

FGAI4H-O-046

Berlin, 31 May – 2 June 2022

Source: H3ABioNet (South Africa)

Title: Workshop: WG-CO - Development of genomics data infrastructure and standards within Africa

Purpose: Discussion

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Abstract: This PPT contains a presentation from the WG-CO workshop on “Equitable data infrastructures to support equitable and effective pandemic intelligence”.



DEVELOPMENT OF GENOMICS DATA INFRASTRUCTURE AND STANDARDS WITHIN AFRICA

Sumir Panji

ITU-WHO Focus Group on Artificial Intelligence for Health
30th May 2022

www.h3abionet.org #h3abionet

Presentation Overview

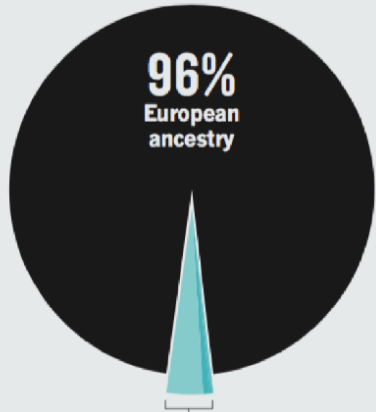
- Introduction
- H3Africa and H3ABioNet
- Overview of H3ABioNet
- Development of data standards
- Development of an H3Africa data archive and data catalogue
- Computational tool development
- Findable, Accessible, Interoperable and Resusable (FAIR) and H3Africa Data and meta-data
- Examples of SARS works and early release of data
- Data Science Initiative Africa (DS-A Africa)
- eLwazi Open Data Science Platform
- Conclusions



The Genomics Revolution and Africa



2009
373 studies
1.7 million samples



2016
2,511 studies
35 million samples

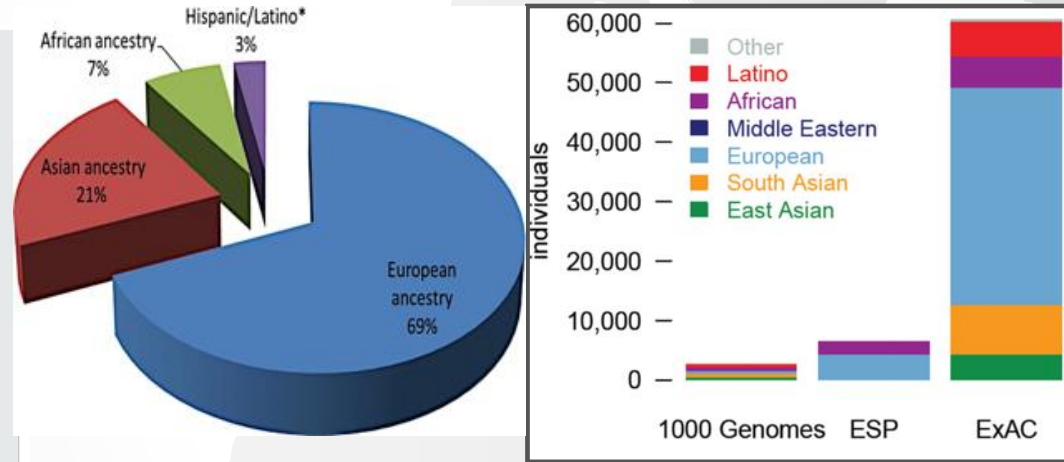


Asian
Other non-European

0.57%

African Ancestry
(Majority African-American)

3%



Broad Ancestry of Evaluation Sample sets used for the PGS Catalogue

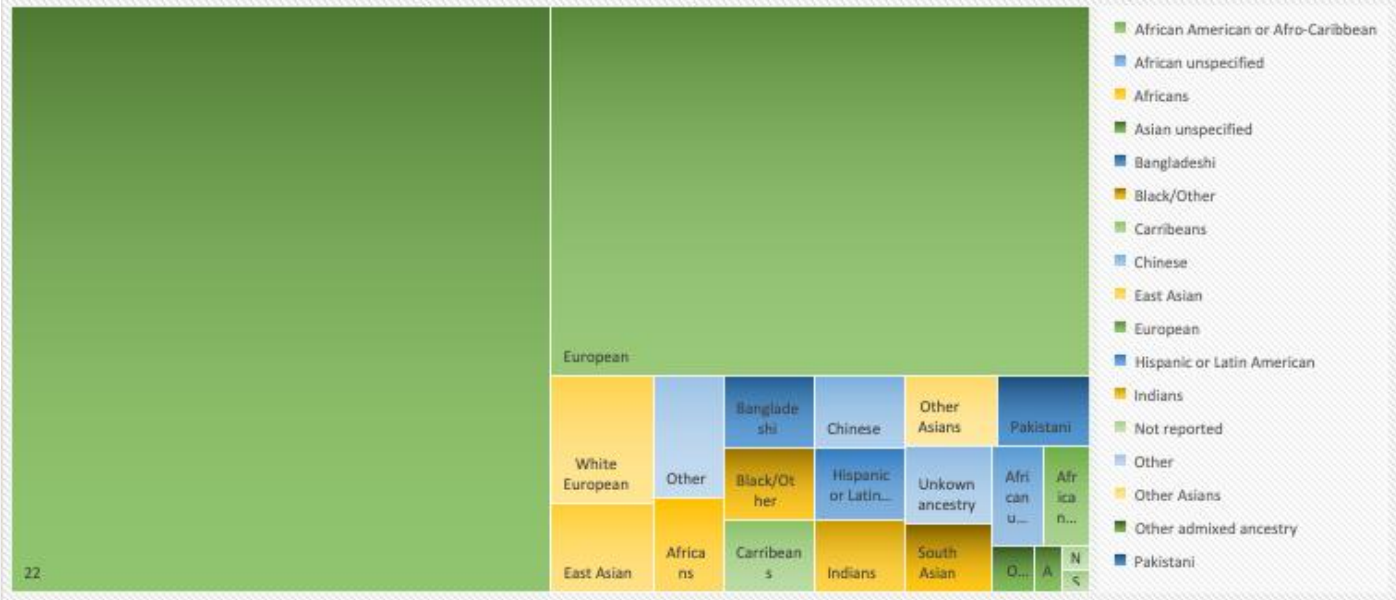


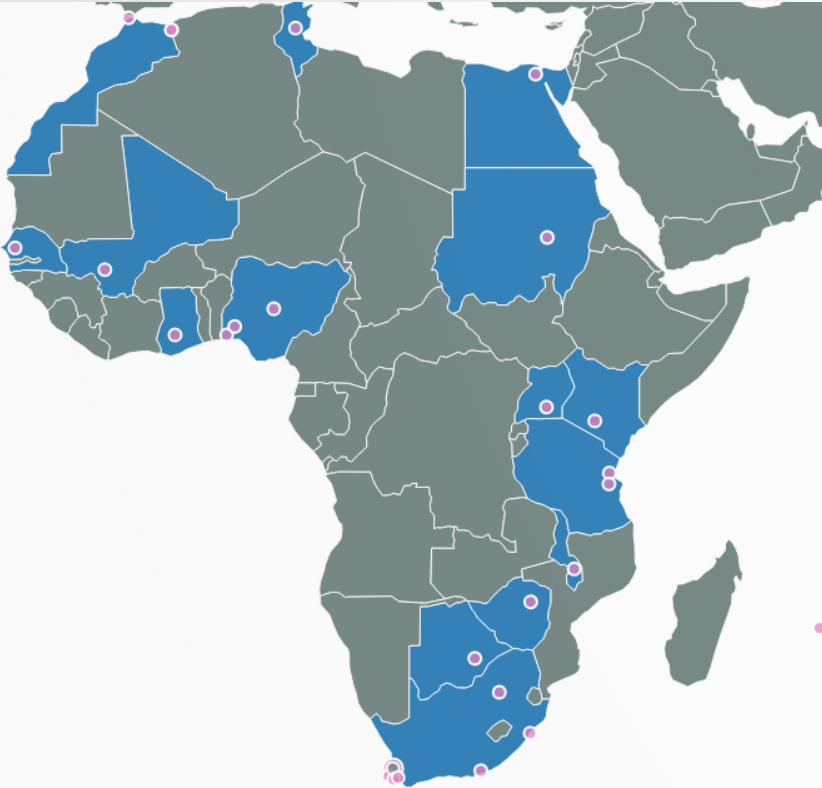
Image credits and reference: Popejoy AB, Fullerton SM. Nature. 2016 Oct 13;538(7624):161-164 <https://doi.org/10.1038/538161a>

Slide credits - Prof. Michèle Ramsay, SBIMB, Wits and Prof. Nicky Mulder UCT C BIO

Data summary from the PGS Catalogue accessed in August 2020: <https://www.pgscatalog.org/>

Pan African Bioinformatics Network for H3Africa (H3ABioNet)

- Pan African Bioinformatics Network to develop bioinformatics capacity in Africa and support the H3Africa research projects



- 28 Institutions
- 16 African Countries

Overview of H3Africa Data



Phenotype data (associated with genotype data)

- Demographic information
- Anthropometric data
- Disease and health related phenotype data



Genetic Variation data human and pathogen

- Genomic Sequence data
- Whole genome exome and targeted sequencing



Genotyping chip array data

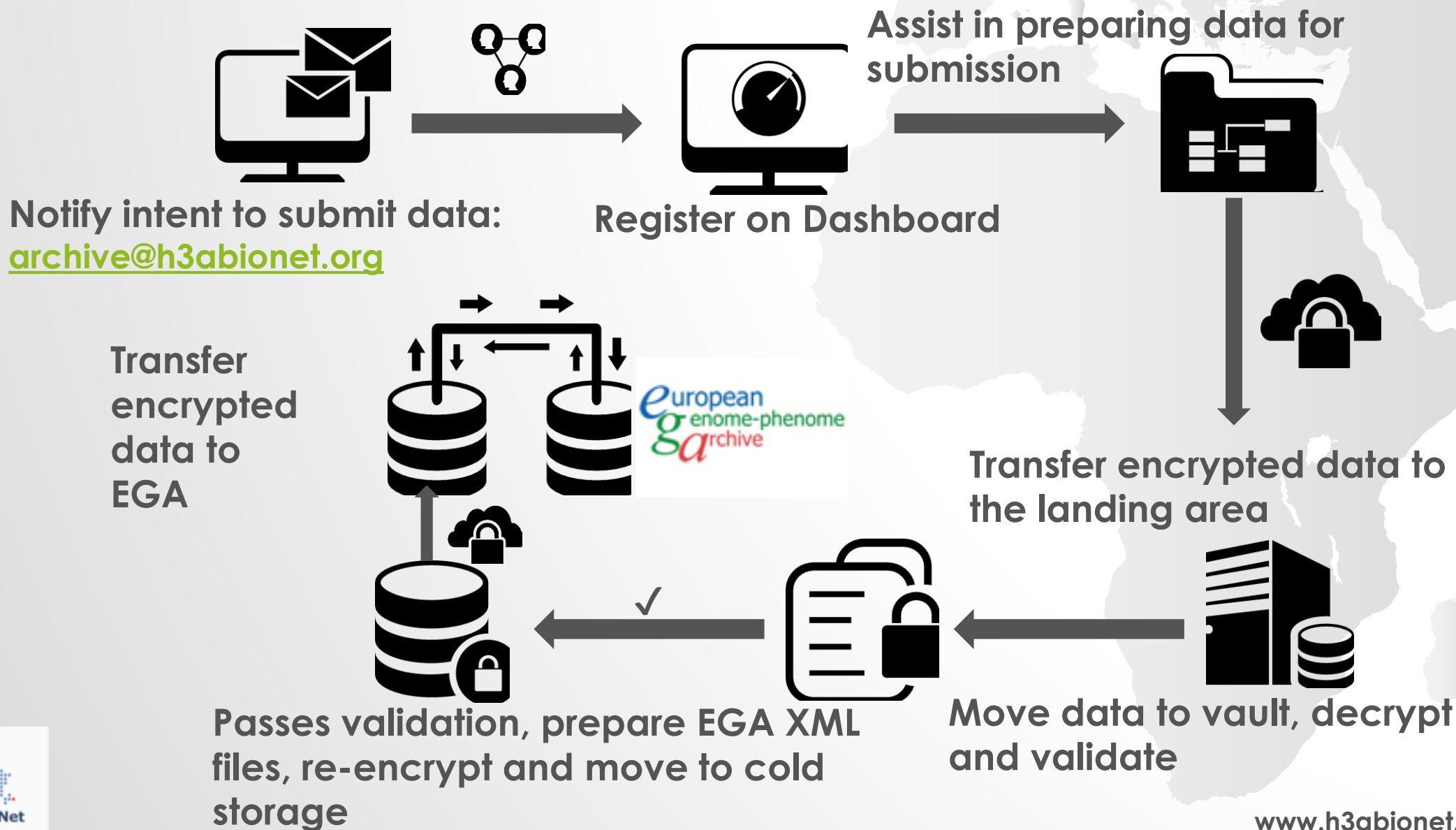
- ~55,000 samples to be run on an H3Africa African custom chip
- Raw intensity files linked



Microbiome sequence data

- Patient/sample phenotypes
- Non-human 16S rRNA sequence data for microbiome
- Non-human full genome sequence data for microbiome
- Possible human sequence contamination

H3Africa Data Archive overview



- 14 African Genomic Datasets in the archive
- 134.9 Terabytes of African Genomics data
- 8 Datasets QC-ed, validated and submitted to EGA
- 77.3 Terabytes transferred for storage at the EGA

Specific Data submission files and checks



- ✓ FastQ files linked with de-identified participant ID (minus adapters, linkers, barcodes)
- ✓ Binary Alignment files (BAMs, de-multiplexed) – linked with participant de-identified ID
- ✓ Variant calling files (VCFs)
- ✓ Phenotypic data
- ✓ Study type e.g. case / control, trio, longitudinal
- ✓ Sequencing platform / technology
- ✓ Mapping file indicating relationship between files



- ✓ Chip platform and model / name
- ✓ Raw intensity files linked with de-identified participant IDs (IDATs, CELs)
- ✓ Manifest file describing SNP or probe content on chip
- ✓ Software used for calling the genotypes
- ✓ Final reports and analyses files generated
- ✓ Phenotypic data
- ✓ Study type e.g. case / control, trio, longitudinal
- ✓ Mapping file indicating relationship between files



- ✓ FastQ files linked with de-identified participant ID (minus adapters, linkers, barcodes)
- ✓ Sequencing platform
- ✓ Final analyses BIOM files (at minimum contain OTUs)
- ✓ Phenotypic data
- ✓ Study type e.g. case / control, longitudinal
- ✓ Mapping file indicating relationship between files



Move data to the vault, decrypt and validate:

- ? Do the checksums match
- ? Are there mapping files present
- ? Do the number of samples match what is expected
- ? Are all the files present for each de-identified participant ID
- ? Is there a mismatch between participant IDs and files
- ? Do all the files have phenotypic data present
- ? Is there a dataset summary description and study abstract present



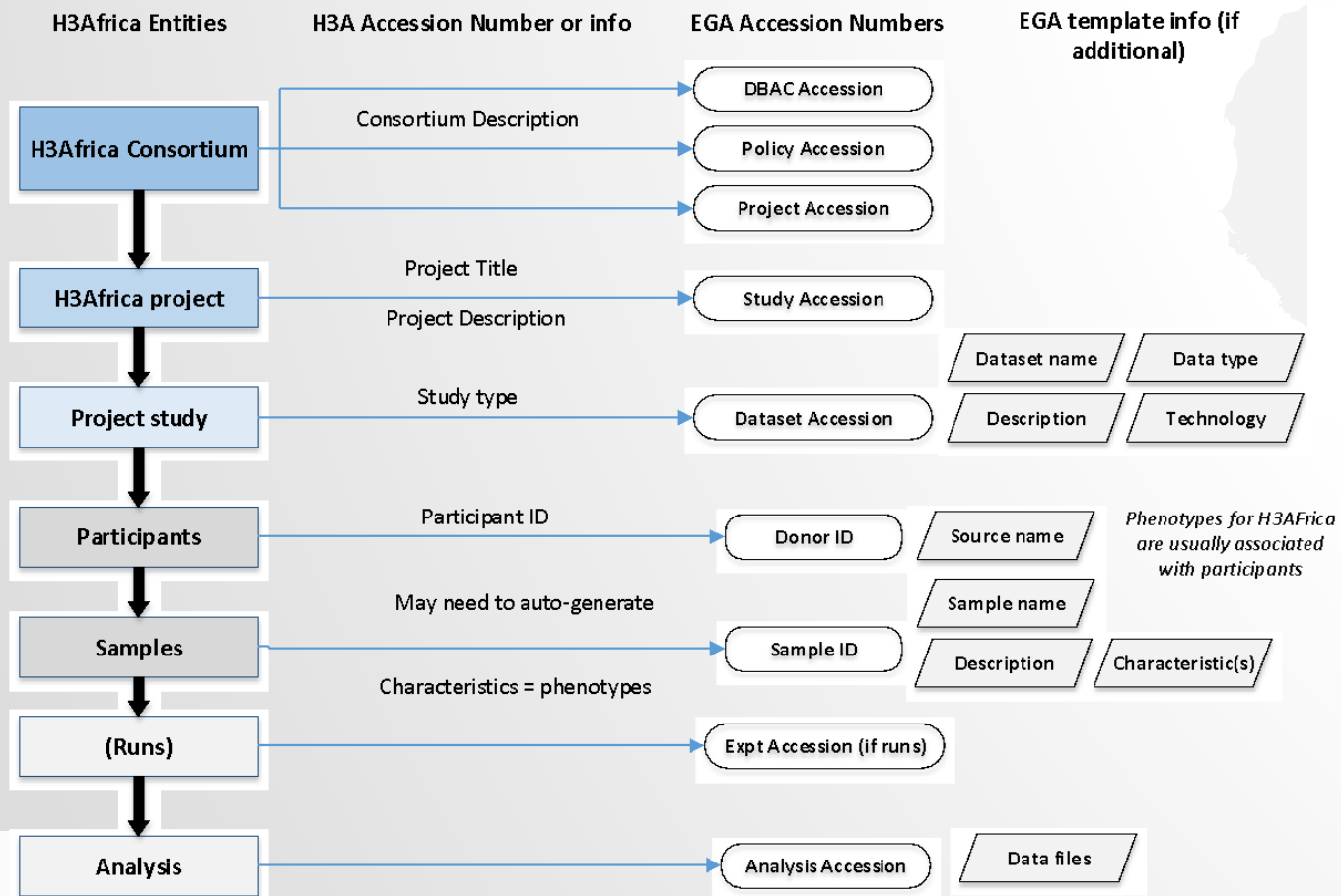
Passes validation:

- Map data to EGA XML schemas
- Re-encrypt the data and move to cold storage
- Submit data to EGA at pre-determined date



Passes validation:

- Map data to EGA XML schemas
- Re-encrypt the data and move to cold storage
- Submit data to EGA at pre-determined date



Search...

Tips on how to search

ABOUT SUBMISSION BROWSE ACCESS DOWNLOAD METADATA Helpdesk Log in

19 search results for H3Africa

All (19) Studies (6) Datasets (13) DAC (0) Documentation (0)

Previous 1 2 Next

H3AFRICA ACCME

...H3AFRICA ACCME...

H3AFRICA MALSIC

...H3AFRICA MALSIC...

Study ID ^	Study Title v	Study Type v
EGAS00001002976	H3Africa Chip Design - Aim of designing a cost-effective GWAS chip with content appropriate for use in genomics studies of individuals from the African continent.	Other

Browse Dataset Files

Back to dataset

H3AFRICA ACCME

Dataset ID	Technology	Samples
EGAD00001004505	Illumina HiSeq 2500	49

Dataset Description

Samples from Nigeria generated for the H3Africa Chip Design Study

1323 Files

12.05 TB

Download as CSV

<https://ega-archive.org/search-results.php?query=H3Africa>

Phenotype harmonization

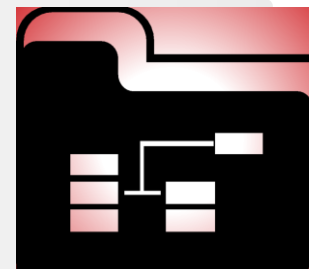
Meta-data (phenotype) data is collected via case report forms (CRFs)

- (1) Age & (2) Sex
- (3) Country of birth
- (4) Current residence
- (5) Native language
- (6) Ethno-linguistic/tribal affiliation
- (7) Country of birth of father and mother
- (8) Native language of father and mother
- (9) Ethno-linguistic/tribal affiliation of mother and father
- (10) Height
- (11) Weight
- (12) Current medications
- (13) Smoking history
- (14) Alcohol history

Same question – data coded in different ways

Same variable – collected in different ways

Project 1 CRF



Female

Daily units

Project 2 CRF



Woman

Weekly units

Project 3 CRF



1

User defined time period

Phenotype harmonization

Table 2. Key variables being collected across different projects in CHAIR

	ACCME		AWI-Gen		DM Group		Kidney group		RHDGen		SIREN	
	n = 11 700	% Complete	n = 11 057	% Complete	n = 3657	% Complete	n = 7964	% Complete	n = 3555	% Complete	n = 3811	% Complete
Age distribution	>18 years	100.0	40-60 years	100.0	≥25/ ≥18 years	100.0	0-74 years	100.0	Paediatrics and adult	100.0	>18-100 years	100.0
Sex	Female only	100.0	M/F	100.0	M/F	100.0	M/F	100.0	M/F	100.0	M/F	100.0
Anthropometrics:												
Weight (kg)	X	99.9	X	100.0	X	100.0	X	100.0	X	98.5	X	87.9
Height (m)	X	99.8	X	100.0	X	100.0	X	99.9	X	95.4	X	91.0
Waist circumference (cm)	X	99.8	X	100.0	X	100.0	X	99.7			X	94.6
Hip circumference (cm)	X	99.9	X	100.0	X	100.0					X	95.5
General health:												
Smoking/tobacco	X	99.8	X	100.0	X	100.0	X	100.0			X	98.3
Alcohol	X	99.8	X	99.9	X	100.0	X	99.9			X	98.9
Cancer history	X	99.8	X	99.9	X	100.0	X	99.8			X	98.5
Diet	X	99.7	X	99.8	X	100.0					X	95.4
Exercise	X	99.8	X	100.0	X	100.0					X	98.3
Cardiovascular health:												
Blood pressure	X	99.9	X	100.0	X	100.0	X	100.0	X	86.8	X	96.6
Atrial fibrillation (ECG)							X	100.0	X	12.1	X	98.3
Stroke and stroke-free status	X	99.8	X	99.9	X	100.0	X	100.0			X	100.0
Myocardial infarction	X	99.8		77.0	X	100.0					X	98.6
Blood collection for biomarkers												
Lipid profile			X	100.0	X	100.0					X	99.0
Fasting plasma Glucose			X	100.0	X	100.0					X	60.6
HbA1c			X		X	100.0					X	41.7
Insulin			X	100.0	X	100.0						
Infection history												
TB infection			X	99.9	X	100.0	X	99.9			X	98.6
HIV status	X	100.0	X	99.9	X	100.0	X	99.9			X	98.6
Malaria			X	99.9	X	100.0	X	100.0			X	98.4
Urine collection												
Albumin (microalbumin)			X	99.8	X	100.0	X	100.0				
Total protein			X	99.9			X	99.9				
Creatinine			X	99.9	X	100.0	X	100.0			X	80.0
Samples to be stored												
DNA	X	99.9	X	99.9	X	100.0			X	99.0	X	99.9
Buffy coat	X	99.9	X	99.9					X	70.0	X	99.9
Plasma	X	99.9	X	99.9	X	100.0			X	82.0	X	99.9
Serum	X	99.9	X	99.9	X	100.0			X	79.9	X	99.9
Urine	X	99.7	X	99.8	X	100.0					X	

M/F, male/female.

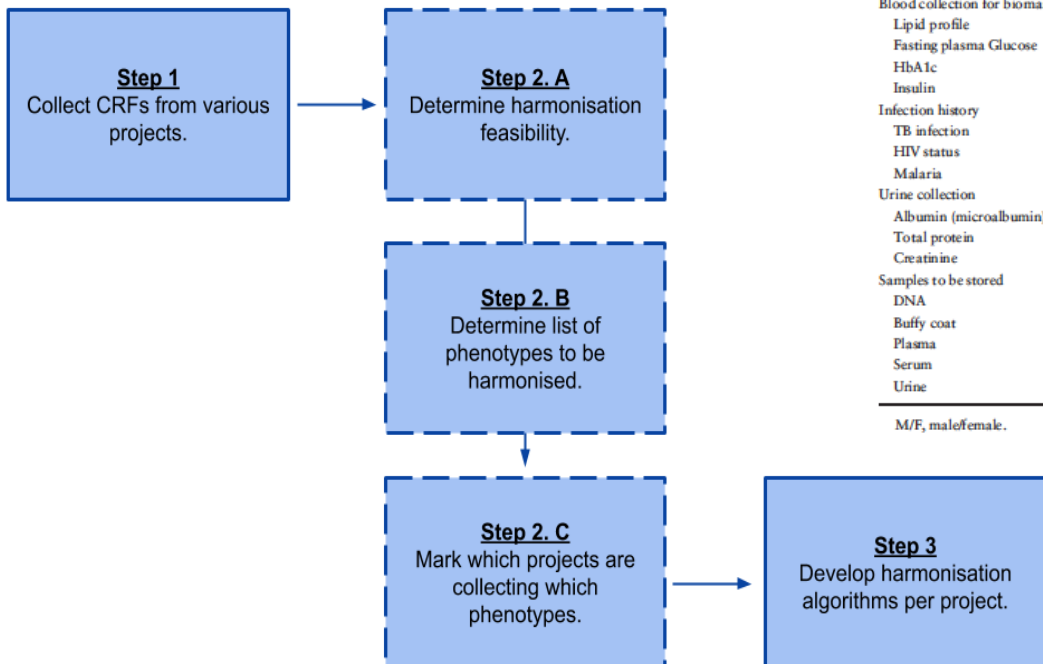
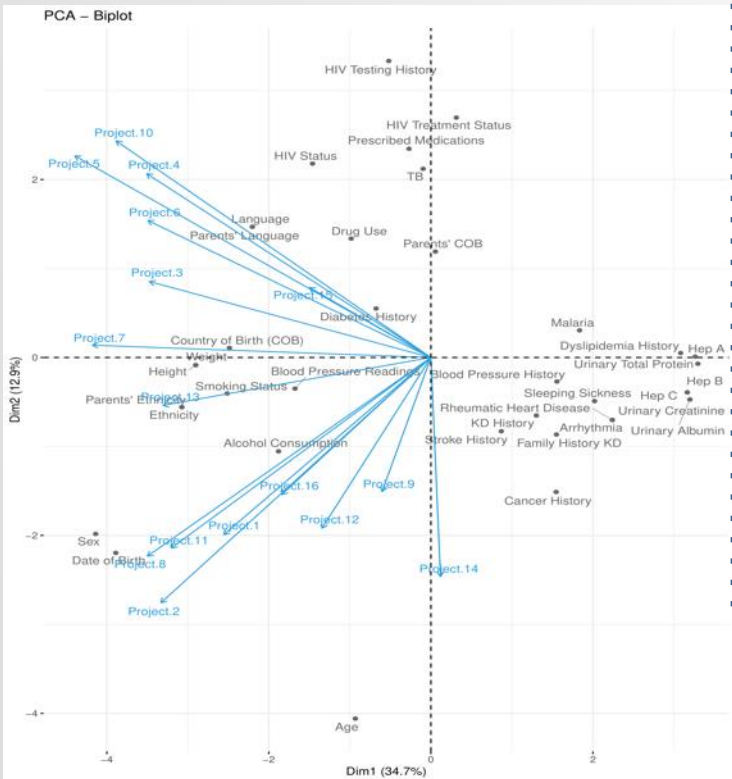


Table reference: Owolabi MO, Akpa OM, Made F, Adebamowo SN, Ojo A, Adu D, Motala AA, Mayosi BM, Ovbiagele B, Adebamowo C, Tayo B, Rotimi C, Akinyemi R, Gebregziabher M, Sarfo F, Wahab KW, Parekh RS, Engel ME, Chisala C, Peprah E, Mensah G, Wiley K, Troyer J, Ramsay M; as members of the CVD Working Group of the H3Africa Consortium. Data Resource Profile: Cardiovascular H3Africa Innovation Resource (CHAIR). Int J Epidemiol. 2019 Apr 1;48(2):366-367g. doi: 10.1093/ije/dyy261. PMID: 30535409; PMCID: PMC6469307.

Phenotype harmonization



Data sources

EGA



Biobanks



CHILD



H3ABioNet



CINECA



Experimental Factor Ontology

Cohort/Resource name	Number of participants	Location	Longitudinal	Diseases	Gender	WGS	WES	RNA-Seq	Epigenetics	Genotyping
CHILD	3.5K	CA	X	Population based developmental health and disease	M & F	X		X	X	X
CARTaGENE	43K	CA	X	Population based cohort	M & F	X		X		X
PROFYLE	450	CA		Pediatric Oncology	M & F	X		X		
CLSA	50K	CA	X	Population based cohort	M & F					X
H3Africa	75K	SA		Multiple communicable and non-communicable diseases in multiple African countries	M & F	X	X			X
BIOS	4K	NL		Population based cohort	M & F	X		X	X	X
Estonian Biobank	51K	EE	X	Population based cohort	M & F	X	X	X	X	X
CoLaus	6.1K	CH	X	Cardiovascular diseases	M & F			X		X
PsyCoLaus	3.6K	CH	X	Mental disorders	M & F			X		X
EGA	700K	UK+ES		Multiple disease and healthy cohorts	M & F	X	X	X	X	X
UK Biobank	500K	UK	X	Population cohort and disease: cancer, heart diseases, stroke, diabetes, arthritis, osteoporosis, eye disorders, depression and forms of dementia	M & F	X	X			X

Human Genomic Data Privacy

H3Africa rich source of meta-data (phenotypes)

- (1) Age & (2) Sex
- (3) Country of birth
- (4) Current residence
- (5) Native language
- (6) Ethno-linguistic/tribal affiliation
- (7) Country of birth of father and mother
- (8) Native language of father and mother
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- (10) Height
- (11) Weight
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- Combination of phenotype and genetic data makes it possible to identify different populations and individuals – controlled access

H3Africa data sharing and access policy

Balance between ensuring adequate safeguards to protect participants, while not being a barrier for scientists to advance research:

- Maximizing the availability of research data, in a timely and responsible manner.
- Protecting the rights and privacy of human subjects who participated in research studies.
- Recognizing the scientific contribution of researchers who generated the data.
- Considering the nature and ethics of the research proposed in establishing the timely release of data, and mechanisms of data sharing.
- Promoting deposition of genomic data in existing community data repositories whenever possible

http://h3africa.org/images/DataSARWG_folders/FinalDocsDSAR/H3Africa%20Consortium%20Data%20Access%20%20Release%20Policy%20Aug%202014.pdf



H3Africa Biospecimen and Data Catalogue



<https://catalog.h3africa.org/>

www.h3abionet.org #h3abionet



<https://catalog.h3africa.org/>

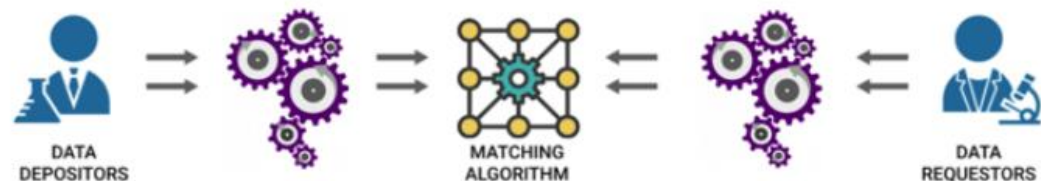
Data Use Conditions

RUO **PS** **MOR** **PUB**

See further information on [Data Use Conditions](#)

Label ▾	Code ▾	Version ▾	Modifier ▾
research use only	DUO-000014	2019-01-07	
general research use	DUO-000042	2019-01-07	
project specific restriction	DUO-000027	2019-01-07	
publication moratorium	DUO-000024	2019-01-07	
publication required	DUO-000019	2019-01-07	

Data Use Ontology



Data submitted by Data Depositor tagged with Data Use Restrictions & Requirements

Data Access Request Application, is represented by DUO data use categories

Data Use Restrictions & Requirements

Data Use Codes

- No Restrictions
- Primary consent code
 - General Research Use (GRU)
 - Health/Medical/Biomedical (HMB)
 - Disease-specific Research (DS-X)
- Secondary code
 - No General Methods (NMDS)
 - Genetic Studies Only (GSO)

Data Use Requirements

- Not-for-profit use only (NPU)
- Publication required (PUB)
- Collaboration required (COL-[XX])
- Publication moratorium/embargo
- Time limit on use
- User-specific
- Project-specific restrictions
- Institution specific restrictions
- Security Terms
- Expunging Terms
- Linking Terms
- Allowed Recontact Terms
- Compulsory Recontact Terms
- IP Claim Terms
- Reporting Terms
- Payment Terms

otherTOS[xx]

MATCHING ALGORITHM



Data Access Requestor - Research Purpose Terms/ Data Use Categories

Purpose of use term

- Research Purpose
 - Methods Development
 - Population Research
 - Ancestry Research
 - Age Categories Research
 - Gender Categories Research
 - Research Control
 - Biomedical Research
 - Genetic Research
 - Drug Development Research
 - Disease Categories Research
 - For Profit Purpose



Data Use Limitation



General Research Use

Motif: Microscope
ID: DUO_0000042



Health/Medical/Biomedical

Motif: Hospital
ID: DUO_0000006



Disease Specific

Motif: Patient and bed
ID: DUO_0000007



Population Origins or Ancestry

Motif: Family tree
ID: DUO_0000011



No Restrictions

Motif: Disabled prohibited mark
ID: DUO_0000004

Data Use Requirements



Not-for-Profit Use

Motif: Dollar and prohibited mark
ID: DUO_0000018



Use Within Geographic Region

Motif: Pinned map
ID: DUO_0000022



Research Ethics Approval Required

Motif: Balance
ID: DUO_0000021

v2.3

Image credits: GA4GH DUO Stream: <https://github.com/EBISPOT/DUO>

FAIR?

Origins and definitions of FAIR

It has long been recognised that it is not sufficient simply to post data and other research-related materials onto the web and hope that the motivation and skill of the potential user would be sufficient to enable reuse. There is a need for various things, including contextual and supporting information (metadata) to allow those data to be discovered, understood and used. This notion has led a number of policy documents to list the key attributes that allow data to be reused and to demonstrate value. Arguably the most influential document is the OECD's 'Principles and Guidelines for Access to Research Data from Public Funding,'⁶ as it demonstrably led to and influenced a series of funder data policies.⁷ Although influential, it is clear from subsequent policies and reports that the OECD attributes needed to be further defined to make them more 'data centric', so that researchers, research institutions and data repositories would have a clearer understanding of the principles underlying useful data sharing.

Primary Recommendations and Actions

Step 1: Define and apply FAIR appropriately

Rec. 1: Definitions of FAIR

FAIR is not limited to its four constituent elements: it must also comprise appropriate openness, the assessability of data, long-term stewardship, and other relevant features. To make FAIR data a reality, it is necessary to incorporate these concepts into the definition of FAIR.

https://ec.europa.eu/info/sites/info/files/turning_fair_into_reality_1.pdf

Why is FAIR important?

Following this approach, we found that the annual cost of not having FAIR research data costs the European economy at least €10.2bn every year. In addition, we also listed a number of consequences from not having FAIR which could not be reliably estimated, such as an impact on research quality, economic turnover, or machine readability of research data. By drawing a rough parallel with the European open data economy, we concluded that these unquantified elements could account for another €16bn annually on top of what we estimated. These results relied on a combination of desk research, interviews with the subject matter experts and our most conservative assumptions.

<https://op.europa.eu/s/sniv>

Who does FAIR apply to?

Stakeholder groups assigned Actions

https://ec.europa.eu/info/sites/info/files/turning_fair_into_reality_1.pdf

1. **Research communities:** practitioners from all fields of humanities and science, clustered in groups around disciplinary interests, data types or cross-cutting grand challenges.
2. **Data services:** domain repositories, Research Infrastructures (ESFRIs) and E-Infrastructures, institutional provision, community and commercial tools and services.
3. **Data stewards:** support staff from research communities and research libraries, and those managing data repositories.
4. **Standards bodies:** formal organisations and consortia coordinating data standards and governing procedures relevant to FAIR, e.g. repository certification, curriculum accreditation.
5. **Global coordination fora:** the Research Data Alliance, CODATA, WDS Communities of Excellence, FORCE11, GO FAIR and other similar initiatives.
6. **Policymakers:** governments, international entities like OECD, research funders, institutions, publishers and others defining data policy.
7. **Research funders:** the European Commission, national research funders, charitable organisations and foundations, and other funders of research activity.
8. **Institutions:** universities and research performing organisations
9. **Publishers:** commercial and not-for-profit, paywall and Open Access publishers of research papers and data.

The NIH Big Data to Knowledge (BD2K, <https://datascience.nih.gov>) program is a trans-NIH data science program that is addressing data sharing by working to make biomedical data Findable, Accessible, Interoperable, and Reusable (FAIR; see <https://www.force11.org/group/joint-declaration-data-citation-principles-final>). Data and analytical resources generated by H3Africa are expected to conform to the FAIR principles.

<https://grants.nih.gov/grants/guide/rfa-files/RFA-rm-16-011.html>

FAIR – H3Africa Data

- RDA FAIR Data Maturity Model (https://www.rd-alliance.org/system/files/FAIR%20Data%20Maturity%20Model%20specification%20and%20guidelines_v1.00.pdf)

- AKA FAIR Indicators
- Indicators of 3 levels
 - Essential
 - Important
 - Useful
- 40 Indicators
 - 7 Findability Indicators
 - 12 Accessibility Indicators
 - 12 Interoperability Indicators
 - 9 Reusability Indicators

Table 1 FAIR data maturity model indicators

FAIR	ID	Indicator	Priority
F1	RDA-F1-01M	Metadata is identified by a persistent identifier	●●● Essential
F1	RDA-F1-01D	Data is identified by a persistent identifier	●●● Essential
F1	RDA-F1-02M	Metadata is identified by a globally unique identifier	●●● Essential
F1	RDA-F1-02D	Data is identified by a globally unique identifier	●●● Essential
F2	RDA-F2-01M	Rich metadata is provided to allow discovery	●●● Essential
F3	RDA-F3-01M	Metadata includes the identifier for the data	●●● Essential
F4	RDA-F4-01M	Metadata is offered in such a way that it can be harvested and indexed	●●● Essential
A1	RDA-A1-01M	Metadata contains information to enable the user to get access to the data	●● Important
A1	RDA-A1-02M	Metadata can be accessed manually (i.e. with human intervention)	●●● Essential
A1	RDA-A1-02D	Data can be accessed manually (i.e. with human intervention)	●●● Essential
A1	RDA-A1-03M	Metadata identifier resolves to a metadata record	●●● Essential
A1	RDA-A1-03D	Data identifier resolves to a digital object	●●● Essential
A1	RDA-A1-04M	Metadata is accessed through standardised protocol	●●● Essential
A1	RDA-A1-04D	Data is accessible through standardised protocol	●●● Essential
A1	RDA-A1-05D	Data can be accessed automatically (i.e. by a computer program)	●● Important
A1.1	RDA-A1.1-01M	Metadata is accessible through a free access protocol	●●● Essential
A1.1	RDA-A1.1-01D	Data is accessible through a free access protocol	●● Important
A1.2	RDA-A1.2-01D	Data is accessible through an access protocol that supports authentication and authorisation	● Useful
A2	RDA-A2-01M	Metadata is guaranteed to remain available after data is no longer available	●●● Essential
I1	RDA-I1-01M	Metadata uses knowledge representation expressed in standardised format	●● Important
I1	RDA-I1-01D	Data uses knowledge representation expressed in standardised format	●● Important
I1	RDA-I1-02M	Metadata uses machine-understandable knowledge representation	●● Important
I1	RDA-I1-02D	Data uses machine-understandable knowledge representation	●● Important
I2	RDA-I2-01M	Metadata uses FAIR-compliant vocabularies	●● Important
I2	RDA-I2-01D	Data uses FAIR-compliant vocabularies	●● Important

FAIR – H3Africa Data

- RDA FAIR indicators assessment

(<https://docs.google.com/spreadsheets/d/1vaRf4PXJ7FvxLZpIKXRIkiEYDSZP5ajsb3KnETz7Em0/edit#gid=1375297260>)

	Sub-principle	ID	Indicator	Priority	Assessment overall	Assessment Essential
Findable	F1	RDA-F1-01M	Metadata is identified by a persistent identifier	Essential	1	1
Findable	F1	RDA-F1-01D	Data is identified by a persistent identifier	Essential		
Findable	F1	RDA-F1-02M	Metadata is identified by a globally unique identifier	Essential		
Findable	F1	RDA-F1-02D	Data is identified by a globally unique identifier	Essential		
Findable	F2	RDA-F2-01M	Rich metadata is provided to allow discovery	Essential		
Findable	F3	RDA-F3-01M	Metadata includes the identifier for the data	Essential		
Findable	F4	RDA-F4-01M	Metadata is offered in such a way that it can be harvested and indexed	Essential		
Accessible	A1	RDA-A1-01M	Metadata contains information to enable the user to get access to the data	Important		

D	Indicator	Description	Assessment details
RDA-F1-01M	Metadata is identified by a persistent identifier	This indicator evaluates whether or not the metadata is identified by a persistent identifier. A persistent identifier ensures that the metadata will remain findable over time and reduces the risk of broken links.	The persistence of an identifier is determined by the commitment of the organisation that assigns and manages the identifier, so the evaluation of this indicator needs to take into account the persistence policy of that organisation. Such a commitment could be expressed by a university or research institute, by a research infrastructure or by an organisation that issues formal identifiers, such as the International DOI Foundation. A possible way to evaluate this indicator is to verify that the identifier used for the metadata is listed in a registry service like the RDA-endorsed FAIRsharing.
RDA-F1-01D	Data is identified by a persistent identifier	This indicator evaluates whether or not the data is identified by a persistent identifier. A persistent identifier ensures that the data will remain findable over time and reduces the risk of broken links.	The persistence of an identifier is determined by the commitment of the organisation that assigns and manages the identifier, so the evaluation of this indicator needs to take into account the persistence policy of that organisation. Such a commitment could be expressed by a university or research institute, by a research infrastructure or by an organisation that issues formal identifiers, such as the International DOI Foundation. A possible way to evaluate this indicator is to verify that the identifier used for the data is listed in a registry service like the RDA-endorsed FAIRsharing.



FAIR – H3Africa Data

- Assessed the H3Africa Chip project data according to the indicators: - 100% on the Assessment for the level of Essential, 76.19% on Assessment Non-Essential with an overall Assessment score of 87.8%

H3ACHIP ACCME RDA FAIR indicators assesment TEMPLATE ☆ 📁 ☁

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	D	E	F	G	H
31	Metadata includes qualified references to other metadata	Important	1	-	1
32	Metadata include qualified references to other data	Useful	1	-	1
33	Plurality of accurate and relevant attributes are provided to allow reuse	Essential	1	1	-
34	Metadata includes information about the licence under which the data can be reused	Essential	1	1	-
35	Metadata refers to a standard reuse licence	Important	0	-	0
36	Metadata refers to a machine-understandable reuse licence	Important	1	-	1
37	Metadata includes provenance information according to community-specific standards	Important	0	-	0
38	Metadata includes provenance information according to a cross-community language	Useful	0	-	0
39	Metadata complies with a community standard	Essential	1	1	-
40	Data complies with a community standard	Essential	1	1	-
41	Metadata is expressed in compliance with a machine-understandable community standard	Essential	1	1	-
42	Data is expressed in compliance with a machine-understandable community standard	Important	1	-	1
43					
44					
45		Sum	36	20	16
46		Score total	87.80%	100.00%	76.19%
47	relevant category)	Score applicable	87.80%	100.00%	76.19%
48		% NA	0.00%	0.00%	0.00%

- **Harnessing Data Science for Health Discovery and Innovation in Africa**
 - **Research Hubs:** Advance and demonstrate feasibility of data science research and innovation to improve health in Africa
 - **Training:** Increase capacity for data science research in Africa
 - **ELSI Research:** Explore Ethical, Legal, and Social Implications of data science research from an African perspective and contribute to policy discussion on the continent
 - **Open Data Science Platform & Coordination Center:** Facilitate the development of a trans-African network of data scientists

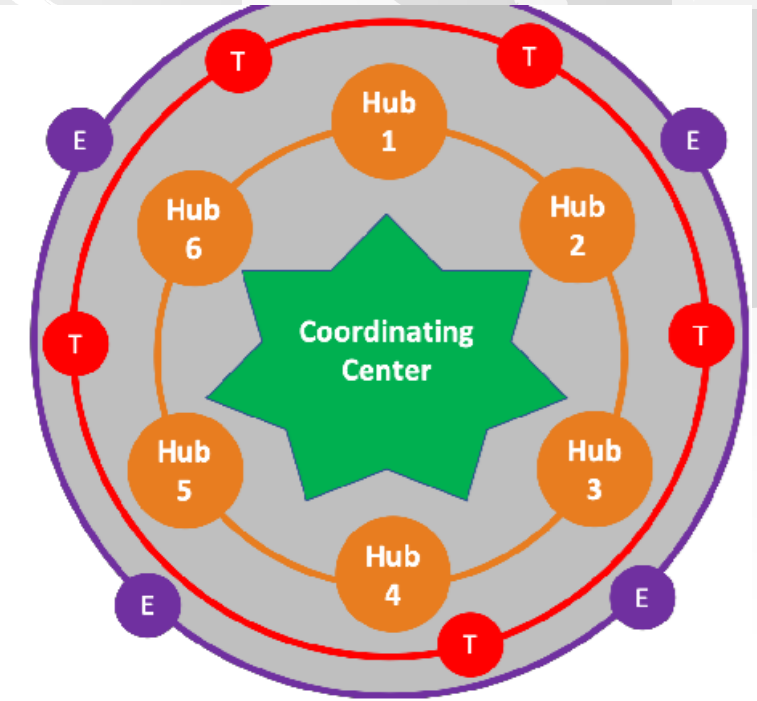


Image and text extracted from: <https://commonfund.nih.gov/africadata>

eLwazi Open Data Science Platform



- To develop an African Open Data Science Gateway, including the eLwazi platform and associated resources, to support the Harnessing Data Science for Health Discovery and Innovation in Africa (DS-I Africa) consortium and beyond. eLwazi, will be a flexible, scalable Open Data Science Platform enabling the implementation of data science for health, that is relevant to the African context.
- Facilitate the development of a trans-African network of data scientists

Ulwazi is the Xhosa word meaning “knowledge” or “information”,
and Olwazi means big rock in Luganda



eLwazi

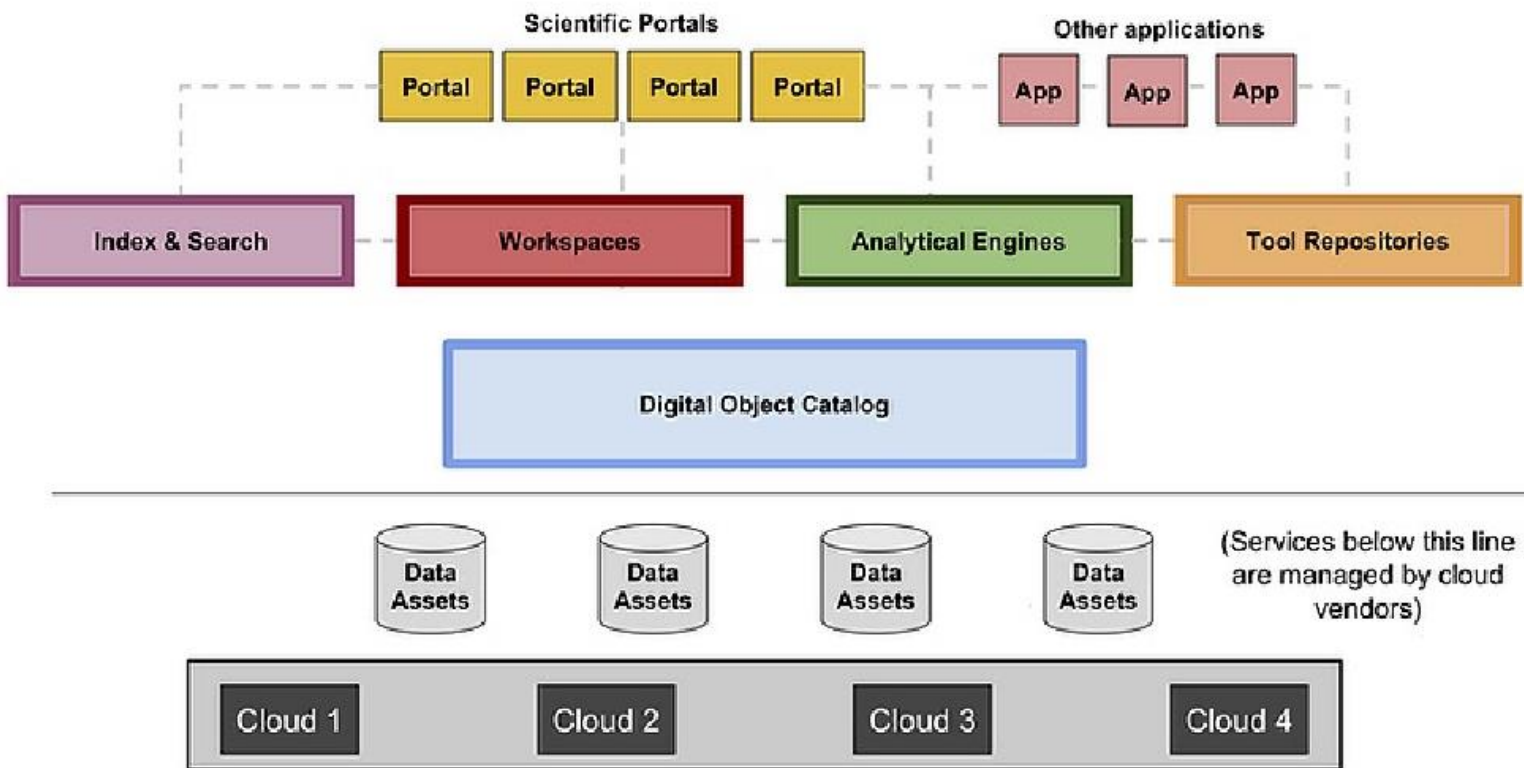


eLwazi Open Data Science Platform

Data biosphere



- <https://www.databiosphere.org/>



Platforms



GEN3
Data Commons



Dockstore

Applications

All of Us
RESEARCH PROGRAM



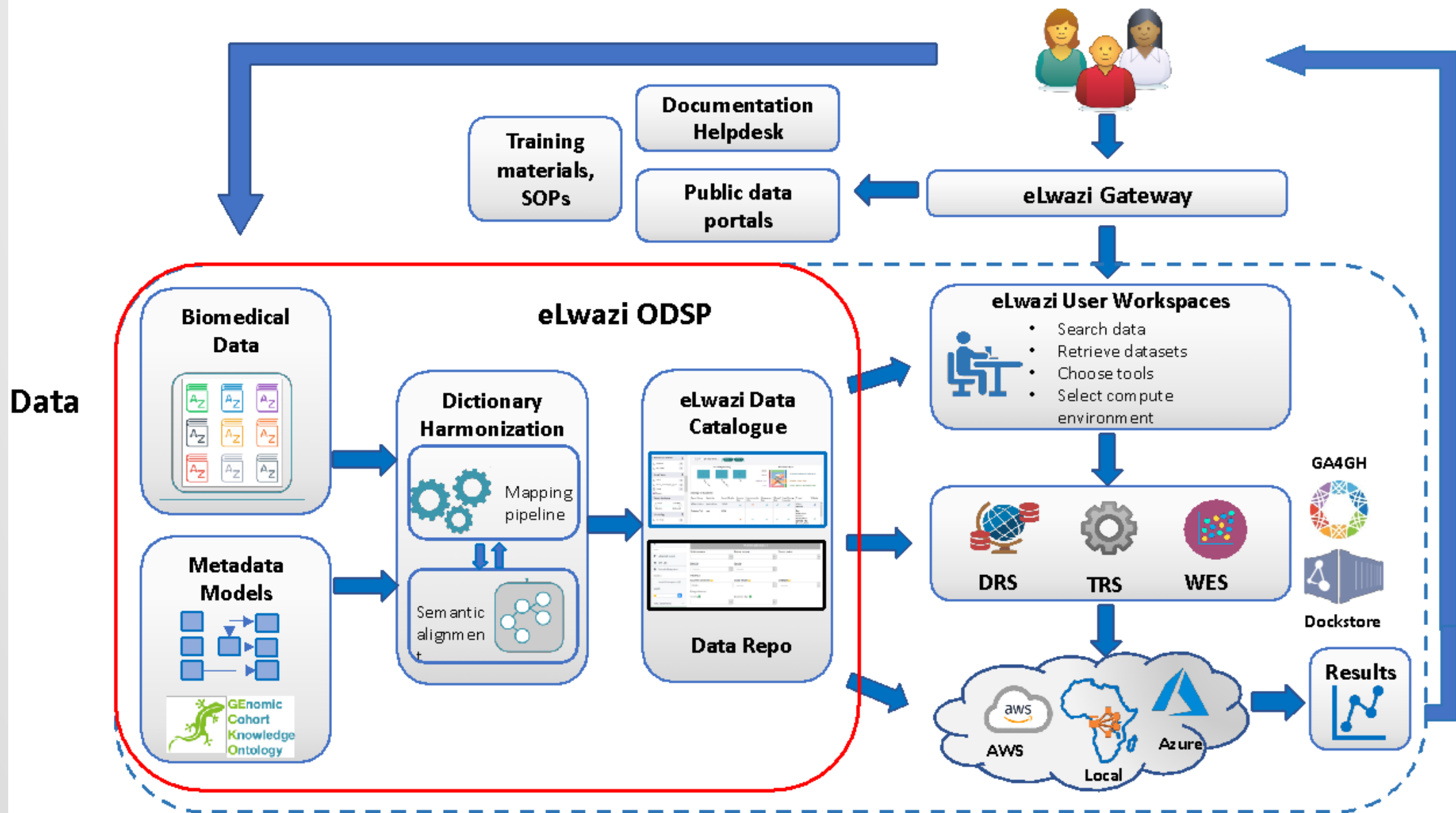
NHGRI AnVIL

biobank^{uk}
Improving the health of future generations

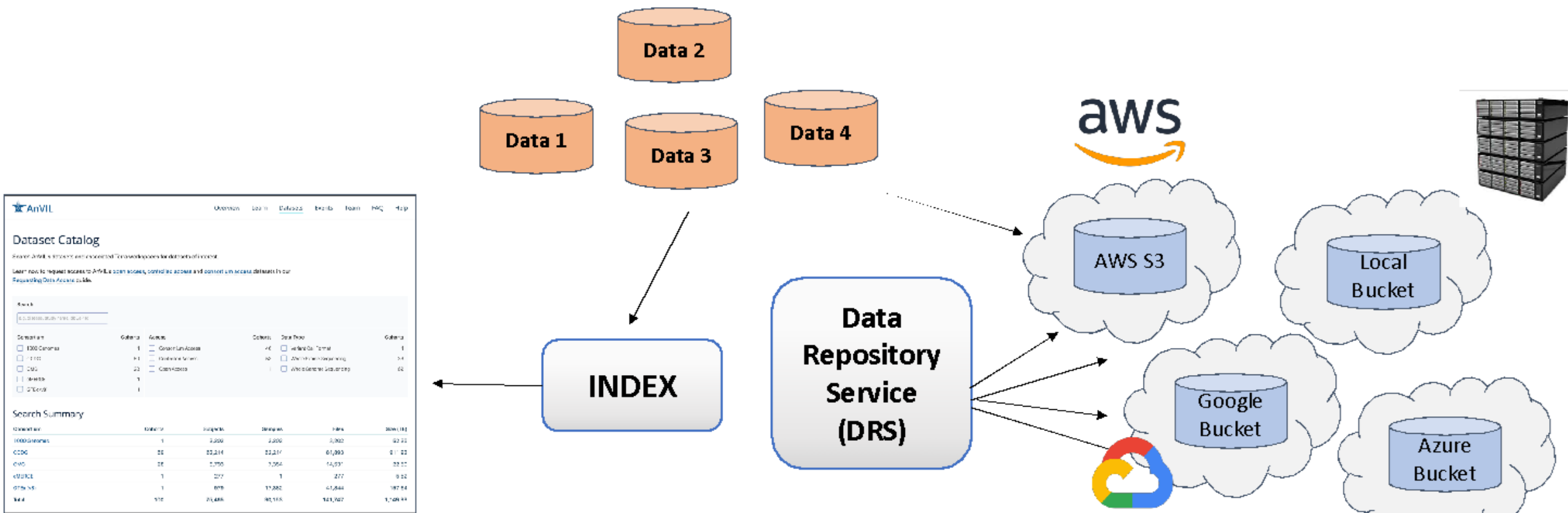


Global Alliance
for Genomics & Health

Overview of the ODSP



Data registry



AnVIL Dataset Catalog

Search Summary

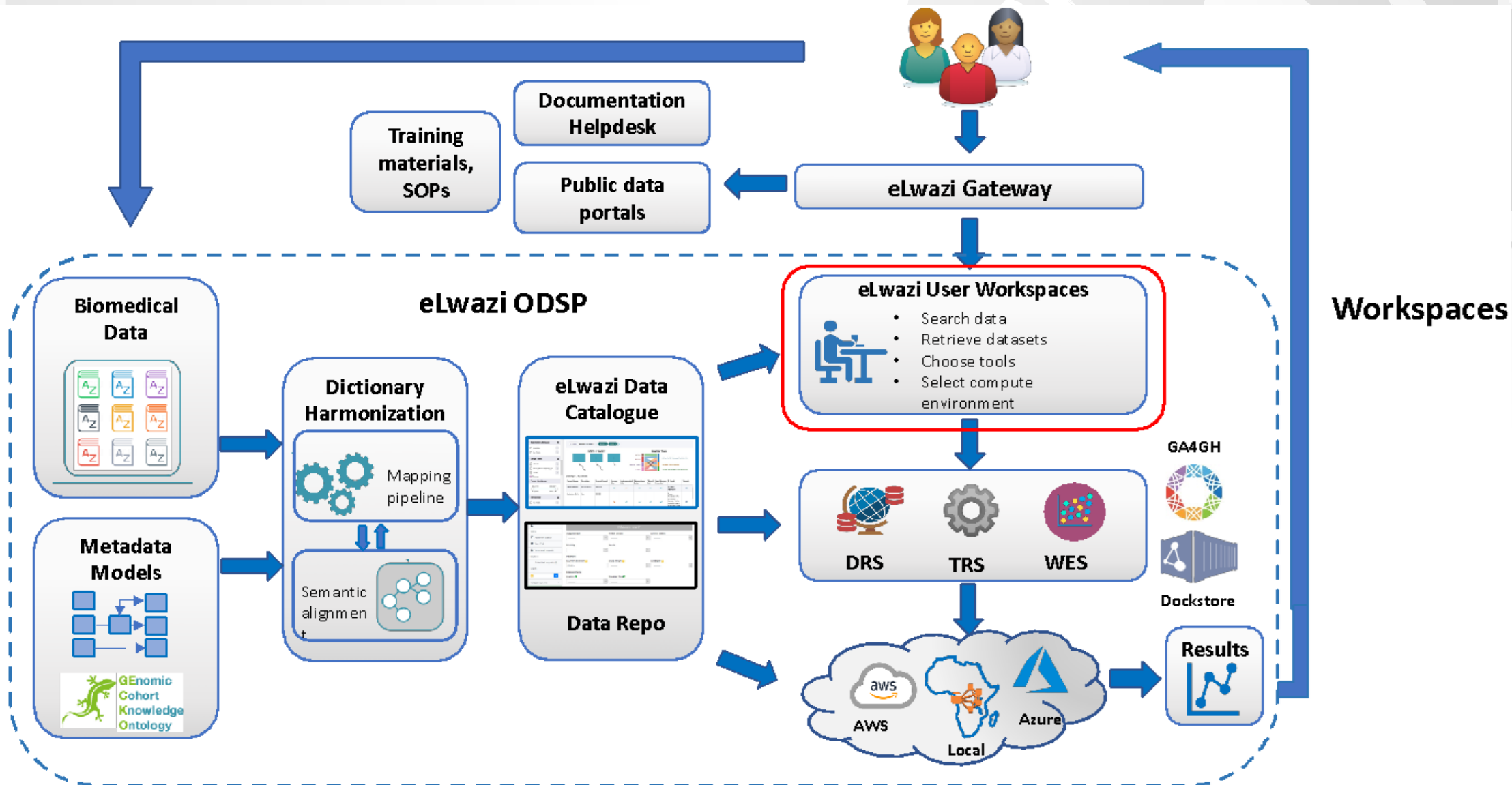
Category	Count	Percentage	Count	Percentage	Count	Percentage
1000 Genomes	1	2.00%	2,200	2.20%	2,200	22.22%
CCDC	22	22.22%	22,217	22.21%	61,993	61.11%
dbSnp	22	2.20%	2,200	2.20%	14,531	14.53%
dbSNP	1	1.00%	227	2.27%	227	2.27%
dbSNP	1	1.00%	678	6.78%	17,282	17.28%
dbSNP	1	1.00%	21,485	21.48%	95,115	95.11%
dbSNP	1	1.00%	141,507	141.50%	1,708,444	1,708.44%

Data Browser
query interface

Implementations for
each computing
environment



Data Analysis



Data Analysis

Welcome to eLwazi

The eLwazi Open Data Science Platform is a project powered by Terra for biomedical researchers to **access data, run analysis tools, and collaborate.**

Find how-to's, documentation, video tutorials, and discussion forums [↗](#)

View Workspaces

Workspaces connect your data to popular analysis tools powered by the cloud. Use Workspaces to share data, code, and results easily and securely.



View Examples

Browse our gallery of showcase Workspaces to see how science gets done.



Browse Data

Access data from a rich ecosystem of data portals.



<https://elwazi.terra.bio/>

Workspaces bring together metadata, data, and compute environments through a web-based user interface. Access for beginner and advanced users

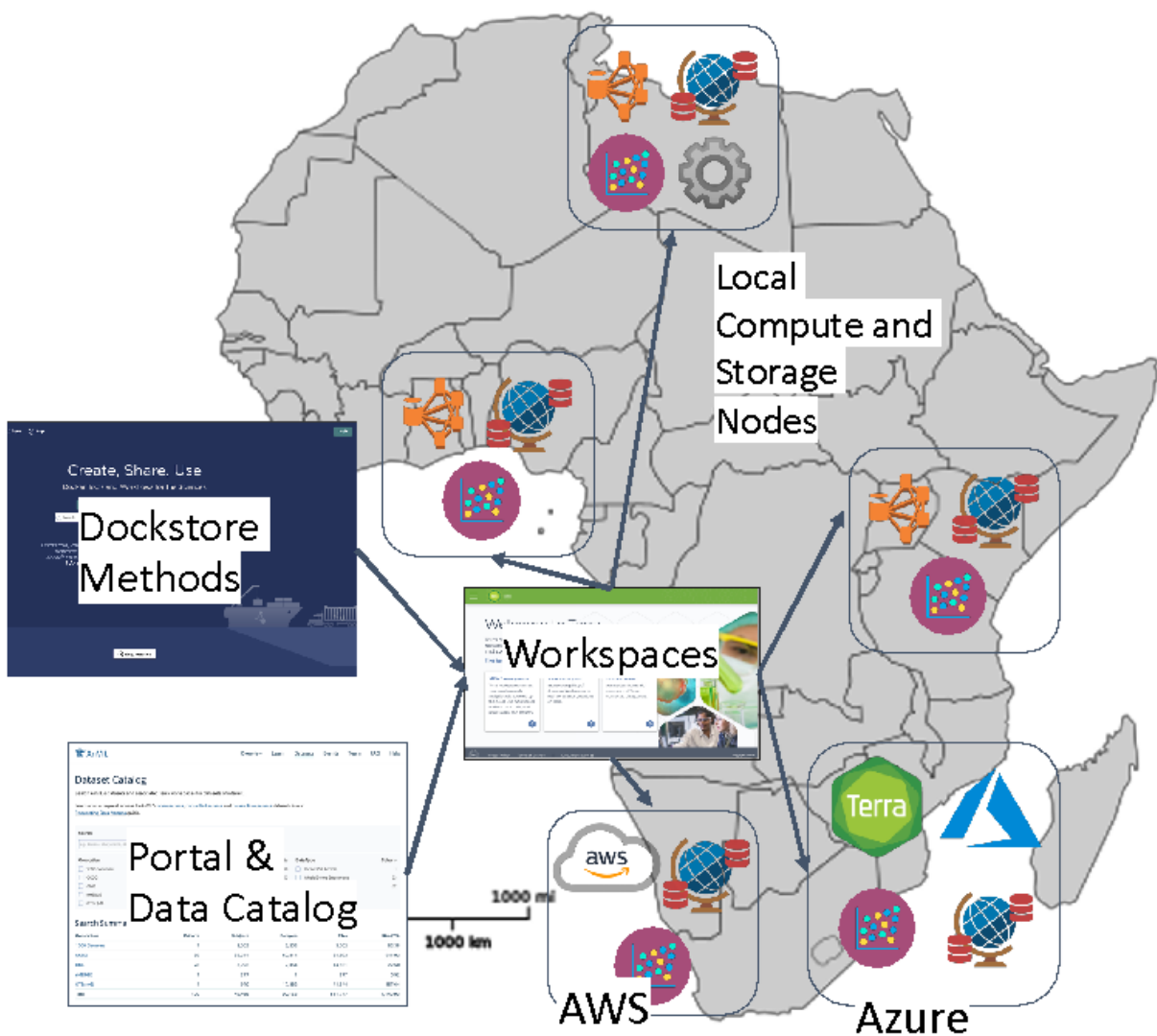
Log into project work space

Identify datasets from registry

Extract relevant data

Select tools from registry

Run in Cloud



Cloud & HPC Environments


-  AWS,  Azure, and  Local

 **DRS** for data access

 **WES** for compute

 **TRS** for workflow sharing

 **Workspaces for:**

-  Terra Data modeling and access
- Compute on data in various locations
- Interactively analyze via Notebooks
- Sharing and Collaborating

Conclusions

GROWTH OF DATA IN THE WORLD TODAY

90 percent of the world's data was created in the last two years.

Image reference: https://data.bsa.org/wp-content/uploads/2015/12/bsadatastudy_en.pdf

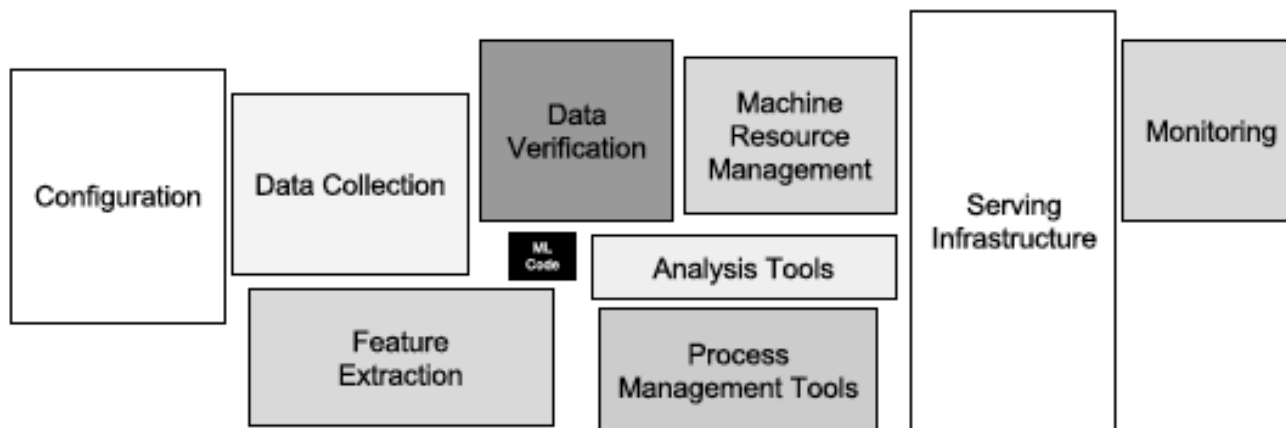
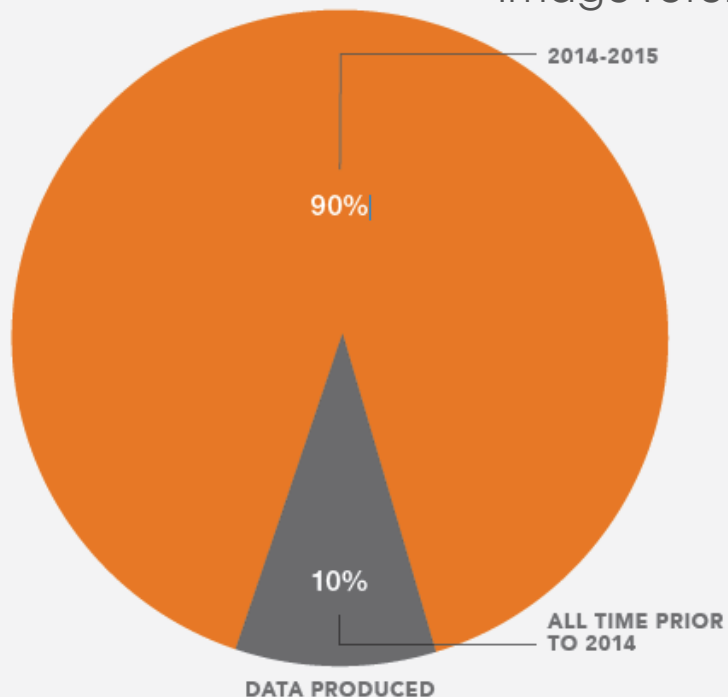


Figure 1: Only a small fraction of real-world ML systems is composed of the ML code, as shown by the small black box in the middle. The required surrounding infrastructure is vast and complex.

Hidden Technical Debt in Machine Learning Systems:

<https://proceedings.neurips.cc/paper/2015/file/86df7dcfd896fcdf2674f757a2463eba-Paper.pdf>

Acknowledgements

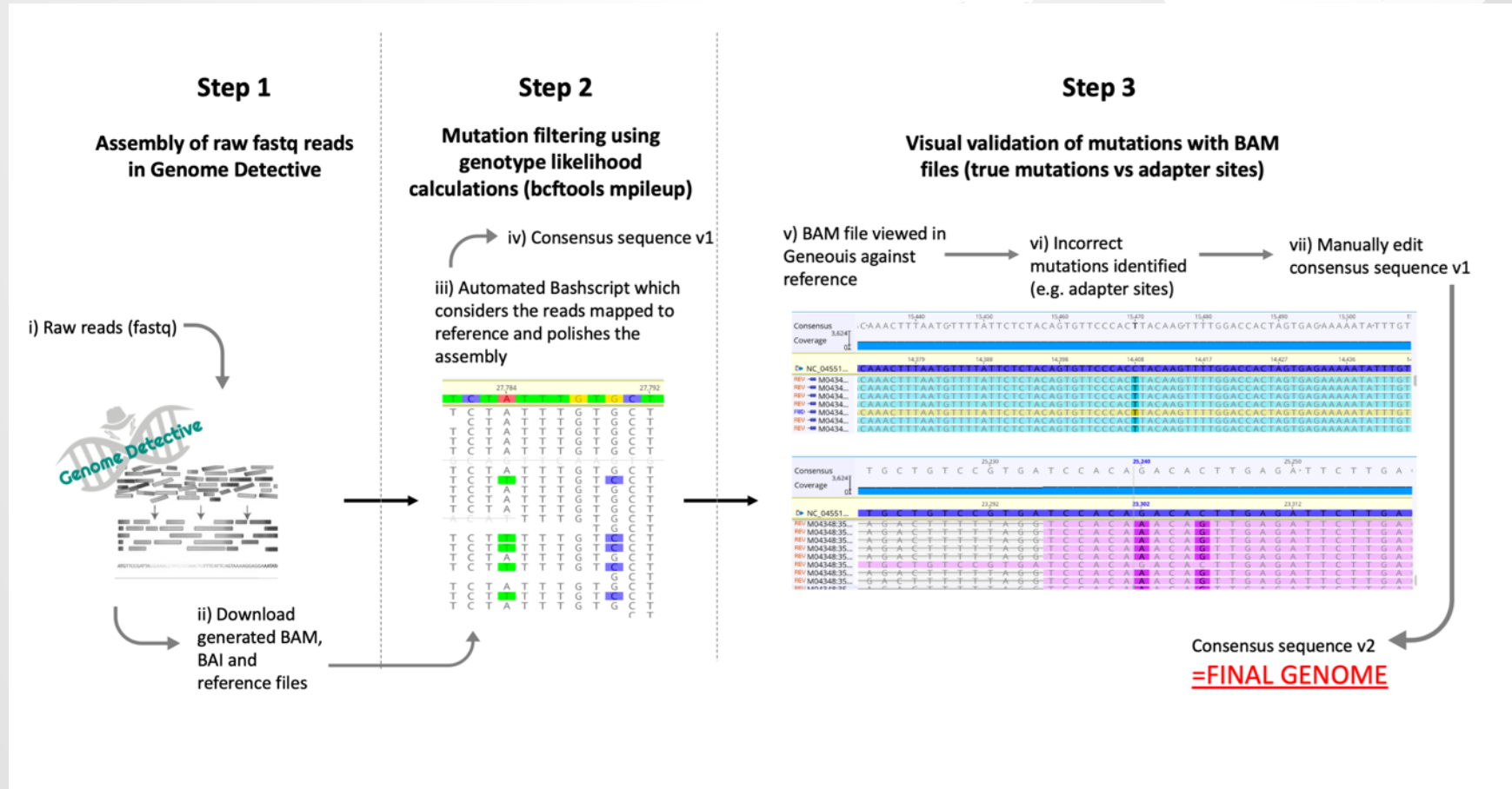


H3ABioNet Consortium

National Human Genome Research Institute: U24HG006941

Assembly of SARS-CoV-2

Objective: To quickly and accurately generate high-quality whole genomes of SARS-CoV-2 that are well annotated and available in public databases.



Slides courtesy of: Houriiyah Tegally, San Emmanuel James and Prof. Tulio De Oliveira, KRISP, UKZN

<https://www.genomedetective.com/app/typingtool/virus/>

<https://www.science.org/doi/10.1126/science.abj4336>

www.h3abionet.org #h3abionet

Data curation for SARS-CoV-2

MicDetect AGM 2020 session description:

The KRISP MicDetect project would like to provide a practical bioinformatics session on cleaning, assembling, depositing and analyzing SARS-CoV-2 genomes.

The hands on practical session will entail:

- Using Genome Detective platform to assemble SARS-CoV-2 genomes
- Using genotype likelihood to perform mutation filtering
- Using next clade to QC mutations and refine consensus sequences
- Using Geneious to perform manual QC of mutations to produce high quality genomes
- Basic phylogenetic analysis for genomic epidemiology studies of COVID-19

The session would require H3ABioNet members to: have access to a laptop or desktop with Linux or Mac OS with the following tools installed:

- **Bedtools:** <https://bedtools.readthedocs.io/en/latest/content/installation.html>
- **BCFTools:** <http://www.htslib.org/download/>
- **Geneious prime** (trial version is sufficient): <https://www.geneious.com/free-trial/>

Consensus v1 Manual correction of wrong mutations Consensus v2 = **FINAL GENOME**

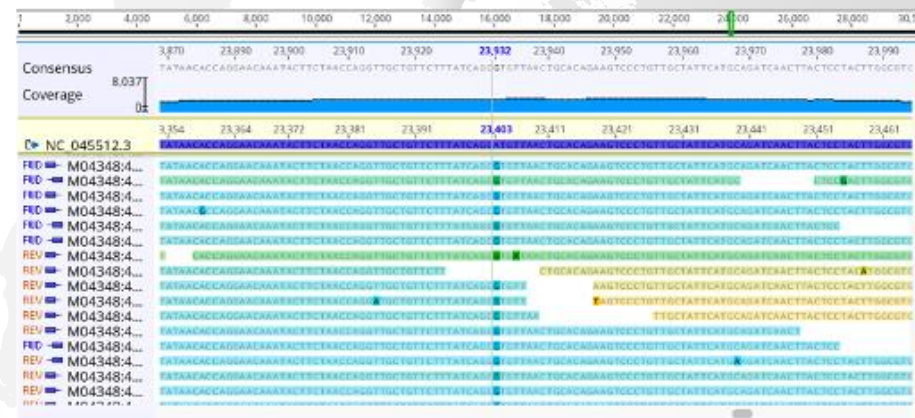
How to spot wrong mutations:

- Indels
- Mutations that cause stop codons in the middle of a genes
- Apparent adapter sites not being trimmed out properly
- “Expected” mutations

How to correct genomes:

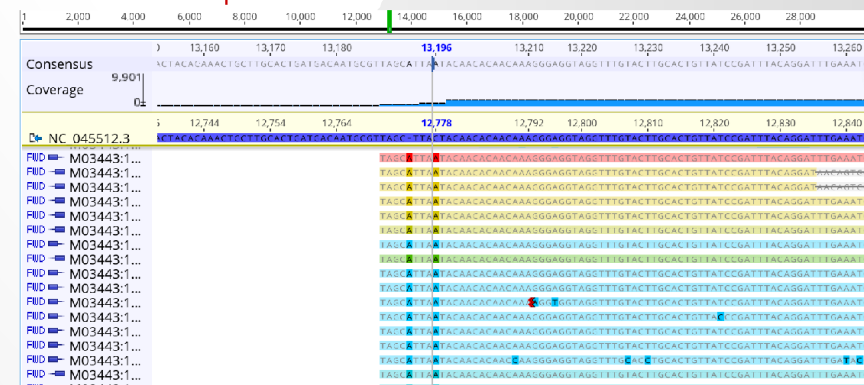
- Upload the FASTA sequencing into Genome Detective
- Find location to be edited by inspecting the mutation on Genome Detective and copying a short length of sequence before or after the mutation
- Search for that sequence in the FASTA text file
- Manually edit sequence in FASTA text file (back to the reference base if mutation is wrong)
- Be very careful at this step to not mistakenly insert other false mutations

Correct mutation



Here, we are looking at a correctly called mutation, indicated by consistent base change (23403A>G), mostly in the middle of long reads.

Untrimmed adapter sites

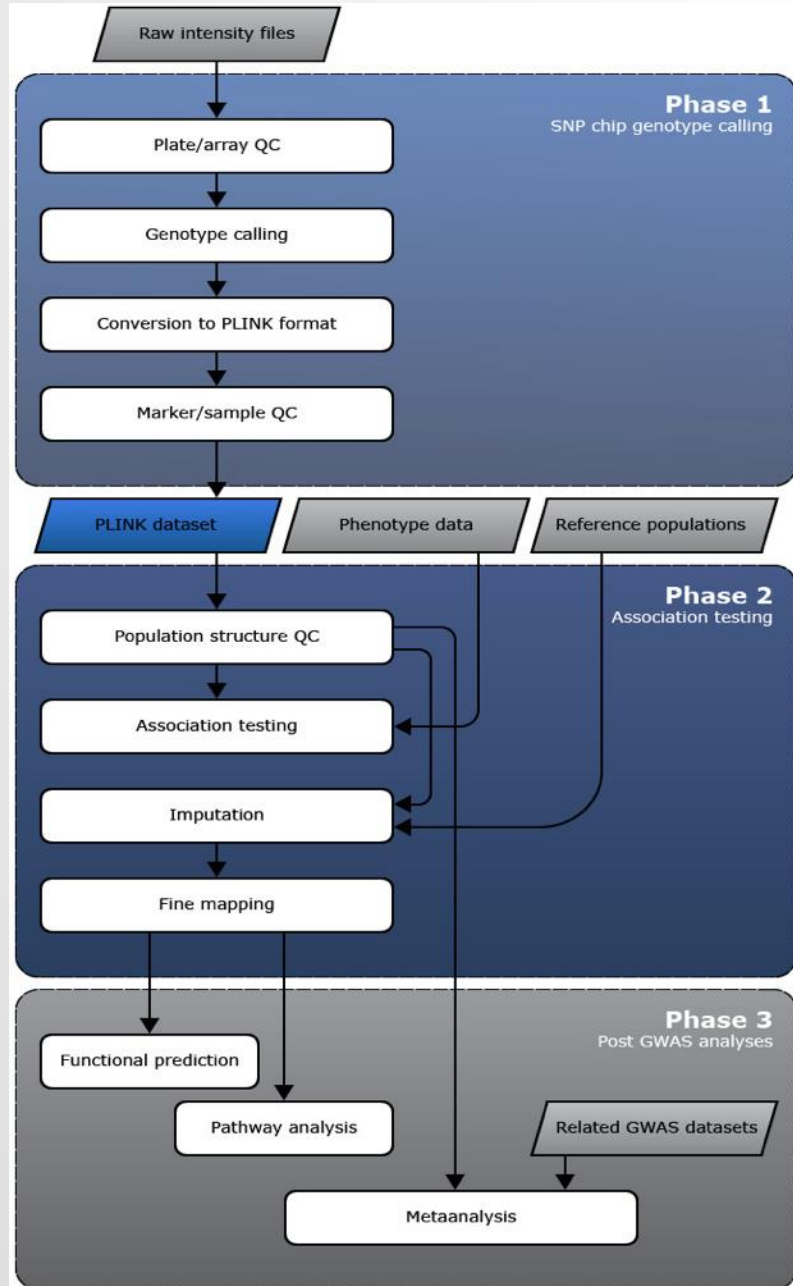


Here, we are looking at an incorrectly called mutation (12778C>A), which appears to be the result of an adapter site that was not trimmed. These sites will often appear at the end of reads.

Slides courtesy of: Houriiyah Tegally, San Emmanuel James and Prof. Tulio De Oliveira, KRISP, UKZN

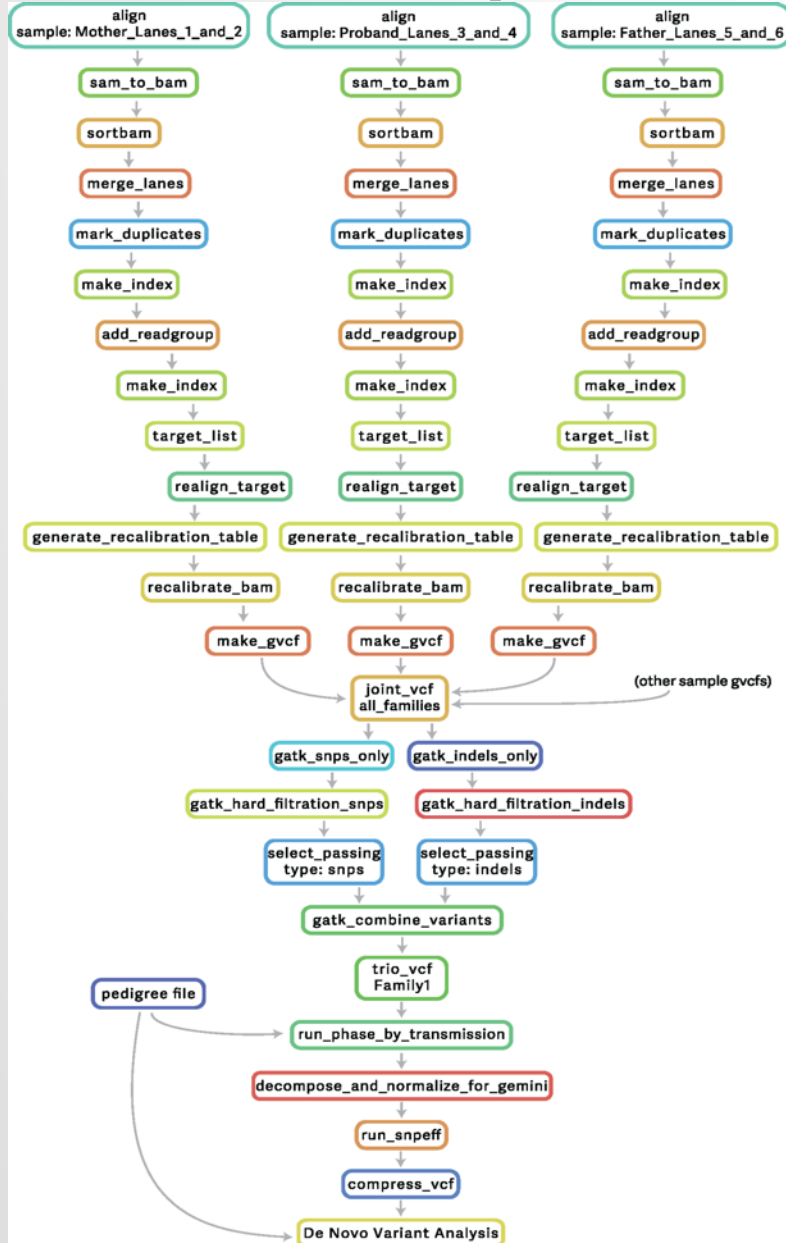
<https://www.genomedetective.com/app/typingtool/virus/>

Reproducible Science - Pipelines



- Bioinformatics analyses directing files through a series of transformations and programs to a final output i.e. a computational pipeline
- Transformations typically done by third-party executable command line software written for Unix-compatible operating systems
- Manually started when previous transformation step completed e.g. qsub command run

Reproducible Science - Workflows



- A workflow is a description of a process (pipeline) that consists of a series of tasks connected in the form a directed graph
- Tasks can be defined as single units of work e.g. split files
- A workflow comprises of an initial unique task and ends with a unique terminal task
- Completion of a task can initiate one or more tasks
- Enables automation e.g. run a pipeline from start to finish without manual input (pipeline)

H3ABioNet Workflows project

