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### Introduction

The goal of this coursework was to put into practice the ontology modelling and semantic data development skills, and to accomplish this I have defined an ontology, populated it from external source (basic task), added extra information into it from the second external source (bonus task) and provided a couple of queries for the ontology populated from the two different sources to demonstrate that it is working correctly and that it can provide the combined information.

## **Constructing the Ontology**

First, the purpose of the ontology needed to be established. The ontology that was created and populated from two external sources is covering the domain of movies and countries and allows such queries as "Find a movie filmed in countries with population more than 100M". Second step was to consider the ways of re-using the existing ontologies. In order to build and populate an ontology, I was using 2 endpoints, namely – DBpedia and Factbook (the corresponding links for these knowledge bases can be found under Bibliography section). Third, a list of terms and their internal organisation needed to be identified. In order to define the structure of the ontology I was using Protégé-OWL2. The resulting ontology can be found in *ontology.owl*. The process of defining the ontology is an iterative process, which is why I had to review my ontology multiple times (both when constructing the ontology structure and when attempting to populate it) in order to estimate if it can store sufficient information from the two endpoints, before I finally came up with the resulting model. In order to accomplish this I was studying several different individuals from the two knowledge bases, for instance I explored such movie entries from the DBpedia knowledge base as "Hysterical blindness" and "The Great Beauty", and such country entries from the Factbook ontology as "Afghanistan" and "Uzbekistan" (the corresponding links are provided under Bibliography).

### Class hierarchy

Next, the schema and skeleton of the ontology had to be defined. Based on the structure of the information contained in the two endpoints, I created 11 classes for my ontology, which are shown in *Image 1. Country, Distributor, Language, Movie* and *Person* are the top level classes, whereas *Actor, Cinematographer, Composer, Director, Producer* and *Screenwriter* are the subclasses of the

Person class. One of the challenges while creating the ontology was to decide which information should be stored and how it should be contained in the ontology. For this purpose I used the movie (DBpedia) and country (Factbook) entries mentioned above in order to choose which properties of the movies and countries I want to represent as a class and in which way it would be more appropriate to store them.

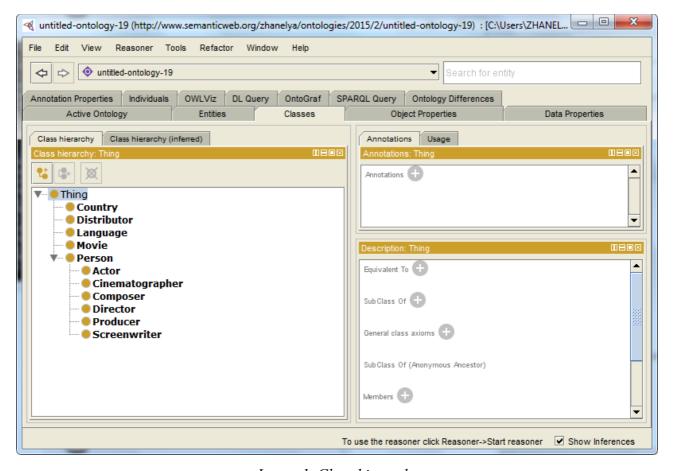


Image 1. Class hierarchy

### · Object properties

In order to enhance the ontology implementation, object properties had to be created to link different concepts together by forming reasonable relations among them. A full list of the object properties defined can be seen in *Image 2* below. For each object property, I defined whether it is functional, inverse functional, transitive, symmetric/asymmetric or reflexive/irreflexive. For instance, *cinematographerOf*, *cinematographyBy*, *directedBy*, *directorOf*, *distributedBy*, *distributorOf*, *filmedIn*, *hasActor*, *playedIn*, *hasMovieLanguage*, *movieLanguageOf*, *producedBy*, *producerOf*, *writtenBy*, *writerOf*, *hasMusicBy* and *wroteMusicFor* are asymmetric because they are

not bi-directional (for instance, if a person is a cinematographer of a movie, the movie itself cannot be a cinematographer of the cinematographer), and irreflexive (for instance, director of a movie cannot be a director of him/herself). On the other hand, *landboundary* is symmetric and irreflexive property, as if a country has a land boundary with another country, then this relation holds in the opposite direction too, and because a country cannot have a land boundary with itself. Additionally, in order to produce a richer ontology, most of the asymmetric properties were defined in pairs, one being an inverse property of another – for example, *cinematographerOf* and *cinematographyBy*. Finally, every object property was given a domain and range restrictions to limit, correspondingly, the values that its subjects and objects can take. For example, the domain of the object property *directorOf* is the *Director* class and its range is the *Movie* class, whereas the *lanboundary* object property has *Country* as both its domain and range restrictions. However, it should be relatively intuitive from the object properties names which object property relates to which classes.

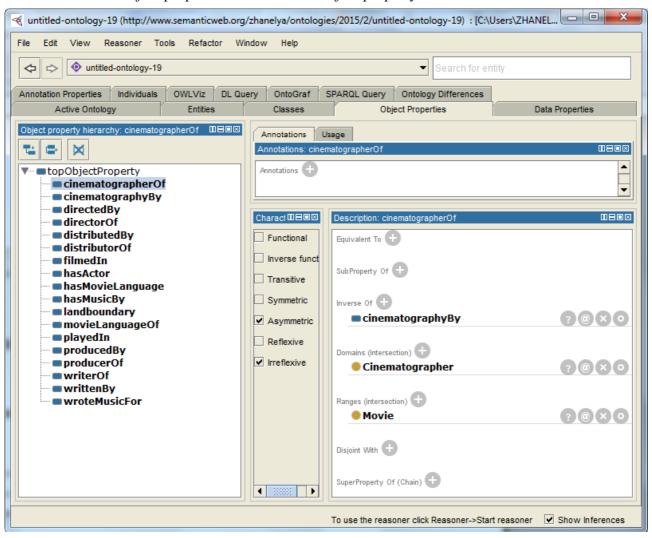
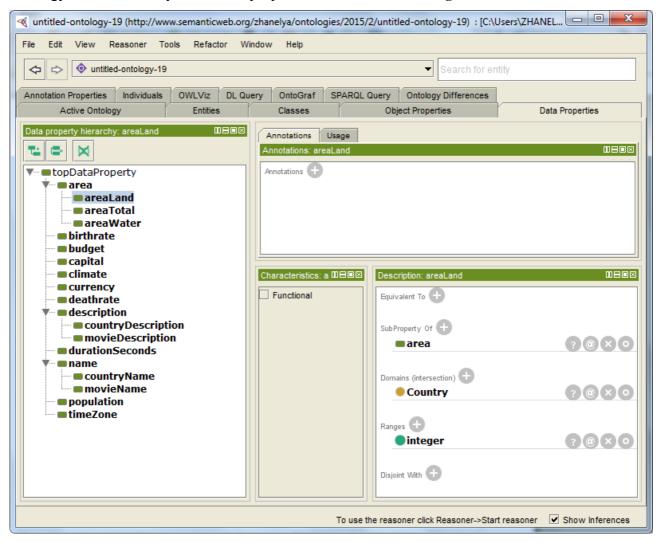


Image 2. Object property hierarchy

#### Data properties

In order to cover atomic properties of movies and countries, I needed to add data properties to my ontology. A full hierarchy of the data properties can be found in *Image 3* below.



*Image 3. Data property hierarchy* 

There are 19 data properties in total defined in the ontology. 12 of them are top level data properties. *Area, birthrate, deathrate, capital, climate, currency, population* and *timeZone* were created to describe countries, whereas *budget* and *durationSeconds* characterise movies. There are also 2 general data properties – *description* and *name*, each having the corresponding sub-properties for movies and countries, namely, *movieDescription* and *countryDescription*, *movieName* and *countryName*. There is also a list of area sub-properties, namely *areaLand*, *areaTotal* and *areaWater*. There are several data properties that are functional, for instance, *capital*, *countryName* and *movieName*, because they can uniquely identify a certain country/movie. Apart from that, every

data property except such general properties as *name* and *description*, has domain and range restrictions, where domain specifies which class of individuals it can refer to, and range specifies the data type of the property. For instance, *areaLand* has a domain restriction of *Country* and a range restriction of *integer* as it is representing the land area of a country in numerical format, whereas *movieName* has domain restriction of *Movie* and range restriction of *string*.

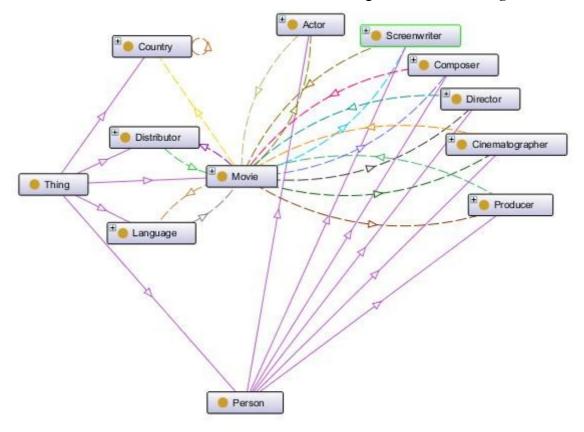


Image 4. Onto Graph ontology representation

A complete ontology represented in a form of an OntoGraph can be seen in *Image 4* above. The links connecting the classes together illustrate the subclass relationship (plain lines) and relations between different concepts (presented as dotted lines on the graph, defined as object properties in the ontology). The main challenge while creating an ontology was to decide which information can be represented as a class and its appropriate object properties and which information should rather be presented in a form of data properties of the existing classes. The assumption that I used at this stage was that basic information that can be stored in a form of literals or numbers (such as country surface area, or movie name) should be represented by data properties, whereas more complex entities such as *Actor* or *Composer*, which would potentially have their own properties, should be

presented as a separate class with the appropriate object properties linking its individuals to some other classes, for example *Movie*. Apart from that, I needed to choose reasonable names for different classes and properties so that they would make sense when the ontology is constructed. For instance, *abstract* property of the movies from DBpedia was renamed into *movieDescription* property within my ontology to make it clearer for the user what kind of information it contains.

### Populating the Ontology from DBpedia

Now, when the ontology skeleton was created, it needed to be populated with external information, as an ontology must be able to integrate and re-use available semantic data. All the concepts of my ontology T-Box were based on the information from DBpedia and Factbook, making it possible to populate its A-Box with already existing data. To populate the ontology from external source I was using SPARQL(1.1). I have chosen DBpedia as my first external ontology, as it contains various information about films. In order to populate the ontology, I produced a python script which can be found in *basic.py*. An outline of the created method will be described below.

First, the appropriate python libraries needed to be imported, such as rdflib, graph and SPARQL to enable rdf parsing, graph processing, and SPARQL queries. Logging module was also imported and configured to ease debugging process of the code.

```
import logging
import rdflib
from rdflib.graph import Graph, URIRef
from SPARQLWrapper import SPARQLWrapper, RDF
from rdflib.plugins.memory import IOMemory
logging.basicConfig()
```

Then, the SPARQL endpoint needed to be configured. For this purpose I was using DBpedia endpoint, which can be accessed through the http://dbpedia.org/sparql link.

```
...
sparql = SPARQLWrapper("http://dbpedia.org/sparql")
```

Query construction is one of the main parts of the ontology population. First, the prefixes were defined, which serve as short-cuts for complete ontology URLs, both from local and external knowledge bases. Then, within the CONSTRUCT statement, a list of movie properties that needed

to be populated was provided together with the location where they should be stored in my local ontology, for instance, the name of the movie was stored under the *movieName* property. Movie attributes that have to be stored as separate class individuals, additionally needed a reference to the name of the class from the local ontology, for instance *filmedIn* has a reference to the *Country* class, as *filmedIn* represents a relation between *Movie* and *Country* classes.

After that, within the WHERE statement, the corresponding properties from the DBpedia ontology were to be specified together with how they could be accessed. Some of the properties are optional, as quite often some non-key properties are missing from the external ontology. Finally, the result of the query was filtered to include only movies having their name and description written in English, and the number of results to be returned was limited to address memory usage/performance issues. The constructed query was then assigned to SPARQL endpoint, whose return format was set to RDF.

```
construct_query="""
   PREFIX movie: <a href="http://www.semanticweb.org/zhanelya/ontologies/2015/2/untitled-ontology-19#">http://www.semanticweb.org/zhanelya/ontologies/2015/2/untitled-ontology-19#</a>
   PREFIX dbpprop: <a href="http://dbpedia.org/property/">http://dbpedia.org/property/</a>
   CONSTRUCT {
   ?film rdf:type movie:Movie.
   ?film movie:movieName ?name .
   ?film movie:filmedIn ?country .
   ?country rdf:type movie:Country .
    WHERE{
    ?film rdf:type dbpedia-owl:Film .
    ?film foaf:name ?name .
    OPTIONAL {?film dbpedia-owl:country}
    FILTER (LANG(?name)="en")
    FILTER (LANG(?abstract)="en")
   LIMIT 100000 """
sparql.setQuery(construct\_query)
sparql.set Return Format(RDF) \\
```

Finally, a graph was created and configured. The results retrieved from the endpoint were passed to this graph, and then merged with the local ontology, producing a new local ontology – result basic.owl, serialised in XML format.

```
memory_store=IOMemory()
graph_id=URIRef("http://www.semanticweb.org/store/movie")
g = Graph(store=memory_store, identifier=graph_id)
rdflib.plugin.register('sparql', rdflib.query.Processor, 'rdfextras.sparql.processor', 'Processor')
rdflib.plugin.register('sparql', rdflib.query.Result, 'rdfextras.sparql.query', 'SPARQLQueryResult')
g = sparql.query().convert()
g.parse("ontology.owl")
g.serialize("result_basic.owl", "xml")
```

The complete ontology populated with movies information can be seen in *Image 5*. Each movie now has the corresponding information retrieved from external ontology, such as movie name, cinematographer, country it was filmed in, etc.

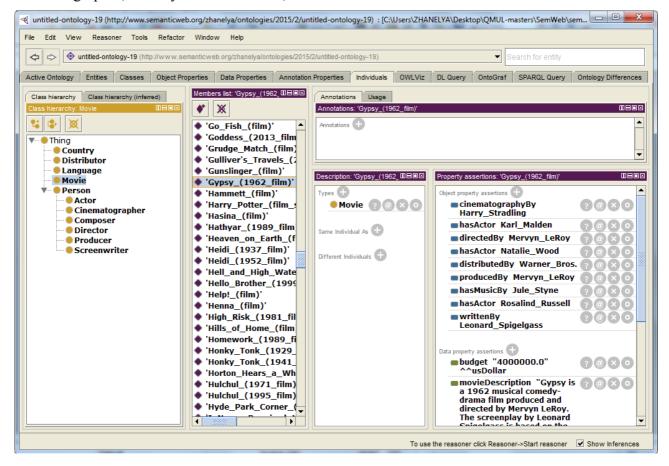


Image5. Populated movies ontology

However, *Movie* class is not the only class that had new individuals added from DBpedia. Other classes have also obtained new instances during the ontology population. However, each such individual does not have its own properties, as the information contained in these classes is not the main purpose of this ontology, and it was added only as these individuals are connected with some Movie instances. Examples of such complementary individuals can be seen in *Image 6* which illustrates part of the instances list for *Country* and *Distributor* classes.

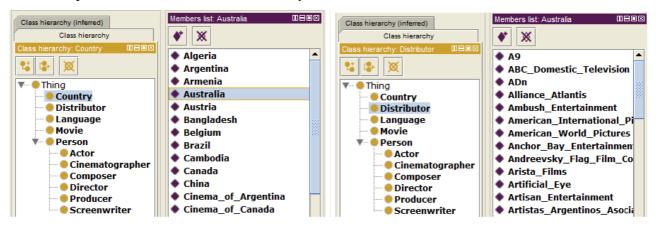


Image 6. Complementary individuals populated from DBpedia

One of the principal challenges when populating the ontology was matching the concepts and properties from the local store with the DBpedia concepts and properties in order to store information appropriately. In order to accomplish that, it was important to understand clearly what information is represented by which names within the external ontology and at which address/location it can be accessed. Apart from that, the results were limited just to those represented in English, as the ontology end user in my case will most likely prefer the information to be presented in English.

## Querying the Ontology populated from DBpedia

In order to test the ontology implementation, I produced a python script containing two queries for my local ontology – *query\_basic.py*. After the appropriate libraries were imported and configured (logging and rdflib), and after the graph was created by parsing the local ontology resulting from populating my ontology with DBpedia (described above), I defined two queries for my ontology. Query1's main purpose is to return movies filmed in United Kingdom. To achieve that, after defining the prefixes for querying my local store, I selected distinct movie names and the

corresponding countries that they were filmed in, limiting the results only to those, where the country name would contain the string "United Kingdom".

```
query1 = """

PREFIX movie: <a href="http://www.semanticweb.org/zhanelya/ontologies/2015/2/untitled-ontology-19#>

PREFIX rdf: <a href="http://www.w3.org/1999/02/22-rdf-syntax-ns#">http://www.w3.org/1999/02/22-rdf-syntax-ns#>

SELECT DISTINCT ?name ?country

WHERE { ?film rdf:type movie:Movie .
    ?film movie:movieName ?name .
    ?film movie:filmedIn ?country .

FILTER regex(str(?country), "United_Kingdom")
    }

"""

results = g.query(query1)
```

Query2 is aimed to retrieve movies that last less than 100 seconds. First, the appropriate prefixes were defined. Then, distinct names and their corresponding duration in seconds were selected. Finally, the results were limited to only those movies that have duration shorter than 100 seconds.

After the queries were constructed, I needed to display the results back to the user in clear

formatting. So, for the first query, I displayed the 2 resulting columns – Name and Country, using an I/O formatting in order to set the column width to 30 and 40 accordingly. The same way were formatted the results returned from the second query, but the column names and width were set differently.

```
print ('{0:30s} {1:40s}'.format("Name","Country"))

print ('{0:30s} {1:40s}'.format("------","-----"))

for name,country in results:

    print ('{0:30s} {1:40s}'.format(name,country))

...

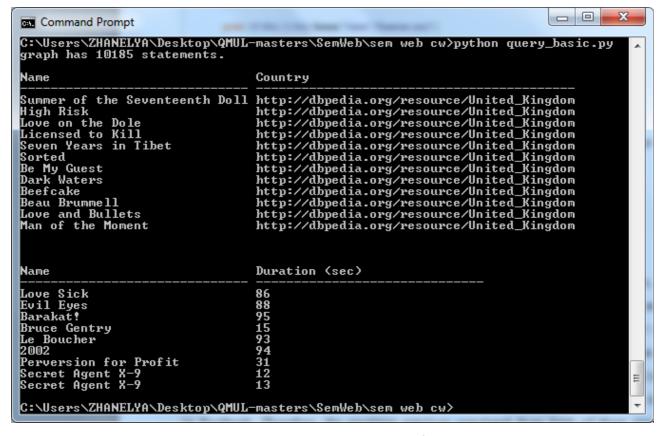
print ('{0:30s} {1:30s}'.format("Name","Duration (see)"))

print ('{0:30s} {1:30s}'.format("-----","----"))

for name,duration in results:

    print ('{0:30s} {1:30s}'.format(name,duration))
```

The output returned from this script querying the local ontology can be seen in *Image 7* below.



*Image 7. Querying movies ontology* 

## Populating the Ontology from Factbook

In order to enrich the ontology already populated with information about movies, and fulfil the requirements set by the bonus task, I had to populate the ontology from the second external semantic data repository. In order to achieve this I chose the Factbook ontology as my second source, because it is fully concentrated on information about countries and thus has more detailed data about each country instance. For example, such country information as birth rate (total number of live births per 1000 citizens) and land boundary cannot be retrieved from DBpedia, but it can be found in Factbook. Therefore, the resulting ontology populated from both of them would be able to answer such questions as "Which movies were filmed in the countries with birthrate between 15 and 20?", and "Which movies were filmed in countries bordering with Kazakhstan?", which would not be possible with either remote knowledge base alone. The python script used for this task – bonus.py – is briefly described below.

After the appropriate python libraries, such as rdflib, graph, SPARQL and logging, were imported and configured, the Factbook SPARQL endpoint was defined:

```
sparql = SPARQLW rapper ("http://wifo5-04.informatik.uni-mannheim.de/factbook/sparql") \\
```

Then, I constructed the query to populate the local ontology from the external Factbook source. First, the prefixes were identified, both for local and remote ontologies. Next, inside the CONSTRUCT statement, a set of country properties to be populated, such as *countryName* and *capital*, were listed and their location within the local storage was identified. One of the most interesting attributes of the countries data set is the land boundary property, as it needs to have a reference to the Country class itself. Within the WHERE clause, the name and location of each country property to be populated was identified within the remote Factbook knowledge base in order to make sure that the data is stored in accordance with its remote structure. As this knowledge base seems to be reasonably stable, none of the properties were set to be optional. Finally, the query was assigned to SPARQL endpoint, and the return type was set to RDF.

```
construct_query="""

PREFIX mc: <a href="http://www.semanticweb.org/zhanelya/ontologies/2015/2/untitled-ontology-19#>
```

```
PREFIX factbook: <a href="http://wifo5-04.informatik.uni-mannheim.de/factbook/ns#">http://wifo5-04.informatik.uni-mannheim.de/factbook/ns#</a>
CONSTRUCT {

?country rdf:type mc:Country .

?country mc:landboundary ?landboundary .

?landboundary rdf:type mc:Country .

}

WHERE {

?country rdf:type factbook:Country .

?country factbook:name ?name .

...

?country factbook:landboundary ?landboundary .

}"""

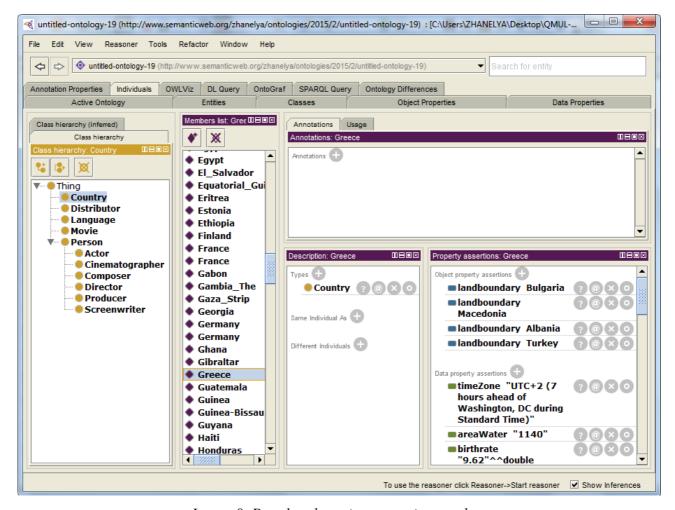
sparql.setQuery(construct_query)

sparql.setReturnFormat(RDF)
```

In the end, a graph was defined and configured. The results obtained from the endpoint were assigned to this graph, which was then merged with the local ontology already populated with movies, resulting in a relatively complete local knowledge base about movies and countries, which can be found in *result bonus.owl*, serialised in XML format.

```
memory_store=IOMemory()
graph_id=URIRef('http://www.semanticweb.org/store/movie_country')
g = Graph(store=memory_store, identifier=graph_id)
rdflib.plugin.register('sparql', rdflib.query.Processor, 'rdfextras.sparql.processor', 'Processor')
rdflib.plugin.register('sparql', rdflib.query.Result, 'rdfextras.sparql.query', 'SPARQLQueryResult')
g = sparql.query().convert()
g.parse("result_basic.owl")
g.serialize("result_bonus.owl", "xml")
```

The resulting ontology populated from the two remote ontologies is illustrated in *Image 8*. It now accommodates appropriate data about each country together with the information about movies.



*Image 8. Populated movies-countries ontology* 

It was quite challenging to map the land boundary property from the Factbook remote ontology onto Country class individuals in the local store, as it was supposed to be referring to the Country class itself. However, after assuming that this type of relation does not really need any specific approach apart from defining it as being symmetric, I simply modelled it the same way that I was modelling a similar relation between two distinct classes.

## Querying the Ontology populated from DBpedia and Factbook

Now, when the ontology was complete, I needed to test its implementation. In order to accomplish that, a python script was developed – *query\_bonus.py* – containing four queries for my local ontology. First, the appropriate libraries were imported and configured (logging and rdflib), then, I created a graph by parsing the local ontology populated from two different sources. Finally, 4

queries were defined for my ontology.

Query1 is aimed to retrieve 10 movies filmed in countries with population more than 100M. Once the prefixes were defined for the local store, I selected distinct movie names and the corresponding countries that they were filmed in. In order to merge the information about movies and countries data, I selected the countries that have the same name as the countries where the movies were filmed in: FILTER regex(str(?countryName\_movie), str(?countryName\_country)). Finally, the movies-countries list was limited to those entries that only have movies filmed in countries with population over 100M: FILTER (?population > 1000000000), and the results were limited to 10 entries. The query results were then displayed in the terminal and the column names and dimensions were appropriately formatted. Any of the following queries were displayed in a similar manner.

```
query1="""
PREFIX mc: <a href="http://www.semanticweb.org/zhanelya/ontologies/2015/2/untitled-ontology-19#">PREFIX mc: <a href="http://www.semanticweb.org/zhanelya/ontologies/2015/2/untitled-ontology-19#">http://www.semanticweb.org/zhanelya/ontologies/2015/2/untitled-ontology-19#</a>
PREFIX rdf: <a href="http://www.w3.org/1999/02/22-rdf-syntax-ns#">http://www.w3.org/1999/02/22-rdf-syntax-ns#</a>
PREFIX xsd: <a href="http://www.w3.org/2001/XMLSchema#">http://www.w3.org/2001/XMLSchema#</a>
SELECT DISTINCT ?movieName ?countryName_country ?population
WHERE { ?film rdf:type mc:Movie .
     ?film mc:movieName ?movieName .
     ?film mc:filmedIn ?countryName_movie .
     ?country rdf:type mc:Country .
     ?country mc:countryName ?countryName_country .
     ?country mc:population ?population .
     FILTER regex(str(?countryName movie), str(?countryName country))
     FILTER (?population > 100000000)
     }LIMIT 10
print ('{0:40s} {1:20s} {2:10s}'.format("Title","Filmed in","Population"))
print ('{0:40s} {1:20s} {2:10s}'.format("-----","-----","-----"))
for x,y,z in g.query(query1):
  print ('{0:40s} {1:20s} {2:10s}'.format(x.encode('utf-8'),y.encode('utf-8'),z))
```

Query2's main purpose is to return movies filmed in countries that have a capital "Buenos Aires",

i.e. in Argentina. After the appropriate prefixes were defined, I selected distinct movie names and the corresponding countries they were filmed in. Next, the countries information was merged with movies information by matching the country name with the country that each movie was filmed in. Finally, the results were limited to those that only have movies filmed in countries with a capital "Buenos Aires".

Query3 was developed to retrieve movies filmed in countries with birthrate between 15 and 20. First, the appropriate prefixes were defined, then distinct movie names and the countries they were filmed in were selected. Next, movies and countries data was merged together and the results were limited to movies filmed in only those countries having their birthrate in a specified range.

```
query3="""

PREFIX mc: <a href="http://www.semanticweb.org/zhanelya/ontologies/2015/2/untitled-ontology-19#">http://www.w3.org/1999/02/22-rdf-syntax-ns#</a>

PREFIX rdf: <a href="http://www.w3.org/1999/02/22-rdf-syntax-ns#">http://www.w3.org/1999/02/22-rdf-syntax-ns#</a>

PREFIX xsd: <a href="http://www.w3.org/2001/XMLSchema#">http://www.w3.org/2001/XMLSchema#</a>

SELECT DISTINCT ?movieName ?countryName_country ?birthrate

WHERE { ?film rdf:type mc:Movie .

?film mc:movieName ?movieName .
```

```
?film mc:filmedIn ?countryName_movie .

?country rdf:type mc:Country .

?country mc:countryName ?countryName_country .

?country mc:birthrate ?birthrate .

FILTER regex(str(?countryName_movie), str(?countryName_country))

FILTER (?birthrate > 10.00 && ?birthrate < 20.00)

}

"""
```

The last query, Query4 was designed to return movies filmed in countries having common land boundary with Kazakhstan. In a similar manner, after prefixes were defined and distinct movie names and corresponding countries were selected, the data from two different ontologies stored now in a local store were merged together. Finally, the results were limited to those movies that were filmed in only those countries that are neighbouring with Kazakhstan.

```
query4="""

PREFIX mc: <a href="http://www.semanticweb.org/zhanelya/ontologies/2015/2/untitled-ontology-19#">http://www.semanticweb.org/zhanelya/ontologies/2015/2/untitled-ontology-19#</a>

PREFIX rdf: <a href="http://www.w3.org/1999/02/22-rdf-syntax-ns#">http://www.w3.org/1999/02/22-rdf-syntax-ns#</a>

PREFIX xsd: <a href="http://www.w3.org/2001/XMLSchema#">http://www.w3.org/2001/XMLSchema#</a>

SELECT DISTINCT ?movieName ?countryName_country ?landboundary

WHERE { ?film rdf:type mc:Movie .
    ?film mc:filmedIn ?countryName_movie .
    ?country rdf:type mc:Country .
    ?country mc:countryName ?countryName_country .
    ?country mc:landboundary ?landboundary .

FILTER regex(str(?countryName_movie), str(?countryName_country))

FILTER regex(str(?landboundary), "Kazakhstan")
}

"""
```

The output returned from this script querying the local ontology can be seen in *Image 9* below.

:\Users\ZHANELYA\Desktop\QMUL-mast	teus\Sembleb\sem ueb e	Southon guesa bonus su
raph has 12649 statements.	cers/semweb/sem web ci	//python query_bonus.py
itle 	Filmed in	Population
ayak	Russia	141377752
uerto Escondido	Mexico	108700891
nnapurna	India	1129866154
namika	India	1129866154
ango ogan	Brazil India	190010647 1129866154
uerto Escondido	Brazil	190010647
a Bruja	Mexico	108700891
ari Nari Naduma Murari	India	1129866154
undarakanda	India	1129866154
itle	Filmed in	Capital
even Years in Tibet	Argentina	Buenos Aires
a Ci únaga	Argentina	Buenos Aires
itle	Filmed in	Birthrate
even Years in Tibet	Argentina	16.53E0
a Ci hinaga	Argentina	16.53E0
omework	Iran	16.57E0
arakat!	Algeria	17.11E0
abies	Israel	17.71E0
utside the Law uerto Escondido	Algeria Brazil	17.11E0 16.3E0
uerto Esconalao osta	Brazii Lebanon	18.08E0
ango	Brazil	16.3E0
eysar	Iran	16.57E0
antouri	Iran	16.57E0
emi cinta belahlah dadaku	Indonesia	19.65E0
utside the Law	Tunisia	15.54E0
es Ambassadeurs	Tunisia	15.54E0
itle File	med in Neighbouring (	country
ayak Rus:	sia http://wifo5-(	 D4.informatik.uni-mannhei
.de/factbook/resource/Kazakhstan rossed Lines Chii .de/factbook/resource/Kazakhstan	na http://wifo5-0	04.informatik.uni-mannhei
.ue/factbook/resource/kazakhstan ream of the Red Chamber Chii .de/factbook/resource/Kazakhstan	na http://wifo5-0	04.informatik.uni-mannhei
ark Waters Rus: .de/factbook/resource/Kazakhstan	sia http://wifo5-0	04.informatik.uni-mannhei

Image 9. Querying movies-countries ontology

One of the main challenges at this step was to match the information about two different concepts in appropriate manner. To accomplish that, a common property was found for these concepts, namely

the country name. Most *Movie* individuals have the *filmedIn* property which contains the country reference. All the *Country* individuals have the *countryName* property storing the country name. By matching these two properties it became possible to merge the data populated into ontology from two different sources. Another difficulty arose from the windows terminal specificity, as it cannot properly display non-utf8 symbols, for instance, symbols from languages other than English. To solve this issue, I was using y.encode('utf-8') for those entries that could potentially cause the problem when formatting the results and displaying them to the screen.

### **List of Files**

#	File Name	Description
1	SemanticWeb_CW_Zhanelya_Subebayeva.	Main report describing the process of building,
	pdf	populating and querying the ontology
2	ontology.owl	An initial Protégé-OWL ontology built
3	catalog-v001.xml	Catalogue generated automatically by Protégé for
		the ontology built
4	result_basic.owl	The ontology resulting from populating the original
		ontology from the first source (DBpedia)
5	result_bonus.owl	The resulting ontology after populating the
		ontology from the second source (Factbook) and
		combining it with the ontology populated from
		DBpedia
6	basic.py	The python script used to populate the ontology
		from the first SPARQL endpoint (DBpedia)
7	bonus.py	The python script used to populate the ontology
		from the second SPARQL endpoint (Factbook)
8	query_basic.py	The python script to query the local store, which is
		the ontology populated from the first SPARQL
		endpoint (result_basic.owl)
9	query_bonus.py	The python script to query the local store, which is
		the ontology populated from both SPARQL
		endpoints (result_bonus.owl)

#### How to run the code

I was running the code under the current settings of my machine, which is running Windows 7.

When building the ontology I was using Protégé version 5.0.0 and Protégé default reasoner – HermiT 1.3.8.3. The python version that I was using for populating and querying my ontology is Python 2.7.5. There were some minor issues with Pyparsing version conflict, which was resolved by installing a specific version of Pyparsing – Pyparsing 1.5.7.

To run the python code, I used Windows Command Prompt – for instance entering *python basic.py* into the terminal populates the ontology stored in *ontology.owl* with movies information from the first source (Dbpedia), storing the resulting ontology in *result\_basic.owl*; *python bonus.py* populates it from the second source (Factbook) storing the outcome in *result\_bonus.owl*. Using *python query\_basic.py* and *python query\_bonus.py* will return the query results from querying the basic (populated from first source) and bonus (populated from 2 sources) ontologies correspondingly and display them on the terminal screen.

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