

# The Semantic Web identity crisis: in search of the trivialities that never were

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**Abstract.** For a domain with a strong focus on unambiguous identifiers and meaning, the Semantic Web research field itself has a surprisingly ill-defined sense of identity. Started at the end of the 1990s at the intersection of databases, logic, and Web, and influenced along the way by all major tech hypes such as Big Data and machine learning, our research community needs to look in the mirror to understand who we really are. The key question amid all possible directions is pinpointing the important challenges we are uniquely positioned to tackle. In this article, we highlight the community’s unconscious bias toward addressing the Paretonian 80% of problems through research—handwavingly assuming that trivial engineering can solve the remaining 20%. In reality, that overlooked 20% could actually require 80% of the total effort and involve significantly more research than we are inclined to think, because our theoretical experimentation environments are vastly different from the open Web. As it turns out, these formerly neglected “trivialities” might very well harbor those research opportunities that only our community can seize, thereby giving us a clear hint of how we can orient ourselves to maximize our impact on the future. If we are hesitant to step up, more pragmatic minds will gladly reinvent technology for the real world, only covering a fraction of the opportunities we dream of.

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## 1. Back to the future

Re-reading the original Semantic Web vision [6] from 2001, we immediately notice where the predictions went wrong. Far less obvious are those that came true; they have become givens in today’s world, part of the *new normal* that now forms our everyday reality. We have forgotten the era ruled by the indestructible Nokia 3310, whose monochrome screen barely counted more pixels than a modern-day app icon, years before most people had Internet access at home—let alone on their *phone*. The crazy thing was imagining that we would be instructing our mobile devices to perform actions for us; the planning and realization of those actions were plausibly explained in the rest of the article. With the unimaginable eventually being solved after

a decade of research, the imaginable may have turned out to be the toughest nut to crack.

The Semantic Web’s roots can be traced further back to the initial Web proposal [1], whose opening diagram presents what we now refer to as a *knowledge graph*, an early glimpse into subject–predicate–object triples rather than the URL–HTTP–HTML triad that would ultimately become the Web. That same Web is currently facing severe threats [3–5], having rapidly gone from a utopian harbor of permissionless innovation to a potentially dystopian environment controlled by only a handful of dominant actors. The Semantic Web seems unaffected by most of this, strangely—until we realize that the Web and the Semantic Web have silently split ways not too long after the first RDF specifications appeared.

Nonetheless, semantic technologies are regularly coined as a means of tackling some of the Web’s most pressing challenges, such as combatting disinformation or fueling its re-decentralization movement [25].

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1 Meanwhile, the Semantic Web research community  
 2 is facing its own battles with some of the latest tech-  
 3 nological hypes, doubting between defending its own  
 4 relevancy next to Big Data, machine learning, and  
 5 blockchain, or surfing atop the waves created by those.  
 6 If you can't beat them, join them; if you can't join them,  
 7 repackage. The days when the keyword "semantics" led  
 8 to guaranteed project funding have faded faster from  
 9 our collective memory than the Nokia 3310 ever will.

10 Granted, cracks have started creeping into these other  
 11 technologies, too. Maybe Big Data is not limitless  
 12 in practice if technical capabilities scale faster than the  
 13 human and legal processes for ethical data management,  
 14 and we do need to link data across distributed sources  
 15 instead of unconditionally aggregating them. Perhaps  
 16 there are problems that machine learning can never solve  
 17 reliably, and the safety provided by first-order logic  
 18 proofs is irreplaceable for crucial decisions. And possi-  
 19 bly it will turn out that decentralized consensus only  
 20 touches a small part of all use cases, that disagreement  
 21 under the "anyone can say anything about anything" flag  
 22 provides a more workable model of the virtual world.

23 So when we are not riding others' waves, what is  
 24 it that unites the Semantic Web research community?  
 25 What makes us truly "us", what are the semantics we  
 26 can attach to our own identity? Having emerged at the  
 27 intersection of the Web, databases, and logic, we have  
 28 since become disconnected from these domains, our  
 29 awareness of which sometimes appears to be frozen  
 30 in time. We tend to disregard that the Web from which  
 31 we spun off is no longer the same as it was, and that  
 32 different approaches are required today. We have held  
 33 on to XML and RPC longer than most, confused the ends  
 34 with the means that were supposed to achieve them.

35 The main danger within an existential crisis is the  
 36 risk of losing our connection to the reality from which  
 37 we originate. The philosophy of our community seems  
 38 to align with Alan Kay's quote that "*The best way to*  
 39 *predict the future is to invent it.*" We build and we  
 40 investigate, expecting the future to wrap its arms around  
 41 the creations we are spawning. In this vision article, we  
 42 rather embrace John Perry Barlow's inversion of the  
 43 quote, in which "*The best way to invent the future is to*  
 44 *predict it.*" Looking back at the dreams from the past and  
 45 recombining those with the aspirations of the present,  
 46 what are the essential missing pieces that require our  
 47 unique dedication as Semantic Web scholars? As in the  
 48 original Semantic Web article, those topics that have  
 49 long been considered trivial might very well be the  
 50 hardest ones in practice [20].  
 51

1 In this article, we make the case for a return to our  
 2 roots of "Web" and "semantics", from which we as  
 3 a Semantic Web community—what's in a name—seem  
 4 to have drifted in search for other pursuits that, however  
 5 interesting, perhaps needlessly distract us from the  
 6 quest we had tasked ourselves with. In covering this  
 7 journey, we have no choice but to trace those meandering  
 8 footsteps along the many detours of our community—  
 9 yet this time around with a promise to come back home  
 10 in the end.  
 11

## 12 2. A little semantics

13 The term "Semantic Web" evidently coincides with  
 14 adding *semantics* to Web content in order to improve  
 15 interpretation by machines. However, after two decades  
 16 of debate, we still seem uncertain about exactly how  
 17 much semantics are in fact useful. The gap between  
 18 data that are published and applications that should  
 19 consume them continues to grow. While the call for  
 20 Linked Data has brought us the eggs, the chickens that  
 21 were supposed to be hatching them are still missing,  
 22 partly because making sense of others' data remains hard.

23 To intertwine data with meaning, we rely on RDF's  
 24 capabilities for exchange and interoperability. But what  
 25 is out there is factual knowledge in a (hyper)graph struc-  
 26 ture, with URIs to uniquely identify terms. The intended  
 27 meaning of the data is captured through knowledge  
 28 representation ontologies such as RDFS or OWL, and  
 29 can be discovered through dereferencing. In that sense,  
 30 data in RDF actually *refer* to their semantics rather than  
 31 *containing* them. And distributing those semantics has  
 32 turned out significantly harder than distributing data.  
 33

34 Early efforts were devoted to the development of  
 35 ontology engineering, and understandably so. Having  
 36 generic software to automatically act on a variety of  
 37 independent data sets was what made the Semantic Web  
 38 vision so appealing. Once domain knowledge had been  
 39 formalized, it could be applied to represent facts, from  
 40 which reasoners could automatically derive *new* facts.  
 41 Yet once we took those endeavors to the Web, it be-  
 42 came apparent we had missed the general practical im-  
 43 plications. As semantics are always consensus-based,  
 44 domain models are only as valuable as the scope of  
 45 the underlying consensus. Hence, their usage cannot  
 46 be guaranteed by parties that were not involved or dis-  
 47 agree with the consensus. Often, people resort to miti-  
 48 gation strategies that disregard the semantics enshrined  
 49 in description logic, by selectively reusing properties  
 50  
 51

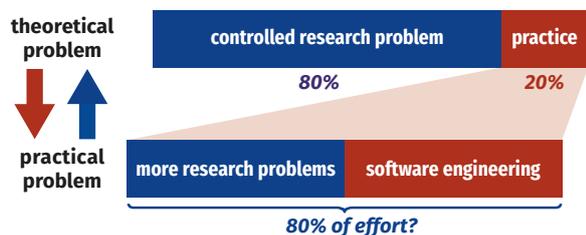


Fig. 1. After having solved the core 80% of a research problem, we often assume that the remaining 20% are practicalities that can be addressed through trivial engineering. In reality, lifting research from controlled experimental environments to the open Web likely leads to other research problems. In addition to bringing problems from theory to practice, we can let practical problems inspire theory.

and classes upon publication, or freely reinterpreting semantics upon consumption.

Core frameworks such as RDF and OWL are sometimes labeled as “by academics, for academics” because of their perceived complexity by developers. Due to a lack of deeper understanding and a shortage of connections to existing development practice, ontologies are in practice often reduced to more prescriptive *vocabularies* that basically again leave semantics up to individual applications. The desire for more simple choices with less flexibility is illustrated by the backing of Schema.org by the major search engines and the increasing popularity of the shape languages SHACL and ShEx. They cover an important gap between data in the wild and applications that need to know what kind of data to expect—one of the aspects we probably want to keep our eyes on.

The disconnect between the need of semantics and the effort to provide it, has cultivated a *heterogeneous* and *underspecified* Web of Data [19]. We cannot afford any longer to handwavingly address practical implementation and usability with deep theories. As depicted in Fig. 1, a strong implicit assumption underlies a lot of our work: that solving the core 80% of a problem is where research is needed, and that the remaining 20% consist of simple engineering to take that research from theory to practice. However, is what we often dismiss as “engineering” *really* just a matter of writing more code? As scientists, we might want to validate that hypothesis, given the considerable problems that arise when we try to deploy semantics at Web-scale.

We need to consider the Web we *have*, before we can have the Web we *want*. After all, what good is high-performance inferencing if ontologies cannot be found or are outdated? What good are unique identifiers for concepts when stating equality with `owl:sameAs` is inadequate for applications [11]? How realistic is SPARQL as a *universal* query language if queries in prac-

tice have to be tailored to specific endpoints, because reasoning is only ever switched on in theory? Meanwhile, enterprises and developers start to give up on the formal semantics, and we risk the baby being thrown out with the bath water. That is the logical result if we leave the completion of the bigger Semantic Web picture to companies with a deadline. Their enthusiastic endorsement of shapes, for instance, could eventually suppress the practice of semantics in data. Researchers understand “a little semantics goes a long way” [14] to not necessarily mean that less semantics would be better than more. But exactly how much is too much for the actual Web? Only through research we can find out.

### 3. Where is the Web?

What arguably sets us apart besides semantics is, well, the Web. In contrast to relational or other databases, our domain of discourse is infinite and unpredictable on multiple levels. Because of the open-world assumption, no single RDF document contains the full truth. Even worse, any sufficiently large collection of Web documents will contain contradictions that, under classical logic, allows us to derive *any* truth—henceforth to be referred to as *ex Tela quodlibet*. Not only can anything be proven from a contradiction, in these days of fake news and dubious political advertising, it has never been easier to find self-consistent documents online in support of virtually any given conclusion or its opposite.

The Web is what we deliver as an answer to any Linked Data skeptic, as an irrefutable argument that all of our perceived or actual complexity is justified, because we are dealing with problems that span the entire virtual address space of the globe and in fact the universe. The Web is the reason why our ontologies are spread all over the place, why the prefix expansion for the OWL ontology counts 30 characters, why FOAF is forever stuck at version 0.9, the Dublin Core vocabulary at 3 different ones, and why we cannot all just settle on Schema.org. The Web is why Open Data exists, why our public SPARQL endpoints are down 1.5 days a month [7], why stable vocabularies suddenly disappear. Everything we do, we do it the way we do, because the Web sets the rules such that anything more simple or logical would *not* do. If the Web is such a self-explanatory answer to the existence of our discipline—then why are so afraid to put our work on top of it?

We are not even talking here about taking our scholarly communication to the Web; let that be the crusade of the dogfooders [8], to whom we dedicate Section 7. We

mean to say that “*it works in our university basement*” has become an acceptable and applauded narrative—and to be fair to both the innocent and the guilty, impressive efforts undertaken in such basements have rightly been awarded scientific stamps of excellence through rigorous non-Web peer review processes. However, we cannot claim the Web as the sole source of our intricacies, while simultaneously ignoring all of the Web’s difficulties by conducting all of our experiments in hermetically controlled environments. By doing so, we pretend that the comfortable 80% cannot significantly be affected by the unpredictable impurities of the 20%, that an  $n$ -fold performance gain in our basements can directly be extrapolated to the same gain for Linked Data in general. As Goodhart’s law states: “*When a measure becomes a target, it ceases to be a good measure*”, except that we can strongly question whether non-Web environments, pure and controlled as they are, have ever fulfilled the role of good measure providers in the first place.

No, we cannot safely assume that the `owl:sameAs` predicate has consistently been used in accordance with at least one of its several meanings [11]. No, we cannot assume that SPARQL endpoints will be available or even return valid RDF, or that any RDF document out there is syntactically valid, coherent, and free of ontological abuse [15]. Yes, people will use the same URL to refer to *different* things, and obviously different URLs to point to the *same* things—without even throwing in as little as a semantically ambiguous `schema:sameAs`. Yes, our precious data sets unnecessarily use different ontologies, so we have to switch on reasoning, even though that makes benchmark results suddenly *worse* than the state of the art—and did we mention that one of those ontologies no longer dereferences but, even back when it still did, was not linked to the others anyway? Upon closer reflection, our fears about testing on the Web are probably justified; our scientific conclusions and their presumed external validity perhaps a little less.

In all honesty, the academic community did take its *publish or perish* adage to heart, and is co-responsible for the billions of RDF triples currently on the public Web as Linked Open Data. But while the Web is a good platform for data publication, it is a pretty bad platform for data consumption [22], which is not coincidentally also where many challenges remain open. This is why we should not ignore the 20% any longer, but embrace the unique challenges and opportunities it brings. Crucial and sometimes counterintuitive insights arise when Web-based techniques are applied to research problems previously only studied in isolation. As an example, link-traversal-based query execution [12] taught us that

SPARQL queries can exist separately from specific interfaces to evaluate them, which in turn are independent from back-ends. Understanding that some of our standardized protocols do not adhere to the constraints of the Web’s underlying REST architectural style, allows us to design interfaces with better scalability properties, which might perform worse in closed environments but yield desirable properties on the public Web [27]. Taking this even further, we can wonder whether the greedy semantics of SPARQL queries are tailored too much to closed databases as opposed to the Web we claim to target. Are we ready for that Web?

We should, however, not become too puristic in our judgment; an important aspect of scientific studies is their ability to zoom in on the isolated contribution of specific factors. Many valid use cases for non-Web RDF applications exist, so not every single undertaking has to embody the omnipotent role ascribed to the mythical Semantic Web agent. Nonetheless, as a community, we want to ensure we combine the 80% sufficiently often with the 20%, such that we obtain at least a more adequate impression of the potentially huge number of research questions hiding in plain sight on the Web.

#### 4. “Linked” as bigger than “Big”

When Big Data became mainstream around 2010, the Semantic Web community was listening with great attention. After all, we had already been working with staggering numbers of facts, hundreds of millions of triples not being an exception. Furthermore, when considering all data on the Web as a whole, we would surely reach the threshold at which Linked Data should be considered Big Data in its own right.

However, Big Data and Linked Data are not necessarily structurally compatible. A main advantage of the RDF data model is that it allows for flexibility, enabling people to capture data that does not lend itself well to the columnar structures of spreadsheets and relational databases. Big Data solutions derive their strength from a strict and rigid structure, which strongly contrasts with RDF’s virtually unbounded freedom. While there have been solutions that leverage Big Data technologies to address RDF use cases such as querying [18], they require reformatting data to fit the Big Data paradigm.

A conceptual issue with the Big Data vision, at least for our purposes, is that it takes the path of the lowest common denominator, as a natural result of an aggregation process. While aggregation definitely has its merits for discovery and analysis, it also flattens unique

1 characteristics and attributes of individual data sets,  
2 dissolving them into a much larger and more homoge-  
3 neous space. An example of how this unintentionally  
4 can become troublesome is found within the Europeana  
5 initiative [16], which serves the noble cause of aggre-  
6 gating highly diverse metadata from cultural institu-  
7 tions all across Europe. However, several individual  
8 institutions felt wronged when they had to upload their  
9 data set—which they knew so well and had taken care of  
10 for so many years—only for it to be mingled with those  
11 of *others* who surely would have different accents and  
12 inferior quality thresholds [24]. What gives Big Data its  
13 attractiveness and efficiency might thus take away what  
14 differentiates us. Time will tell if similar arguments can  
15 be made about the Wikidata project [28], which aims  
16 to be a global knowledge base.

17 For some time, we have been mildly apologetic about  
18 not doing Big Data, at one point hastily rebranding our-  
19 selves as “Semantics and Big Data” [9] before realizing  
20 that, indeed, there is another research community out  
21 there that is better positioned to tackle those challenges.  
22 Considering the 2001 article [6] as the official birth date  
23 of the Semantic Web, let us conveniently ignore those  
24 teenage years during which we should be forgiven for  
25 rapidly cycling through different phases as we were in  
26 fact just constructing our own identity. We should not  
27 aspire to be that popular kid from high school, who, as it  
28 turned out later, had merely peaked early in life. Nearing  
29 our twenties now, let us stop apologizing already for  
30 just being ourselves.

31 If we conceptually think about Big Data versus what  
32 we are aiming to achieve with Linked Data, our chal-  
33 lenges might very well be the bigger ones. Notwith-  
34 standing impressive research and engineering efforts to  
35 scale up Big Data solutions the way they do, harvesting  
36 an enormous amount of homogeneous data in a single  
37 place creates ideal conditions for processing and anal-  
38 ysis. A small number of very large data sets is easier  
39 to manage than a very large number of small data sets.  
40 Size does matter, just not always in the way others think:  
41 the heterogeneity and distribution of Linked Data is  
42 currently at a level that cannot be adequately tackled  
43 with Big Data techniques. Instead of being ashamed  
44 about practicing Small Data, we should proudly flaunt  
45 its multitude and diversity. In times of increasing calls  
46 for inclusion, let this be a good thing.

47 Because even if we technically would be able to  
48 centralize everything in one place, we could only serve  
49 the relatively small space of public data, not all of  
50 the private data that is the focus point of Big Data  
51 applications. After all, there are very good reasons for

1 data to live in different places, not in the least legal or  
2 privacy concerns. Those needs are only becoming more  
3 pressing, given important drivers such as the GDPR legal  
4 framework in Europe, and a strong world-wide call for  
5 more choice and control over personal data. By keeping  
6 each individual’s data close by in a small personal store,  
7 people will be in a much better position to safeguard  
8 their most precious digital assets. The challenge then of  
9 course is in connecting these distributed pieces of data  
10 at runtime, which the Solid project [17] does through  
11 Linked Data.

12 In a distributed future, there will not be less data,  
13 but more; if it cannot reside in one place for whatever  
14 reason, it will have to be linked. This is yet another  
15 reason why we need to be prepared for Web-scale dis-  
16 covery and querying over federations that are magni-  
17 tudes more challenging than our current experimental  
18 environments.

## 21 5. AI beyond ML

22  
23 There is no question the age of *deep learning* is very  
24 much upon us. As the latest one to mature, deep learning  
25 has spawned numerous research efforts, techniques,  
26 and even production-ready applications with machine  
27 learning, elevating the state of AI once again. Semantic  
28 Web research has not been resilient to the siren song, and  
29 started exploiting RDF knowledge bases as fertile soil for  
30 deep learning and other machine learning approaches.  
31 Popular topics that emerged, such as *embeddings* and  
32 *concept learning* enable model training from description  
33 logics to complete and extend any semantic information  
34 present. Developing such approaches reduces the high  
35 manual effort currently required for participating in the  
36 Semantic Web.

37  
38 Semantic technologies were originally considered  
39 part of the AI family and in essence still are [10]. In-  
40 ference of logical consequences from data can drive a  
41 machine’s autonomy. Yet in the shadow of advanced  
42 machine learning, the “cool kids” perceive us as apostles  
43 of an old, inflexible, and outdated rule-based approach.  
44 However, maturation in the machine learning field also  
45 uncovered the gaps where semantic technology can  
46 prove its relevance. Use cases prone to *decision accu-*  
47 *racy*, such as healthcare or privacy enforcement, profit  
48 from the exact outcomes of first-order logic. Further-  
49 more, the ability of some semantic reasoners to *explain*  
50 their actions through proofs [26] is a much desired trait  
51 by the primarily black-box machine learning methods.

1 As both angles have their merits, the future is very  
 2 likely *hybrid*, and we need to further explore compli-  
 3 mentary roles. For instance, semantics and inference  
 4 can pre-label data to improve the accuracy of models.  
 5 Post-execution explainability could be achieved by rea-  
 6 soning over semantic descriptions of nodes. In the area  
 7 of personal digital assistants, declarative AI can append  
 8 a human representation of the world to representations  
 9 trained on raw data. This would fill knowledge gaps  
 10 of current assistants such as Siri and Alexa, increase  
 11 their associative ability, and eventually improve the au-  
 12 thenticity of their interactions. Some more fundamental  
 13 questions also need to be answered, such as training  
 14 a model under the open world assumption. Appropriate  
 15 strategies exist, but there are many more unknowns.

16 Semantic inference and first-order logic might lead  
 17 to less spectacular conclusions, but they will nonethe-  
 18 less be crucial to advanced machine learning systems.  
 19 Also here, it is important to solve the engineering side  
 20 of things. Several machine learning tools are readily  
 21 available for developers, who, through testing, discover  
 22 further challenges. When machine learning solutions  
 23 “just work”, developers do not need to know what is  
 24 inside; importantly, such simplicity is the result of  
 25 *research*, not just engineering. Getting rid of the “trivial”  
 26 problems with semantic inference hopefully means  
 27 providing these more spectacular results, on the Web.  
 28 Maybe this is the better way to position ourselves in the  
 29 next waves to come, such as reinforcement learning.

## 30 6. Challenging until proven trivial

31  
 32  
 33  
 34 Ultimately, all of above indicates a need to guard  
 35 ourselves from conducting research in a vacuum. Not  
 36 all science requires practical purposes, but many of  
 37 the research problems we study will never actually  
 38 occur if the Semantic Web does not take off any further,  
 39 so we should at least consider—for our own sake—  
 40 prioritizing those urgent problems that are blockers  
 41 to its adoption. Part of our hesitance might be that,  
 42 having fought hard for recognition as a scientific domain,  
 43 we are afraid to be pushed back into the corner of  
 44 engineering. We usually zoom in on very focused, often  
 45 incremental research problems, which tend to bring us  
 46 progress. Our conferences and journals strive to find  
 47 a high threshold for what qualifies as research, with  
 48 a strong focus on qualitative experimentation. Thereby,  
 49 we risk optimizing for familiarity and purity rather than  
 50 for originality and impact, because the scientific merits  
 51 of novel directions are inherently much harder to assess.

1 While high thresholds in general are commendable, they  
 2 also result in a higher percentage of false negatives,  
 3 both in submitted works that never get accepted, and  
 4 in stellar research ideas that never materialize because  
 5 fear of such rejections encourages safer bets.

6 As much time as we spend justifying ourselves toward  
 7 other communities, those efforts sometimes pale to how  
 8 other reviewers expect authors to justify their choice to  
 9 address pragmatic concerns that, all things considered,  
 10 should be no less of a scientific contribution. Pareto’s  
 11 law from Fig. 1 lures around the corner: we consider  
 12 the core 80% of a hard problem and assume that the  
 13 remaining 20% is a non-issue. Converting technological  
 14 research into digestible chunks for developers is consid-  
 15 ered trivial and outside of our scientific duty, despite  
 16 the considerable scientific challenges of creating simple  
 17 abstractions to complex technology, as the machine  
 18 learning community shows time and time again.

19 Yet everything that reeks of engineering is shunned.  
 20 However, most researchers in our community have not  
 21 built a single Semantic Web app, so we cannot pretend  
 22 to understand the insides of the 20%. As such, it is  
 23 impossible to tell whether that remainder is trivial or not.  
 24 We do not get in touch with some of the most pressing  
 25 issues, because we already ruled them out as trivial, and  
 26 then wonder about the reasons for the low adoption of  
 27 the otherwise excellent 80% research.

28 Since the Semantic Web started, Web development  
 29 has massively changed. Many apps are now built by  
 30 *front-end developers*, for whom Semantic Web technol-  
 31 ogies are inaccessible—explaining the success of sub-  
 32 stantially less powerful but far more developer-friendly  
 33 technologies such as GraphQL. The GraphQL commu-  
 34 nity, who have prided themselves on simplicity com-  
 35 pared to the Semantic Web technology stack, are slowly  
 36 discovering that they were merely solving simpler prob-  
 37 lems. Queries with local semantics indeed become prob-  
 38 lematic if data needs to come from multiple sources.  
 39 Instead of reusing the lessons from years of SPARQL  
 40 federation research, the GraphQL community rather  
 41 reinvents ontologies by calling them “schema stitch-  
 42 ing” [21]. Persisting on the pragmatic road, which they  
 43 initially took because our alternative was deemed too  
 44 complex, they might ironically end up with something as  
 45 difficult but less powerful, because they did not have the  
 46 same forethought. Even more ironic is that we remain  
 47 stuck in that forethought and wonder when adoption is  
 48 coming. We compensate by drawing such new technol-  
 49 ogies back into the research domain [13], but gloss over  
 50 a crucial point: bringing SPARQL levels of expressivity  
 51 to front-end developers is in fact a *research* problem.

1 Designing an appropriate Linked Data *developer experience* [23] is so challenging because, while regular  
2 apps are hard-coded against one specific well-known  
3 back-end, Linked Data apps need to expect the unex-  
4 pected as they interface with heterogeneous data from all  
5 over the Web. Building such complex behavior involves  
6 a sophisticated integration of many branches of our  
7 research, which requires designing and implementing  
8 complex program code. Exposing such complex behav-  
9 ior into simple primitives, as is needed for front-end  
10 developers, requires *automating* the generation of that  
11 complex code, likely at runtime. Such endeavours have  
12 not been attempted at the research level, let alone would  
13 they be ready for implementation by skilled engineers.

14 This research gap between current research solutions  
15 and practice means that much of our work cannot be  
16 applied. Some find it acceptable that nothing works in  
17 practice yet. Unfortunately, such a lax attitude leaves  
18 us with an all too comfortable hiding spot: why would  
19 my research have to work in the real world if others'  
20 does not? As a direct consequence of this line of thought,  
21 we cannot meaningfully distinguish research that could  
22 eventually work from research that never will.

23 Until we have examined whether or not something is  
24 trivial, we should not make any implicit assumptions.  
25 Perhaps we should consider scoring manuscripts on the  
26 80/20 Pareto scale, and ensure that we have enough of  
27 both sides at our conferences and in our journals. By  
28 also judging applicability, we abandon our filter bubbles  
29 and extend our action radius to urgent problems in the  
30 way of adoption—which will only grow our research  
31 community.

## 32 7. Practice what we preach

33  
34  
35 Not only do many of us lack Semantic Web experi-  
36 ence as app developers, our even bigger gap is experi-  
37 ence as users. Although a significant amount of our  
38 communication (not in the least toward funding bodies)  
39 consists of technological evangelism, we rarely succeed  
40 in leveraging our own technologies. If we keep on find-  
41 ing excuses for not using our own research outcomes,  
42 how can we convince others? The logicians among us  
43 will undoubtedly recognize the previous statement as  
44 a *tu quoque* fallacy: our reluctance to dogfood is factu-  
45 ally independent of our technology's claim to fame. Yet  
46 if all adoption were solely based on sound reasoning,  
47 our planet would look very different today. Credibility  
48 and fairness aside, we are not in the luxury position  
49 to tell others to “do as I say, not as I do.” The burden

1 of proof is entirely upon ourselves, and the required  
2 evidence extends beyond the scientific.

3 In addition to being an instrument of persuasion,  
4 dogfooding addresses a more fundamental question:  
5 which parts of our technology are ready for prime time,  
6 and which parts are not? By becoming users of our own  
7 technologies, we will gain a better understanding of the  
8 elusive 20% that clearly, had it actually been so trivial,  
9 would already have been there. Never underestimate the  
10 power of frustration: feeling frustrated about unlocked  
11 potential is what prompted Tim Berners-Lee to invent  
12 the Web [2]. Only by managing almost his entire life  
13 with Linked Data, he is able to keep a finger on the  
14 Semantic Web's pulse, and his eyes on its Achilles' heel.

15 If we similarly had a deeper understanding of real-  
16 world Linked Data flows and obstacles, would we not  
17 be in a better position to make a difference? We might  
18 want to address concrete problems happening today, in  
19 addition to targeting those that will hopefully arise—  
20 conditional on today's problems ending up solved—after  
21 several more years.

## 22 8. In conclusion

23  
24  
25 After almost two decades, the Semantic Web should  
26 step out of its identity crisis into adolescence. In search  
27 of a target market for adoption, research in semantic  
28 technologies has ridden others' waves perhaps a little  
29 too often. While those bring in useful lessons to be  
30 learned, we should not forget to learn our own on the  
31 place where we can make a major difference: the Web.  
32 There, new technologies still emerge every day—just  
33 not ours. Investing in theoretically interesting problems  
34 without also delivering the necessary research to achieve  
35 practical implementations seems to have singled us out.

36 A Semantic Web has data and semantics intertwined,  
37 yet distributing those semantics has been proven hard.  
38 Can we focus on the practice and implications of sharing  
39 and preserving semantics? If not, we might leave the  
40 original vision to die in the hands of a more short-term  
41 and pragmatic agenda. No doubt, the need for full-  
42 scale data integration will eventually reappear, possibly  
43 reinventing the solutions and methods we are working  
44 on today. But that realization might take another decade.

45 The Web might not be our only target market, but it  
46 is the one that sets us apart. Yet it does not pop up in  
47 the average “threats to validity” section of articles—if  
48 there even is one. The rules are set in a unique way,  
49 which requires overcoming specific hurdles to make  
50 things work. To really test the external validity of our  
51

work, we should submerge in the practical side of things and thus make the Web a better suited place for data consumption. Our experimental environment should not be the same as that of Big Data; we should thrive with a lot of small data sets instead of a few large ones, and in heterogeneity instead of homogeneity. We could differentiate ourselves as the main driver for the much needed re-decentralization of the Web, where, backed by privacy and data legislation, Web-scale federation is the next big thing. To this end, positioning semantic technologies as a complement to machine learning is a necessity. The future of AI is hybrid: descriptive logic can bring accuracy, explainability and, of course, meaningful data to the table.

In order to succeed, we will need to hold ourselves to a new, significantly higher standard. For too many years, we have expected engineers and software developers to take up the remaining 20%, as if they were the ones needing to catch up with us. Our fallacy has been our insistence that the remaining part of the road solely consisted of code to be written. We have been blind to the substantial research challenges we would surely face if we would only take our experiments out of our safe environments into the open Web. Turns out that the engineers and developers have moved on and are creating their own solutions, bypassing many of the lessons we already learned, because we stubbornly refused to acknowledge the amount of research needed to turn our theories into practice. As we were not ready for the Web, more pragmatic people started taking over.

And if we are honest, can we blame them? Clearly, the world will not wait for us. Let us not wait for the world.

## References

- [1] Tim Berners-Lee. Information management: A proposal. Technical report, CERN, March 1989. <https://www.w3.org/History/1989/proposal.html>.
- [2] Tim Berners-Lee. The next Web, 2009. [https://www.ted.com/talks/tim\\_berners\\_lee\\_on\\_the\\_next\\_web](https://www.ted.com/talks/tim_berners_lee_on_the_next_web).
- [3] Tim Berners-Lee. Three challenges for the Web, according to its inventor. Web Foundation, March 2017. <https://webfoundation.org/2017/03/web-turns-28-letter/>.
- [4] Tim Berners-Lee. The Web is under threat. join us and fight for it. Web Foundation, March 2018. <https://webfoundation.org/2018/03/web-birthday-29/>.
- [5] Tim Berners-Lee. 30 years on, what's next #ForTheWeb? Web Foundation, March 2019. <https://webfoundation.org/2019/03/web-birthday-30/>.
- [6] Tim Berners-Lee, James Hendler, and Ora Lassila. The Semantic Web. *Scientific American*, 284(5):34–43, May 2001. doi: 10.1038/scientificamerican0501-34.
- [7] Carlos Buil-Aranda, Aidan Hogan, Jürgen Umbrich, and Pierre-Yves Vandenbussche. SPARQL Web-querying infrastructure: Ready for action? In Harith Alani, Lalana Kagal, Achille Fokoue, Paul Groth, Chris Biemann, Josiane Xavier Parreira, Lora Aroyo, Natasha Noy, Chris Welty, and Krzysztof Janowicz, editors, *Proceedings of the 12<sup>th</sup> International Semantic Web Conference*, volume 8219 of *Lecture Notes in Computer Science*, pages 277–293, Heidelberg, Germany, 2013. Springer. doi: 10.1007/978-3-642-41338-4\_18.
- [8] Sarven Capadisli. *Linked Research on the Decentralised Web*. PhD thesis, University of Bonn, 2019. <https://csarven.ca/linked-research-decentralised-web>.
- [9] Philipp Cimiano, Oscar Corcho, Valentina Presutti, Laura Hollink, and Sebastian Rudolph, editors. *The Semantic Web: Semantics and Big Data*, volume 7882 of *Lecture Notes in Computer Science*, Heidelberg, Germany, 2013. Springer. doi: 10.1007/978-3-642-38288-8.
- [10] Harry Halpin. The Semantic Web: the origins of artificial intelligence redux. In *Third International Workshop on the History and Philosophy of Logic, Mathematics, and Computation*, 2004.
- [11] Harry Halpin, Patrick J. Hayes, James P. McCusker, Deborah L. McGuinness, and Henry S. Thompson. When owl:sameAs isn't the same: An analysis of identity in Linked Data. In Peter F. Patel-Schneider, Yue Pan, Pascal Hitzler, Peter Mika, Lei Zhang, Jeff Z. Pan, Ian Horrocks, and Birte Glimm, editors, *Proceedings of the 9<sup>th</sup> International Semantic Web Conference*, Lecture Notes in Computer Science, pages 305–320, Heidelberg, Germany, 2010. Springer. doi: 10.1007/978-3-642-17746-0\_20.
- [12] Olaf Hartig, Christian Bizer, and Johann-Christoph Freytag. Executing SPARQL queries over the Web of Linked Data. In Abraham Bernstein, David R. Karger, Tom Heath, Lee Feigenbaum, Diana Maynard, Enrico Motta, and Krishnaprasad Thirunarayan, editors, *Proceedings of the 8<sup>th</sup> International Semantic Web Conference*, pages 293–309, Heidelberg, Germany, 2009. Springer. doi: 10.1007/978-3-642-04930-9\_19.
- [13] Olaf Hartig and Jorge Pérez. Semantics and complexity of GraphQL. In Pierre-Antoine Champin, Fabien Gandon, Mounia Lalmas, and Panagiotis G. Ipeirotis, editors, *Proceedings of the 2018 World Wide Web Conference*, pages 1155–1164, Geneva, Switzerland, 2018. ACM. doi: 10.1145/3178876.3186014.
- [14] James Hendler. The dark side of the Semantic Web. *IEEE Intelligent Systems*, 22(1):2–4, 2007. doi: 10.1109/MIS.2007.17.
- [15] Aidan Hogan, Andreas Harth, Alexandre Passant, Stefan Decker, and Axel Polleres. Weaving the Pedantic Web. In Christian Bizer, Tom Heath, Tim Berners-Lee, and Michael Hausenblas, editors, *Proceedings of the 3<sup>rd</sup> Workshop on Linked Data on the Web*, volume 628 of *CEUR Workshop Proceedings*, Aachen, Germany, 2010. CEUR-WS.
- [16] Antoine Isaac and Bernhard Haslhofer. Europeana Linked Open Data – data.europeana.eu. *Semantic Web Journal*, 4(3):291–297, July 2013. doi: 10.3233/SW-120092.
- [17] Essam Mansour, Andrei Vlad Sambra, Sandro Hawke, Maged Zereba, Sarven Capadisli, Abdurrahman Ghanem, Ashraf Aboul-naga, and Tim Berners-Lee. A demonstration of the Solid platform for social Web applications. In *Companion Proceedings of the 25<sup>th</sup> International Conference on World Wide Web*, pages 223–226, Geneva, Switzerland, 2016. ACM. doi: 10.1145/2872518.2890529.
- [18] Alexander Schätzle, Martin Przyjacieli-Zablocki, Antony Neu, and Georg Lausen. Sempala: Interactive SPARQL query processing on Hadoop. In Peter Mika, Tania Tudorache, Abraham

- 1 Bernstein, Chris Welty, Craig Knoblock, Denny Vrandečić, Paul  
2 Groth, Natasha Noy, Krzysztof Janowicz, and Carole Goble,  
3 editors, *Proceedings of the 13<sup>th</sup> International Semantic Web  
4 Conference*, volume 8796 of *Lecture Notes in Computer Science*,  
5 pages 164–179, Heidelberg, Germany, 2014. Springer. doi:  
6 10.1007/978-3-319-11964-9\_11.
- [19] Max Schmachtenberg, Christian Bizer, and Heiko Paulheim.  
7 Adoption of the Linked Data best practices in different topical  
8 domains. In Peter Mika, Tania Tudorache, Abraham Bernstein,  
9 Chris Welty, Craig Knoblock, Denny Vrandečić, Paul Groth,  
10 Natasha Noy, Krzysztof Janowicz, and Carole Goble, editors,  
11 *Proceedings of the 13<sup>th</sup> International Semantic Web Conference*,  
12 volume 8796 of *Lecture Notes in Computer Science*, pages 245–  
13 260, Heidelberg, Germany, 2014. Springer. doi: 10.1007/978-3-  
14 319-11964-9\_16.
- [20] Clay Shirky. The Semantic Web, syllogism, and worldview,  
15 2003. [http://www.shirky.com/writings/hercomeseverybody/  
16 semantic\\_syllogism.html](http://www.shirky.com/writings/hercomeseverybody/semantic_syllogism.html).
- [21] Sashko Stubailo. The next generation of schema  
17 stitching, 2018. [https://blog.apollographql.com/  
18 the-next-generation-of-schema-stitching-2716b3b259c0](https://blog.apollographql.com/the-next-generation-of-schema-stitching-2716b3b259c0).
- [22] Frank van Harmelen. 10 years of Semantic Web: does it  
19 work in theory?, 2011. [https://www.cs.vu.nl/~frankh/spool/  
20 ISWC2011Keynote/](https://www.cs.vu.nl/~frankh/spool/ISWC2011Keynote/).
- [23] Ruben Verborgh. Designing a Linked Data developer ex-  
21 perience, 2018. [https://ruben.verborgh.org/blog/2018/12/28/  
22 designing-a-linked-data-developer-experience/](https://ruben.verborgh.org/blog/2018/12/28/designing-a-linked-data-developer-experience/).
- [24] Ruben Verborgh. One flew over the cuckoo’s nest – The role of  
23 aggregation on a decentralized Web, 2018. [https://rubenverborgh.  
24 github.io/EuropeanaTech-2018/](https://rubenverborgh.github.io/EuropeanaTech-2018/).
- [25] Ruben Verborgh. Re-decentralizing the Web, for good this time.  
25 In Oshani Seneviratne and James Hendler, editors, *Linking the  
26 World’s Information: Tim Berners-Lee’s Invention of the World  
27 Wide Web*, pages 225–237. ACM, 2020.
- [26] Ruben Verborgh, Dörthe Arndt, Sofie Van Hoecke, Jos De Roo,  
28 Giovanni Mels, Thomas Steiner, and Joaquim Gabarro. The  
29 pragmatic proof: Hypermedia API composition and execution.  
30 *Theory and Practice of Logic Programming*, 17(1):1–48, 2017.  
31 doi: 10.1017/S1471068416000016.
- [27] Ruben Verborgh, Miel Vander Sande, Olaf Hartig, Joachim  
32 Van Herwegen, Laurens De Vocht, Ben De Meester, Gerald  
33 Haesendonck, and Pieter Colpaert. Triple Pattern Fragments:  
34 a low-cost knowledge graph interface for the Web. *Journal of  
35 Web Semantics*, 37–38:184–206, March 2016. doi:  
36 10.1016/j.websem.2016.03.003.
- [28] Denny Vrandečić and Markus Krötzsch. Wikidata: A free  
37 collaborative knowledgebase. *Communications of the ACM*,  
38 57:78–85, 2014. doi: 10.1145/2629489.