### Summary

Machine Learning and Data Challenges
 Aymeric Floyrac & Maximilian Oppelt

- Data Challenge SFP-SFD : Visio-Mel Search for a digital signature evaluating the risk of metastatic <u>evolution of localized melanomas</u> Dr Frédéric Staroz
- Data Challenge Hôpital Foch : DigiLut Detection of graft rejection following lung transplantation Dr Antoine Roux
- Data Challenge SFCTCV : Benefits of Al in thoracic oncology research: about the C PLu NET challenge
   Romain Vergé
- <u>Sage Bionetworks Organizing DREAM Challenges for Engaged Health Research</u> <u>Thomas Schaffter, PhD</u>
- Data Challenge SFMN : Dat-Hub Parkinson and dopamine transporter imaging: a data challenge from the Health Data Hub and the French Society of Nuclear Medicine
   Pr Eric Guedj
- Data Challenge HUS : D-IA-GNO-DENT HDH Data Challenge Rare oral diseases / Artificial <u>intelligence / Diagnostic assistance</u> Pr Agnès Bloch Zupan
- Data Challenge SFA : Allergen Chip Challenge: artificial intelligence at the service of the allergic patient

Pr Joana Vitte

• Open science: sharing and reusing research data Dr Lorien Benda & Isabelle Blanc





## WELCOME TO THE DATA CHALLENGES FOR HEALTH EVENT !

### 2:15 - 6:00













## Data Challenges for Health - Introduction



#### **David Sainati**

eHealth projects director at the Digital Health Ministerial Delegation

David Sainati has a PhD in pharmacy and a graduate of HEC Paris. He founded and managed a digital health start-up for 8 years. With this entrepreneurial experience, he joined the Ministerial Delegation for eHealth (DNS) in 2019 as a project manager and he is in charge of national projects related to entrepreneurship and compliance.



#### Pr Hossein Khonsari

Data Challenges

Medical director of the Health Data Hub

Prof. Roman Hossein Khonsari is Medical Director of the Health Data Hub. Professor of Plastic and Maxillofacial Surgery at the University of Paris and Maxillofacial Surgeon at the Necker - Enfants Malades Hospital, trained in Nantes and Paris, he also holds a master's degree in biology from the Ecole Normale Supérieure de Paris and a PhD in craniofacial development from King's College London.

2:20 - 2:30













### Data Challenges for Health - Introduction



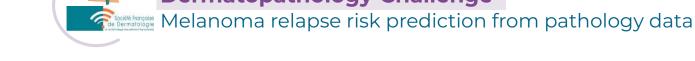
<sup>es</sup>Allergies# SFA

Société Rancaise d'Allers

#### **Thoracic Surgery Challenge**

Classification of lung tumours from cellular and clinical imaging data

**Dermatopathology Challenge** 



**Allergology Challenge** Machine learning on biological data from allergology



#### Lung Transplant Challenge

Development of an algorithm capable of detecting graft rejection following transplantation from cellular imaging data



#### **Rare Dental Diseases Challenge**

L'INNO

TION

Detection of rare oral and dental diseases from photographs of teeth



#### **Nuclear Medicine Challenge**

bpifrance

Diagnosis of Parkinson's syndromes from brain imaging data













## **Machine Learning and Data Challenges**

Data Challenges for Health



#### Aymeric Floyrac

Data Scientist at Health Data Hub

INSEE administrator, Aymeric Floyrac has been working as a data scientist at Health Data Hub for nearly two years. He works with HDH pilot projects on machine learning pipelines and on a variety of topics including synthetic data generation and data pseudonymization.

Aymeric Floyrac graduated from Ecole polytechnique in applied mathematics in 2020.



#### Maximilian Oppelt

Research Scientist at the Fraunhofer Institute for Integrated Circuits

Maximilian P. Oppelt is a Research Scientist, at the Fraunhofer Institute for Integrated Circuits and a PhD Candidate at the Machine Learning and Data Analytics Lab of the Friedrich Alexander University. After working for medical device manufacturers, he joined the academic world again and is now working on novel end-to-end deep learning methods to extract stable representations of biosignals. He participated in past data challenges himself and is now supervising students and young researchers in upcoming competitions.

2:30 - 2:45











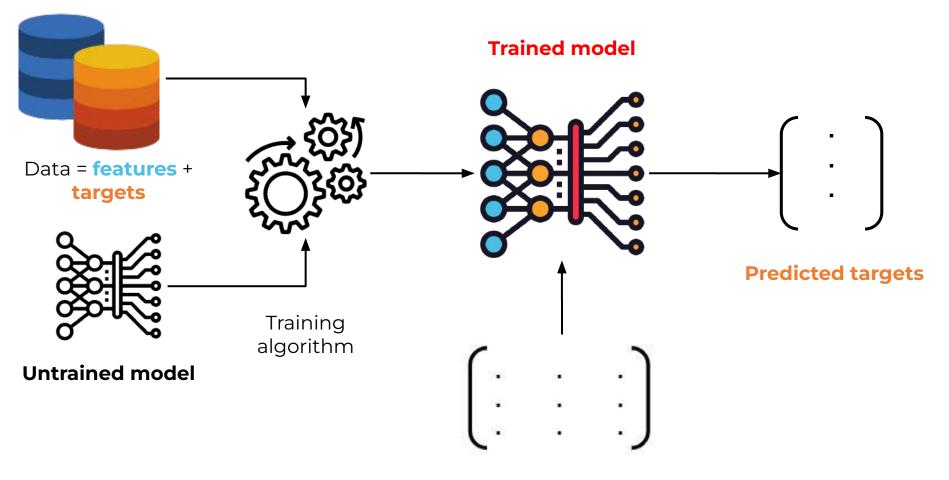




### Machine learning Health Data Hub

24th May 2022

# Machine learning: what and what for?



**New features** 

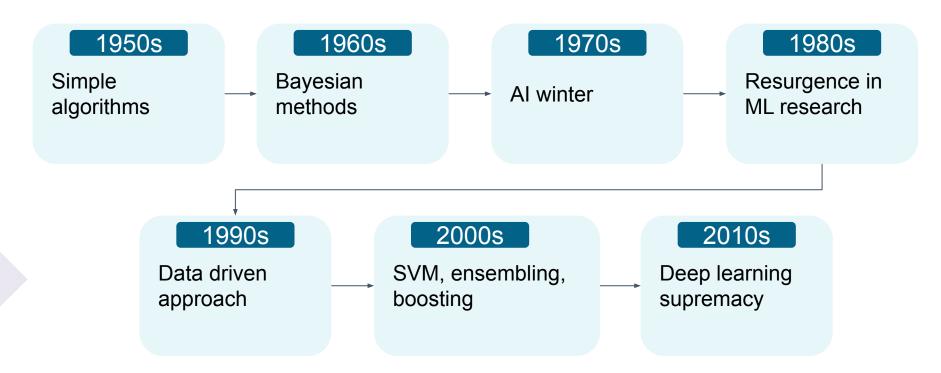


# A little bit of history

"Learning is any process by which a system improves from experience."

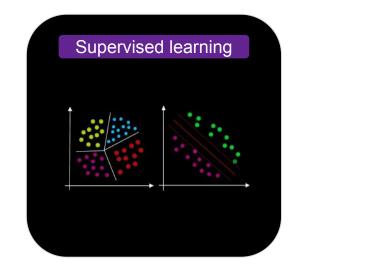
Herbert Simon

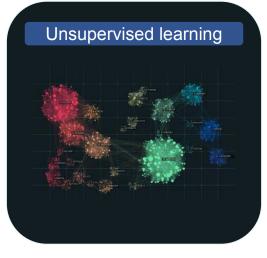






# A little bit of theory







#### Some popular applications

- Natural Language Processing
- Computer Vision
- Speech Recognition

- Recommendation systems
- ✤ Generative models
- Trends prediction



# Machine learning in action



#### Problem

A chronic disease: heart failure A downward spiral of hospital stays Most hospital stays could be prevented



#### Solution

Use remote monitored pacemakers Deploy an alerting system for cardiologists

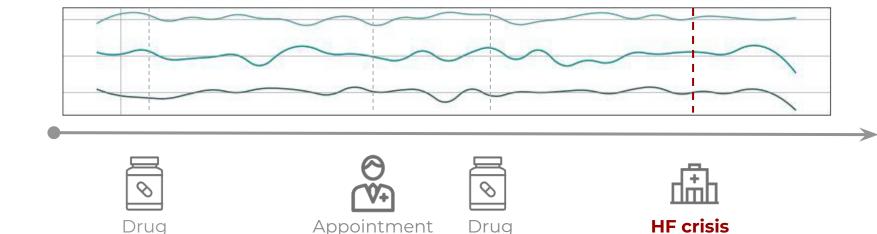


# **Machine learning in action**

Data and model training

Connected medical device







Claims (national health insurance) Drug

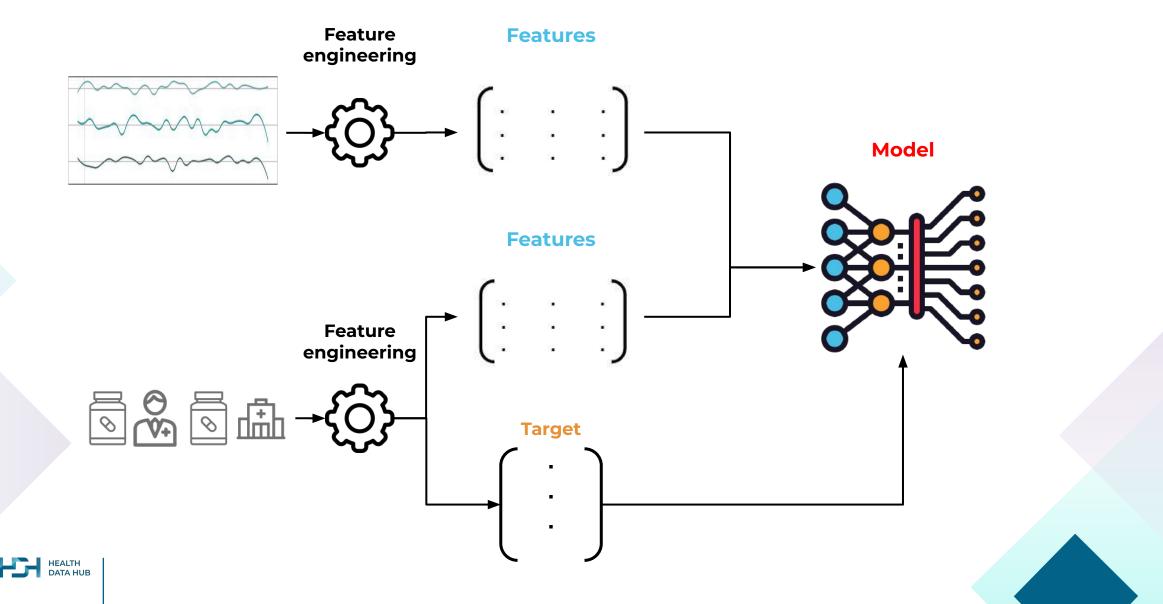
Appointment

**HF crisis** 



# Machine learning in action

Data and model training



# Thank you





Fraunhofer Institute for Integrated Circuits IIS



Friedrich-Alexander-Universität Erlangen-Nürnberg



Maximilian P. Oppelt

# Data Challenges for Health A Data Scientist Perspective

#### The Data Scientist Perspective Data Challenges Can Be Life Changing For Engineers

#### Data Challenge: My Journey



**B.Eng Mechatronics** 

Friedrich-Alexander-Univers Erlangen-Nürnberg

M.Sc. Medical Engineering

Image Quality Engineer

Image Processing Software

GE Healthcare

Fraunhofer



& DATA ANALYTICS

Research Scientist Physionet 2020 Computers in Cardiology Challenge on ECG Arrhythmia Detection

PhD Candidate

Engineer

#### Institute for Integrated Circuits & MaD Lab



- Biomedical Signal Analysis
- Software Engineering and Embedded Design
- Clinical Studies
- Machine and Deep Learning







#### Motivation for Participation in Data Challenges Key Messages



#### Combining Scatter Transform and Deep Neural Networks for Multilabel Electrocardiogram Signal Classification

Maximilian P Oppelt<sup>1</sup>, Maximilian Riehl<sup>1</sup>, Felix P Kemeth<sup>2</sup>, Jan Steffan<sup>1</sup> <sup>1</sup>Department of Image Processing and Medical Engineering, Fraunhofer IIS <sup>2</sup>Department of Chemical and Biomolecular Engineering, Whiting School of Engineering, Johns Hopkins University









#### Recipes For Designing a Great Data Challenge Insights from A Challenger

Building a portfolio of projects! Win price money! Learn new techniques and explore new domains! Benchmark your skill. Participants can help refine the goal and discuss the metric. This increases identification with the challenge and motivate competitors on the long run and encourage data scientists from diverse backgrounds.



Build A Community: Google Groups, Discord, Live Events or Forums. This helps competitors too keep motivated in phases of slow progress.



Choose your metric wisely: Classic machine learning metrics such as accuracy might not lead to most desirable outcome and choose your validation/test-set wisely!



Keep the pedagogical motivation of competitors in mind: Young researchers will compete to learn!







#### Participating in Future Challenges Skills and Tools and Requirements

Tips for Participants

- Read the rules and report flaws.
- Make sure the challenge topic is a good fit.
- Network and communicate with other teams.
- Start simple and submit continuously.
- Try to gather domain knowledge and work with the data.
- Learn from similar challenges and publications.
- Track you experiments.
- Have Fun!

#### **Participation in Future Competitions**

- Compete with students and new team members for training and faster onboarding.





Fraunhofer Institute for Integrated Circuits IIS

# Have Fun!

# Data Challenge SFP-SFD : Visio-Mel - Search for a digital signature evaluating the risk of metastatic evolution of localized melanomas





#### Dr Frédéric Staroz

Surgical pathologist at OuestPathologie, Quimper, France

Frédéric Staroz is a dermatopathologist in his practice. He is president of the national professional council in pathology, involved in medical training and innovation within his specialty. He is also a member of the French Society of Pathology and as such coordinates the VisioMel DataChallenge supported by the HDH and the BPI. VisioMel aims to develop a digital fingerprint predictive of recurrence in melanoma.













#### 2:45 – 2:55



Search for a digital signature evaluating the risk of metastatic evolution of localized melanomas

Dr Data



RIC•Me

(DPath

FRANÇAISE DE PATHOLOGIE

SOCIÉTÉ

### - Dr F. Staroz -

Pathologist at Ouest Pathologie, Quimper site.

**Member** of the board of directors of the **French Society of Pathology**, and of the **VisioMel** association.

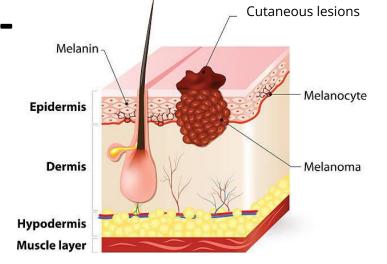
**President** of the **National Professional Council of Pathologists** (CNPath).











#### Melanoma:

Multifactorial disease of melanocytes

Current public health concern

#### Origin:

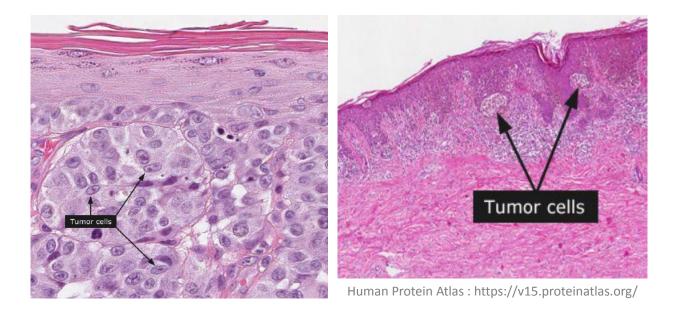
 Mainly depends on the interaction between UV exposure (period and intensity), host factors (presence of atypical nevi, high number of nevi, skin phototype) and genetic factors

#### **Statistics:**

- 3rd most frequent skin cancer but the one of **most serious**
- Increasing incidence in white population

#### **Treatment:**

- Surgical resection of primary melanomas
- Targeted therapy, immunotherapy : improve the outcome of patients with advanced melanoma but severe adverse events can occur



#### Importance to target patients with increased risk of metastatic evolution

# - So far -



The risk of **metastastic evolution** is currently **estimated** by anatomopathologists through **histological analysis** of **microscopic lesions** and caracterisation of **cancer stage** 

BUT

<u>Main purpose:</u> There is a high variability in the evolution of early stages melanomas (from 0 to IIC)

□ **Lack of reliability** in the prognosis

Secundary purpose: Targeted therapy is only active in

patients with BRAF mutated melanoma

Artificial intelligence could predict which patient is going to relapse

Artificial intelligence could predict if BRAF is mutated

# - Aim of this data challenge -

In order to complete the clinical and pathological prognosis, challengers aim to build an algorithm able to:

- 1. Identify a **digital fingerprint** that predict the **metastatic evolution** of patients within **5 years following initial diagnosis**
- 2. Predict the **mutational status** of the tumor for targeted therapy



# Expected available anonymised data: ≈ 3000 patients with primary melanomas -

• Clinical data: from RIC-Mel database:



Network of 49 French centers: - 43 dermatological centers

- 6 cancer centers

Since the creation in 2012, ≈ 40 000 patients have been included

 -Age, sex, medical history, melanoma clinical and pathological characteristics, 5 years outcome

# Expected available anonymized data: ≈ 3000 patients with primary melanoma -



#### • Clinical data:

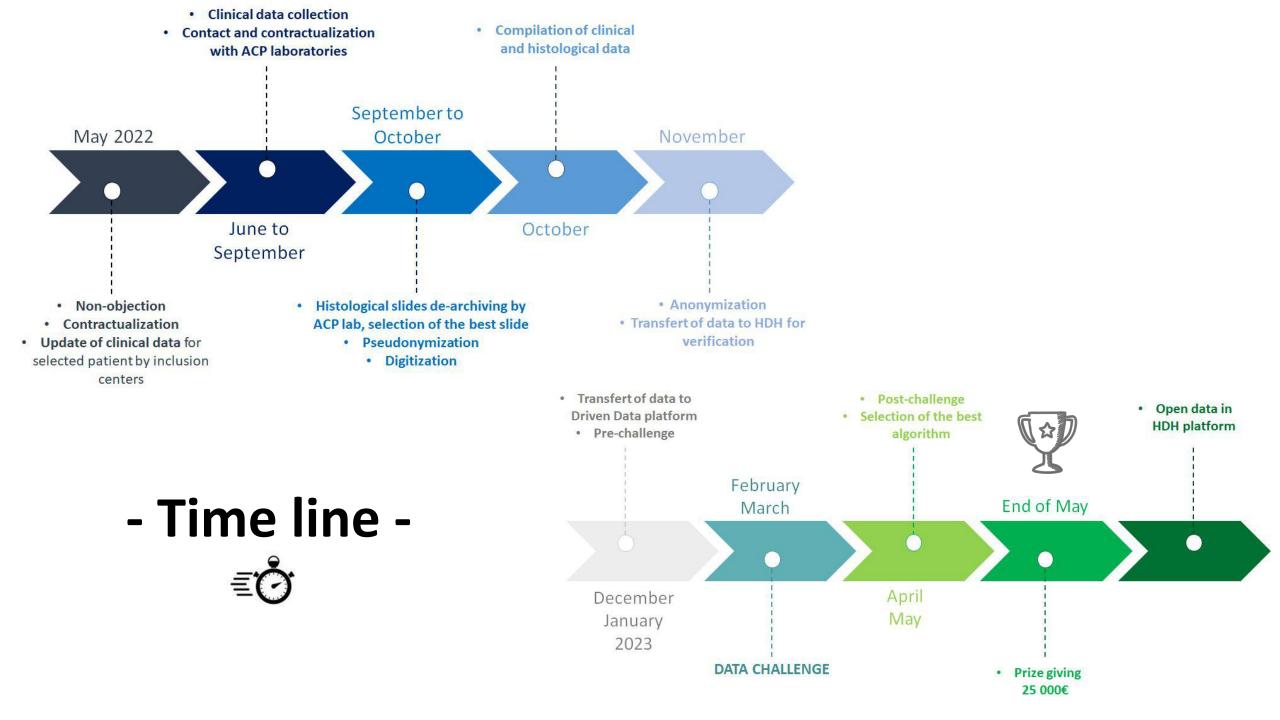
Age, sex, medical history, melanoma clinical and pathological characteristics, 5 years outcome

#### ACP data:

 Cancer stage
 (AJCC, from stage 0 to stage IIC), virtual histological slides.

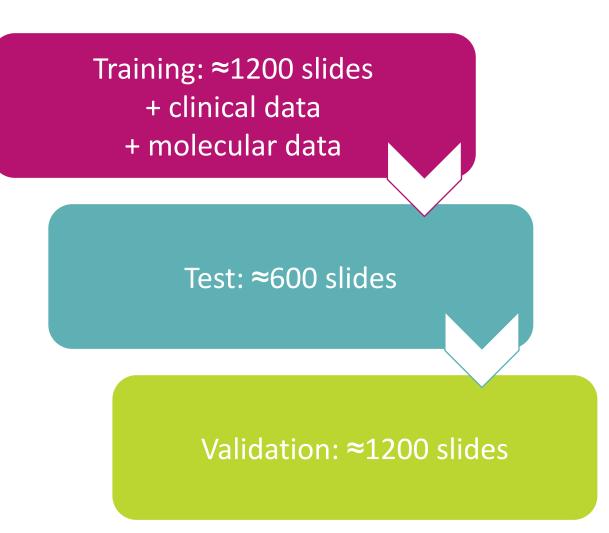
#### • Molecular data:

# - Presence/absence of **BRAF mutation**



- The data challenge -





- The data will be available on the driven data plateform that is hosting this challenge, for a seven weeks period
- The best challengers will be rewarded for their contributions (25 000€)
- Data will be exported to the Health Data Hub and will remain available in open source

# Data Challenge Hôpital Foch : DigiLut - Detection of graft rejection following lung transplantation





#### **Dr Antoine Roux**

(MD, PhD) MCU-PH UMR 0892 VIM, INRAE/U. of Versailles St-Quentin-en-Yvelines

Antoine Roux is a pneumologist and has a PhD in immunology, he is interested in the humoral response to lung transplantation as a post-doctoral fellow at the University of California (UCLA). His current research focuses on the clinical, histological and immunological aspects of rejection in order to develop targeted therapeutic strategies.

2:55 – 3:05

GOUVERNEMENT Liberté Égalité Fraternité





**bp**ifrance







# DigiLUT

Dr Antoine Roux MCU PH Hôpital Foch UVSQ- Paris Saclay









# **LUNG Transplantation**

#### Only curative treatment for end stage Respiratory disease in selected patient

- □ Remove 2 unfunctional Lung
- □ Implant 2 functional Lung

4000 LTx/year worldwide 400 LTx/year in France

Main indications (>80%)

- □ COPD
- □ Pulmonary Fibrosis
- □ Cystic Fibrosis

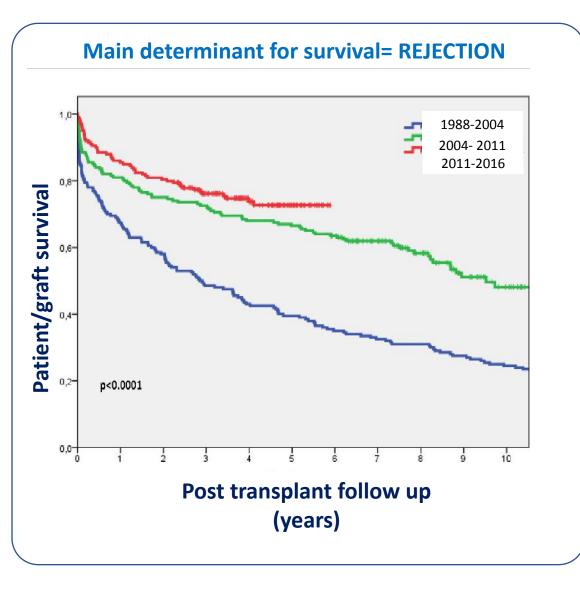
High mortality rate without Transplantation







## **LUNG Transplantation Challenge**



#### Rejection

- Heterogenous process
- Cellular Rejection= A lesion
- Graded 0 to 4

#### **TREATMENT for saving Graft**

## Histological definition **POOR Reproducibility**

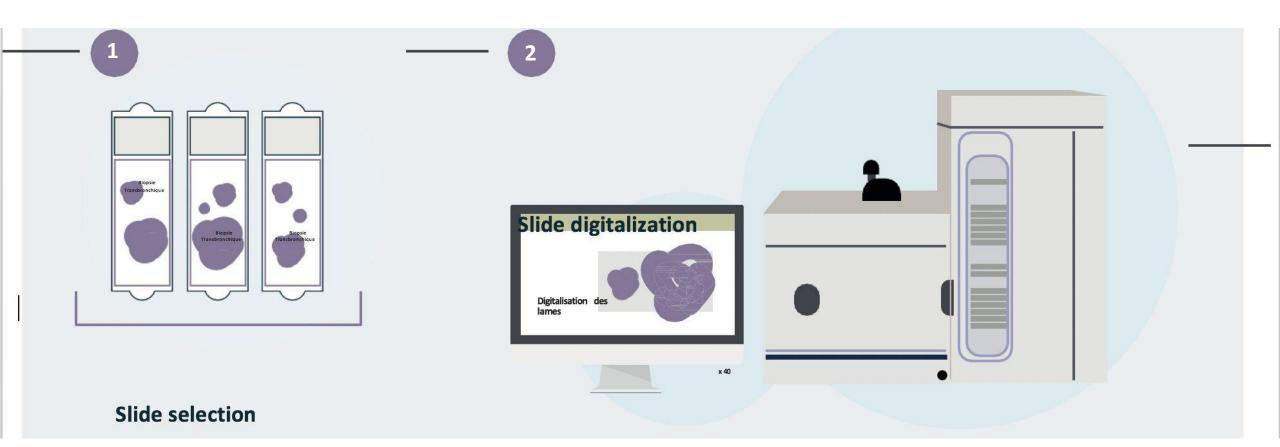
| Variable                   | ĸ Values<br>Median (range) |
|----------------------------|----------------------------|
| Biopsy adequacy            | 0.28 (-0.03 to 0.42)       |
| Acute cellular rejection   | 0.40 (0.24-0.62)           |
| Airway inflammation        | 0.23 (0.11-0.56)           |
| Obliterative bronchiolitis | 0.18 (0.04-0.58)           |
| Acute lung injury DAD      | 0.20 (0.03-0.63)           |
| Endotheliitis              | 0.22 (-0.04 to 0.47)       |
| Alveolar hemosiderosis     | 0.40 (0.19-0.62)           |
| Capillary inflammation     | 0.17 (0.03-0.31)           |
| Suspicion for aspiration   | 0.14 (-0.02 to 0.66)       |
| C4d                        | 0.40 (0.24-0.78)           |

C4d, complement 4d; DAD, diffuse alveolar damage

Banff study of pathologic changes in lung allograft biopsy specimens with donor-specific antibodies

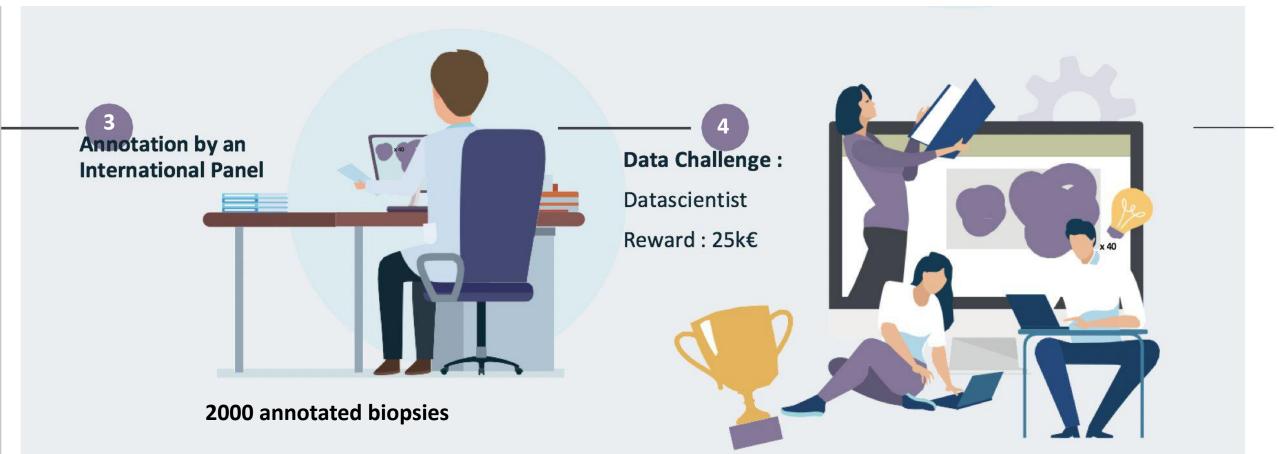
William Dean Wallace, MD.<sup>a</sup> Ning Li, PhD.<sup>b</sup> (Laus B. Andersen, DMSc.<sup>c</sup> A. Valeria Arrossi, MD.<sup>d</sup> Medhat Askar, MD, PhD.<sup>d</sup> Gerry J. Berry, MD.<sup>a</sup> Matthew M. DeNicola, MD.<sup>a</sup> Desley A. Neil, MBBS, PhD, FR(Path, Elizabeth N. Pavlisko, MD.<sup>b</sup> Elaine F. Reed, PhD.<sup>a</sup> Myriam Remmelink, MD.<sup>b</sup> S. Sam Weigt, MD.<sup>b</sup> Birgit Weynand, MD.<sup>J</sup> Jennifer Q. Zhang, PhD.<sup>a</sup> Marie M. Budev, DD.<sup>b</sup> and Carol F. Farver, MD<sup>a</sup>

# IMPROVE Acute rejection diagnosis DigiLUT



# **IMPROVE Acute rejection diagnosis**

# DigilUT



# **IMPROVE Acute rejection diagnosis**

# -DigilUT





### <u>Challenge for the data scientist community:</u>

Help us to improve Rejection detection and Graft outcome!!!





### Data Challenge SFCTCV : Benefits of AI in thoracic oncology research: about the C PLu NET challenge





#### **Romain Vergé**

Resident in thoracic surgery at the Toulouse University Hospital, member of the French Society of Thoracic and Cardiovascular Surgery

### 3:05 – 3:15

Romain Vergé is a resident in Thoracic Surgery with a special interest in oncology. He is particularly involved in lung neuroendocrine tumors and is currently conducting a national study with the French Society of Thoracic and Cardiovascular Surgery using the French national database Epithor and classical statistical approaches.

The use of AI appears to him as an additional way to increase knowledge about these rare tumors.













Benefits of AI in thoracic oncology research: about the C PLu NET challenge : Classification and Prognosis of LUng NeuroEndocrine Tumors







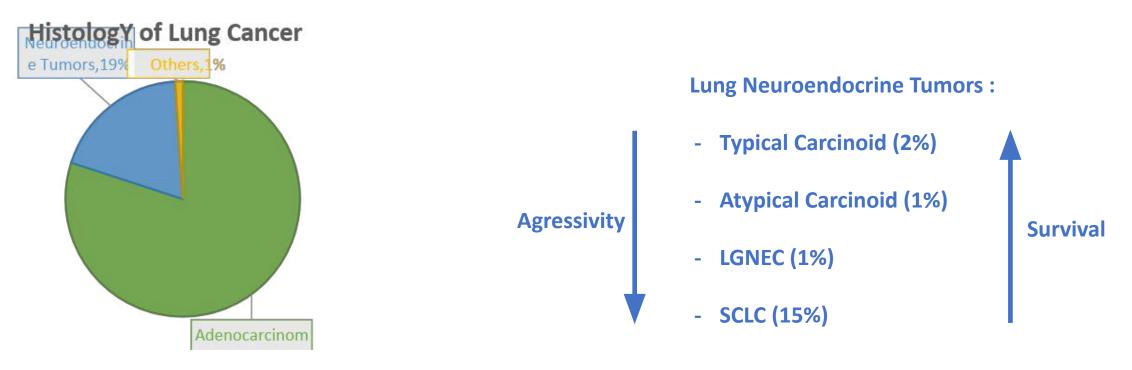




## Introduction Lung Cancer or Lung Cancers ?



- Leading cause of cancer deaths worldwide
- Heterogeneity of lung cancer



## **Scientific Problem**



### • Classification = Predictive of survival + Necessary for management

| Descriptor  | Definition   |
|-------------|--|
| T descripto |  |
| тх          | Primary tumor cannot be assessed or tumor proven by the presence of malignant cells in spu-<br>tum or bronchial washings but not visualized with imaging or bronchoscopy   |
| T0          | No evidence of primary tumor   |
| Tis         | Carcinoma in situ  |
| T1          | Tumor ≤ 3 cm in greatest dimension, surrounded by lung or visceral pleura, without broncho-<br>scopic evidence of invasion more proximal than the lobar bronchus   |
| Tla         | Tumor $\leq 1$ cm in greatest dimension  |
| T1b         | Tumor > 1 cm but ≤ 2 cm in greatest dimension  |
| Tlc         | Tumor > 2 cm but ≤ 3 cm in greatest dimension  |
| T2 descript | or   |
| T2          | Tumor > 3 cm but ≤ 5 cm or tumor with any of the following features: involvement of a main<br>bronchus regardless of the distance from the carina; invasion of the visceral pleura; associated<br>with partial or complete lung atelectasis or pneumonitis                                     |
| T2a         | Tumor > 3 cm but $\leq$ 4 cm in greatest dimension   |
| T2b         | Tumor > 4 cm but $\leq$ 5 cm in greatest dimension   |
| T3          | Tumor > 5 cm but ≤ 7 cm in greatest dimension or one that directly invades any of the follow-<br>ing structures: parietal pleura, chest wall (including superior sulcus tumors), phrenic nerve,<br>parietal pericardium; or separate tumor nodule or nodules in the same lobe                  |
| T4          | Tumor measuring >7 cm in greatest dimension that invades any of the following structures: medi-<br>astinum, diaphragm, heart, great vessels, trachea, recurrent laryngeal nerve, esophagus, vertebra<br>body, carina; or separate tumor nodule or nodules in a different lobe of the same lung |
| N descripto | r  |
| NX          | Regional lymph nodes cannot be assessed  |
| N0          | No regional lymph node metastasis  |
| NI          | Metastasis in ipsilateral peribronchial and/or ipsilateral hilar lymph nodes and intrapulmonary<br>nodes, including involvement by direct extension  |
| N2          | Metastasis in ipsilateral mediastinal and/or subcarinal lymph nodes  |
| N3          | Metastasis in contralateral mediastinal, contralateral hilar, ipsilateral or contralateral scalene, or<br>supraclavicular lymph nodes  |
| M descripto | n -  |
| M0          | No distant metastasis  |
| M1          | Distant metastasis   |
| MIa         | Separate tumor nodule or nodules in contralateral lung; malignant pleural effusion or pleural<br>thickening or nodules or masses; malignant pericardial effusion or pericardial thickening or<br>nodules or masses   |
| M1b         | Single distant (extrathoracic) metastasis in a single organ  |
| M1c         | Multiple distant (extrathoracic) metastases in a single organ or multiple organs   |



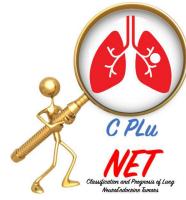
Compounding



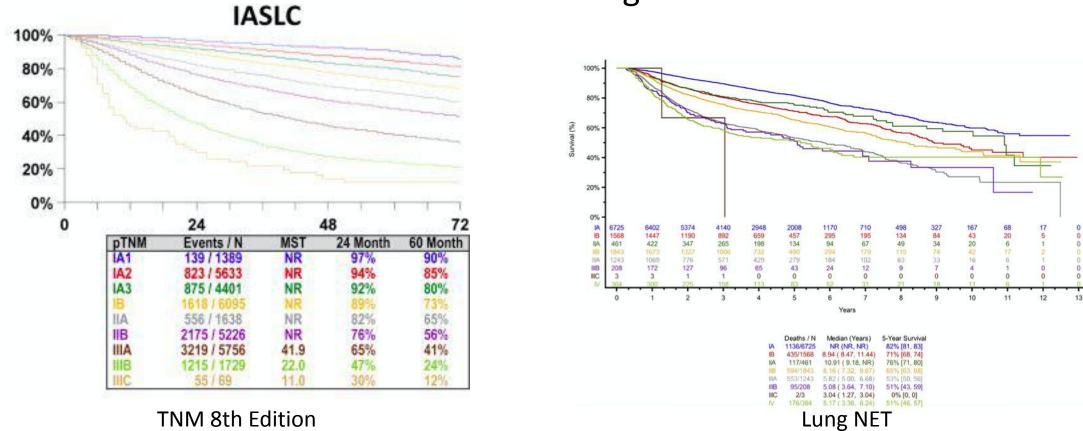


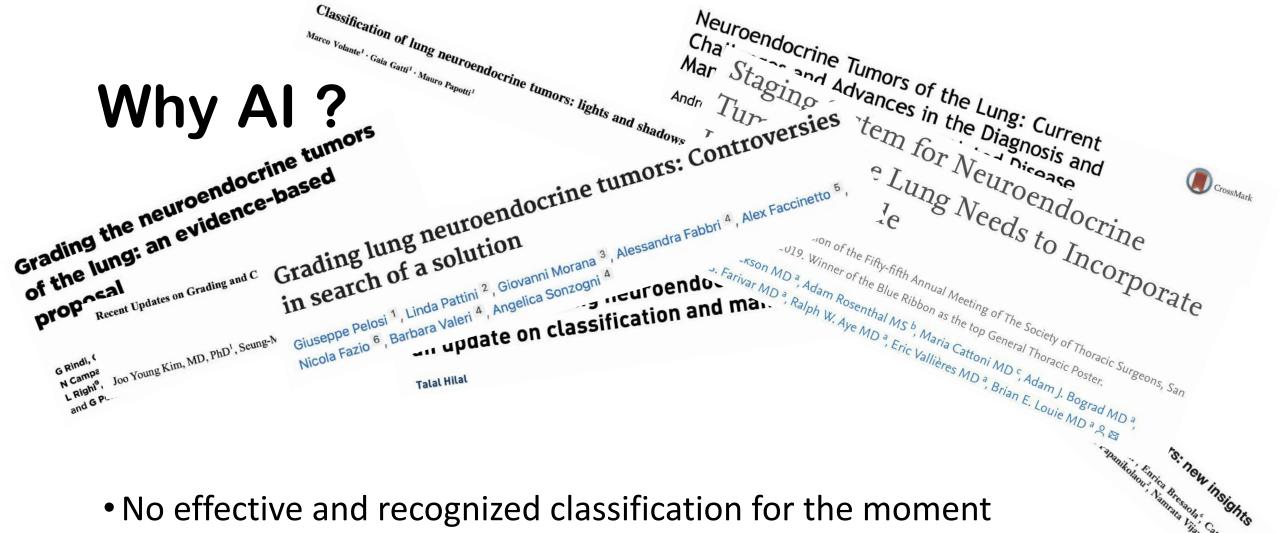
STAGE 4

## **Scientific Problem**



- Classification = Predictive of survival + Necessary for management
- Classical Classification = Useless for Lung NET





No effective and recognized classification for the moment

• Ongoing work of SFCTCV using classical statistical methods

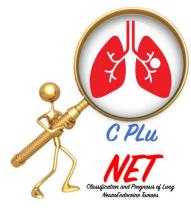
## **Challenge Objective**



 Create a histopronostic classification to predict survival of Lung NET and decide on standardized management

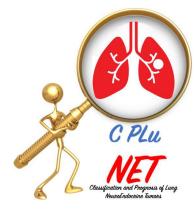
•Obj : Develop an algorithm to predict patient survival using patient clinical data and tumour-specific data

## Data Challenge



- 2 types of datasets
  - Patient clinical data from EPITHOR
  - Tumour-specific data from pathologist reports and pathology slides +++
- For 1500-2000 Patients
- Creation of an unique bank of digital and imaging data

## **EPITHOR**



- French National Database of General Thoracic Surgery
- Created 2002
- Prospective +++
- 111 centers = most surgical procedures performed in French thoracic surgical departments
- 52 variables per patient, covering information about the patients' characteristics, associated illness, pulmonary function, surgical procedures, cancer staging, and postoperative outcome
- Source of many scientific publications +++

## **EPITHOR**

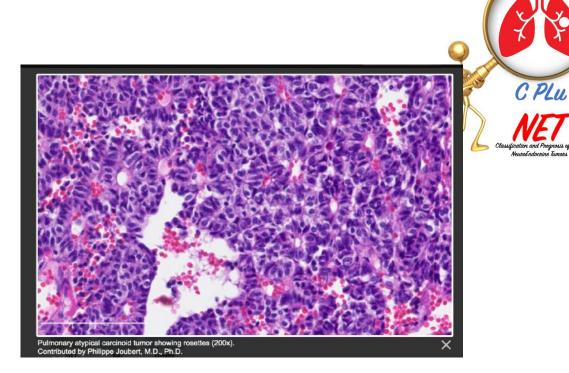


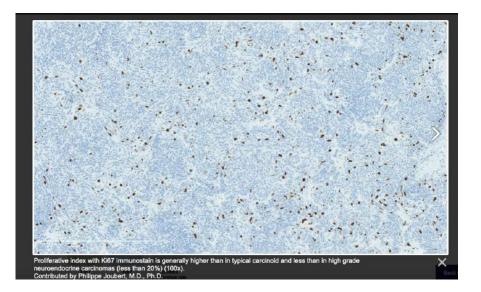
| 1  | Α                     | В      | С          | D         | E          | F   | G     | Н      | 1     | J     | К    | L      | м      | N      | 0    |
|----|-----------------------|--------|------------|-----------|------------|-----|-------|--------|-------|-------|------|--------|--------|--------|------|
| 1  | PATIENT               | Sexe 🗸 | DDNais 💌   | SEJOUR_D/ | ENTREE 🗖   | AGE | POIDS | TAILLE | BMI 모 | OMS 🗖 | FUME | VEMS 🔽 | VEMS 🗖 | TIFFEN | DLCC |
| 2  | 640000162_01-00000062 | М      | ########## | 2016      | 07/04/2016 | 79  | 75    | 173    | 25,1  | 1     | 15   | 83     | 56     |        |      |
| 3  | 630000479_01-CF_05893 | F      | ########## | 2014      | 07/01/2014 | 42  | 80    | 150    | 35,6  | 0     |      |        |        |        |      |
| 4  | 630000479_01-CF_04140 | М      | ########## | 2011      | 09/11/2011 | 67  | 90    | 170    | 31,1  | 0     |      | 87     | 87     |        |      |
| 5  | 630000479_01-CF_03087 | F      | ########## | 2010      | 16/02/2010 | 74  | 68    | 171    | 23,3  | 1     |      | 93     | 93     |        |      |
| 6  | 630000479_01-CF_03088 | F      | ########## | 2010      | 17/02/2010 | 37  | 61    | 150    | 27,1  | 0     |      |        |        |        |      |
| 7  | 630000479_01-CF_05758 | М      | ########## | 2013      | 21/10/2013 | 38  | 79    | 177    | 25,2  | 0     |      | 108    | 108    |        |      |
| 8  | 630000479_01-CF_05753 | М      | ########## | 2013      | 03/11/2013 | 73  | 98    | 171    | 33,5  | 1     |      | 121    | 121    |        |      |
| 9  | 630000479_01-CF_07469 | F      | ########## | 2015      | 20/10/2015 | 45  | 74    | 170    | 25,6  | 0     |      | 69     | 69     |        |      |
| 10 | 630000479_01-CF_04048 | М      | ########## | 2010      | 16/12/2010 | 76  | 60    | 174    | 19,8  | 1     |      |        |        |        |      |
| 11 | 630000479_01-CF_04041 | F      | ########## | 2011      | 13/09/2011 | 45  | 50    | 170    | 17,3  | 1     |      | 76     | 76     |        |      |
| 12 | 630000479_01-CF_04081 | М      | ########## | 2011      | 07/10/2011 | 64  | 88    | 175    | 28,7  | 1     |      | 110    | 110    |        |      |
| 13 | 630000479_01-CF_07393 | F      | ########## | 2015      | 01/10/2015 | 56  | 70    | 158    | 28    | 0     |      | 65     | 65     |        |      |
| 14 | 370004467_01-00000237 | F      | ########## | 2016      | 17/04/2016 | 76  | 53    | 168    | 18,8  | 0     | 10   | 109    | 86     |        |      |
| 15 | 370004467_01-00000238 | М      | ########## | 2016      | 18/05/2016 | 65  | 68    | 168    | 24,1  | 0     | 0    | 94     | 74     |        |      |
| 16 | 630000479_01-CF_04534 | М      | ########## | 2012      | 19/04/2012 | 74  | 78    | 180    | 24,1  | 0     |      | 52     | 52     |        |      |
| 17 | 630000479_01-CF_07120 | М      | ########## | 2015      | 22/06/2015 | 57  | 103   | 193    | 27,7  | 0     |      | 88     | 88     |        |      |
| 18 | 630000479_01-CF_07051 | F      | ########## | 2015      | 02/06/2015 | 65  | 59    | 150    | 26,2  | 0     |      | 86     | 86     |        |      |
| 19 | 630000479_01-CF_07063 | F      | ########## | 2015      | 07/06/2015 | 66  | 53    | 163    | 19,9  | 0     |      | 99     | 99     |        |      |
| 20 | 630000479_01-CF_07240 | F      | ########## | 2015      | 30/08/2015 | 30  | 72    | 173    | 24,1  | 0     |      |        |        |        |      |
| 21 | 630000479_01-CF_05985 | F      | ########## | 2014      | 11/02/2014 | 70  | 57    | 159    | 22,5  | 1     |      | 74     | 74     |        |      |
| 22 | 630000479_01-CF_05989 | F      | ########## | 2014      | 20/01/2014 | 45  | 74    | 168    | 26,2  | 1     |      | 83     | 83     |        |      |
| 23 | 630000479_01-CF_04288 | М      | #########  | 2011      | 22/06/2011 | 66  | 70    | 165    | 25,7  | 2     |      |        |        |        |      |

## **Pathology Slides**

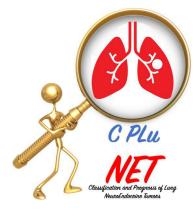
- Identify tumour-specific characteristics -> Prognosis
  - Cell « design »
  - Mitoses
  - . . . .
- From nearly 15 centers
- Scanning +++ (1Go for each slide)
- 2 for each patient :
  - -> Classical : H&E

-> Ki67





## **Pathology Slides**



- Review of the pathology slides by an expert committee
  - Validate the inclusion of neuroendocrine tumors
  - Underline the areas of interest on which the machines will work
- Creation of a unique virtual pathology slide bank for these rares tumors

## Data Challenge

- 2008-2020 : Typical Carcinoid, Atypical Carcinoid, LGCNEC(SCLC?)
- 1500-2000 patients /15 centers in France
  - 20 variables for each patient extracted from Epithor
  - 2 pathology slides per patient = 3-4000 pathology slides = 4To of data
  - + Macroscopic characteristics from pathology reports
  - Survival data from INSEE Database
- No missing data

### □ Primary outcome : Overall survival

• Challenge objective : Test the ability of the algorithm to predict patient survival with both types of data



## You're a Data Scientist ?

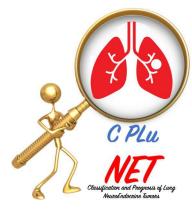
 First time that clinical data and pathology slides of exceptional quality are combined to create a classification using Al

• Unique way to participate in an exciting challenge for data scientists and clinicians

• Last but not least: 25000€



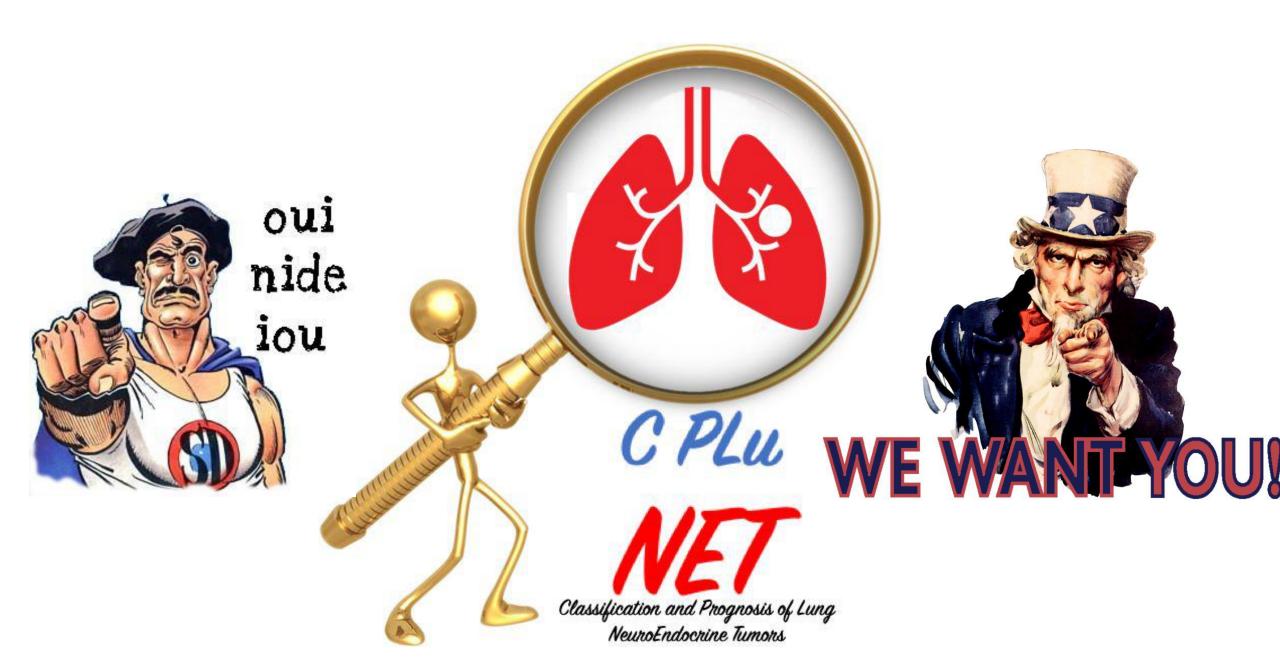
## **Dates to remember**



- Creation of datasets: May-December 2022
- Review by expert committee : November 2022-March 2023

Challenge : March-May 2023

• Prize-giving ceremony : June 2023





### **BREAK TIME !**



## 

### 3:15 - 4:05

### **STAY INFORMED!**

Scan the QR code to get the latest informations on our Data Challenges !









**b**pifrance





#### Sage Bionetworks - Organizing DREAM Challenges for Engaged Health Research





#### **Thomas Schaffter, PhD**

PhD, Lead of the Research & Benchmarking Technology Workstream Sage Bionetworks

Thomas Schaffter is the Lead of the Research & Benchmarking Technology Work Stream at Sage Bionetworks where he develops cloud infrastructure that enables the organization of biomedical scientific challenges.

4:05 – 4:15













### Organizing DREAM Challenges for Engaged Health Research

Thomas Schaffter, PhD Sage Bionetworks

PFUE Data Challenges for Health event May 24, 2022



powered by Sage Bionetworks



### **Thomas Schaffter**

PhD in Biotechnology & Bioengineering, EPFL, Lausanne, Switzerland.

Post-Doctoral Researcher @ IBM, New York, USA.

Senior Solutions Architect @ Sage Bionetworks, Seattle, USA.

**Mission**: accelerate biomedical discoveries by improving methods for scientific collaboration and communication

Better Science Together



https://www.linkedin.com/in/tschaffter/



https://www.github.com/tschaffter



Sage Bionetworks Office, Seattle.



### Our mission is ...

- to contribute to the solution of important **biomedical** problems
- to foster **collaboration** between research groups
- to **democratize access** to data
- to accelerate research
- to **objectively assess** algorithms and their performance

# SYNAPSE

#### **People and Teams**

- Group permissions
- Integrated auth
- ORCID links
- ARs, ACLs, and the ACT

#### **Data and Results**

- Versioning for data, code
- Provenance to link results
- Distributed data stores

+

• Continuous benchmarking

#### **Research Process**

- Connecting pieces of the broader research cycle
- Real-time collaboration

+

• Sharing across groups, teams, departments

#### **Synapse Platform and Integrated Tools**

Modular platform supports many different pieces of the research "puzzle", and all features can be leveraged in the web client, the analytic (R/Python/CLI) clients, or directly via the API enabling custom integrations/functionality

### **Recent challenge achievements**





First modern model-to-data challenge using Docker (2017)

First DREAM challenge on sensitive EHR data (2019)



First continuous challenge - EHR data (2020)





First multi-site evaluation - EHR data (2021)

https://dreamchallenges.org



### 

Out of every 1000 women screened, only 5 will have breast cancer. But 100 will be recalled for further testing. We can do better.

Build a model to help reduce the recall rate for breast cancer screening.

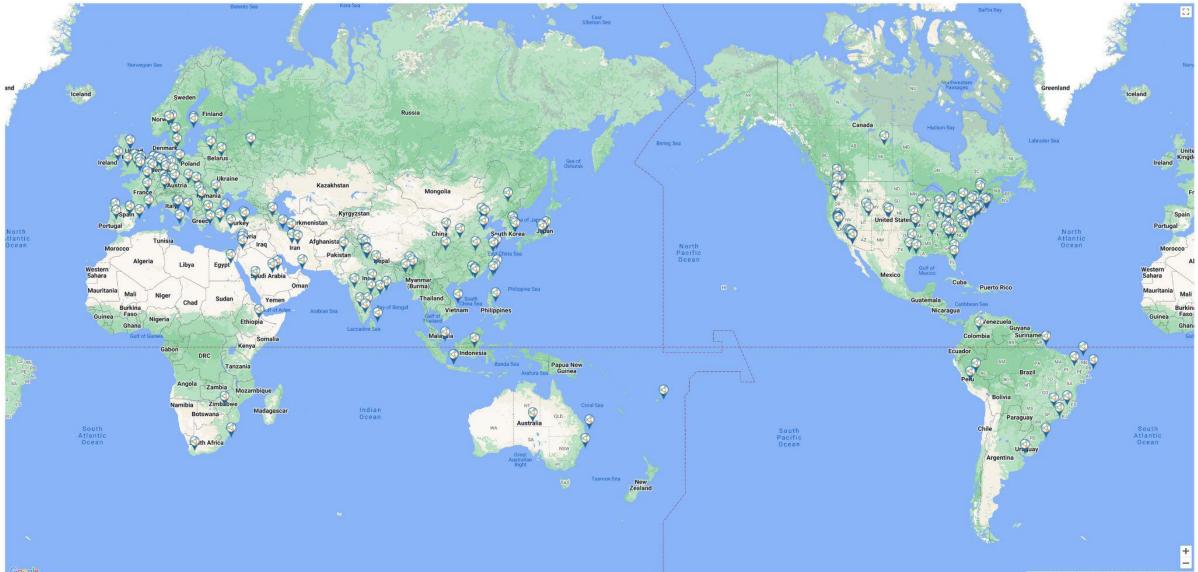
Calling all coders to join the Challenge.

Up to a **\$1,000,000** in cash prizes for winning models.

May the best model win.

https://www.synapse.org/Digital Mammography DREAM Challenge

### **DM DREAM Challenge Participants**



Keyboard shortcuts Map data ©2022 Google, INEGI Terms of Use







#### New 510(k) clearance expands the use of I mammography, opening up the entire breat

PARIS November 28, 2021 - Therapixel, a company leading the use of Al-based pro the U.S. Food and Drug Administration (FDA) for MammoScreen®, the explainable a breast cancer screening. This new 510(k) clearance (K211541) expands the use of I

This FDA clearance, received after submitting results from a multi-reader study conmammography market. Study findings revealed improvement in readers' performan radiologists alone. Additionally, the study showed significant time saving when usin

MammoScreen® automatically detects and characterizes suspicious soft tissue les their likelihood of malignancy. The results are summarized by the MammoScreen Sc 1, being least likely to reveal malignancy and 10 most likely.

"Receiving FDA clearance for MammoScreen is a major mileston of Paris-based Therapixel. "Thanks to a deep and fruitful collabo 2017 DREAM challenge winning 2D algorithm to a powerful FDAcan now assist all radiologists in their day-to-day-work whichever

### LOS ANGELES BUSINESS JOURNAL

THE COMMUNITY OF BUSINESS<sup>™</sup>

The Lists 🗸 Real Estate Health Care Tech Finance Manufacturing Special Editions 🗸 Events 🗸 Custom Content 🗸 Classifieds 🗸

HOME + HEALTH CARE

HEALTH CARE > MEDICAL DEVICES

#### FDA Approves RadNet Cancer Technologies

RadNet Inc. announced that two of its subsidiaries received approval from the Food

and Drug Administration for their mammography and prostate artificial

RadNet is a publicly traded diagnostic imaging company based in Los Angeles. Revenue for the company increased 8.4% year-over-year to \$341.8 million in the first quarter of 2022, according to the company's most recent quarterly report.

BY ANTONIO PEQUEÑO IV MAY 23, 2022

intelligence tools.

L.A. County Unemployment Rate Falls to 5.5% in April as **Employers Add Another** 13,000 Payroll Jobs

RELATED ARTICLES



ECONOMY







130M Loan Given for Self-



Schaffter, T., Buist, D.S., Lee, C.I., Nikulin, Y., Ribli, D., Guan, Y., Lotter, W., Jie, Z., Du, H., Wang, S. and Feng, J., 2020. Evaluation of combined artificial intelligence and radiologist assessment to interpret screening mammograms. JAMA network open, 3(3), pp.e200265-e200265.

## **Thanks!**

Data Challenge SFMN : Dat-Hub - Parkinson and dopamine transporter imaging: a data challenge from the Health Data Hub and the French Society of Nuclear Medicine





#### Aix-Marseille University, Timone hospital, SFMN leader of Brain Work

Pr Eric Guedj

4:15 - 4:25

Eric Guedj is Professor of Biophysics and Nuclear Medicine at Aix-Marseille University, and head of the Nuclear Medicine Department at Timone hospital. He is an expert in brain molecular imaging and theranostics, and the current SFMN leader of the Brain Work Group. Eric Guedj scientifically coordinates the "DaT-HUB" Data Challenge, with the objective to improve the diagnosis of Parkinson's disease by brain SPECT imaging of dopamine transporter.

Group

GOUVERNEMENT Liberté Égalité Fraternité





bpifrance





Société Française de Médecine Nucléaire et Imagerie Moléculaire

### **DaT-HUB** project

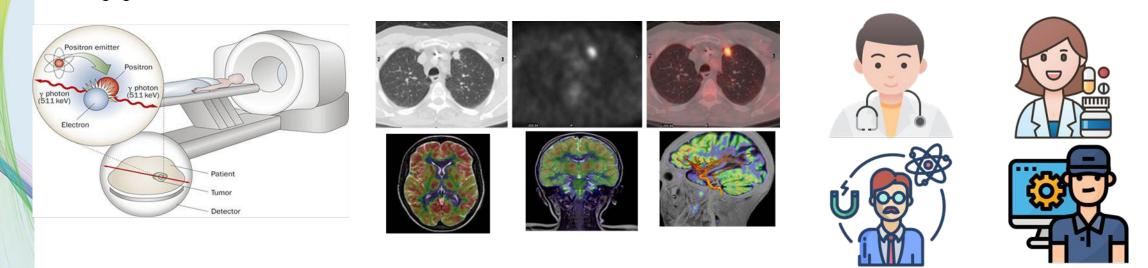


### SFMN - HDH Partnership Data Challenge Prof. Eric GUEDJ

## Nuclear Medicine in France



« The activity of nuclear medicine consists of the use, in diagnosis or therapy, of a radio-pharmaceutical or an active implantable medical device, in unsealed sources, emitting ionizing radiation, administered to the patient, including the use a single photon emission computed tomography [SPECT] or positron emission tomography [PET] camera and integrated, if applicable, other imaging systems », Art. R. 6123-134, Public Health Code



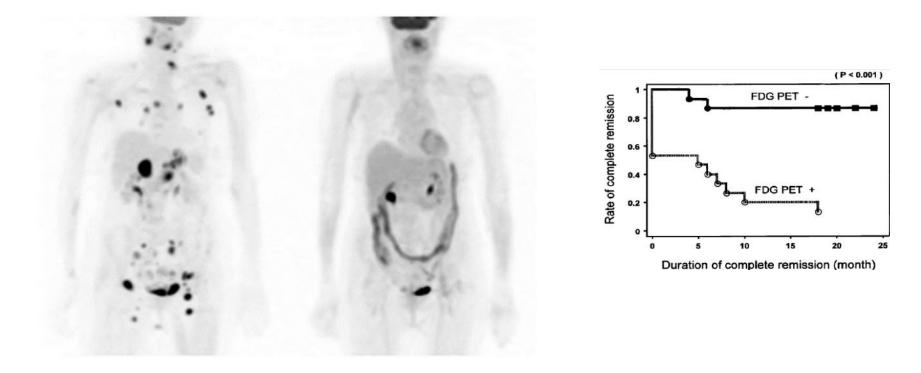
620 cameras in France, and 1.679.388 imaging exams in 2019 (+4.6% per year since 2008)
 927 physicians, 172 pharmacists, 192 physicists, 1,518 technologists

## The medical motives

- Need of **precise** and **early** characterization to:
  - diagnose and guide treatment of diseases otherwise incurable at later stages
  - evaluate efficiency of (expensive) therapeutics to not delay more effective treatment, and not cause unnecessary side effects
- Evolution towards personalized medicine
   groups of patients → the individual patient
- Diagnostic and therapeutic advances mainly driven by non-invasive bio-imaging, particularly at individual level, for molecular signature
   molecular changes precede morphological ones (*more sensitive for early diagnosis and evaluation*)

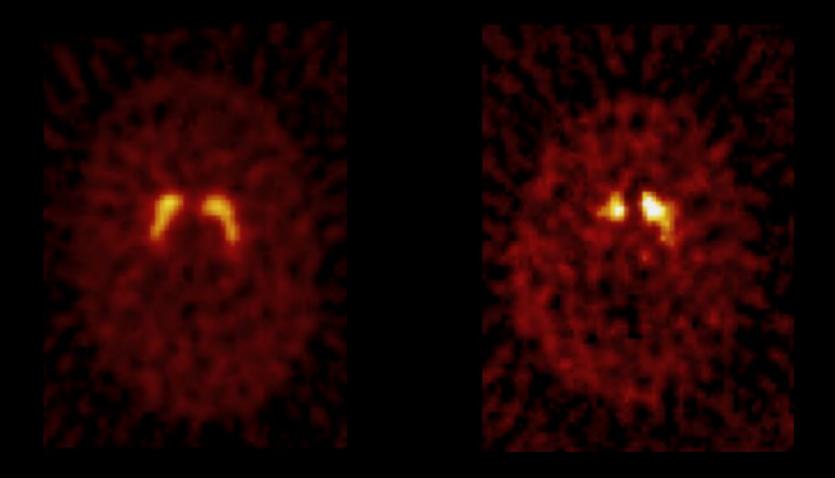
### PET Predicts Prognosis After 1 Cycle of Chemotherapy in Aggressive Lymphoma and Hodgkin's Disease

Lale Kostakoglu, MD<sup>1</sup>; Morton Coleman, MD<sup>2</sup>; John P. Leonard, MD<sup>2</sup>; Ichiei Kuji, MD<sup>1</sup>; Holly Zoe<sup>1</sup>; and Stanley J. Goldsmith, MD<sup>1</sup>



The Journal of Nuclear Medicine • Vol. 43 • No. 8 • August 2002

### 50% of brain dopaminergic loss before first symptoms of Parkinson



Healthy subject

Patient with Parkinson's disease

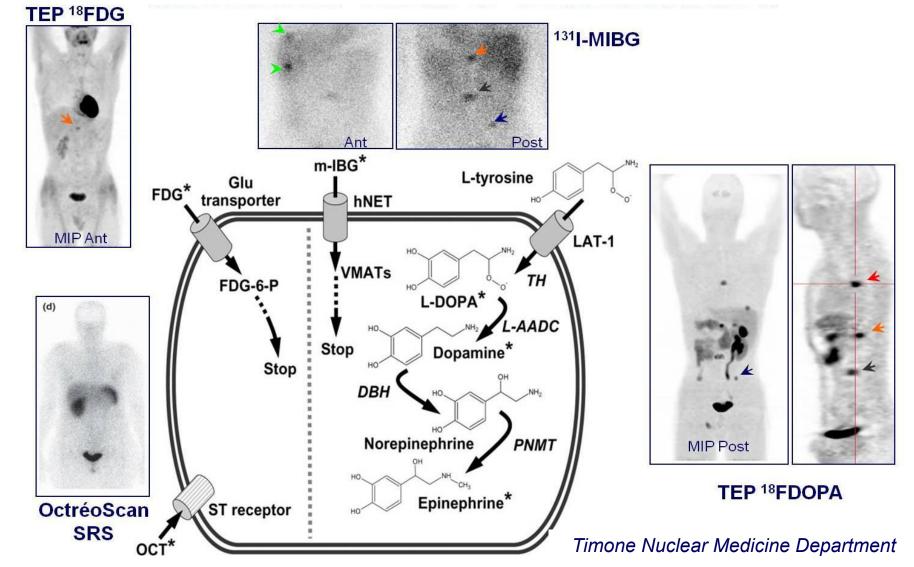
Parkinson's disease: 2<sup>nd</sup> most frequent neurodegenerative disease after AD, 1% of subjects after 60yrs, ≈ 200,000 patients in France, 30,000 new patients per year; 30% of diagnostic error (autopsy) with positive impact of early treatments

## The medical motives

- Need of **precise** and **early** non-invasive characterization to:
  - diagnose and guide treatment of diseases otherwise incurable at later stages
  - evaluate efficiency of (expensive) therapeutics to not delay more effective treatment, and not cause unnecessary side effects
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  - molecular changes precede morphological ones (*more sensitive for early diagnosis and evaluation*)
  - molecular biomarkers better evaluate complexity of each disease at patientand lesion- level (more specific for precise diagnosis and evaluation)

# Molecular signature of neuro-endocrine tumors



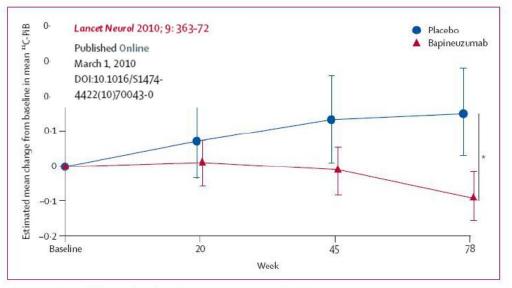


# The medical motives

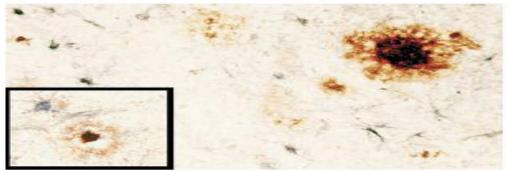
- Need of **precise** and **early** non-invasive characterization to:
  - diagnose and guide treatment of diseases otherwise incurable at later stages
  - evaluate efficiency of (expensive) therapeutics to not delay more effective treatment, and not cause unnecessary side effects
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- Diagnostic and therapeutic advances mainly driven by non-invasive bio-imaging, particularly at individual level, for molecular signature
  - molecular changes precede morphological ones (*more sensitive for early diagnosis and evaluation*)
  - molecular biomarkers better evaluate complexity of each disease at patientand lesion- level (more specific for precise diagnosis and evaluation)
    - ✓ this molecular complexity is linked to prognosis, and therapy (*companion drugs*)

<sup>11</sup>C-PiB PET assessment of change in fibrillar amyloid-β load in patients with Alzheimer's disease treated with bapineuzumab: a phase 2, double-blind, placebo-controlled, ascending-dose study

Juha O Rinne, David J Brooks, Martin N Rossor, Nick C Fox, Roger Bullock, William E Klunk, Chester A Mathis, Kaj Blennow, Jerome Barakos, Aren A Okello, Sofia Rodriguez Martinez de Llano, Enchi Liu, Martin Koller, Keith M Gregg, Dale Schenk, Ronald Black, Michael Grundman



**Figure 2: Estimated change from baseline over time in mean** <sup>11</sup>C-PiB PET Data are least squares means and 95% CIs. \*Difference between patients in the placebo group and those in the bapineuzumab group at week 78=-0.24 (p=0.003). PiB=Pittsburgh compound B.



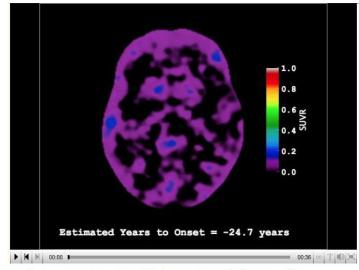
### The NEW ENGLAND JOURNAL of MEDICINE

ESTABLISHED IN 1812

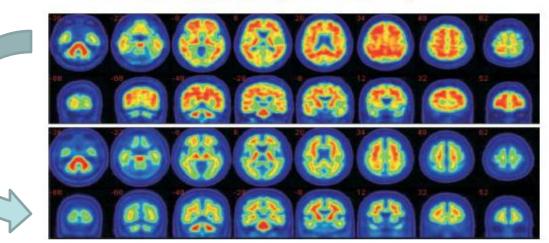
AUGUST 30, 2012

VOL. 367 NO. 9

Clinical and Biomarker Changes in Dominantly Inherited Alzheimer's Disease

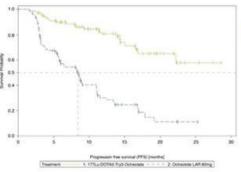


• 18F-PIB Flutemetamol (GE Healthcare)



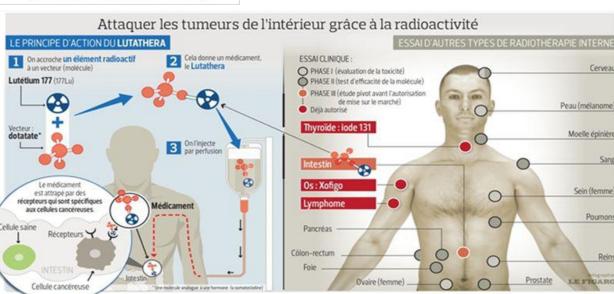
# **Molecular Imaging & theranostics**

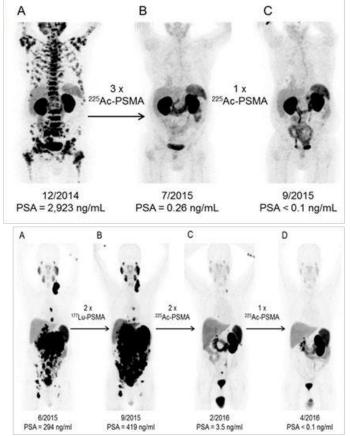
- Molecular process targetted by a tracer
- Labeled by a radio-isotope: <u>diagnostic & therapeutic (companion pair)</u>



<sup>177</sup>Lu-Dotatate Improves Progression-Free Survival in Patients with Midgut Neuroendocrine Tumours: Results of the Phase III **NETTER-1** Trial

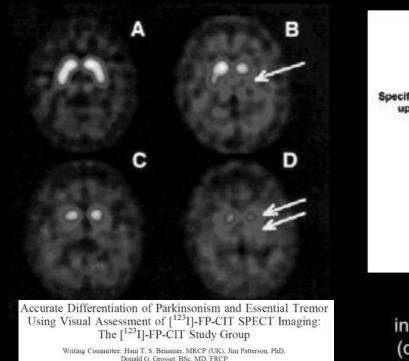
N Engl J Med. 2017 Jan 12;376(2):125-135



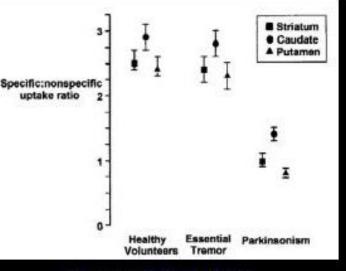


J Nucl Med. 2016 Jul;57(7):1006-13

# **DaT SPECT Imaging for Parkinsonism**

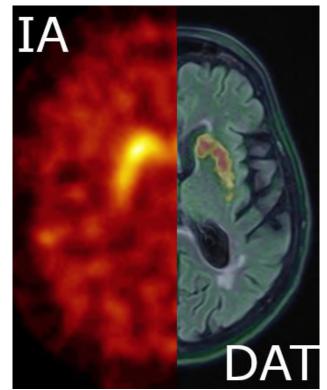


Institute of Neurological Sciences, Glasgow, U.K.

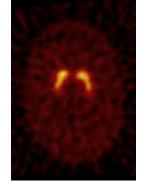


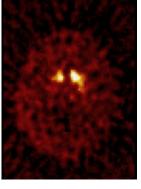
Se = 97%; Sp = 100%

inter-operator concordance = 95% (qualitative score on 4-step scale)



- Brain SPECT imaging of dopamine transporter activity (DaTSCAN©):
  - a binary biomarker of degenerative parkinsonism:
    - ✓ No dopaminergic loss in non-degenerative causes (≈ 1/3 of exams)
    - ✓ Dopaminergic loss in degenerative cause (≈ 2/3 of exams)
  - 50% of dopaminergic loss at diagnosis in case of degeneration
  - 17,250 exams per year in France, in  $\approx$  200 centres
  - Se/Sp: 97%/100%, inter-operator concordance: 95%
  - 20% of uncertain classification in non-expert interpretation in practice
  - □ Visual interpretation based on morphological/intensity criteria



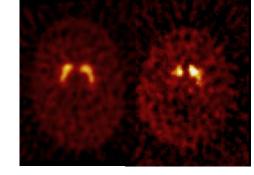


Normal Pathological



DATA HUB

# The DaT-HUB data challenge



- A multicentric French national DaTSCAN<sup>©</sup> database to further develop in 2023 an automated diagnostic aid algorithm identifying normal vs pathological exams a data challenge
- ≈ 5,000 exams to be collected (3D images, < 10 Mo per scan) with consensual binary interpretation among 3 national experts, and additional data (age, sex, acquisition parameters, confidence index, asymmetry index, severity index)
- The winner will be the algorithm with the performance of maximal concordance with the expert interpretation (25,000 € prize)
- Further scientific objectives on possible interaction with French healthcare database (SNDS):
  - Correlation between the algorithm score and the quantitative visual score, according to the clinical severity
  - □ Identification of possible iatrogenic or neuroprotective drugs
  - Impact of certain conditions on dopamine activity (such as long COVID)
  - ....



# Data Challenge HUS : D-IA-GNO-DENT - HDH Data Challenge - Rare oral diseases / Artificial intelligence / Diagnostic assistance





#### \_\_\_\_\_

4:25 - 4:35

#### Pr Agnès Bloch Zupan

(PU-PH) Hôpitaux Universitaires de Strasbourg,Centre de référence coordonnateur des maladies rares orales et dentaires CRMR O-Rares Université de Strasbourg, Faculty of Dental Surgery and Institute of Genetics and Molecular and Cellular Biology, CNRS- UMR7104, INSERM U1258

Agnès Bloch-Zupan is Professor in Oral Biology at the Faculty of Dentistry, Université de Strasbourg, France. She coordinates the Reference Center for Rare Oral and Dental Diseases, CRMR O-RARES of the University Hospitals of Strasbourg and affiliated French constitutive and competence centers. She supervises clinical research projects and is the scientific director of the D(4)/phenodent database (www.phenodent.org) as well as the genetic diagnosis NGS tool GenoDENT.













# D\_IA\_GNO\_DENT : Artificial intelligence helping diagnosis of rare diseases with oral expressions

#### Pr Agnès Bloch-Zupan

PU-PH professor, clinician, rare disease expert, researcher at IGBMC UMR Unistra 7104, Inserm U1258

Coordinator CRMR O-Rares Rare diseases Reference Center HUS and network, French Filière TETE COU, ERN CRANIO

#### Guillaume Icre Innovation and Valorization HUS Christine Lecomte E-Health HUS



Rare oral and dental Diseases (O-RARES) Network

> FILIÈRE SANTÉ MALADIESRARES TeteCou







# **Context and challenges**



**Developmental abnormalities** of the oral cavity and teeth



900 rare diseases/syndromes with dental anomalies and other associated symptoms



**Dental development is regulated by** numerous genes



Mineralization fixes in time the anomalies



**Dental defects** exist in possible combination with disorders of other systems : retina, skeleton, kidney, skin, epilepsy...



Currently there is an important diagnostic wandering : Patients are not diagnosed with "rare disease" and do not benefit from specific personalized care.

# The idea A virtual expert for

diagnosis help. Transdisciplinary analysis for improved patient care.

# How



An Al-based solution analyzing images of the oral cavity, for automatic recognition of diagnostic signatures.

# Why



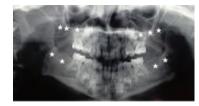
Offer the expertise of the O-Rares network to health professionals.

Move from diagnostic wandering to a personalized diagnostic itinerary!

Rare diseases with oral expression

Patients, data D4/phenodent database, genetic testing NGS GenoDENT 567 genes, ARS

- Life and Health research expertise
- An AI-based solution analyzing panoramic radiographs i-DENT AMI Grand-Est 2019
- For automatic recognition of radiographic diagnostic signatures

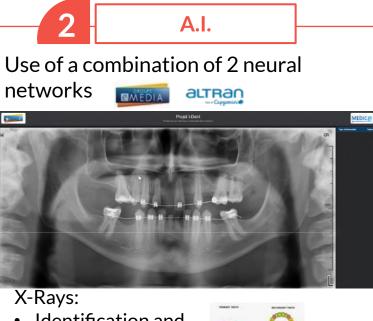


Les Hopitaux

PAX9 – Absence of molars AD / Heterozygous deletion PAX9 (NM\_006194.3): c.(?\_-115)\_(\*62\_?)del.







- Identification and segmentation of teeth types
- f O =
- (I,Ca,PreM,Mo) Coming soon :
- Identification of teeth according
   to the FDI nomenclature
- Dental age
- Correspondance to a diagnostic signature pattern
- From the signature to the gene.



# **Current work**

HDH : to go further...



- Combining radiographic and clinical signs (photos) to identify rare diseases.
- Other parameters may come into play:



shape,

color.

texture.



Patient history...



**Data/Cohorts** 

Amelogenesis imperfecta 200 + 20 - Dentine anomalies 50 - Environmental anomalies 50 - Control 100

Intraoral colour photographs (4 to 6) Colour, Surface texture, Shape/size







Key-words D4/phenodent – HPO : AI hypoplastic, hypomineralized, hypomature; Medical information; Gene - GenoDENT; Diagnosis



# **Developped solution**

### For whom?



Designed for rare disease patients

- Decrease diagnostic wandering
- Accelarate care in the « right » care pathway

### By whom?



- Access to enhanced knowledge of rare diseases
- Acces to rapid expertise

### How?



#### Input and output data

- Inputs : images, reports, diagnostic signatures...
- Output : a (pre-)diagnosis of suspected rare disease with oral manifestations

## What?



- An expert tool to help with pre-diagnosis, or even diagnosis when faced with a suspected rare disease with oral expressions.
- Thanks to a collection of descriptive data, managed by an artificial intelligence engine, it wilg guide the professional in a personalized way.



Expertise in Health Research and Patient Data

Creation and testing of an A.I. based

teleexpertise software

Platform

**Experts** 

Secure and adapted data transfer/development

### Challengers

A.I. expertise







#### PREPARE HDH/BPI

#### **BUILD Data Challenge**

HDH MOU BPI application BPI official notification

BPI signature /HUS

MAY 2022

Scientific committee
Executive committee
Contracts
Data protection, security
RGPD
Data preparation
Communication/challengers
IUNE 2022-FEBRUARY 2023

CHALLENGE DATA

PLATFORM COMPETITION

**MARCH-MAY 2023** 

POST Data Challenge

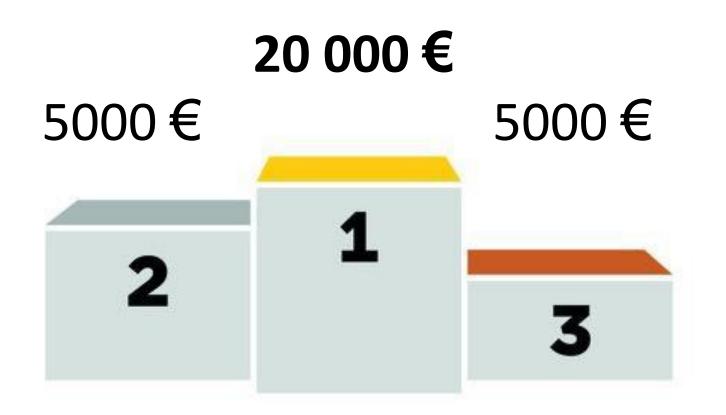
Algorithms tests REWARD

Administrative closure Publications Open data Communication JUNE-OCTOBER 2023



Challenge and reward

## European challengers : 30 000 euros





# Data Challenge SFA : Allergen Chip Challenge: artificial intelligence at the service of the allergic patient





#### Pr Joana Vitte

MD, PhD, Associate Professor - Hospital Practitioner in Immunology, Aix-Marseille University and Montpellier University, President of the Scientific Committee of the French Society of Allergology.

#### 4:35 - 4:45

Joana Vitte is a specialist in immunological mechanisms and tools in allergology, combining hands-on hospital innovative diagnostic. translational work. and epidemiological research and undergraduate and postgraduate training. She currently acts as the coordinator of the Allergen Chip Challenge project in collaboration with Health Data Hub and funded by the Banque Publique d'Investissement, aiming at further developments in the field of open data and artificial intelligence tools for allergists and immunologists.















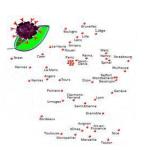
#### Principal investigator: Société Française d'Allergologie



# **The Allergen Chip Challenge**

#### Speaker: Dr Joana Vitte, MD, PhD

Associate Professor, Aix-Marseille University, MEPHI



AllergoBioNet Le Réseau d'ImmunoAllergologie Biologique Hospitalière

Associate Researcher, Montpellier University, Desbrest Institute of Epidemiology and Public Health

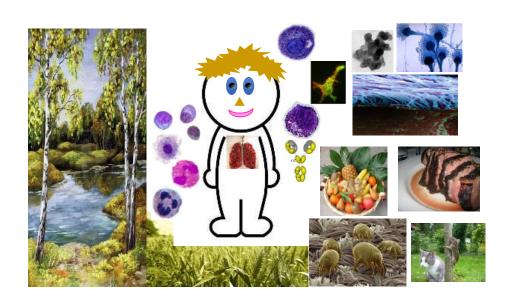
President of the Scientific Council of the French Society of Allergology







# Context: Exposure to potential allergens is universal, but not everyone develops IgE antibodies and allergy



The exposome concept and the immune response

## Allergy = antigen-specific loss of tolerance

Allergen = an antigen that causes an allergic disease by inducing an adaptive immune response

Allergen-specific immunoglobulin (Ig)E = the hallmark of allergic immune response

Allergy diagnosis = clinical history + demonstration of immunoglobulin (Ig) E to culprit allergen

Allergens come in a virtually unlimited variety, and so do IgE

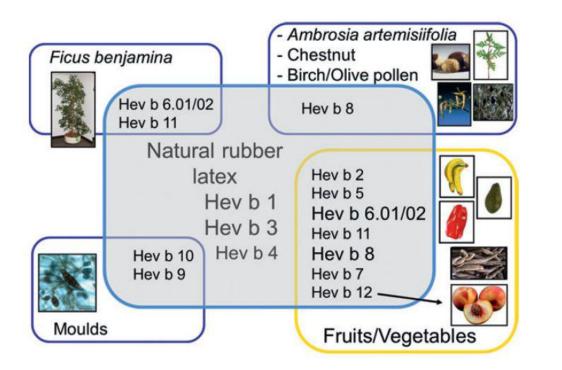


ALLERGEN NOMENCLATURE WHO/IUIS Allergen Nomenclature Sub-Committee Financial contribution from IUIS, EAACI, and AAAAI organizations



Bacher P & Scheffold A, *J Allergy Clin Immunol* 2018 Chruszcz M *et al*, *Allergy* 2021

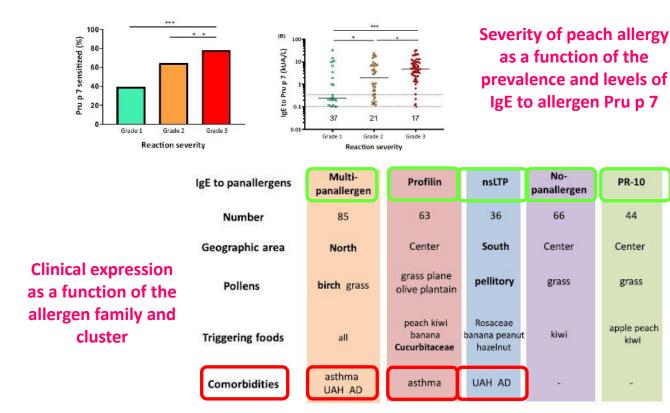
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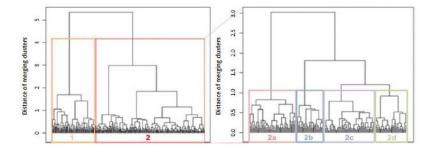


= cross-recognition by IgE and induction of multiple symptoms

Matricardi PM *et al, Pediatr Allergy Immunol* 2016; Mastrorilli C *et al, Allergy* 2016; Klingebiel C *et al, Clin Exp Allergy* 2019

### 





# Visit card of an allergic patients: IgE landscape + clinical history

#### 2. IgE results sorted by protein group

| Egg white             | nGald 1            | Ovomucoid                               | <0.3 ISU-E | Weed pollen          |
|-----------------------|--------------------|---|------------|----------------------|
|                       | nGal d 2           | Ovalbumin                               | <0.3 ISU-E | Ragweed              |
|                       | nGal d 3           | Conalbumin/Ovotransferrin               | <0.3 ISU-E | Mugwort              |
| Egg yolk/chicken meat | nGald 5            | Livetin/Serum albumin                   | 2,8 ISU-E  | Goosefoot            |
| Cow's milk            | nBos d 4           | Alpha-lactalbumin                       | <0.3 ISU-E | Wall pelitory        |
|                       | nBos d 5           | Beta-lactoglobulin                      | <0.3 ISU-E | Plantain             |
|                       | nBos d 8           | Casein                                  | <0.3 ISU-E | Saltwort             |
|                       | nBos d lactoferrin | Transferrin                             | <0.3 ISU-E | Animal               |
| Cod                   | rGad c 1           | Parvalbumin                             | <0.3 ISU-E | Dog                  |
| Shrimp                | nPen m 2           | Arginine kinase                         | <0.3 ISU-E |                      |
|                       | nPen m 4           | Sarcoplasmic calcium binding<br>protein | <0.3 ISU-E |                      |
| Cashew nut            | rAna o 2           | Storage protein, 11S globulin           | 1 ISU-E    | Horse                |
| Brazil nut            | rBer e 1           | Storage protein, 2S albumin             | <0.3 ISU-E | Cat                  |
| Hazelnut              | nCor a 9           | Storage protein, 11S globulin           | 0,5 ISU-E  |                      |
| Walnut                | rJug r 1           | Storage protein, 2S albumin             | <0.3 ISU-E | Mouse                |
|                       | nJug r 2           | Storage protein, 7S globulin            | <0.3 ISU-E | Mold                 |
| Sesame seed           | nSes i 1           | Storage protein, 2S albumin             | <0.3 ISU-E | Alternaria           |
| Peanut                | rAra h 1           | Storage protein, 7S globulin            | 57 ISU-E   |                      |
|                       | rAra h 2           | Storage protein, Conglutin              | >100 ISU-E | Aspergillus          |
|                       | rAra h 3           | Storage protein, 11S globulin           | >100 ISU-E | Asperginus           |
|                       | nAra h 6           | Storage protein, 2S albumin             | >100 ISU-E |                      |
| Soybean               | nGly m 5           | Storage protein, Beta-conglycinin       | <0.3 ISU-E | 0.1                  |
|                       | nGly m 6           | Storage protein, Glycinin               | 60 ISU-E   | Cladosporium         |
| Buckwheat             | nFag e 2           | Storage protein, 2S albumin             | <0.3 ISU-E | Mite                 |
| Wheat                 | rTri a 19.0101     | Omega-5 gliadin                         | <0.3 ISU-E | B. tropicalis (HDM)  |
|                       | nTri a aA_TI       | Alpha-amylase / Trypsin inhibitor       | <0.3 ISU-E | D. farinae (HDM)     |
| Kiwi                  | nAct d 1           | Cysteine protease                       | <0.3 ISU-E |                      |
|                       | nAct d 5           | Kiwellin                                | <0.3 ISU-E | D. pteronyssinus (HD |

| Grass pollen   |           |                              |            |   |
|----------------|-----------|------------------------------|------------|---|
| Bermuda grass  | nCyn d 1  | Grass group 1                | <0.3 ISU-E |   |
| Timothy grass  | rPhl p 1  | Grass group 1                | <0.3 ISU-E |   |
|                | rPhl p 2  | Grass group 2                | 0,6 ISU-E  | - |
|                | nPhI p 4  | Berberine bridge enzyme      | <0.3 ISU-E |   |
|                | rPhl p 5  | Grass group 5                | <0.3 ISU-E |   |
|                | rPhl p 6  | Grass group 6                | <0.3 ISU-E |   |
|                | rPhi p 11 | Ole e 1-related protein      | <0.3 ISU-E |   |
| Free pollen    |           |                              |            |   |
| Birch          | rBet v 1  | PR-10 protein                | <0.3 ISU-E |   |
| Japanese cedar | nCry j 1  | Pectate lyase                | 0,5 ISU-E  | - |
| Cypress        | nCup a 1  | Pectate lyase                | 0,5 ISU-E  |   |
| Olive pollen   | rOle e 1  | Common olive group 5         | 19 ISU-E   | - |
|                | rOle e 9  | Beta-1,3-glucanase           | <0.3 ISU-E |   |
| Plane tree     | rPla a 1  | Putative invertase inhibitor | 13 ISU-E   |   |
|                | pPla a 2  | Polygalacturonase            | <0.3 ISULE |   |

IgE

| Weed pollen                  |          |                                |            |
|------------------------------|----------|--------------------------------|------------|
| Ragweed                      | nAmb a 1 | Pectate lyase                  | <0.3 ISU-E |
| Mugwort                      | nArt v 1 | Defensin                       | <0.3 ISU-E |
| Goosefoot                    | rChe a 1 | Ole e 1-related protein        | <0.3 ISU-E |
| Wall pelitory                | rParj2   | Lipid transfer protein (nsLTP) | <0.3 ISU-E |
| Plantain                     | rPla I 1 | Ole e 1-related protein        | <0.3 ISU-E |
| Saltwort                     | nSal k 1 | Pectin methylesterase          | <0.3 ISU-E |
| Animal                       |          |                                |            |
| Dog                          | rCan f 1 | Lipocalin                      | <0.3 ISU-E |
|                              | rCan f 2 | Lipocalin                      | 2,3 ISU-E  |
|                              | rCan f 5 | Arginine Esterase              | <0.3 ISU-E |
| Horse                        | rEqu c 1 | Lipocalin                      | <0.3 ISU-E |
| Cat                          | rFel d 1 | Uteroglobin                    | 1,2 ISU-E  |
|                              | rFel d 4 | Lipocalin                      | <0.3 ISU-E |
| Mouse                        | nMus m 1 | Lipocalin                      | <0.3 ISU-E |
| Mold                         |          |                                |            |
| Alternaria                   | rAlt a 1 | Acidic glycoprotein            | <0.3 ISU-E |
|                              | rAlt a 6 | Enolase                        | <0.3 ISU-E |
| Aspergillus                  | rAsp f 1 | Mitogillin family              | <0.3 ISU-E |
|                              | rAsp f 3 | Peroxysomal protein            | <0.3 ISU-E |
|                              | rAsp f 6 | Mn superoxide dismutase        | <0.3 ISU-E |
| Cladosporium                 | rCla h 8 | Mannitol dehydrogenase         | <0.3 ISU-E |
| Mite                         |          |                                |            |
| B. tropicalis (HDM)          | rBlot 5  | Mite group 5                   | <0.3 ISU-E |
| D. farinae (HDM)             | nDer f 1 | Cysteine protease              | <0.3 ISU-E |
|                              | rDer f 2 | NPC2 family                    | <0.3 ISU-E |
| D. pteronyssinus (HDM)       | nDer p 1 | Cysteine protease              | <0.3 ISU-E |
|                              | rDer p 2 | NPC2 family                    | <0.3 ISU-E |
| L. destructor (storage mite) | rLep d 2 | NPC2 family                    | <0.3 ISU-E |
| Cockroach                    |          |                                |            |
| Cockroach                    | rBlag 1  | Cockroach group 1              | <0.3 ISU-E |
|                              | rBla g 2 | Aspartic protease              | <0.3 ISU-E |
|                              | rBla g 5 | Glutathione S-transferase      | <0.3 ISU-E |

| PR-10 protein |          |               |            |
|---------------|----------|---------------|------------|
| Peach         | rPru p 1 | PR-10 protein | <0.3 ISU-E |
| Soybean       | rGly m 4 | PR-10 protein | <0.3 ISU-E |
| Peanut        | rAra h 8 | PR-10 protein | 1,7 ISU-E  |
| Kiwi          | rAct d 8 | PR-10 protein | <0.3 ISU-E |
| Celery        | rApi g 1 | PR-10 protein | <0.3 ISU-E |

... and counting... to >100 and up to 300 allergen-specific

Female, 12 years Life-threatening allergic reaction to peanut and soybean

Previously allergic to egg, now tolerates it Olive pollen rhinitis and asthma

Questions: should this patient carry an epinephrine autoinjector? be allowed to eat cashew nuts? is this patient going to outgrow the peanut/soybean allergy or rather react to further related plant foods? Should the family dog be banned from the sofa in the living room?



How to make the most of this set of data for the patient's management?

#### Allergen Chip Challenge: the idea behind the project



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#### Data flow for allergy diagnosis using allergen multiplex investigation: current situation and projection of Allergen Chip Challenge outcome.

- **a**, Current situation: multiple pathways and significant disconnection for *in vitro* and clinical data, hampering integrated interpretation.
- **b**, Allergen Chip Challenge objective: data collection, database implementation, data challenge for an AI solution efficiently bridging clinical and *in vitro* data for diagnosis, prognosis, therapeutic choice and follow-up of allergic diseases using allergen multiplex investigation.

#### **Allergen Chip Challenge**

Artificial Intelligence project in Molecular Allergology Partnership: French Society of Allergology and Health Data Hub

| Tin | neline                         | Background:  |  |  |  |  |  |
|-----|--------------------------------|--|--|--|--|--|--|
| *   | May 2021:<br>application       | <ul> <li>allergen multiplex investigation          hundreds of numerical data for each patient          complex issue for allergists     </li> <li>Health Data Hub has initiated a « Grand Challenge program on Artificial Intelligence for improving medical diagnosis » and will partner         with medical societies to implement applied research projects     </li> </ul> |  |  |  |  |  |
| *   | September                      | Objectives:  |  |  |  |  |  |
|     | 2021:<br>accepted              | Data collection: allergen multiplex and clinical data  |  |  |  |  |  |
| •   |                                | Data challenge   |  |  |  |  |  |
| *   | January 2022:<br>financial and | Development of an AI solution for the clinical interpretation of IgE sensitization data obtained through allergen multiplex investigation  |  |  |  |  |  |
|     | regulatory                     | □ Open science   |  |  |  |  |  |
|     | application                    | Scientific partners:   |  |  |  |  |  |
| *   | Notification                   | French Society of Allergology = principal investigator   |  |  |  |  |  |
|     | of                             | Desbrest Institute of Epidemiology and Public Health (IDESP, INSERM UMR UA 11, Montpellier, France)  |  |  |  |  |  |
|     | acceptance:<br>May 2022        | AllergoBioNet (French association of allergy lab specialists from university and public hospitals)   |  |  |  |  |  |
|     | 1010 2022                      | Data:  |  |  |  |  |  |
| *   | Expected                       | 3,000 allergen multiplex (ISAC, ALEX, FABER) and corresponding clinical data (retrospective collection) 🗆 > 500 000 IgE data with corresponding  |  |  |  |  |  |
|     | completion:<br>December        | clinical information.  |  |  |  |  |  |
|     | 2023                           | Challenge:   |  |  |  |  |  |
|     |                                | Database hosted by a data challenge platform, open to international competitors. Selection of the algorithm with the best ability to predict   |  |  |  |  |  |
|     |                                | clinical data based on IgE sensitization   |  |  |  |  |  |
|     |                                | Expected outcome:  |  |  |  |  |  |
|     |                                | Open source of AI solution or partnership with the French Society of Allergology   |  |  |  |  |  |

# Key messages

# Hot topic for Allergology

- To get hold of the relationship exposome – IgE – clinical expression of allergy
- To efficiently diagnose and stratify allergic patients
- To better understand the IgE clusters as a function of age, symptoms, and climate area



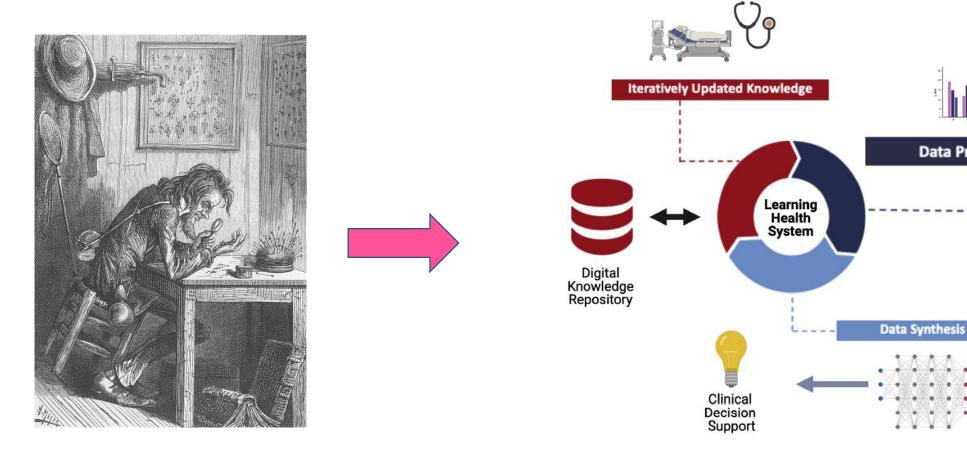


# Hot topic for data challenge competitors

- Dataset Excel sheet
- Tackling immune response complexity
- Open data and open science approach
- Data challenge winner: prize: 25 k€



### Thank you for your attention



**Data Processing** 

AI

### **Open science: sharing and reusing research data**





#### **Dr Lorien Benda**

Open science project manager at Health Data Hub

Lorien Benda has a PhD in chemistry and is a project manager at the Health Data Hub. She is in charge of the open science, open source strategy and interoperability strategy aiming at decompartmentalising knowledge around health databases in order to facilitate their reuse.



#### **Isabelle Blanc**

National Director of Data, Algorithms and Source Codes for Research and Innovation of the Ministry of Higher Education, Research and Innovation in France.

Isabelle Blanc, originally a researcher in the field of human nutrition, has a leading role in the public policies of Open Science, especially on opendata and opensource. She drives all actions aiming at structuring, sharing and opening up research data, as well as opening up and promoting software source code produced by research.

#### 4:45 - 5:00









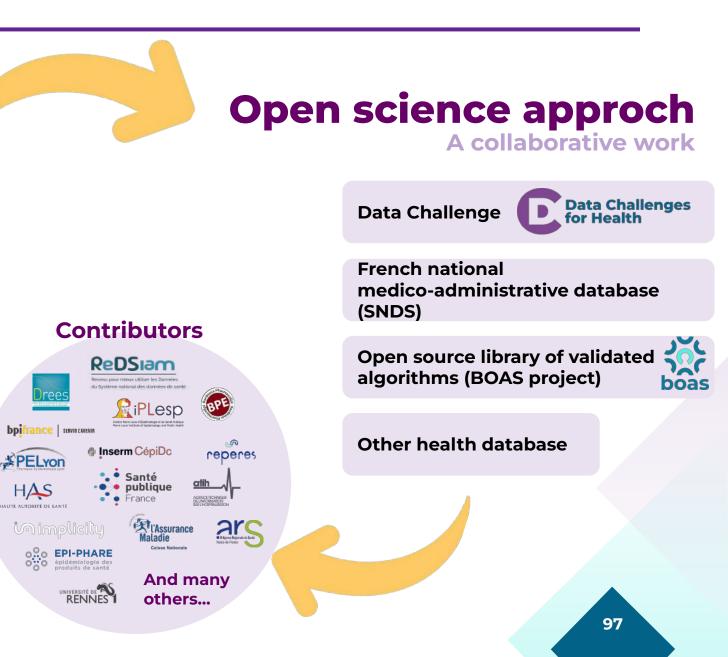




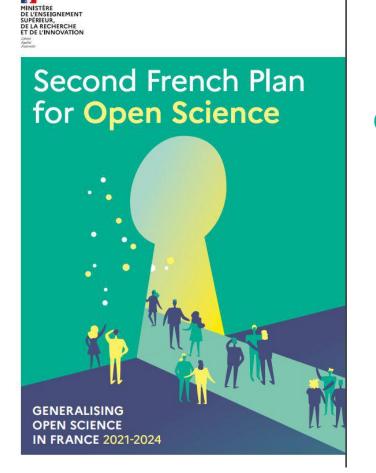
# The Health Data Hub's open science approach

The Health Data Hub its and partnerships are developing, making available and updating a series of tools designed to facilitate the manipulation of data. health **Because** understanding data their organisation, but also their biases, potentials and limits - represents a considerable effort for each project leader. the Health Data Hub proposes to share knowledge: this is the open science approach.





MINISTÈRE DE L'ENSEIGNEMENT SUPÉRIEUR, DE LA RECHERCHE ET DE L'INNOVATION Liberté Égatité Fratemité



www.ouvrirlascience.fr/home/

French open science policy : a dedicated chapiter to research data *Structuring, sharing and opening up research data* 

« We will encourage practices that favor research data reuse. »

#### Give recognition to and boost the reuse of research data

 $\rightarrow$  Encourage best practices in **data citation**.

 $\rightarrow$  Award an annual **research data prize** to highlight the work of exemplary projects and teams in preparing to reuse or reusing research data.

 $\rightarrow$  Track the changing dynamics in opening up data sets associated with publications that are stored in a selection of repositories, through the Open Science Barometer.

 $\rightarrow$  Launch a **Europe-wide call for proposals on reusing research data** as part of the European Research Area Network.

#### **Isabelle Blanc**

National Director of Data, Algorithms and Source Codes Ministry of Higher Education, Research and Innovation



# Launch of a new call for projects!



# NEW CALL FOR PROJECTS 2022 HEALTH DATA CHALLENGE

The Ministerial Delegation for eHealth (DNS) and the Health Data Hub are pleased to announce the launch of a new call for projects in 2022!

### For more information, contact us on : data.challenge@health-data-hub.fr



Would you like to be a **project leader** and propose a new **Data Challenge**?



Apply to be part of the upcoming edition of Health Data Challenges !















# **THANK YOU!**





#### **STAY INFORMED!**

Scan the QR code to get the latest informations on our Data Challenges !













