Trabalho prático individual nº 2

Inteligência Artificial Ano Lectivo de 2022/2023

16 de Dezembro de 2022

I Observações importantes

- 1. This assignment should be submitted via GitHub within 27 hours after its publication. The assignment can be submitted after 27 hours, but will be penalized at 5% for each additional hour.
- 2. Complete the requested methods in module "tpi2.py", provided together with this description. Keep in mind that the language adopted in this course is Python3.
- 3. Include your name and number and comment or delete non-relevant code (e.g. test cases, print statements); submit only the mentioned module "tpi2.py".
- 4. You can discuss this assignment with colleagues, but you cannot copy their programs neither in whole nor in part. Limit these discussions to the general understanding of the problem and avoid detailed discussions about implementation.
- 5. Include a comment with the names and numbers of the colleagues with whom you discussed this assignment. If you turn to other sources, identify these sources as well.
- All submitted code must be original; although trusting that most students will do this, a plagiarism detection tool will be used. Students involved in plagiarism will have their submissions canceled.
- 7. The submitted programs will be evaluated taking into account: performance; style; and originality / evidence of independent work. Performance is mainly evaluated concerning correctness and completeness, although efficiency may also be taken into account. Performance is evaluated through automatic testing. If necessary, the submitted modules will be analyzed by the teacher in order to appropriately credit the student's work.

II Exercícios

Together with this description, you can find modules semantic_network, bayes_net and constraintsearch. They are similar to the ones used in practical lectures, but small changes and additions were introduced.

Module tpi2 contains some derived classes. In the following exercices, you are asked to complete certain methods in these classes. Any other code that you develop and integrate in other modules will be ignored.

The module tpi2_tests contains some test code. You can add other test code in this module. Don't change the semantic_network, bayes_net and constraintsearch modules.

You can find the intended results of tpi2_tests in the file tpi2_results.txt

The responses to the main questions asked by students during this TPI will be collected in section III below.

- 1. The small semantic networks module used in practical classes was designed to facilitate the initial learning of the subject by the students. For this reason, many aspects were left out. In a more professional system, they would have to be considered. The attached module semantic_network was designed based on the one you already know, but has some differences related to type checking. For example, in the classes module, it is not possible to know if an association is established between two objects or between two types. There is also no way to define whether an association accepts only one value or multiple values. The Association class constructor was therefore modified and now has the following arguments:
 - entity1 First argument of the association;
 - name Association name;
 - entity2 Second argument (or value) of the association;
 - card Cardinality of the association, which can be:
 - None To be used between objects, i.e. entity1 e entity2 are necessarily names of objects.
 - "one" To be used between types. Specifies that the association, when used for objects, will accept a single value. Example: a person has a single father.
 - "many" To be used between types. Specifies that the association, when used for objects, will accept multiple values (zero or more). Example: a person can have multiple children.
 - default To be used for associations between types. It specifies the default value (second argument) of the association. Only associations with cardinality "one" can have a default value. Example: by default, the hight of a "man" is 1.75.

Since it does not carry out any validation, the original module (from lectures) allows any string of characters "xpto" to be used by the same user simultaneously as the name of an object and as the name of a type. In the module semantic_network now provided, the class SemanticNetwork has three new methods to create relationships, which return True if it is possible to create the relationship and False otherwise.

- add_member(user,obj,type)
- add_subtype(user,subtype,supertype)
- add_association(user,e1,name,e2,card,default)

Note that the implementation of these methods is very incomplete due to the lack of auxiliary methods to use for type checking. It is on these auxiliary features that you should now focus your attention:

a) Develop a method is_object(user,obj) in the MySN class that checks if the object objects in user declarations

- b) Develop a method is_type(user, type) in the MySN class that checks if the type type exists in user declarations
- c) In order to maintain consistency, it is necessary to check, for each new relation to be added, whether the types of entities in that relation are consistent with the types already existing in the network. Since, in this semantic network, it is possible to declare a concrete association between two objects without declaring the general characteristics of that association at the type level, it becomes necessary to develop type inference mechanisms. So, develop the following methods in the MySN class:
 - infer_type(user,obj) Infers the type of an object based on the declarations of user. The result must be:
 - <type> the type of the object, provided that it can be inferred;
 - "_unknown_" if obj exists, but its type cannot be inferred;
 - None if obj doesn't exist.
 - infer_signature(user,assoc) Infers the signature of the association based on the declarations of user. The result must be:
 - (t1,t2) where t1 and t2 are the types of the entities involved in the association;
 - None if the association does not exist.

These two functions depend on each other, so to be completely correct they will have to be implemented as mutually recursive functions. However, you can obtain part of the score by solving the simpler cases.

2. Bayesian networks have several mathematical properties. One such property is the *Markov blanket*. It is defined as a set containing all the variables that protect a given variable from the rest of the network. All knowledge about this variable can be extracted from the knowledge of its Markov blanket. The Markov blanket of a variable A in a Bayesian network is the set of nodes consisting of the parents and children of A and the other parents of the children.

Implement in the MyBN class a markov_blanket() method that, given a variable of the network, var, returns a list of the variables that make up the Markov blanket of that variable. Not that the Bayesian network implementation in bayes_net module is based on lists of true and false variables and not on explicit boolean values.

For testing purposes, the network represented in Figure 1 is included in the tpi2_tests module.

- 3. The following exercices cover the topic of constraint-based search for assignment problems.
 - a) Implement in class MyCS the method propagate(var) which is called in search() after choosing a specific value for var.
 - b) Implement in class MyCS the method higherorder2binary(ho_c_vars,unary_c), which will be used to transform a higher-order constraint involving the variables in list ho_c_vars, into a set of binary constraints involving those variables and a new auxiliary variable (see slides). This method updates the domains and constraints dictionaries and does not return a result. unary_c is a function that describes the higher-order constraint as unary constraint on the new auxiliary variable.

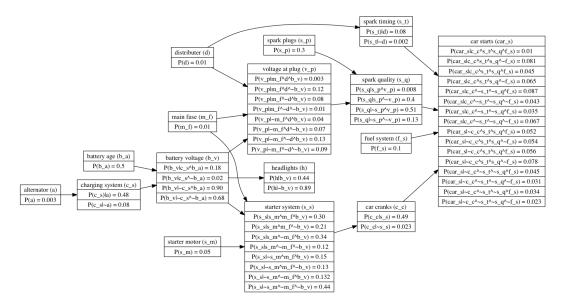


Figura 1: Bayesian network for a car that does not start

III Clarification of doubts

This work will be followed through http://detiuaveiro.slack.com. The clarification of the main doubts will be placed here.