

Applying Linked Data Approaches to Pharmacology: Architectural Decisions and Implementation

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Abstract. The discovery of new medicines requires pharmacologists to interact with a number of information sources ranging from tabular data to scientific papers, and other specialized formats. In this application report, we describe a linked data platform for integrating multiple pharmacology datasets that form the basis for several drug discovery applications. The functionality offered by the platform has been drawn from a collection of prioritised drug discovery business questions created as part of the Open PHACTS project, a collaboration of research institutions and major pharmaceutical companies. We describe the architecture of the platform focusing on seven design decisions that drove its development with the aim of informing others developing similar software in this or other domains. The utility of the platform is demonstrated by the variety of drug discovery applications being built to access the integrated data.

An alpha version of the OPS platform is currently available to the Open PHACTS consortium and a first public release will be made in late 2012, see <http://www.openphacts.org/> for details.

Keywords: pharmacology, linked data, data integration

1. Introduction

The exploration and development of new drugs requires scientists to draw knowledge from multiple sources of information. These range from online databases of proteins (e.g. UniProt [26] and Enzyme [3]) and chemicals (e.g. ChEMBL [19], ChemSpider [37], and DrugBank [30]), to models of biological pathways (e.g. Reactome [33], WikiPathways [29], and KEGG [28]) as well as the scientific literature.

These information sources are often held in different formats and sourced from a wide variety of organiza-

tions. Together they cover a wide area of the scientific space of interest, but at the same time overlap scope, and thus frequently record different (or even inconsistent) representations of the same data. In recent years, several key datasets for drug discovery have been published on the Semantic Web including those provided by Chem2Bio2RDF [11] and Linking Open Drug Data (LODD) [40]. In the latter case, the data is linked following linked data principles [8,24].

The integration of knowledge from these disparate sources presents a significant problem to scientists, with the intellectual and scientific challenges often

being overshadowed by the need to repeatedly perform error-prone and tedious mechanical tasks. The first challenge is to identify the entities of interest in the many data sources, and to relate and map these to one another. This allows complementary information from the data sources to be collated in a single record. For example, ChemSpider contains data about chemical compounds and where they can be sourced, while ChEMBL complements this with data about the bioactivity of drug-like molecules and DrugBank provides information on the clinical use of drugs which contain the molecules. These data sources can be linked based on the chemical structure of the compounds. However, differences in scientific or technical approaches to molecular structure representation mean that different data sources will not always be in agreement. For example, searches for the chemical “Fluvastatin” on ChemSpider¹ and DrugBank² return different compounds: although their basic chemical structure matches, the compounds differ in their stereochemistry³.

A further challenge is the lack of semantics associated with links in traditional database entries. For example, the entry in UniProt for the protein “kinase C alpha type homo sapien”⁴ contains a link to the Enzyme database record⁵, which has complementary data about the same protein and thus the identifiers can be considered as being equivalent. One approach to resolve this, proposed by Identifiers.org, is to provide a URI for the concept which contains links to the database records about the concept [27]. However, the UniProt entry also contains a link to the DrugBank compound “Phosphatidylserine”⁶. Clearly, these concepts are not identical as one is a protein and the other a chemical compound. The link in this case is representative of some interaction between the compound and the protein, but this is left to a human to interpret. Thus, for successful data integration one must devise

strategies that address such inconsistencies within the existing data.

In this application paper, we present the architectural decisions made in implementing the Open Pharmacology Space (OPS) platform within the Open PHACTS project⁷. The key goal of the project is to support a variety of common tasks in drug discovery through a technology platform that will integrate pharmacological and other biomedical research data using open standards such as RDF. The OPS platform is being developed by a public-private partnership to address the problem of public domain data integration for both academia and major pharmaceutical companies [44], and will make its first public release in late 2012. A key driver of the project is that integrated data should address concrete pharmacological research questions and integrate with the workflows and applications already used within the drug discovery pipeline.

This paper focuses on two key contributions:

- A set of seven architectural decisions for a data integration platform driven by pharmacological business questions. These decisions should inform others developing similar or related software.
- A discussion of the use of this platform by three separate drug discovery applications.

The rest of this paper is organized as follows. Related work is discussed in Section 2. A summary of the concrete pharmacological research questions that the functionality provided by the OPS linked data platform seeks to address are presented in Section 3. Section 4 discusses the architectural decisions made to provide an integration framework that is capable of integrating public data sources in order to answer the top priority research questions. Section 5 gives details of the implementation of a linked data platform for pharmacology that enables enriched functionality on three separate, existing, drug discovery applications. Finally conclusions are drawn in Section 6.

2. Related Work

The OPS platform deliberately builds upon a large amount of prior work delivered by the Semantic Web community in two key areas: infrastructure and data sources.

¹<http://www.chemspider.com/Chemical-Structure.393587.html> accessed 17 Sept 2012.

²<http://www.drugbank.ca/drugs/DB01095> accessed 17 Sept 2012.

³For details see <http://www.chemconnector.com/2012/07/29/> accessed 17 Sept 2012.

⁴<http://www.uniprot.org/uniprot/P17252> accessed 17 Sept 2012.

⁵<http://enzyme.expasy.org/EC/2.7.11.13> accessed 17 Sept 2012.

⁶<http://www.drugbank.ca/drugs/DB00144> accessed 17 Sept 2012.

⁷<http://www.openphacts.org> accessed 17 Sept 2012

In terms of infrastructure, there have been several semantic data integration platforms proposed and developed. ODEMapster [5] is a global-as-view data integration approach for exposing relational data sources through an existing ontology. MASTRO [10] provides support for more expressive queries to be evaluated over a subset of *DL-Lite* ontologies. Fox *et al.* [18] discuss the requirements for domain ontologies and relationships to upper ontologies when deploying a semantic integration system. FedX [41] provides a framework for federating queries over a set of distributed SPARQL endpoints. All of these systems present a single integrated view of the data to the users and build upon the large body of work on data integration within the database community [22,31]. Central to these data integration systems are the mappings between the data sources (which in many cases are relational databases) and the intended semantics of the integrated result (represented as ontologies in semantic integration systems). Such mappings need to be supplied by domain experts with a good understanding of the data sources and the (ontological) view of the integrated data to be created. Dataspaces [21] is an approach to lower the effort required by the domain expert to generate and maintain mappings. The dataspace system starts with a minimal set of mappings which is incrementally enriched based on user feedback [6].

Another important area of infrastructure work are scientific workflow systems such as Taverna [34], Galaxy [20], and Kepler [32]. These systems are designed to enable users to compose a series of operations into a *workflow* that will extract data from sources, and execute data manipulation and computational services to analyse it. An important aspect of these systems is capturing the provenance of each operation performed by the workflow. A key idea behind workflows is enabling their reuse so that results can be reproduced. Community portals such as MyExperiment [15] support the sharing and reuse of workflows. The OPS infrastructure builds on the Large Knowledge Collider (LarKC) system that provides a pluggable workflow environment for developing scalable Semantic Web applications that integrate large datasets [13].

The Open PHACTS project is, of course, not the first to realize that the use of Semantic Web standards, along with common ontologies, can ease the integration burden in the pharmacology domain [36]. Samwald *et al.* [39] present the findings of the W3C Health Care and Life Sciences interest group with regards to the availability of data sources, their potential for linking, and present recommendations for best

practices in publishing pharmacology data as Linked Data. The Bio2RDF [7], Neurocommons [38], Linking Open Drug Data (LODD) [40], Linked Life Data [35] and Chem2Bio2RDF [11] projects have all made significant sets of biology and chemistry data available in RDF. We built upon the comprehensive work of the community in creating RDF-based data sources to power the OPS platform.

3. Motivation: Pharmacology Research Questions

A key driver of the OPS system is that the integrated data should address concrete ‘real world’ pharmacology research questions and integrate with the workflows and applications already used within the drug discovery pipeline. A collection of 83 questions has been created and prioritised by the scientific and pharmaceutical partners in the Open PHACTS project. Full details of the research questions, the generation and prioritisation process, and analysis of the results can be found in [2,25]. We provide a brief summary so as to motivate the functionality provided by the OPS system, and the data sets that have been integrated to provide an information space.

The business questions span a range of areas, however, the focus in developing the OPS platform is on those concerning pharmacology data, specifically the interactions between compounds and molecular targets. Many of the questions can be answered now by industry researchers or by academic researchers with their internal systems and via bio- and chem-informaticians. However, a shared, fully interoperable and routinely updated, scientist-friendly platform for performing these analyses in a non-manual way does not currently exist.

In the simplest case, the questions would be of the form:

“Retrieve all information available about aspirin.”

This requires discovering all of the entries for aspirin in the available data sources, e.g. ChemSpider, ChEMBL, and DrugBank, and combining the data. Such a task is complex in itself due to the different focus of the data sources and the fact that each uses its own set of identifiers, not to mention the inconsistencies in the overlapping data that must be resolved or displayed in an appropriate form to the user.

The more complex research questions involve filtering compounds based on their properties and experimental results, e.g.

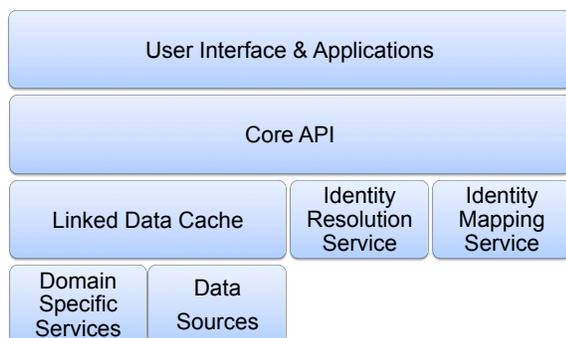


Fig. 1. OPS Platform Architecture Overview

“Give all oxidoreductase inhibitors with IC50 activity value < 100nM in both human and mouse.”

To answer such questions requires the filtering and comparison of values from different data sources. This is complicated by the fact that the data sources use a variety of units to express this value. For example, even within ChEMBL, a single curated source, the IC50 value is represented in various bases of molar ranging from nano- to milli- and even in other units such as microgram per millilitre ($\mu\text{g.ml}^{-1}$).

Another group of the questions are driven by genetics, pathways, or diseases. An example from this group would be:

“For a given disease related molecular pathway, give me all targets and known active compounds that modulate them.”

These require a larger set of data sources, e.g. UniProt and WikiPathways, to be associated to compounds, targets, and the interactions between them. Similar issues arise but at a larger scale.

4. Architectural Decisions

The OPS platform provides a semantic platform for pharmacology that integrates data from a set of distributed open data sources.

Figure 1 provides an overview of the OPS Platform architecture as it is currently implemented. It consists of seven components:

1. Data Sources
2. Linked Data Cache (LDC)
3. Identity Resolution Service (IRS)
4. Identity Mapping Service (IMS)
5. Domain specific services

6. Core API

7. User Interfaces/Applications

These components arise out of corresponding architectural decisions made variously for pragmatic, technical and social reasons. We now discuss each of these seven decisions. Components resulting from these decisions are in bold.

4.1. Rely on existing RDF-based datasets

As a result of the Linked Data movement [8,24], the Semantic Web community have made a large number of existing datasets available as RDF. Particularly relevant datasets to drug discovery are the ChEMBL RDF conversion⁸ [45], the ChEBI ontology sourced from the EBI⁹ [16], the conversion of DrugBank provided by the LODD project¹⁰ [40], the conversion of the Enzyme database sourced from UniProt¹¹, and the UniProtKB/Swissprot dataset itself¹² [4]. These **Data Sources** are the foundation of the OPS platform.

The decision to rely on existing data sources was both pragmatic and social. Pragmatically, relying on existing data allowed the project to quickly develop a working system. Socially, it encourages both originating data providers (e.g. UniProt) and third parties to continue to provide RDF-data for pharmacology sources. Such RDF conversions are crucial to the success of Open PHACTS. This decision however has a cost, as data sources will use different schemes instead of one common global schema. Additionally, OPS is reliant on data providers continually updating RDF versions of important datasets. For example, an early version of the OPS platform used the Chem2Bio2RDF conversion of ChEMBL which has not been kept up-to-date with the underlying ChEMBL database from the EBI¹³. This severely limited the data available to the OPS platform and caused a great deal of confusion

⁸http://semantics.bigcat.unimaas.nl/chembl/v13_ops/ accessed 17 Sept 2012

⁹<ftp://ftp.ebi.ac.uk/pub/databases/chebi/ontology/chebi.owl> accessed 17 Sept 2012

¹⁰http://www4.wiwiw.fu-berlin.de/drugbank/drugbank_dump.nt accessed 17 Sept 2012

¹¹ftp://ftp.uniprot.org/pub/databases/uniprot/current_release/rdf/enzyme.rdf.gz accessed 17 Sept 2012.

¹²<http://www.uniprot.org/uniprot/?query=reviewed%3ayes&force=yes&format=rdf> accessed 17 Sept 2012

¹³<https://www.ebi.ac.uk/chembl/> accessed 17 Sept 2012

to our internal users. We have now moved to a more recent conversion and are working with the EBI to incorporate the RDF conversion process into their data release pipeline.

4.2. Centralize the data

A classic trade-off in the design of data integration systems is whether to federate queries or warehouse data. In this case, we chose to warehouse the data for reasons of reliability and performance. In terms of performance, interactive query speeds are necessary, thus, there may be significant latency overheads when relying on remote services (e.g. SPARQL endpoints). Furthermore, third party services (i.e. external to the consortium) cannot and should not be relied upon to provide consistent access. Essentially, it would be impolite for us to make use of others resources for answering SPARQL queries especially when they may be large or frequent. To address these concerns, the aforementioned datasets are centralized into a **Linked Data Cache (LDC)**. The term cache is used to emphasize that the platform does not maintain data on its own and is just a temporary store to enable data to be queried quickly at interactive speeds.

An issue with caching the data is ensuring that the datasets reflect the latest versions of the underlying datasets. However, it is common for scientific datasets to follow release cycles and these can be used to reload the data in the cache. See for example UniProt [14].

4.3. Separate keyword search and structured queries

A key entry point into the platform is through keyword search. In the pharmacology domain, this is more than just text matching as keywords can often match to multiple often very distinct concepts. For example, when typing “*menthol*” the user could quite reasonably mean the chemical menthol, or the menthol receptor protein. This sort of conceptual keyword search is different from queries over structured databases, thus, a new architectural component was introduced, the **Identity Resolution Service (IRS)**. The role of the IRS is to translate user-entered entity names (in free text form) into known entities within the system (i.e. that have a defined URI). These known entities can then be used in structured queries. The IRS also supports the disambiguation of concepts through interaction with the user interface.

4.4. Equality is context dependent

Datasets often provide links to equivalent concepts in other datasets. These result in a profusion of “equivalent” identifiers for a concept. Identifiers.org provide a single identifier that links to all the underlying equivalent dataset records for a concept. However, this constrains the system to a single view of the data, albeit an important one.

A novel approach to instance level links between the datasets is used in the OPS platform. Scientists care about the types of links between entities: different scientists will accept concepts being linked in different ways and for different tasks they are willing to accept different forms of relationships. For example, when trying to find the targets that a particular compound interacts with, some data sources may have created mappings to gene rather than protein identifiers: in such instances it may be acceptable to users to treat gene and protein IDs as being in some sense equivalent. However, in other situations this may not be acceptable and the OPS platform needs to allow for this dynamic equivalence within a scientific context. As a consequence, rather than hard coding the links into the datasets, the OPS platform defers the instance level links to be resolved during query execution by the **Identity Mapping Service (IMS)**. Thus, by changing the set of dataset links used to execute the query, different interpretations over the data can be provided.

4.5. Leverage domain-specific services

There are a variety of important pharmacological operations that are specific to a domain and have reliable and performant implementations. Thus, instead of creating new implementations, the OPS platform relies on these existing **domain-specific services**. For example, of critical importance for drug discovery is that identical small molecule compounds from across different data sources are integrated accurately, based on structure rather than name or identifier mapping which is less accurate. Instead of reimplementing this feature, we rely on an existing chemistry registration and normalisation service: ChemSpider [37]. This takes each compound from all of the data sources and maps it using structure-based methods to a unique ChemSpider identifier. The mappings between each chemical in each data source and ChemSpider are then loaded into the system to provide an accurate set of chemical mappings between different databases.

Going forward, we aim to make use of the SADI Framework for accessing additional domain specific services [12].

4.6. Provide a simple API

The initial prototype of the OPS platform only provided a SPARQL endpoint through which the integrated data could be queried. This required each of the drug discovery applications to have an intimate knowledge of the data exposed and the ability to write the required SPARQL queries to retrieve the data desired. This approach also left the OPS platform exposed to poorly formed queries or simply queries that are too open ended (e.g. a `select * where {?s ?p ?o}`). This both impacted developer productivity and hurt the reliability of the service.

To address this problem, an additional component, the **Core API** was introduced into the architecture. The Core API provides a set of common methods that applications can call. This benefits application developers as they no longer need to formulate their own SPARQL queries.

The OPS Core API does more than to simply provide an abstraction layer for the application developer. Such interfaces provide assurances that the data layer will only be exposed to well designed queries, and allows the API developer to make extensive use of optimizations present in RDF stores. We believe this to be key to the successful deployment of Semantic Web infrastructures which consume large scale datasets.

4.7. Leverage the user interface

A key driver of the OPS platform is to enable user-focused drug discovery applications. Thus, the entirety of the OPS platform architecture is designed around enabling these sorts of applications. This also means that we can take advantage of the user interface, for example, by encapsulating queries in an API, thus ensuring that only well-defined queries are run, or using the user interface to ensure that correct URIs are given to the system when performing a query. We also believe that this decision will allow for scalability in the future. However, there is a trade-off as this means that the OPS platform in its current form is not designed for other tasks such as large offline analytics.

5. Open Pharmacology Space Implementation

This section provides details on the implementation of the OPS prototype platform as shaped by the ar-

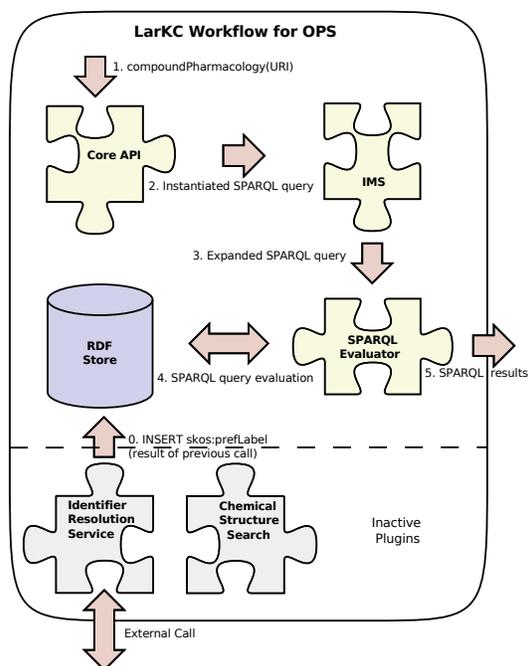


Fig. 2. LarKC plugins and workflow activation for a `compoundPharmacology(URI)` API call.

chitectural decisions given in Section 4 and the needs of the drug discovery scientists in the Open PHACTS consortium. To clarify the platform architecture, consider the simplest of the three example pharmacology research questions provided in Section 3:

“Retrieve all information available about aspirin.”

This can be achieved through the use of two OPS core API methods: `compoundLookup(text)` and `compoundPharmacology(URI)`.

The IRS is invoked by the core API through the `compoundLookup("aspirin")` method and provides a number of alternatives, including *“Aspirin”*, *“FIORINAL”* (which is a compound that contains aspirin), and *“Hyppirin”* (which is an aluminium complex ion of aspirin). The returned items are each annotated with additional metadata (e.g. descriptions and synonyms) to aid the users in their decision.

Once the user selects a particular concept, e.g. Aspirin (`cw:aspirin`¹⁴), the user interface will make

¹⁴We use the notation `cw:aspirin` to represent the URI <http://www.conceptwiki.org/concept/dd758846-1dac-4f0d-a329-06af9a7fa413/> accessed 17 Sept 2012.

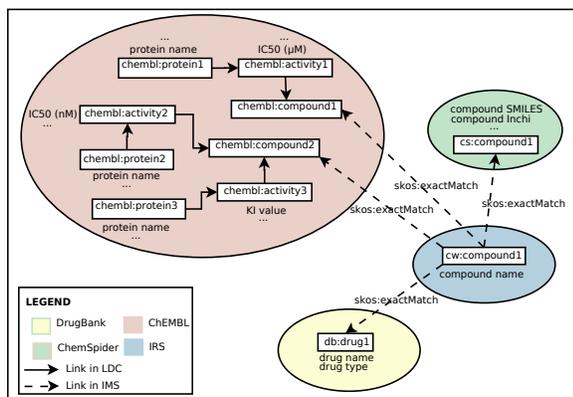


Fig. 3. Sample results of a `compoundPharmacology(URI)` API call.

the call `compoundPharmacology(cw:aspirin)`. The workflow to evaluate this method call is depicted in Figure 2 and is executed as follows¹⁵. First the compound URI is inserted into the query which is then expanded with equivalent URIs provided by the IMS. The final query is evaluated against the Linked Data Cache. An example result of this API call is shown in Figure 3, which shows the concept identified by the URI retrieved from the IRS and its links to concepts in ChemSpider, ChEMBL, and DrugBank. Note that it is possible for there to be multiple mappings within a single dataset. The figure provides example properties retrieved from each dataset, e.g. SMILES and InChI from ChemSpider, and activity values from ChEMBL.

5.1. The Large Knowledge Collider

The platform is implemented as a connection of a number of separate software systems. These software systems are coordinated using the Large Knowledge Collider (LarKC), exploiting its pluggable architecture to call external services [13].

Workflows in LarKC are defined in terms of their input, output and the connections between the plugins that participate in them. In addition, LarKC endpoints are a special type of plugin which expose an interface for receiving input and providing output, accessible through HTTP. A valid workflow must contain a single endpoint and a single plugin connected to its output. When the execution of all participating plugins

is completed, the endpoint collects any output from the workflow and responds to the original HTTP request.

Figure 2 presents the LarKC plugins that participate in the main OPS workflow, illustrated as jigsaw puzzle pieces. The figure gives a trace of the information flow between the various plugins as a result of a `compoundPharmacology(URI)` call to the OPS API.

5.2. Linked Data Cache

Additionally, LarKC acts as the Linked Data Cache (LDC) component within the OPS platform. During the OPS platform implementation, an alternative data layer was developed for LarKC, which implements the OpenRDF Sesame SAIL interface [9]. SAIL stands for “Storage and Inference Layer” and is a connection oriented interface that is implemented by a number of RDF stores to enable their interoperability with the Sesame platform. The SAIL implementation has been contributed to the LarKC open source codebase, enabling the deployment of LarKC with a wide range of alternatives to the default BigOWLIM (which is now deprecated and replaced by Ontotext with OWLIM-SE). More importantly it allows OPS users and developers to set up local OPS instances with an RDF store that is specifically selected to suit their particular data needs and hardware specifications. For example an offline demonstration instance to be installed on a mid-range laptop using a small sample dataset can avoid the installation and configuration overhead of a scalable RDF store by using an in-memory alternative.

We have successfully tested the integration of LarKC with four additional RDF stores: the default Sesame in-memory store [9], bigdata [43], OpenLink Virtuoso [17], and Garlik’s 4store [23]. Currently we use the Sesame in-memory store as a number of instability issues were discovered with the remaining alternatives. The in-memory store is deployed on a server with 392 GB of RAM, and an overview of the data currently loaded in the LDC is given in Table 1. As we are fast approaching the limitations of this solution, we are now in the process of moving to a hosted solution provided by OpenLink¹⁶, who have recently joined the Open PHACTS consortium. The Virtuoso triplestore will be enhanced to meet the challenges of scale and complexity faced in deploying the OPS platform.

In addition to the main OPS workflow, the current prototype makes use of an auxiliary workflow for

¹⁵Note that the IRS plugin was used to support the `compoundLookup(text)` call. The Chemical Structure Search plugin will be discussed in Section 5.5

¹⁶<http://www.openlinksw.com/> accessed 17 Sept 2012

| Data property | Value |
|---------------------|-------------|
| Distinct resources | 93 118 400 |
| Distinct predicates | 293 |
| Named graphs | 11 |
| Total quads | 460 037 674 |
| Size on disk | 63 GB |
| Size in RAM | 292 GB |

Table 1

Characteristic properties of RDF data currently present in the LDC.

which the IMS, IRS and external services are not used. The purpose of the workflow is to identify the URIs for which data is available in OPS from a large list of candidates. As RDF stores will provide the worst-case performance when queried for URIs that do not appear in their indices (as they do not exist), such queries benefit from circumventing the IMS.

5.3. Identity Resolution Service

As discussed in Section 4.3 the mapping between user provided text and URIs is a non trivial task that has to take into account the type of entity the user is interested in. This functionality is thus decoupled from the LDC and is provided by the Identity Resolution Service (IRS) plugin.

This plugin is able to make calls to an API provided by ConceptWiki¹⁷ to retrieve a list of resources of the required type with canonical labels curated by the community. The response is then dynamically imported in the underlying RDF store. The IRS plugin responds to explicitly defined predicates in the SPARQL query¹⁸ which are also used to relate the resulting RDF to existing data in the LDC. Updates take place immediately, making the result of the external call directly available to the SPARQL query that triggered it, but also to all future queries.

If such predicates are not present in the query the plugin remains inactive, as is the case in the example of Figure 2. However, the example call will retrieve information dynamically asserted by the IRS plugin during a previous `compoundLookup` call—namely the object of the `skos:prefLabel` predicate (arrow #0 in the figure).

¹⁷<http://ops.conceptwiki.org> accessed 17 Sept 2012.

¹⁸`PREFIX :<http://wiki.openphacts.org/index.php/ext_function#>`

5.4. Identity Mapping Service

The IMS plugin responds to all queries provided by the API endpoint and invokes the IMS service for any instance URIs that appear in the query. Once an appropriate list of mappings for each URI has been retrieved from the IMS service, the plugin will expand the original SPARQL query to include equivalent URIs.

The expanded query is then forwarded to the SPARQL Evaluator plugin which will query the RDF store and write to the output of the workflow. The need to store equivalent URIs in the LDC is thus eliminated, and the platform is able to provide appropriate results to identical queries with different entity equivalence requirements.

5.5. Domain Specific Services

So far only a single domain-specific service has been integrated with the OPS platform – namely the Chemical Structure Search service provided by ChemSpider. This service enables the retrieval of small molecule compound URIs that match, contain, or are similar to a user provided chemical structure.

As such functionality requires the use of highly specialised and computationally expensive methods and algorithms, we developed a LarKC plugin that invokes the ChemSpider API, dynamically converts search results to RDF and updates the LDC. The inserted data is persistent and reused to answer queries with identical parameters.

Similar to the IRS plugin, the Chemical Structure Search will scan each SPARQL query it receives for a set of predefined predicates which correspond to the types of search supported by ChemSpider.

5.6. Core API

As a key focal point of the OPS platform is to enable the rapid development of user-focused applications in the drug discovery domain, we have developed a simple REST-based API to facilitate the retrieval of data from the LDC. Applications can thus be developed without a requirement for extensive knowledge of the schemas used in each underlying dataset. While the application developers are limited to the methods of the core API, these have been collaboratively defined to meet their functionality needs.

The core API is implemented as an endpoint in LarKC that will receive the input to each OPS workflow, and extract the method name and correspond-

ing parameters from the HTTP POST request. Each method name corresponds to a parameterized SPARQL query which is instantiated using the parameters provided, e.g. a URI corresponding to the compound to lookup. The resulting query is provided as input to all plugins directly connected to the endpoint in the workflow (i.e. the IMS, IRS and Chemical Structure Search plugins).

The parameterized SPARQL queries are used to join the data across the datasets at a schema level, i.e. the specific columns to retrieve from each of the datasets is determined by the SPARQL query issued. For each data retrieval task, e.g. `compoundLookup`, `targetLookup`, etc., the set of columns to retrieve from each of the datasets has been determined in consultation with the scientific users of the platform. The addition of new datasets and the evolution of existing schemas are facilitated since the changes to SPARQL queries are made at a single point. Additionally, the parameterized SPARQL queries have been optimized for performance. This is particularly important due to the potential for large volumes of data and the complexity of performing query time mapping of identifiers.

5.7. User interfaces

Within the Open PHACTS project there are several exemplar user interface applications being developed each with a different drug discovery task as a focus. The OPS platform provides a common semantic web platform which can be accessed by these drug discovery applications through the core API.

For a general browsing interface over the linked data, a version of an internal system used in the pharmaceutical company Lundbeck, named LSP4All, was extended and connected to the platform. This interface, now branded the OPS Explorer, allows for searching, retrieving and browsing of data. As a result some of the typical user interfaces experienced by a pharmaceutical company researcher have been converted to rely on Semantic Web data. Figure 4 shows the OPS Explorer returning the integrated information about the compound aspirin.

In addition to browsing and searching for data, researchers utilize scientific literature in their work. To support this process the Utopia Documents [1] software was connected to the platform to allow users to view chemical compounds associated with a PDF article, see Figure 5.

Finally, a more specialized interface for curating biological pathways, PathVisio [42], was connected to

the platform, see Figure 6. This connection allowed PathVisio to provide additional pathway information to the user as they review a pathway.

Thus, based on the same common platform, we have created a prototype for an enriched pharmacology workspace software platform. Additional exemplar applications using the platform are being developed by the Open PHACTS consortium.

6. Conclusions

We presented the architectural decisions for implementing a linked data platform to support drug discovery that integrates the pre-competitive openly available data from many semantic web sources. The motivation for the functionality provided by the OPS linked data platform was driven by real business questions for drug discovery in pharmacology gathered from the Open PHACTS partners. The platform supports different views over the data through stand-off mappings which are applied at query time, i.e. it is possible to have the system relate genes and proteins in one case but not another. The versatility of the platform has been demonstrated by the wide variety of drug discovery applications that can be connected to the integrated data.

Future work is to expand the functionality of the system to a wider set of the business questions. This will also require linking a larger number of data sources, and thus increasing the size and complexity of the linked data cache.

An alpha version of the OPS platform is currently available to the Open PHACTS consortium. A first public release will be made in late 2012, see <http://www.openphacts.org/> for details.

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Open PHACTS Explorer

Compound by name: Aspirin

Compound name: Aspirin

Compound by Name search results

Aspirin

Pharmacology Data | View in ChemBioNavigator

The prototypical analgesic used in the treatment of mild to moderate pain. It has anti-inflammatory and antipyretic properties and acts as an inhibitor of cyclooxygenase which results in the inhibition of the biosynthesis of prostaglandins. Aspirin also inhibits platelet aggregation and is used in the prevention of arterial and venous thrombosis. (From Martindale, The Extra Pharmacopoeia, 30th ed, p5)

Aspirin is rapidly hydrolyzed primarily in the liver to salicylic acid, which is conjugated with glycine (forming salicylic acid) and glucuronic acid and excreted largely in the urine.

ChemSpider ID: 2157
Molecular Formula: C₉H₈O₄
SMILES: CC(=O)Oc1ccccc1C(=O)O
Standard InChI: InChI=1S/C9H8O4/c1-c(10)13-8-5-3-2-4-7(8)9(11)12/h2-5H,1H3,(H,11,12)
Standard InChIKey: BSYNRYMUTXBKSQ-UHFFFAOYSA-N

Affected Organism: Humans and other mammals

Indications: For use in the temporary relief of various forms of pain, inflammation associated with various conditions (including rheumatoid arthritis, juvenile rheumatoid arthritis, systemic lupus erythematosus, osteoarthritis, and ankylosing spondylitis), and is also used to reduce the risk of death and/or nonfatal myocardial infarction in patients with a previous infarction or unstable angina pectoris.

Protein Binding: High (99.5%) to albumin. Decreases as plasma salicylate concentration increases, with reduced plasma albumin concentration or renal dysfunction, and during pregnancy.

Melting Point: 135 °C (boiling point 140 °C)

AllogP: 1.2
H-Bond Receptors: 4
H-Bond Donors: 1
Mol Weight: 180.157
MW Freebase: 180.157
Polar Surface Area: 63.6
Rotatable Bonds: 3

Fig. 4. Compound by name look up through the OPS Explorer interface

Utopia Documents – Development of a Resistance-like Phenotype to Sorafenib by Human Hepatocellular Carcinoma Cells Is Reversible and Can Be Delayed...

Neoplasia Vol. 12, No. 11, 2010 | Preclinical Model of Sorafenib Resistance in HCC Tang et al. 931

Figure 1. Effect of sorafenib regimens on orthotopically implanted human Hep3B cells in SCID mice. Hep3B cells were tagged with βHCG and implanted into the left lobe of the liver (n = 5 per treatment group). (A) Urine βHCG levels of pooled samples, normalized to urine creatinine levels, of mice treated with sorafenib monotherapies (at 15, 30, or 60 mg/kg per day by gavage) or sorafenib (at 30 mg/kg per day) combined with either metronomic CTX (20 mg/kg per day, per os) or metronomic UFT (15 mg/kg per day by gavage). Treatments were carried out during the periods indicated with boxes in **Figure 1B**, ceasing starting when toxicity (i.e., >10% average body weight loss) was observed in each treatment group. The graph also shows the beneficial effect of the sorafenib regimens compared with controls treated by gavage of vehicle alone, on the relative tumor burden. Green asterisk indicates a mouse treated with 15 mg/kg per day of sorafenib that died at that time point. Red asterisk indicates a mouse treated with 60 mg/kg per day of sorafenib that died at that time point. (B) Corresponding weight measurements of the tumor mass in the liver of mice in each treatment group at the time of sacrifice (note that because of the different treatment efficacies, mice from the groups were killed at different weeks after the initiation of treatment, as indicated). (C) Corresponding mouse weight curves as a measure of relative toxicity of the different sorafenib treatment regimens. Note that for individual groups, sorafenib treatment was suspended for the indicated times whenever toxicity was observed in each group.

the following regimens: 1) 60 mg/kg per day sorafenib, 2) 30 mg/kg per day sorafenib plus UFT, and 3) 30 mg/kg per day sorafenib plus CTX. The sudden drop in βHCG levels in the 15-mg/kg per day sorafenib group on day 56 (Figure 1A) coincided with the death of one mouse in the group before urine collection. Similarly, one mouse died in the control group on day 42, and the remaining groups became necrotic 3 days later; the control group was therefore terminated on day 45 (i.e., 3 weeks after treatment initiation). For the 15-mg/kg per day sorafenib monotherapy group, all the remaining mice also began to show weight loss or manifested a moribund state by day 75, at which point mice from this group were also killed. However, all the other treatment groups did not show signs of advanced disease until day 84 (i.e., week 10 after treatment initiation), at which point the experiment was terminated, and mice from all groups were killed. The mean weights of tumor mass in the livers of the mice at the time of sacrifice (Figure 1B) were consistent with the βHCG data. Thus, they showed that the sorafenib treatment regimens caused a reduction in the tumor burden compared with controls. Taken together, these results suggest that the sorafenib regimens are effective in treating orthotopically implanted established Hep3B tumors, although the prolonged sorafenib treatment regimen using the higher doses are associated with significant host toxicity in SCID mice.

Effect of Sorafenib Treatment on Subcutaneous HCC Tumors in SCID Mice

Hep3B cells were also injected subcutaneously in SCID mice to further evaluate the effect of sorafenib treatment in this more conventional model of therapy testing (Figure 2A). Furthermore, caliper measurements of the subcutaneously growing tumors in this experiment allowed us to confirm the reliability of the βHCG marker as a readout for relative tumor burden (Figure 2B). Similar to our observations using the orthotopic tumor transplant model, we again noted that the sorafenib monotherapies we tested (at 15, 30, or 60 mg/kg per

Back | sorafenib

DrugBank
Related drugs and their targets

Wikipedia entries
Community curated encyclopedia entries

Protein Data Bank structures
Biological macromolecular structures

ChEMBL
Related bioactive small molecules

Open PHACTS
Sorafenib

C₂₁H₁₉ClF₃N₃O₃

Description: Sorafenib (rINN), marketed as Nexavar by Bayer, is a drug approved for the treatment of advanced renal cell carcinoma (primary kidney cancer). It has also received "Fast Track" designation by the FDA ... [more]

Biological Transformation: Sorafenib is metabolized primarily in the liver, undergoing oxidative metabolism, mediated by CYP3A4, as well as glucuronidation mediated by UGT1A8. Sorafenib accounts for approximately 70-85% of the circulating analytes in plasma at ... [more]

Articles from PubMed
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Fig. 5. Article enrichment in Utopia Documents using the OPS Platform

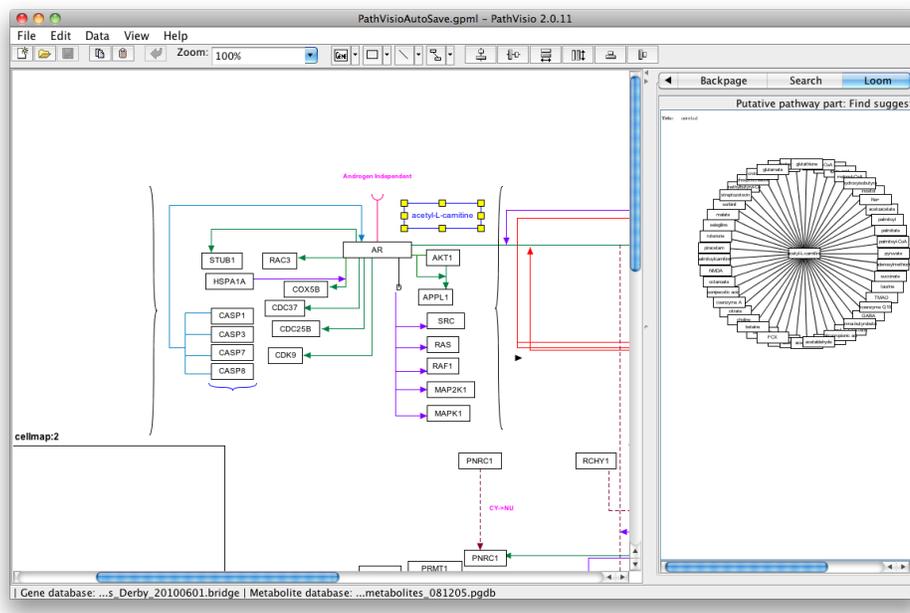


Fig. 6. PathVisio using the OPS Platform to suggest relationships with new genes or metabolites (right side window) for a gene selected in a pathway (left side window)

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