

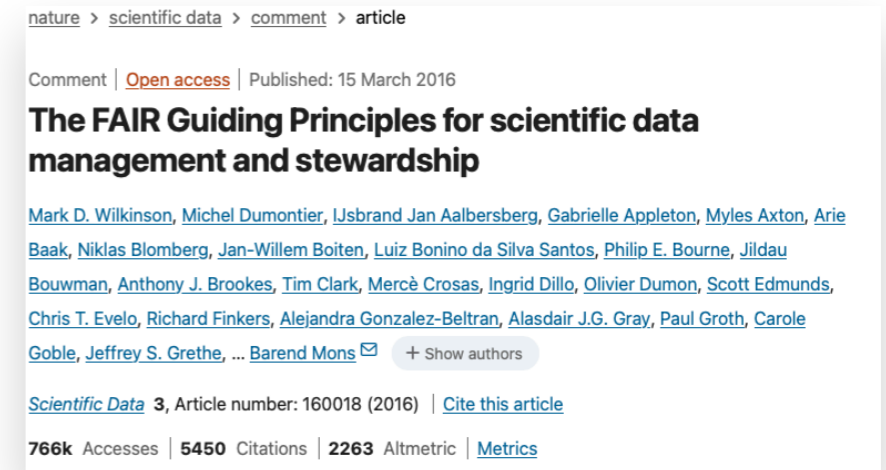
A generic framework to better understand and compare FAIRness measures

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FAIR principles \neq technical specifications

- ▶ Largely **adopted** by research funders and organizations
- ▶ Key for more open & reproducible sciences

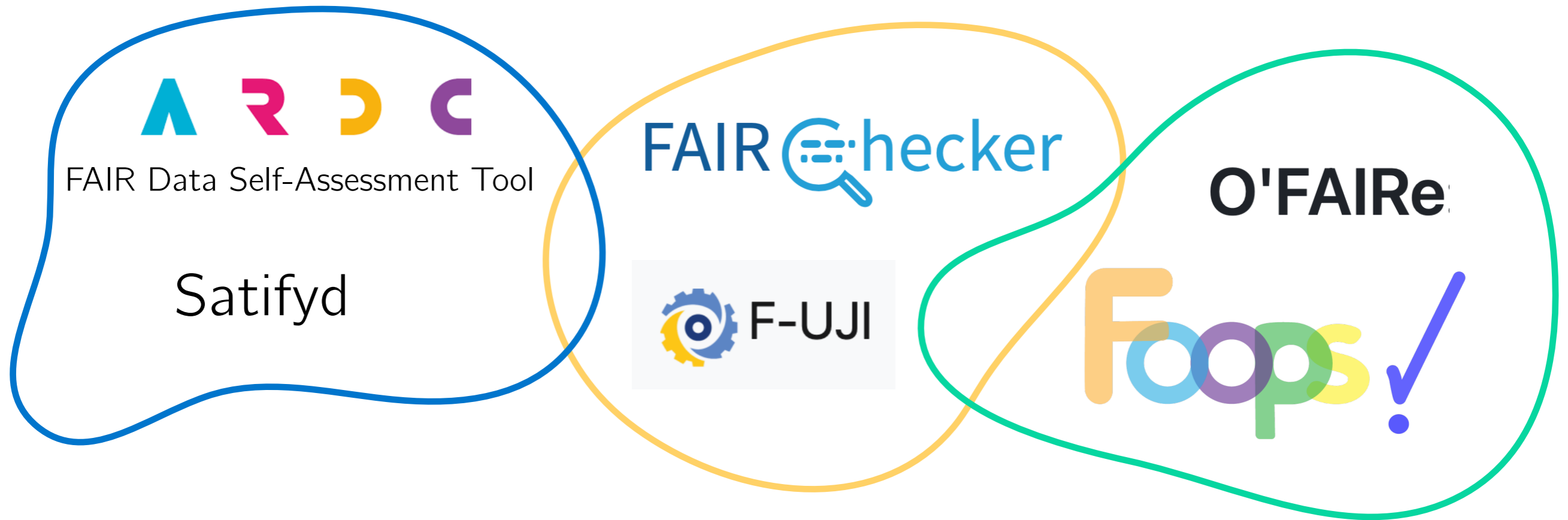


However

- ▶ Principles, non-technical guidelines
- ▶ can be **interpreted** differently by specific communities
- ▶ **many** ways of **implementing** them

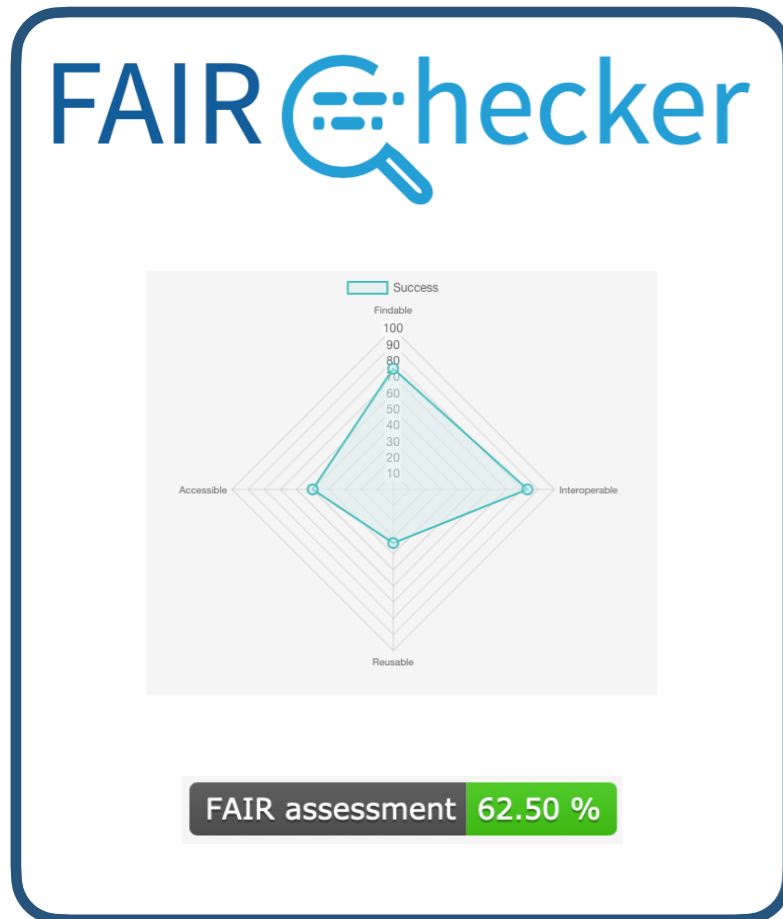


Multiple FAIR assessment tools

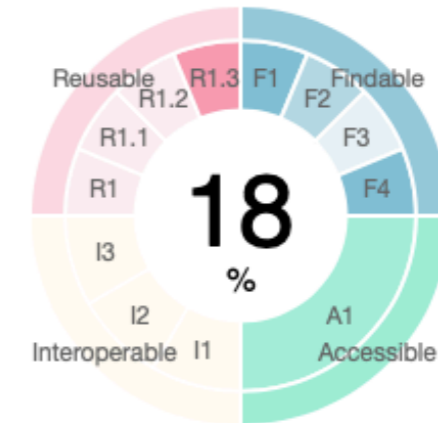


- ▶ Check lists / questionnaires
- ▶ Automated tools
- ▶ Some tools are "community oriented"
- ▶ Some tools are "technologically opinionated"

How FAIR is a bioinformatics software ?



<https://bio.tools/seurat>



	Score earned:	Fair level:
Findable:	2.5 of 7	○ initial
Accessible:	1 of 3	○ initial
Interoperable:	0 of 4	○ incomplete
Reusable:	1 of 10	○ initial

- ▶ Why are results so different?
- ▶ Which tool should I use?
- ▶ What's behind the scene?

One digital object ...

... but different FAIR results 🤔

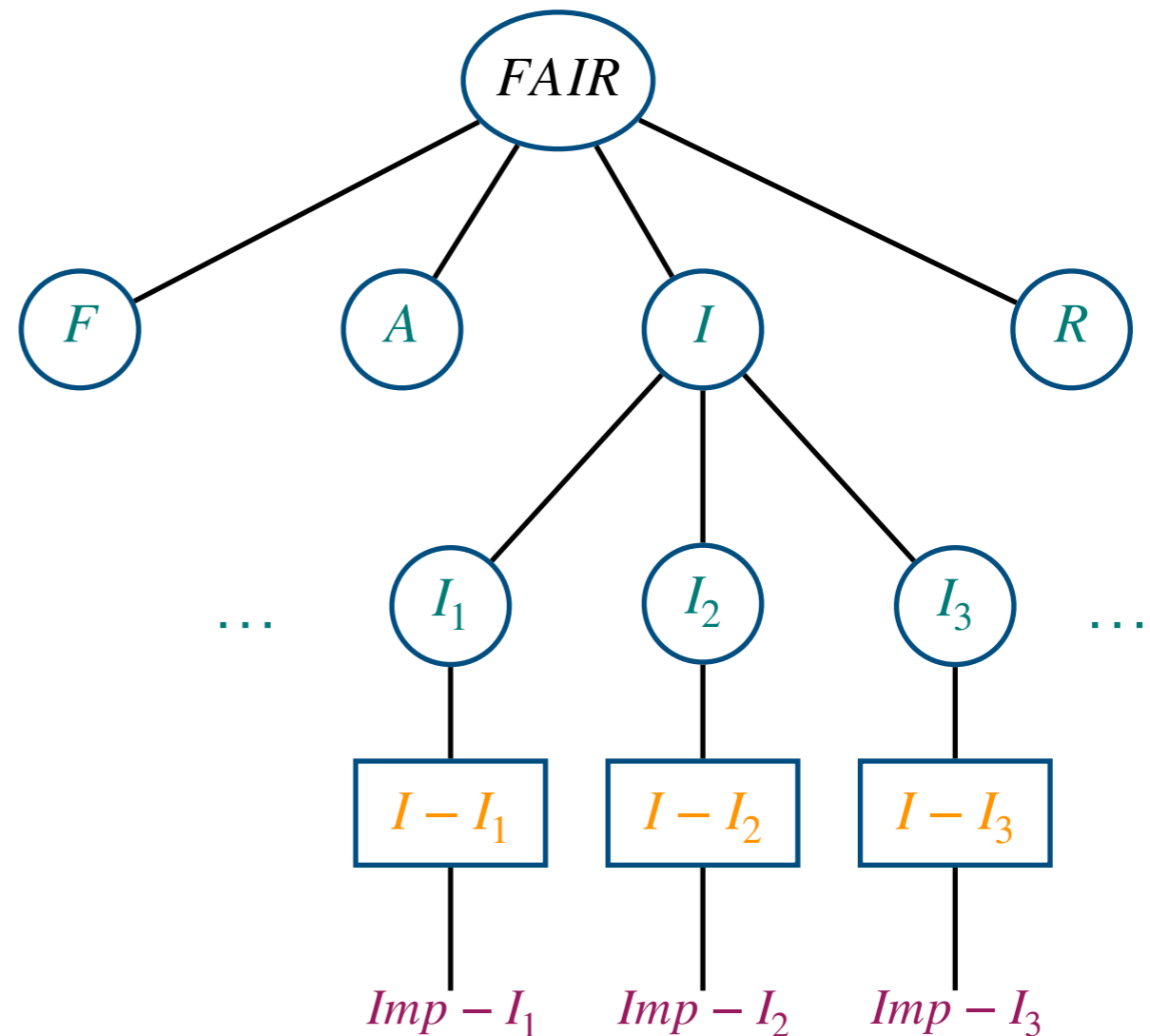
① We need a computable framework for *modeling* FAIR measures

Objectives

- (i) **specify measures** in a uniform and computable **model**
- (ii) propose specific **quantities** to **analyse and compare** FAIR assessment approaches

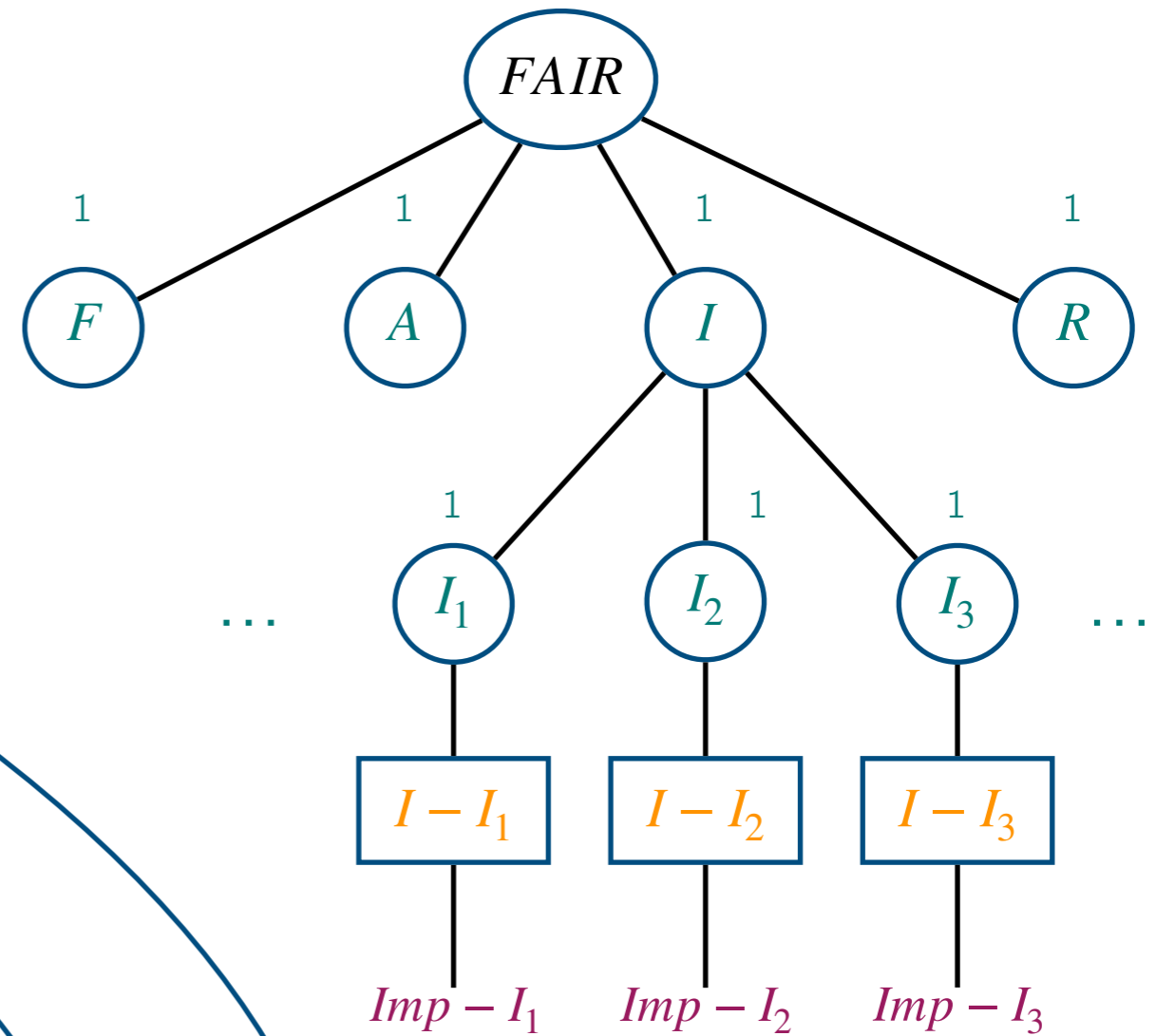
Typical structure for FAIR assessment methods

- ▶ Principles P : the FAIR principles and their sub-principles
- ▶ Indicators $I(\mathcal{M})$: the specification of principles, (i.e. what has to be verified)
- ▶ Implementations $Imp(\mathcal{M})$: the implementations of the principles



$$\mathcal{M} = \left(\overbrace{(V, E, FAIR)}^{\text{structure}}, \underbrace{\diamond, w, v_{max}, D}_{\text{score}} \right)$$

Score computation



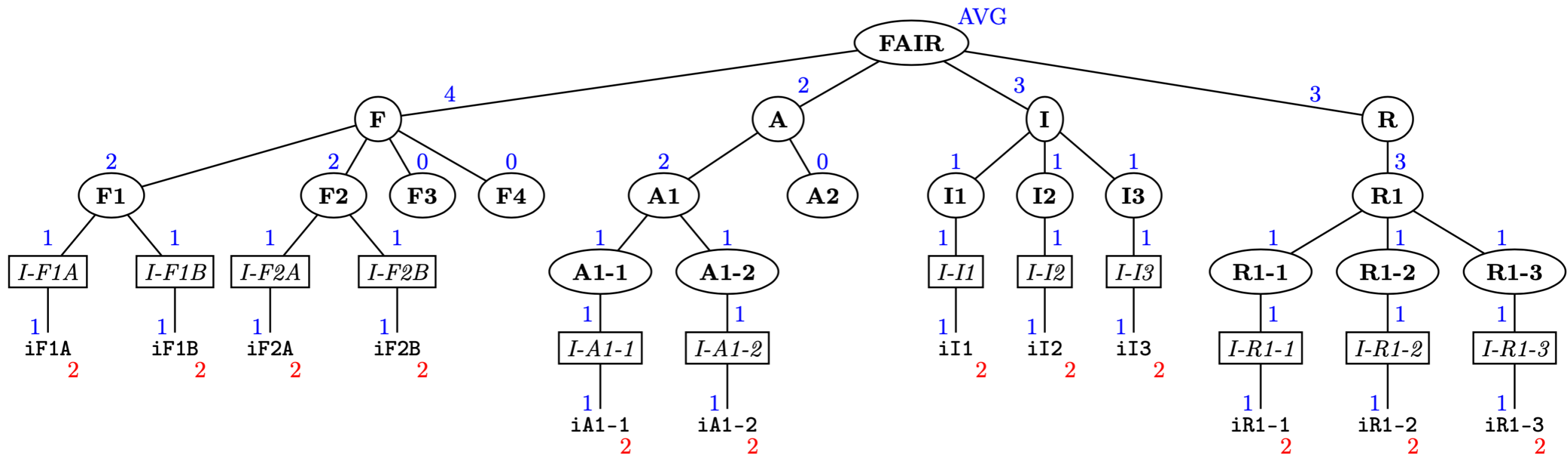
Max function: $v_{max}(i)$ returns the maximum reachable score for an indicator or an implementation

Weighting function: $w(n)$ returns the weight of the node n compared to its siblings

Aggregation function : weighted sum or weighted average to collect scores

$$\mathcal{M} = (\overbrace{(V, E, FAIR)}^{\text{structure}}, \underbrace{\diamond, w, v_{max}, D}_{\text{score}})$$

Model instantiation with FAIR-Checker

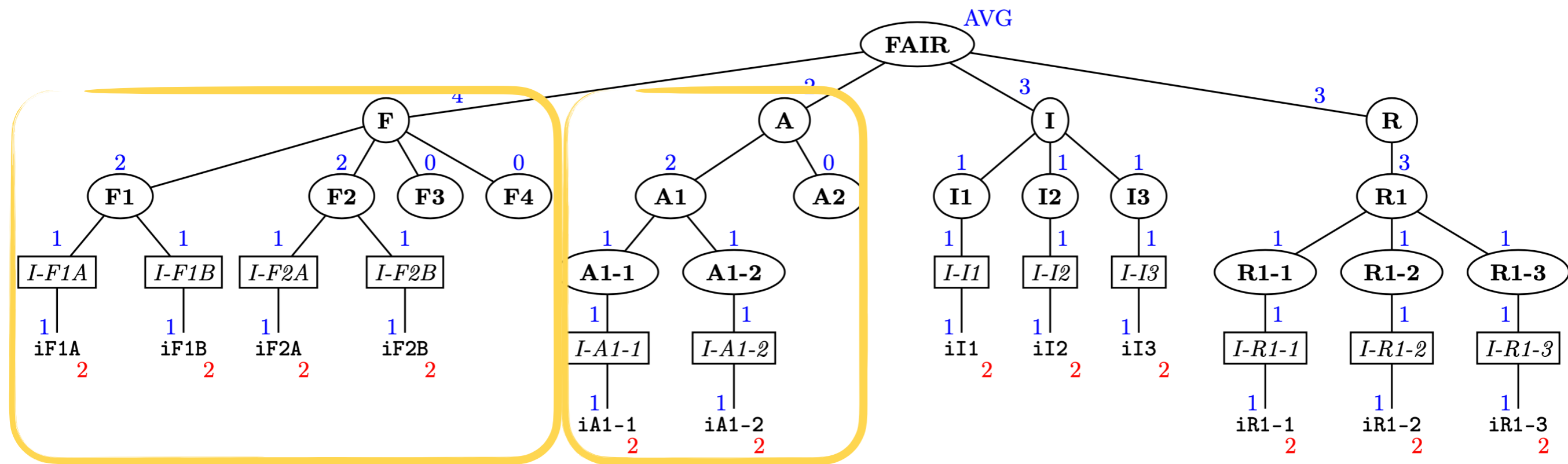


- ▶ The tool fits the tree-based structure $\{P, I, Imp\}$
- ▶ needs the maximum reachable score per implementation $\rightarrow 2$
- ▶ needs an aggregation function $\rightarrow AVG$ (normalized scores)
- ▶ needs cumulative weights for each principle $\rightarrow w \in [0, 4]$

② We need quantities to characterize and compare FAIR measures

Quantifying the *granularity* of a measure

Idea: a global metric computed as the mean number of indicators per principles with at least one indicator. *If we have one principle with many implementations → important granularity.*

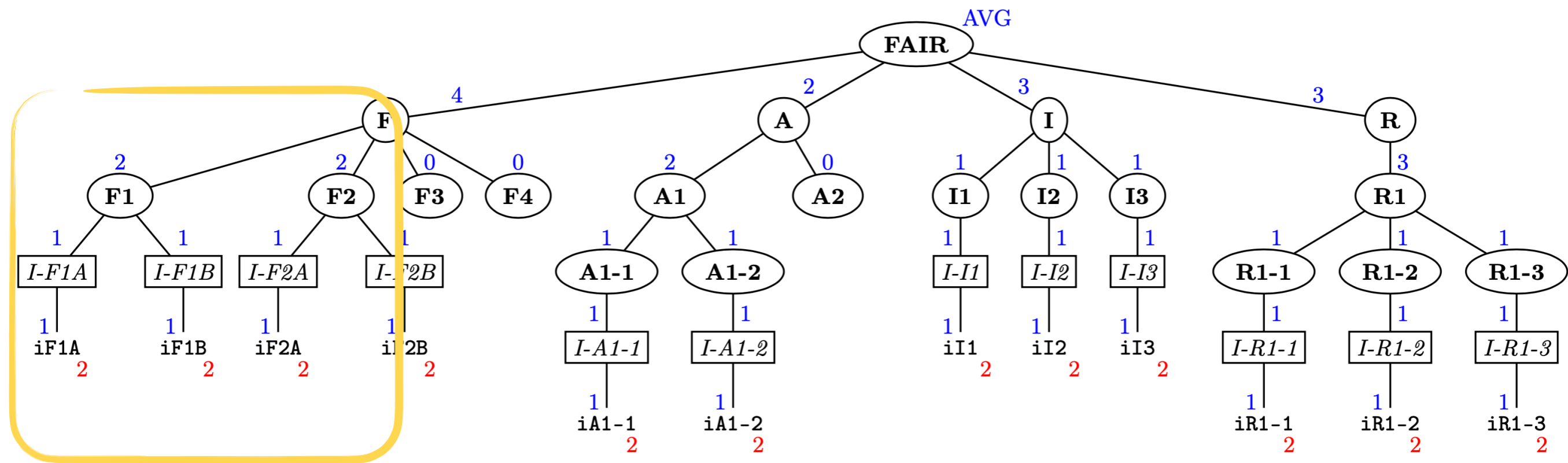


F has 4 indicators, A has only 2 → principles have not the same granularity

$$\text{granularity}(F) = 4/2 = 2 ; \text{granularity}(\text{FAIR}) = 12/10 = 1.2$$

Quantifying the *coverage* of a measure

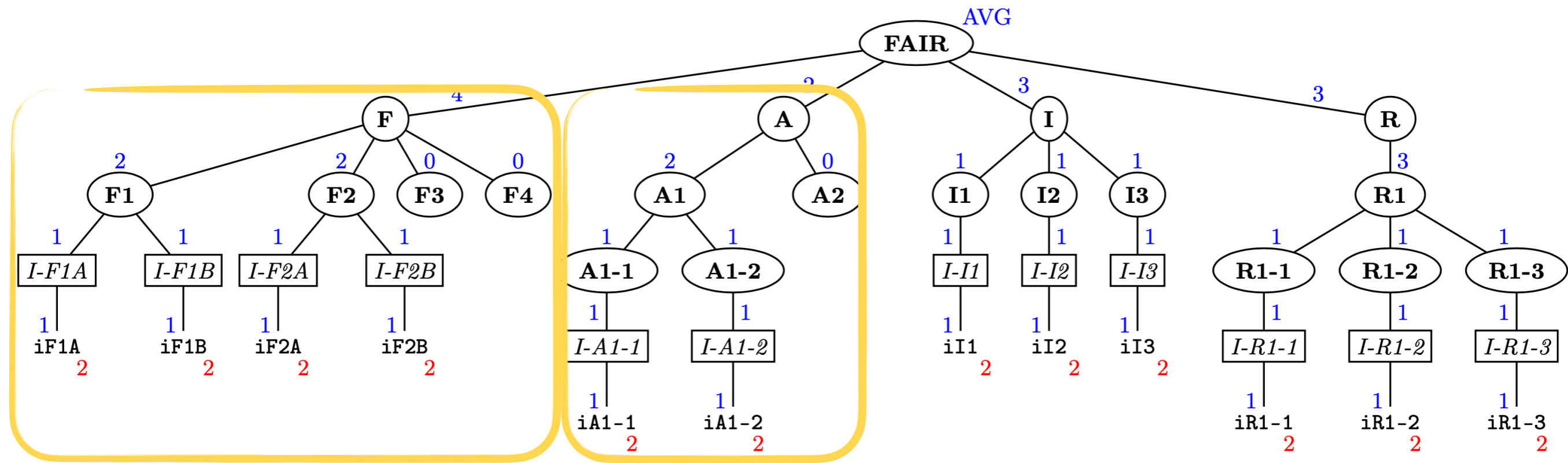
Idea: a principle is considered as covered if it exists at least one implementation for the principle or, its sub-principles.



F3 & F4 are never evaluated \rightarrow F has 50% coverage

Quantifying the *impact* of principles

Idea: the impact of a principle corresponds to the global score obtained when all its underlying implementations are successful (without considering other principles).



$$\text{imp}(A) = 2 \cdot 2 / 12 \cdot 2 = 1/6$$

$$\text{imp}(F) = 4 \cdot 2 / 12 \cdot 2 = 2/6$$

→ F contributes 2 times more compared to A in the global FAIR score

Experimental results

Experimental setup

- ▶ Evaluated tools

automated: F-UJI, FAIR Evaluator, FOOPS!, O'FAIRe, FAIR-Checker,
questionnaires: ARDC, SATiFYD

- ▶ 10 selected web resources for FAIR assessments, covering

- datasets descriptions
- ontologies
- online courses
- bioinformatics software
- RDF files

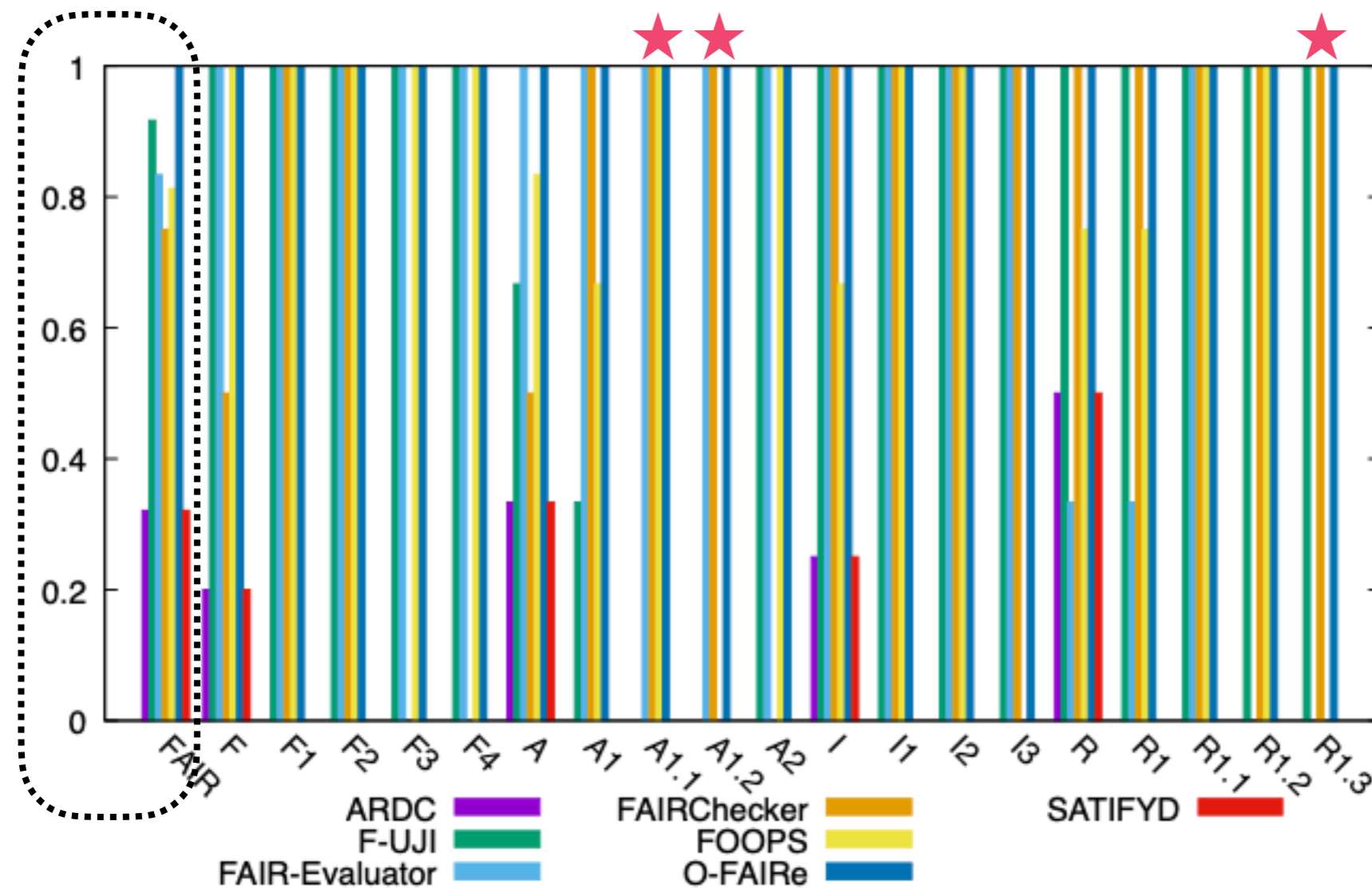
Do engines reach consensus on FAIR assessment ?

Resource	F-UJI (%)	FAIR-Checker (%)	Std dev
Dataset (PANGAEA) [31]	91	91.70	0.49
Gene Ontology (OLS) [21]	18	16.70	0.92
Dataset (Harvard Dataverse) [23]	75	79.20	2.97
Dataset (Kaggle) [26]	60	70.80	7.64
Online course (Moodle) [28]	4	16.70	8.98
Dataset (Governmental platform) [22]	52	70.80	13.29
Dataset (WHO) [39]	27	50.00	16.26
Training material (TeSS) [36]	39	70.80	22.49
Bioinformatics tool (bio.tools) [6]	18	54.20	25.60
Dataset (RDF metadata) [33]	43	87.50	31.47



- ▶ Higher scores for FAIR-Checker
- ▶ Last two entries: std. dev. > 25 % ?

Coverage rates

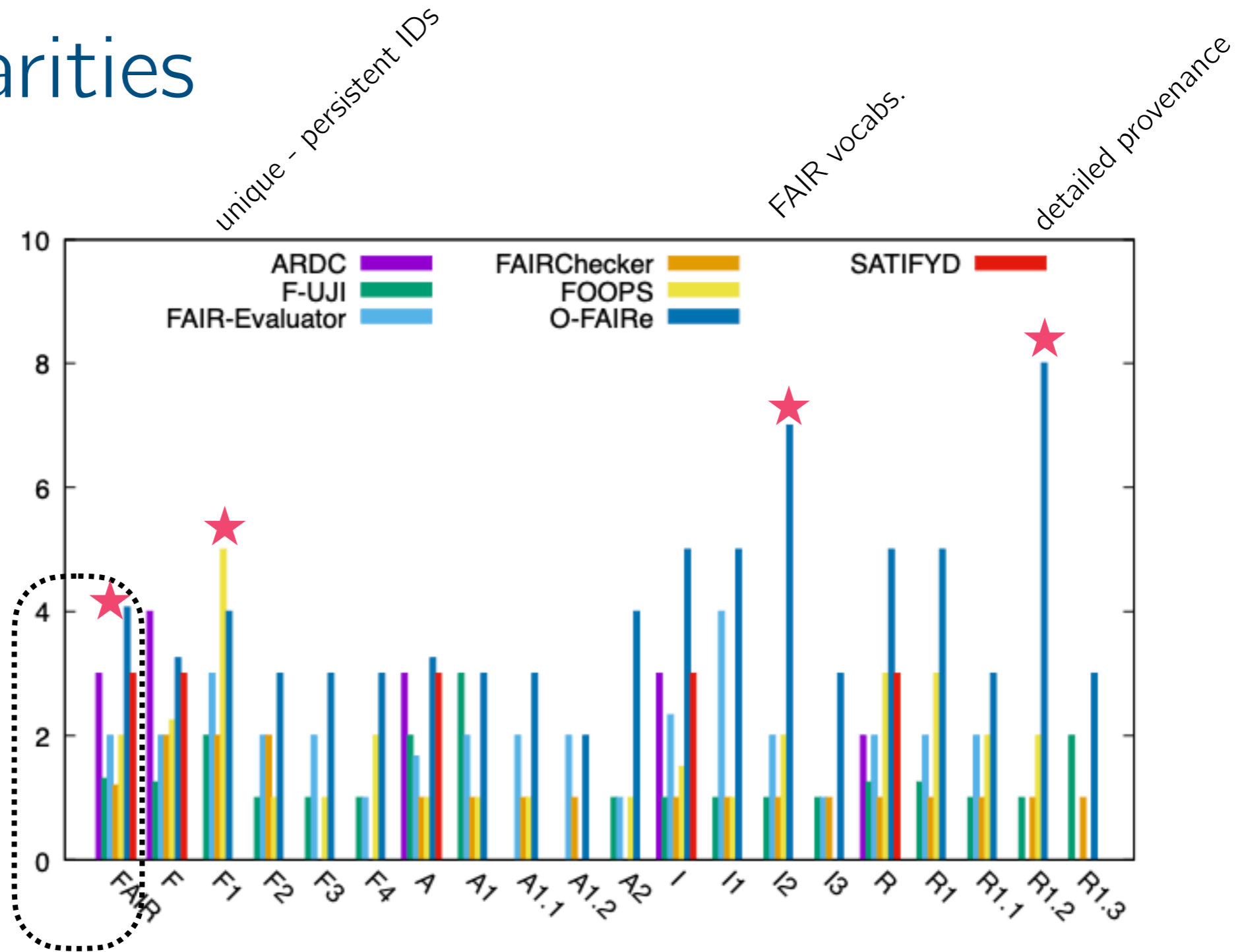


some principles (★) are not implemented
 → technical issues ?
 → interpretation ?

{O-FAIRe} → the only one with a full coverage (1)

Low coverage for questionnaires (no sub-principles)

Granularities

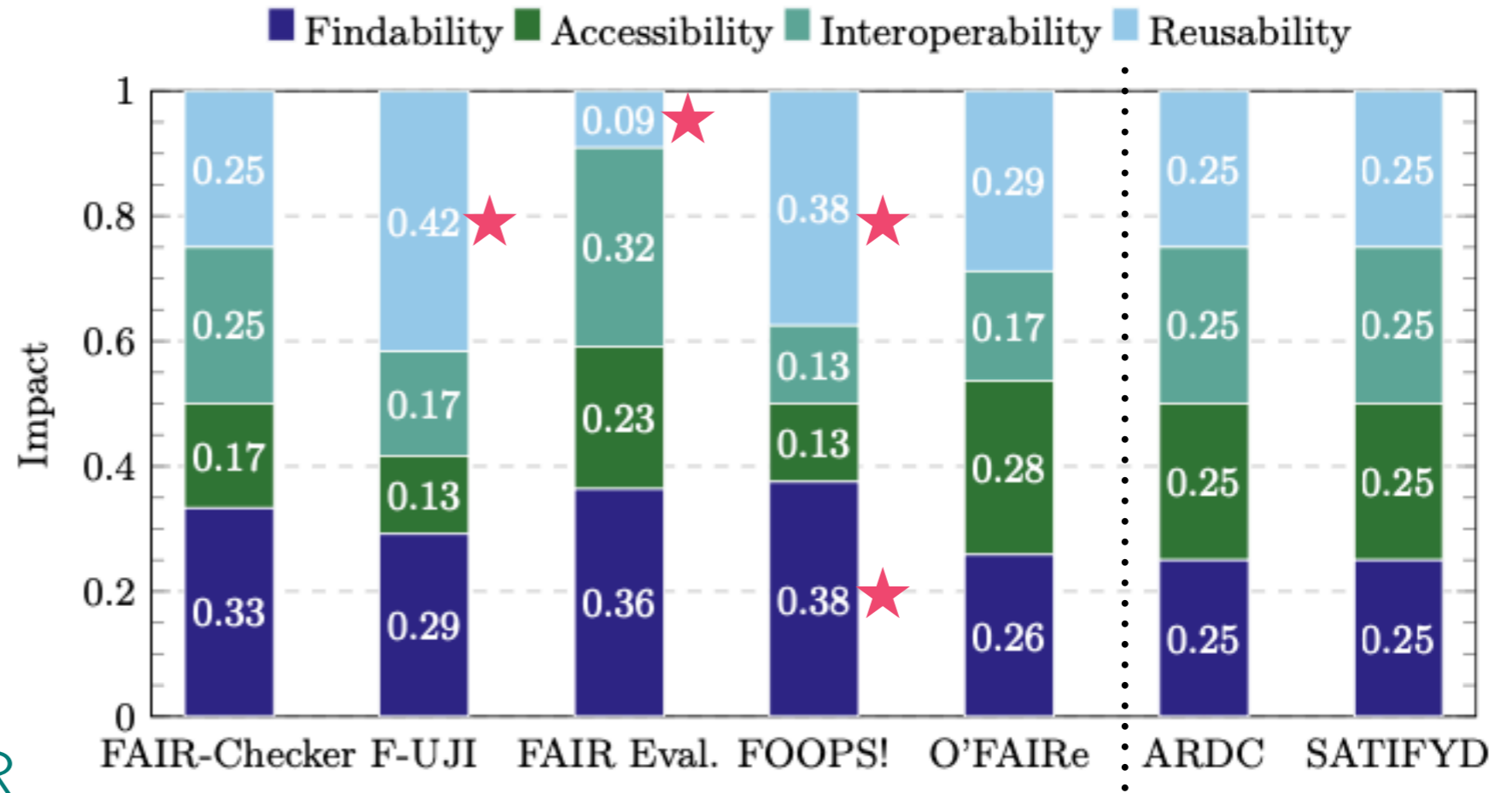


- ▶ O-FAIRe (targeting ontologies) has the most fine-grained evaluation (e.g. I2 and R1.2) → great diversity in metadata for ontologies
- ▶ FOOPS deeply investigate F1 (identifiers)

Impacts

Are all principle equally contributing to the global FAIR assesment score ?

→ "No ..."



How to get a good FAIR score with a minimal effort ?

- ▶ Pay attention to identifiers (F), license + provenance + domain-specific standards (R) if you use FOOPS!
- ▶ Not useful to spend energy on provenance or domain ontologies if you use FAIR-Evaluator ...
- ▶ ... but pay attention to it if you use F-UJI.

Conclusion and future works

Take-home message

- ▶ We need to understand the specificities of FAIR measurement tools
- ▶ We introduced a generic **model** for representing FAIR measures and computing their **granularity**, **coverage** and **impact**
- ▶ Some tools are "biased" (intentionally or not):
 - they explore more deeply some dimensions
 - which has an impact on the scores
- ▶ **Future works** include
 - better investigating implementations (e.g. dependencies), with tools developers
 - share these metadata on the web (e.g. DQV ontology)
 - contribute to FAIR harmonization efforts with other communities (softwares, workflows, machine learning resources ...)

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