

Project Assignment 1

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1. Introduction

Knowledge graphs are information represented in the form of graphs where nodes are entities and the relationships between them are the edges. The fact that they help form better connections between data helps machines to process and connect loads of information for better results. Knowledge graphs are the foundation for artificial intelligence and are the main tool that is required to build intelligent agents such as chatbots that help answer questions based on the user's input. Each fact in a knowledge graph is represented as a triple (subject, predicate, object) which are interconnected with each other to form the whole graph.

In this project, we create a knowledge graph that stores the information about Universities and its academic details. Here, we concentrate on information about Concordia University alone. The goal of this project is to build a knowledge graph in such a way that it can answer questions about the University through SPARQL queries. This project is the first step of building an intelligent agent that can reply to questions related to the university which would be further done in our next project. Knowledge graphs are created from the data extracted from the web. Here, we take information about the courses, topics related to it and further details regarding it from the Concordia University websites and its open databases.

We have constructed the University knowledge graph using RDF and RDFS standard in turtle format. The graph is built to store information about universities along with its DBpedia entry URI. But since we restrict our project to just one university, the graph stores information about all the courses offered by Concordia University and the topics that are covered in each course. It also stores student information such as name, email etc. For testing purposes, we have used dummy information for ten students and their course history.

2. Schema Description

2.1 Base University Knowledge Graph Schema

The University graph is stored in the form of RDF schema. It contains five classes (University, Course, Topic, Student and CourseGrade) and four properties (coversCourse, hasPart, studiesAt and tookCourse). These are stored in the form of the classes. Each of these components has attributes that store more information about them. Each component has a Label and Comment for better understanding.

2.1.1 University (Class)

This contains the name and the DBpedia URI for the given university. Since we use only Concordia University in our project, this has only once instance.

2.1.2 Course (Class)

This contains the name of the course, course subject (like COMP), course number (like 691), the description for the course and additional link from which the course information is taken.

2.1.3 Topic (Class)

This class refers to the topics extracted from course name and description. This contains the name of the topic and the DBpedia URI entry for the topic. Additional, analysis is done to analyze the URI extracted.

2.1.4 Student (Class)

This contains the name of the student; first name and last name, ID number and email address of the student.

2.1.5 Course Grade (Class)

This contains a course class instance along with the grade that the student scored for that course.

2.1.6 coversCourse (Property)

This property links the courses to the university with courses in range and university in domain.

2.1.7 hasPart (Property)

This property links a course with all the topics related to it with course in domain and topics in range.

2.1.8 studiesAt (Property)

This property links a student with the university with student in domain and university in range.

2.1.9 tookCourse (Property)

This property links a student with the courses taken with the grade attained by the student with student in domain and CourseGrade Class in range.

The RDF Schema for the University graph is stored in Turtle format, in the files “universityKG.ttl”, the basic description of the base schema (contains classes and properties listed above) and “DataGraph.ttl” with all the data extracted and linked using triplets with the base schema.

2.2 Vocabularies Used in Schema

2.2.1 Reused Vocabularies

2.2.1.1 RDFS Schema

rdfs: <http://www.w3.org/2000/01/rdf-schema#>

rdfs:label : Used to provide a human readable name for all the entities in our KG.

rdfs:comment : Used to provide a additional comment all the entities in our KG.

rdfs:domain : Used to define the Domain of the property.

rdfs:range : Used to define the Range of the property.

rdfs:Class : Used to define a class.

rdfs:seeAlso : Used to provide additional links for course.

rdfs:subClassOf : Used to provide extension of Person Class to the Student Class.

2.2.1.2 RDF Schema

rdf: <http://www.w3.org/1999/02/22-rdf-syntax-ns#>

rdf:Property : Used to define a property.

2.2.1.3 XML Schema

xsd: <http://www.w3.org/2001/XMLSchema#>

xsd:string : Used to define a string literal.

xsd:int : Used to define an integer literal.

2.2.1.4 FOAF Vocabulary

foaf: <http://xmlns.com/foaf/0.1/>

foaf:name : Used to provide names for entities such as course, topic and university.

foaf:givenName : Used to provide first name of the student.

foaf:familyName : Used to provide last name of the student.

foaf:mbox : Used to provide email of the student.

foaf:Person : Used to define Student as sub class of the Person Class.

2.2.1.5 DBpedia Property

dbp: <http://dbpedia.org/property/>

dbp:score : Used to provide the grade for a student.

dbp:id : Used to provide the identity number for a student.

2.2.1.6 DCMI Metadata Terms

dc: <http://purl.org/dc/elements/1.1/>

dc:source : Used to link dbpedia links for University and Topic instances.

dc:subject : Used to connect Course(like COMP 691) for a CourseGrade instance(further to a student) and provide the course subject (like COMP) for a Course instance .

dc:identifier : Used to provide course number for Course instance.

dc:description : Used to provide course description for Course instance.

2.2.2 Our Schema

2.2.2.1 ISP

isp: <http://intelligentsystemproj1.io/schema#>

isp:University : Class for a University.

isp:Course : Class for a Course.

isp:Topic : Class for a Topic.

isp:Student : Class for a student.

isp:CourseGrade : Class for courses with grades for a student.

isp:studiesAt : Property that links the university that the student studies at .

isp:tookCourse : Property that links the Courses completed by a student.

isp:hasPart : Property that links the Topics covered for a Course.

isp:coversCourse : Property that links the Courses covered in a University.

2.3 Schema Description in Image

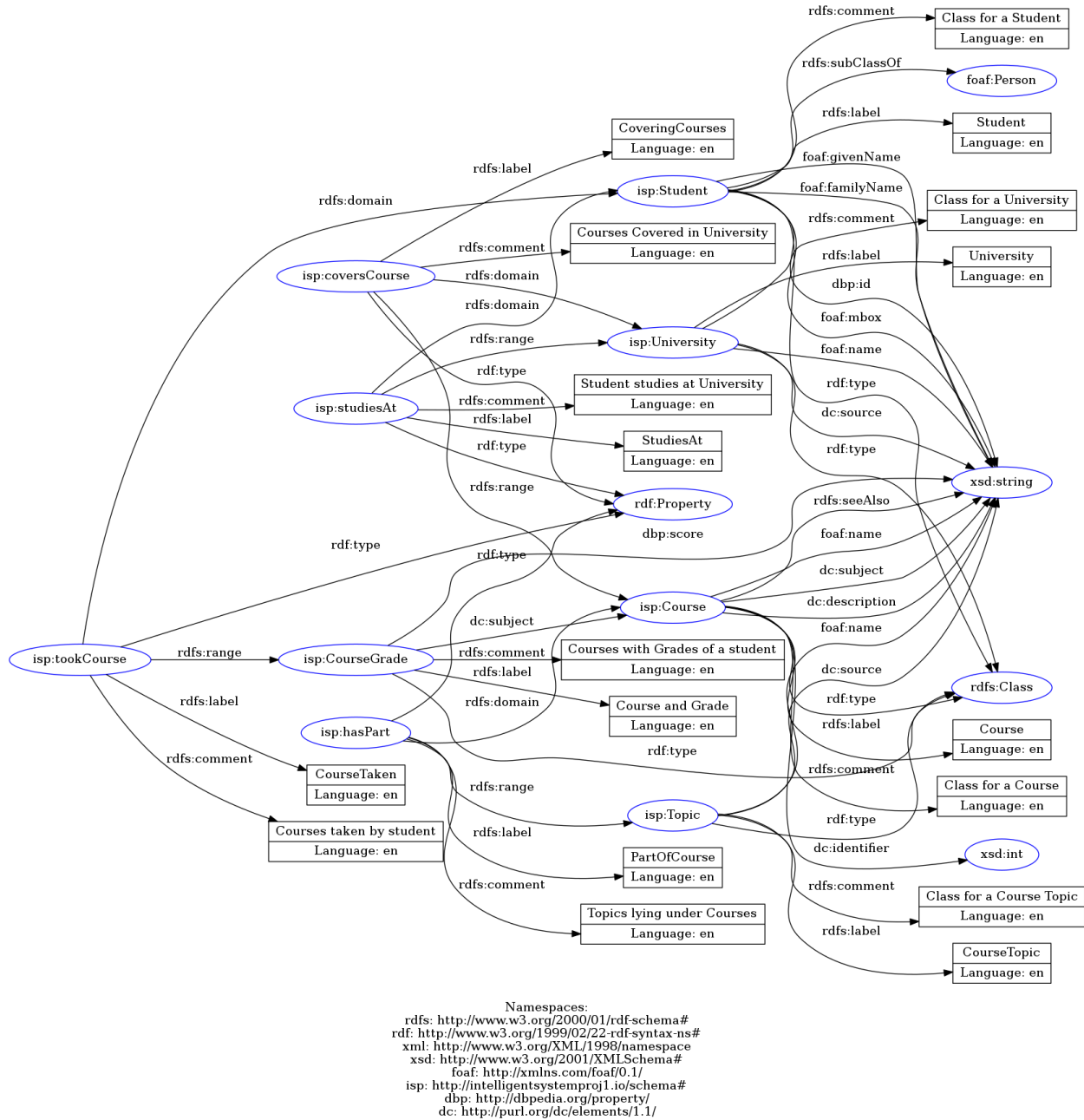


Figure 1: Shows the RDF graph visualisation for the base knowledge graph.

3. Automated Knowledge Base Construction

3.1 Dataset Created

In this step, we have first generated .csv files that stores information about the instances.

Courses.csv stores the data related to courses and its properties.

Grades.csv stores the courses and the grade scores by each student.

Student.csv stores information about students such as name, email, id etc.

Topics.csv stores information and DBpedia links for Topics for each course.

Universities.csv stores information about universities and its DBpedia entries.

The data was scraped from Concordia University Graduate websites. We have scraped the data from over 99 URLs which gave information about the courses that are offered by the University, along with its course code and course description. We have found 2742 graduate courses and 6702 topics offered by Concordia from all the Departments.

3.2 Tools/Libraries Used

Pandas

We use pandas here to process all the data stored in the csv files.

Rdflib

Rdflib is used in our project to create and parse the Knowledge graph.

Spotlight

We use spotlight library to find access the DBpedia spotlight to identify entities from the course name and description to extract the topics with their DBpedia links for a course.

BeautifulSoup

BeautifulSoup library is used for scraping data from the Concordia University URLs to get academic data such as courses provided by the university with the course description, course code and course name.

3.3 Tools Created

We have created 3 tools to build the knowledge graph and 1 tool to run the 6 queries given. All the tools are developed in python.

1LoadCoursesStudentsGrades

This tool is used to create student and course data. It uses pandas and Beautiful Soup. The data for courses such as course name, course number and course description are extracted by doing web scraping of 99 Concordia University URLs with Beautiful Soup. With these links, we were able to attain 2742 graduate courses. The student data includes 15 students' names hard coded. The courses are randomly allotted to the students with various grades. The data extracted is saved using pandas in respective csvs.

2LoadLinkTopics

This tool is used to get the topics with the dbpedia links for a course name and description. It uses pandas and Spotlight. This tool extracts the courses data from courses.csv. The course name and description are joined to make a call to spotlight annotate function. The link "<https://api.dbpedia->

spotlight.org/en/annotate”, 0.5 confidence, support=20 and the course data is sent to get the topics(surfaceFrom) and dbpedia links. Various confidence levels were tried started from 0.3, which was further increased and chose 0.5.

3CreateKnowledgeGraph

This tool is used to extract the data from the course, university, course, grade and topics csvs with the help of pandas library. Further, using rdflib is used to create the knowledge graph which is further saved in “DataGraph.ttl” file.

4RunQueries

This tool is used to run the 6 queries given in the assignment with various inputs taken from console.

3.4 How to use the tools to create knowledge graph

- Run 1LoadCoursesStudentsGrades.py which takes input the “universityKG.ttl” file and saves the data in the respective files in CSV folder.
- Run 2LoadLinkTopics which takes input the Courses.csv and saves the data in respective files in CSV folder.
- Run 3CreateKnowledgeGraph which takes input the “universityKG.ttl” file and the files in CSV folder and saves the triplets in “DataGraph.ttl”.

4. Queries

We use SPARQL queries to ask and get information from the RDF Knowledge graph. We have tested our knowledge graph for various queries. The queries are run on python using rdflib but we have also tested our queries on Apache Jena Fuseki.

Following prefixes are used or all queries:

```
PREFIX dbr: <http://dbpedia.org/resource/>
PREFIX db: <http://dbpedia.org/>
PREFIX is: <http://purl.org/ontology/is/core#>
prefix dbp: <http://dbpedia.org/property/>
prefix dc: <http://purl.org/dc/elements/1.1/>
prefix isp: <http://intelligentsystemproj1.io/schema#>
prefix rdf: <http://www.w3.org/1999/02/22-rdf-syntax-ns#>
prefix rdfs: <http://www.w3.org/2000/01/rdf-schema#>
prefix xsd: <http://www.w3.org/2001/XMLSchema#>
```

4.1 Query 1

The assignment asked to get the total number of triples in knowledge graph in this query. Following query was designed to solve this:

```
SELECT (COUNT(*) as ?triples) WHERE { ?s ?p ?o . }
```

QUERY EXPLAINED

The query iterates all the triples and filter using the statement in the where clause and count them using count function.

RESULT

We were able to attain following result:

Total number of triples: 46254

4.2 Query 2

The assignment asked to get total number of students, courses, and topics. We tried two queries:

a. Following query was tried first. But, because of the huge database, this was taking a lot of time while running in python. Thus, we tried another query listed in b part.

```
SELECT
(COUNT(DISTINCT ?student) as ?scount)
(COUNT(DISTINCT ?course) as ?ccount)
(COUNT(DISTINCT ?topic) as ?tcount)
WHERE{
    ?student rdf:type isp:Student .
    ?course rdf:type isp:Course .
    ?topic rdf:type isp:Topic .
}
```

b. This query was taking very less time comparatively.

```
SELECT ?scount ?ccount ?tcount
WHERE{
    {SELECT
    (COUNT(DISTINCT ?student) as ?scount)
    WHERE { ?student rdf:type isp:Student . } }
    UNION
    {SELECT
    (COUNT(DISTINCT ?course) as ?ccount)
    WHERE { ?course rdf:type isp:Course .} }
    UNION
    {SELECT
    (COUNT(DISTINCT ?topic) as ?tcount)
    WHERE { ?topic rdf:type isp:Topic .} }
}
```

QUERY EXPLAINED

The query iterates all the triples and filter using the statements in the where clause and calculates the number of students, courses and topics. The second query use Union which helps in reducing the time for the query.

RESULT

We were able to attain following results:

Total number of students : 15
Total number of courses : 2742
Total number of topics : 6702

4.3 Query 3

The assignment asked to list all covered topics using their (English) labels and their link to DBpedia for a course c. Following query was designed to solve this:

```
SELECT DISTINCT ?name ?link
WHERE{
  ?course rdf:type isp:Course .
  ?course dc:subject "MAST" .
  ?course dc:identifier 830 .
  ?course isp:hasPart ?topic .
  ?topic dc:source ?link .
  ?topic foaf:name ?name .
}
```

QUERY EXPLAINED

The query iterates all the triples and filter using the statements in the where clause. Here, we match all the nodes of the type `isp:Course` that has given course subject (like MAST) and Course number (like 830), these are the example values. We match those course nodes to the topics covered by that course, given by `isp:hasPart`. We then display the names (`foaf:name`) of these topics and their dbpedia links (given by `dc:source`).

RESULT

We were able to attain following results:

1. For COMP 7251, we got

```
Please enter the Course Subject(like COMP): COMP
Please enter the Course Number(like 691): 7251
```

| TopicName | TopicLink |
|------------------|---|
| wireless | http://dbpedia.org/resource/Wireless_LAN |
| topology | http://dbpedia.org/resource/Network_topology |
| MAN | http://dbpedia.org/resource/Metropolitan_area_network |
| sensor networks | http://dbpedia.org/resource/Wireless_sensor_network |
| PAN | http://dbpedia.org/resource/Personal_area_network |
| mobile computing | http://dbpedia.org/resource/Mobile_computing |

2. For MAST 830, we got

Please enter the Course Subject (like COMP): MAST

Please enter the Course Number (like 691): 830

| TopicName | TopicLink |
|-------------------|---|
| Gauss | http://dbpedia.org/resource/Carl_Friedrich_Gauss |
| class number | http://dbpedia.org/resource/Ideal_class_group |
| L-series | http://dbpedia.org/resource/L-function |
| Cyclotomic Fields | http://dbpedia.org/resource/Cyclotomic_field |
| theorem | http://dbpedia.org/resource/Theorem |

4.4 Query 4

The assignment asked to list all courses this student completed, together with the grade for a given student. Following query was designed to solve this:

```
SELECT ?courseSub ?courseNum ?courseName ?grade
WHERE{
  ?student rdf:type isp:Student .
  ?student dbp:id "40083902" .
  ?student isp:tookCourse ?courseGrade .
  ?courseGrade dbp:score ?grade .
  ?courseGrade dc:subject ?course .
  ?course foaf:name ?courseName.
  ?course dc:subject ?courseSub .
  ?course dc:identifier ?courseNum .
}
```

QUERY EXPLAINED

The query iterates all the triples and filter using the statements in the where clause. We start by matching all the nodes of the type isp:student and has "40083902" (example instance) as id (dbp:id). We then take the courses with the grades scored by the student by using isp:tookCourse. Dbp:score gives the grade and dc:subject gives the course whose name is extracted using foaf:name, subject using dc:subject and course number using dc:identifier. The details of the course with grade is outputted.

RESULT

We were able to attain following result:

1. For Student id 40083902, we got:

Please enter the Student id: 40083902

| CourseSubject | CourseNumber | CourseName | Grade |
|---------------|--------------|---|-------|
| COMP | 6321 | Machine Learning (4 credits) | F |
| COMP | 6341 | Computer Vision (*) (4 credits) | A+ |
| COMP | 6331 | Advanced Game Development (*) (4 credits) | B+ |

- For Student id 40083898, we got:

Please enter the Student id: 40083898

| CourseSubject | CourseNumber | CourseName | Grade |
|---------------|--------------|--|-------|
| COMP | 7251 | Mobile Computing and Wireless Networks (4 credits) | A+ |
| COMP | 6281 | Parallel Programming (*) (4 credits) | F |
| COMP | 7241 | Parallel Algorithms and Architectures (4 credits) | B+ |

4.4 Query 5

The assignment asked to list all students that are familiar with the topic (i.e., took, and did not fail, a course that covered the topic) for a given topic. Following query was designed to solve this:

```
SELECT DISTINCT ?id (CONCAT(?firstName, " ", ?lastName) as ?name)
WHERE{
?student rdf:type isp:Student .
?student dbp:id ?id .
?student foaf:givenName ?firstName .
?student foaf:familyName ?lastName .
?student isp:tookCourse ?courseGrade.
?courseGrade dbp:score ?grade .
?courseGrade dc:subject ?course .
?course isp:hasPart ?topic .
?topic foaf:name "Game engine" .
FILTER(?grade < "F") }
```

QUERY EXPLAINED

The query iterates all the triples and filter using the statements in the where clause. We take all the nodes related to a topic identified by a topics name (foaf:name), say for example, "CORBA" by matching all the Course node (isp:tookCourse) taken by a student (isp:student) which has a course with topic(isp:hasPart) named "CORBA". We then check if the student has passed in this course by filtering out the fail grades using FILTER. This says that the student is indeed familiar with this topic.

RESULT

We were able to attain following result:

- For "Game engine", we got:

Please enter the topic name: Game engine

| StudentID | StudentName |
|-----------|-----------------|
| 40083901 | Mary Smith |
| 40083902 | Maria Hernandez |

- For "Computer Science", we got:

Please enter the topic name: Computer Science

| StudentID | StudentName |
|-----------|-------------|
| 40083895 | James Smith |

4.4 Query 6

The assignment asked to list all topics (no duplicates) that this student is familiar with (based on the completed courses for this student that are better than an “F” grade) for a given student. Following query was designed to solve this:

```
SELECT DISTINCT ?tName
WHERE{
  ?student rdf:type isp:Student .
  ?student dbp:id "40083895" .
  ?student isp:tookCourse ?courseGrade.
  ?courseGrade dbp:score ?grade .
  ?courseGrade dc:subject ?course .
  ?course isp:hasPart ?topic .
  ?topic foaf:name ?tName .
  FILTER(?grade < "F") .
}
```

QUERY EXPLAINED

The query iterates all the triples and filter using the statements in the where clause. We find the student nodes linked to a specific student id (dbp:id or we can use any other identification attribute such as foaf:name or foaf:mbox), say “40083895”. We then retrieve the courses taken by the student using isp:tookCourse and through that we take the topics covered under it using dc:subject. We then check if the grade(dbp:score) for those courses are better than F by using FILTER.

RESULT

We were able to attain following result:

1. For student id 40083895,

Please enter the Student id: 40083895

```
TopicName
-----
server
fault tolerance
concurrency
remote procedure call
distributed computing
scalability
CORBA
concurrency control
Client-server
communication
fault-tolerant
interprocess communication
distributed systems
COMP
Computer Science
```

2. For student id 40083902,

Please enter the Student id: 40083902

TopicName

computer

3D

Artificial Intelligence

AI

mobile gaming

Game engine

collision detection

pathfinding

OpenCV

COMP

perceptual organization

Computer Vision

5. Tests

Link analysis

We have used link analysis to calculate the accuracy for DBpedia spotlight entity detection to find the URI entry in DBpedia for the topics that are from the course name and description text. We have tested for any ambiguity or errors in the entities identified from 100 links detected by python spotlight and provided correct URL in case of wrong URIs. The data for link analysis is present in "LinkAnalysis.csv" file. The analysis showed 96% success.

| | Course Su | Course Nu | Topic | URI Generated | Correct/Not Correct | Correct URI |
|-----|-----------|-----------|--------------------------------|---|---------------------|--|
| 1 | COMP | 691 | Computer Science | http://dbpedia.org/resource/Computer_science | Correct | |
| 2 | SOEN | 6011 | Version control | http://dbpedia.org/resource/Version_control | Correct | |
| 3 | SOEN | 6011 | code review | http://dbpedia.org/resource/Code_review | Correct | |
| 4 | SOEN | 6011 | Continuous integration | http://dbpedia.org/resource/Continuous_integration | Correct | |
| 5 | SOEN | 6021 | software metrics | http://dbpedia.org/resource/Software_metric | Correct | |
| 6 | COMP | 6231 | distributed computing | http://dbpedia.org/resource/Distributed_computing | Correct | |
| 7 | COMP | 6231 | scalability | http://dbpedia.org/resource/Scalability | Correct | |
| 8 | COMP | 6231 | concurrency | http://dbpedia.org/resource/Concurrent_computing | Correct | |
| 9 | COMP | 6231 | fault tolerance | http://dbpedia.org/resource/Fault_tolerance | Correct | |
| 10 | COMP | 6231 | server | http://dbpedia.org/resource/Web_server | Incorrect | https://en.wikipedia.org/wiki/Server_(computing) |
| 11 | COMP | 6281 | heterogeneous computing | http://dbpedia.org/resource/Heterogeneous_computing | Correct | |
| 12 | COMP | 6281 | Parallel programming | http://dbpedia.org/resource/Parallel_computing | Correct | |
| 13 | SOEN | 6311 | software life-cycle | http://dbpedia.org/resource/Software_release_life_cycle | Correct | |
| 14 | SOEN | 6311 | Software development | http://dbpedia.org/resource/Software_development | Correct | |
| 15 | COMP | 6321 | logistic regression | http://dbpedia.org/resource/Logistic_regression | Correct | |
| 16 | COMP | 6321 | neural networks | http://dbpedia.org/resource/Artificial_neural_network | Correct | |
| 17 | COMP | 6321 | Unsupervised learning | http://dbpedia.org/resource/Unsupervised_learning | Correct | |
| 18 | COMP | 6321 | k-means | http://dbpedia.org/resource/K-means_clustering | Correct | |
| 19 | COMP | 6321 | Reinforcement learning | http://dbpedia.org/resource/Reinforcement_learning | Correct | |
| 20 | COMP | 6331 | Artificial Intelligence | http://dbpedia.org/resource/Artificial_intelligence | Correct | |
| 21 | COMP | 6331 | AI | http://dbpedia.org/resource/Artificial_intelligence | Correct | |
| 22 | COMP | 6341 | Computer Vision | http://dbpedia.org/resource/Computer_vision | Correct | |
| 23 | COMP | 6341 | OpenCV | http://dbpedia.org/resource/OpenCV | Correct | |
| 24 | COMP | 6351 | partial differential equations | http://dbpedia.org/resource/Partial_differential_equation | Correct | |
| 25 | COMP | 6361 | Numerical Analysis | http://dbpedia.org/resource/Numerical_analysis | Correct | |
| 26 | COMP | 6381 | reverse engineering | http://dbpedia.org/resource/Reverse_engineering | Correct | |
| 27 | SOEN | 6431 | software maintenance | http://dbpedia.org/resource/Software_maintenance | Correct | |
| 28 | SOEN | 6441 | unit tests | http://dbpedia.org/resource/Unit_testing | Correct | |
| 29 | SOEN | 6441 | multi-threading | http://dbpedia.org/resource/Thread_(computing) | Incorrect | https://en.wikipedia.org/wiki/Multithreading_(computer_architecture) |
| 30 | SOEN | 6441 | code reuse | http://dbpedia.org/resource/Code_reuse | Correct | |
| 31 | COMP | 6461 | multicasting | http://dbpedia.org/resource/Multicast | Correct | |
| 32 | COMP | 6461 | UDP | http://dbpedia.org/resource/User_Datagram_Protocol | Correct | |
| 33 | COMP | 6461 | Network security | http://dbpedia.org/resource/Network_security | Correct | |
| 34 | SOEN | 6461 | object-oriented | http://dbpedia.org/resource/Object-oriented_design | Correct | |
| 35 | SOEN | 6461 | design patterns | http://dbpedia.org/resource/Software_design_pattern | Correct | |
| 36 | SOEN | 6471 | cloud computing | http://dbpedia.org/resource/Cloud_computing | Correct | |
| 37 | SOEN | 6481 | requirements engineering | http://dbpedia.org/resource/Requirements_engineering | Correct | |
| 38 | SOEN | 6481 | formal languages | http://dbpedia.org/resource/Formal_language | Correct | |
| 39 | SOEN | 6481 | quality assurance | http://dbpedia.org/resource/Quality_assurance | Correct | |
| 40 | SOEN | 6481 | UML | http://dbpedia.org/resource/Unified_Modeling_Language | Correct | |
| 41 | COMP | 6521 | optimization | http://dbpedia.org/resource/Query_optimization | Correct | |
| 42 | COMP | 6521 | Data warehouse | http://dbpedia.org/resource/Data_warehouse | Correct | |
| 43 | COMP | 6521 | Data mining | http://dbpedia.org/resource/Data_mining | Correct | |
| 44 | COMP | 6521 | XML | http://dbpedia.org/resource/XML | Correct | |
| 45 | COMP | 6521 | multimedia | http://dbpedia.org/resource/Multimedia | Correct | |
| 46 | COMP | 6591 | first-order logic | http://dbpedia.org/resource/First-order_logic | Correct | |
| 47 | COMP | 6591 | relational algebra | http://dbpedia.org/resource/Relational_algebra | Correct | |
| 48 | COMP | 6591 | relational calculus | http://dbpedia.org/resource/Relational_calculus | Correct | |
| 49 | COMP | 6641 | Turing machines | http://dbpedia.org/resource/Turing_machine | Correct | |
| 50 | COMP | 6641 | recursive functions | http://dbpedia.org/resource/Computable_function | Incorrect | https://en.wikipedia.org/wiki/Recursion_(computer_science) |
| 51 | COMP | 6651 | algorithm | http://dbpedia.org/resource/Algorithm | Correct | |
| 52 | COMP | 6651 | complexity analysis | http://dbpedia.org/resource/Analysis_of_algorithms | Correct | |
| 53 | COMP | 6651 | NP-complete | http://dbpedia.org/resource/NP-completeness | Correct | |
| 54 | COMP | 6711 | point location | http://dbpedia.org/resource/Point_location | Correct | |
| 55 | COMP | 6711 | robot | http://dbpedia.org/resource/Robotics | Correct | |
| 56 | COMP | 6711 | motion planning | http://dbpedia.org/resource/Motion_planning | Correct | |
| 57 | COMP | 6711 | computer graphics | http://dbpedia.org/resource/Computer_graphics | Correct | |
| 58 | COMP | 6731 | Pattern Recognition | http://dbpedia.org/resource/Pattern_recognition | Correct | |
| 59 | COMP | 6731 | Feature extraction | http://dbpedia.org/resource/Feature_extraction | Correct | |
| 60 | COMP | 6751 | natural language processing | http://dbpedia.org/resource/Natural-language_processing | Incorrect | https://en.wikipedia.org/wiki/Natural_language_processing |
| 61 | COMP | 6751 | text mining | http://dbpedia.org/resource/Text_mining | Correct | |
| 62 | SOEN | 6751 | User interface design | http://dbpedia.org/resource/User_interface_design | Correct | |
| 63 | SOEN | 6751 | User interface | http://dbpedia.org/resource/User_interface | Correct | |
| 64 | COMP | 6761 | software engineering | http://dbpedia.org/resource/Software_engineering | Correct | |
| 65 | COMP | 6761 | collision detection | http://dbpedia.org/resource/Collision_detection | Correct | |
| 66 | COMP | 6761 | physics | http://dbpedia.org/resource/Game_physics | Correct | |
| 67 | COMP | 6771 | frequency domain | http://dbpedia.org/resource/Frequency_domain | Correct | |
| 68 | COMP | 6771 | image segmentation | http://dbpedia.org/resource/Image_segmentation | Correct | |
| 69 | COMP | 6771 | Hough transform | http://dbpedia.org/resource/Hough_transform | Correct | |
| 70 | COMP | 6771 | edge detection | http://dbpedia.org/resource/Edge_detection | Correct | |
| 71 | COMP | 6781 | N-gram | http://dbpedia.org/resource/N-gram | Correct | |
| 72 | COMP | 6781 | part-of-speech tagging | http://dbpedia.org/resource/Part-of-speech_tagging | Correct | |
| 73 | COMP | 6781 | parsing | http://dbpedia.org/resource/Parsing | Correct | |
| 74 | COMP | 6791 | Information Retrieval | http://dbpedia.org/resource/Information_retrieval | Correct | |
| 75 | COMP | 6791 | vector space | http://dbpedia.org/resource/Vector_space | Correct | |
| 76 | COMP | 6791 | Tokenization | http://dbpedia.org/resource/Lexical_analysis | Correct | |
| 77 | COMP | 6811 | dynamic programming | http://dbpedia.org/resource/Dynamic_programming | Correct | |
| 78 | COMP | 6821 | database | http://dbpedia.org/resource/Database | Correct | |
| 79 | SOEN | 6841 | risk management | http://dbpedia.org/resource/Risk_management | Correct | |
| 80 | SOEN | 6861 | Service Oriented Architecture | http://dbpedia.org/resource/Service-oriented_architecture | Correct | |
| 81 | SOEN | 6861 | Web services | http://dbpedia.org/resource/Web_service | Correct | |
| 82 | SOEN | 6861 | SOAP | http://dbpedia.org/resource/SOAP | Correct | |
| 83 | COMP | 7251 | mobile computing | http://dbpedia.org/resource/Mobile_computing | Correct | |
| 84 | COMP | 7251 | PAN | http://dbpedia.org/resource/Personal_area_network | Correct | |
| 85 | COMP | 7251 | MAN | http://dbpedia.org/resource/Metropolitan_area_network | Correct | |
| 86 | COMP | 7251 | topology | http://dbpedia.org/resource/Network_topology | Correct | |
| 87 | COMP | 7451 | distributed systems | http://dbpedia.org/resource/Distributed_computing | Correct | |
| 88 | COMP | 7521 | cryptographic | http://dbpedia.org/resource/Cryptography | Correct | |
| 89 | COMP | 7521 | authentication | http://dbpedia.org/resource/Authentication | Correct | |
| 90 | COMP | 7521 | data integrity | http://dbpedia.org/resource/Data_integrity | Correct | |
| 91 | COMP | 7521 | confidentiality | http://dbpedia.org/resource/Confidentiality | Correct | |
| 92 | COMP | 7521 | access control | http://dbpedia.org/resource/Access_control | Correct | |
| 93 | SOEN | 6481 | IEEE | http://dbpedia.org/resource/Institute_of_Electrical_and_Electronics_Engineers | Correct | |
| 94 | SOEN | 6481 | ISO | http://dbpedia.org/resource/International_Organization_for_Standardization | Correct | |
| 95 | SOEN | 6481 | Agile | http://dbpedia.org/resource/Agile_software_development | Correct | |
| 96 | SOEN | 6481 | user stories | http://dbpedia.org/resource/User_story | Correct | |
| 97 | SOEN | 6481 | formal specification | http://dbpedia.org/resource/Formal_specification | Correct | |
| 98 | SOEN | 6861 | RESTful | http://dbpedia.org/resource/Representational_state_transfer | Correct | |
| 99 | SOEN | 6491 | Refactoring | http://dbpedia.org/resource/Code_refactoring | Correct | |
| 100 | SOEN | 6211 | Semantic Web | http://dbpedia.org/resource/Semantic_Web | Correct | |

6. Conclusion

We have successfully created a system that scrapes data from Concordia University websites and stores the data in a Knowledge Graph. We have implemented a program that answers questions about our database through SPARQL queries. All the queries implemented in our program gave fast and accurate results. It was found by linking analysis of spotlight generated URI's of the entities, our project correctly identifies 96% of the DBpedia entities linked to the entities in our knowledge graph. Our project qualifies for the properties:

Scalable: The knowledge graph created can be expanded to as many numbers of courses or universities as needed. It can be expanded to contain information about other Universities or Departments too.

Flexible: Since we are taking all our University data from the university websites, any changes that occur in a course attributes or an addition of a new course would get updated in our knowledge graph every time we run the code. This ensures flexibility of the data in the KG.

Convenience: The program has a pleasant user interface which makes it easier to understand the knowledge graph and the queries related to it.

Accuracy: We have eliminated all duplicate entries and therefore data is stored only once.

7. Future Scope and Limitations

In this project we have limited our database to Concordia University alone. We could expand our knowledge graph to many Universities for our future scope. We could also store more information about a university such as the services offered by the universities, information about the faculties and their course and research topics, schedules of the courses, locations of the buildings, contact information of the departments etc. As we grow our knowledge graph, it also calls for us to optimise our code to handle much larger data so that the code doesn't take too long to run for bigger databases. This project can be improved to form a complete chatbot that answers any question regarding any specific university by including natural language algorithms to our code.