



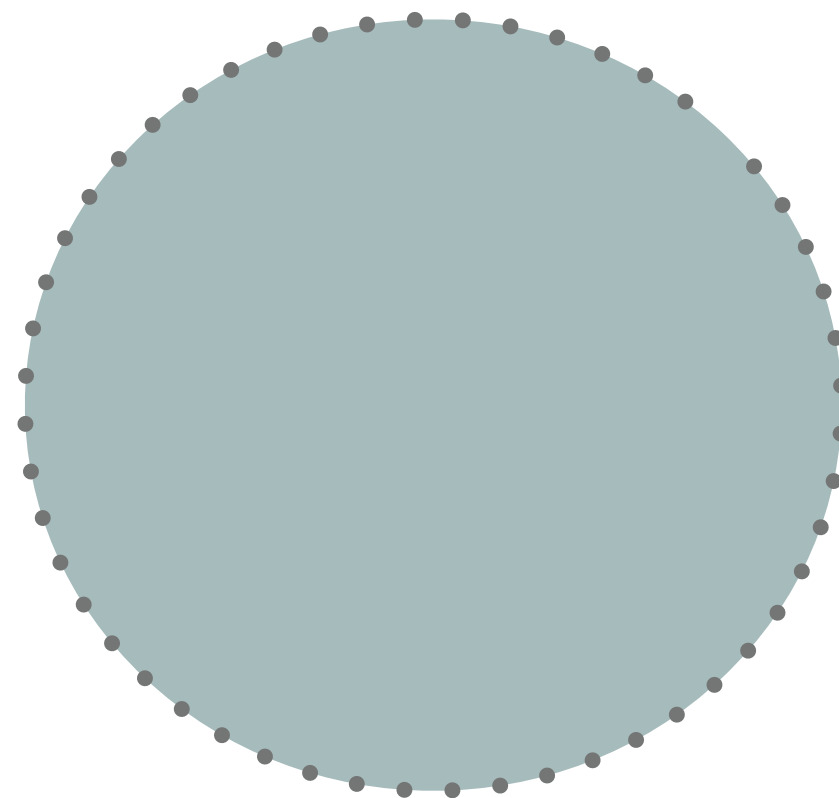
HANDLING CONFLICTING REQUIREMENTS

*Paolo Ceravolo - SESAR Lab - Dipartimento di Informatica
Università degli Studi di Milano*



OUTLINE

- Introduction
- Examples
- Techniques
- Summary



INTRODUCTION

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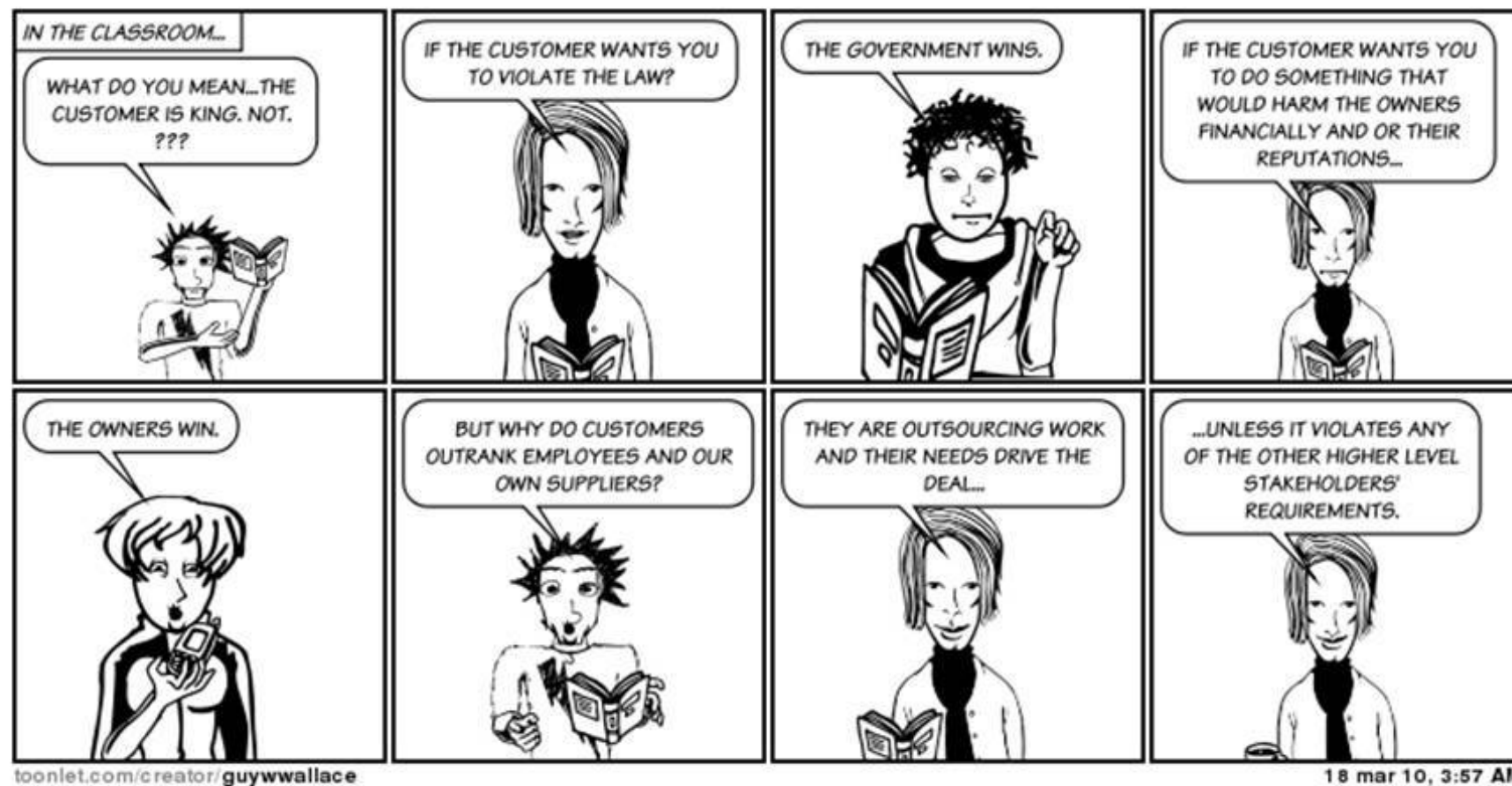
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EXAMPLE 1 – MULTIPLE STAKEHOLDERS

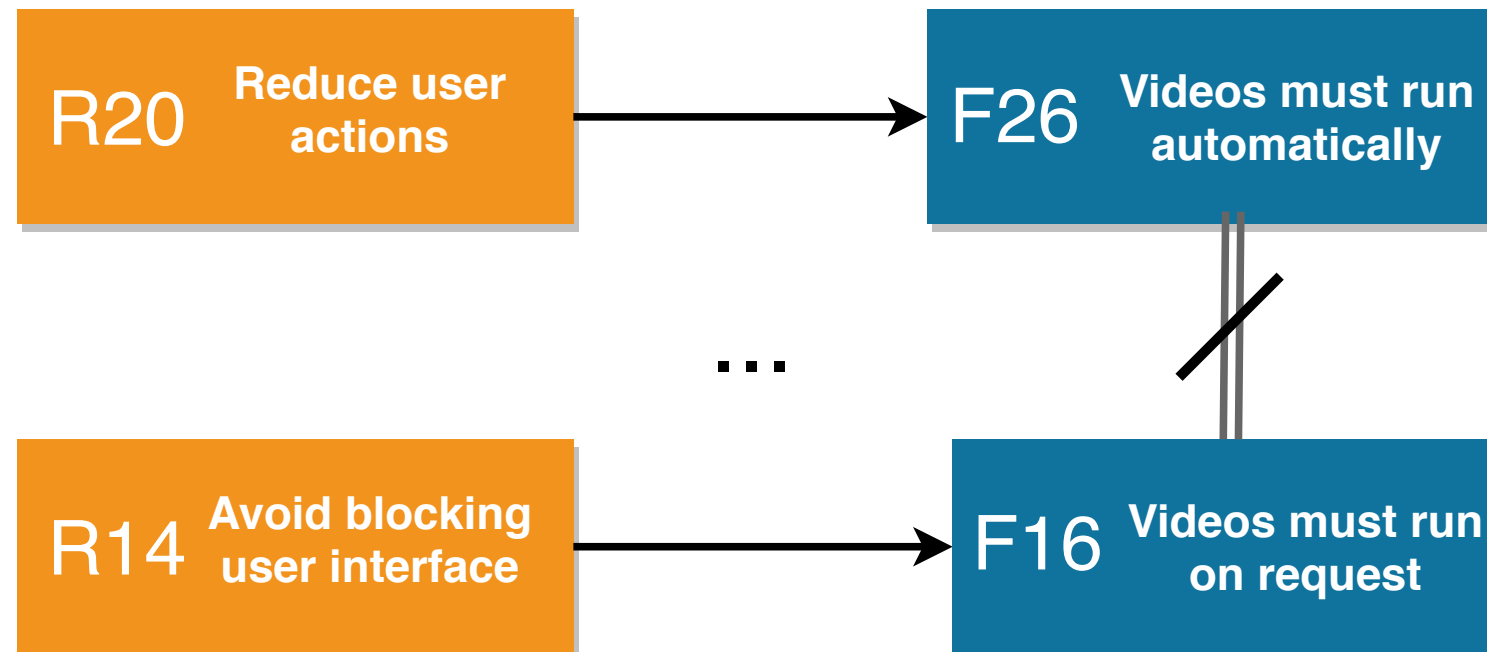
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The Customer Is King. Not. Not in a Stakeholder Hierarchy.



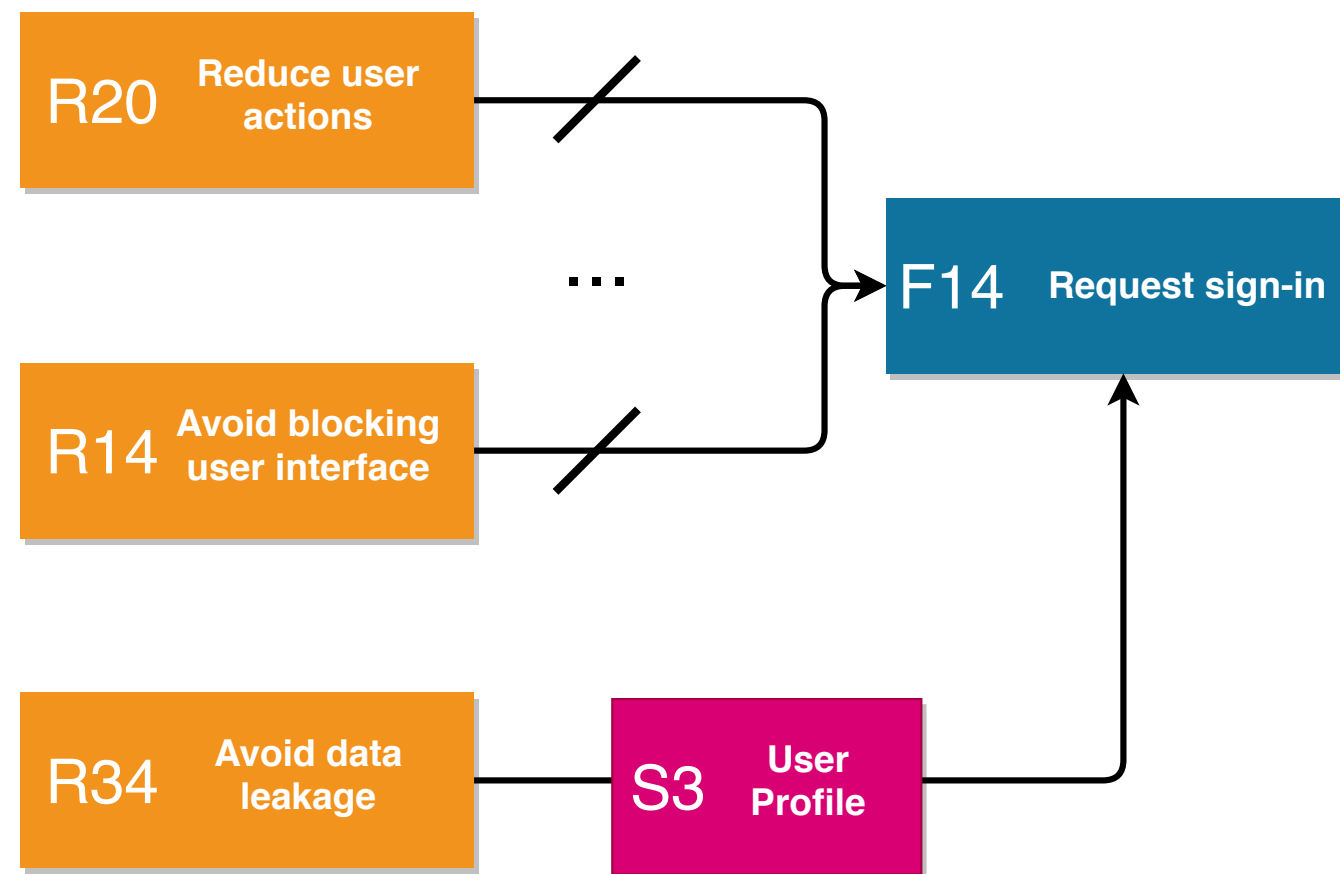
EXAMPLE 2 – CONFLICTING FEATURES

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EXAMPLE 3 – CONFLICTING DEPENDENCIES

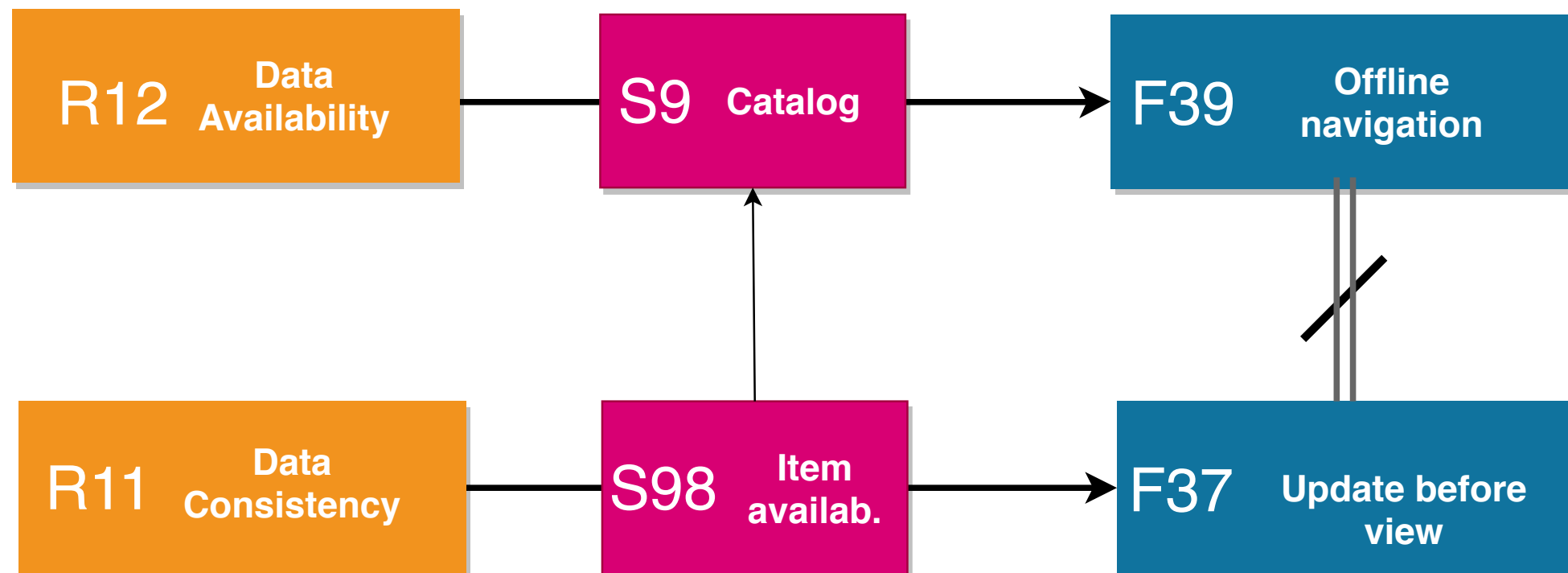
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EXAMPLE 4 – CONFLICTING DEPENDENCIES FOR KR



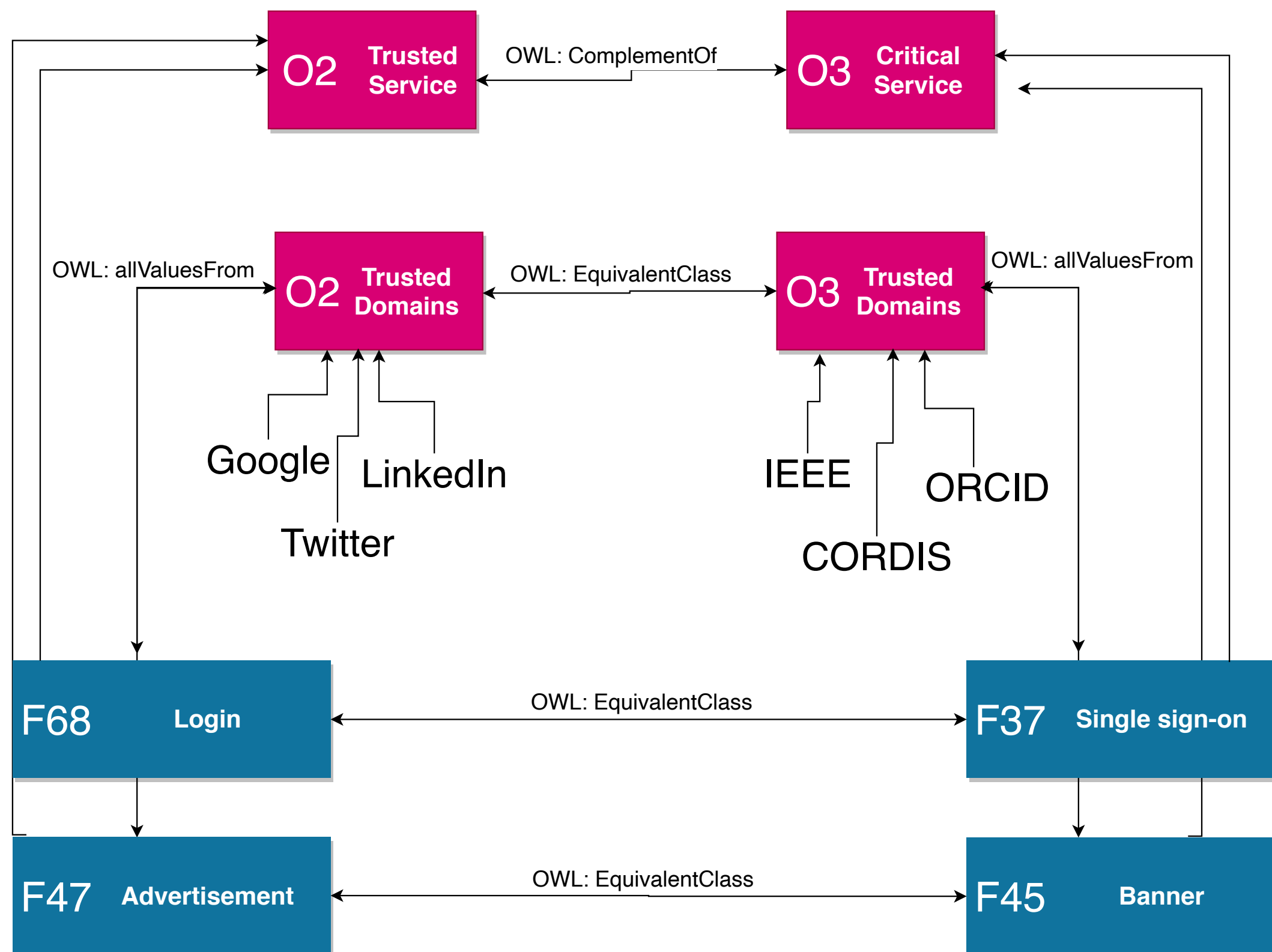
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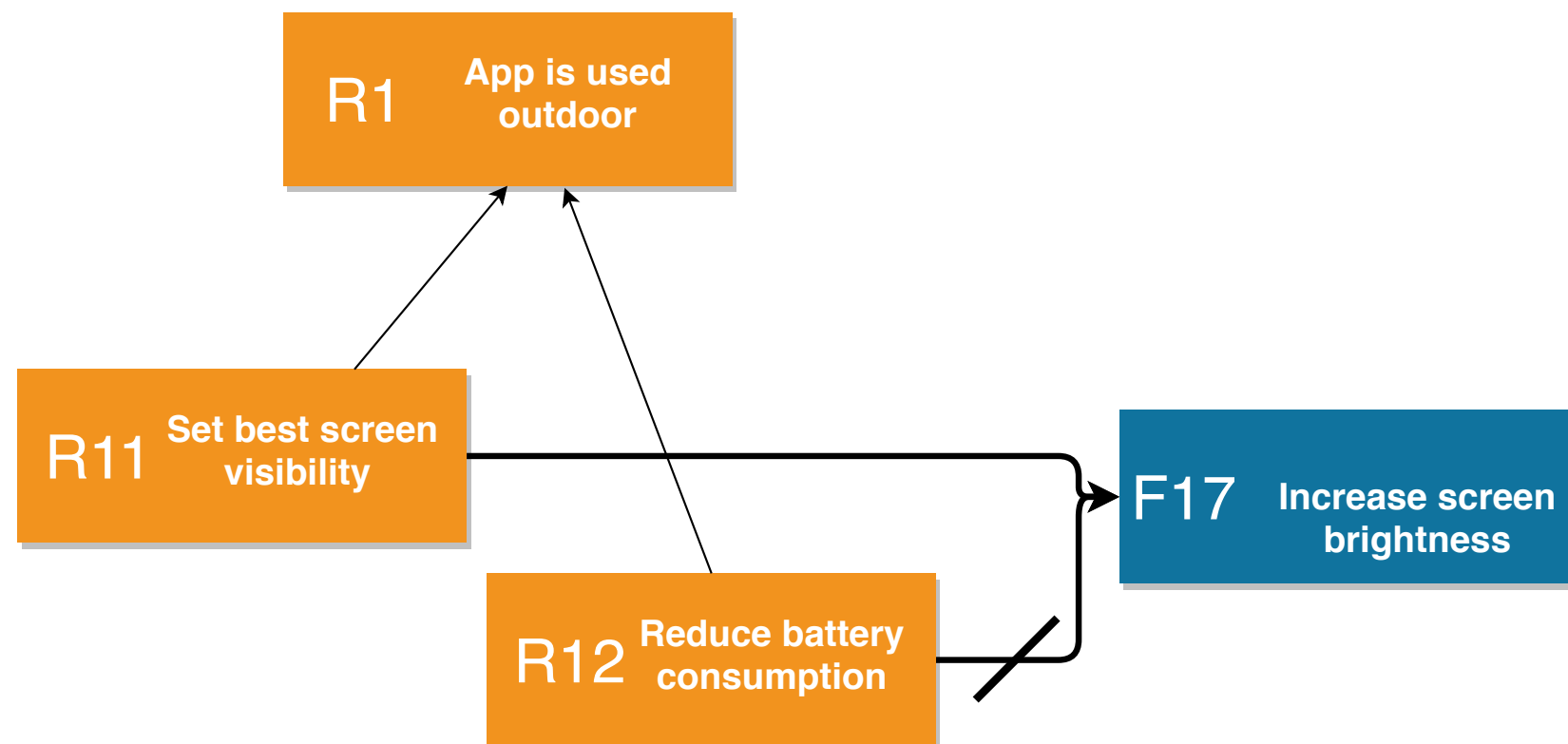
EXAMPLE 5 – CONFLICTING KR



EXAMPLE 6 – APPARENT CONFLICTING KR



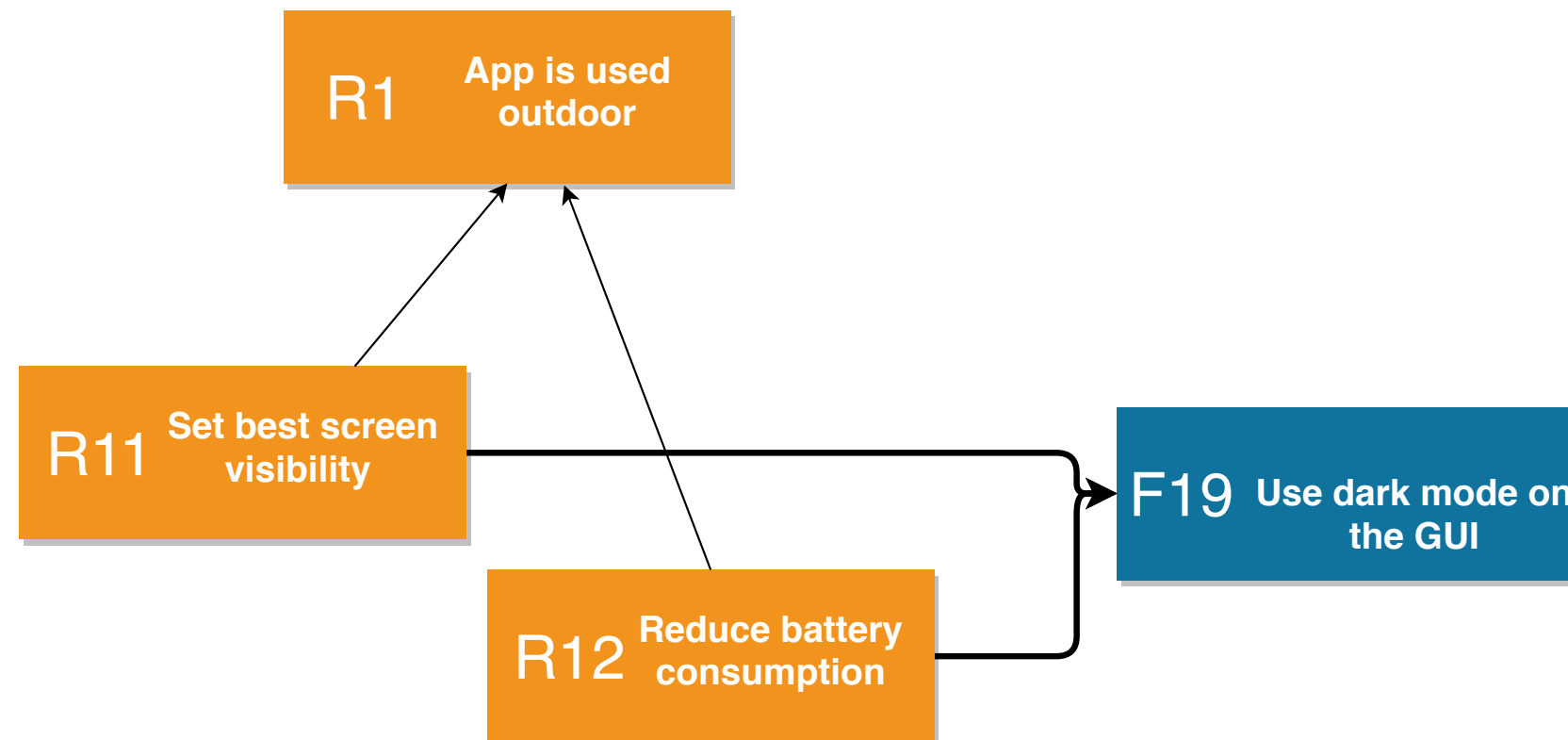
EXAMPLE 6 – APPARENT CONFLICT



EXAMPLE 6 – APPARENT CONFLICT: NEW SOLUTION

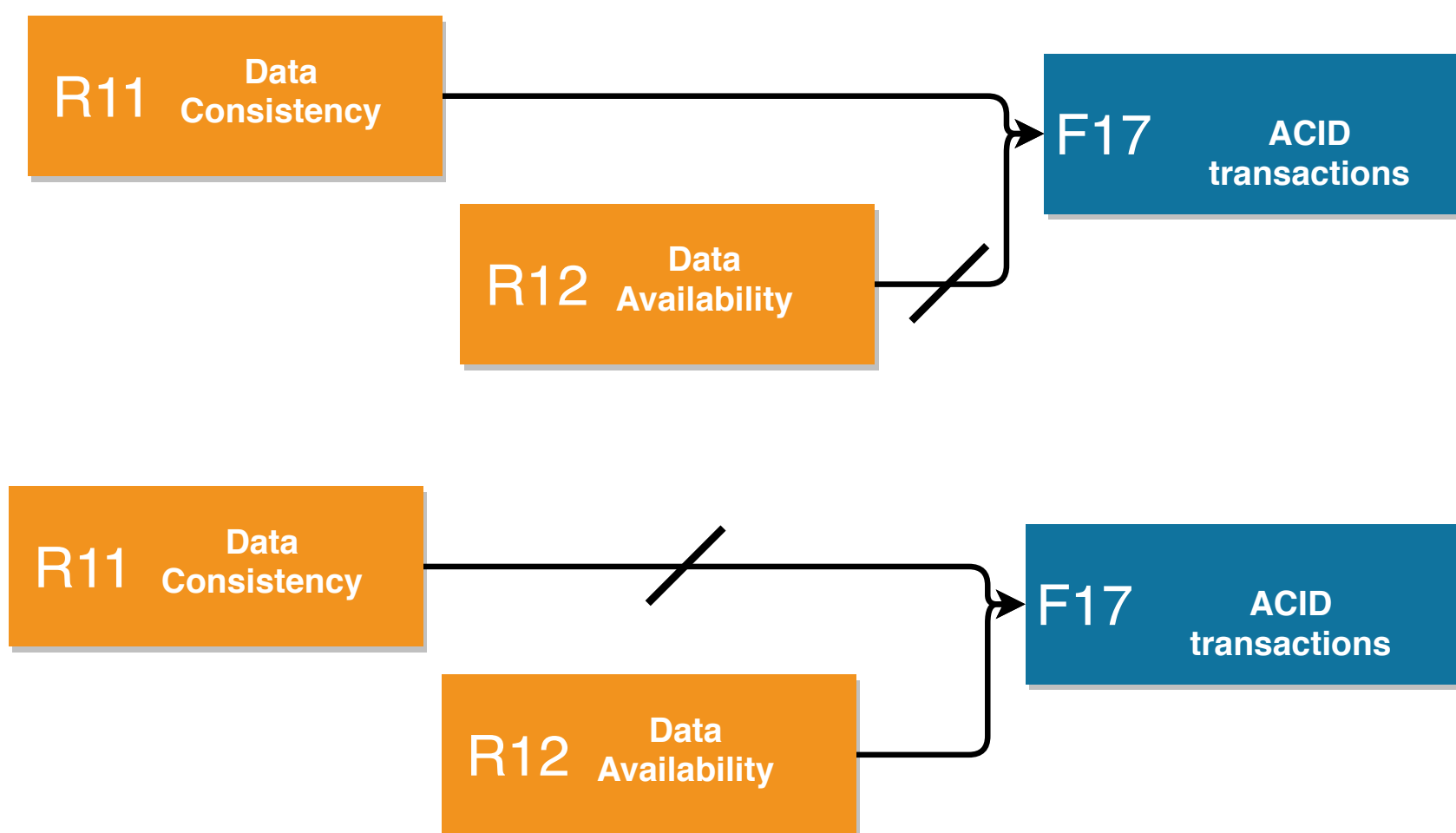


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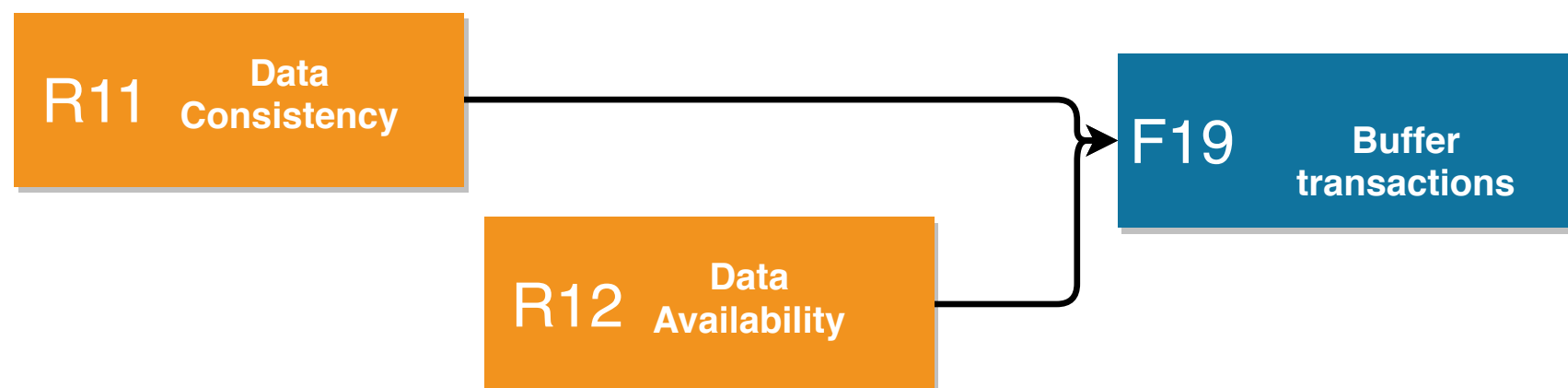
EXAMPLE 7 – APPARENT CONFLICT

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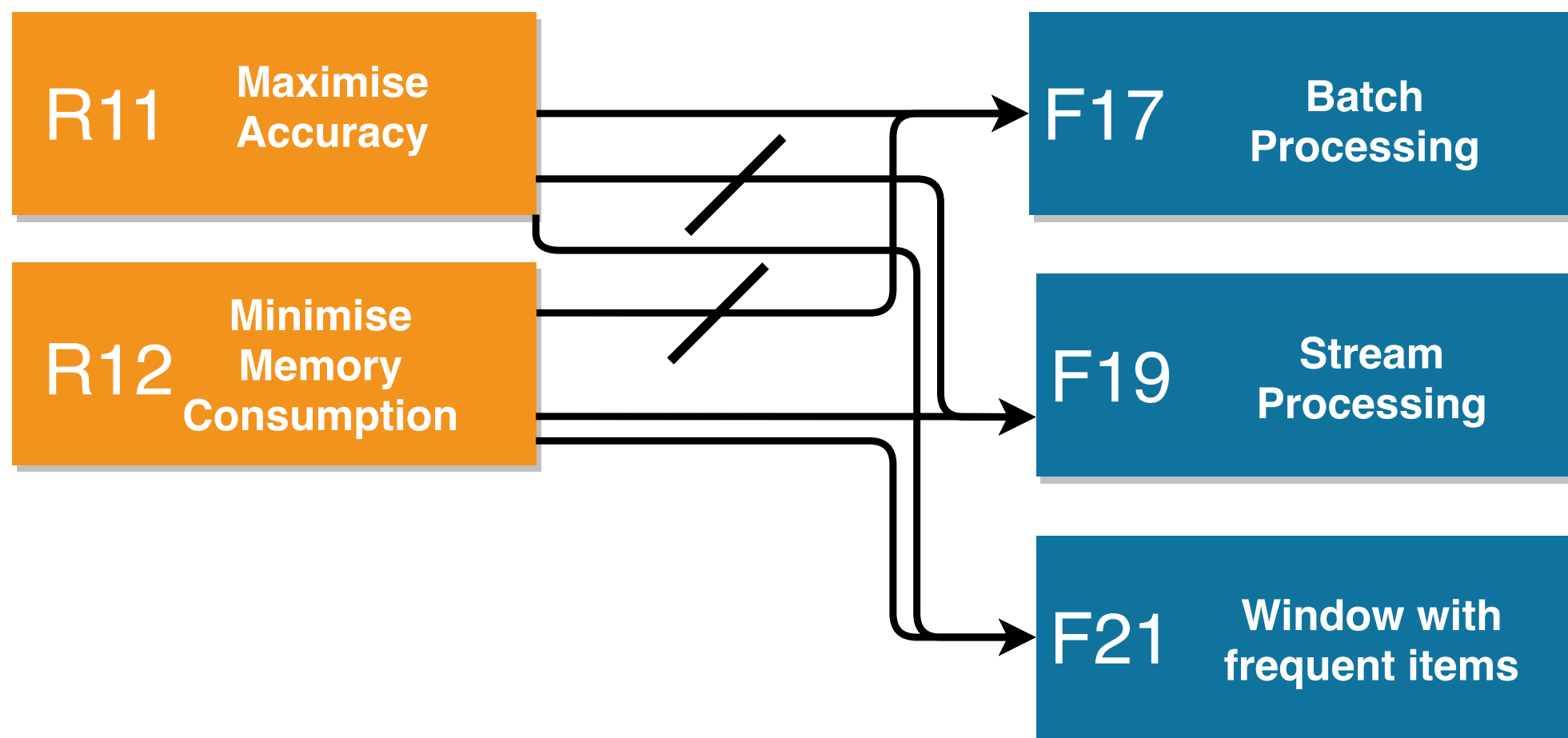


EXAMPLE 7 – APPARENT CONFLICT: TRADE-OFF



EXAMPLE 8 – APPARENT CONFLICT: TRADE-OFF

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- If requirements and specifications are associated with a **formal semantics** we can use automated reasoning to detect inconsistencies
- Following the **classical principle of non-contradiction**, KR languages consider invalid any set of assertions bringing to a contradictory proposition
- To solve inconsistencies the goal is **isolating the minimal inconsistent subset**
 - This operation **grows exponentially** with the number of assertions in the knowledge base
 - Greedy methods can be used but human intervention cannot be avoided

- In Hussain et al. an example following the structure below generated 98132 possible **resolve candidates**

```
# Ontology Org A
o1:UserAccount
  rdfs:subClassOf
    [ a owl:Restriction ;
      owl:maxCardinality 1;
      owl:onProperty o1:recoveryPhone;
    ]

# Data Org B
o2:UserA o2:recovery "+393287738" .
o2:UserA o2:recovery "+393287654" .

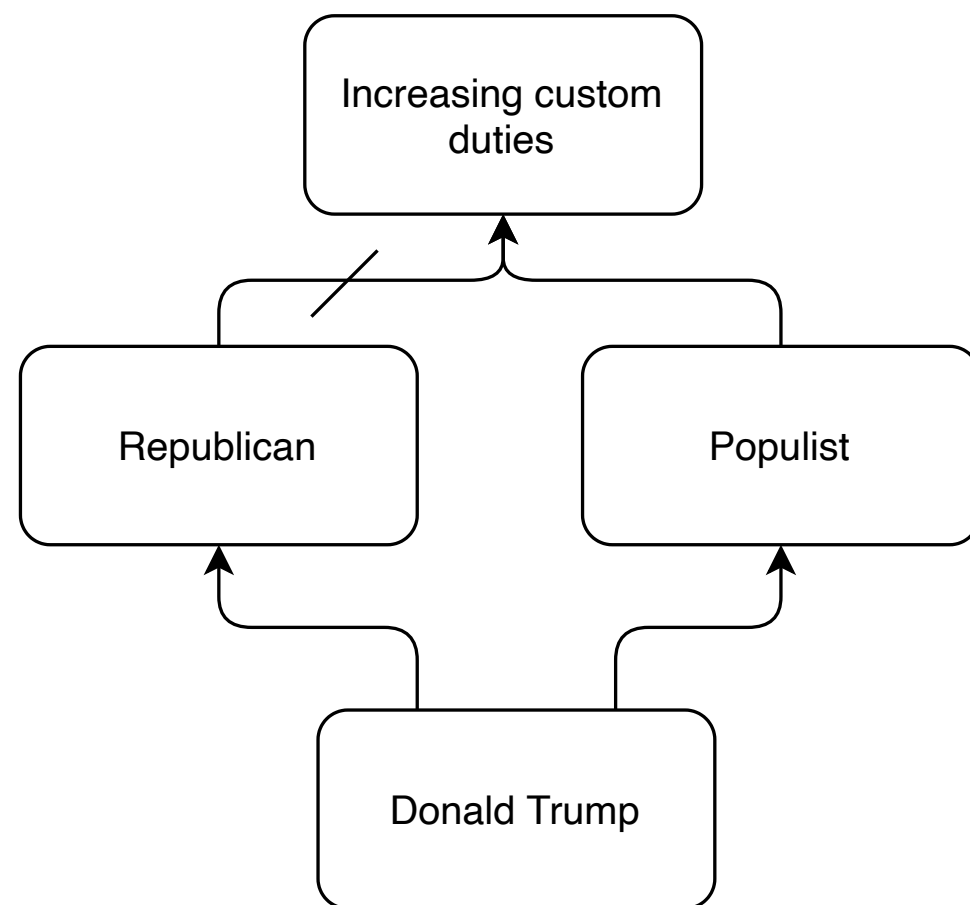
# Mapping ontologies A and B
o2:UserA rdf:type o1:UserAccount .
"+393287738" owl:differentFrom "+393287654" .
o1:recoveryPhone owl:equivalentProperty o2:recovery .
```

- Defeasible reasoning is an alternative
 - Reasoning does not produce a full or final demonstration
 - Its conclusions go beyond the premises
 - Integrating common sense and expert knowledge

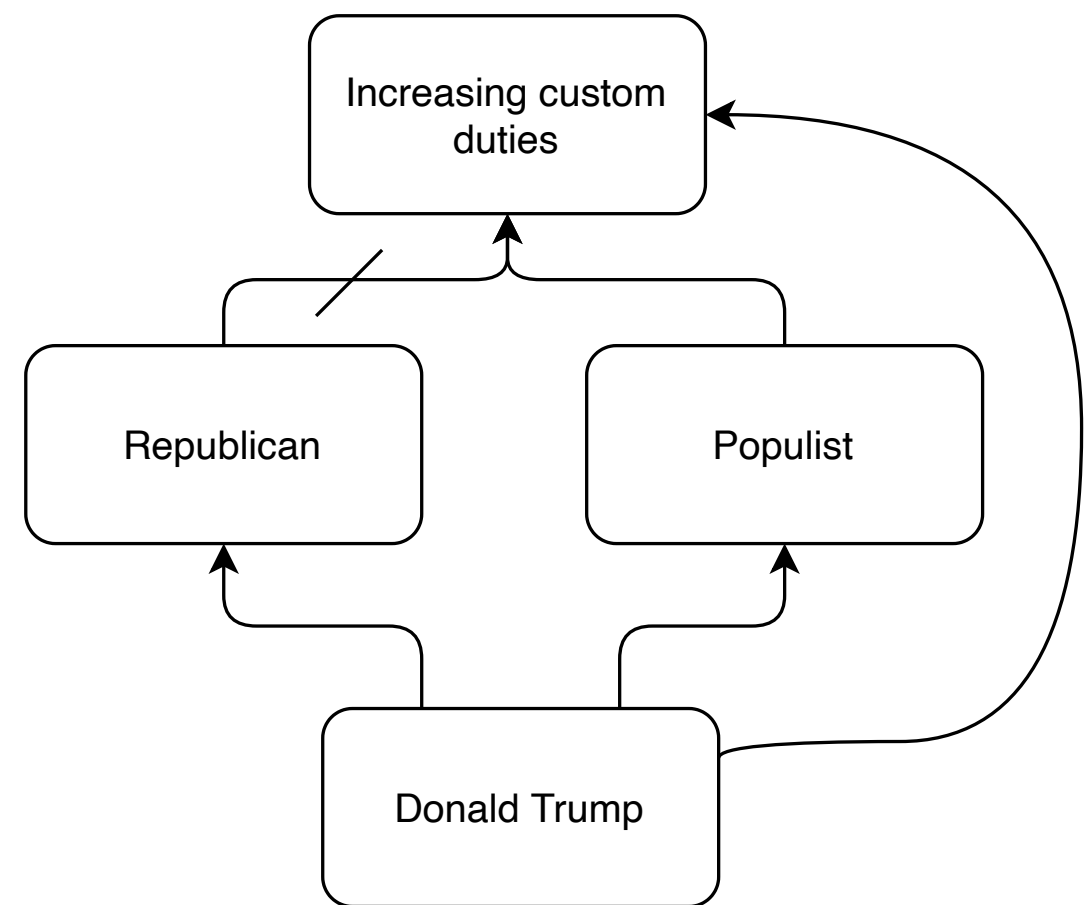
$$T_{8.4} = \left\{ \begin{array}{l} r_1 : \text{Bird}^* \sqsubseteq \text{Flies}^* \sqcap \text{Animal} \\ r_2 : \text{Eagle} \sqsubseteq \text{Bird} \\ r_3 : \text{Penguin}^* \sqsubseteq \text{Bird}^* \sqcap (\neg \text{Fly})^* \end{array} \right\}.$$

$$Lam_{\text{Penguin}}(T_{8.4}) = \text{Bird}, \text{Flies}, \text{Penguin}, (\neg \text{Flies}).$$

- Defeasible reasoning is an alternative



t_1



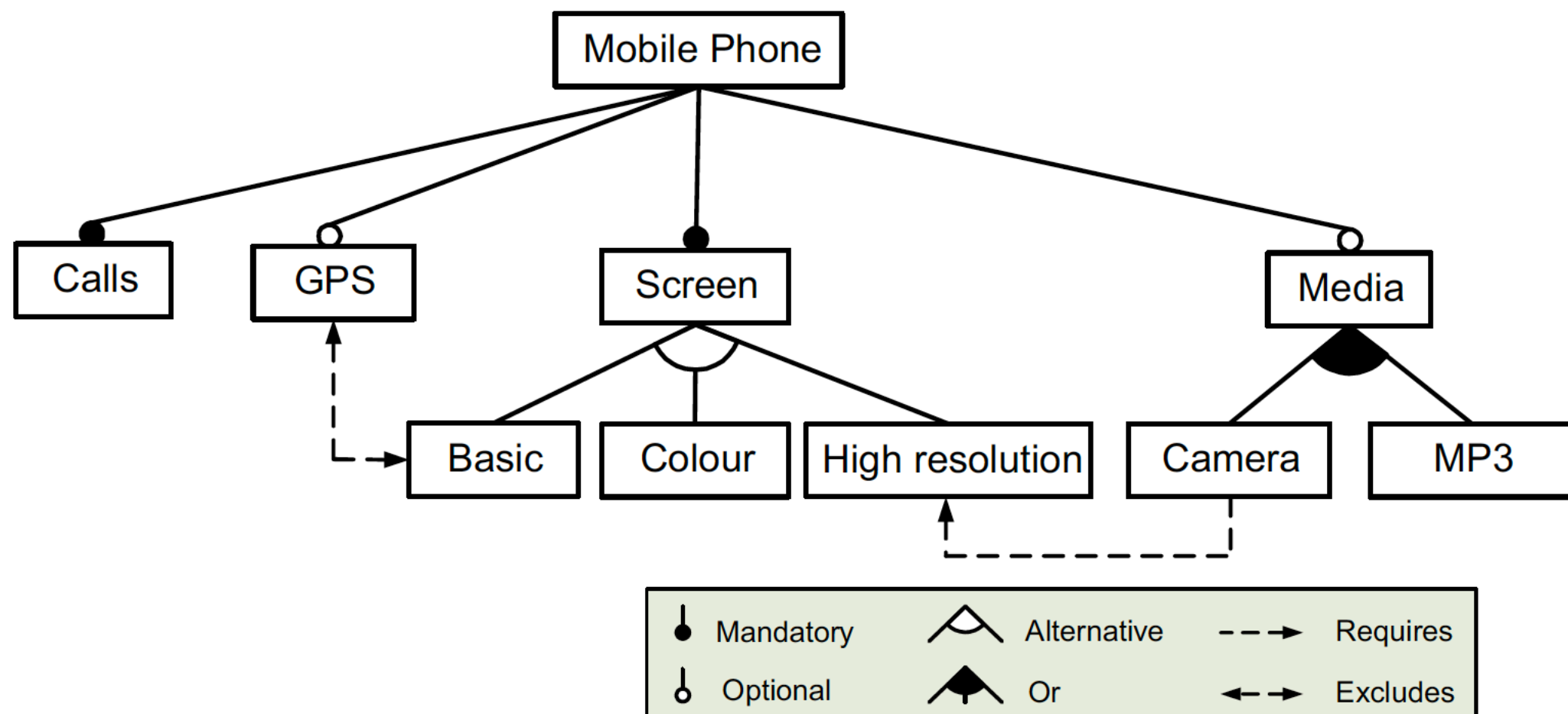
t_2

Horty, John F., Richmond H. Thomason, and David S. Touretzky. "A skeptical theory of inheritance in nonmonotonic semantic networks." *Artificial intelligence* 42, no. 2-3 (1990): 311-348.

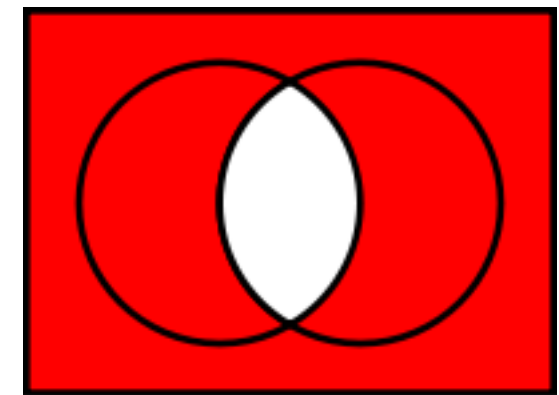
- Knowledge unification bring to intractable knowledge
- Explicit representation of conflicts in knowledge simplify both detection and resolution



- The explicit representation of dependencies between features and requirements is often adopted in Software Engineering



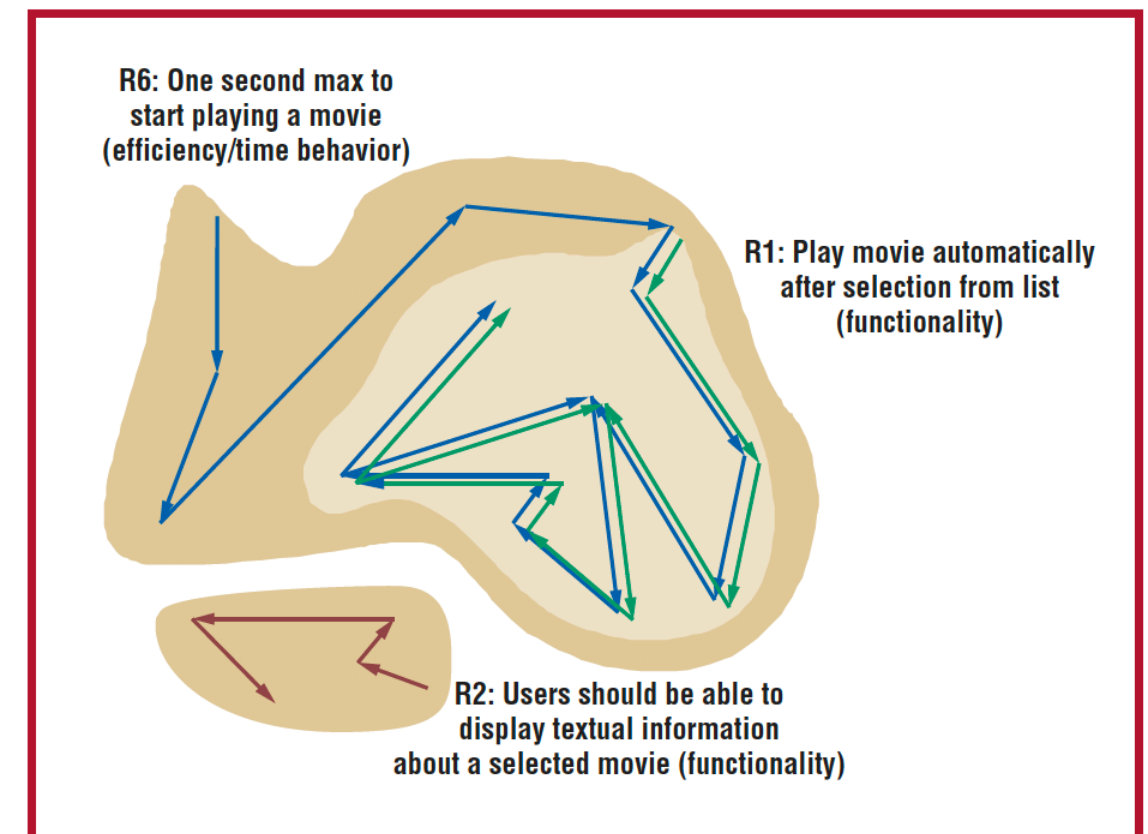
- In terms of propositional calculus a conflicting requirement can be interpreted as an **alternative denial**, the negation of the conjunction
 - $P \rightarrow \neg Q$ equivalent to $\neg P \vee \neg Q$
- When propositions are correctly gathered detection is easy
- To resolve conflicts priority and specificity are typical criteria



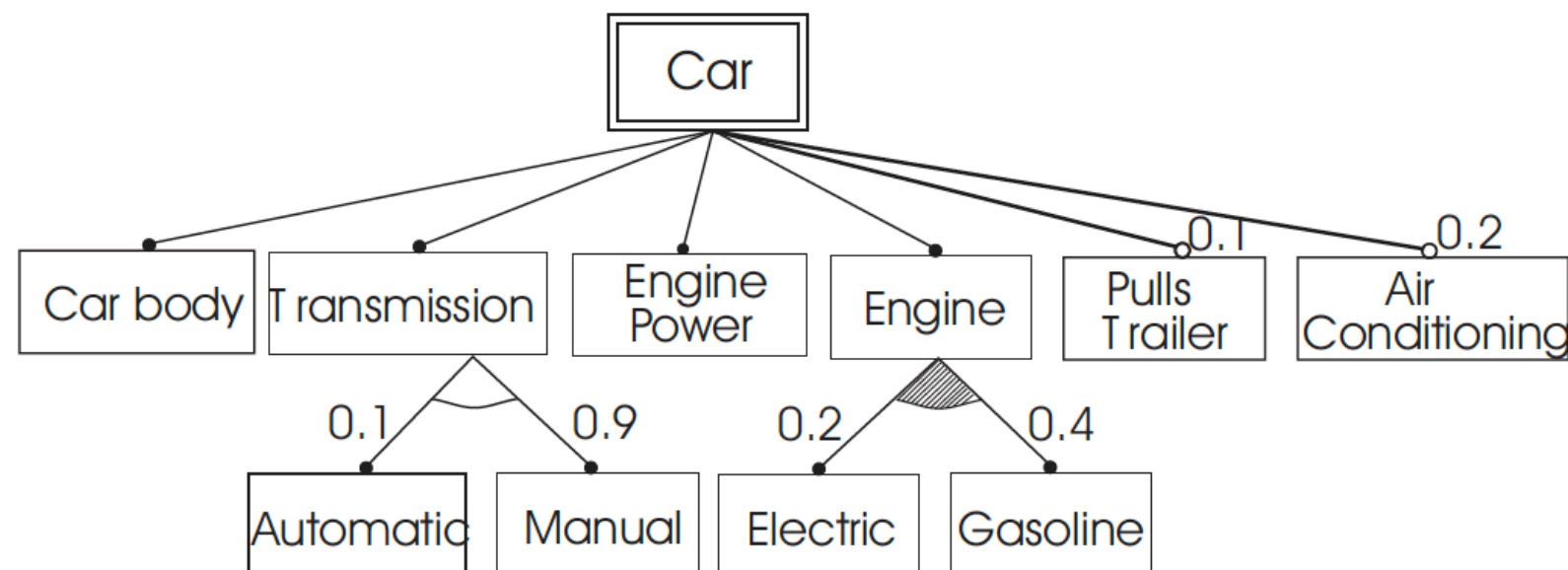
Model of potential conflict and cooperation*

Requirement attribute	Effect							
	Functionality	Efficiency	Usability	Reliability	Security	Recoverability	Accuracy	Maintainability
Functionality	+	–	+	–	–	0	0	–
Efficiency	0	+/-	+	–	–	0	–	–
Usability	+	+/-	+	+	0	+	+	0
Reliability	0	0	+	+	0	0	0	0
Security	0	–	–	+	+	0	0	0
Recoverability	0	–	+	+	0	+	0	0
Accuracy	0	–	+	0	0	0	+	0
Maintainability	0	0	0	+	+	0	0	+

Egyed, A., & Grunbacher, P. (2004). Identifying requirements conflicts and cooperation: How quality attributes and automated traceability can help. *IEEE software*, 21(6), 50-58.



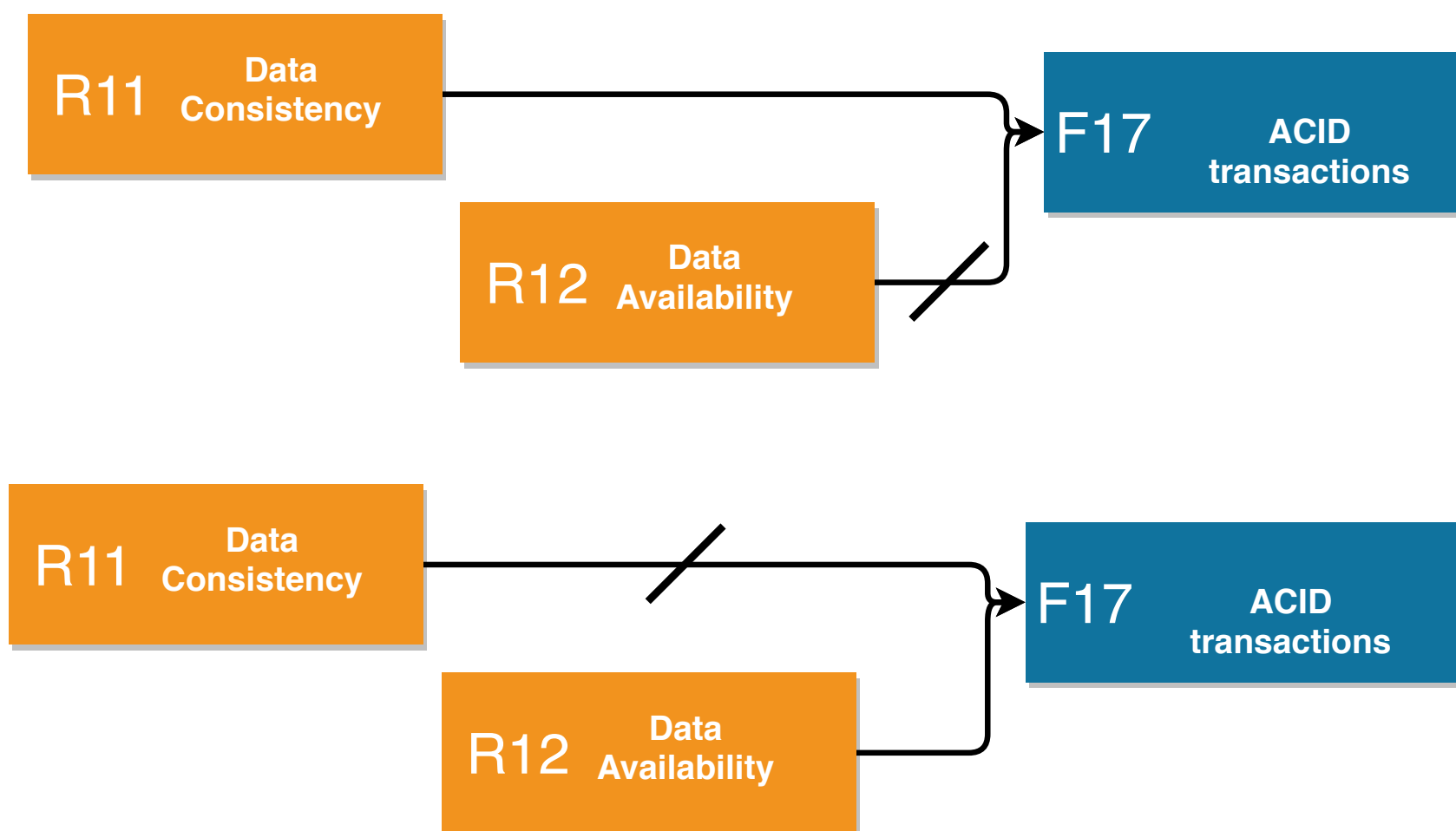
DEPENDENCIES



Robak, Silva, and Andrzej Pieczynski. "Employing fuzzy logic in feature diagrams to model variability in software product-lines." In 10th IEEE International Conference and Workshop on the Engineering of Computer-Based Systems, 2003. Proceedings., pp. 305-311. IEEE, 2003.

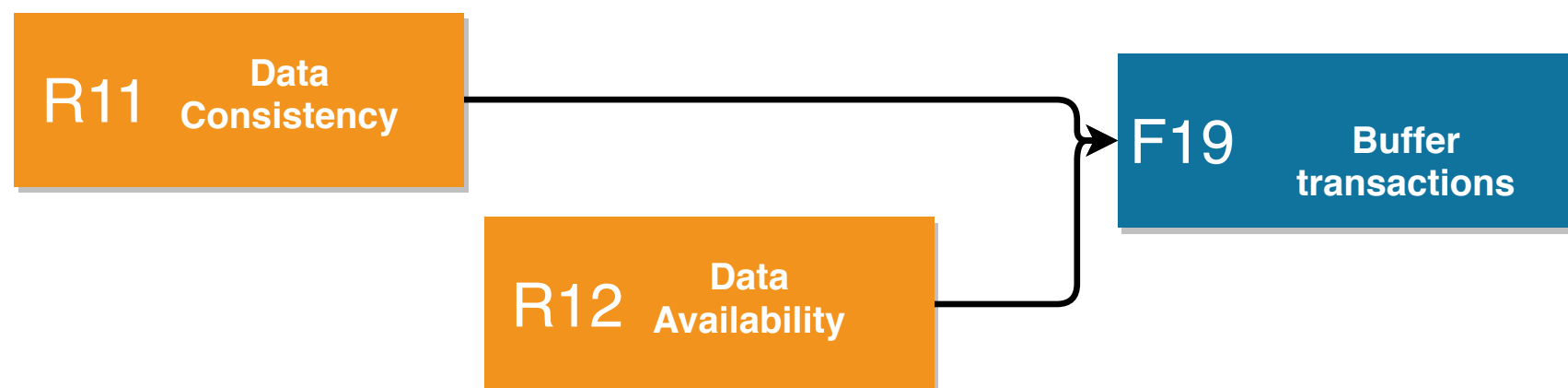
EXAMPLE 7 – APPARENT CONFLICT

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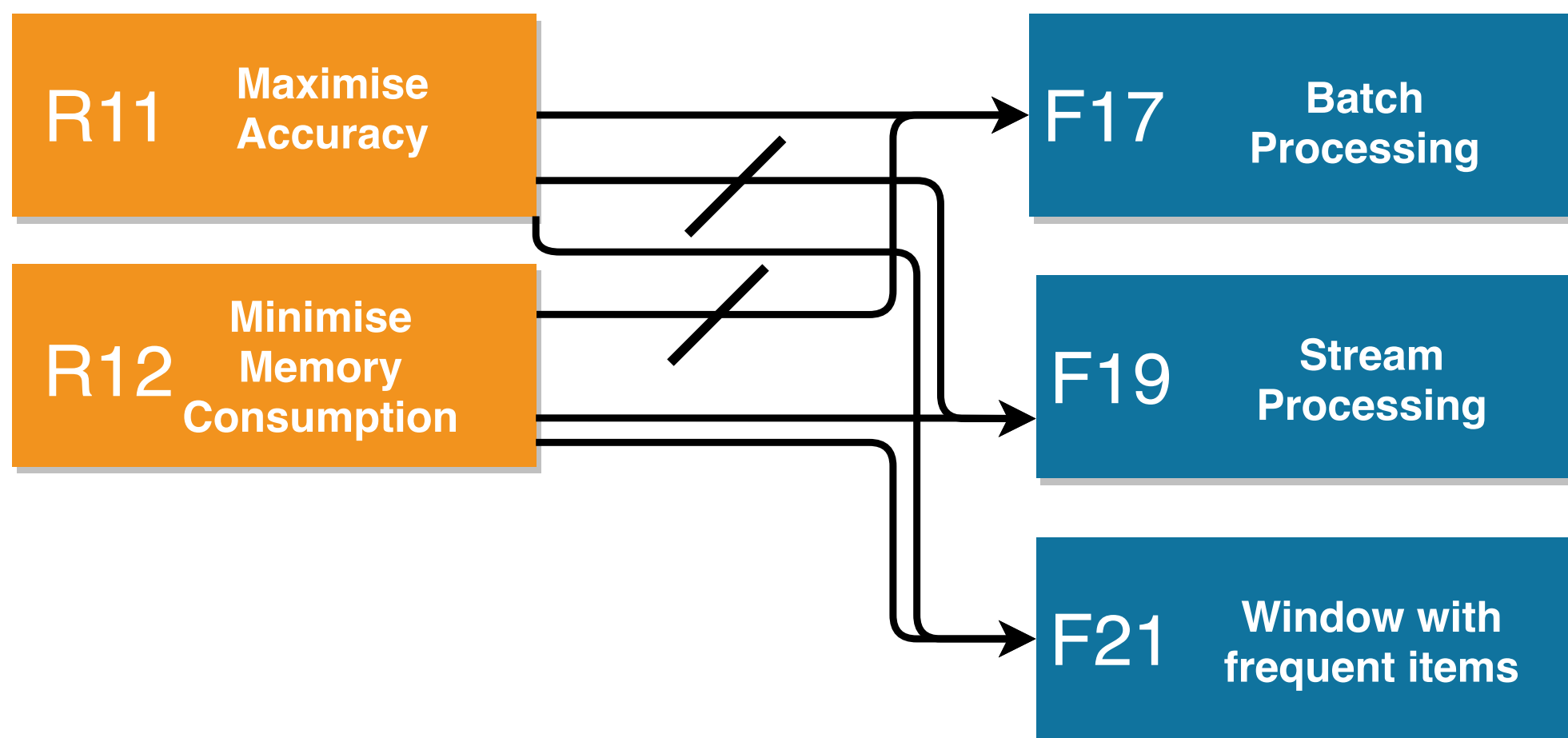
EXAMPLE 7 – APPARENT CONFLICT: TRADE-OFF

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EXAMPLE 8 – APPARENT CONFLICT: TRADE-OFF

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D – LESSON LEARNED



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- Explicit representation of conflicts in knowledge simplify both detection and resolution
- The definition of conflicts using a priori knowledge is not always feasible
- Partial conflicts requires a detailed understanding of the correlation between feature values

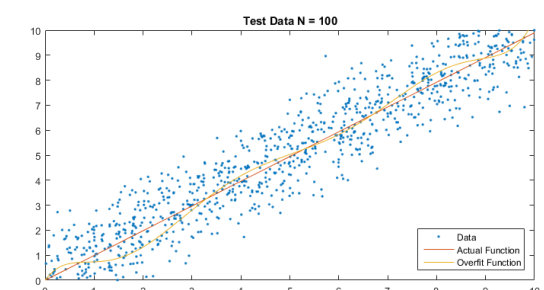
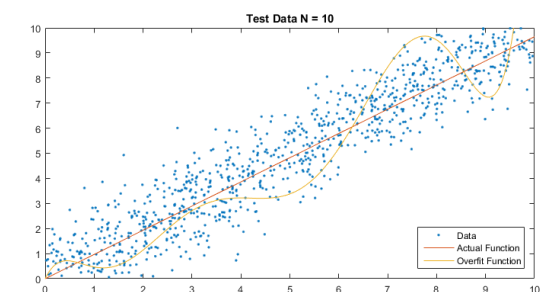


DOMAIN DEPENDENCIES

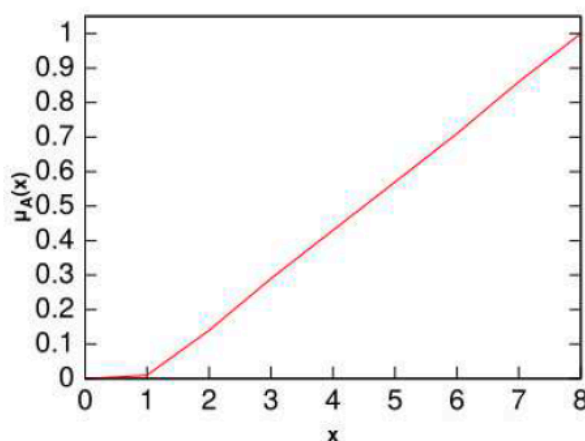


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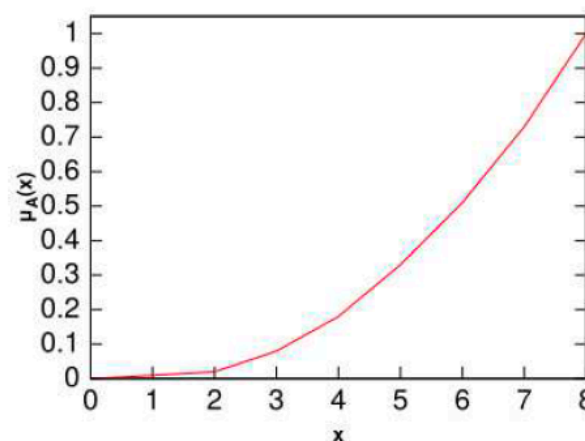
- Representing the dependency between two domains of values permits to constraints their values
 - Different scales can apply
 - Different distributions shapes can apply
 - Statistical generalisation may not apply
 - Non stationary behaviour may apply



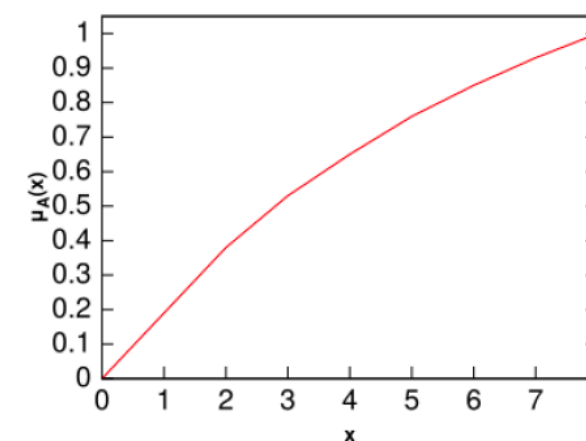
- Fuzzy membership functions can be used to map a domain to a normalised degree of satisfaction and to put it in relationship with other domains



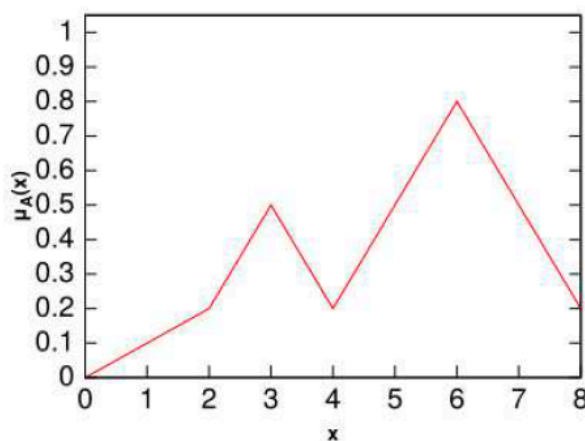
(a)



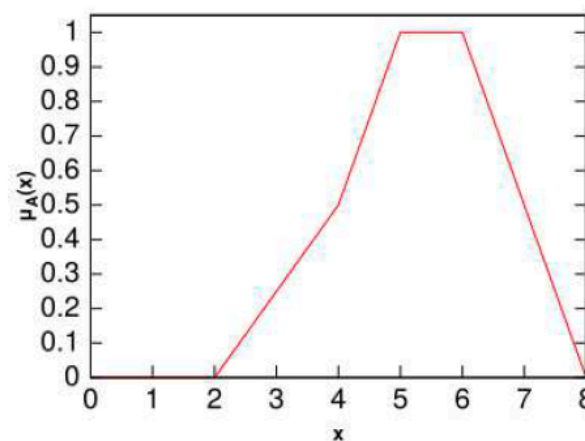
(b)



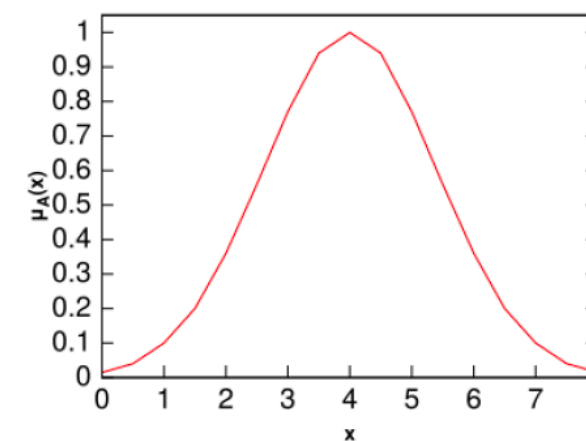
(c)



(d)



(e)



(f)

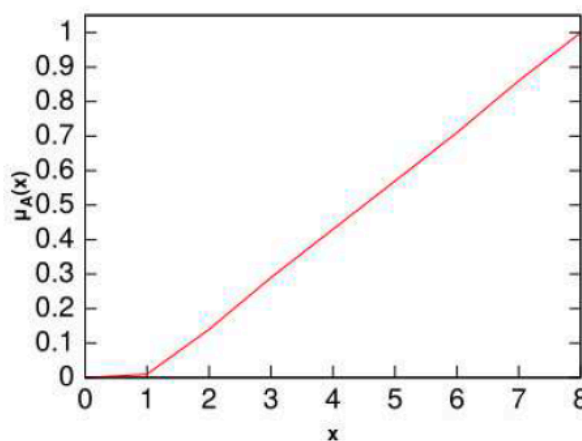
Ardagna, Claudio Agostino, Valerio Bellandi, Michele Bezzi, Paolo Ceravolo, Ernesto Damiani, and Cedric Hebert. "Model-based big data analytics-as-a-service: take big data to the next level." *IEEE Transactions on Services Computing* (2018).

- Fuzzy membership functions can be used to map a domain to a normalised degree of satisfaction and to put it in relationship with other domains
- Using fuzzy logic the same truth function used with predicate logic can be used with domain dependency
 - $P \rightarrow \neg Q$ equivalent to $\neg P \vee \neg Q$
 - $\neg P$ equivalent to $1-P$
 - $P \vee Q$ equivalent to $\max(P, Q)$
 - $P \rightarrow \neg Q$ equivalent to $\max(1-P, 1-Q)$

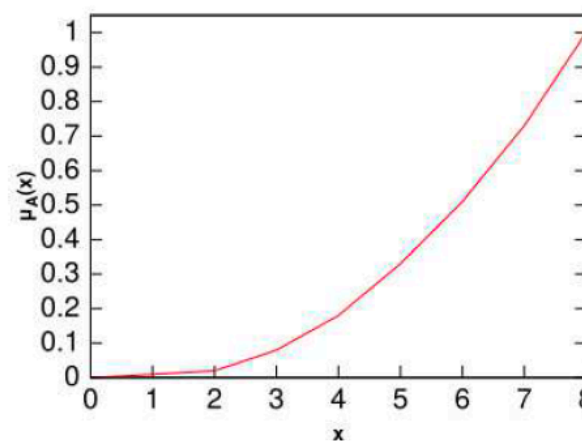
- Given two variables A and B we can define the dependency between their domains using the following steps:
 1. Map domain D_A to a $[0, 1]$ interval using a fuzzy MF
 2. Map domain D_B to a $[0, 1]$ interval using a fuzzy MF
 3. The truth function $\max(1-P, 1-Q)$ tell us if two values of the valuables are conflicting or not
- Example: is μ_B conflicting with μ_A given that μ_A is 0.5 and μ_B is 0.6?

$$\max(0.5, 0.4) = 0.5 \leq \mu_B$$

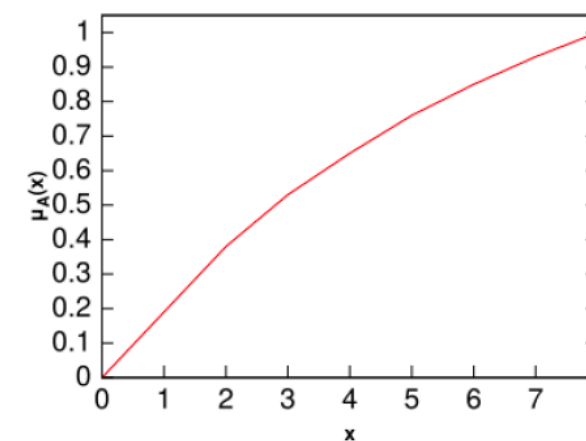
- A domain is view as the interval of values (either discrete or continue) contained in a lower limit and an upper limit
- $D: [a, b]$



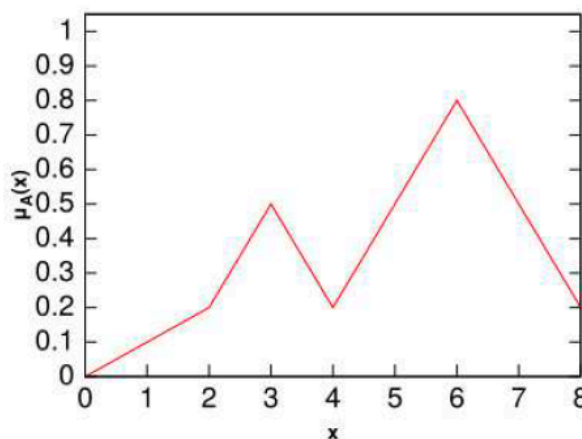
(a)



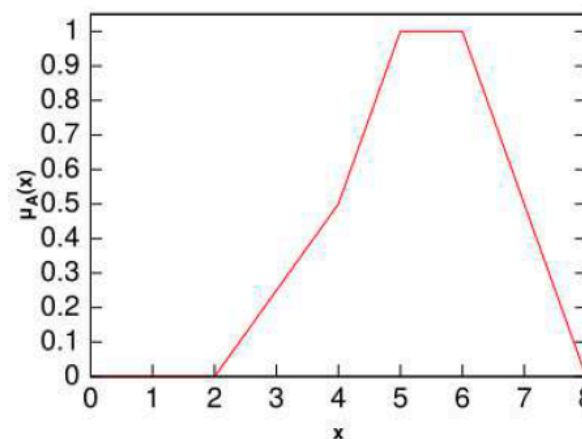
(b)



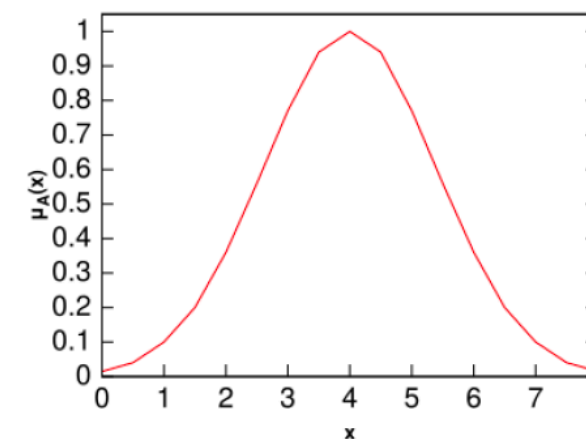
(c)



(d)



(e)



(f)

Ardagna, Claudio Agostino, Valerio Bellandi, Michele Bezzi, Paolo Ceravolo, Ernesto Damiani, and Cedric Hebert. "Model-based big data analytics-as-a-service: take big data to the next level." *IEEE Transactions on Services Computing* (2018).

- A domain is view as the interval of values (either discrete or continue) contained in a lower limit and an upper limit
D: [a, b]

a) linear and monotonic

$$\mu_A(v \in \mathcal{D}) = \frac{a - v}{b - a}.$$

b) quadratic monotonic

$$\mu_A(v \in \mathcal{D}) = \frac{(a - v)^2}{(b - a)^2}$$

d) triangular non-monotonic

$$\mu_A(v \in D) = \begin{cases} 0, & v \leq a \\ \frac{v-a}{m-a}, & a \leq v \leq m \\ \frac{b-v}{b-m}, & m \leq v \leq b \\ 1, & v \geq b \end{cases}$$

f) gaussian non-monotonic

$$\mu_A(v \in \mathcal{D}) = e^{-\frac{v-m}{2k^2}}$$

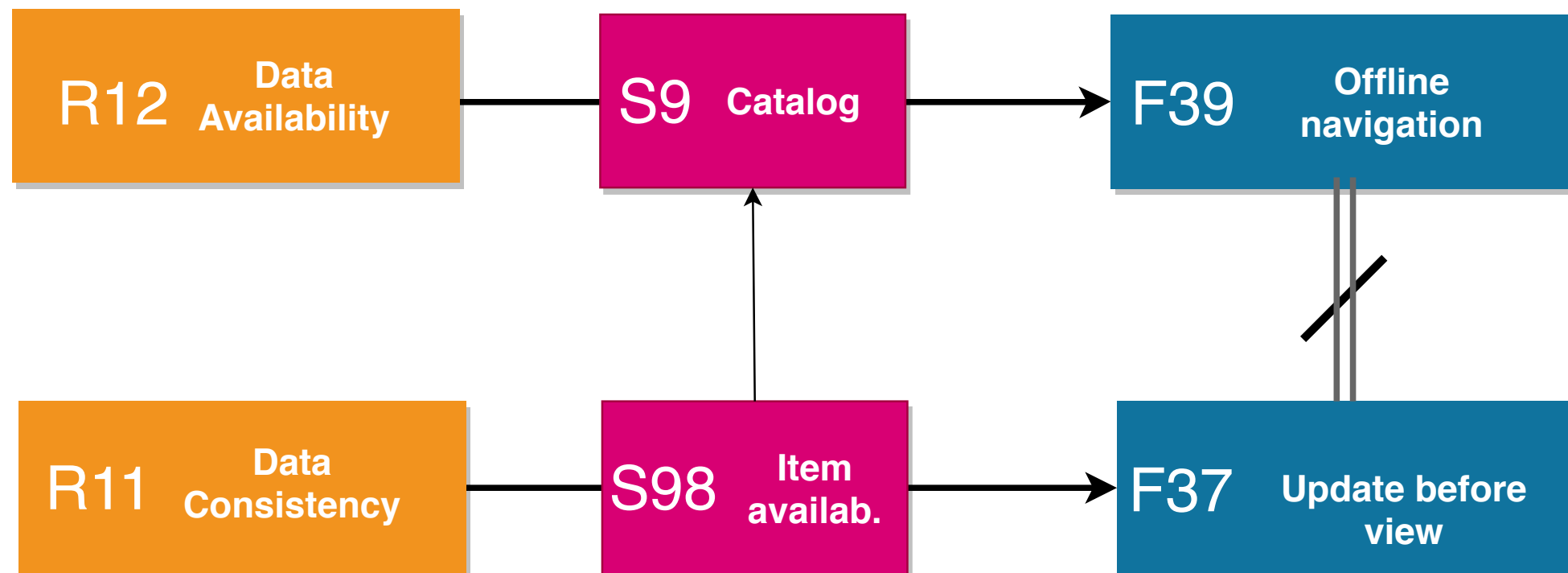
- A key point to represent dependencies between two variables' domains is studying their correlation
 - Regression analysis
 - Nonparametric regression
 - Multiple regression analysis
 - Random forest
 - Neural networks

Marquez, Leorey, Tim Hill, Reginald Worthley, and William Remus. "Neural network models as an alternative to regression." In Proceedings of the Twenty-Fourth Annual Hawaii International Conference on System Sciences, vol. 4, pp. 129-135. IEEE, 1991.

EXAMPLE 4 – CONFLICTING DEPENDENCIES FOR KR



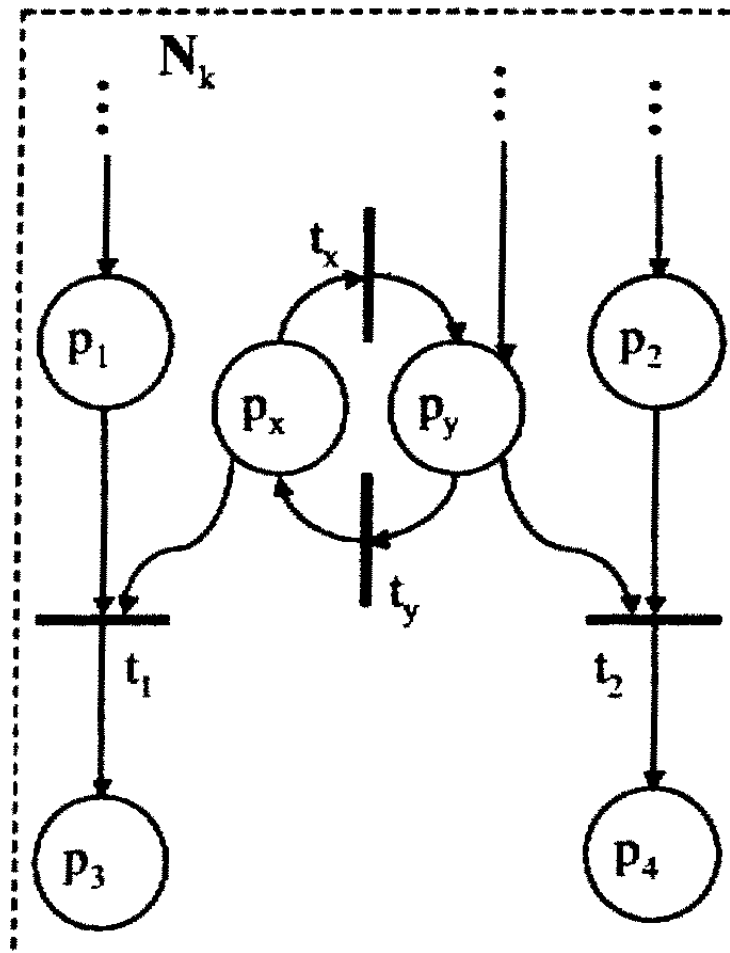
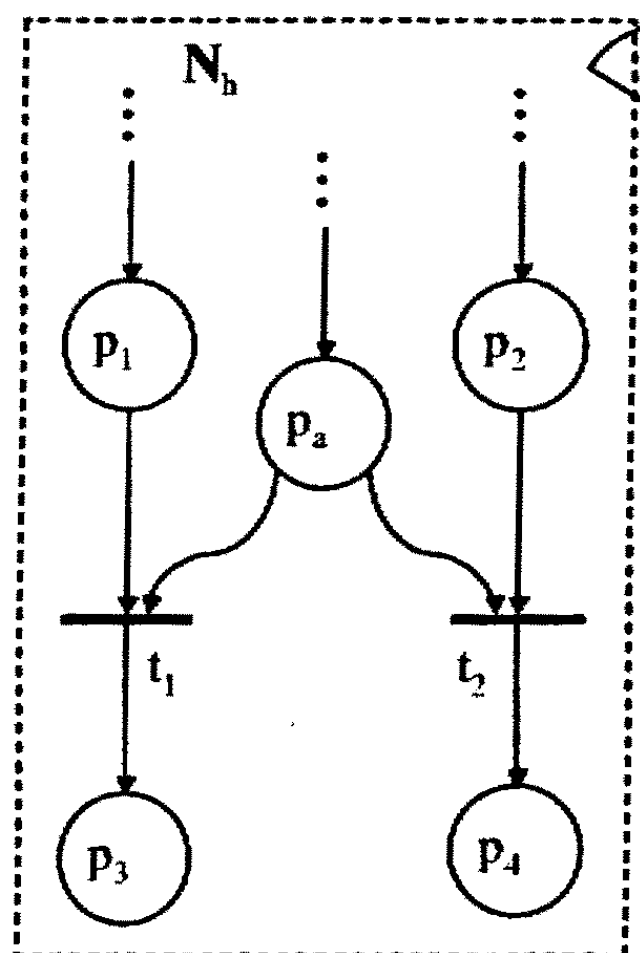
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- Explicit representation of conflicts in knowledge simplify both detection and resolution
- The definition of conflicts using a priori knowledge is not always feasible
- Partial conflicts requires a detailed understanding of the correlation between feature values
- In dynamic systems conflicts may arise from concurrency

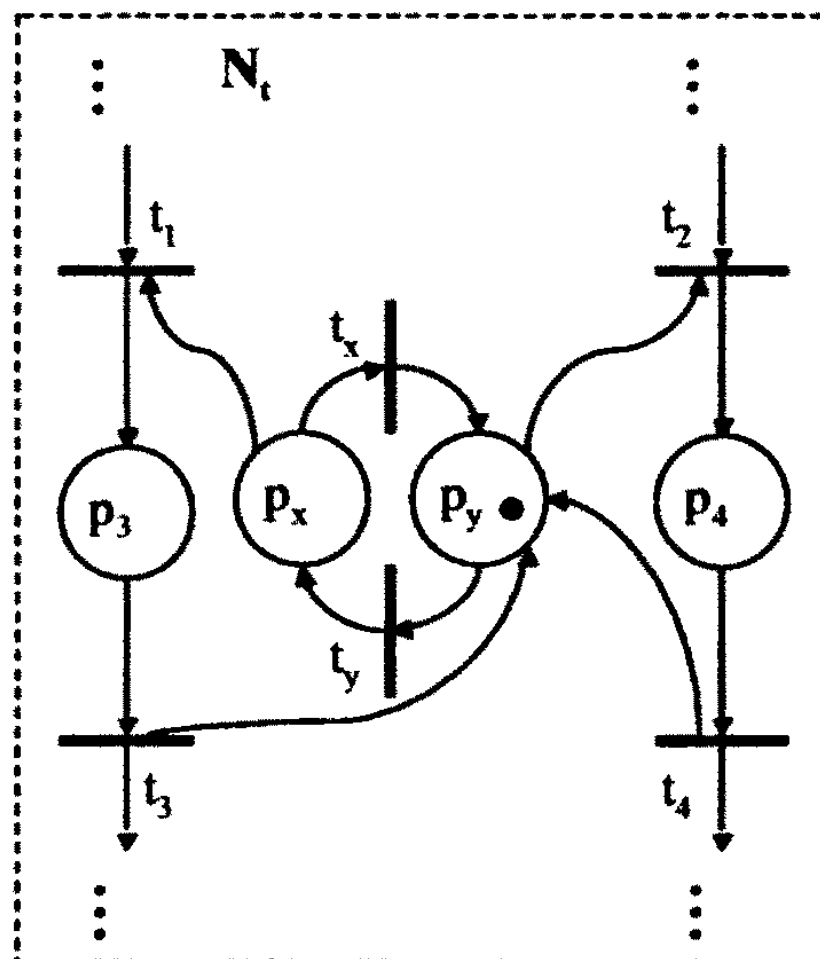
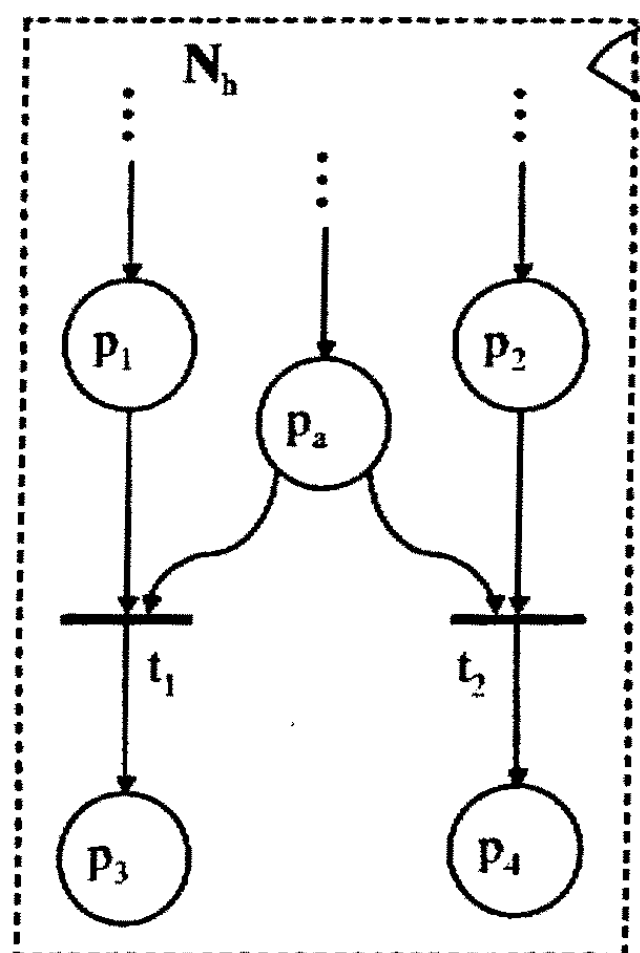


- If a resource can be used by mutually exclusive access only, conflicts can be addressed by resource sharing
- Concurrency must be represented and handled



Gomes, Luis. "On conflict resolution in Petri nets models through model structuring and composition." In *INDIN'05. 2005 3rd IEEE International Conference on Industrial Informatics*, 2005., pp. 489-494. IEEE, 2005.

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Gomes, Luis. "On conflict resolution in Petri nets models through model structuring and composition." In *INDIN'05. 2005 3rd IEEE International Conference on Industrial Informatics*, 2005., pp. 489-494. IEEE, 2005.

- Explicit representation of conflicts in knowledge simplify both detection and resolution
- The definition of conflicts using a priori knowledge is not always feasible
- Partial conflicts requires a detailed understanding of the correlation between feature values
- In dynamic systems conflicts may arise from concurrency
- Resources may be consumed after a number of accesses



IN SUMMARY

	Knowledge Representation	Dependencies		Dynamic Systems
Detect	Alignment	Manual Tagging	Detect Correlation	Discovery
	*	**	**	**
Resolve	Partitioning	Priorities Specificity	Domain Mapping	Reachability
Predicates	**	**	*	*
Dis & Cont Variables	*	*	**	*
Concurrency				**
Factual Impossible				

MOTIVATIONS



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	Knowledge Representation		Dependencies	Dynamic Systems
Detect	Alignment	Manual Tagging	Detect Correlation	Discovery
	More inconsistencies that needed	May implies incompleteness	Data may be incom. Concept Drift applies	Multiple models
Resolve	Partitioning	Priorities Specificity	Domain Mapping	Reachability
<i>Predicates</i>	Well with priorities	Well with priorities and specificity	Overcomplicated	Overcomplicated
<i>Dis & Cont Variables</i>	You have to accept possibility	You have to accept possibility	Well if the mapping is appropriate	Limited to what can be expressed by tokens
<i>Concurrency</i>				Well
<i>Factual Impossible</i>	Nope	Nope	Nope	Nope



Thank You

HANDLING CONFLICTING REQUIREMENTS

*Paolo Ceravolo - SESAR Lab - Dipartimento di Informatica
Università degli Studi di Milano*