## FAIR Principles for AI/ML Data Sharing

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#### **Outline**



- Q1. What are FAIR Principles?
- Q2. Why FAIR Principles are important for AI/ML data sharing?
- Q3. How to implement FAIR Principles for AI/ML data sharing in precision oncology?
  - a. Initiatives to support interoperability and reusability in data collection
    - Clinical data
    - ii. Genomic data
    - iii. Imaging data
  - b. Initiatives to support findability and accessibility in data sharing
    - Network architecture
    - ii. Access control

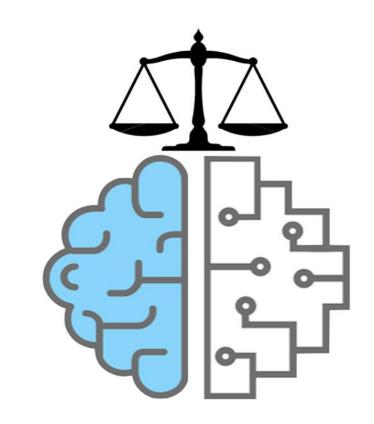
Take home messages

# **Question One:**What are FAIR Principles?



## Findability, Accessibility, Interoperability, and Reusability (FAIR)

- 'Findability' implies data can be found online, typically through indexing in search engines.
- 'Accessibility' means data can be retrieved directly or via an approval process.
- 'Interoperability' imposes data to follow standards.
- 'Reusability' requires the context of the data generation (metadata) is **documented** so it can be compared to or integrated with other data sets.



Charles Vesteghem, Briefings in Bioinformatics, 2020,



#### F1. (Meta)data are assigned a globally unique and persistent identifier.

• **Identifiers** consist of an internet link (e.g., a uniform resource identifier (URI) that resolves to a web page ).

URI http://purl.uniprot.org/uniprot/A0A022YWF9
URI pattern Local ID



• **Metadata** is data that provides information about other data, but not the content of the data.

emlployee_id	first_name	last_name	nin	department_id	Metadata		
44	Simon	Martinez	HH 45 09 73 D	1			
45	Thomas	Goldstein	SA 75 35 42 B	2			
46	Eugene	Comelsen	NE 22 63 82	2	Column	Data Type	Description
47	Andrew	Petculescu	XY 29 87 61 A	1	emlployee_id	int	Primary key of a table
48	Ruth	Stadick	MA 12 89 36 A	15	first_name	nvarchar(50)	Employee first name
49	Barry	Scardelis	AT 20 73 18	2	last_name	nvarchar(50)	Employee last name
50	Sidney	Hunter	HW 12 94 21 C	6	nin	nvarchar(15)	National Identification Number
51	Jeffrey	Evans	LX 13 26 39 B	6	position	nvarchar(50)	Current postion title, e.g. Secretary
52	Doris	Bemdt	YA 49 88 11 A	3	department_id	int	Employee department. Ref: Departments
53	Diane	Eaton	BE 08 74 68 A	1	gender	char(1)	M = Male, F = Female, Null = unknown
54	Bonnie	Hall	WW 53 77 68 A	15	employment_start_date		Start date of employment in organization.
55	Taylor	لا	ZE 55 22 80 B	1	employment_end_date		Employment end date. Null if employee st



- F1. (Meta)data are assigned a globally unique and persistent identifier.
- F2. Data are described with rich metadata.
  - **Metadata** can (and should) be generous and extensive, including descriptive information about the context, quality and condition, or characteristics of the data.
  - Rich metadata allow a computer to automatically accomplish routine and tedious sorting and prioritizing tasks.



- F1. (Meta)data are assigned a globally unique and persistent identifier.
- F2. Data are described with rich metadata.
- F3. Metadata clearly and explicitly include the identifier of the data they describe.
  - The metadata and the dataset they describe are usually **separate files**. The association between them should be made explicit by mentioning a dataset's globally unique and persistent identifier in the metadata.
  - Many **repositories** will generate globally unique and persistent identifiers for deposited datasets that can be used for this purpose.



- F1. (Meta)data are assigned a globally unique and persistent identifier.
- F2. Data are described with rich metadata (defined by R1 below).
- F3. Metadata clearly and explicitly include the identifier of the data they describe.
- F4. (Meta)data are registered or indexed in a searchable resource.
  - If the **availability** of a digital resource such as a dataset, service or repository is not known, then nobody (and no machine) can discover it. Identifiers and rich metadata descriptions alone will not ensure '**findability**' on the internet.
  - There are many ways in which digital resources can be made discoverable, including indexing.





- A1. (Meta)data are retrievable by their identifier using a standardized communications protocol.
  - Most users of the internet retrieve data by 'clicking on a link'. This is a high-level interface to a low-level protocol called **Transmission Control Protocol** (TCP), that the computer executes to load data in the user's web browser.
  - FAIR data retrieval should be mediated **without** specialized tools or communication methods. This principle focuses on how data and metadata can be retrieved from their identifiers.



## Accessibility Principle\_1.1

A1. (Meta)data are retrievable by their identifier using a standardized communications protocol.

#### A1.1 The protocol is open, free, and universally implementable.

- To maximize data reuse, the protocol should be free (no-cost) and open (-sourced) and thus globally implementable to facilitate data retrieval.
- Anyone with a computer and an internet connection can access at least the metadata. Hence, this criterion will impact your choice of the repository where you will share your data.





A1. (Meta)data are retrievable by their identifier using a standardized communications protocol.

A1.1 The protocol is open, free, and universally implementable.

A1.2 The protocol allows for an authentication and authorization procedure, where necessary.

- The 'A' in FAIR does not necessarily mean 'open' or 'free'. Rather, it implies that one should provide the exact conditions under which the data are accessible. Hence, even heavily protected and private data can be FAIR.
- Ideally, accessibility is specified in such a way that a **machine** can **automatically** understand the requirements, and then either **automatically** execute the requirements or alert the user to the requirements.

## **Accessibility Principle\_2**



- A1. (Meta)data are retrievable by their identifier using a standardized communications protocol.
  - A1.1 The protocol is open, free, and universally implementable.
  - A1.2 The protocol allows for an authentication and authorization procedure, where necessary.

#### A2. Metadata are accessible, even when the data are no longer available.

- Datasets tend to degrade or disappear over time because there is a cost to maintaining an online presence for data resources.
- Storing the metadata generally is much **easier** and **cheaper**. Hence, metadata should persist even when the data are no longer sustained.





- I1. (Meta)data use a formal, accessible, shared, and broadly applicable language for knowledge representation.
- Humans should be able to exchange and interpret each other's data. Also, data should be **readable for machines** without the need for specialized or ad hoc algorithms, translators, or mappings.
- Each computer system at least has knowledge of the other system's data exchange formats. It is critical to use
  - (1) commonly used controlled vocabularies, ontologies
  - (2) a well-defined framework to describe and structure (meta)data.

## **Interoperability Principle\_2**



I1. (Meta)data use a formal, accessible, shared, and broadly applicable language for knowledge representation.

#### 12. (Meta)data use vocabularies that follow FAIR principles.

 The controlled vocabulary used to describe datasets needs to be documented and resolvable using globally unique and persistent identifiers. This documentation needs to be easily findable and accessible by anyone who uses the dataset.

## **Interoperability Principle\_3**



- 11. (Meta)data use a formal, accessible, shared, and broadly applicable language for knowledge representation.
- 12. (Meta)data use vocabularies that follow FAIR principles.
- 13. (Meta)data include qualified references to other (meta)data.
- A qualified reference is a **cross-reference** that explains its intent.
- In particular, the scientific links between the datasets need to be described. Furthermore, all datasets need to be properly **cited**.





## R1. (Meta)data are richly described with a plurality of accurate and relevant attributes.

- The data publisher should provide not just metadata that allows **discovery**, but also metadata that richly describes the **context** under which the data was generated.
- This may include the experimental protocols, the manufacturer and brand of the machine or sensor that created the data, etc.

## Reusability Principle\_1.1



R1. (Meta)data are richly described with a plurality of accurate and relevant attributes.

#### R1.1. (Meta)data are released with a clear and accessible data usage license.

- 'I' principles covered elements of **technical interoperability**. R1.1 is about **legal interoperability**. What usage rights do you attach to your data? This should be described clearly.
- Clarity of **licensing status** will become more important with automated searches involving more licensing considerations.

## Reusability Principle\_1.2



R1. (Meta)data are richly described with a plurality of accurate and relevant attributes.

R1.1. (Meta)data are released with a clear and accessible data usage license.

#### R1.2. (Meta)data are associated with detailed provenance.

• Include a description of the **workflow** that led to your data: Who generated or collected it? How has it been processed? Has it been published before? Ideally, this workflow is described in a machine-readable format.

## Reusability Principle\_1.3

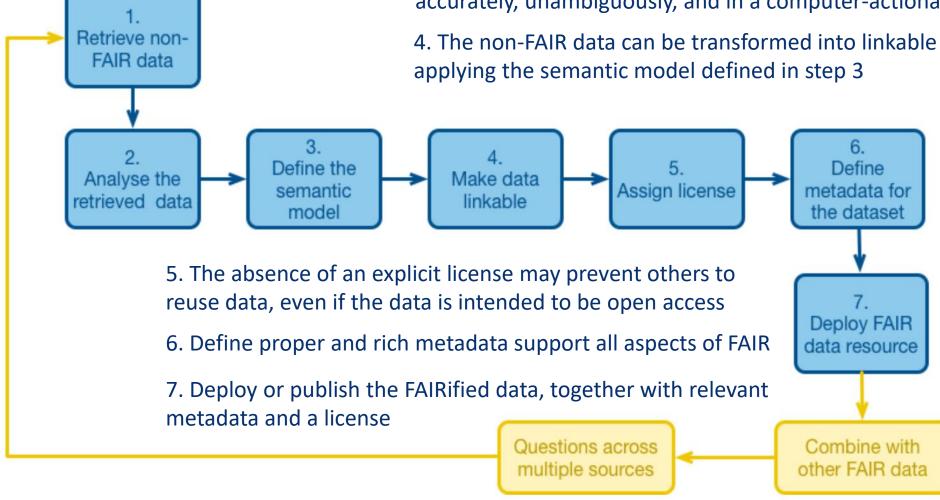


- R1. (Meta)data are richly described with a plurality of accurate and relevant attributes.
  - R1.1. (Meta)data are released with a clear and accessible data usage license.
  - R1.2. (Meta)data are associated with detailed provenance.
  - R1.3. (Meta)data meet domain-relevant community standards.
  - It is easier to reuse data sets if they are **similar**: same type of data, data organized in a standardized way, etc.
  - Many communities have minimal **information standards**. FAIR data should at least meet those standards.

#### **FAIRification Process**

1. Gain access to the data to be FAIRified

- 2. Inspect the content of the data
- 3. Describes the meaning of entities and relations in the dataset accurately, unambiguously, and in a computer-actionable way
- 4. The non-FAIR data can be transformed into linkable data by



https://www.go-fair.org/fair-principles/

# Question Two: Why FAIR Principles are Important for Data Sharing?



## Benefits for Implementation of FAIR Principles



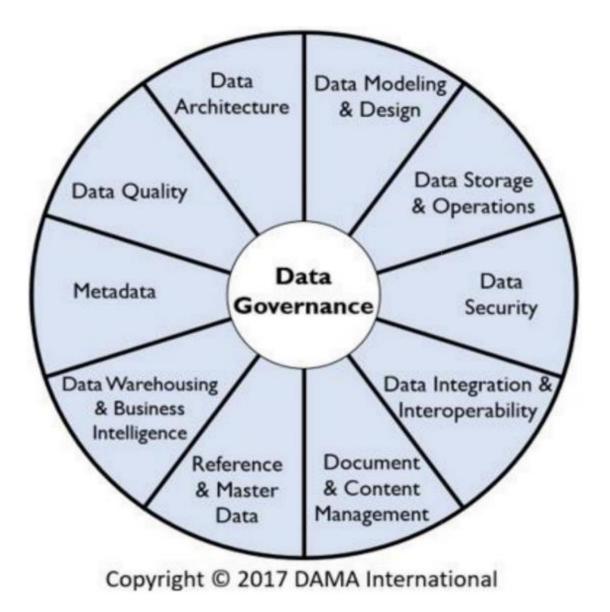
The entire process of running data through the **value chain** from acquisition, semantic alignment, integration to analytics to generate insights will become streamlined and, as a result, more effective.

- Accelerating innovation owing to availability of FAIR data for primary use and secondary reuse;
- Developing more-segmented or –personalized medicines by exploiting FAIR real-word data to match best treatment to relevant patient cohorts;
- Enabling data sharing and collaborations across institutions and companies.

#### **Data Governance**



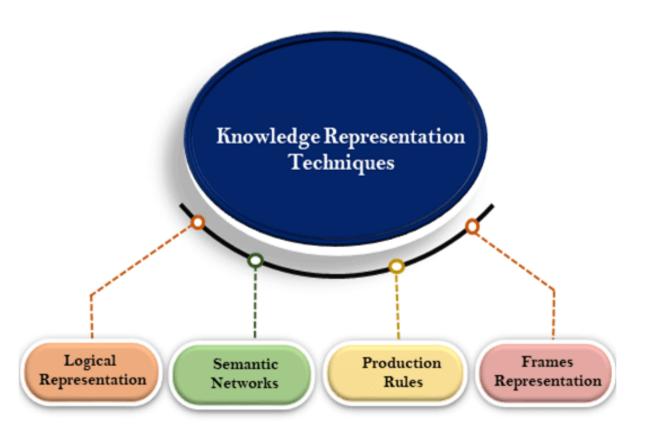
- Data governance is a collection of processes, roles, policies, standards, and metrics that ensure the effective and efficient use of information in enabling an organization to achieve its goals.
- The FAIR principles provide an important but not exhaustive foundation to define it.



## **Knowledge Representation**



- Knowledge representation is the ability for computer systems to understand information about the world with sufficient accuracy to utilize that information for an intelligent purpose.
- Successful deployment of FAIR
  will require a standardized
  information architecture, which
  helps reach a robust consensus on
  the ontologies in capturing
  specific types of data.

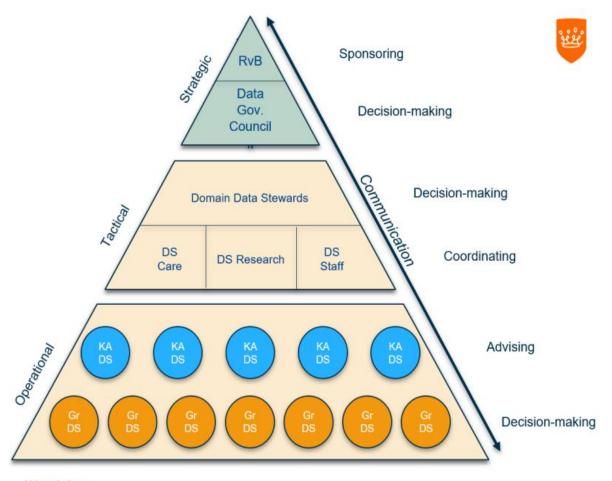


Javatpoint.com, Techniques of knowledge representation

## **Data Stewardship**



- Data Stewardship has the role of ensuring an adequate level of data and metadata quality throughout the organization utilizing data governance processes.
- Data integration challenges and quality issues usually emerge late at the time of analysis and reuse; the best point of action for quality control is at the point of data collection.



#### Abbreviations

Data Governance Council
Domain Data Stewards (for Research: Pls)

DS: Data Steward KA: Knowledge Area

Gr: Group

Model by Chrissie Taselaar, PMC

## **FAIR Data Principles in Precision Oncology**



- The goal of precision medicine is to take a detailed view of each patient and their cancer to tailor their treatment accordingly. Research has recently shown that cancer is so **heterogeneous** that single research centers cannot produce enough data to fit prognostic and predictive models of sufficient accuracy. **Data sharing** in precision oncology is therefore of utmost importance.
- However, in addition to privacy and ethical issues, various local and national health care systems and reporting traditions are often incompatible, making it complicated, expensive, and time-consuming to aggregate data from different sources due to the amount of data management involved. Then various initiatives have been launched to tackle these issues

#### FAIR Principles: the Moffitt Bioinformatics Experience

#### Findable

- Sample identifiers should be unique, machine sortable, non-PHI
- Should indicate project, PI, sample number. (not just PI\_1, PI\_2, ...)
- Organization of analysis metadata and processes (including tools and versions used) helps answer later questions
- At Moffitt, patient samples should always be linkable to a patient!
  - Can be done in a de-identified manner
  - Allows association of molecular and other data with clinical information.

#### Accessible

- Metadata should be in a standard format to allow easy searching
- Locally stored data must be organized in a way that is findable in the future!



#### FAIR Principles: the Moffitt Bioinformatics Experience

#### Interoperable

- Used standardized terms, ontology, data formats, references to enable analysis with a greater number of tools.
- Standardization important for data reuse, but also for impactful application development.
- Always include reference/acknowledgement of others' data!!!

#### Reusable

- When depositing data, include all fields needed to reproduce work.
- When submitting patient data, protect our patients by using protected data repositories. Data are still available, but protected from misuse.
- The exact analysis used is important to understand and interpret the data. Stay tune for Best Practices lecture!



## Question Three:

## How to Implement FAIR Principles for AI/ML Data Sharing in **Precision Oncology**?

- a. Initiatives to support interoperability and reusability in data collection
  - i. Clinical Data
  - ii. Genomic Data
  - iii. Imaging Data



## **Clinical Data**

Clinical data encompass the information about **patient status** and **disease phenotype**.

- The **patient status** includes demographic information, medication, comorbidities, exposures, blood test results, and treatment information.
- The disease phenotype is characterized by morphology and topography.
  - Morphology details the cellular structure of the cancer.
  - Topography defines the morphology's location.
- Usually, these data are collected by healthcare personnel and stored in electronic health records (EHRs) or in a research and clinical trials context, such as case report forms (CRFs).

## **Major Initiatives for Clinical Data**

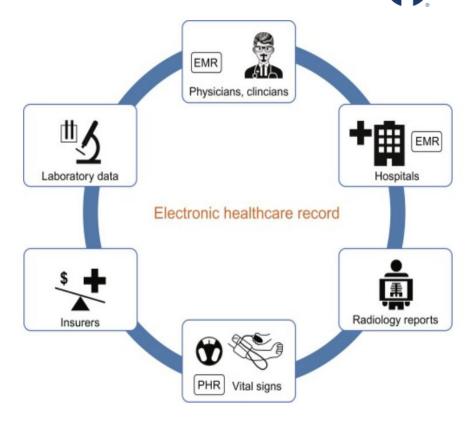
- Data Structure Models
  - > EHR, FHIR, CRF, REDCap, GDC
- Ontologies
  - > caDSR, LOINC, SNOMED CT
- Classifications
  - > ICDs, ICD-O, ATC
- Provenance Information
  - > FAIR-Health



#### **Data Structure Models - EHR**

#### **Electronic Health Record (EHR)**

 An EHR is an electronic version of a patient's medical history, that is maintained by the provider over time, and may include all of the key administrative clinical data relevant to that person's care under a particular provider, such as demographics, progress notes, problems, medications, vital signs, past medical history, immunizations, laboratory data and radiology reports.



S. Rakesh Kumar, 2019

• Several countries have largely solved the problem at a regional level but is still facing **interoperability** issues at the national level. Large efforts are needed to converge to a more broadly accepted open standard.

#### **Data Structure Models - FHIR**



#### **Fast Healthcare Interoperability Resources (FHIR)**

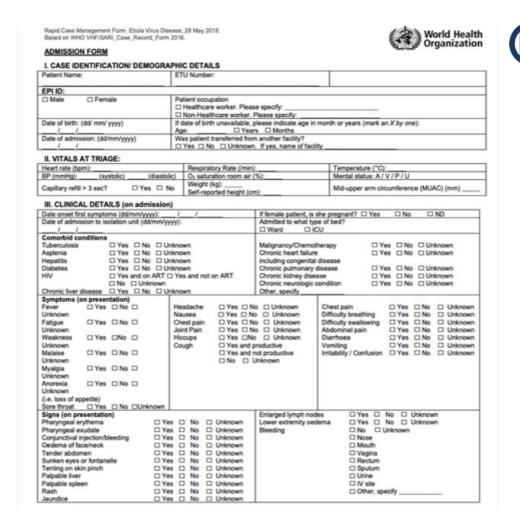
- The FHIR is a standard describing data formats and elements (known as "resources") and an application programming interface (API) for exchanging electronic health records (EHR). This standard is related to another older initiative from Health Level Seven, focusing exclusively on clinical data.
- Due to its **specificity**, this format might be cumbersome outside of an **EHR** context. Furthermore, the format supports a multitude of data types but does not provide guidance on what to share.



#### **Data Structure Models - CRF**

#### **Case Report Form (CRF)**

• A CRF is a paper or **electronic questionnaire** specifically used in clinical trial research. It is the tool used by the sponsor of the clinical trial to collect data from each participating patient. The sponsor of the clinical trial develops the CRF to collect the specific data they need in order to test their **hypotheses** or answer their **research questions**.



World Health Organization, 2018

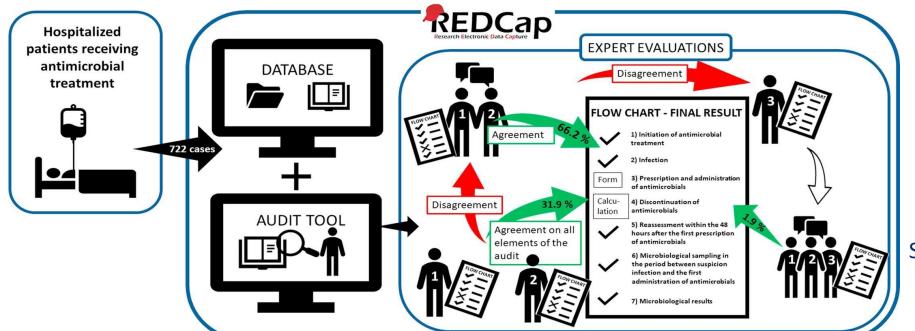
 Before being sent to the sponsor, the CRFs are usually de-identified (not traceable to the patient) by removing the patient's name, medical record number, etc., and giving the patient a unique study number.

## Data Structure Models - REDCap



#### Research Electronic Data Capture (REDCap)

- The REDCap solution allows one to design CRFs, but the emphasis is more on implementation than on good practices, as it is not designed for specifications but for actual data collection.
- The problem remains that these solutions do not provide a clear guideline on what to collect in the context of data sharing.



Signe H.Kragelund. 2018

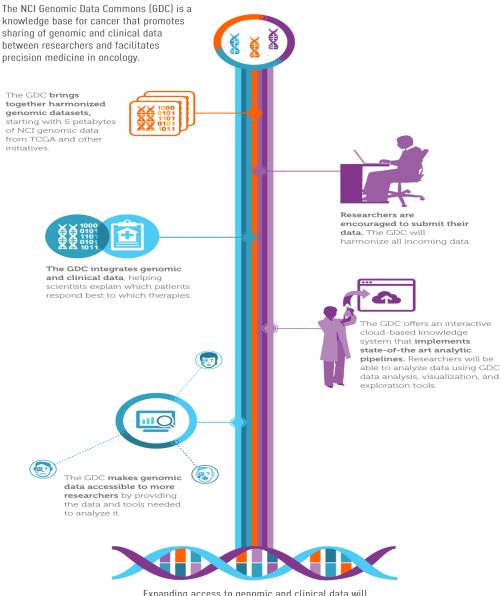
### **Data Structure Models - GDC**

#### **Genomic Data Commons (GDC)**

- The GDC's goal is to share linked clinical and genomic data from the Therapeutically Applicable Research to Generate Effective Treatments (TARGET) and The Cancer Genome Atlas (TCGA) projects. This is the largest public data repository to date linking these two types of data.
- A major accomplishment of the GDC was its ability to successfully gather and share data from disparate sources in a harmonized way as detailed harmonization requirements and procedures were designed for that purpose.

#### NATIONAL CANCER INSTITUTE

#### Genomic Data Commons



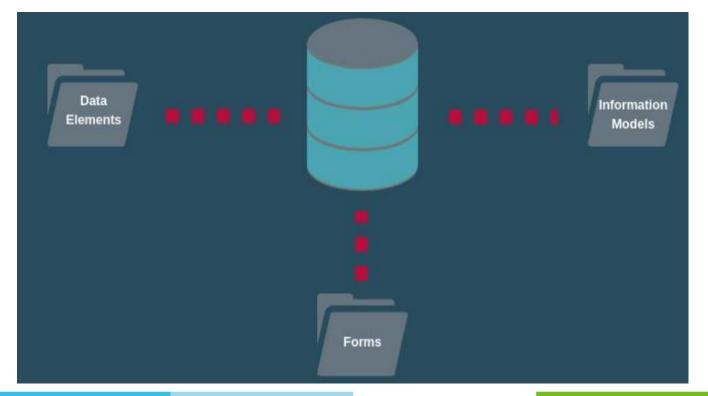
Expanding access to genomic and clinical data will accelerate cancer research and help improve the diagnosis and treatment of each cancer patient.

### Ontology - caDSR



An **ontology** is a system of carefully defined terminology, connected by logical relationships, and designed for both humans and computers to use.

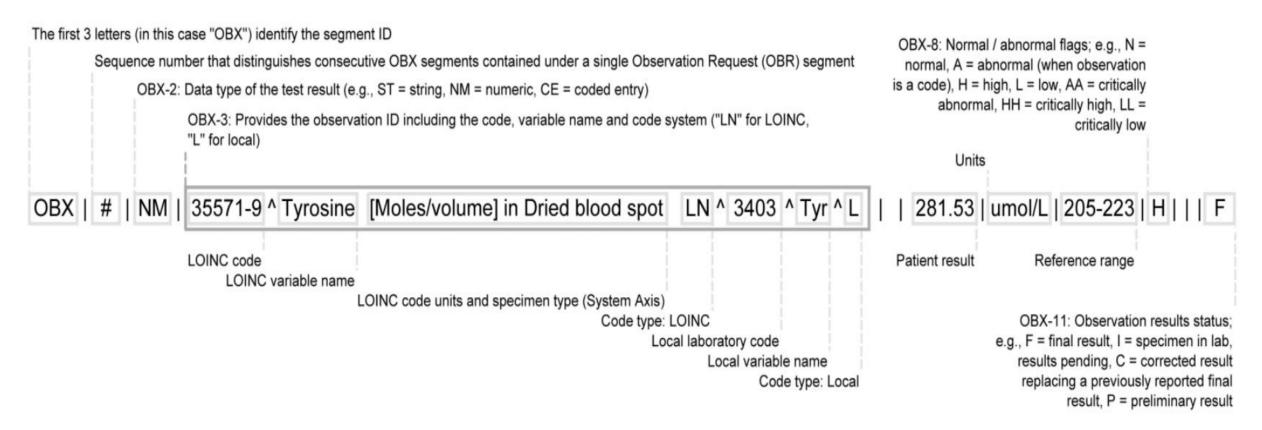
- The GDC project uses the simple ontology Cancer Data Standards Registry and Repository (caDSR) developed by the National Cancer Institute (NCI), which builds upon the common data elements (CDEs) to define data and metadata.
- Research protocols use many CRFs to collect the data researchers are studying. An information model is a software engineering representation of the concepts about cancer research and clinical care.



### **Ontology - LOINC**



• The Logical Observation Identifiers Names and Codes (LOINC) system is supposed to facilitate **interoperability**, and it is the federally required code for exchanging laboratory data.

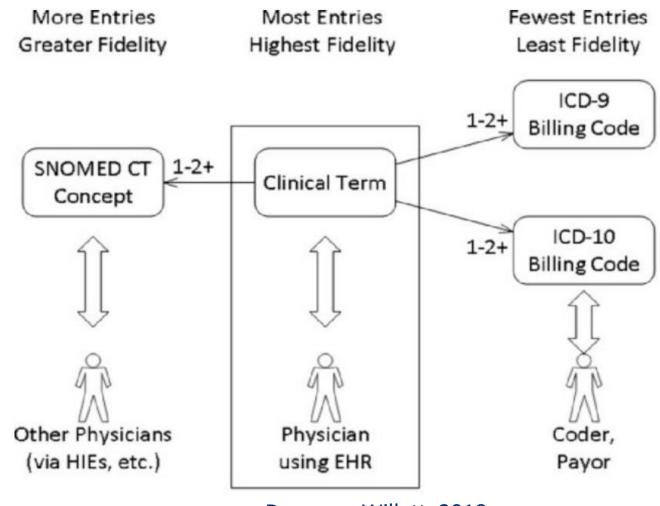


### **Ontology - SNOMED CT**



#### Systematized Nomenclature of Medicine—Clinical Terms (SNOMED CT)

- The SNOMED CT is a systematically organized computer-processable collection of medical terms used in clinical documentation and reporting.
- SNOMED CT is considered to be the most comprehensive, multilingual clinical healthcare terminology in the world. It is now a globally accepted nomenclature and is notably collaborating with LOINC.



### Classification – ICD, ICD-O



• Classifications are similar to ontologies as they define a common language, but they are much narrower in terms of scope. Moreover, there has been a much stronger movement towards convergence regarding classifications than ontologies. Most of this convergence was made possible through the World Health Organization (WHO).

- The International Classification of Diseases (ICD) is designed to promote international comparability in the collection, processing, classification, and presentation of mortality statistics.
- The ICD for Oncology (ICD-O) is a domainspecific extension of the International Statistical Classification of Diseases and Related Health Problems for tumor diseases



### **Classification - ATC**



#### **Anatomical Therapeutic Chemical (ATC)**

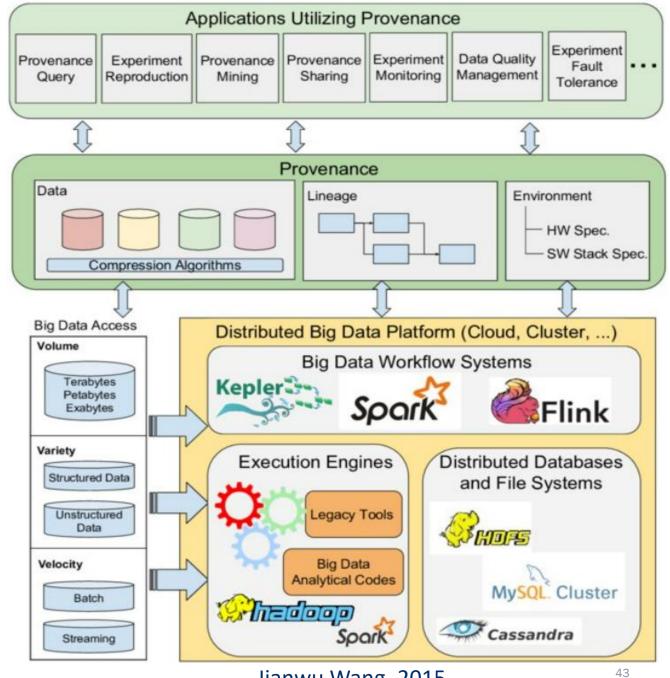
- The ATC Classification System is a drug classification system that classifies the active ingredients of drugs according to the organ or system and their therapeutic and chemical properties.
- Its purpose is an aid to monitor drug use and for research to improve quality medication use.

Symbol	Description
A	Alimentary Tract And Metabolism
В	Blood And Blood Forming Organs
C	Cardiovascular System
D	Dermatological
G	Genito Urinary System And Sex Hormones
Н	Systemic Hormonal Preparations Excl. Sex Hormones
J	General Anti-Infectives For Systemic Use
L	Anti-Neoplastic & Immunomodulating Agents
M	Musculo-Skeletal System
N	Nervous System
P	Anti-Parasitic Products
R	Respiratory System
S	Sensory Organs
V	Various
U	Other
X	Total

Lisa Nissen, 2005

### **Provenance Information**

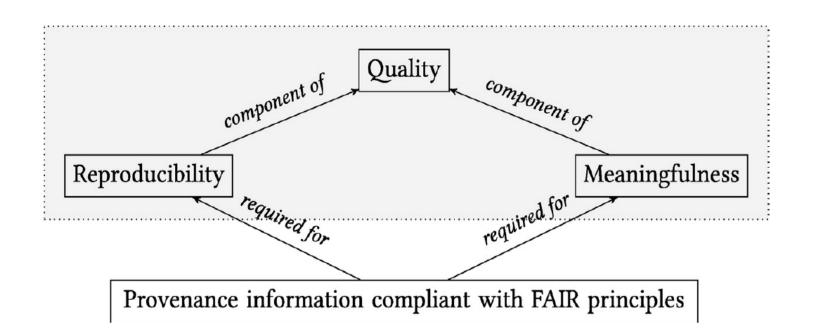
- **Provenance** is information about entities, activities, and people involved in producing a piece of data or thing. FAIR Principles are likely to address only parts of the problem.
- Complete provenance information should view data, lineage, and environment as a unified resource for reproducibility and meaningful integration of the data.



### **FAIR-Health Principles**



An extension of the FAIR Principles was proposed to FAIR-Health principles, which
include additional quality aspects related to research reproducibility and meaningful
reuse of the data.



Petr Holub, Biopreservation and Biobanking, 2018

### **Genomic Data**

- Contrary to clinical data that can be stored and shared directly upon collection, a genomic data analysis starts with biospecimens of various origins (biopsy, blood, bone marrow, etc.) from which DNA or RNA is extracted.
- The genomic data are then generated from this material and often require further bioinformatics processing before it can be interpreted. The entire work-flow needs to be standardized and documented to guarantee interoperability and reusability.
- More samples may not be available: data is all that is left!
- More details in the upcoming Best Practices lecture!



#### Standardized File Formats – Critical for FAIR Analyses

- FASTQ file: text file of raw sequence data and base quality scores: <u>https://en.wikipedia.org/wiki/FASTQ\_format</u>
- SAM/BAM file: text and binary representation of sequence alignment to a reference genome. Indexed for fast access anywhere in a large file. Well documented: <a href="https://samtools.github.io/hts-specs/SAMv1.pdf">https://samtools.github.io/hts-specs/SAMv1.pdf</a>
- VCF file: Variant Call Format. Text file containing genetic variants in a nested data structure. Many tools exist to handle analysis of this file type. <a href="https://samtools.github.io/hts-specs/VCFv4.2.pdf">https://samtools.github.io/hts-specs/VCFv4.2.pdf</a>
- MAF file: Mutation Annotation Format. Another format containing genetic variants that is easier for humans to read as it is a tabular file format. Tools exist to analyze/summarize these files as well. Used by TCGA.
- Text files usually compressed with gzip, and tools available to index text file for rapid access: <a href="http://www.htslib.org/doc/tabix.html">http://www.htslib.org/doc/tabix.html</a>

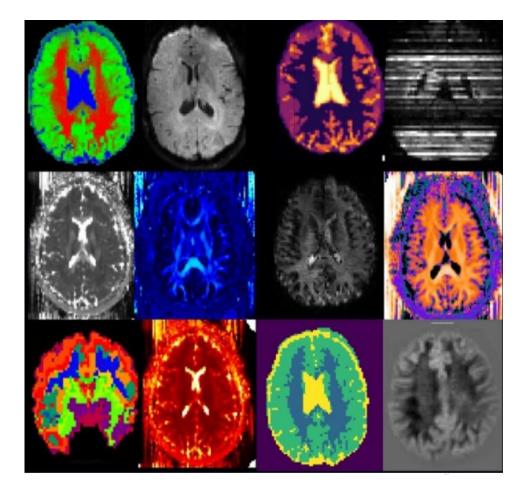
### Considerations for Sharing Molecular Data

- Many computational analyses use molecular data, including DNA and RNA sequencing, proteomics, metabolomics, expression microarrays, etc. These data are now required to be shared by NIH when included in a grant!
- Genome Data Sharing Required by NIH: Genomic Data Sharing (GDS) Policy NOT-OD-14-124
  - https://osp.od.nih.gov/scientific-sharing/genomic-data-sharing/
- For human subjects research, patients must be consented for sharing, and Human Subjects regulations must be followed (ie, protection of PHI)
  - The Total Cancer Care consent protocol is compatible.
- Moffitt is revising the data sharing process: contact BBSR or Susan Sharpe if you plan to share molecular data.
- PLAN AHEAD!! The process can take a long time!



# **Imaging Data**

- Imaging biomarkers hold tremendous promise for precision medicine clinical applications. Development of such biomarkers relies heavily on image post-processing tools for automated image quantitation.
- Their deployment in the context of clinical research requires interoperability with the clinical systems.

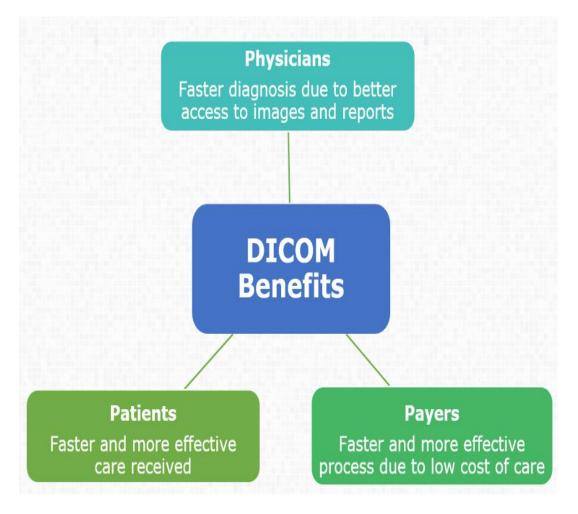


Jana Hutter, 2020

### Digital Imaging and Communications in Medicine (DICOM)



- DICOM is the standard for the communication and management of medical imaging information and related data. It is most commonly used for storing and transmitting medical images enabling the integration of medical imaging devices.
- DICOM is primarily used to support interoperability between clinical systems for image interchange. Consumption of the DICOM images is widely supported in research tools.



National Resource Centre for EHR Standards, 2022

## Question Three:

# How to Implement FAIR Principles for AI/ML Data Sharing in **Precision Oncology**?

- a. Initiatives to support interoperability and reusability in *data collection* 
  - i. Clinical Data
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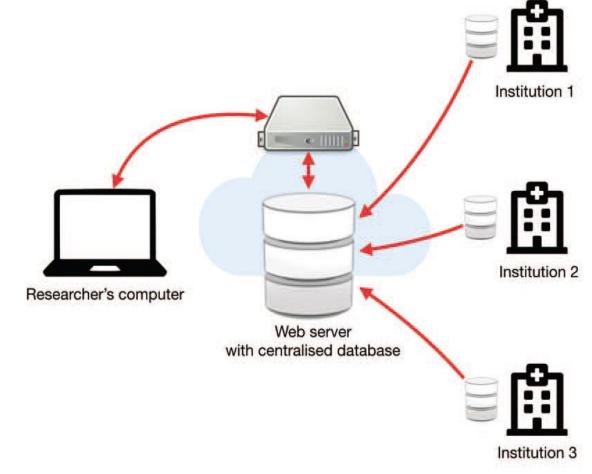
- b. Initiatives to support findability and accessibility in data sharing
  - i. Network Architecture
  - ii. Access Control



### **Network Architecture - Centralized Architectures**



- In the centralized network architecture, each institution must upload their data to a centralized web server, where everything is gathered in one place.
- This architecture guarantees a better harmonization of the data, but it also faces major challenges.
  - The drawbacks of the strict harmonization
  - Very large size of the project
  - The massive quantity of data generated by high-throughput sequencing.



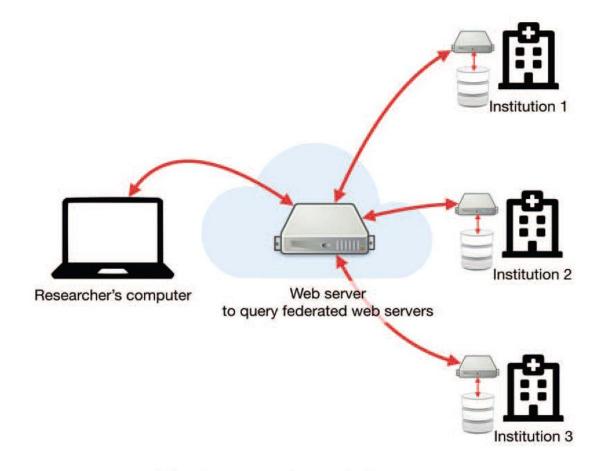
#### **Centralized Architectures**

Charles Vesteghem, 2020

### **Network Architecture - Federated Architectures**



- In the federated network architecture, data stay at their respective institutions, but each institution must implement an interface to make the data findable but not necessarily accessible.
- Here, the goal is to make data easily searchable by defining an interface rather than a structure. However, the trade-off is that the stored data are not likely to be as interoperable as in the centralized scenario.



#### **Federated Architectures**

Charles Vesteghem, 2020



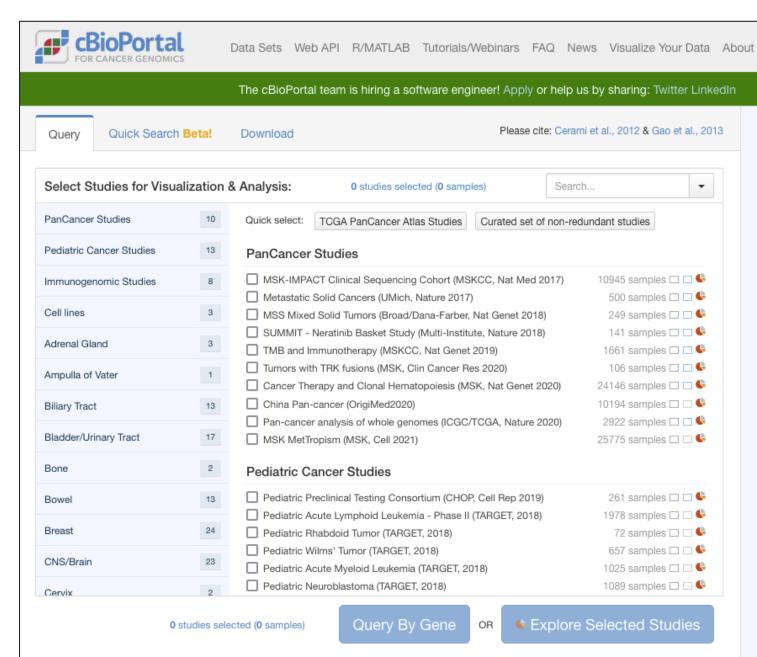


There are mainly two approaches for accessibility:

- In the **gate-keeper** approach, data are not directly accessible and a request to access data is required. This approach usually guarantees data of better quality and improves the FAIRness of the stored data, notably **reusability**.
- In contrast, the **open access** approach implies that data are available without restriction and its goal is to build common genetic resources to foster research. The main aim is accessibility, potentially to the detriment of other FAIR aspects.
- There is a trend to have a mix of the two approaches. The mixed approach followed by the GDC and ICGC seems to be the most pragmatic one, which allows one to keep more sensitive data under control while making less sensitive data easily accessible.

#### Findable and Accessible Sequence Data - cBioPortal

- Many sequencing projects share high-level results via a web-based tool.
- Includes The Cancer Genome Atlas (TCGA) and others.
- Can easily find different diseases.
- Can access clinical and molecular results.

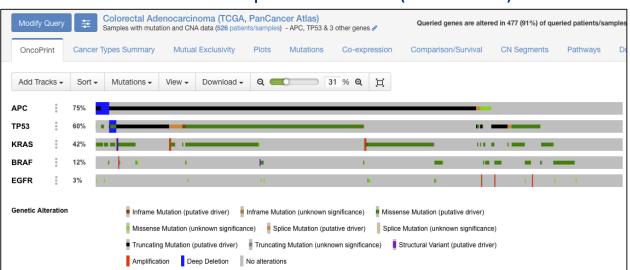


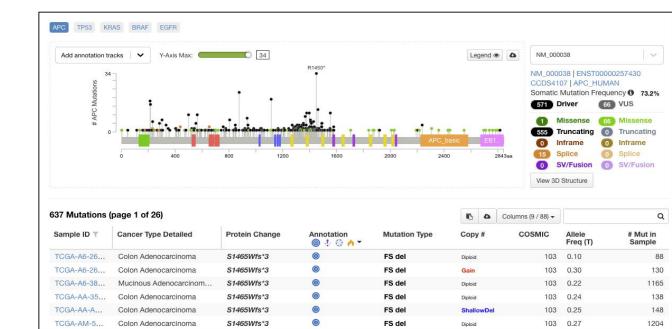
#### cBioPortal, Colorectal Cancer Example

#### **Clinical Data**

#### Colorectal Adenocarcinoma (TCGA, PanCancer Atlas) Click gene symbols below or e Colorectal Adenocarcinoma TCGA PanCancer data, The original data is here, The publications are here, PubMed CN Segments Selected: 594 patients | 594 samples Clinical Data Category Cancer Type Detailed Freq Molecular Profile Freq 100% Colon Adenocarcinoma 378 63.6% 594 100.0% Putative arm-level copy-number fr. Rectal Adenocarcinoma 155 26.1% mRNA expression z-scores relativ. 592 99.7% Mucinous Adenocarcinoma of the.. 61 Log2 copy-number values 592 99.7% mRNA Expression, RSEM (Batch) 592 99.7% 50 100 150 200 Search. Select all mRNA expression z-scores relativ. 592 99.7% KM Plot: Disease Free Survival (m... KM Plot: Disease-specific Survival... mRNA expression z-scores relativ. 592 99.7% KM Plot: Progression Free Survival. 592 99.7% Putative copy-number alterations . 100% 100% 100% Methylation (HM27 and HM450 m. □ 591 99.5% 583 Microbiome Signatures (log RNA 98.1% 50% 50% 50% 534 Fusions 89.9% Mutations 534 89.9% 50 100 150 200 0 50 100 150 200 50 100 150 200 Search. Mutation Count vs Fraction Genome Altered **Overall Survival Status** Mutated Genes (534 profiled samples) Freq Gene 12k-# samples □ 387 639 72.5% 332 314 TP53 58.8% TTN 864 257 48.1% KRAS 223 218 40.8% PIK3CA 178 147 27.5% -0.3310 MUC16 382 □ 146 27.3% SYNE1 337 144 27.0% -0.3420 124 FAT4 231 23.2% p=0.00 201 RYR2 105 19.7% 173 OBSCN 99 18.5% 0 0.1 0.2 0.3 0.4 0.5 0.6 0.7 0.8 0.9 1 ZFHX4 147 95 17.8% Fraction Genome Altered Search. Structural Variant Genes (534 profiled samples) CNA Genes (592 profiled samples) **Ethnicity Category** Gene Freq Gene Freq Cytoband TCF25 7 129 1.3% RBFOX1 16p13.3 21.8% DPEP1 7 1.3% WWOX 16q23.1-q. 11.0% 4 TCF7L2 0.7% CCSER1 4g22.1 4 ITCH 0.7% MACROD2 20p12.1 ETV6 3 0.6% POFUT1 20q11.21 55 9.3%

#### Sequence Results (Mutations)





### **Take Home Messages**



- The FAIR Data Principles must be taken into account in the **conception phase** of the project.
- For the actual data collection, **REDCap** is an excellent resource due to its flexibility and open API, allowing it to be easily integrated with existing solutions.
- The **GDC** data structure can be considered the de facto standard and therefore a logical choice for structuring data collection.
- The implementation of existing standards, notably from **WHO**, is mandatory for **interoperability**.
- Findability can be achieved through a federated infrastructure.

# **Questions?**

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