\perp)$ 
  - $\ell^1$  : likelihood that the decision is permit
  - $\ell^0$  : likelihood that the decision is deny
  - $\ell^\perp$  : likelihood that the decision is NA
  - $\ell^1 + \ell^0 + \ell^\perp = 1$

$\llbracket (a, v) \rrbracket_T(q)$	=	$\max\{\sigma(v_i, v) \mid (a, v_i) \in q\}$
$\llbracket \neg t \rrbracket_T(q)$	=	$1 - \llbracket t \rrbracket_T(q)$
$\llbracket t_1 \vee t_2 \rrbracket_T(q)$	=	$\llbracket t_1 \rrbracket_T(q) + \llbracket t_2 \rrbracket_T(q) - \llbracket t_1 \rrbracket_T(q) \llbracket t_2 \rrbracket_T(q)$
$\llbracket t_1 \wedge t_2 \rrbracket_T(q)$	=	$\llbracket t_1 \rrbracket_T(q) \llbracket t_2 \rrbracket_T(q)$
$\llbracket 1 \rrbracket_P(q)$	=	$(1, 0, 0)$
$\llbracket 0 \rrbracket_P(q)$	=	$(0, 1, 0)$
$\llbracket (t, p) \rrbracket_P(q)$	=	$\llbracket t \rrbracket_T(q) \cdot \llbracket p \rrbracket_P(q) + (1 - \llbracket t \rrbracket_T(q)) \cdot (0, 0, 1)$
$\llbracket p_1 \nabla p_2 \rrbracket_P(q)$	=	$(\ell_1^1 + \ell_2^1 - \ell_1^1 \ell_2^1, \ell_1^0 \ell_2^0 + \ell_1^0 \ell_2^\perp + \ell_1^\perp \ell_2^0, \ell_1^\perp \ell_2^\perp)$
$\llbracket p_1 \Delta p_2 \rrbracket_P(q)$	=	$(\ell_1^1 \ell_2^1 + \ell_1^1 \ell_2^\perp + \ell_1^\perp \ell_2^1, \ell_1^0 + \ell_2^0 - \ell_1^0 \ell_2^0, \ell_1^\perp \ell_2^\perp)$

# Risk Resolution – Server side

- Responsible for policy enforcement, similar to the Policy Enforcement Point (PEP) in the XACML
- Decision needs to be made conclusively and not by likelihood
- The uncertainty in likelihood estimations needs to be quantified
- "Not-applicable" decisions need to be translated into either "permit" or "deny"
- Conservatively deciding, NAs are considered "deny"

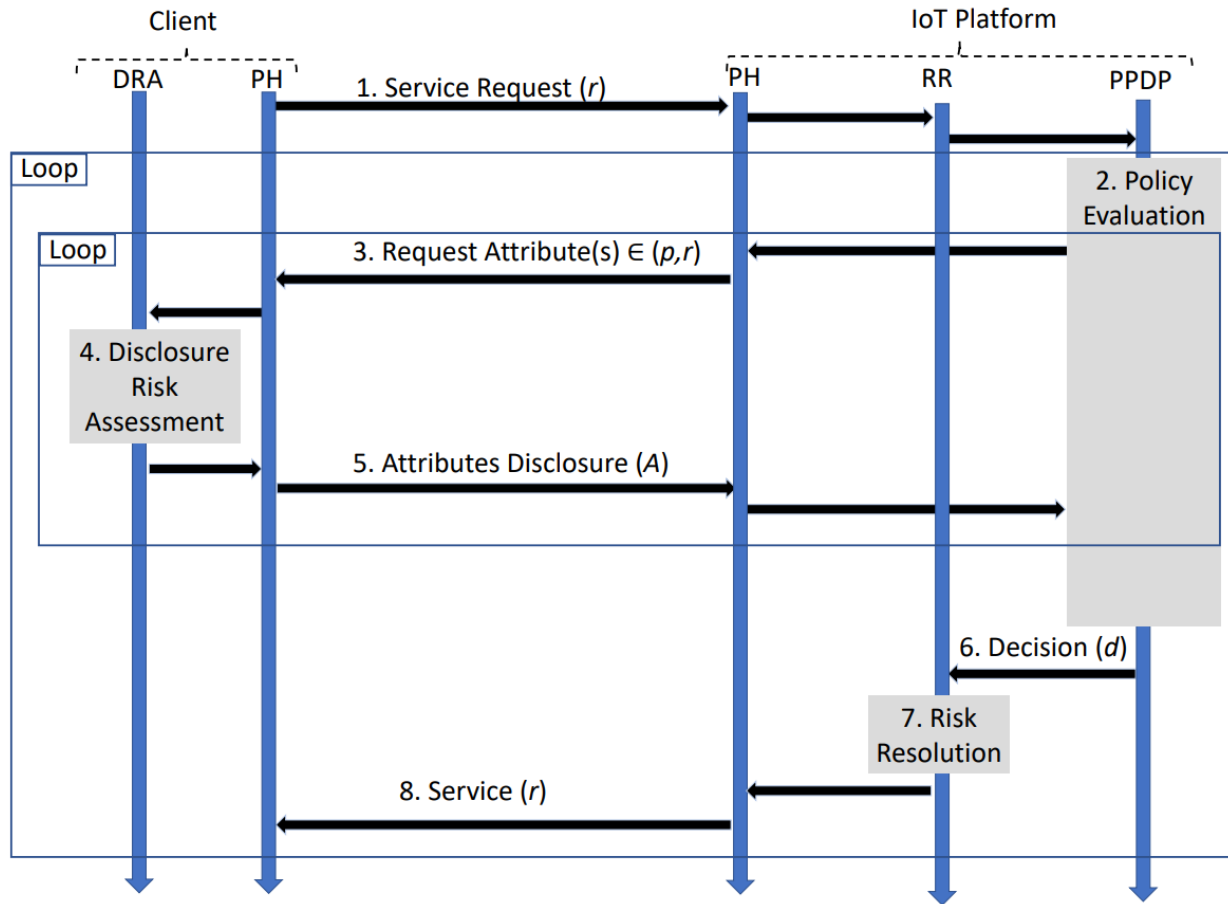
## □ Risk Tolerance

$$\ell^1 \geq \alpha (\ell^0 + \ell^\perp)$$

$\alpha$  : risk factor (showing criticality of resources)

$\alpha (\ell^0 + \ell^\perp)$  : permission lower band

# Protocol



- Sequence diagram of the protocol used by the client to access services provided by the server IoT platform

## Notes

- Client can decide to disclose her attributes in different levels of granularity.
- The protocol can be continued until:
  - The resource is granted
  - The client refuses to reveal more info
- Information disclosure increases the disclosure risk, and refusal to expose coarse-grained information is time and energy consuming
- A trade-off between disclosure risk and time/energy consumption is applied in client's behavior.



# Implementation

- ❑ Each disclosed attribute value is embedded into a CWT

## ❑ CWT

- CBOR Web Token
- A compact means of representing **claims** to be transferred between two parties
- Optimized for IoT use-cases
- Consisting of 3 parts:
  1. Header (for correct decoding)
  2. Claims (standardized or private)
  3. Signature (for integrity and authenticity verification)

## ❑ Claims

- A piece of information asserted about a subject
- represented as a name/value pair consisting of a Claim Name and a Claim Value
- Claims in a CWT are encoded using CBOR

- Standardized claims are used for the: *issuer (iss)*, *unique identifier (cti)*, *subject (sub)*, and *expiration date (exp)*
- Ad-hoc private claim is used for *attribute values (atv)*
- HMAC-SHA-256 is used to generate CWT signature
- A single CWT is 233 bytes

## ❑ Delivering CWT over IoT wireless networks

- CoAP message (adding 4 bytes header)
- UDP (adding 8 bytes header)
- 6LoWPAN (adding 40 bytes header)
- IEEE 802.15.4 (21 bytes header and 4 bytes trailer)
- Sum: 76 bytes
- MTU of IEEE 802.15.4 is 127 bytes
- Available payload for application-layer information: 50 bytes
- Number of MAC-layer messages needed to deliver a single CWT: five IEEE 802.15.4 messages

## ❑ Energy Consumption Estimations

- Openmote-b IoT device, same protocol stack, same MAC-layer message
- 802.65 mJ for a transmission slot
- 778.51 mJ for a reception slot

# Experiment

- **Attribute disclosure approaches:**

1. Direct disclosure of non-sensitive attribute values with the highest disclosure risk (*A1*)
2. Incremental disclosure of non-sensitive attribute values, from low risk to high (*A2*)

- **Attribute delivery modes:**

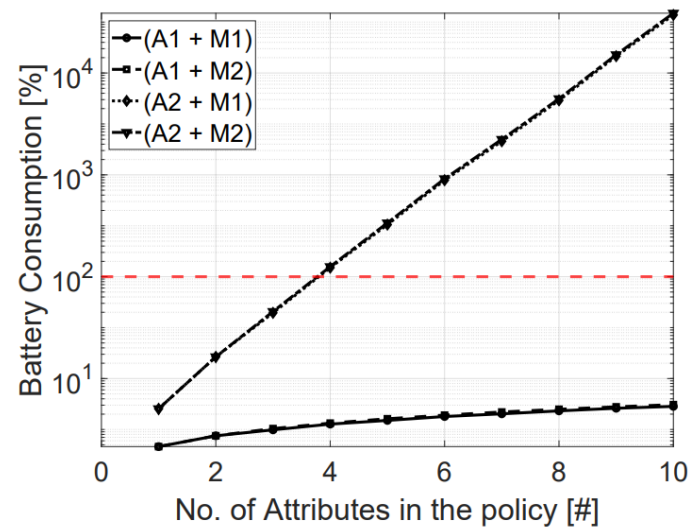
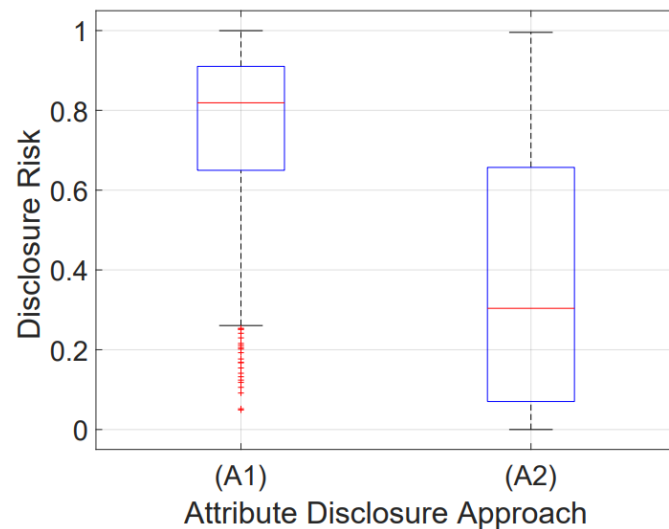
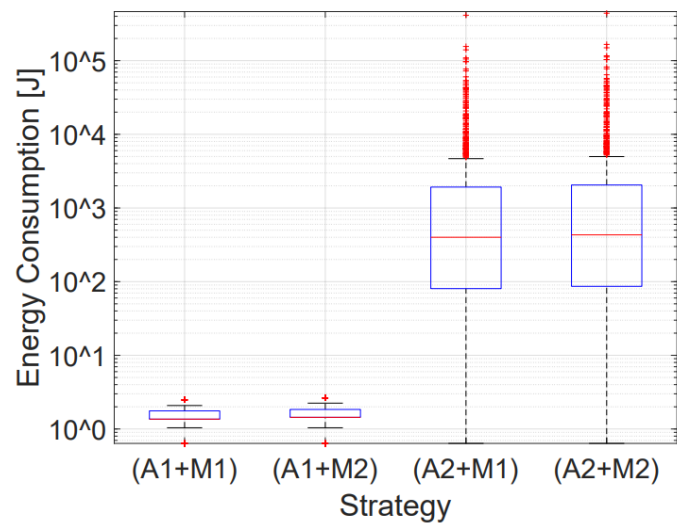
1. All CWTs delivered together in a single stream (*M1*)
2. CWTs delivered one-by-one in different streams (*M2*)

- **Operational strategies:**

(*A1+M1*), (*A1+M2*), (*A2+M1*), (*A2+M2*)

- **Considerations:**

- Number of attributes: 6 (fixed in first experiment, ranging from 1 to 10 in second experiment)
- Attribute hierarchies: Static, Height-balanced binary trees, predefined depth (9, 10, or 11)
- Random semantic closeness values (siblings' semantic closeness values with the parent node sum up to 1)
- Random policies on the server side : Random combination of attributes (from 1 to 6) + Random risk factor



# Results

# ***Ref:***

*Sciancalepore, Savio & Zannone, Nicola. (2022). PICO: Privacy-Preserving Access Control in IoT Scenarios through Incomplete Information. 10.1145/3477314.3508379.*