BioLink/PubChem Group - Final Report:

Motivation:

BioLink is a biological database that leverages Neo4J's functionality to create a graph of biological data, complete with relationships between different types of biological information. BioLink currently features genes and diseases, as well as the relationships between them, and is working on expanding other biological entities into the graph. In its completed state, BioLink will serve as an information exploration/retrieval tool that will allow researchers to discover interesting relationships between different biological entities. The final product will also feature a UI that will have completely abstracted away Neo4J, allowing for querying without knowledge of Cypher.

PubChem is the world's largest chemical database. Operated by the United States National Institute of Health, PubChem offers API access to millions of chemical compounds and bioassays. While there is plenty of data in PubChem, it is hard to traverse, and the relational components of compounds and bioassays are abstracted out by a middle-man class of data called a substance. A substance is essentially a compound that was tested on by a specific bioassay, and the existence of substances in PubChem slows down a user's ability to understand the relationships between bioassays and compounds. If this data were to be imported into a graph database, the connectivity of compounds and bioassays to each other (and even genes and diseases) would be much easier to visualize and draw conclusions from.

Project Goals:

In this project, we aimed to add node types and associations from the data found in PubChem. This database was primarily focused around three major biological entities: the substance, the compound, and the bioassay. Contributors to the PubChem database can add information about a particular chemical substance, which are the Substance entities. But different contributors could upload slightly different information for one specific chemical structure, so there ends up being multiple substance records for one actual substance. This is where the compound entity comes in, by merging the most important content from all of these substance

records into one normalized representation of that chemical structure. Knowing all this, we decided to focus on the compound entity over each specific substance entity. An additional goal was to extract from PubChem details about bioassays, which represent tests of the compound entities. Once we have these two entities from PubChem, we planned to make sure to identify the specific associations between them. Finally, a major goal for our project was to make sure that all these new node types and links were not only added into the BioLink database but also integrated by additionally including links from all this new data back to the existing data. We aimed to do this by finding connections between bioassays and genes, the latter being an already prevalent feature in the database.

Because PubChem houses data for millions of compounds/bioassays, we chose to import only the first 3000 compounds and their associated bioassays. Because we were not importing all the data, it was important to make sure that our work was easily extensible, so that, in the future, more PubChem data could be easily imported using our extensions to the existing BioLink import tool, provided the data was formatted as a CSV, as we made sure our data was.

Significance:

The reason this project is significant is because it lets us leverage graph databases in order to derive new associations that are otherwise not easily accessible. In the context of the project, we can easily understand far-away relationships between compounds and the respective genes they interact with by using their common factor in bioassays. By tying in more information with what BioLink already has in the form of proteins and diseases, we can help the medical field more quickly understand various causes of diseases.

Additionally, this platform helps serve as a multi-modal database that can support various types of queries within its domain. As expected, by adding support for an even greater domain, we can help expand not only the number of features we can query about but also the types of complexity of the queries themselves.

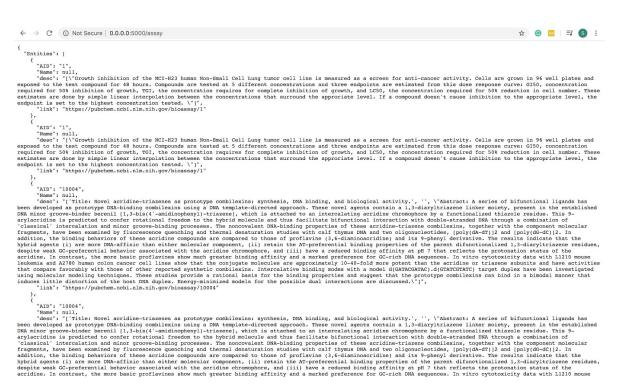
As a final result, now that we have added compounds and their associated bioassays, we can understand the effects of compounds on various diseases - since compounds are tested by specific bioassays, which interact with genes, which are associated with diseases.

Methodology:

We used the PubChemPy API and the PubChem REST API to load our data. We ran a python script to load 3000 compounds, find and load each compound's associated bioassays, and locate associated genes to those bioassays. We then exported the data on compounds, assays, genes, and the interconnections between them as CSVs to upload via a slightly modified version of the existing BioLink data import tool (specifically the config.json and db.py files).

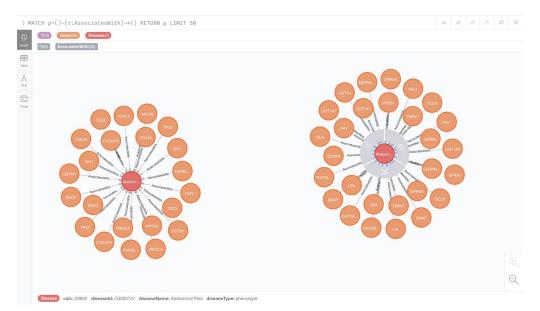
Results

After we wrote code to extract the information present in the PubChem and added the data to the BioLink database, here are the results of our successful work. We specifically show in the screenshots below how we added bioassays and compounds to the backend of the database, how we added different edge types (InteractsWith and TestsFor) to the biolink graph database, and how these changes translate to the front end querying:

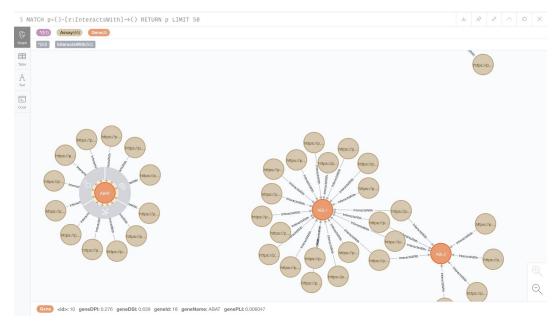


This is a picture of how bioassays, which we added to the biolink database, are stored in the backend of the biolink application. A bioassays here is made up of a bioassay ID (AID), name, and in-depth description (desc).

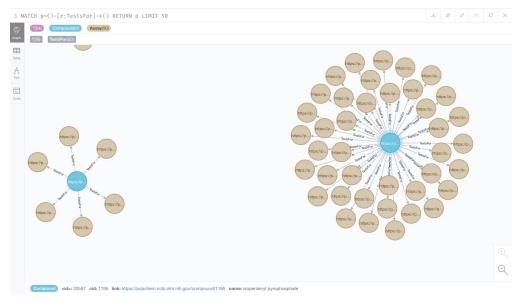
This is a picture of how compounds, which we added to the biolink database, are stored in the backend of the biolink applications. A compound here is made up of a compound ID (cid), name, and link to more information about the compound (link).



This is a picture of how genes are linked to diseases, the AssociatesWith edge. These edge and node types already existed in biolink before our project, but since we connect bioassays and compounds to genes we wanted to show how those gene nodes are connected to diseases to start off.



This is a picture of how bioassays are linked to genes, the InteractsWith edge. We added the bioassay node types to the biolink database and were able to create this edge type through the data pubchem provides on bioassay to gene interactions.



This is a picture of how compounds are linked to bioassays, the TestsFor edge. We added the bioassay and compound node types to the biolink database and were able to create this edge type through the data pubchem provides on compound to bioassay interactions.

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- Home
- Genes
- Diseases
- Compound
- Assay

Standard	

This is a picture of the front-end of the biolink application, done in React. We added the compound and assay links.

Conclusion

Ultimately, our project goals were successfully met in this final project. The goals we accomplished were understanding the complexities of PubChem and gaining some primitive domain knowledge on the biological data it stored, then formulating a procedure for how to extract the data we felt was important to add to the BioLink database through the use of the API's, and then we updated the files that import data directly into the database in a scalable way. In the future, contributors to BioLink can leverage our work by adding in more data from PubChem relating to bioassays and compounds, as well as their interactions, to gain a more complete understanding of how bioassays and compounds interact with different genes and diseases.