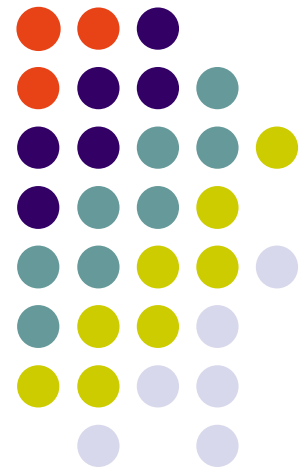


# Configuring data flows in the Internet of Things for security and privacy requirements

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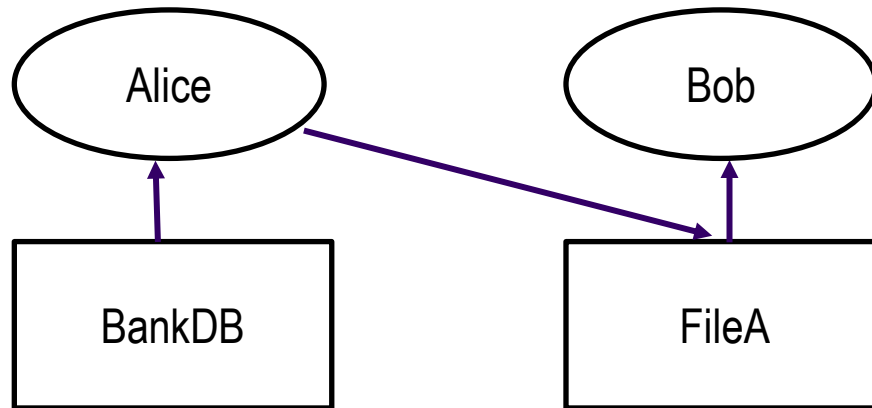
***Université du Québec en Outaouais***

*Foundation and Practice of Computing, November 2018*



# Data flow control vs access control

- Access control:
  - Controls access of subjects to objects
- Data flow control:
  - Controls where data can end up in a network



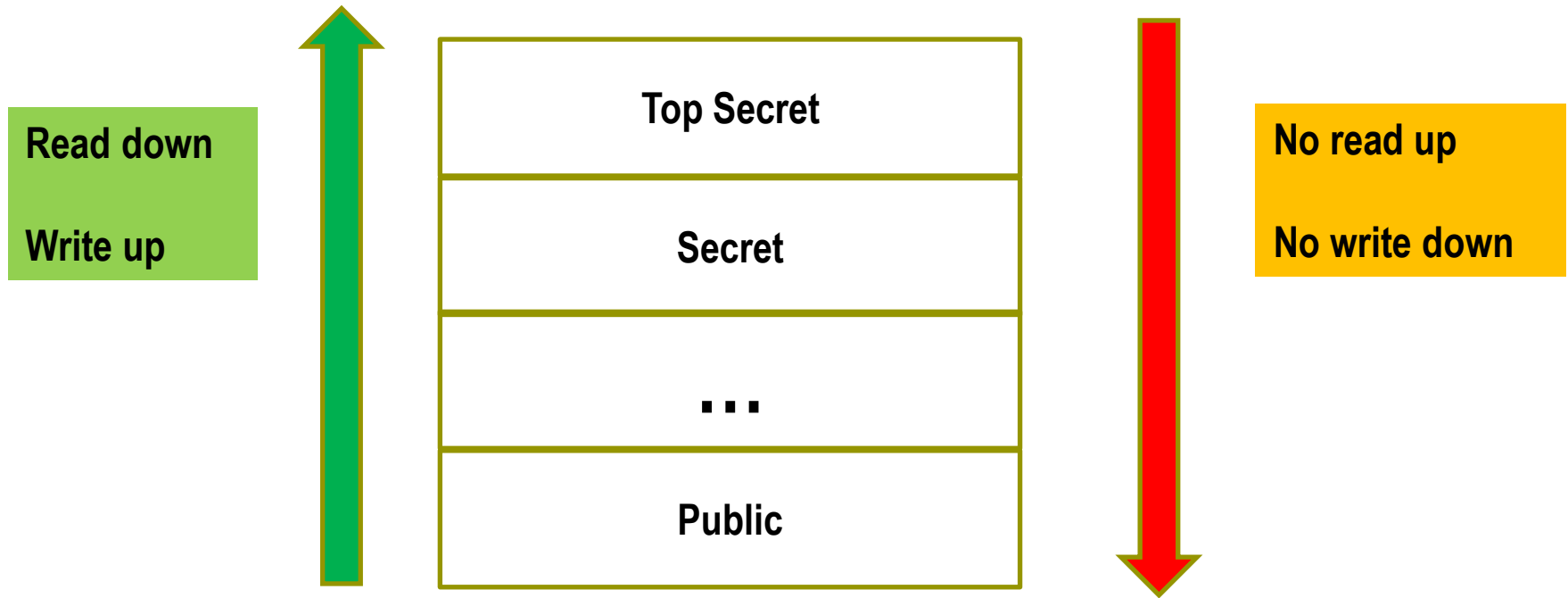
- Bob can know the data in BankDB although it has no direct access to it (see Trojan horse etc.)

# MAC

## Mandatory Access Control data security models

- Subjects and objects are labelled
  - Subjects are labelled by the data that they can read
  - Objects are labelled by the data that they can contain
- There are label-based rules that determine
  - Which subjects can read which objects
  - Which subjects can write on which objects
- Simultaneously guarantees access control and flow control
  - Often considered too restrictive

# The Bell-La Padula model

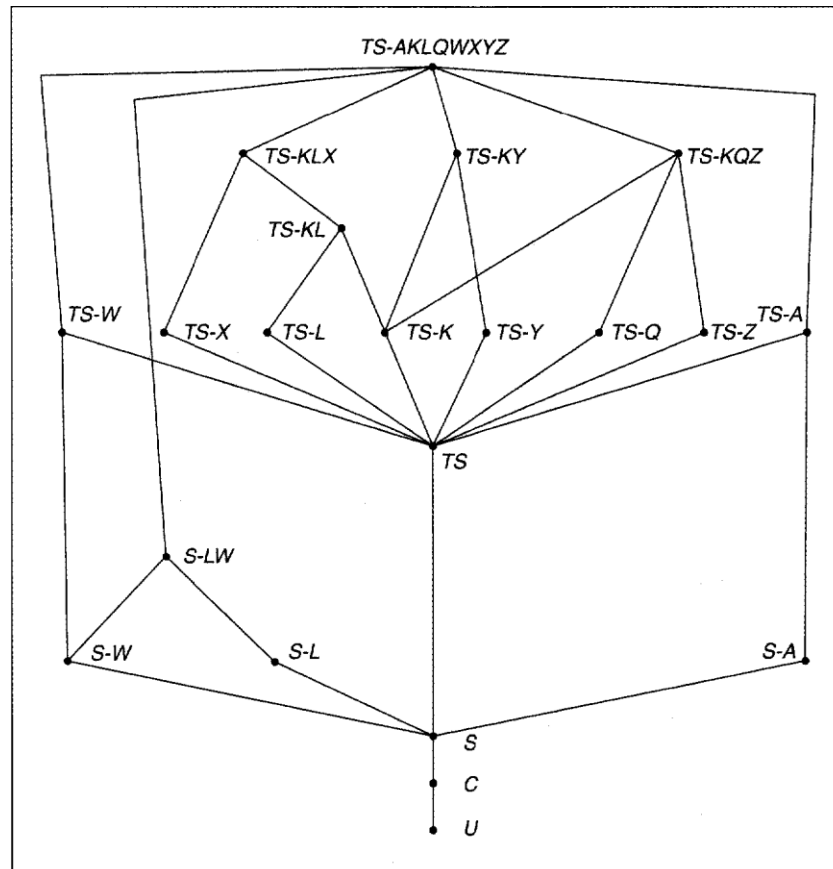


Data can flow only upwards

We generalize this model

# An established generalization: Lattice model (Denning 1974)

- Data can move only upwards in the lattice



Source: Sandhu, 1993

# Success and critique of the lattice model

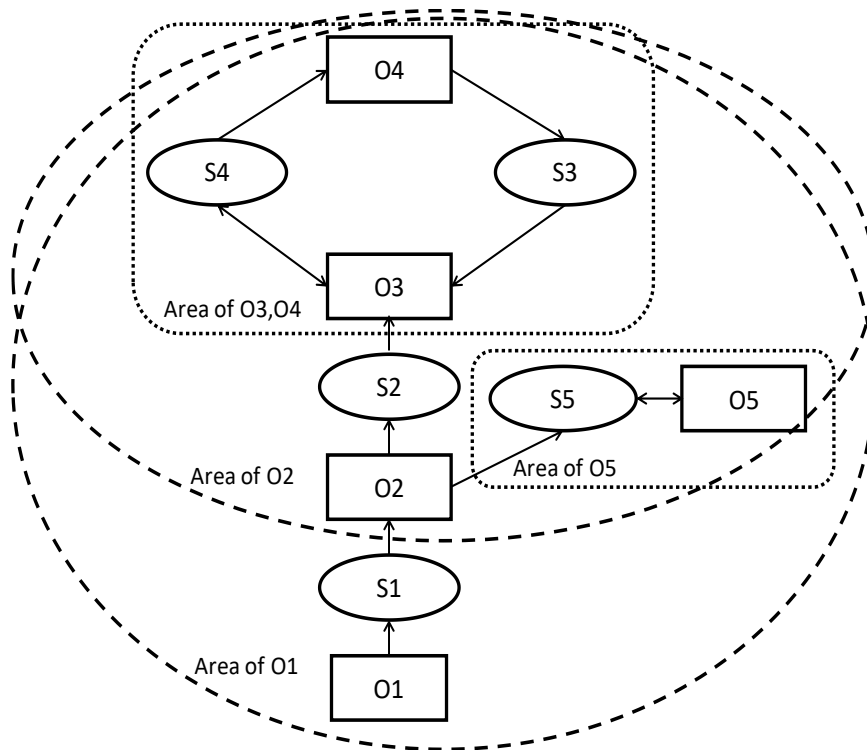
- **All** security data flow models in the literature are based on the lattice model
- But: The lattice model defines security properties in terms of itself!
  - Essentially: “A data flow is secure iff data can only move up in a pre-defined lattice of security classes”
- Also: it may make it necessary to include inexistent or impossible entities in order to have a lattice structure, e.g.
  - an entity that can know everything and
  - another that can know nothing
  - entities that contradict security constraints
    - E.g. if no entity is supposed to know both Bank1 and Bank2 data, it is still necessary to assume the existence of an entity that knows both!

# Last year's FPS paper (Logrippo)

- Secrecy property for data item  $O$  is defined as a partitioning of a set of entities in at least two areas:
  - Area where the entities can know  $O$
  - Area where the entities cannot know  $O$
  - With an order relationship  $\text{CanFlow}$  such that  $\text{CanFlow}(X,Y)$  is true iff entity  $Y$  can know all the data items that entity  $X$  can
    - An order relationship generated by an inclusion property
    - It is a quasi-order
    - It can be represented as a digraph
  - Quasi-orders become partial orders if fully connected components are condensed into one component
    - Entities that can freely exchange data are condensed into one entity

# Example:

not a lattice, only a quasi-order that can be transformed into a partial order of components



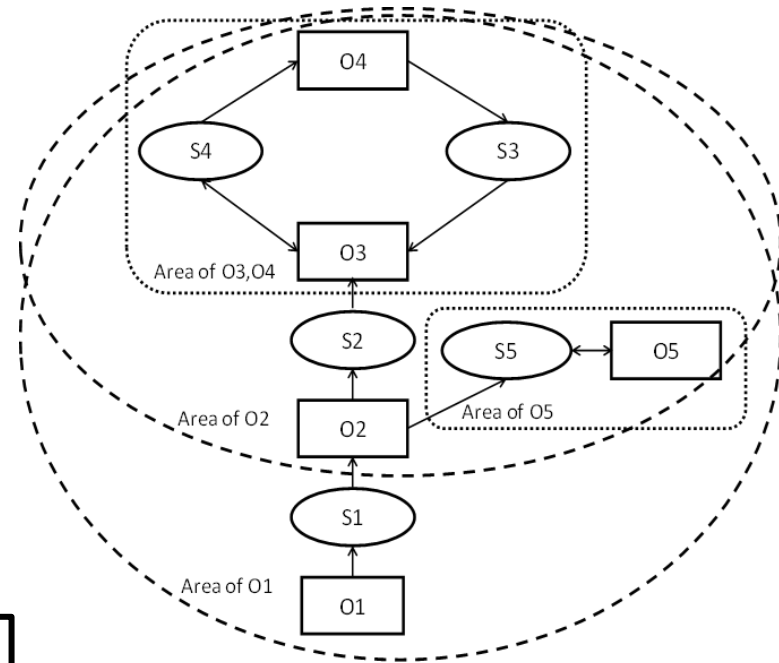
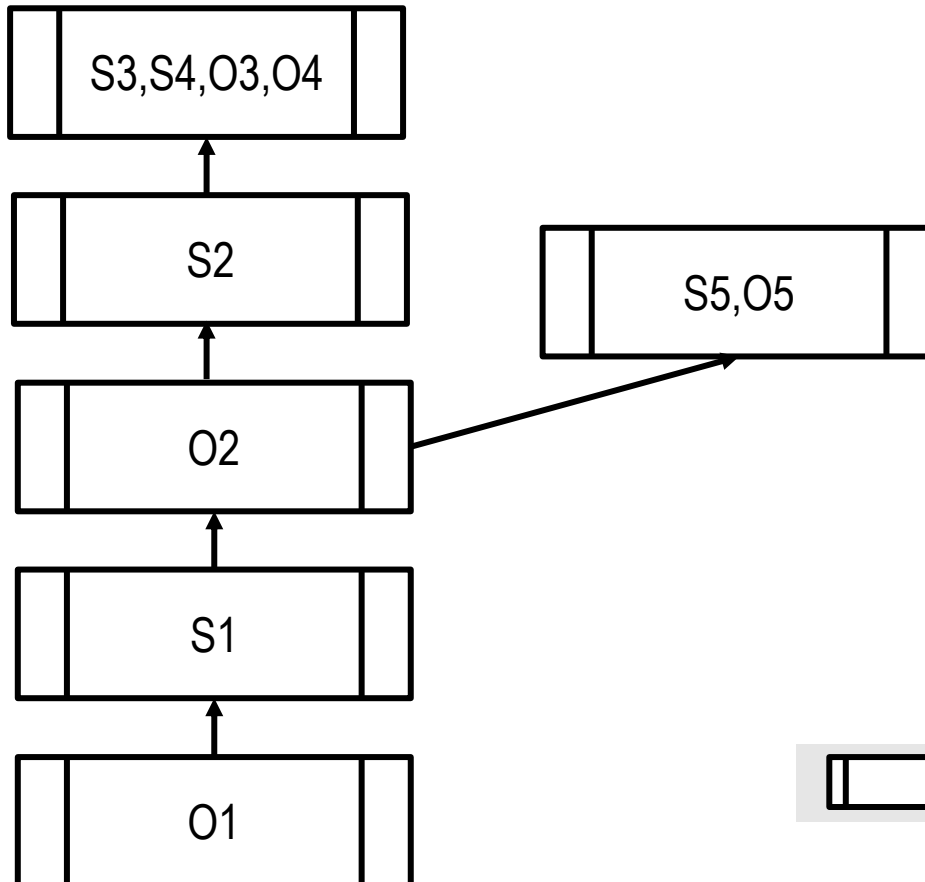
○Subjects, □Objects, → data flow

- Data in O5 cannot end up in O2, S4 etc,
  - The contents of O5 is a *secret* for O2, S4 etc.
  - Secret known only to O5 and S5
- Data in object O1 are the least secret
- Data in objects O3, O4, O5 are the most secret
  - Components are levels of security



# The partial order

Collapsing contents-equivalent entities into single nodes



 Equivalence class

# Basic theoretical results

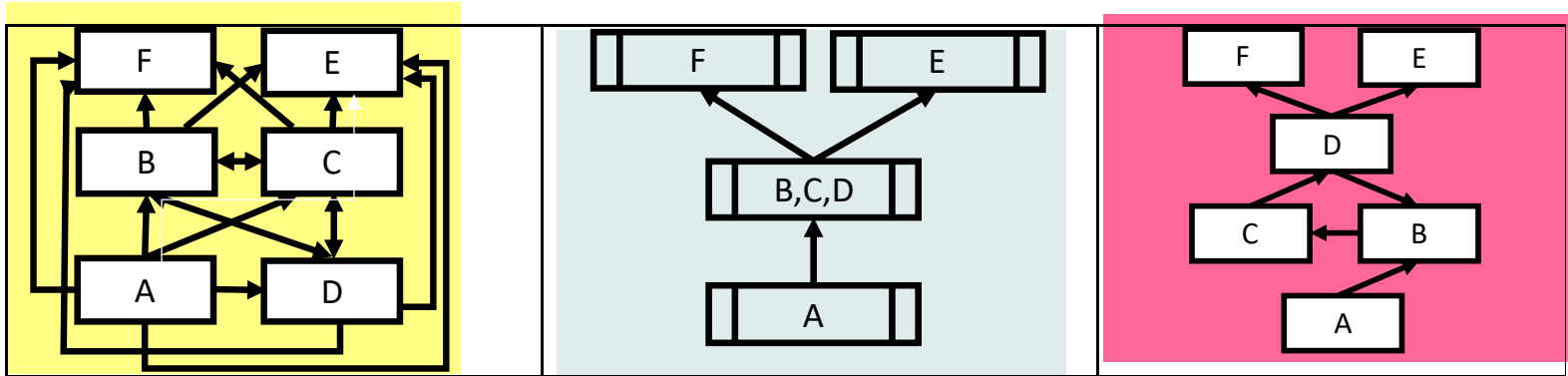
- Graph theory, order theory, relation theory say:
  - Any transitive, reflexive relation can generate a partial order of components
    - Components encapsulate equivalence classes generated by symmetries
- Algorithm theory says:
  - These partial orders can be found in linear time
    - Tarjan, Kosaraju algorithms

# Our conclusions (references at the end)

- Any data network has levels of security
  - From top secret to public (if we want to call them so)
    - Possibly only one level if there is no secrecy whatsoever (only one component)
      - ❖ (our FPS2017 paper)
- Lattice model is a sufficient model for secrecy, *partial order model is necessary and sufficient*
  - And partial orders always exist!
- Partial orders, i.e. levels of security can be found efficiently
  - (our IPL paper)
- Given security constraints, secure data networks can be constructed efficiently
  - (this paper)

# Streamlining the results:

detecting components and their partial order



A data flow graph that could be obtained using a data inclusion relationship

Identifying the partial order of components (linear time algo.)

An equivalent, streamlined graph

# Dynamic configuration of networks

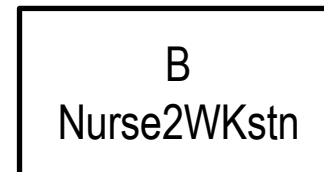
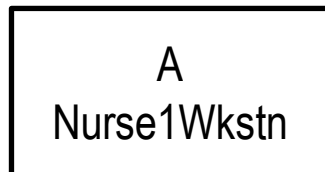
- Entities can be added according to needs
- When a new entity is added, it must come with labels stating what data it can contain
  - CanHold or CH
- Communication channels are created according to inclusion relationships
  - $\text{CanFlow}(X,Y)$  iff  $\text{CanHold}(X) \subseteq \text{CanHold}(Y)$
- The (efficient) partial order detection algorithm is run to clean the graph and leave only the necessary channels

# Application to the Internet of Things

- IoT is a highly distributed, highly dynamic environment where data can flow among “things” in complex data flow configurations
- It is important that “**all and only**” secure data flows be allowed;
  - Available to all intended destinations
  - But only to those
  - As permissive as possible, but also as forbidding as necessary
- Very few **generic** solutions have been proposed for data flow security in the IoT
  - But here is one ...

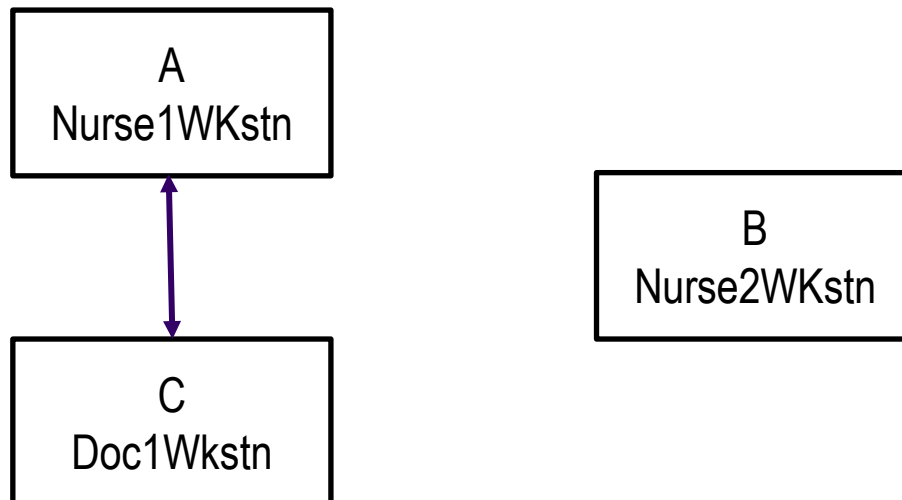
# Hospital devices example (1)

- $\text{New}(A) = \text{Nurse1Wkstn}\{\mathbf{SamPress}, \mathbf{BobPulse}, \mathbf{Stats1}\}$ .
- $\text{New}(B) = \text{Nurse2Wkstn}\{\text{SallyPulse}, \text{Stats2}\}$ .
  - No relation between  $\text{CH}(A)$  and  $\text{CH}(B)$
  - No flow relationships



## Hospital devices example (2)

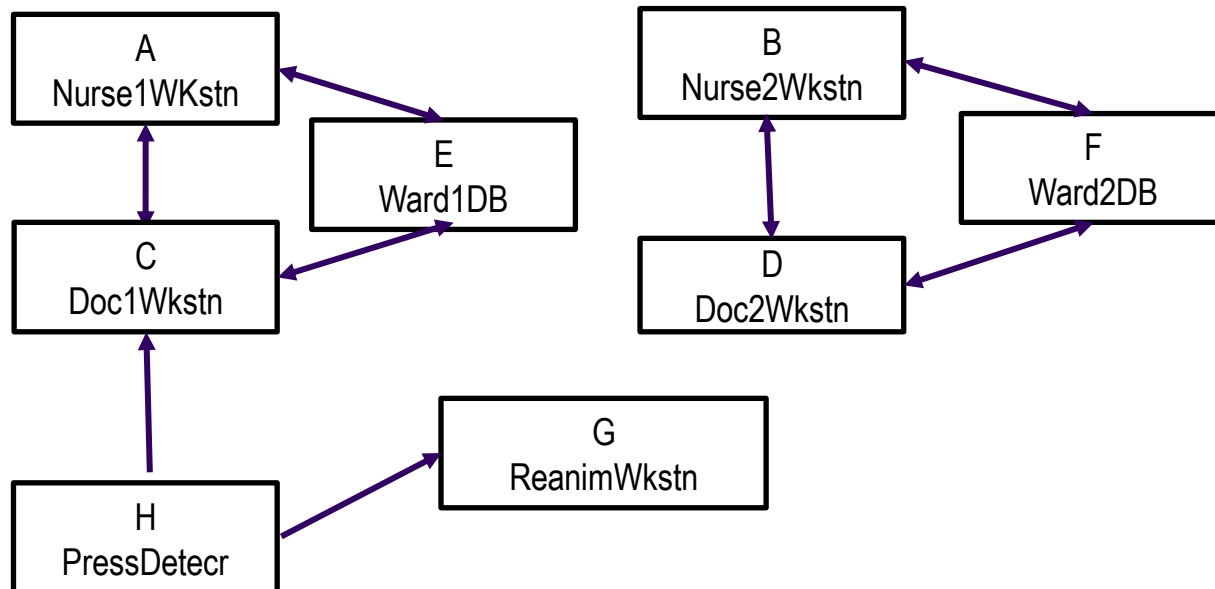
- $\text{New}(A) = \text{Nurse1Wkstn}\{\mathbf{SamPress}, \mathbf{BobPulse}, \mathbf{Stats1}\}$ .
- $\text{New}(B) = \text{Nurse2Wkstn}\{\text{SallyPulse}, \text{Stats2}\}$ .
- $\text{New}(C) = \text{Doc1Wkstn}\{\mathbf{SamPress}, \mathbf{BobPulse}, \mathbf{Stats1}\}$ .
  - Now,  $\text{CH}(C) = \text{CH}(A)$  so  $\text{CF}(C,A)$  and  $\text{CF}(A,C)$



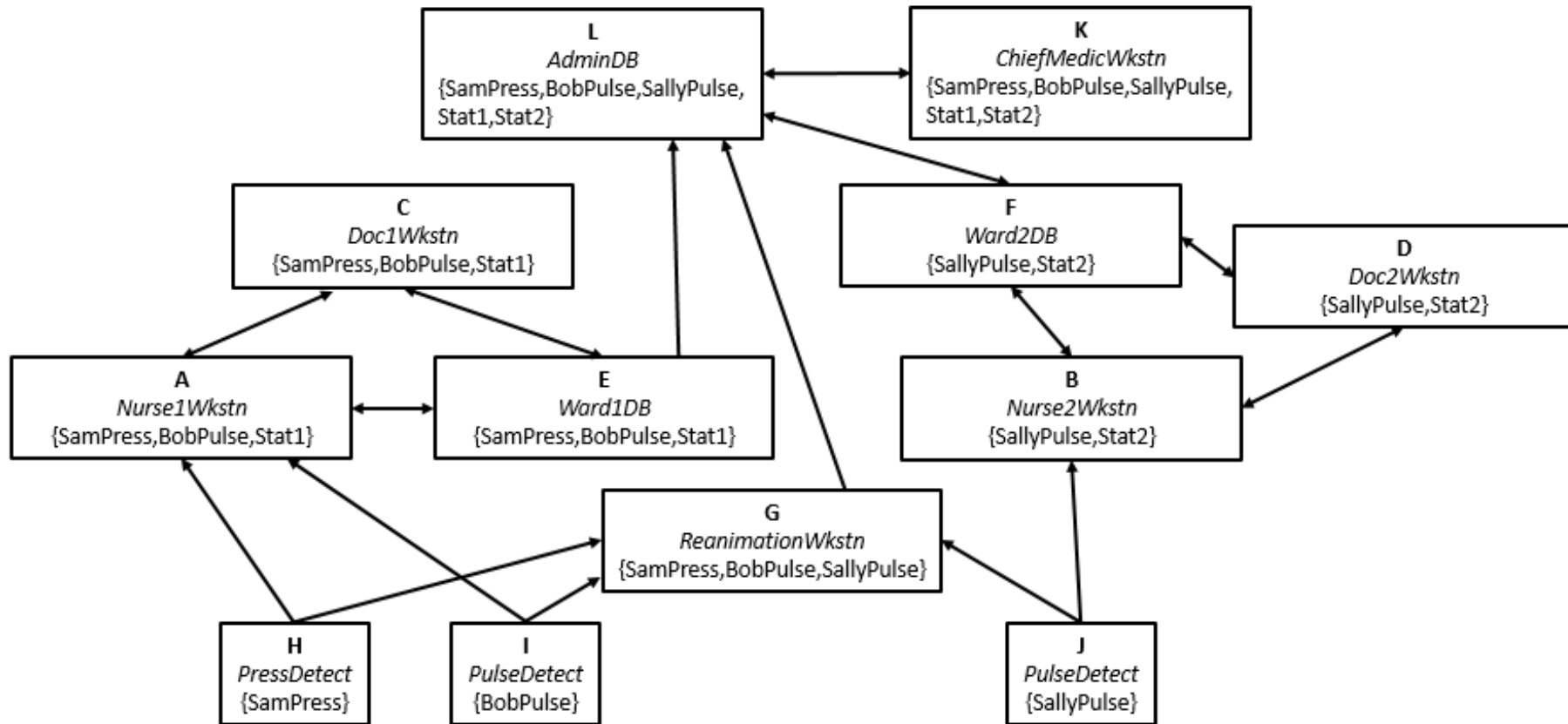


# Hospital devices example (3)

- $\text{New}(D) = \text{Doc2Wkstn}\{\text{SallyPulse}, \text{Stats2}\}$ .
- $\text{New}(E) = \text{Ward1DB}\{\text{SamPress}, \text{BobPulse}, \text{Stats1}\}$ .
- $\text{New}(F) = \text{Ward2DB}\{\text{SallyPulse}, \text{Stats2}\}$ .
- $\text{New}(G) = \text{ReanimationWkstn}\{\text{SamPress}, \text{BobPulse}, \text{SallyPulse}\}$ .
- $\text{New}(H) = \text{PressDetect}\{\text{SamPress}\}$

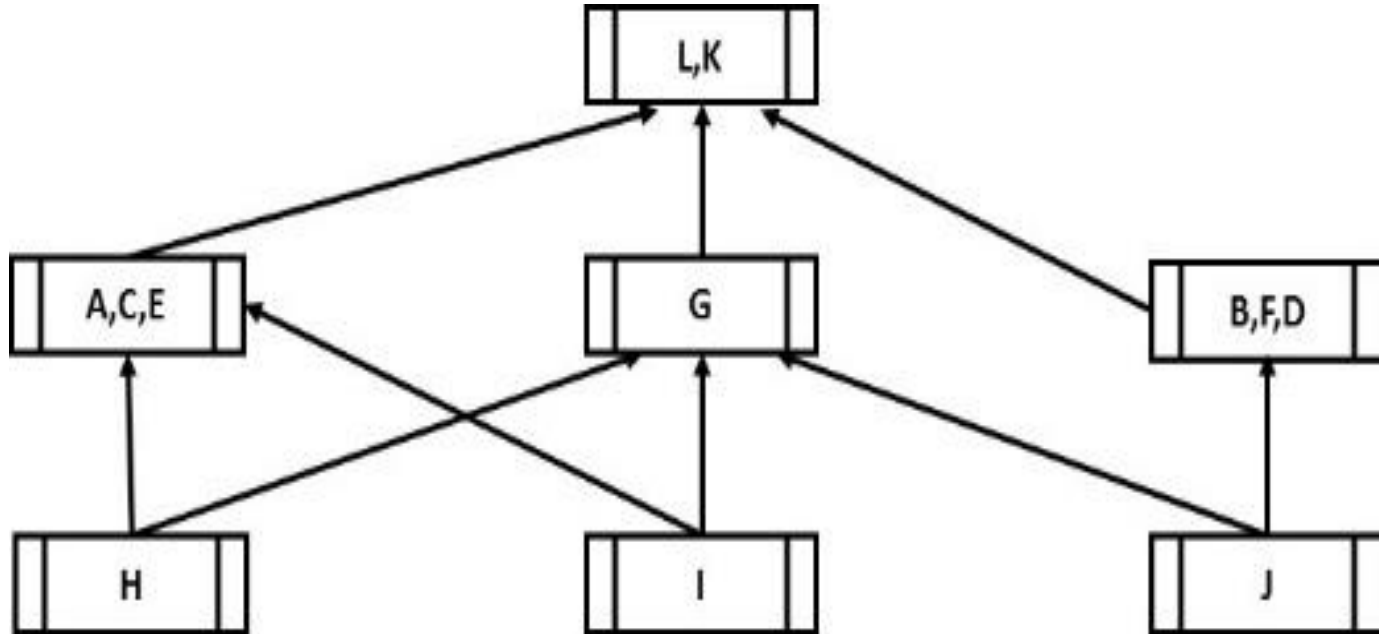


# The full example in the paper



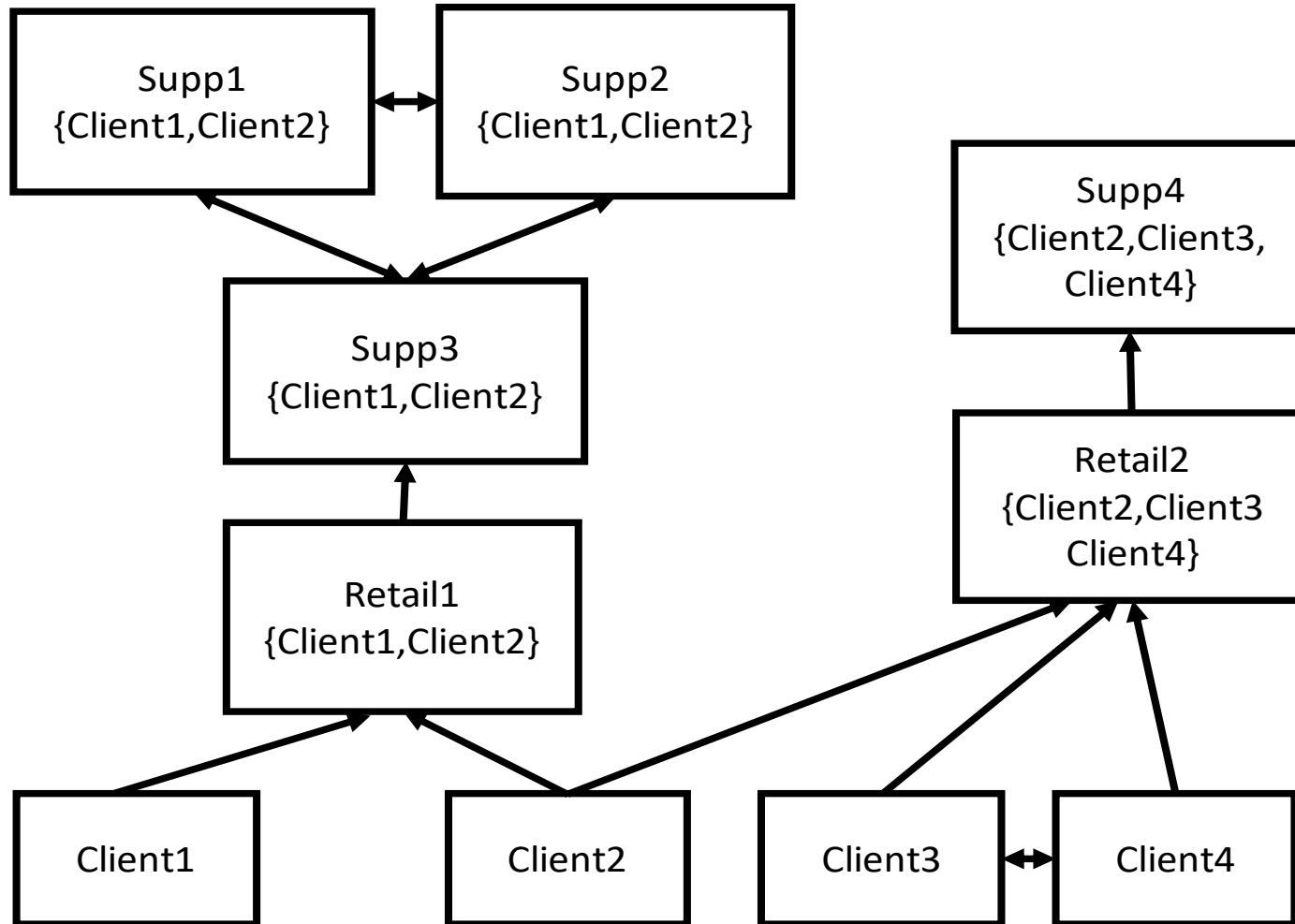
We will get to this structure independently of the order of creation of the entities

# Its partial order of components

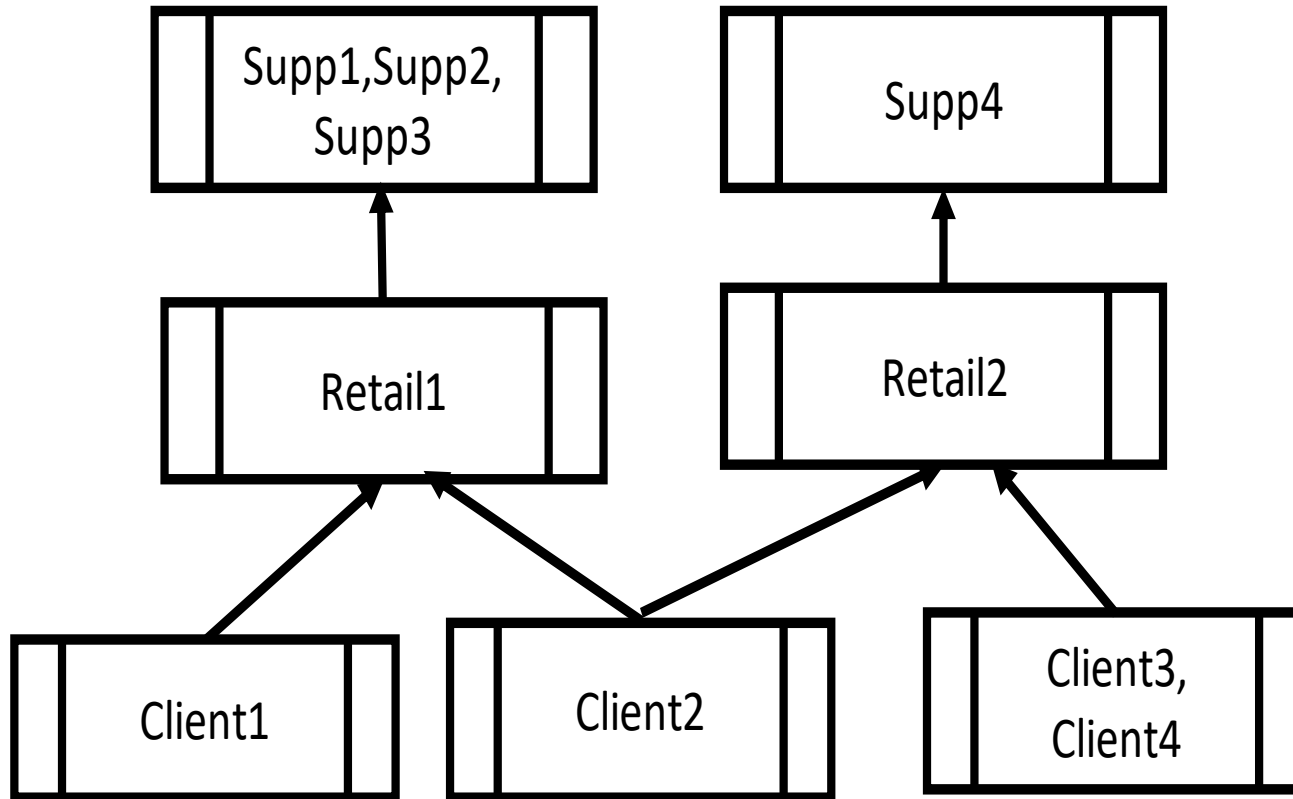


Secrecy grows together with knowledge as we move up

# E-commerce example: orders data flow

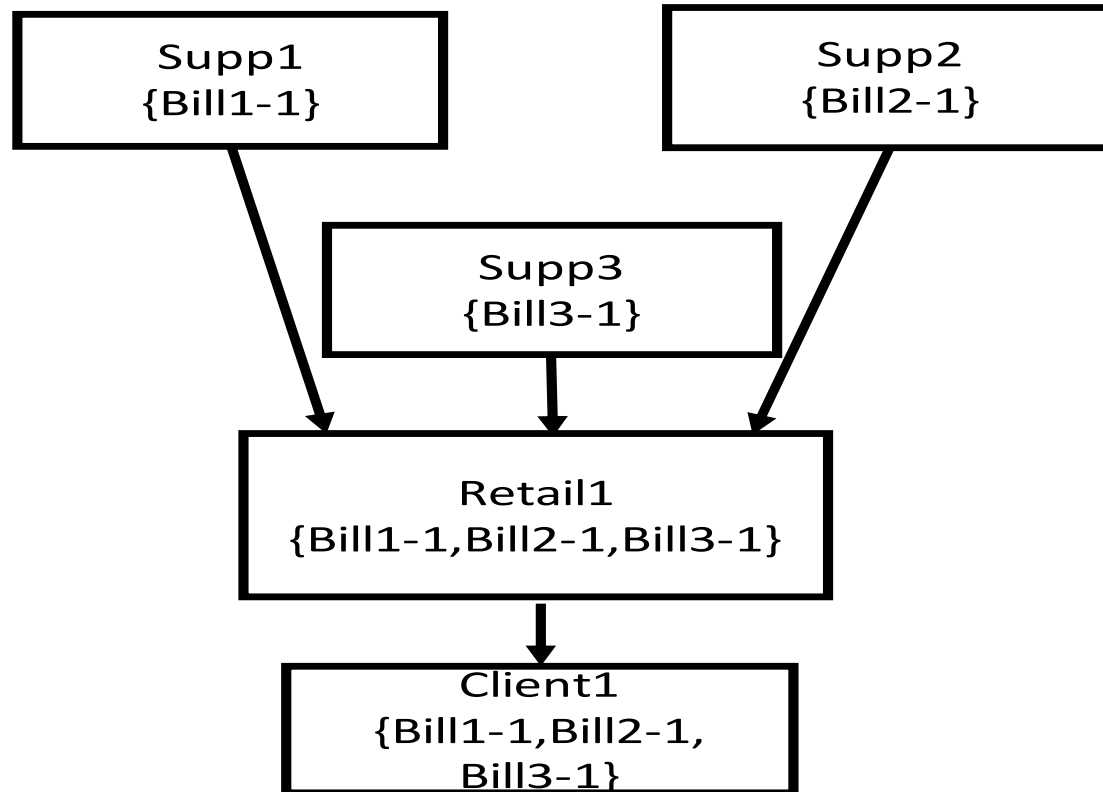


## Partial order of components in e-commerce example



# Another type of data flow in e-commerce

## billing data flow for Client1



# Coexisting data flows

- So, several data flows can coexist in a network
- Our method can handle them, by tagging data according to the data flow to which they belong

# Language primitives for hospital IoT network definition

- For hospital examples, we could have the following types:
  - LType Patient(PatientId) (to define logical type Patient)
  - TType PressDetect(DetectId) (to define a device PressDetect with a DetectId)
  - TType PulseDetect(DetectId) (to define a device PulseDetect)
  - LType Ward(WardId) (to define a logical type Ward)
  - LType Nurse(NurseId)
  - TType NurseWkstn(WkstnId)
- and the following operators:
  - Assign (DetectId, PatientId) (assign a detector to a patient)
  - Assign (PatientId, WardId) (assign a patient to a ward)
  - Assign (NurseId, WardId) (assign a nurse to a ward)
  - Assign (WkstnId, WardId) (assign a workstation to a ward)
  - Etc.



## And CanHold assertions, such as

- ***CH(WkstnId, DetectId)***

- *if Assign(PatientId, WardId(WkstnId)) and Assign(DetectId, PatientId)*

- If a workstation is assigned to a ward and a patient is also assigned to the same ward,
- then there is a data flow (channel) between the workstation and the detectors for the patient

# Progressive network construction:

- New Ward (Emerg).
- New NurseWkstn(EmergWkstn)
- New Nurse (Alice)
- Assign (Alice,Emerg).
- Assign (EmergWkstn,Emerg)
- New Patient (Sam).
- Assign (Sam,Emerg)
- New PressDetect(PRD0001)
- Assign (PRD0001, Sam)



- At this point, by the CH assertion, a channel is created from device PRD0001 to previously created device EmergWkstn
  - The emergency workstation has been assigned to the emergency ward and then Sam has been assigned to device PRD0001 and the same ward

# Re-configurations

- IoT systems should be able to continuously reconfigure
  - Data de-classification and other updates due to changing requirements
  - Entity disappearance
- It might be possible to repair the network locally, or in the worst it might be necessary to execute our generic configuration algorithm

# How to implement this?

- By access control mechanisms
- By routing mechanisms
- By encryption, to implement secure channels
  - If data flow from A to C through B, but B cannot read it, then the channel is only from A to C

# RPL routing for IoT networks

- RPL: Routing Protocol for Low-Power and Lossy Networks
- Uses Directed Acyclic Graphs (DAGs) to express routing in IoT systems
- New devices are placed on DAGs according to Objective Functions (OFs)
- In current use, OFs express mainly efficiency constraints: minimum power usage
- Can our own DAGs be combined with RPL DAGs to include security constraints in RPL routing?
  - If so, it could be possible to program RPL routing to avoid certain nodes if these should not be part of the flow
  - Permissible data paths should run over existing links by using encryption?
    - Research topic ...

# Conclusions

- **Necessary and sufficient** conditions for data secrecy in networks can be obtained by generalizing traditional MAC and lattice concepts
- **Exactness:** By labeling entities in IoT networks according to the type of data they can hold, it is possible to configure data transfer channels such that **all and only** logically allowed flows are possible
  - Both **secrecy and integrity** are taken care of
- **Scalability:** Efficient algorithms exist for such configurations, which makes the solution **scalable** and practical
- **Implementability:**
  - Data must be tagged, entities must be labelled
  - Further research on protocols and encryption is necessary

## Related work

- Although the literature in security and access control in the IoT is vast, there are very few papers with solutions for data flow control in IoT networks
- Previous to us, they were all based on the lattice model
- Many papers on security in IoT do not provide specific solutions

## Some basic references (as of 2018)

- S. Khobragade, N. V. Narendra Kumar, R. K. Shyamasundar. Secure synthesis of IoT via readers-writers flow model. Proc. Intern. Conf. on Distrib. Computing and Internet Techn. (ICDCIT 2018), LNCS 10722, 86–104.
- T. Pasquier, J. Bacon, J. Singh, D. Eysers. Data-Centric Access Control for Cloud Computing. Proc. 21st ACM Symp. on Access Control Models and Technologies (SACMAT '16), 81-88. (+ other papers by same authors)
- Search ‘Luigi Logrippo papers’ and ‘Luigi Logrippo presentations’ and look at the most recent titles