Topic 9. Data integration

Why, how, promise and pitfalls. Standards, ontologies, ID mapping and metadata

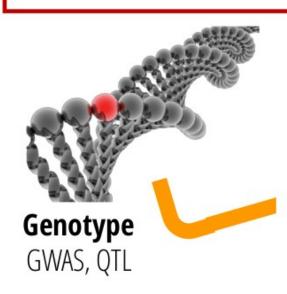
The Genotype to Phenotype challenge

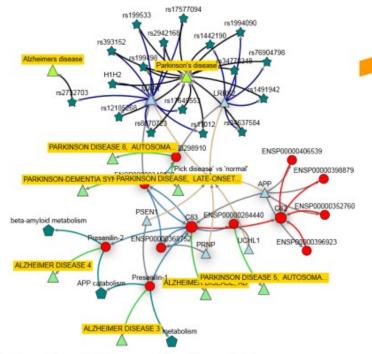
Topic 5 - Genes & Genomes

Topic 6 – Functional genomics

Topic 7 – Networks & Pathways

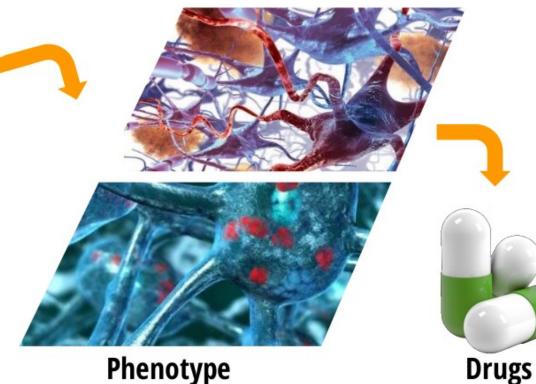
Topic 8 – Phenotypes & Diseases







Data selection, processing, transformation, integration, interpretation

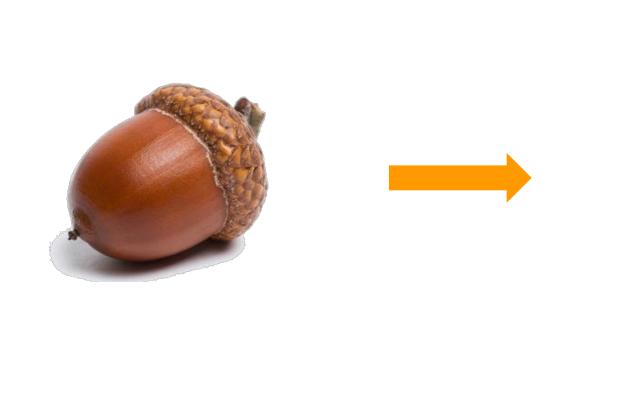


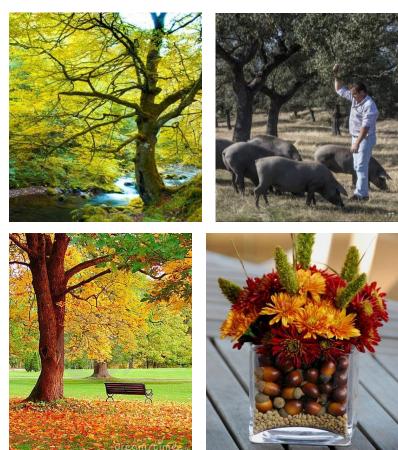
Alzhaimar Parkinso

Alzheimer, Parkinson

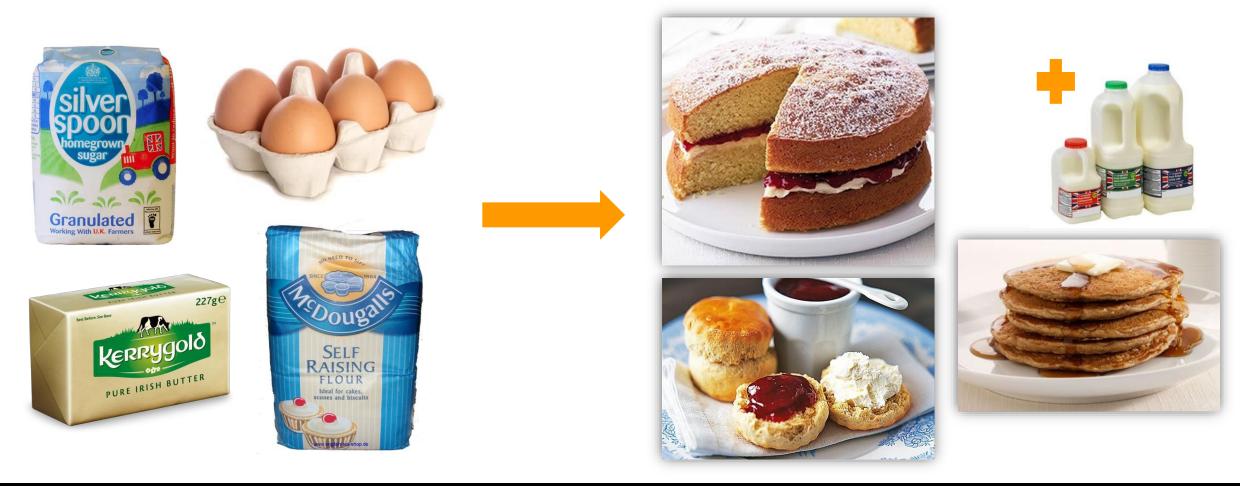
Precision medicine

1. You need to understand the limitations/context of the data you start with, to be able to get something useful from a larger integrated dataset





2. Even when you have the same data to work with, integrating it in different ways can give completely different views of the bigger biological picture

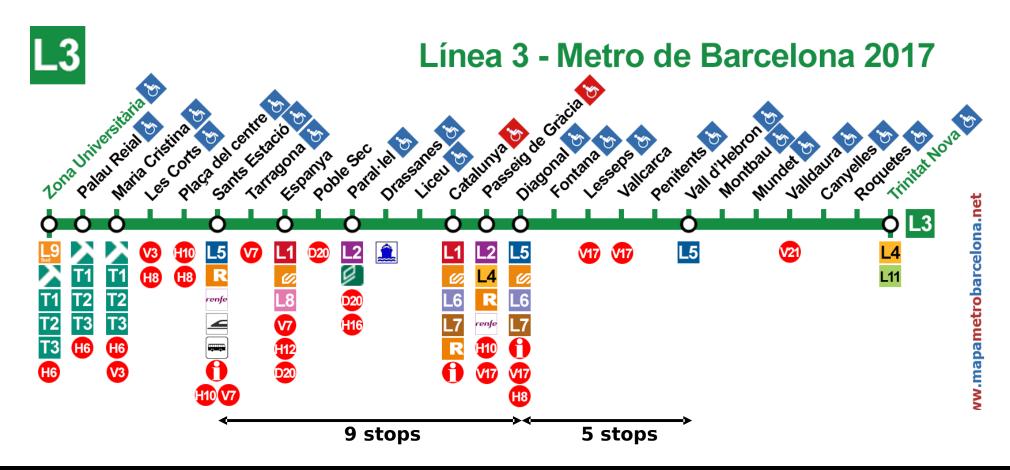


3. Even if you understand the data and have a good knowledge of the biology, integrating data without the right skills can give a poor result or false impression of reality



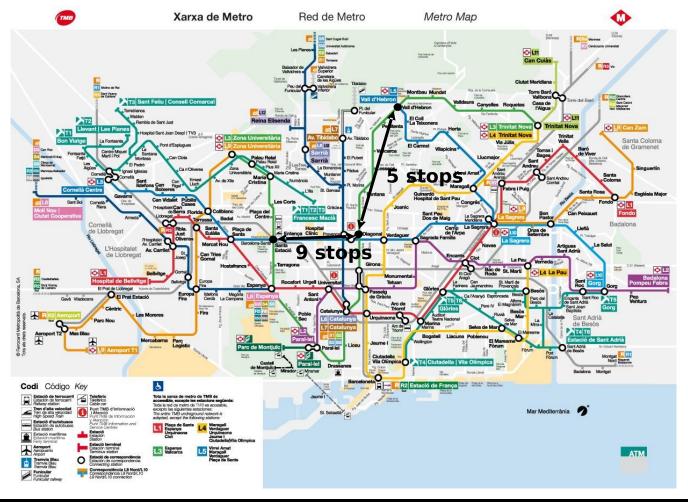


4. How (and which) data is represented has a significant effect on our understanding - aiding us in some areas, but can mis-represent the truth in others



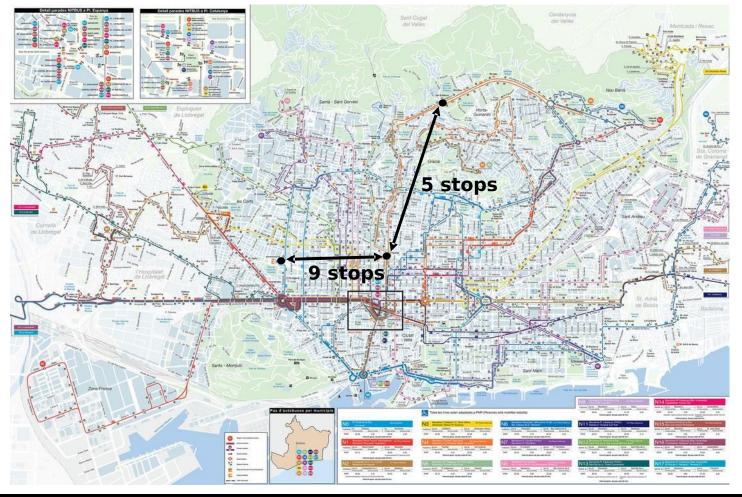
4. How (and which) data is represented has a significant effect on our understanding - aiding us in some areas, but can mis-represent the truth in

others



4. How (and which) data is represented has a significant effect on our understanding - aiding us in some areas, but can mis-represent the truth in

others



5. We do not (usually) have all the data - unless we recognize this we can be led to false conclusions.







Debunking Lunar Landing Conspiracies with Maxwell and VXGI https://www.youtube.com/watch?v=O9y_AVYMEUs

Big data, bioinformatics, and particularly data integration have the potential to help us understand biology better.

However, it never provides a definitive answer - it can be used to guide experimentation; and ideas developed from *in silico* approaches should be validated by laboratory (and preferably orthogonal) methods.

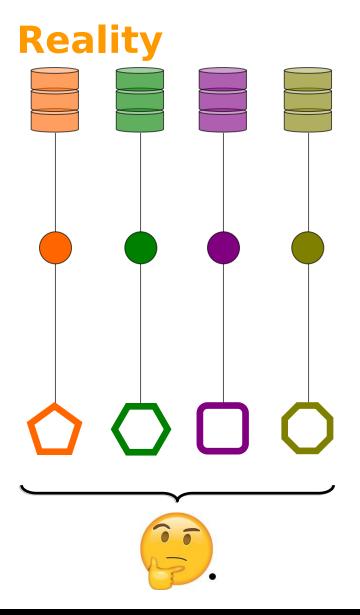
Data integration: state of the art

Ideally





Data integration: state of the art



Many data resources

- Many to maintain
- New appearing
- Few have a sustained future
- Not easy to find them

Different query interfaces

Variable results

- Formats
- Schemas
- Data content

Data integration

- Redundancy
- Inconsistency



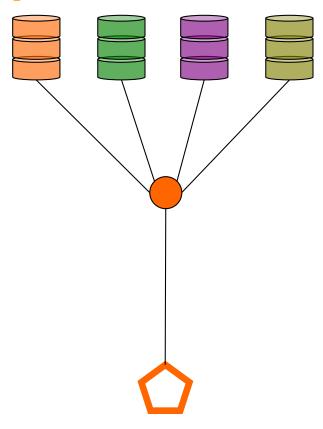
A catalogue of databases, described according to the BioDBcore guidelines, along with the standards used within them; partly compiled with the support of Oxford University Press (NAR Database Issue and DATABASE Journal).

Compiles 2062 databases, 1710 standards & 168 policies*

*20/11/2023

Data integration: state of the art

Compromise

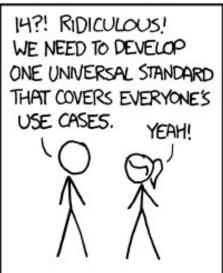


Standards & Ontologies



HOW STANDARDS PROLIFERATE: (SEE: A/C CHARGERS, CHARACTER ENCODINGS, INSTANT MESSAGING, ETC.)

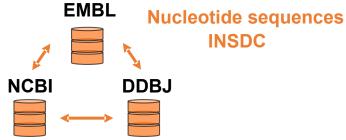
SITUATION: THERE ARE 14 COMPETING STANDARDS.





Collaboration among data providers

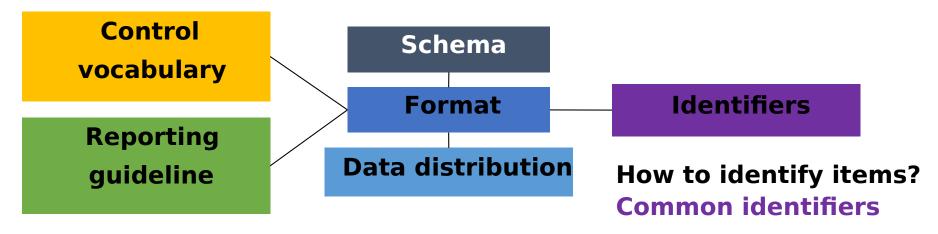
- More data coverage
- Less redundancy
- Less inconsistency
- Better data management: access, exchange, sharing, portability, interoperability, annotation, comparison, verification, representation, integration, reusability



How to annotate it?

Controlled vocabularies (ontologies)

How/where to store it?
Database, common schemas



What to store? How to put the data in context?

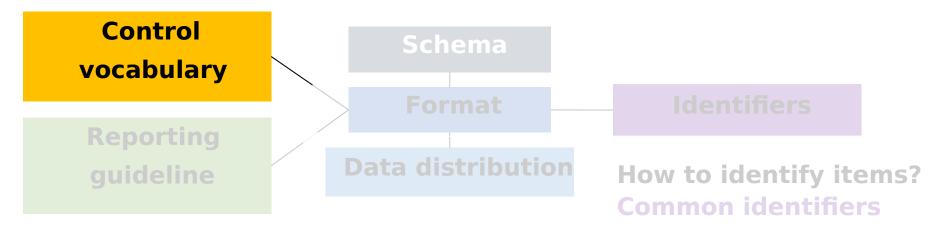
Minimum information guidelines, Metadata How to make it available?

Common formats, common query interfaces

How to annotate it?

Controlled vocabularies (ontologies)

How/where to store it?
Database, common schemas



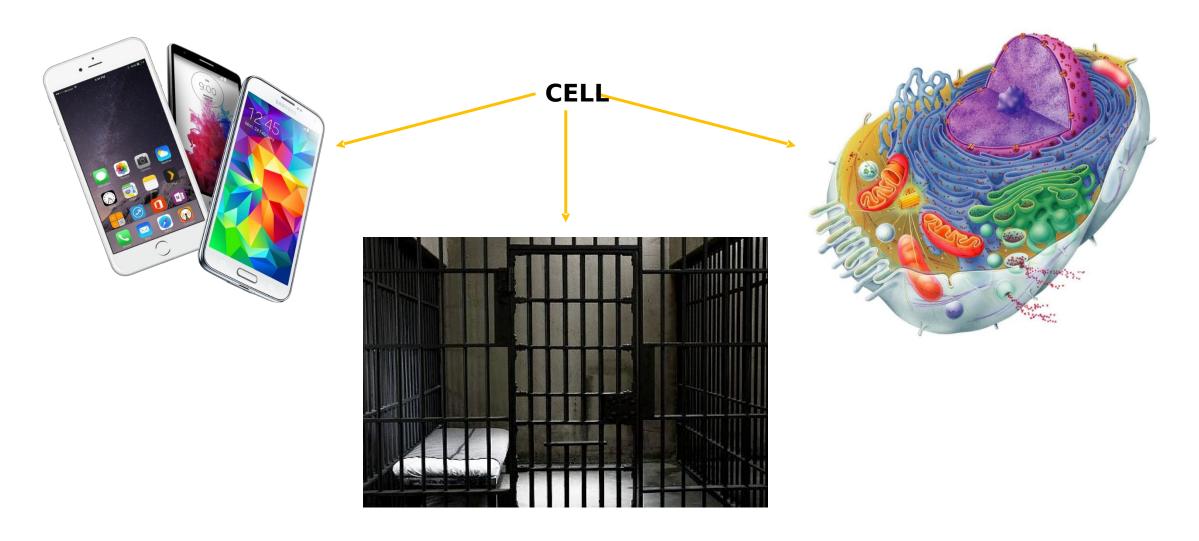
What to store? How to put the data in context?

Minimum information guidelines,
Metadata

How to make it available? Common formats, common query interfaces

Controlled vocabularies (ontologies)

• Inconsistency in natural language: same name, different concepts



Controlled vocabularies (ontologies)

- Controlled vocabularies provide a way to organize knowledge for subsequent retrieval
- Mandate the use of defined preselected terms
- The Gene Ontology (GO)
- A way to capture biological knowledge for individual gene products in a written and computable form
- A set of concepts and their relationships to each other arranged as a hierarchy

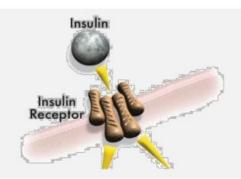


http://www.geneontology.org/

Controlled vocabularies: The Gene Ontology (GO)

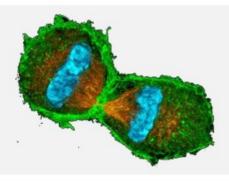
1. Molecular Function

An elemental activity or task or job



- protein kinase activity
- insulin receptor activity

· cell division

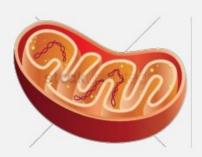


2. Biological Process

A commonly recognized series of events

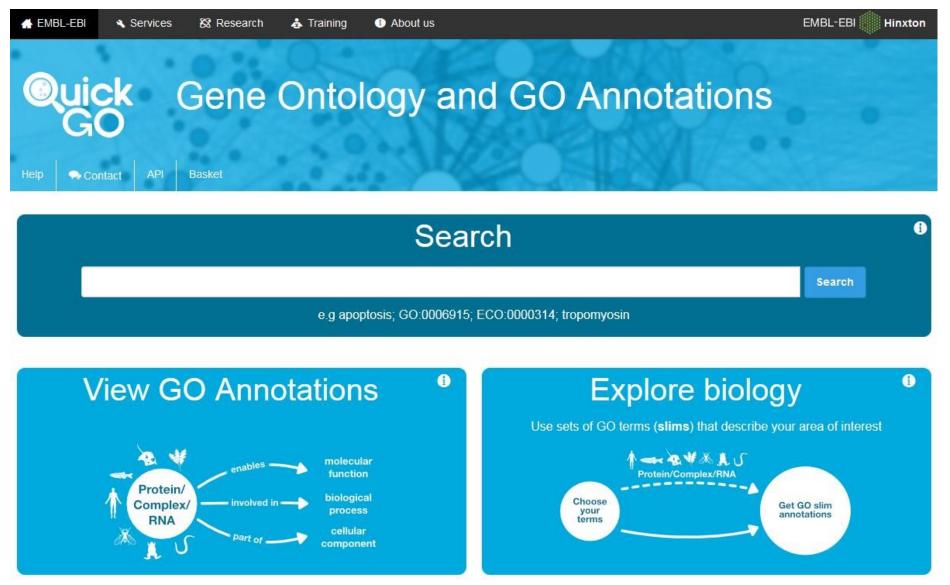
3. Cellular Component

Where a gene product is located



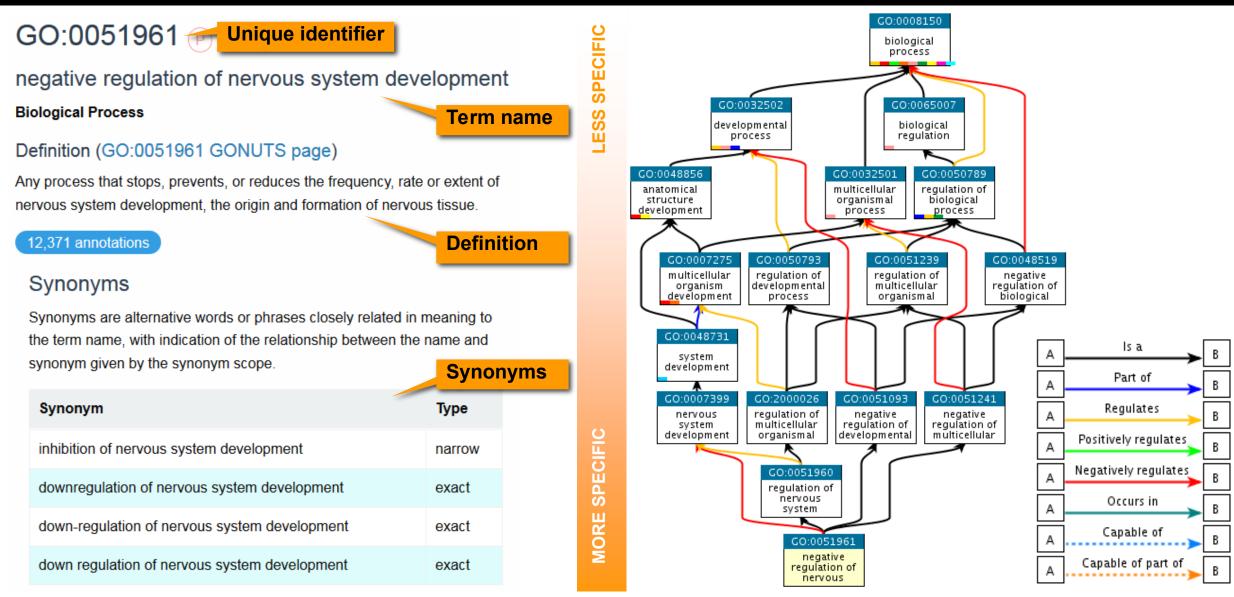
- mitochondrion
- mitochondrial matrix
- · mitochondrial inner membrane

Controlled vocabularies: The Gene Ontology (GO)



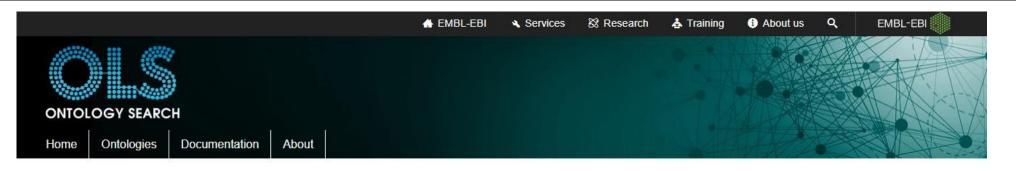
https://www.ebi.ac.uk/QuickGO/

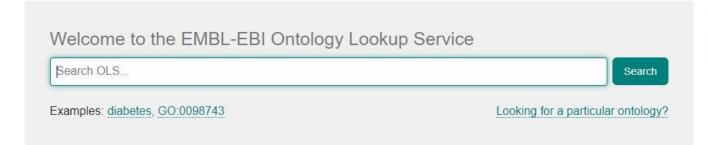
Controlled vocabularies: The Gene Ontology (GO)



https://www.ebi.ac.uk/QuickGO/term/GO:0051961

Controlled vocabularies: The Ontology Lookup Servi





About OLS

The Ontology Lookup Service (OLS) is a repository for biomedical ontologies that aims to provide a single point of access to the latest ontology versions. You can browse the ontologies through the website as well as programmatically via the OLS API. OLS is developed and maintained by the Samples, Phenotypes and Ontologies Team (SPOT) at EMBL-FBI

Related Tools

In addition to OLS the SPOT team also provides the OxO, Zooma and Webulous services. OxO provides cross-ontology mappings between terms from different ontologies. Zooma is a service to assist in mapping data to ontologies in OLS and Webulous is a tool for building ontologies from spreadsheets.

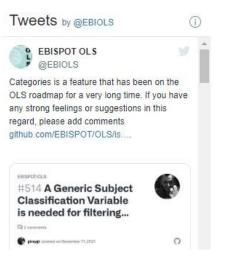
A Report an Issue

For feedback, enquiries or suggestion about OLS or to request a new ontology please use our GitHub issue tracker. For announcements relating to OLS, such as new releases and new features sign up to the OLS announce mailing list

■ Data Content

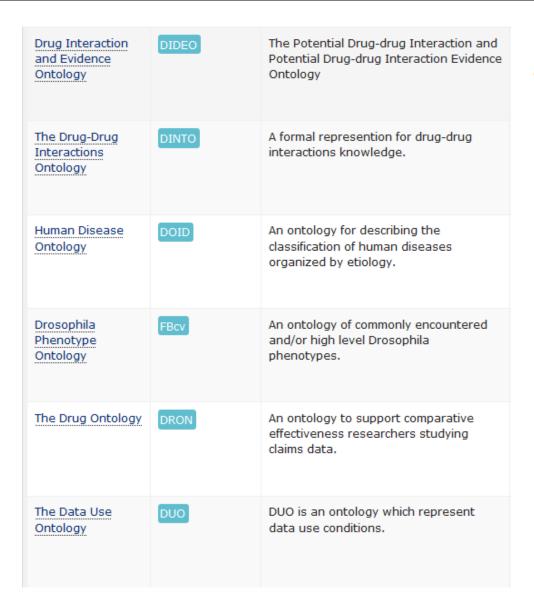
Updated 19 de nov. 2021 06:28

- · 268 ontologies
- 6.940.171 terms
- · 34,337 properties
- 492.213 individuals



https://www.ebi.ac.uk/ols/index

Controlled vocabularies: The Ontology Lookup Service (OLS)



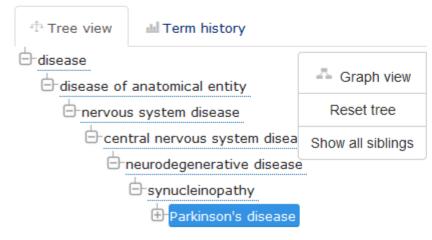


Search DOID Q

http://purl.obolibrary.org/obo/DOID_14330

A synucleinopathy that has_material_basis_in degeneration of the central nervous system that often impairs motor skills, speech, and other functions. [url:http://en.wikipedia.org /wiki/Parkinson%27s_disease]

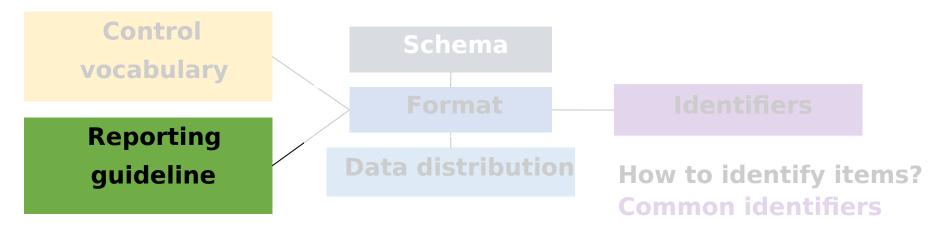
Synonyms: Parkinson disease, paralysis agitans



UMLS_CUI:C0030567

How to annotate it?
Controlled vocabularies (ontologies)

How/where to store it?
Database, common schemas



What to store? How to put the data in context?

Minimum information guidelines, Metadata How to make it available? Common formats, common query interfaces

Minimum information guidelines

- The reader can understand, and potentially reproduce, the authors experiments
- The data can be moved into a database by a manual curator
- The data can be captured, in some form, by text-mining

```
Nature Genetics 29, 365 - 371 (2001)
doi:10.1038/ng1201-365
```

Minimum information about a microarray experiment (MIAME)—toward standards for microarray data

nature biotechnology

The minimum information about a proteomics experiment (MIAPE)

Gigascience. 2: 13.

Published online . doi: 10.1186/2047-217X-2-13

The role of reporting standards for metabolite annotation and identification in metabolomic studies

PMCID: PMC3853013

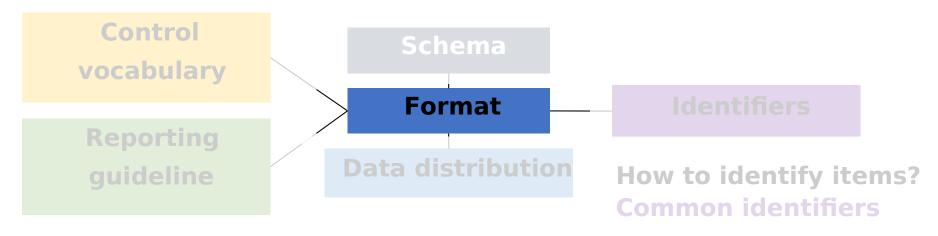
Metadata

- Metadata is data that describes other data
- In the case of experimental data, this is a description of experimental conditions; e.g., patient/environmental details, cell lines, instrumentation
- Common problems: incomplete and inconsistent metadata

37 year old male	initial phase male	male fetus	six males mixed
600 yr. old male	m	male plant	stallion
adult male	make	male, 8 weeks old	steer
bull	makle	male, castrated	sterile male
castrated male	mal e	male, pooled	strictly male
cm	male	males	tetraploide male
dioecious male	male (7-2872)	man	type i males
diploid male	male (7-3074)	men	type ii males
drone	male (m-a)	normale male	virgin male
engorged male	male (m-o)	ram	winged and wingless males
fertile male	male caucasian	rooster	young male
four males mixed	male child	s1 male sterile	, ,
individual male	male fertile	sex: male	female*

How to annotate it?
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What to store? How to put the data in context?

Minimum information guidelines,
Metadata

How to make it available?

Common formats, common query interfaces

Data formats

- Merge data from different data sources with minimal effort
- Parse the data
- Build tools to use the data
- PSI-MI XML format (molecular interactions)



Methods Mol Biol. 2016;1418:3-17. doi: 10.1007/978-1-4939-3578-9_1.

Overview of Sequence Data Formats.

Zhang H1.

Author information

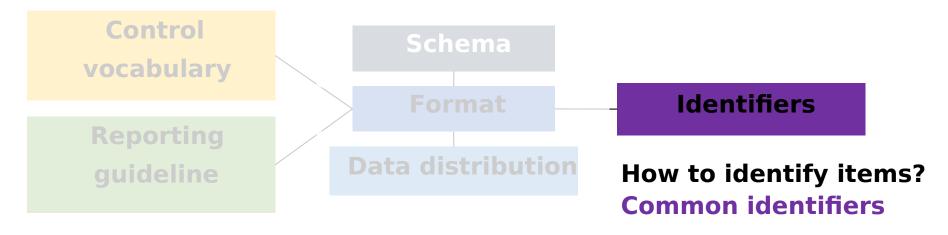
Abstract

Next-generation sequencing experiment can generate billions of short reads for each sample and processing of the raw reads will add more information. Various file formats have been introduced/developed in order to store and manipulate this information. This chapter presents an overview of the file formats including FASTQ, FASTA, SAM/BAM, GFF/GTF, BED, and VCF that are commonly used in analysis of next-generation sequencing data.

KEYWORDS: BED; FASTA; FASTQ; GFF/GTF; Next-generation sequencing; SAM/BAM; Sequencing data; Sequencing data file format; VCF

How to annotate it?
Controlled vocabularies (ontologies)

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What to store? How to put the data in context?

Minimum information guidelines,
Metadata

How to make it available? Common formats, common query interfaces

- Identifiers: The use of identifiers allows for unambiguous identifications of molecules or conceptual entities, and their representation in databases
- ID mapping: There is a large number of identifiers that aim to represent the same entity

ENSP00000269305

P04637

NP 001119584.1

P53_HUMAN

HGNC:11998

MEEPQSDPSVEPPLSQETFSDLWKLLPENNVLSPLPSQAMDDLMLSPDDIEQWFTEDPGP DEAPRMPEAAPPVAPAPAAPTPAAPAPAPSWPLSSSVPSQKTYQGSYGFRLGFLHSGTAK SVTCTYSPALNKMFCQLAKTCPVQLWVDSTPPPGTRVRAMAIYKQSQHMTEVVRRCPHHE RCSDSDGLAPPQHLIRVEGNLRVEYLDDRNTFRHSVVVPYEPPEVGSDCTTIHYNYMCNS SCMGGMNRRPILTIITLEDSSGNLLGRNSFEVRVCACPGRDRRTEEENLRKKGEPHHELP PGSTKRALPNNTSSSPQPKKKPLDGEYFTLQIRGRERFEMFRELNEALELKDAQAGKEPG GSRAHSSHLKSKKGQSTSRHKKLMFKTEGPDSD

NX P04637

Antigen NY-CO-13

Cellular tumor antigen p53

TP53

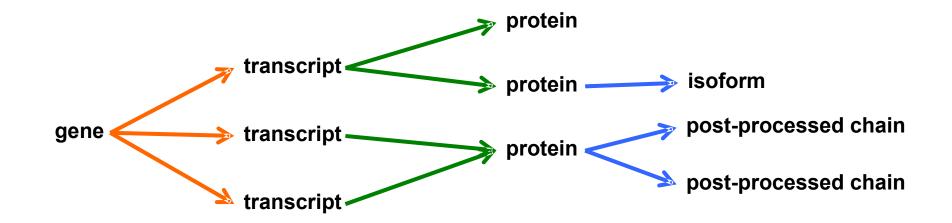
P04637-1

uc002gii.2

CHEMBL2221344

- Most commonly used accessions:
 - Entrez GenelDs: Gene-centered identifier: DNA consensus sequence, no isoform or variants.
 - **UniProt**: Represents proteins, taking into account isoforms. Additional identifiers for variants and post-processed chains.
 - RefSeq: Represents sequences of DNA, RNA and proteins.
 - **Ensembl:** Identifiers that represent genes and their different products: gene, gene tree, protein, regulatory feature, transcript, exon and protein family.
 - International Protein Index: Proteomics reference database (protein sequences). Now obsoleted, but still used in proteomics.
 - **HUGO gene symbols:** Unique symbols and names for human loci (protein-coding genes, RNA genes and pseudogenes).
 - Organism centered databases: FlyBase, TAIR, WormBase, SGD, ...

Identifiers most of the times do not have a 1:1 relationship: gene ≠ transcript ≠ protein



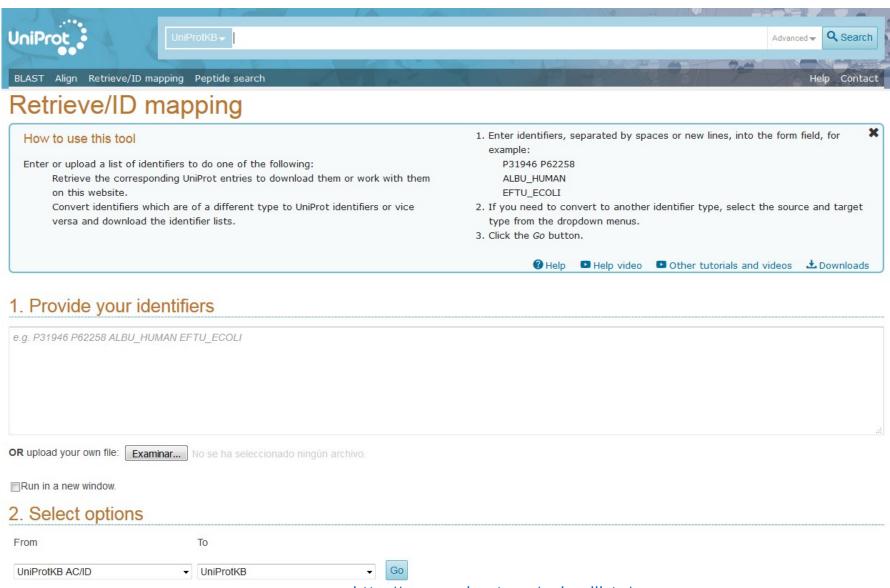
- Identifiers (and sequences!) disappear and get updated
- Identifiers are **misused**; *e.g.*, gene identifiers used to represent proteins

ID mapping tools / mapping tables

UniProt ID mapping
Ensembl BioMart
DAVID GeneID Conversion Tool
HGNC project

- https://www.uniprot.org/uploadlists/
- https://www.ensembl.org/biomart/martview
- https://david.ncifcrf.gov/conversion.jsp
- http://www.genenames.org/

- Common problems:
 - You are not able to map all the IDs
 - You get 1:n mappings



http://www.uniprot.org/uploadlists/

Frameworks for finding and mapping equivalent database identifiers for genes, proteins, and metabolites

BridgeDb

- Cytoscape app: http://apps.cytoscape.org/apps/bridgedb
- bridgeDbR Bioconductor package: <u>http://bioconductor.org/packages/devel/bioc/html/BridgeDbR.html</u>
 Web service API: http://webservice.bridgedb.org/



BioMart

- Cytoscape app: http://apps.cytoscape.org/apps/biomartwebserviceclient
- biomaRt Bioconductor package: https://bioconductor.org/packages/release/bioc/html/biomaRt.html
- Web service API: http://www.biomart.org/martservice.html



Points to remember

- In order to merge/integrate data it needs to have as many similarities as possible – the same entity/concept described in different ways will not merge
- Data transfer/conversion almost always means data loss if you want detail, go to the source
- More resources does not necessarily mean more data many databases import from the few that curate
- If using a resource which imports data check the date of the last import
- Web-services run directly from the most recent release may be a better source of data than a data-warehouse, which imports infrequently
- One advantage of database submission data will then be available in standard formats

Some final recommendations

- Curation, curation
 - Contribute to databases that accept submissions
 - Use credible data sources
- Collaboration, constant feedback, agree on standards
 - Use standard data formats
 - Use ontology terms to annotate metadata
- Openness, reproducibility
 - Use open-source software
 - Manage and share code
 - Transparent data sources and analysis procedures
 - Document as much as you can





