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ARNAUD

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CENTRE DE SOINS

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CLINICAL DATA FOR GENOMIC ANALYSIS

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Diagnostic challenge in rare diseases

A (very) large knowledge

- There are more than **10000** rare diseases
- A geneticist only has **1** brain

Linkeropathies

Kabuki
Syndrome



Ciliopathies

Familial
mediterranean
fever

22q11.2
Syndrome

Williams-Beuren
Syndrome

Increasing numbers of discoveries

2023	Jan	Feb	Mar	Apr	May	Jun
New	54	38	41	53	48	39
Updated	422	257	531	415	310	380

Jul	Aug	Sep	Oct	Nov	Dec
20	42	32	37	46	37
415	449	316	411	306	334

OMIM.org





Physicians needs computer helps since long time...

Clinical geneticists were early adopters of software as clinical decision support

"For precision medicine human and artificial intelligence need to join efforts." (Peter Krawitz)

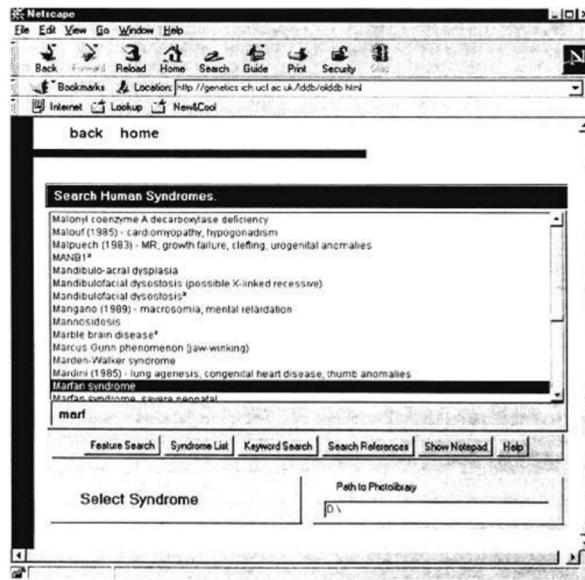
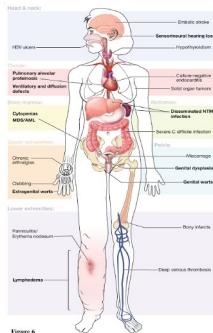


FIG. 1. Syndrome list search screen.



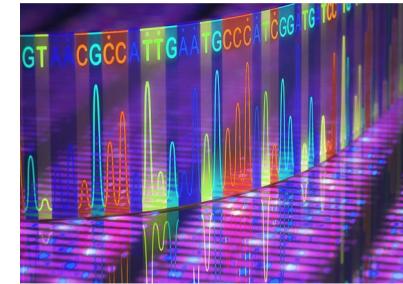
Physicians : how we used to do



1. Recognition of "gestalt"
- and 2. association of symptoms : "phenotype"

Physicians select a targeted sequencing analysis according to their knowledge

"Phenotype - first"

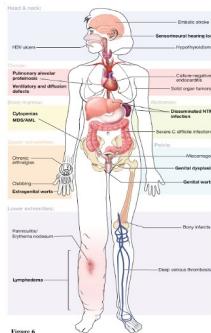


3. DNA analysis/sequencing : "genotype"

Which rare diseases do I think ? (precise query)



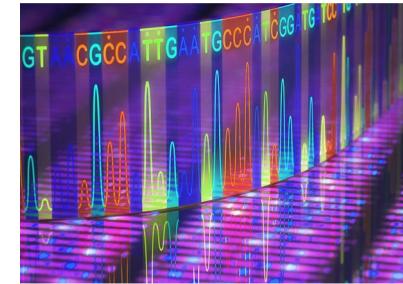
Physicians : today in the genomic medicine era



1. Recognition of "gestalt"
- and 2. association of symptoms : "phenotype"

Look into the complete genome to identify variants that causes patient's disease: precision medicine

"Genotype - first"



3. DNA analysis/sequencing : "genotype"

Which indication do I choose when I prescribe ? (large query)



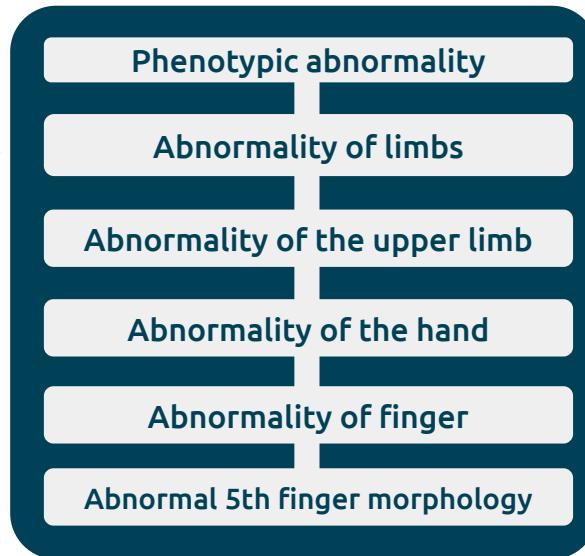
The need for computational phenotype analysis

Phenotyping :

Physicians identification of characteristics deviating from normal morphology, physiology, and behavior

Ontology :

Standardized symptom terms linked according to the human development architecture



Computational phenotype analysis :

Identification of diagnostic hypothesis, clinically relevant groups of patients, ...

A common language between human and machine is necessary for computer support (for now)



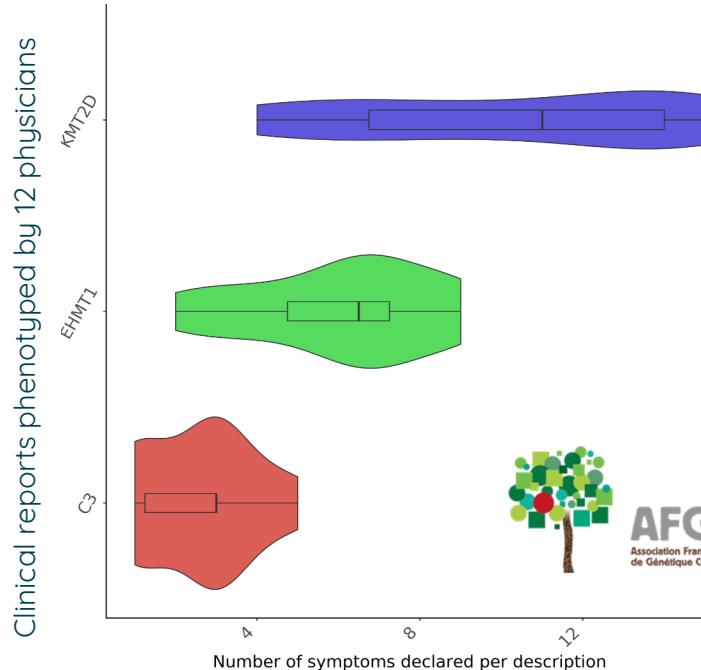


Very heterogeneous and sparse phenotyping practices...

- Phenotypes described in a cohort of **1686 patients**
- **47%** of HPO terms declared only once

Three clinical reports described by 12 physicians

**Physicians phenotyping diversity
explains the observed
heterogeneity**

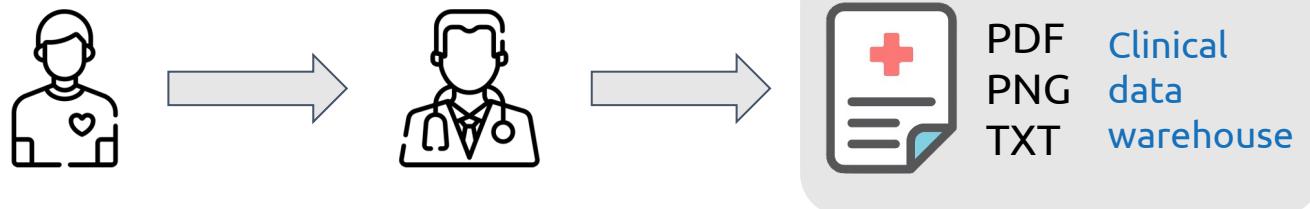


AFGC
Association Francophone
de Génétique Clinique



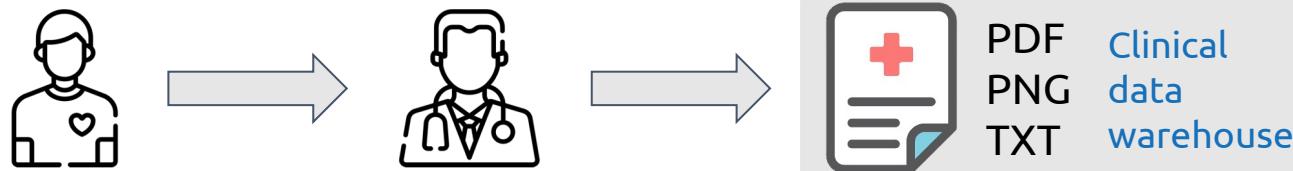


The clinical data pipeline for genomic analysis





The clinical data pipeline for genomic analysis



Extract data

Data access
Optical character recognition (OCR)

Transform data

Ontology terms extraction (HPO, OMOP...)

Exploit data

Ontology to variant prioritization and more...



The clinical data pipeline for genomic analysis



WORK IN PROGRESS

Extract data

Data access
Optical character
recognition (OCR)



Entreprise de Données de Santé du Languedoc



ClinFly

Transform data

Ontology terms
extraction (HPO,
OMOP...)



Exploit data

Ontology to variant
prioritization
and more...



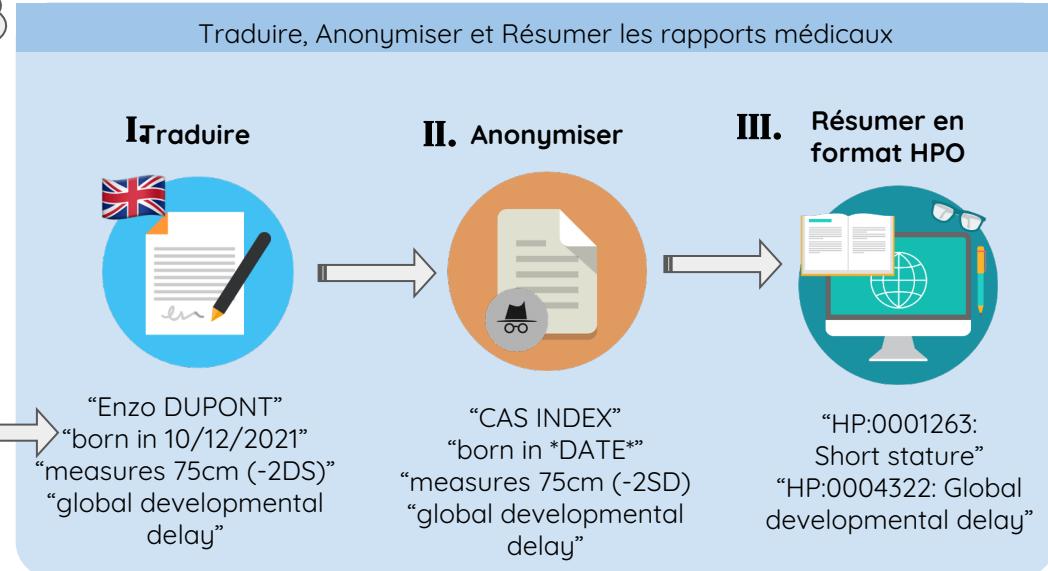


Challenges in data extraction and sharing

1. Tous les outils d'analyse phénotypique computationnelle et d'extractions de termes d'ontologie n'utilisent que l'anglais comme langue.
2. Il est très difficile d'obtenir des médecins des descriptions cliniques en termes de HPO
3. Difficile de partager des informations cliniques anonymes avec d'autres médecins dans le monde entier (GeneMatcher)



What we dream to improve clinical data extraction and sharing



Let's try !



Traduire, Anonymiser et Résumer les rapports médicaux

I. Traduire



II. Anonymiser



III. Résumer en format HPO



Développer les abréviations

Traduction avec OpenNMT

Correction des erreurs courantes

Une liste de termes PHI français à rendre anonymes

Une liste de termes cliniques à ne pas anonymiser

Anonymiser

Extraire les biométries

Flags des phrases qui ne concernent pas le cas d'index

Résumer

OpenNMT:

Open-Source Toolkit for Neural Machine Translation

microsoft/presidio

Context aware, pluggable and customizable data protection and de-identification SDK for text and Images



ClinPhen

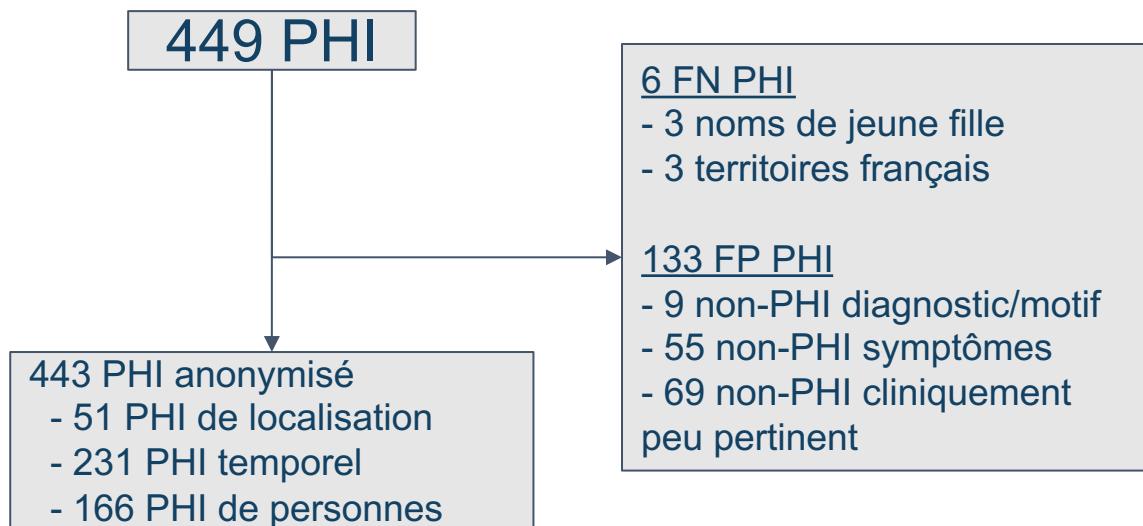


Résultats sur 50 courriers



1. ClinFly identifie les PHI et gère la protection des données de santé, atteignant :

- Recall rate of 99%
- Precision rate of 77%



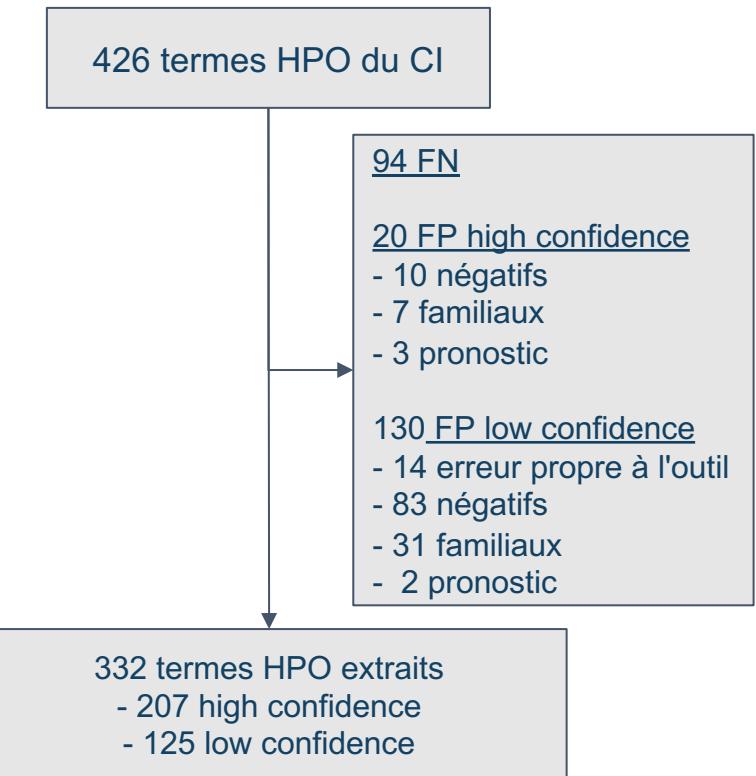
Lucas
Gauthier
CHU
Lyon



Résultats sur 50 courriers

2. Les performances de ClinFly pour le résumé d'un rapport médical par le format HPO :

- Precision of 92%
- Maintained an average of 4.5 HPO terms per report = 49% recall





Graphical interface and command line

Graphical User Interface

For single report usage with interactive analysis, ClinFly provides a web application accessible at <https://huggingface.co/spaces/kyauy/ClinFly>.

To run the Streamlit application on your local computer, activate the poetry shell and run the `clinfly_app_st.py` file:

```
poetry shell  
streamlit run clinfly_app_st.py
```

Command Line Interface

For processing multiple reports with offline options, use the command line interface provided by `clinfly_app_cli.py`.

The input should be a TSV .txt file structured as follows (see `data/test.tsv` for an example):

```
Report_id_1 Doe John Report text  
...  
Report_id_X Doe John Report text
```

Outputs will be placed in the `results` folder according to the file extension, using first three columns in filename.

- The deidentify report will be generated and placed in the `results/Reports` folder.
- Three HPO extraction outputs will be generated in `TSV`, `TXT` and `JSON` folders.

To run the CLI application on your local computer :

```
poetry shell  
<python running version> clinfly_app_cli.py --file <input txt file with the reports> --languag
```



Enody
Gernet
CHU
Montpellier



ClinFly



The clinical data pipeline for genomic analysis



WORK IN PROGRESS



Extract data

Data access
Optical character
recognition (OCR)

Transform data

Ontology terms
extraction (HPO,
OMOP...)



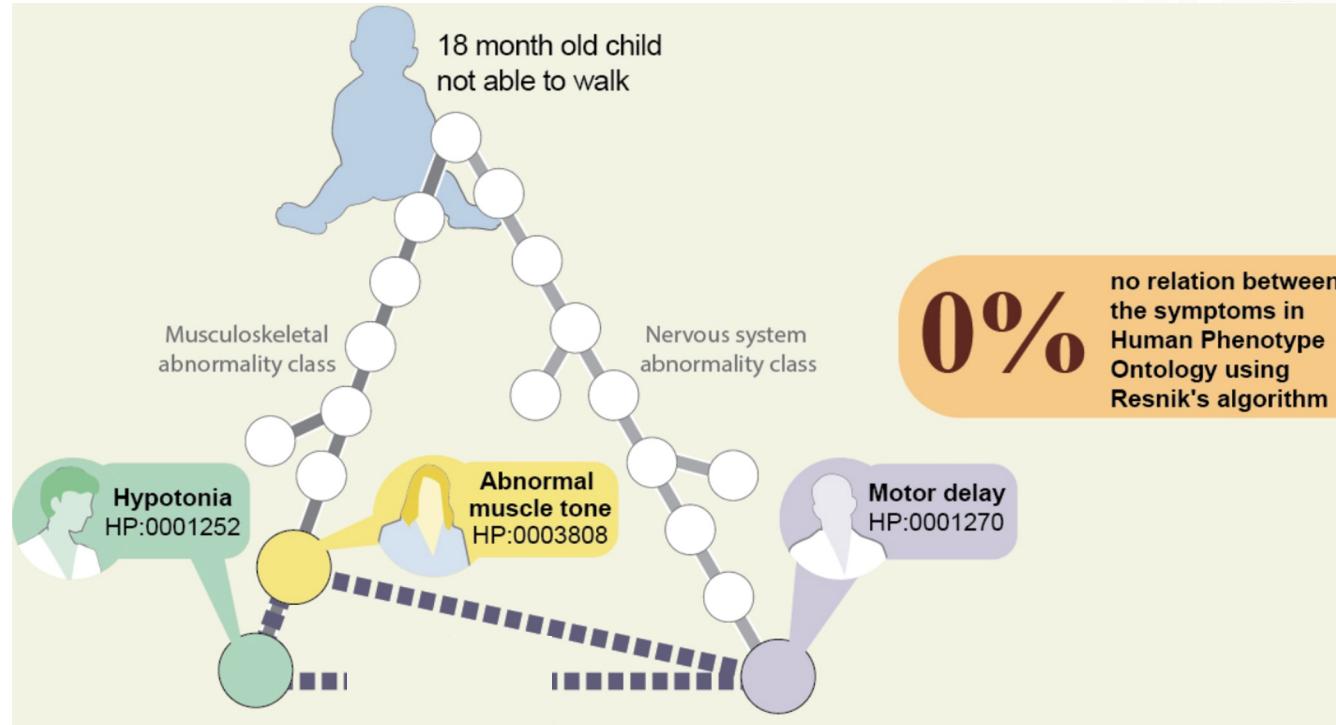
Exploit data

Ontology to variant
prioritization
and more...



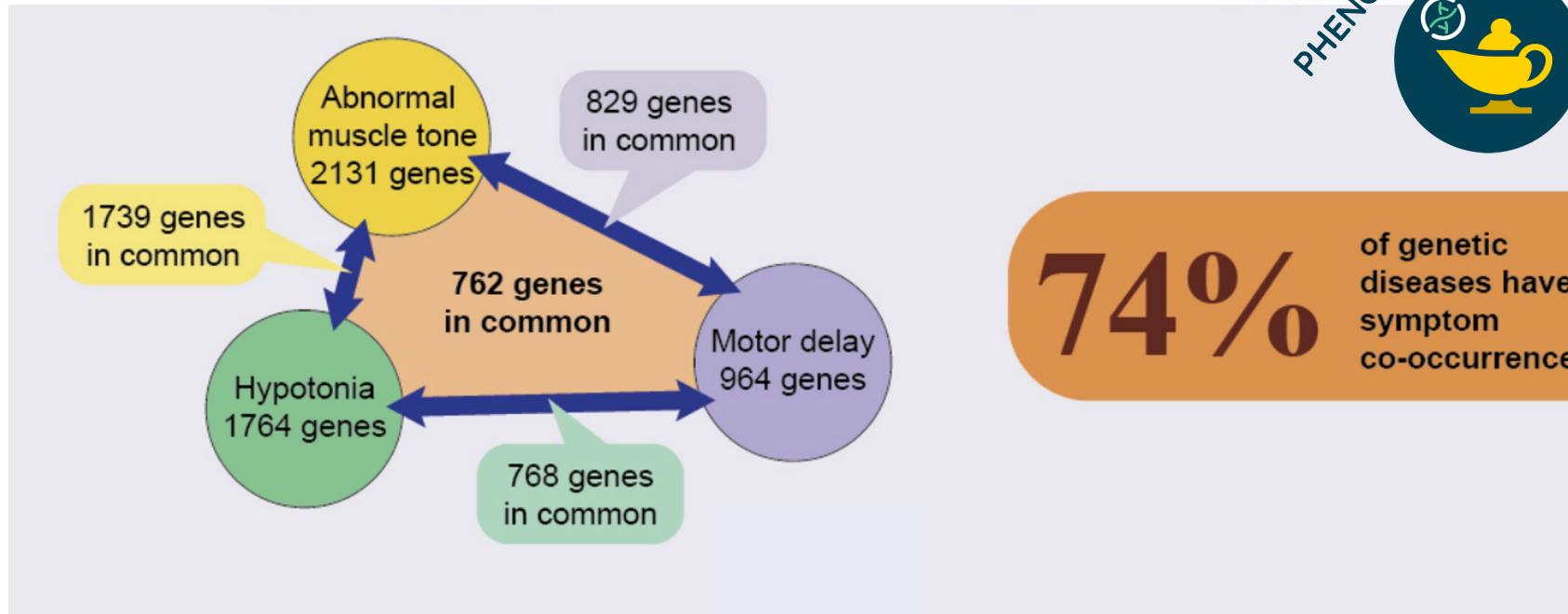


PROBLEM Heterogeneity of clinical descriptions between physicians, which cannot be captured by the HPO ontology

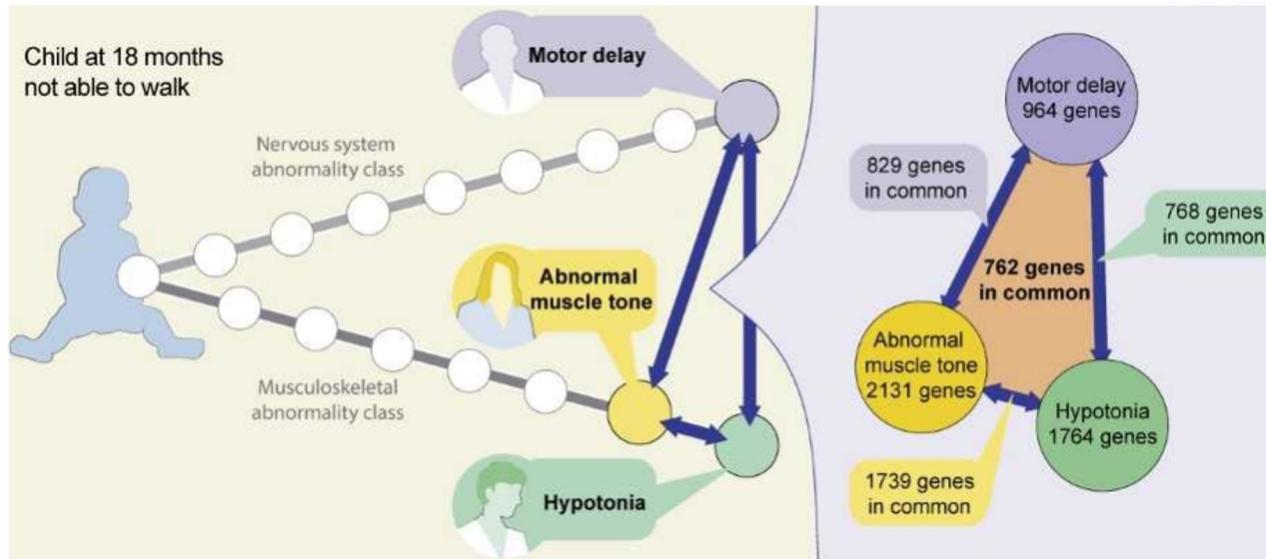




SOLUTION Learning phenotypic similarity using a knowledge graph based on gene-HPO associations



Multiple ways to describe patients



Physicians acquire intricate cognitive frameworks to solve diagnostic problems

→ Link between symptoms are very different from Human Phenotype Ontology structure

Inductive reasoning



Microcephaly
(HP:0000252)

Delayed ability to walk
(HP:0031936)



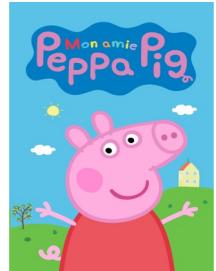
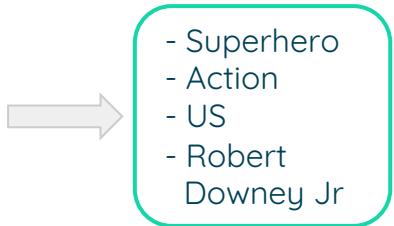
Disease /Phenotypic
patterns I know



Intellectual deficiency
(HP:0031936)



Inductive reasoning... through modeling

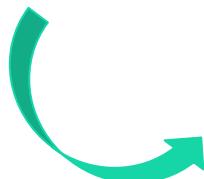


NETFLIX

Recommendation
algorithms are
really efficient

Apply inductive reasoning in symptom-gene associations

	Unstructured data			Structured databases			
Gene	Text-matching via OMIM	PubMed	NCBI	Merge all HPO-gene association available	The Jackson Laboratory	orphanet	HP:0000006 Probability
BRCA1	1	1	0	0	0	0	0.5



$$\begin{array}{ccccccc} & \text{HP01} & \text{HP02} & \text{HP03} & \dots & \dots & \dots & \text{HP15785} \\ \text{Gene 1} & 0.23 & 0.41 & 0 & \dots & \dots & \dots & 0 \\ \text{Gene 2} & 0 & 0.21 & 0.32 & \dots & \dots & \dots & 0.42 \\ \dots & \dots \\ \text{Gene 4531} & 0.11 & 0.27 & 0.42 & \dots & \dots & \dots & 0.42 \end{array}$$

Building a symptom-gene association matrix with ~16000 symptoms



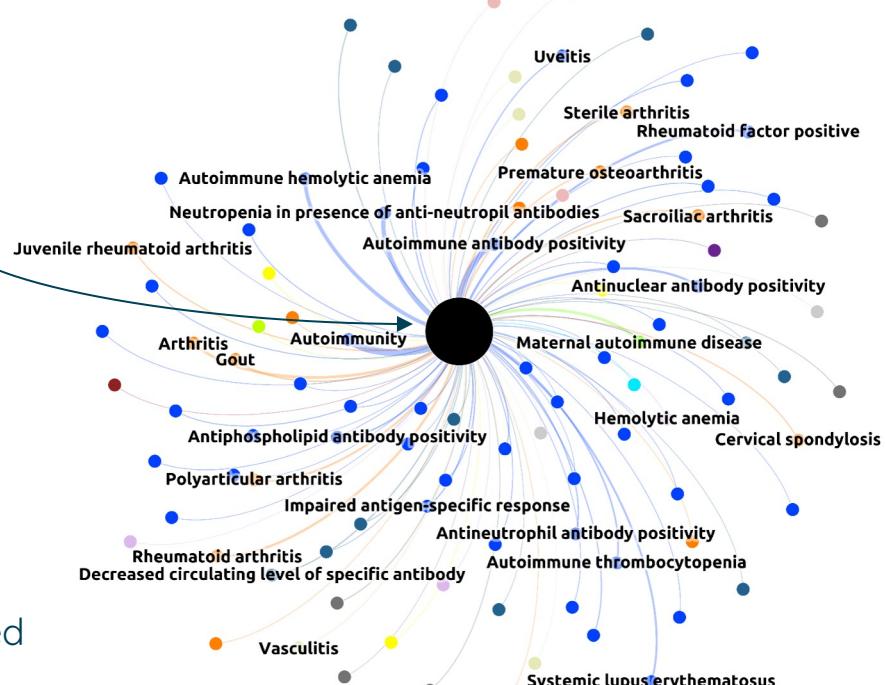
Symptom-gene graph

Retrieving Symptom-gene associations with graphs

- **390 groups** of symptoms
(n=43,308, 10% of symptom-group associations)
- 5,971,755 pairs of symptoms
- **3,222,053 additional** NMF-based symptom-gene associations
- “only” 2% cohort symptom-gene association missing !

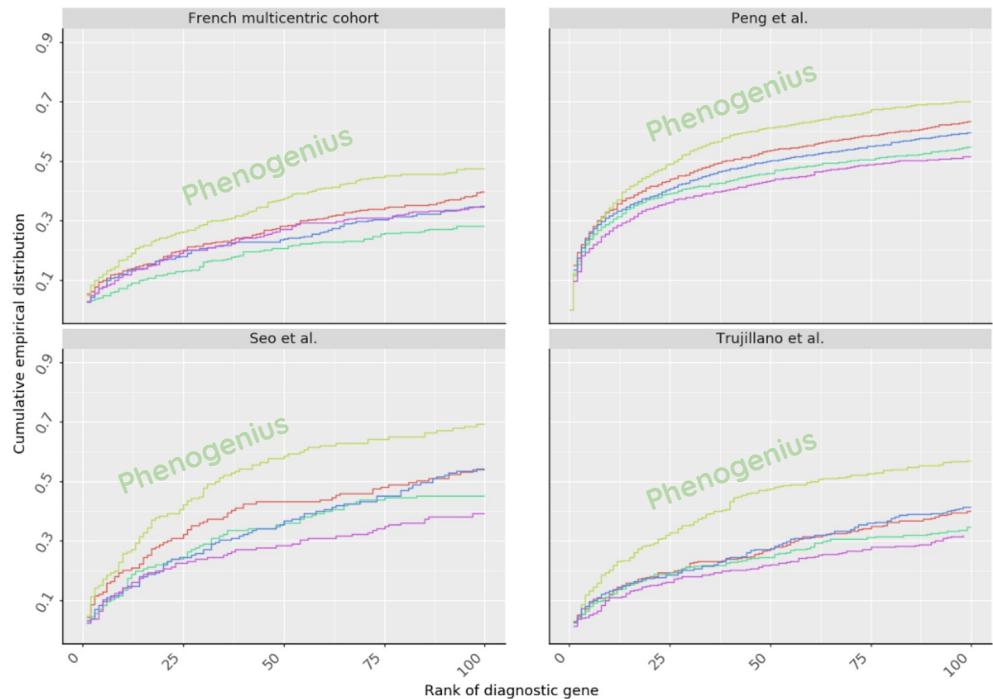
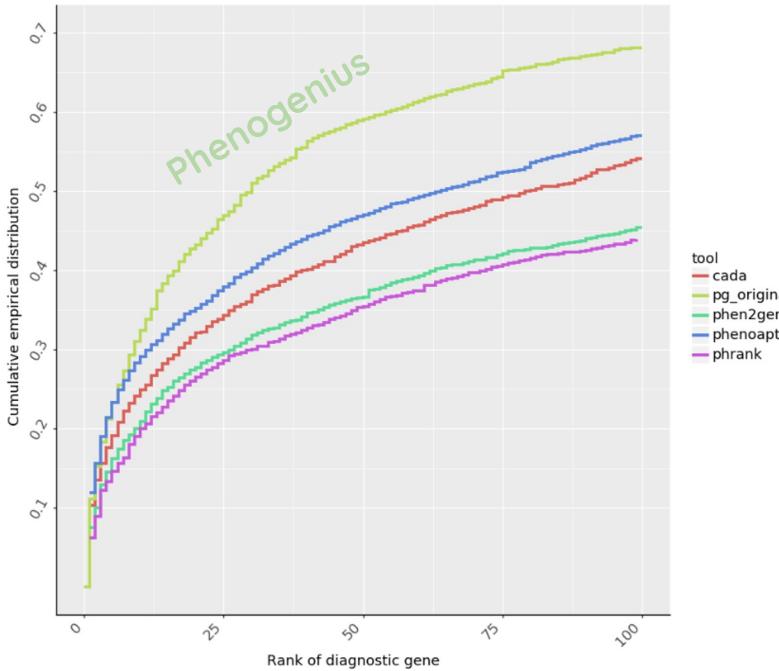
An example: Autoimmunity group

- Main symptom :
Autoimmunity
- 99 HPOs included
- 14 different HPO classes
- 545 genes associated to the group



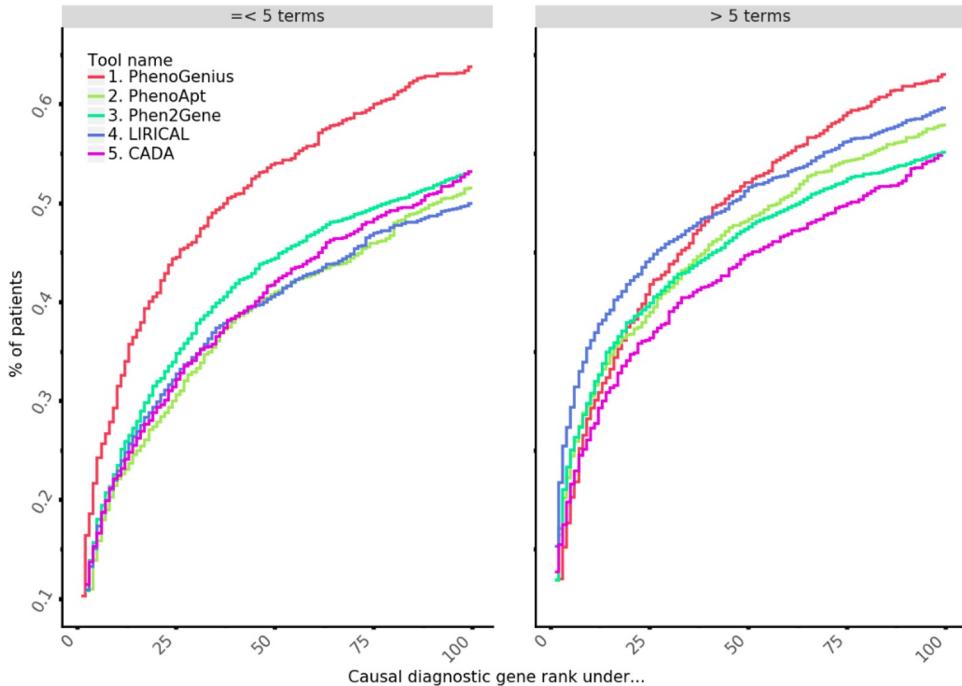
Performance on 4 cohorts (1686 patients)

Outperforming 2022 state-of-the-art ranking methods



Performance on 4 cohorts (1686 patients)

A way better semantic similarity using disease and not ontology architecture



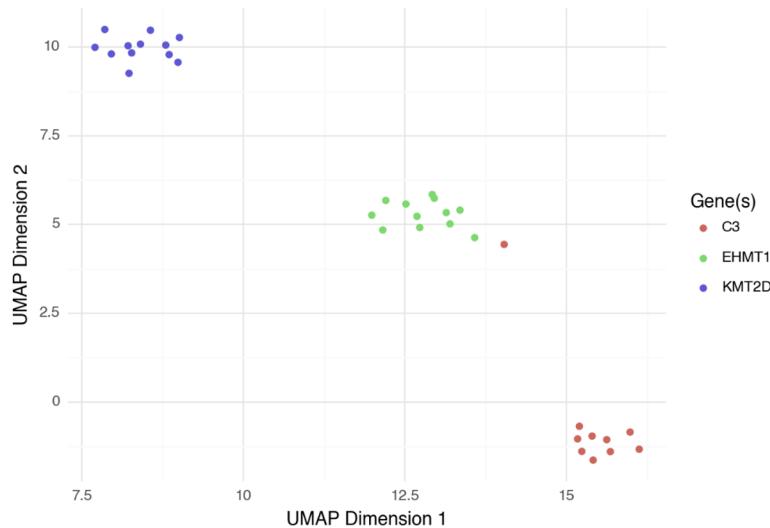
Median diagnostic rank = < 5 terms :
PhenoGenius : 37
Phen2Gene (best competitor): 78

% of cohort with <= 5 terms :
50% (853/1686)

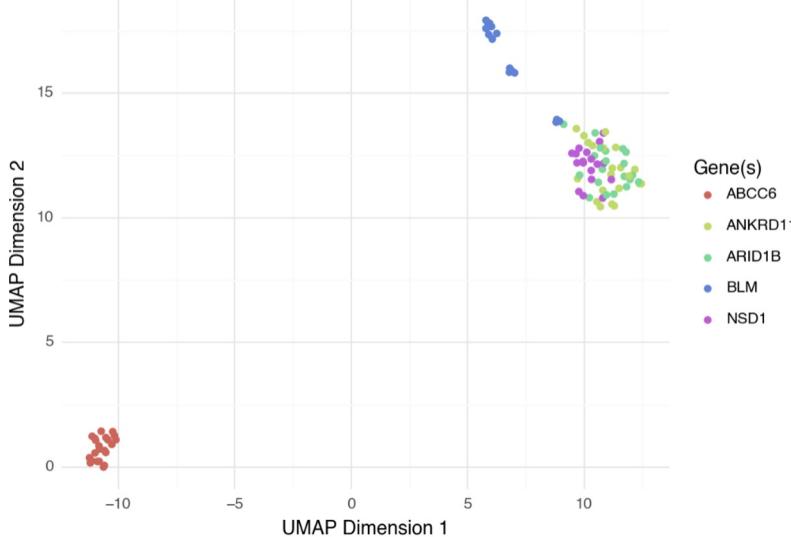


Use phenotypes as phenomics

Three clinical reports described by 12 physicians

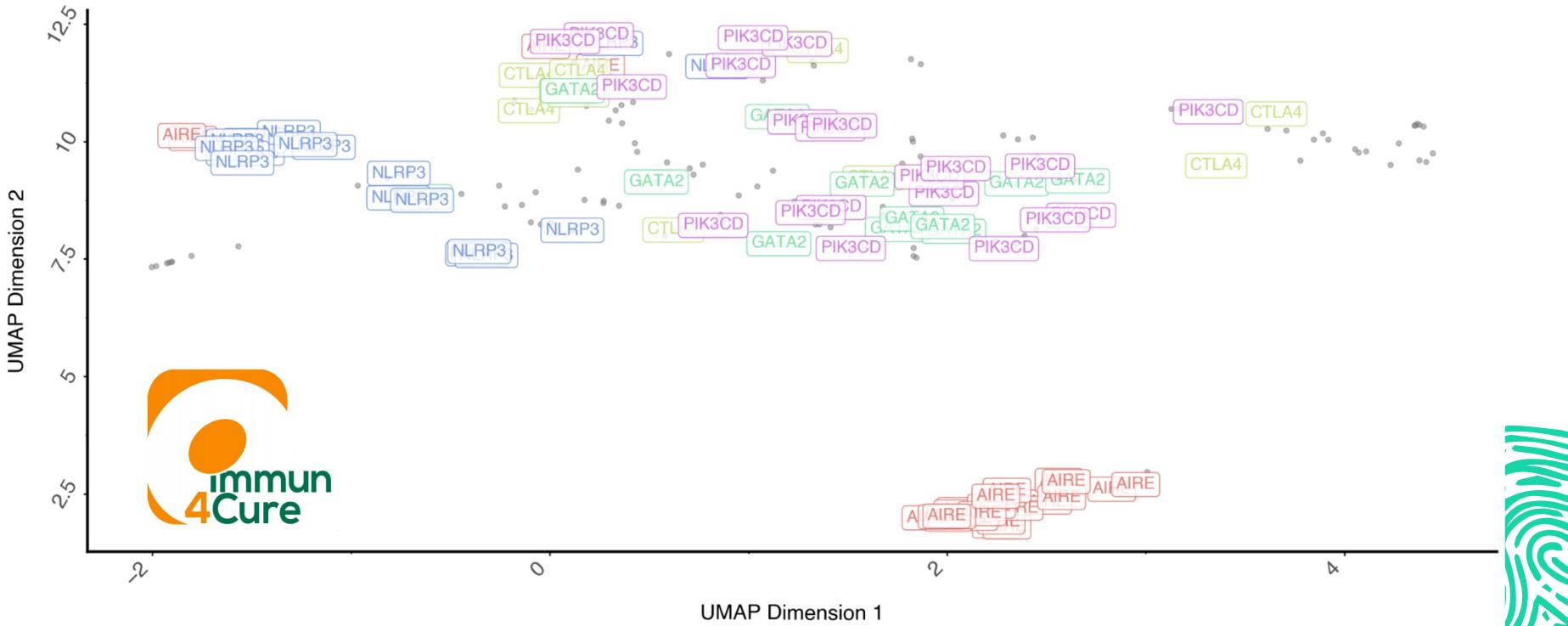


Cohort : 5 most frequent diagnostic genes



-> Vectorial projection available for all patients

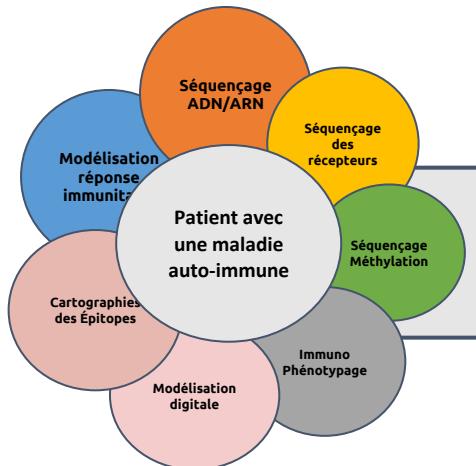
I Cohort phenomics projection



Clinical data as Phen-Omics, an omic type among others A 5 to 10 years vision ?



Integrative immunology



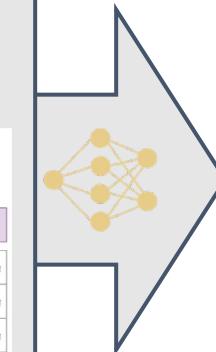
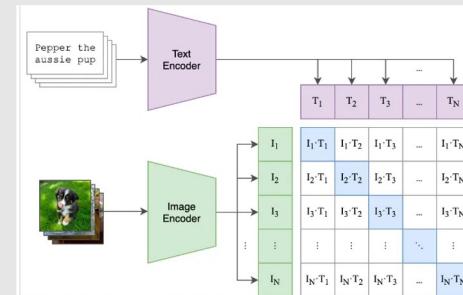
Multi-omics analysis using structured data



Generative AI to structure clinical data

Multimodal embeddings into vector database

Omics data wrangling to structured text or imaging



Inference & clinical applications

Chat Agents using LLM

Patient stratification



Thanks for your attention!

 kyauy/ClinFly

kyauy/PhenoGenius

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★ CHU Grenoble Alpes