**RDF/OWL and SPARQL instead of NoSQL databases**

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**Abstract.** NoSQL database technology has become increasingly popular in the data management field over the past few years with more than 25 percent adoption in 2014[[1]](#footnote-1). This phenomenal growth has occurred in spite of  some striking shortcomings in NoSQL technologies. In particular, NoSQL databases currently lack a standard query language and do not support the critical standard transaction processing support found in modern relational data management systems. This paper proposes that RDF/OWL and Sparql technology is an appropriate alternative to current NoSQL implementations. Sparql can act as a standard query language for NoSQL implementations, and existing RDF/Owl database implementations today already provide full transaction processing support. This paper also argues that RDF/OWL database implementations are well suited for Database as a service environments, DBaaS.

In addition this paper outlines the game-changing potential of Inferencing support in data management. Inferencing is natively supported in OWL-based systems, but not available in relational or current NoSQL databases.

**Keywords:** Data Management, Query Translation, RDF, OWL, SPARQL, SQL, and Inference.

1. **Introduction**

NoSQL database systems, especially MongoDB[[2]](#footnote-2) (MongoDB, 2014), have been gaining notoriety over the past several years purportedly for several reason 1) their data model (e.g., JSON) and its “schema-less flexibility” is a better match for data-driven web applications; 2) they are well suited for Database-as-a-Service (DBaaS) implementations, e.g., (mongolab, 2014); and 3) they can scale across small, inexpensive systems, i.e., they can “scale-out”. However, the underlying “schema-less flexible” data model for these systems is very similar to technology that has a very long history in the form of Entity-Attribute-Value (EAV) database systems (Stead, Hammond, & Straube, 1982) (McDonald, Blevins, Tierney, & Martin, 1988).

EAV technology has evolved into the standards-based RDF/OWL and SPARQL technology of today[[3]](#footnote-3) and this paper will show that RDF/OWL and SPARQL are practical alternatives to current NoSQL databases for building data-driven web applications and for DBaaS implementations. (Scale-out won’t be discussed in this paper because it is believed to be orthogonal to the discussion.)

As a case study to demonstrate the capabilities of RDF/OWL and SPARQL, this paper will present an implementation of a simple, flask-based (Flask, 2014) “book” website developed using the MongoDB NoSQL database system and the same website developed with RDF/OWL and SPARQL embedded in a system called ReL[[4]](#footnote-4), This paper will also show how to build DBaaS applications using RESTful ReL. Both of these examples will show that RDF/OWL and SPARQL are very well suited for these types of applications.

Also, it is well known that a major weakness of NoSQl databases is that there is no standard, universal query language. The “book” example demonstrates that the SPARQL query language is a suitable, standard query language for NoSQL-type databases.

In addition, this paper will show that, when using Oracle’s RDF implementation (Oracle Graph, 2014), support for standard transaction properties such as read consistency, serializable (Concurrency, 2011), and ACID (Transactions, 2011) can be made available to NoSQL-like applications. This addresses another major weakness of NoSQL databases; that they do not provide this level of transaction support. MongoDB, for example, only supports single statement transactions and the entire Database[[5]](#footnote-5) is locked by the transaction statement.

Figure 1 shows the main menu for the “book” website example used in this paper.

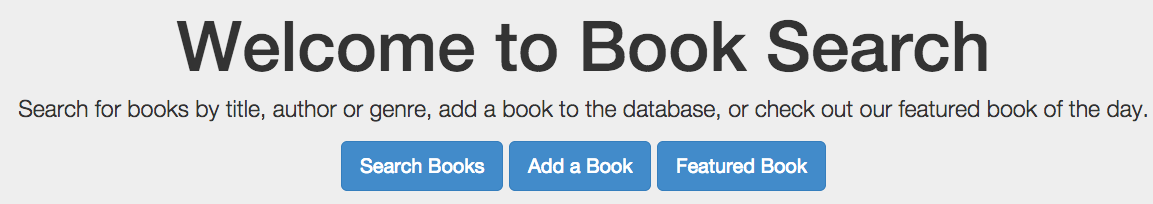


Figure 1

1. **Inserting data into the book website database**

Figure 2 shows the menu for adding a book in this web application.

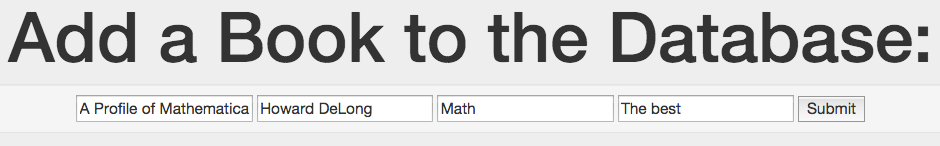


Figure 2

When the submit button is clicked, the following code is executed.

* 1. **For the MongoDB application**

First, a MongoDB connection is established:

from flask import Flask, render\_template, redirect, request

import pymongo

app = Flask(\_\_name\_\_, static\_url\_path = "")

connection\_string = "mongodb://127.0.0.1"

connection = pymongo.MongoClient(connection\_string)

database = connection.books

books = database.mybooks

Then, the insert is done using the MongoDB insert API:

@app.route('/add/', methods=['GET', 'POST'])

def add():

if request.method == 'POST':

new\_data = {k : v for k, v in request.form.items()}

**books.insert(new\_data)**

return render\_template('add.html', alert =

"success")

else:

return render\_template('add.html', alert="")

* 1. **For the ReL application**

First, a connection to an Oracle RDF datastore is established:

from flask import Flask, render\_template, redirect, request

app = Flask(\_\_name\_\_, static\_url\_path = "")

conn = connectTo 'jdbc:oracle:thin:@host:1521:orcl' 'user'

'password' 'rdf\_mode' 'bookApp'

Then, the insert is done using a standard SQL insert syntax[[6]](#footnote-6); however, no table named “books” was created beforehand, so this is “scheme-less,” and “flexible” just like MongoDB:

@app.route('/add/', methods=['GET', 'POST'])

def add():

if request.method == 'POST':

new\_data = {k : v for k, v in request.form.items()}

values = (str(new\_data['title']),

str(new\_data['author']), str(new\_data['genre']),

str(new\_data['description']))

**SQL on conn """insert into books(title, author,**

**genre, description) values"""values**

return render\_template('add.html', alert =

"success")

else:

return render\_template('add.html', alert="")

Behind the scenes, ReL converts the SQL insert into a series of several RDF/OWL insert statements as follows[[7]](#footnote-7) (data level triples are shown in bold below and OWL level triples are shown in italics and underlined).

**BEGIN**

**commit ;**

**set transaction isolation level serializable[[8]](#footnote-8) ;**

INSERT INTO BOOK\_DATA VALUES ( BOOK\_APP\_SQNC.nextval, SDO\_RDF\_TRIPLE\_S('FALL2014\_CS347\_PROF:<owl>', **'owl#89', 'owl#title', '"A Profile of Mathematical Logic"^^xsd:string'**));

INSERT INTO BOOK\_DATA VALUES ( BOOK\_APP\_SQNC.nextval, SDO\_RDF\_TRIPLE\_S('FALL2014\_CS347\_PROF:<owl>', *'owl#title', 'rdf:type', 'owl:DatatypeProperty'*));

INSERT INTO BOOK\_DATA VALUES ( BOOK\_APP\_SQNC.nextval, SDO\_RDF\_TRIPLE\_S('FALL2014\_CS347\_PROF:<owl>', *'owl#title', 'rdfs:domain', 'owl#books'*));

INSERT INTO BOOK\_DATA VALUES ( BOOK\_APP\_SQNC.nextval, SDO\_RDF\_TRIPLE\_S('FALL2014\_CS347\_PROF:<owl>', *'owl#title', 'rdf:range', 'rdfs:xsd:string'*));

INSERT INTO BOOK\_DATA VALUES ( BOOK\_APP\_SQNC.nextval, SDO\_RDF\_TRIPLE\_S('FALL2014\_CS347\_PROF:<owl>', *'owl#title', 'rdf:type', 'owl:FunctionalProperty'*));

INSERT INTO BOOK\_DATA VALUES ( BOOK\_APP\_SQNC.nextval, SDO\_RDF\_TRIPLE\_S('FALL2014\_CS347\_PROF:<owl>', **'owl#89', 'owl#author', '"Howard DeLong"^^xsd:string'**));

INSERT INTO BOOK\_DATA VALUES ( BOOK\_APP\_SQNC.nextval, SDO\_RDF\_TRIPLE\_S('FALL2014\_CS347\_PROF:<owl>', *'owl#author', 'rdf:type', 'owl:DatatypeProperty'*));

INSERT INTO BOOK\_DATA VALUES ( BOOK\_APP\_SQNC.nextval, SDO\_RDF\_TRIPLE\_S('FALL2014\_CS347\_PROF:<owl>', *'owl#author', 'rdfs:domain', 'owl#books'*));

INSERT INTO BOOK\_DATA VALUES ( BOOK\_APP\_SQNC.nextval, SDO\_RDF\_TRIPLE\_S('FALL2014\_CS347\_PROF:<owl>', *'owl#author', 'rdf:range', 'rdfs:xsd:string'*));

INSERT INTO BOOK\_DATA VALUES ( BOOK\_APP\_SQNC.nextval, SDO\_RDF\_TRIPLE\_S('FALL2014\_CS347\_PROF:<owl>', *'owl#author', 'rdf:type', 'owl:FunctionalProperty'*));

INSERT INTO BOOK\_DATA VALUES ( BOOK\_APP\_SQNC.nextval, SDO\_RDF\_TRIPLE\_S('FALL2014\_CS347\_PROF:<owl>', **'owl#89', 'owl#genre', '"Math"^^xsd:string'**));

INSERT INTO BOOK\_DATA VALUES ( BOOK\_APP\_SQNC.nextval, SDO\_RDF\_TRIPLE\_S('FALL2014\_CS347\_PROF:<owl>', *'owl#genre', 'rdf:type', 'owl:DatatypeProperty'*));

INSERT INTO BOOK\_DATA VALUES ( BOOK\_APP\_SQNC.nextval, SDO\_RDF\_TRIPLE\_S('FALL2014\_CS347\_PROF:<owl>', *'owl#genre', 'rdfs:domain', 'owl#books'*));

INSERT INTO BOOK\_DATA VALUES ( BOOK\_APP\_SQNC.nextval, SDO\_RDF\_TRIPLE\_S('FALL2014\_CS347\_PROF:<owl>', *'owl#genre', 'rdf:range', 'rdfs:xsd:string'*));

INSERT INTO BOOK\_DATA VALUES ( BOOK\_APP\_SQNC.nextval, SDO\_RDF\_TRIPLE\_S('FALL2014\_CS347\_PROF:<owl>', *'owl#genre', 'rdf:type', 'owl:FunctionalProperty'*));

INSERT INTO BOOK\_DATA VALUES ( BOOK\_APP\_SQNC.nextval, SDO\_RDF\_TRIPLE\_S('FALL2014\_CS347\_PROF:<owl>', **'owl#89', 'owl#description', '"The best"^^xsd:string'**));

INSERT INTO BOOK\_DATA VALUES ( BOOK\_APP\_SQNC.nextval, SDO\_RDF\_TRIPLE\_S('FALL2014\_CS347\_PROF:<owl>', *'owl#description', 'rdf:type', 'owl:DatatypeProperty'*));

INSERT INTO BOOK\_DATA VALUES ( BOOK\_APP\_SQNC.nextval, SDO\_RDF\_TRIPLE\_S('FALL2014\_CS347\_PROF:<owl>', *'owl#description', 'rdfs:domain', 'owl#books'*));

INSERT INTO BOOK\_DATA VALUES ( BOOK\_APP\_SQNC.nextval, SDO\_RDF\_TRIPLE\_S('FALL2014\_CS347\_PROF:<owl>', *'owl#description', 'rdf:range', 'rdfs:xsd:string'*));

INSERT INTO BOOK\_DATA VALUES ( BOOK\_APP\_SQNC.nextval, SDO\_RDF\_TRIPLE\_S('FALL2014\_CS347\_PROF:<owl>', *'owl#description', 'rdf:type', 'owl:FunctionalProperty'*));

**END ;**

**/**

As noted above, some of the INSERT statements insert appropriate OWL schema information into the RDF tuple-store, e.g., for the “title” attribute, the following is inserted:

**INSERT** INTO BOOK\_DATA VALUES ( BOOK\_APP\_SQNC.nextval, SDO\_RDF\_TRIPLE\_S('FALL2014\_CS347\_PROF:<owl>', *'owl#title',* ***'rdf:type', 'owl:DatatypeProperty'***))

**INSERT** INTO BOOK\_DATA VALUES ( BOOK\_APP\_SQNC.nextval, SDO\_RDF\_TRIPLE\_S('FALL2014\_CS347\_PROF:<owl>', *'owl#title',* ***'rdfs:domain', 'owl#books'***))

**INSERT** INTO BOOK\_DATA VALUES ( BOOK\_APP\_SQNC.nextval, SDO\_RDF\_TRIPLE\_S('FALL2014\_CS347\_PROF:<owl>', *'owl#title',* ***'rdf:range', 'rdfs:xsd:string'***))

**INSERT** INTO BOOK\_DATA VALUES ( BOOK\_APP\_SQNC.nextval, SDO\_RDF\_TRIPLE\_S('FALL2014\_CS347\_PROF:<owl>', *'owl#title',* ***'rdf:type', 'owl:FunctionalProperty'***))

This schema information is used when querying the book database.

1. **Searching the book website database**

Figure 3 shows the menu for searching for a book in this web application.



Figure 3

When the submit button is clicked, the following code is executed:

* 1. **For the MongoDB application**

The search is done using the MongoDB “find” API:

@app.route('/search/', methods=['GET', 'POST'])

def search():

if request.method == 'POST':

query = request.form['query']

return render\_template('search.html', posting=True,

query=query, title\_results = **books.find**({

'title':query}),author\_results = **books.find**({

'author':query}))

else:

return render\_template('search.html', posting =

False)

* 1. **For the ReL application**

The ReL search is done using standard SQL select statements as follows (the SQL select statements return a python tuple of tuples):

@app.route('/search/', methods=['GET', 'POST'])

def search():

if request.method == 'POST':

query = request.form['query']

titles[[9]](#footnote-9) = SQL on conn "select title, author from

books where title = '"query"'"

authors = SQL on conn "select title, author from

books where author = '"query"'"

title\_dict = convert\_to\_dict(titles)

author\_dict = convert\_to\_dict(authors)

genre\_dict = {}

no\_results = title\_dict == 0 and author\_dict == 0

and genre\_dict == 0

return render\_template('search.html', posting=True,

query=query, no\_results=no\_results,

title\_results=title\_dict,

author\_results=author\_dict,

genre\_results=genre\_dict)

else:

return render\_template('search.html',

posting=False)

Behind the scenes, ReL converts the SQL select statements into SPARQL statements, as shown below[[10]](#footnote-10),[[11]](#footnote-11):

SELECT v1 "title", v2 "author"

FROM TABLE(SEM\_MATCH('SELECT \* WHERE {

?s1 rdf:type :books .

OPTIONAL { ?s1 :title ?v1 }

OPTIONAL { ?s1 :author ?v2 }

**?s1 :title ?f1** .

FILTER(?f1 = "Howard DeLong") }' ,

SEM\_MODELS('FALL2014\_CS347\_PROF'), null,

SEM\_ALIASES( SEM\_ALIAS('', 'http://www.example.org/people.owl#')), null) )

SELECT v1 "title", v2 "author"

FROM TABLE(SEM\_MATCH('SELECT \* WHERE {

?s1 rdf:type :books .

OPTIONAL { ?s1 :title ?v1 }

OPTIONAL { ?s1 :author ?v2 }

**?s1 :author ?f1** .

FILTER(?f1 = "Howard DeLong") }' ,

SEM\_MODELS('FALL2014\_CS347\_PROF'), null,

SEM\_ALIASES( SEM\_ALIAS('', 'http://www.example.org/people.owl#')), null) )

1. **Transaction support in ReL**

As discussed in a previous footnote, the SPARQL statements shown above can be included in a regular Oracle transaction (see the example below). The same is true for all other database operations in ReL. So, ReL can provide traditional transaction support (i.e., read committed, serializable, and ACID) for all of its database operations. (Notice in Section 2 above, the INSERTS were wrapped in PL/SQL BEGIN and END statements, which maked the INSERTS ACID). One of the major criticisms of NoSQL databases is that they don’t provide traditional transaction support. ReL does not suffer from this problem when using Oracle’s implementation of RDF/OWL and SPARQL.

**commit ;**

**set transaction isolation level serializable ;**

SELECT v1 "title", v2 "author"

FROM TABLE(SEM\_MATCH('SELECT \* WHERE {

?s1 rdf:type :books .

OPTIONAL { ?s1 :title ?v1 }

OPTIONAL { ?s1 :author ?v2 }

?s1 :title ?f1 .

FILTER(?f1 = "Howard DeLong") }' ,

SEM\_MODELS('FALL2014\_CS347\_PROF'), null,

SEM\_ALIASES( SEM\_ALIAS('', 'http://www.example.org/people.owl#')), null) );

SELECT v1 "title", v2 "author"

FROM TABLE(SEM\_MATCH('SELECT \* WHERE {

?s1 rdf:type :books .

OPTIONAL { ?s1 :title ?v1 }

OPTIONAL { ?s1 :author ?v2 }

?s1 :author ?f1 .

FILTER(?f1 = "Howard DeLong") }' ,

SEM\_MODELS('FALL2014\_CS347\_PROF'), null,

SEM\_ALIASES( SEM\_ALIAS('', 'http://www.example.org/people.owl#')), null) );

1. **DBaaS - RESTful ReL**

ReL can also be run as a RESTful (Fielding, 2000) server, which means ReL SQL calls (e.g., """select title, author from books where title = '"""query"""'""", which was discussed above) can be embedded in any environment that supports the CURL function. For instance, we use RESTful ReL in R (Project, 2014) to access data and convert it to R data frames for analysis. The same “data model-to-RDF/OWL and SPARQL” translations that were discussed above can be used with Restful ReL. Here’s how ReL can be invoked from R to query the standard Oracle emp table:

d = getURL( URLencode('host:5000/rest/native/?query = "**select \* from emp**"'), httpheader = c(DB='jdbc:oracle:thin:@host:1521:orcl', USER='user', PASS='password', MODE='rdf\_mode', MODEL='Fall2014' , returnFor = 'R'), verbose = TRUE)

The “returnFor = 'R'” httpheader parameter value above directs RESTful ReL to return data in the following format:

d

"list(c('COMM', 'HIREDATE', 'JOB', 'DEPTNO', 'SAL', 'ENAME', 'MGR', 'EMPNO'), list(c('NULL', 1400, 'NULL', 'NULL', 500, 'NULL', 300, 'NULL', 'NULL', 'NULL', 'NULL', 'NULL', 'NULL', 'NULL'),c('23-JAN-1982', '28-SEP-1981', '1-MAY-1981', '3-DEC-1981', '22-FEB-1981', '9-JUN-1981', '20-FEB-1981', '8-SEP-1981', '12-JAN-1983', '09-DEC-1982', '17-NOV-1981', '17-DEC-1980', '3-DEC-1981', '2-APR-1981'),c('CLERK', 'SALESMAN', 'MANAGER', 'ANALYST', 'SALESMAN', 'MANAGER', 'SALESMAN', 'SALESMAN', 'CLERK', 'ANALYST', 'PRESIDENT', 'CLERK', 'CLERK', 'MANAGER'),c(10, 30, 30, 20, 30, 10, 30, 30, 20, 20, 10, 20, 30, 20),c(1300, 1250, 2850, 3000, 1250, 2450, 1600, 1500, 1100, 3000, 5000, 800, 950, 2975),c('MILLER', 'MARTIN', 'BLAKE', 'FORD', 'WARD', 'CLARK', 'ALLEN', 'TURNER', 'ADAMS', 'SCOTT', 'KING', 'SMITH', 'JAMES', 'JONES'),c(7782, 7698, 7839, 7566, 7698, 7839, 7698, 7698, 7788, 7566, 'NULL', 7902, 7698, 7839),c(7934, 7654, 7698, 7902, 7521, 7782, 7499, 7844, 7876, 7788, 7839, 7369, 7900, 7566)))"

Then, the data can be converted to an R data frame using the following two commands:  
  
df <- data.frame(eval(parse(text=substring(d,1)))[2])  
  
colnames(df) <- unlist(eval(parse(text=substring(d,1)))[1])  
  
Finally, head(df) results in the following:

head(df)

COMM HIREDATE JOB DEPTNO SAL ENAME MGR EMPNO

1 NULL 23-JAN-1982 CLERK 10 1300 MILLER 7782 7934

2 1400 28-SEP-1981 SALESMAN 30 1250 MARTIN 7698 7654

3 NULL 1-MAY-1981 MANAGER 30 2850 BLAKE 7839 7698

4 NULL 3-DEC-1981 ANALYST 20 3000 FORD 7566 7902

5 500 22-FEB-1981 SALESMAN 30 1250 WARD 7698 7521

6 NULL 9-JUN-1981 MANAGER 10 2450 CLARK 7839 7782

1. **Inference**

TBD

1. **Summary**

In an InfoWorld article (Oliver, 2014), the author claims, “the time for NoSQL standards is now”. But, RDF/OWL and SPARQL are standards that exist now and this paper has demonstrated that they are perfectly well suited for building “schema-less”, flexible NoSQL-type applications and DBaaS applications. What’s more, unlike other NoSQL systems, RDF/OWL and SPARQL systems like ReL can support standard, read committed, serializable, and ACID transaction processing. So, maybe “the time for NoSQL to use existing standards is now”.

The Oracle implementation of RDF/OWL and SPARQL (Oracle Graph, 2014) was used for the applications in this paper; however, any similar implementation of these standards could be used instead.

Inference has the potential to be a very powerful technology in data management, especially the notions of “type” and “inverse” . . .

This paper did not discuss the “scale-out rather than scale-up” proposition of NoSQL databases, but that debate has a decades long history and needs no more discussion here. However, there is no reason that RDF/OWL and SPARQL could not be used just as effectively in a “scale-out” system. As a matter of fact, ReL and RESTful ReL would run blazingly fast in Oracle’s Exadata (Exadata, 2014) environment, which is effectively “scale-out” albeit not inexpensive.

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1. 2014 Forrester Research, Inc. [↑](#footnote-ref-1)
2. Similarities with the Cassandra (Cassandra, 2014) NoSQL system will also be shown in Section 3 of this paper. [↑](#footnote-ref-2)
3. In this paper, we take the very simple view that RDF is a standard format for storing objects in a triple-store database, OWL is a standard format for storing metadata (i.e., schema information) about the objects in the triple store database, and SPARQL is a standard triple-store query language. More details can be found in the RDF Primer (RDF, 2014), OWL 2 Primer (OWL, 2014), and the SPARQL 1.1 Overview (SPARQL, 2014), however, these references often obscure this simple view. [↑](#footnote-ref-3)
4. ReL (Relation Language) is a python-based, data management system that uses RDF/OWL and SPARQL as its tuple manager. In addition to supporting RDF/OWL and SPARQL, ReL is data model agnostic and allows data manipulation and retrieval using a mix and match of many different higher-level data models, including the Relational Model, a Semantic Model based upon the work of Hammer and McLeod (Hammer & McLeod, 1981), and the OO python model. The ReL Relational Model, which is automatically translated to RDF/OWL and SPARQL, is used in this paper. However, because ReL is a python-based system, it’s trivial to also support JSON by translating JSON into one of the other supported data models and to translate results back into JSON. This has been done in several ReL applications. [↑](#footnote-ref-4)
5. A Mongo Database is a container for a set of Collections and a Collection is a container for a set of JSON Documents. [↑](#footnote-ref-5)
6. It will be shown later in the paper that these SQL inserts are converted to RDF inserts. [↑](#footnote-ref-6)
7. URIs have been abbreviated to help with readability. [↑](#footnote-ref-7)
8. Notice that Oracle allows RDF triple-store statements to be wrapped in standard SQL. This means standard transaction processing can be done with RDF triple-store statements. The same is true for Oracle SPARQL statements as will be seen in Section 4. [↑](#footnote-ref-8)
9. The SQL statement returns a python tuple of tuples. [↑](#footnote-ref-9)
10. Notice, the SPARQL statement is in an Oracle Table function that is a part of a standard SQL statement. [↑](#footnote-ref-10)
11. The use of the OPTIONAL pattern in the SPARQL statements means that each of the attributes modified by the OPTIONAL pattern will optionally be part of a returned tuple (row). This is like the CASSANDRA NoSQL database system where “Unlike a table in an RDBMS, different rows in the same column family do not have to share the same set of columns, and a column may be added to one or multiple rows at any time.” (Cassandra, 2014) [↑](#footnote-ref-11)