PROJECT SPECIFICATION

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AIM: given two datasets named disease_evidences.tsv and gene_evidences.tsv derived from DisGeNET COVID-19 data collection as input, write a program able to read the two files, compute operations on them exploiting classes, Object-Oriented programming and Python packages (Pandas, Flask and ABC) and present the user the relevant outcomes on a web application.

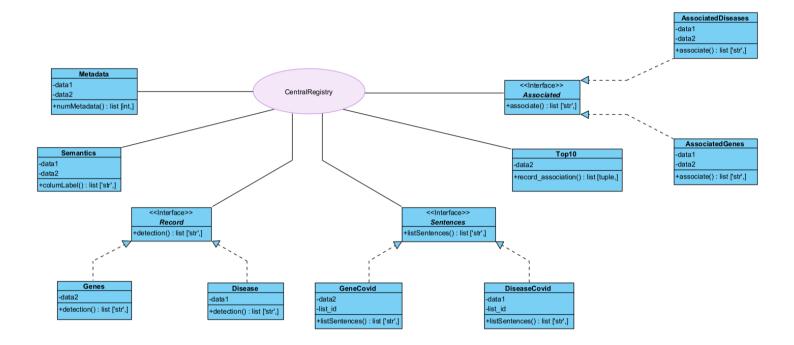
WORKFLOW

 After reading the given files and having understood the instructions to build our web application, first we focus on how to define our normal classes, one per each operation and then, when possible, we try to link some classes together by implementing a couple of abstract classes serving as template from which we can generate subclasses to perform similar operations in a faster and more efficient way. Then, we move to the design of the CRC cards basing on the classes we are going to implement in part 2.

Below, there are the examples of an abstract and a normal class:

Abstract Record	Data
Record	Genes, Disease
	disease_evidences
	gene_evidences
Genes	Data,Record
The list of different genes detected in literature sorted in ascending order	• gene_evidences

Then, we build the complete UML diagram by connecting the abstract classes
with their realizations (through blue dashed lines) and normal ones with their
associations (through dark straight lines). All the classes are characterized by
their respective attributes and methods.



The next step consists of the implementation of three parts: part1.py with the
two dataset readers using pandas and the registry listing the different types of
operations specified in part2.py with their relative classes. While part3.py
contains the codes to build the web-based user interface and its Flask
application. Such a UI provides a list of choices, where each choice enables an
analytical objective.

REGISTRY CLASS OF PART_1

- → It defines the readers of the two datasets, the list of operations needed to be performed and the list of links of each web page, one per operation.
- → It provides a function for each operation calling the result from the imported file of part2.py. This is necessary for the implementation of part3.py.

Grouping all functions of part1.py in a unique CentralRegistry class that takes as parameters the two DataFrames (named data1 and data2 in our program) facilitates the performance of the operations and the calling of functions in part2.py.

CLASSES OF PART 2

Class Metadata:	Input: the two datasets
Class Metadata.	Output: list of 4 integers
	Functions: return the number of rows/columns of the two
	DataFrames
	Technique(s): built-in function .shape() of Pandas
Class Semantics:	Input: the two datasets
	Output: a list of strings
	Functions: return the labels of the columns of the two
	DataFrames
	Technique(s): built-in function .columns() of Pandas
Class Record(ABC):	abstract
Class Genes(Record):	Input: gene_evidences.tsv
	Output: a list of the geneid
	Functions: return distinct genes in an ascending order
	Technique(s): built-in function .groupby() of Pandas
Class Disease(Record):	Input: disease_evidences.tsv
	Output: a list of the diseaseid
	Functions: return distinct diseases in an ascending order
	Technique(s): same as Genes
Class Sentences(ABC):	abstract
Class GeneCovid(Sentences):	Input: gene_evidences.tsv
	Output: list of strings
	Functions: return all sentences associating COVID with genes
	Technique(s): usage of ['column_label'] to extract the columns
	of data of interest and index them to look for evidence of covid
	exploiting the " <span "="" and<="" class="disease covid cdisease" th="">
	" <span "="" class="disease covid cvirus" inside="" sentences.<="" strings="" th="">
Class DiseaseCovid(Sentences):	Input: disease_evidences.tsv
	Output: list of strings
	Functions: return all sentences associating COVID with diseases
	Technique(s): same as GeneSentences without the " <span "="" class="disease covid cvirus" string.<="" th="">
	class- disease covid cvii us stillig.
Class Top10:	Input: the two datasets
	Actual output: list of tuples((geneid, diseaseid): frequency)
	Output: list of lists[[geneid], [diseaseid], [frequency]]
	Functions: return the top 10 distinct disease-gene associations
	Technique(s): exhaustive search exploiting the pmid values
01	
Class Associated(ABC):	abstract
Class Associated(ABC): Class AssociatedDiseases(Associated):	Input: the two datasets
<u>-</u>	Input: the two datasets Output: list of strings
i	Input: the two datasets Output: list of strings Functions: provide disease list given gene symbol/ID
Class AssociatedDiseases(Associated):	Input: the two datasets Output: list of strings Functions: provide disease list given gene symbol/ID Technique(s): usage of .loc[] and the pmid values
<u>-</u>	Input: the two datasets Output: list of strings Functions: provide disease list given gene symbol/ID Technique(s): usage of .loc[] and the pmid values Input: both datasets
Class AssociatedDiseases(Associated):	Input: the two datasets Output: list of strings Functions: provide disease list given gene symbol/ID Technique(s): usage of .loc[] and the pmid values Input: both datasets Output: list of strings
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WEB APPLICATION PART 3

Using Flask and importing the file of part1.py, we create a homepage with the list of provided operations and for each of them, bound to the corresponding class defined in part2.py, we build a web page using distinct HTML files to return the final result.

- → self.__links in part1.py allows to have short links for their relative webpages. Each one corresponds to their relative operation.
- → Implementing clickable "Homepage", "Go back" and "Continue" buttons allows to facilitate webpage navigation.

EFFICIENCY

From operation 1 to 6, we can observe a high efficiency, as the computing time is very short taking only a few seconds. Points 8 and 9 take a few more seconds because of the double for loops iterating on large DataFrames. While point 7 takes around 10 hours to compute the output due to the exhaustive search approach performed on huge datasets. In this case, we have preferred the optimality of the solution at the expense of the time efficiency. Consequently, we saved the output and assigned it to a variable.

WORKING APPROACH

- logical division of the different parts of the program to build them one after another in order to facilitate their connection increasing the final working efficiency
- reasoning together to understand the general path to follow and then working on the same parts either simultaneously or separately
- all the parts of the project (such as PNGs, CRC Cards, further comments etc.) are found in the same folder

```
def Neurons(hours_of_programming):
    if hours_of_programming > o:
        return 'RIP'
    else:
        return 'see ya next time'
print(Neurons(118))
```