Argumentation-Based Security for Social Good

Erisa Karafili

Imperial College London

April 25, 2018

Erisa Karafili, Antonis C. Kakas, Nikolaos I. Spanoudakis, Emil C. Lupu

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"AF-Cyber: Logic-based Attribution and Forensics in Cyber Security"
Funded from the EU's Horizon 2020 under the Marie Skłodowska-Curie grant agreement No 746667.

Agenda

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Introduction to the Solution

- Two important problems in "social context"
- They can both be seen as decision making problems
- Argumentation reasoning solves problems under partial, conflicting and context dependent knowledge
- Our solution captures different types of conflicts
- We introduce a conflict resolution procedure via priorities between rules

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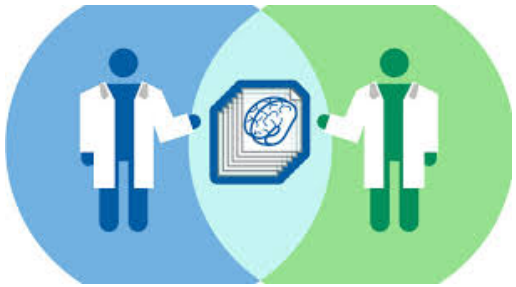
Data Sharing

- Data services are increasing in popularity
- They enable service optimisation and personalisation
- The necessity to protect and ensure the security properties of the data



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Data Sharing Agreements

- Different entities are involved during the sharing of data
- A data sharing agreement is made between the involved entities
  - Data security requirements
  - User preferences
  - Business rules
  - Legislation rules

Challenges:

- Difficult to represent these agreements
- The agreements are applied to the same data in different contextual environment
- The rules of the agreements can create conflicts or not be efficient

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Secure Data Sharing with Argumentation

Solution

A technique based on a policy language and argumentation reasoning for representing and analysing data sharing agreements

Contributions:

- Representation of the rules through arguments
- Efficiency and consistency analysis
- Solve the conflicts by introducing priorities between rules
- An automated decision process decides how and who can access/share/use the data
- The decision process is made using the GorgiasB tool[1]

[1] `http://gorgiasb.tuc.gr/`

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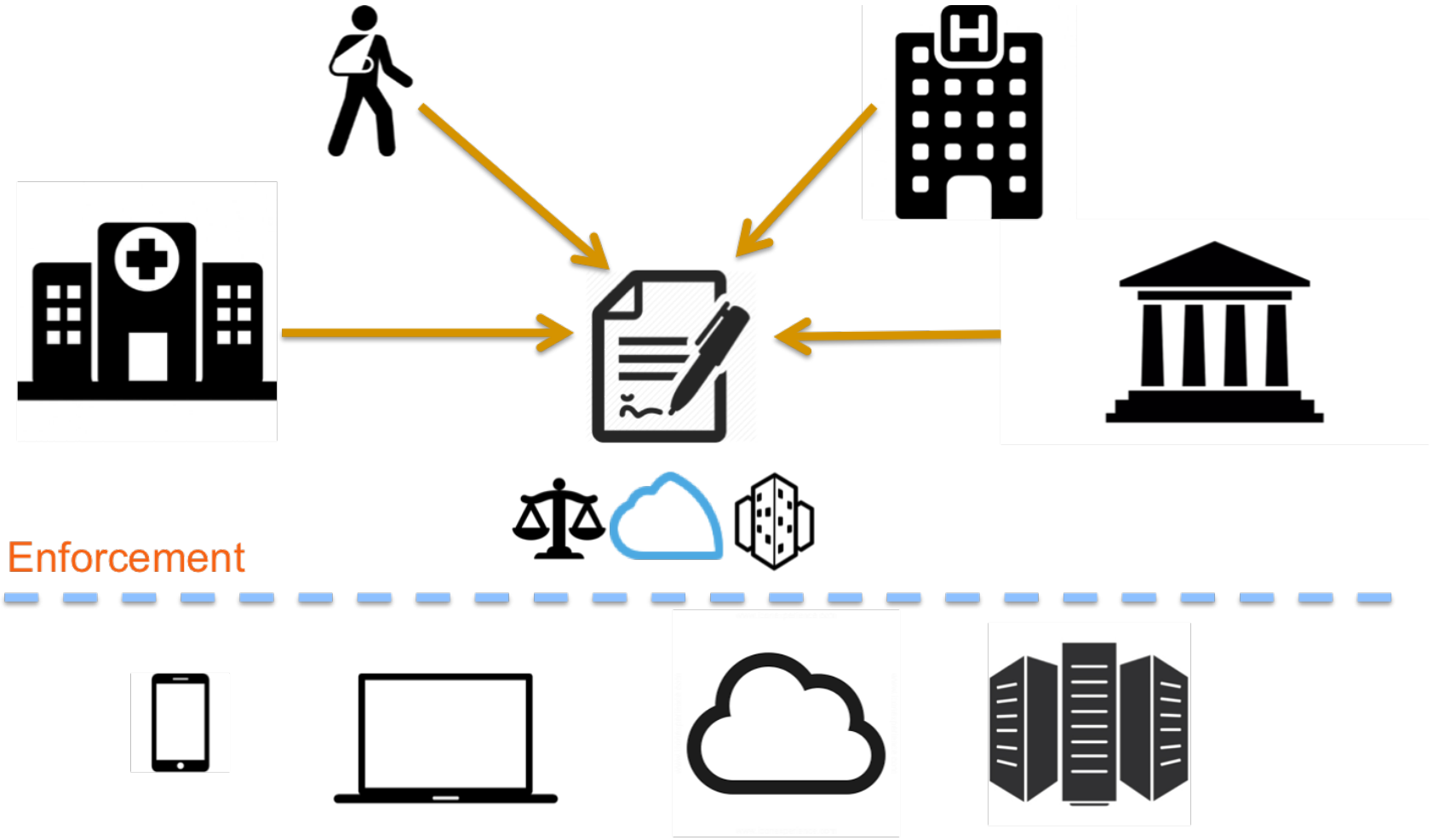
An E-Health Example: Coco Cloud



http://www.coco-cloud.eu/

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DSAs Rules and their Representation

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Some of the rules included in the DSAs:

(1) The patient can access her/his data

$$Access(Data, Patient, Permitted) \leftarrow Owner(Patient, Data)$$

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Some of the rules included in the DSAs:

(1) The patient can access her/his data

(2) The treating doctor can access the patient's data, when s/he is inside the hospital and during her/his shift

$$Access(Data, Doctor, Permitted) \leftarrow TreatD(Doctor, Patient) \wedge Owner(Patient, Data) \wedge shift(D) \wedge hospP(H, L_2) \wedge position(Doctor, L_1) \wedge same(L_1, L_2)$$

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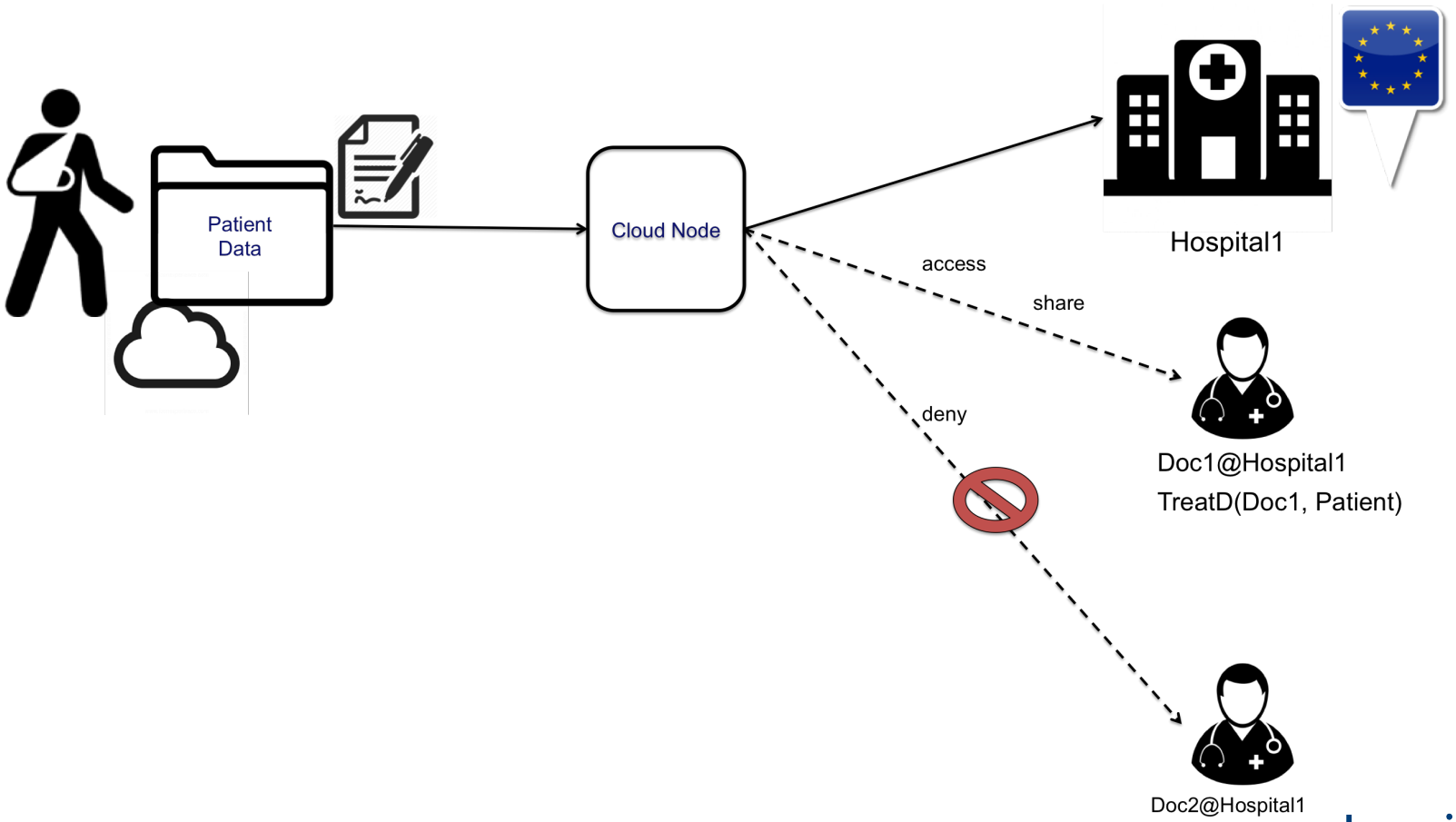
DSAs Rules and their Representation

(2) The treating doctor can access the patient's data, when s/he is inside the hospital and during her/his shift



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(2) The treating doctor can access the patient's data, when s/he is inside the hospital and during her/his shift

(3) The data can be shared inside the EU/EEA, e.g., a second opinion

$$Access(Data, Doctor, Permitted) \leftarrow Owner(Patient, Data) \wedge TDoc(D_1, Patient) \wedge SecondOp(D_1, Doctor) \wedge Work(Doctor, H) \wedge EU^*(H)$$

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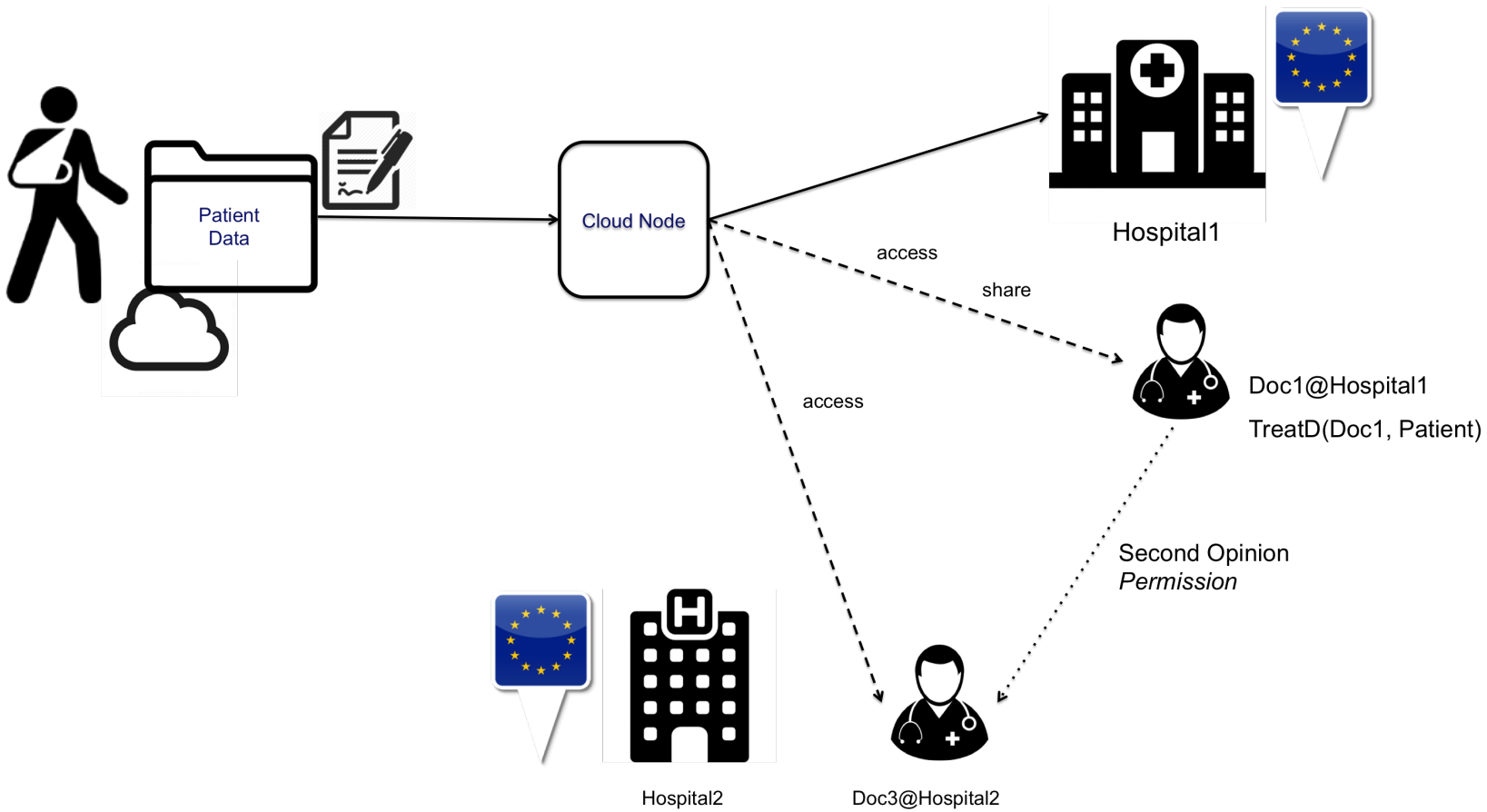

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(4) The data cannot be shared outside EU or EEA

$$Access(Data, Doctor, Denied) \leftarrow Owner(Patient, Data) \wedge Work(Doctor, H) \wedge not\ EU^*(H)$$

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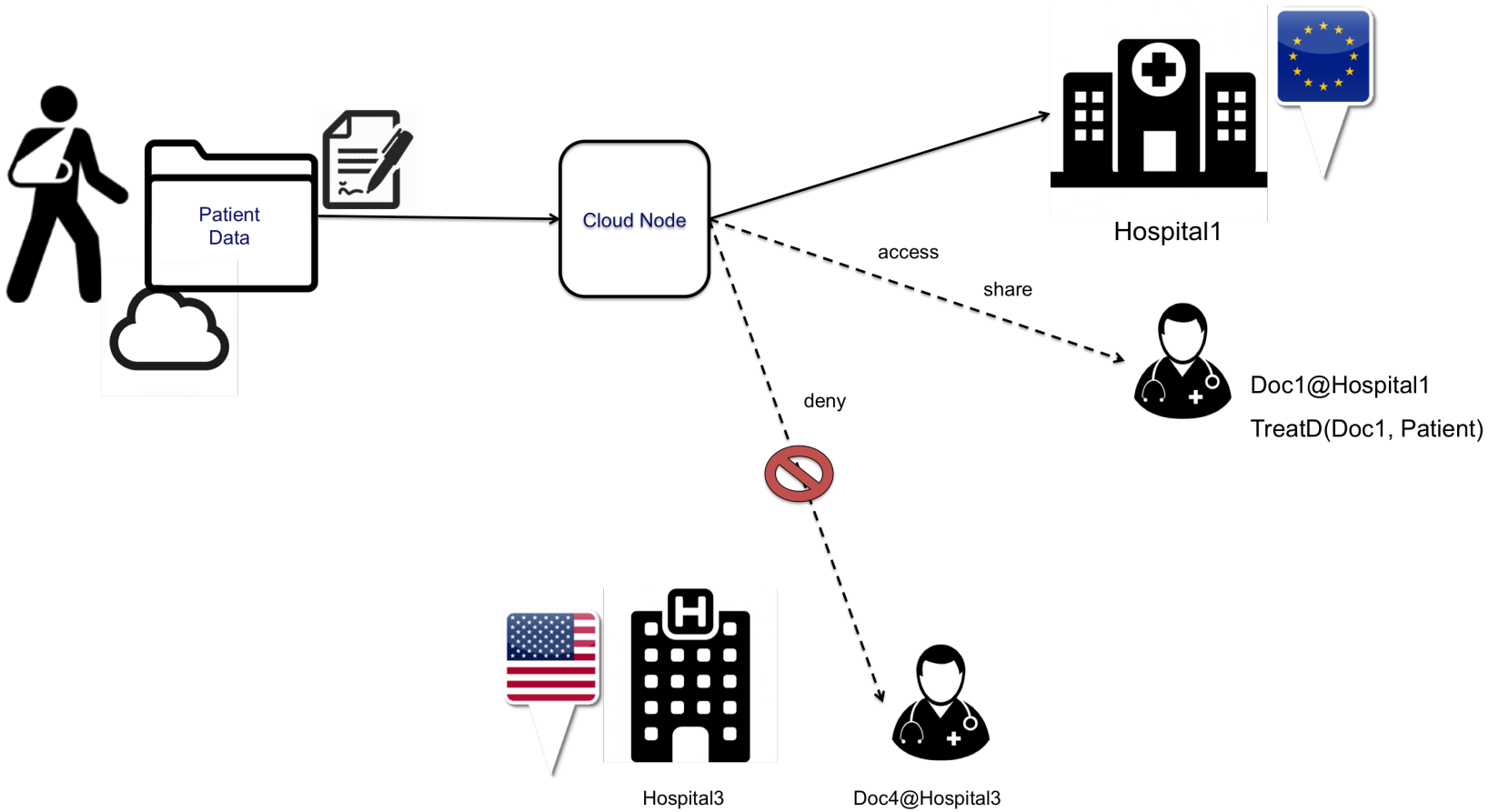
DSAs Rules and their Representation

(4) The data cannot be shared outside EU or EEA



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Conflicting Rules

(5) In case, the patient is in an emergency not in an EU/EEA country, then part of his data can be shared with an entity of that country, if that country has legal agreements for cross borders flow of information with EU

$$Access(Data, Doctor, Permitted) \leftarrow Emergency(Patient, H) \wedge Owner(Patient, Data) \wedge Work(Doctor, H) \wedge \textbf{not } EU^*(H) \wedge Agreement(H)$$

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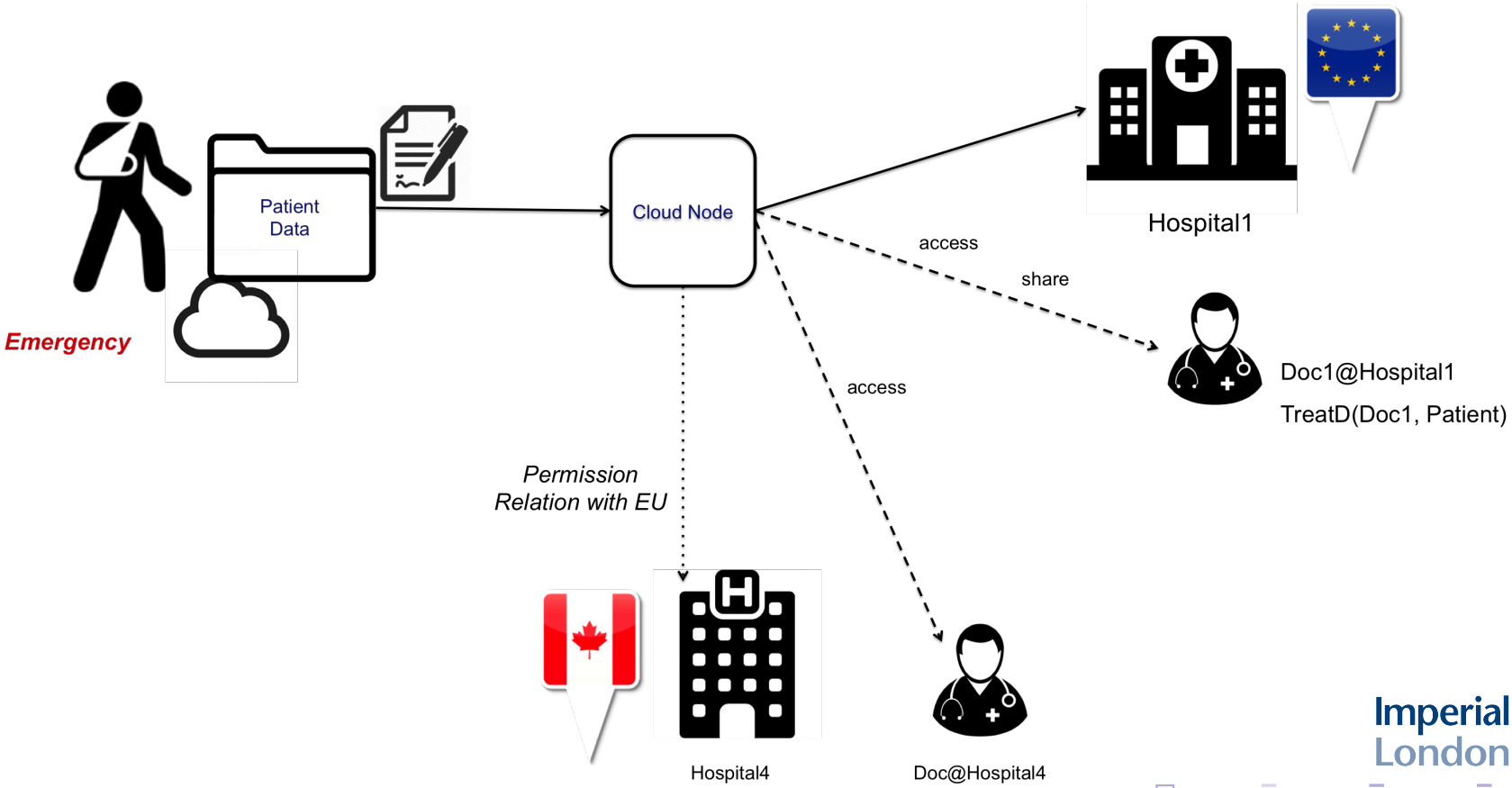

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$$Access(Data, Doctor, Permitted) \leftarrow Emergency(Patient, H) \wedge Owner(Patient, Data) \wedge Work(Doctor, H) \wedge \textbf{not } EU^*(H) \wedge Agreement(H)$$

- The introduced policy analysis is able to find the conflict between rules (4) and (5)
- The argumentation based decision process solves this conflict by introducing a priority between the rules

$$(5) > (4)$$

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The Future is Interconnected

In 2020 there is an expectation of more than 20 billions of IoT devices connected. (McAfee labs)

- The growing of connectivity increases the security challenges

"Every minutes, we are seeing about half a million attack attempts that are happening in Cyber Space"(Fortinet)

- The cost of Cyber Crime Damage by 2021 will reach $6 Trillion (Cybersecurity Ventures)

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The Attribution Problem

Attribution in cyber attacks is the process of assigning an action to a particular actor/entity/country

Problem

Given evidence of an attack, decide who did/performed/instigated the attack

- Forensics helps in the attribution process
- The evidence is incomplete and/or conflicting

Solution

A methodology based on argumentation reasoning and social science techniques

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Attribution in Cyber Attacks

- We propose a methodology based on Adbuctive and Argumentation reasoning
- The attribution reasoner is based on logical rules
- The knowledge based is structured through a Social Science model (Q-model)
- Implementing *physical* as well as *social attribution*

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Attribution through Argumentation

- Pieces of evidence are represented as facts and defeasible knowledge
- The rules are defined as arguments for certain conclusions
- Hierarchies are introduced between arguments
- The reasoner decides the winning argument
- The reasoner is implemented using tools for preference-based argumentation
- An explanation is provided for the given attribution

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Attribution with Argumentation and Social Science

- The evidence is categorised and analysed following a social science approach
- The reasoner can answer if a given entity performed the attack

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An example of Attribution in Cyber Attack

- HIDS logs check: SSH brute force/dictionary attack
- Firewalls logs check: IP's sources of the attack
  - Geolocation of the IP's
  - IP's spoofed, that country did not performed the attack
  - The attack is designed to avoid a certain country, then that country performed the attack

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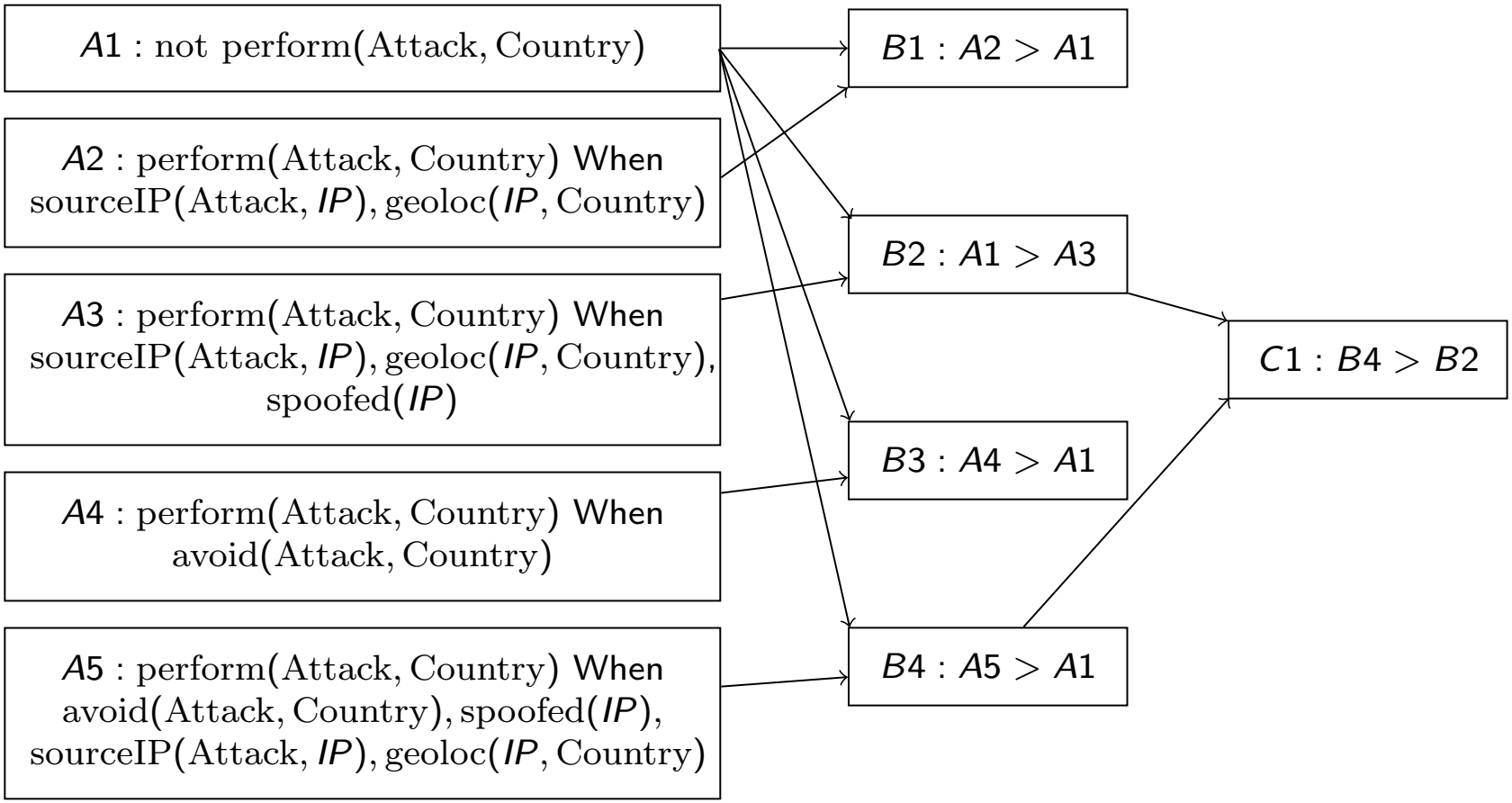
Decision Diagram for the Attribution example



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Further Reasoning Rules and Priorities

Consider complex examples of attacks, where social attribution is involved

- Language(Attack, Country)
- Motive(Attack, Country)
- Capable(Attack, Country)
- Target(Attack, Country)

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Conclusions

- We presented a solution for
  - Regulatory data sharing
  - Cyber attack attribution
- The solution is based on argumentation reasoning
- Decision making mechanism under incomplete, conflicting and context dependent information

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Ongoing and Future Work

Ongoing Work:

- Collect and categorise the various pieces of evidence
- Extract the reasoning rules applied in various use cases
- Construct and enrich the reasoner
- Extend the attribution solution to guide the analysts during evidence collection/analysis

Future Work:

- Quantitative arguments strength
- Construct a Logical Framework for Attribution
- Work on human cognitive reasoning for the social evidence
- Fully automate the conflict resolution process

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Questions?

e.karafili@imperial.ac.uk
http://www.imperial.ac.uk/people/e.karafili
http://rissgroup.org/
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Imperial College
London

References

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2. Erisa Karafili, Emil C. Lupu "Enabling Data Sharing in Contextual Environments: Policy Representation and Analysis" in SACMAT 2017, 231-238, 2017.
3. Erisa Karafili, Konstantina Spanaki, Emil C. Lupu "An argumentation reasoning approach for data processing" in *Journal of Computers in Industry*, Elsevier, Volume 94, 52-61, 2018.
4. Thomas Rid, Ben Buchanan "Attributing cyber attacks" in *Journal of Strategic Studies*, 38(1-2):4–37, 2015.

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