



IDEA4RC vision

- All observational data for all adult patients with rare cancer in Europe is made available for research, innovation, and clinical decision making
- With no registration burden for clinicians
- With a simple and transparent mechanism to make data available for all stakeholders
- While protecting privacy of individuals and maintaining autonomy and data control of data providers

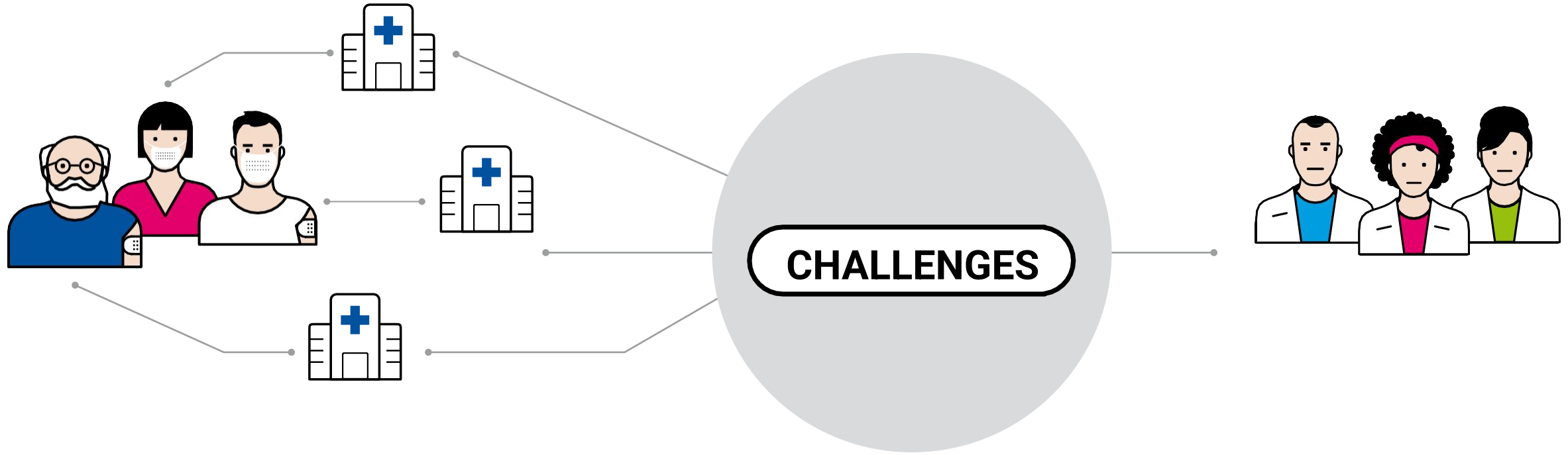


European
Reference
Networks

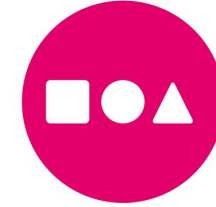


EURACAN





**Data heterogeneity
and availability**



**Data findability
and understanding**



**Data security
and trust**



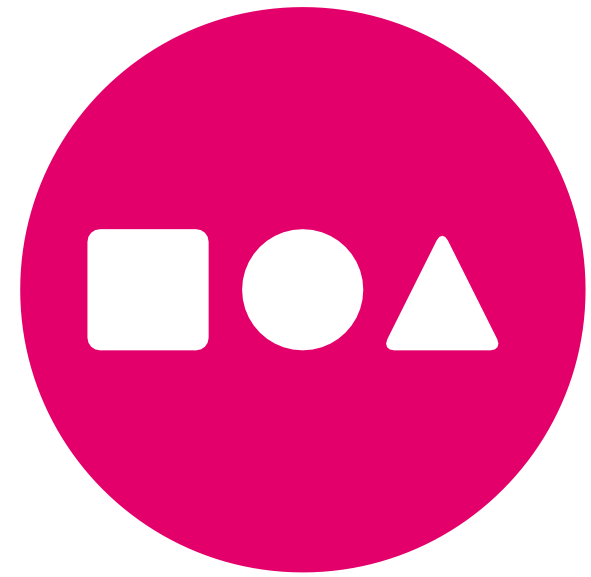
Data access



Challenge 1

Data heterogeneity and availability

- Electronic Health Records (EHRs) contain structured and unstructured data
- EHR are written in different EU languages Data
- Quality is highly variable
- Different formats, often incompatible
- Different standards and data models





Challenge 2

Data findability and Understanding

- Lack of knowledge about the possibility that a dataset already exists
- Metadata without details: difficult to understand whether the dataset meets the needs
- Different languages
- Different vocabulary/terminology used

Challenge 3

Data security and trust



- Traditional perimeter-based security is not enough in the modern IT landscape
- Complying with the security requirements stated in the GDPR, EHDS, and NIS2 entails an increase in cost and expertise that are generally not available at hospitals
- How to maintain the control over data within each data owner?
- How to enable personal health data analysis if data cannot be transferred across different institutions in context like rare cancers or rare diseases

Challenge 4

Data access

- Difficult to understand which is the organisation/institution responsible to grant data access
- Lengthy, heterogeneous and complex approval processes
- Different governance restrictions to data access
- Lack of interoperability/collaborations across data access applications
- Lack of trust: what happens after access is granted? Who are the actors involved?



IDEA4RC solutions

1. NLP for Medical Entity Extraction
2. NLP Query/cohort Builder for Sarcomas and Head & Neck Cancer Data Models
3. Conversational AI in IDEA4RC Applications
4. Federated AI in IDEA4RC

NLP for Medical Entity Extraction

An example

Me llamo **Ignacio**, tengo **58 años**. En junio de 20254 **sufrí un infarto de miocardio** por el que fui atendido de urgencia en el hospital. Me **realizaron un cateterismo** y estuve ingresado varios días en la unidad coronaria. Tras el alta, me prescribieron medicación diaria: **ácido acetilsalicílico (aspirina)** como antiagregante plaquetario, **atorvastatina** para el control del colesterol, **bisoprolol como betabloqueante**, y **ramipril**. En la **revisión de enero de 2025**, me **diagnosticaron hipertensión arterial**. Actualmente estoy en **seguimiento con el cardiólogo** y el médico de cabecera para **controlar la tensión** y prevenir nuevos eventos cardiovasculares.

Patient

Condition

Procedure

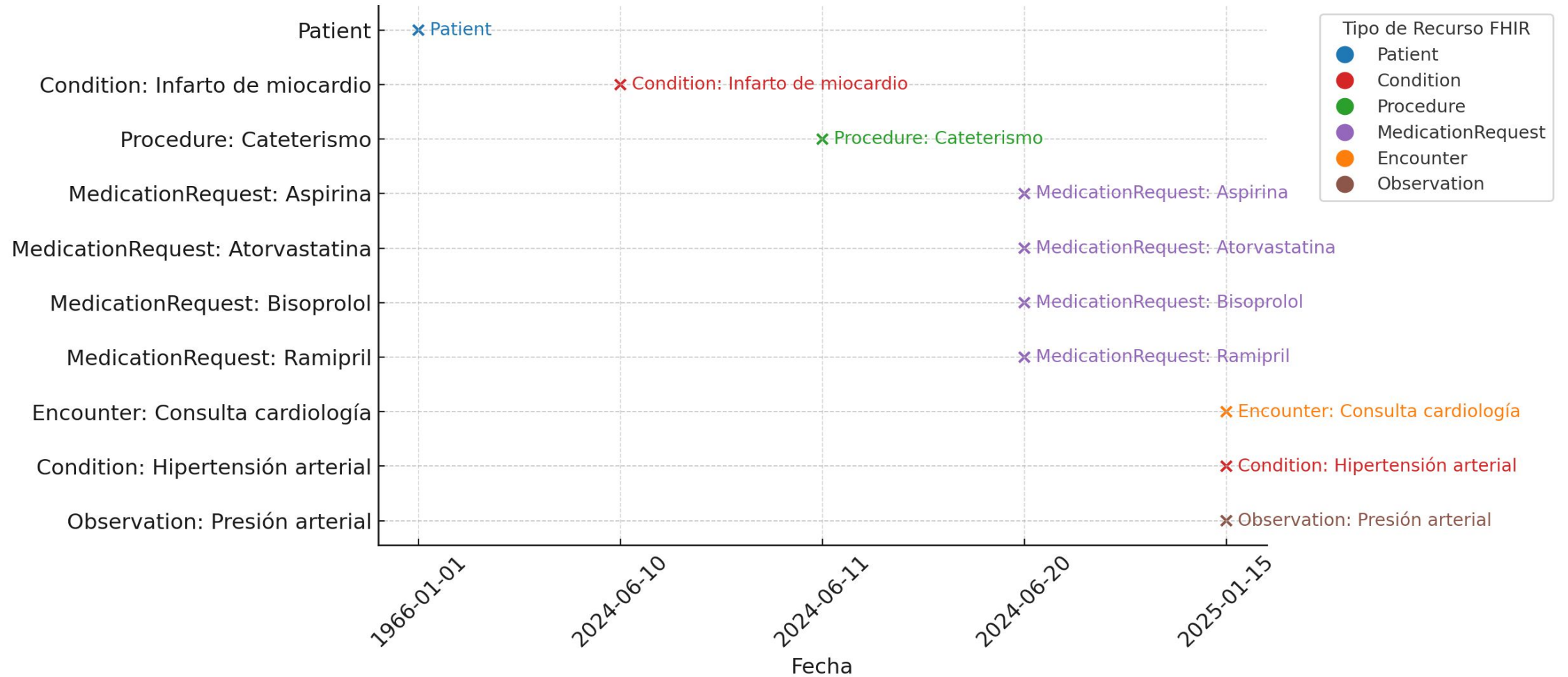
MedicationRequest

Encounter

Observation

NLP for Medical Entity Extraction

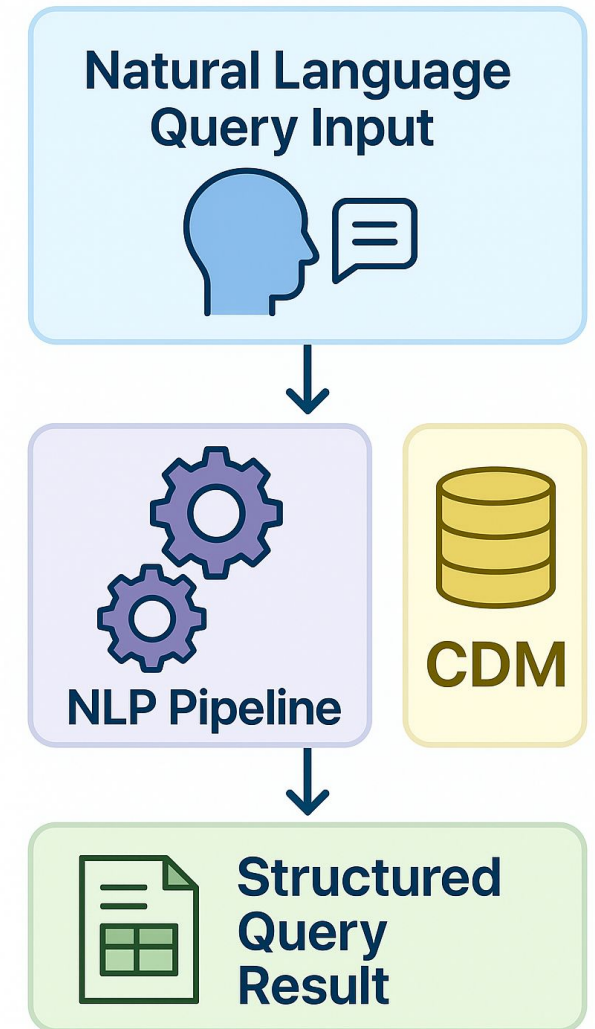
An example



NLP Query/cohort Builder for Cancer Data Models

The Scenario in IDEA4RC

- Clinicians want to use directly the tools
- No previous knowledge about OMOP or FHIR data structure need to be assumed
- The tools need to provide easy access to the most common queries



NLP Query Builder for Cancer Data Models

Innovation in IDEA4RC



Validate the identified data:

patients diagnosed in the

lower inner quadrant of breast ✓

that went under

lumpectomy ✓

All terms have been validated!

Cancel

 Send to AI

Federate Learning

what is it?

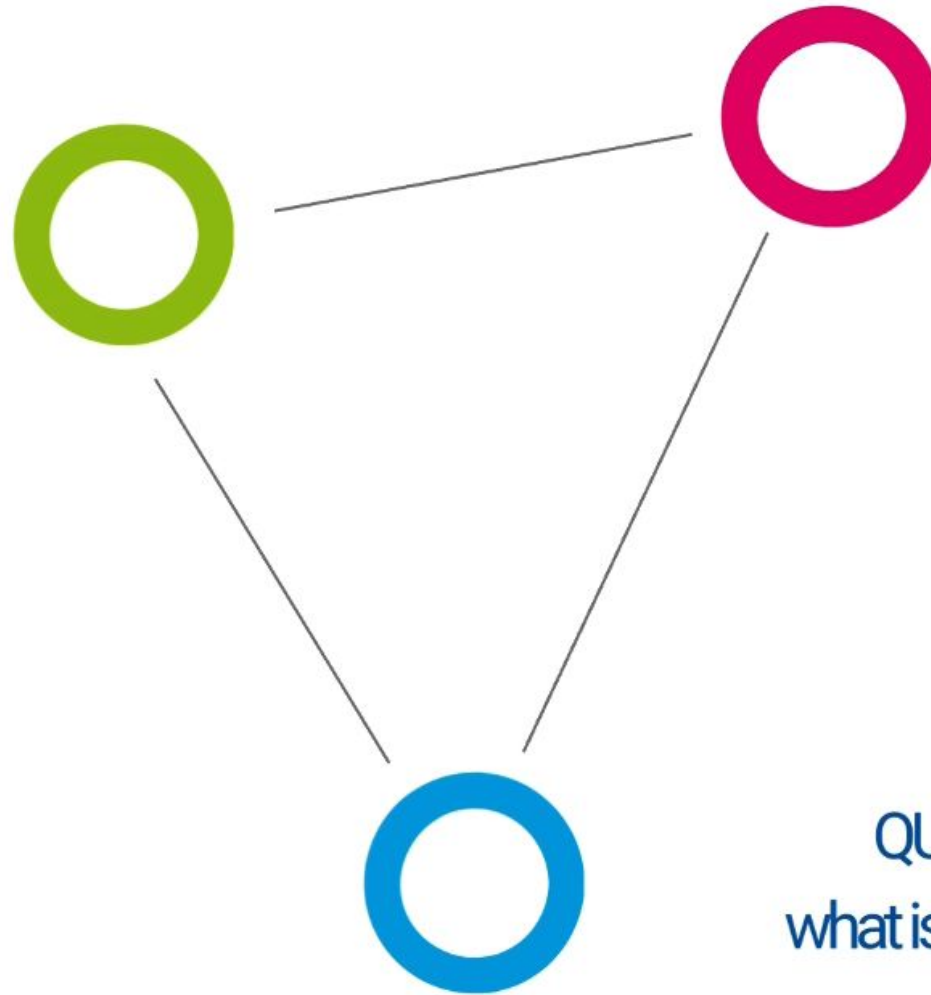
Federated Learning is a machine learning technique that enables multiple devices to collaboratively train a model without sharing raw data.

Instead, each device trains the model locally with its own data, and only the model updates (gradients) are sent to a central server for aggregation.

This approach is useful for privacy-sensitive data, as it keeps individual data secure and private throughout the training process

Federate Learning

An example



QUESTION:
what is the average
age of patients?

AI-Powered Innovations in IDEA4RC

NLP for Clinical Text

- Extracts insights from multilingual free text
- Standard structured data

Query Builder

- No need to know OMOP/FHIR
- Easy cohort creation for clinicians

<https://www.idea4rc.eu/>

Conversational AI

- Smart assistant for platform navigation
- Trained on IDEA4RC docs

Federated Learning

- Enable privacy-preserving AI across 11 centers
- Zero Trust, GDPR-compliant