

Arabic Ontology-Based Approach for Chest Diseases Diagnosis

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INTRODUCTION

No one can ignore the importance of the respiratory system as one of the most important systems in the human body which provides it with oxygen. Therefore, any trouble in its function will lead to death.

Several diseases can infect the respiratory system, each of them attacks at least one organ and has a set of signs and symptoms. Therefore, the early detection of any defect in the respiratory system functions stands at the top of doctor's tasks.

However, the diagnostic process is not an easy process due to the complexity of the human body and overlapping phenotypes. For that reason, computer science can be helpful to support physicians in the diagnosing's process. So, creating a system based on Abstract: Chest diseases are a subgroup of respiratory system diseases. The symptoms of these diseases are similar and this makes the diagnosis process difficult. Therefore, to ease the diagnosis process we gathered Information about fourteen diseases which attack the chest with their symptoms and investigations about them. In this study, we present an Arabic ontology-based approach for chest diseases diagnosis. We focus on ontology building process. This ontology can be used to help physicians and other users, determine the chest disease that a patient is suffering from and what are the investigations that should be applied. While experts can easily gather information from this data, lay users lack the expertise needed to deal with it. Most of the efforts to solve this problem focus on the English language. So also, in this study, we present a new approach based on natural language processing to translate Arabic language query to SPARQL query.

ontology will facilitate the diagnosis for doctors, by providing it with the patient's symptoms and signs, it will query the ontology and provides feedback about the expected disease. Since, there is no research-to our knowledge-that uses ontologies to serve chest diseases in Arabic.

In addition, we propose a natural language processing model which will enable users to query the semantic data using their natural language. Our approach starts a translation process by removing stop words then identifying the query target after that form triples by mapping the rest of words in the query or their stems or their meanings with the ontology.

The rest of this papergives a background about ontology development, disease diagnosis and translating natural language queries to SPARQL queries, presents the methodology of ontology development, the implementation of the ontology using Protege tool and presents the proposed model and the evaluation of the ontology. Finally, contains the conclusion and future work.

CHALLENGES AND IMPORTANCE OF THE RESEARCH

The challenge in designing such an ontology is that it requires medical experts to revise and evaluate the gathered information. and the challenges in the field of natural language processing of Arabic language is the complexity from syntactical level including the convoluted Arabic grammar.

There are different word derivations and there is an impact of the diacritics marks on the word's meaning. Besides, there are other challenges related to the usage of the stem in which a word's stem provides words with different meanings. The contributions of this research is represented in:

- Design Arabic ontology for chest disease diagnosis
- Help lay users to query the ontology using their natural language
- Enrich Arabic content on the web since it is an Arabic ontology

Literature review: In this study, we discus many works in the domain of ontology development, disease diagnosis and translate natural language query to SPARQL query and vice versa.

Ontology development: In Alfonse, etc., researchers suggested a methodology consists of three basic modules: the diagnostic module, the staging module and the treatment recommendation module. In order to detect patient disease, the patient provides his/her signs and symptoms to the diagnostic module which detects what type of cancer the patient is suffering from. Then the staging module finds the current stage of cancer based on the cancer type, signs and symptoms that are provided by the patient. Based on the determined cancer type and cancer stage, the treatment recommendation module can recommend a specific treatment for the case.

In Wang and Tansel^[1] researchers discussed the problem of knowledge acquisition. They proposed an inference method according to the case at hand. The system takes the ontology and the symptoms as input. Then in the first phase, the system returns a set of diseases. Thus, if the number of these returned diseases is one, then the process of diagnosing is finished. However, if the number of returning diseases is greater than one, the differential diagnosing process begins where the system retrieves similar cases through semantic way depending on a knowledge model.

In Tris^[2] Cristina Romero Tris proposed a decision support system based on a knowledge base to help physicians in the diagnosis process and to verify the diagnosis made by the doctor. Therefore, if the doctor inputs a disease not related to the symptoms, the system will notify the doctor of the inconsistency. Moreover, it will suggest the disease that best fits the symptoms. The proposed system aims to personalize existing knowledge base through the extraction of a partial ontology which contains the medical terms that refer to the patient.

In Jayaratne Lakshman Jayaratne proposed a decision support system based on ontology. The proposed system composed of two components, a genotype component and phenotype component. The initial input of the system is genetic sequences of a patient. Firstly, the system identifies whether the gene in input is mutated or not with respect to the referenced gene sequences. Then, according to those mutations, a corresponding common list of phenotypes are identified and shown to the physician. They are listed down according to the frequency and probability of phenotypes that might occur. The physician must first make sure that the phenotypes given by the system are really in the patient's body and can then get the diagnostics report after choosing the matching phenotype from those listed by the system.

In Joshi and Dhariwal^[3] researchers proposed a framework composed of three phases wherein the first phase is a text analysis technique would be applied to medical records. In the second phase, the semantic analysis process was applied to store the information extracted from the first phase in a knowledge base and the last phase involves querying the knowledge base to get the related medical information. Here, they used medical rules to infer additional medical facts about the patients and to generate a rich knowledge base of patient facts.

In Baraka and El-Askary^[4] researchers proposed an ontology-based approach to diagnose the date palm disease and suggest appropriate treatment by identifying anomalous observations on the parts of the tree. The approach consists of three interrelated modules: knowledge base, reasoning engine and a server-side application. The knowledge base was built using OWL ontology and contains information related to date palm. The reasoning engine accepts user input queries and responses to data through the I/O interface by analyzing the acquired dynamic information together with the static knowledge stored in the knowledge base. In addition, the ontology was evaluated using the task-based framework and it indicates an accuracy of 100 and 97.6% when using the precision and recall method.

Translating natural language query to SPARQL query: In AlAgha^[5] researchers proposed a Generic approach to translate Arabic NL queries to SPARQL. The proposed approach utilizes the off-the-shelf Arabic language toolkit to build the parse tree of the user query. It then analyses the syntactic structure of the tree to extract NPs, identify head and modifier words and represent the query in a triple format, i.e., subject-predicate-object. The proposed approach is independent from the ontology, since, it can be easily ported from one ontology to another.

In Shaik *et al.*^[6] researchers proposed a new version of LODQA which takes as an input natural language query and returns it SPARQL form. It consists of three modules which are:

First module is responsible for handling the natural language query, parsing it and producing a graph representation called as Pseudo Graph Pattern (PGP) which contains nodes and relations. Typically, the nodes correspond to the nouns and relations to the dependency paths between them. Additionally, PGP specifies the focus of the query.

Second module is responsible for finding the URIs and values of the nodes in the PGP through mapping PGP with the ontology. A natural language term may be normalized to more than one RDF terms due to ambiguity. Therefore, more than one Anchor PGPs may be produced.

Third module is responsible for searching the target dataset for corresponding parts, considering possible variations. To absorb structural discrepancy between the APGP and actual structure in the target dataset, it attempts to generate SPARQL queries for all the possible structural variations. Then the SPARQL queries were submitted to the target endpoint and the answers were gathered and provided to the user.

In Sander *et al.*^[7] researchers proposed an approach to translate English queries to SPARQL queries based on knowledge stored in ontologies and lexicons. The translation process composed of seven steps which are:

- Delete stop words
- Stem the remaining words in the query and delete punctuation and extract dates
- Map the remaining tokens against the ontology. Concepts, individuals, relations and properties are matched, the appliance labels are extracted
- Tokens without corresponding entries in the ontology are checked against the lexicon. These are either KPIs or keywords which do not belong to the ontology
- After identifying the tokens in the ontology or lexicon, the corresponding SPIN rule which provide the main structure of the SPARQL query is extracted
- The last step merges the defined SPARQL body from the SPIN rule with the additional information from the ontology and the extracted dates

In Ngomo *et al.*^[8] researchers proposed SPARQL2NL, a novel approach that can verbalize SPARQL queries. The approach is tailored towards SPARQL constructs typically used in keyword search and question answering. It consists of four main steps: a preprocessing step which normalizes the query and extracts type information for the occurring variables. Processing step which constructs a generic representation of the query. A post-processing step which applies reduction and replacement rules in order to improve the legibility of the verbalization and a realization step which generates the final natural language representation of the query.

In Pradel *et al.*^[9] researchers proposed an approach that enabled end-users to query knowledge bases. This approach is based primarily on predefined samples that lead user query translation. The researchers relied on the hypothesis that end-user queries varied among fixed queries. The process of translating the natural language query into the SPAROL query proposed in this search consists of two basic steps. The first step was converting the natural language query to Pivot Query, Where the nominal entities were selected, the dependencies were analyzed, the target was derived from the query and a set of predefined rules was applied to the reliability statement to extract all elements of the query. The second step involved mapping the predefined samples with the Pivot query. A set of possible interpretations of the user query which were arranged according to their expected relationship with the user query, were then suggested to the user as a set of rewritten queries.

In Kim and Cohen^{[10]⁻} researchers proposed a prototype version of LODQA (Linked Open Data Question Answering) which build SPARQL query through two steps. In the first one, it uses Enju English parser which produce predicate-argument relation and in the second step, LODQA searches the ontology for corresponding terms. Researchers used OntoFinder which search in medical ontologies.

Ontology development: In this study, we present the steps which we follow to build our ontology. The proposed CDDOnto (Chest Disease Diagnostic Ontology) will be very important for diagnosing chest diseases. We gather the ontology content from Haslett *et al.*^[11] and from a domain expert who helps to identify concepts and relationships between them. There is a wide range of tools available for creating ontologies such as SWOOP, Onto Track and protege but we choose protege because according to Khondoker and Muller^[12] it is the most domain-independent tool. According to Noy and McGuinness^[13] building an ontology requires executing the following steps:

Name in English	Name in Arabic
Investigation	استقصاء
Symptom	عرض
Disease	مرض
Cough	سعال
Pulmonary embolism	الانصمام الرئوى
Interstitial pulmonary disease	داء الرئوى الخلالي
Tuberculosis	ً السلَّ
Inherited	وراڈی

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Table 2: Object properties of the classes				
Object property	English name	Domain	Range	
له_عرض	has_symptom	مرض	عرض	
عرض_ل	symptom_of	عرض	مرض	
يشخص	Diagnosis	استقصاء	مرض	
يشخص_ب	diagnosed_by	مرض	استقصاء	

Determine the domain of the ontology. We cannot build an ontology without specifying its domain. The domain of our ontology is diagnosing chest diseases. Our ontology can answer different types of questions like:

- What are the symptoms of a given disease?
- What are the investigations of a given disease?
- What is the disease of giving symptoms?
- Reuse existing ontologies

We build CDDOnto from scratch since there is no such ontology in the Arabic language. Overview the Ontology. We gathered information about fourteen diseases, their symptoms and investigation. Then we organized them in an ontology. Table 1 shows all classes of the ontology.

Enumerate main terms in the ontology. We find three main classes, a disease (مرض) which has four subclasses, symptom (عرض) which has one subclass and investigation (استقصاء).

Define the properties of the classes. There are two types of properties: object property and data property. While the first links object to object, the other link object to XML schema. We add data properties such as السم , سبب, to the ontology. They are used for giving a value to an instance of a class. For example, "سبب" cause of اللتدرن البئ primary tuberculosis is Mycobacterium tuberculosis "المتفطرات السلية". The data property is applicable to each instance of the class. Table 2 illustrates the object properties while Table 3 illustrates data properties.

Define the attributes of the data properties. Each data property has attributes and each of them is specifying different things such as value types, allowed types and the cardinality of the values which attribute can take.

- Value type: describes the types of values a property can have. For example, "سبب" property has a value of type string
- Allowed values: describes the allowed values for the properties. For example, the allowed values for the "استقصاء" and "عرض""مرض" "اسم ومرض السم ومرض الم

Data property	In English	Domain	Range
اسم	Name	مرض, عرض, استقصاء	String
سبب	Cause	مرض	String
توصيف	Description	مرض	String



Fig. 1: Data cardinality of data properties

Cardinality: defines how many values a property can have. For example, the property "اسم" has multiple cardinalities but it has at least one. Figure 1 shows the value type and cardinality of some of the properties

Creating instances of CDDOnto. The creation of individuals allows all the properties of the classes to be recorded. We defined 101 instances that represent all ontology concepts including diseases, symptoms and investigations.

Applying ontology reasoner: We apply an ontology reasoner (e.g., Hermit reasoned) to check that everything is ok and to identify new relations from existing ones.

CDDONTO IMPLEMENTATION IN PROTEGE

This section describes the development of CDDOntoin protege as an owl ontology.

Classes and subclasses: In CDDOnto disease (مرض), symptom (استقصاء) and investigation (استقصاء) are subclasses of class thing. Figure 2 shows the main classes in the ontology whereas Fig. 3 shows all.

Instances, properties and facets: In CDDOnto, individuals are defined with their data properties. In addition to object properties between them. Figure 4 shows data taxonomy such as "الانصمام الرئوي" pulmonary embolism class which contains three instances. Pulmonary embolism is a blockage in the pulmonary artery which supplies the blood to the lungs. Data Properties are shown in Fig. 5 which contains three properties, these properties are explained as follows:



Fig. 2: Protege snapshot of the class hierarchy



Fig. 3: Taxonomy of main classes



Fig. 4: Instances of pulmonary embolism

- اسم "name": اسم the refers to the word of circulation of the disease in the world
- سبب "Cause": express an event in the human body which Cause a disease
- وصف"Description": describes either a disease or an investigation



Fig. 5: Data properties of an individual

ARABIC TO SPARQL TRANSLATION APPROACH

In this study, we will go through the steps of the proposed approach which is illustrated in Fig. 6.

Removing stop words: The stop words were known from the beginning of information retrieval which are the most commonly used words in the language and are not meaningful but are used to complete it such as subjective pronouns, demonstrative pronouns, prepositions, negation, WH questions, adverbs and adverbs of place. Deleting these words will not affect the content of the query.

Identifying query goal: We mean by the query goal the word which come after the question word and it, will be the first word after the previous step (remove stop word). It is mapped to one of the ontology classes or data properties. After that, the goal is passed to light stemming process like removing the "IL" from the beginning of each word.

Generating NGram: NGram is a sequence of N elements of a given text string, used in several areas, such as determining the probability of automatic machine translation, predicting the next most likely word in a sentence, checking the spelling of text or identifying the feeling of part of a given text^[14].

Getting ontology concepts: In this step, we query the ontology to get all ontology classes and data properties and then we normalize them by removing the underscore (_) between the terms. After that, we stem them all using the same stemming algorithm which is used in stemming the query terms to get shared roots between the query terms and ontology concepts.

Mapping query terms to ontology concepts: In this step, we map the query terms, one of their meanings using Arabic WordNet, their stems or one of the meaning of their stems to the ontology concepts and when we find a correspondence, we replace the query term with it. The stemming process is useful for obtaining the same root for different word forms. There are different algorithms available for Arabic word stemming but, in this research, we use ISRI Arabic stemmer Taghva^[15].



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Fig. 6: The approach of translating Arabic natural language query to SPARQL query



Fig. 7: Query to get individuals in the ontology

Specifying query individuals: In this step, we specify all query individuals resulted from the previous step. We can specify individuals in the ontology through the query which is illustrated in Fig. 7.

Generating duals: In this step, we group the query goal with each individual to form pairs (query goal, individual).

Generating triples: In this step, we generate the triples using the duals generated in the previous step. We get the relation linked between them through querying the ontology. It is known that each relation has two sides, domain and range and we can get the property using the query which is shown in Fig. 8. Note that the line 2 in Fig 7 and 8 is the ontology URI.

In case the range of the natural language query is an individual, the previous query will return an empty result set. Therefore, we use the query shown in Fig. 9 to get the type of the individual then use the query that is shown in Fig. 8 to get the relation.

PREFIX rdfs:< <u>http://www.w3.org/2000/01/rdf-schema#</u> > PREFIX my: < <u>http://www.semanticweb.org/ali/ontologies/2017/2/CDDOnto#</u> > SELECT ?property WHERE { ?property rdfs: domain my: domain rdfs: range my: range.}

Fig. 8: Query to get object property between two classes

 PREFIX my: < http://www.semanticweb.org/ali/ontologies/2017/2/CDDOnto# >

SELECT ?class WHERE {my: Individual rdf:type ?class}

Fig. 9: SPARQL query to get the type of an individual

Forming the query: The SPARQL query consists of three sections which are: SELECT statement, WHERE clause and solution modifiers. In this research, we focus on forming the first two sections and we do not discuss the forming of the solution modifier section. The RDF triple that was built in the previous step forms the WHERE clause for the SPARQL query. Now we have to construct the SELECT statement and associate it with WHERE clause to form a meaningful SPARQL query. To construct the SELECT statement, we have to identify the target of the query which we specified in Step 2, now we have to link the target of the query with the WHERE clause. We have done this by the following:

Add a projection variable to the SELECT clause which we name it as ? goal. The forming of the WHERE clause differs depending on the purpose of the query. In case the purpose of the query is a data property then the WHERE clause is formed as following: individual, data_property, goal. Then, we replace all occurrences of the goal with the variable ? goal. After that, we add the Ontology URI to all expressions in the query-except the variables-and all required namespaces as prefix to the query.

Processing conjunction and disjunction: The previous approach fails when the query contains a conjunction or disjunctions as the following example: له عرض حمى و الداء , الذي , i.e., what is the disease that has a symptom of fever and nasal polyps? (1). In order to solve it we do not delete the conjunction or disjunction letters which are 'و' i.e. 'and' and 'i' i.e. 'or' when deleting the stop words. Then we form the query and concatenate the triples using the conjunctions and disjunctions.

Evaluation of CDDOnto: In this section, we evaluate the ability of the ontology in representing all terms, properties and relations through ontology querying using description logic query which is a standard Protege plugin and the SPARQL query to verify and validate the ontology in accordance to competency questions. In addition to verify it with the help of a domain expert since we gathered 22 real case for patients then order doctor to diagnose it. After that, compare it with the result of the ontology.

Ontology evaluation through ontology querying: We present three querying examples which answer the main questions that are asked in the development process of the ontology. Example 1:

- The question: what are the diseases that have the symptom السهال?
- DL_Query: اسهال value له_عرض and مرض
- Figure 10 shows the result of DL_Query and illustrates the individuals of the disease class

Example 2:

- The question: What are the symptoms of cystic fibrosis disease?
- DL_Query is: التليف الكيسى value عرض له value عرض
- Figure 11 shows the result of DL_Query and illustrates the individuals of symptom class

Example 3:

- In this example, we show a SPARQL
- The question: what are the symptoms of cystic fibrosis disease?

Figure 12 shows SPARQL query and Fig. 13 shows the result of the SPARQL query which illustrates the symptom's individuals. That means all these symptoms are symptoms of the cystic fibrosis disease.



Fig. 10: Individuals of a disease

DL query:
Query (class expression)
النَّبْتِ Value مرض من value مرض لَّ symptom and symptom_of value cystic fibrosis (english transaltion)
Execute Add to ontology
Query results
Instances (9 of 9)
spontaneous pneumothorax استرواح صدر_عقوی 🔶
میلینات_انفیه (nasal polyps
productive cough of a green sputum سعال_منتع_نفتح_أخضر 🔷
productive cough of yellow purulen سعال_منتع_لفتيع_فيص_أصفر 🔷
🗄 سکری diabetes mellitus د cough
respiratory failure قصور_تنفسي 🔷
🔷 کی_رلوی pulmonary heart
🔶 نفت_سوى hemoptysis

Fig. 11: Individuals of cystic fibrosis's symptom

SPARQL query:		
PREFIX my: <http: www.semantic<br="">SELECT ?symptom</http:>	web.org/ali/ontologies/2017/2/CDDOnto#>	
WHERE { ?symptom	{ التليف_الكيس:•my عرض_لــmy عرض	
WHERE {?symptom :	ny: symptom_of my:cystic_fibrosis} (english transaltion)	

Fig. 12: SPARQL query

Ontology evaluation with real cases: In this step, we evaluate the ontology using 22 real cases for patients. Then order a specialist doctors to diagnose them. After that, we compare it with the results we gathered from the ontology. The ontology shows an accuracy of 72.7%.

ستان منتج القتي قيمي أصفر Productive cough of yellow purulent sputum استراح مدر عنوی spontaneous pneumothorax اسرونیات أنتیه diarhea قال بولیات أنتیه pulmonary heart محری diabetes mellitus محر انتها respiratory failure سمال منتج القتي أخضر Productive cough of green sputum انت موری hemoptysis

Fig. 13: SPARQL query result

CONCLUSION

In this study, we proposed an Arabic ontology-based approach for diagnosing chest diseases. We focused on the process of building an ontology. The ontology content is related to the medical domain and It was gathered from^[7] and from a domain expert. The ontology provides a knowledge base of chest diseases, symptoms and investigations. The proposed ontology is evaluated with a help of domain experts and showed an accuracy of 72.7%.

Then, we developed a natural language processing model which enables users query the ontology using their Arabic natural language and get the desired knowledge. In the future, we intend to save the translation results of the queries, then using data mining techniques to get the translation based on the closest case to the case at hand.

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