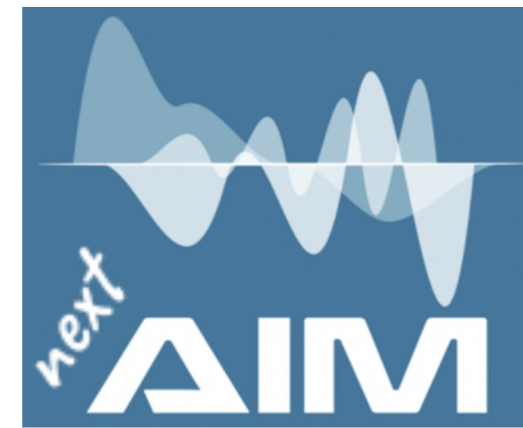


NEXT_AIM: breve descrizione ed obiettivi

[2022-2024]

<https://www.pi.infn.it/aim/>

Artificial Intelligence in Medicine (AIM): *next* steps
focus on *n*o-so-big data and *ex*plainable *t*echniques



next_AIM

Resp. Nazionale: A. Retico

Resp. Locali:

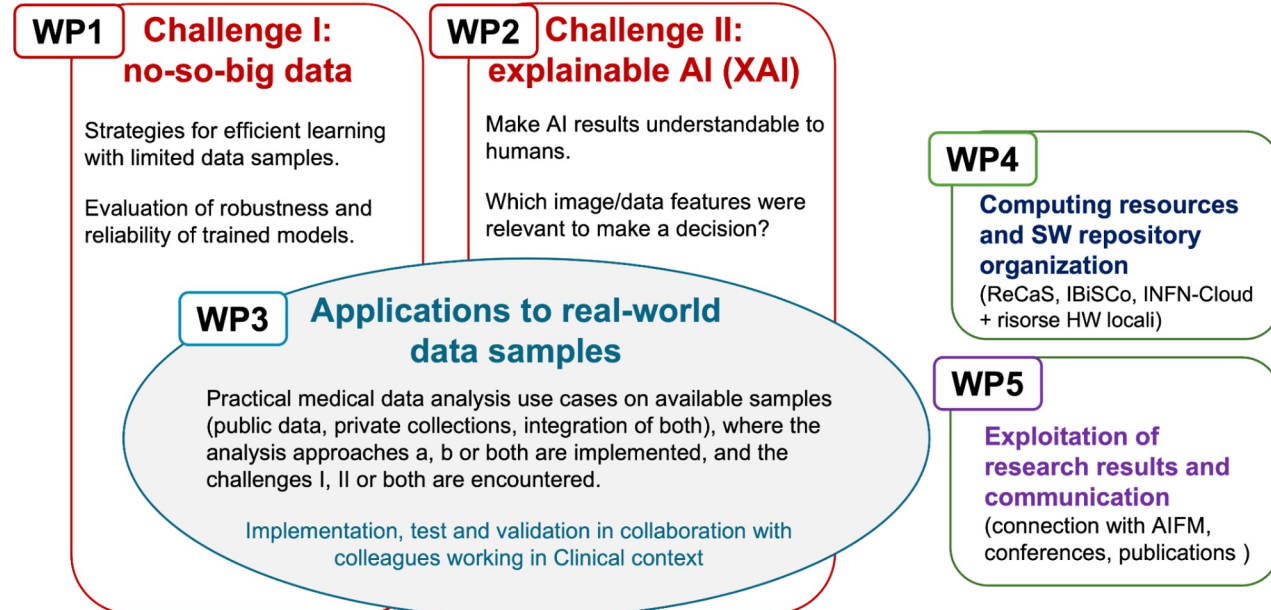
Bari (S. Tangaro)
Bologna (D. Remondini)
Cagliari (P. Oliva)
Catania (M. Marrale)
Ferrara (G. Paternò)
Firenze (C. Talamonti)
Genova (A. Chincarini)
Lab. Naz. Sud (G. Russo)
Milano (C. Lenardi)
Napoli (G. Mettivier)
Pavia (A. Lascialfari)
Pisa (M.E. Fantacci)

+ Padova (Alberto Zucchetta) from 2023

next AIM goal: to take steps towards developing robust and explainable AI algorithms and validating them on realistic use cases in the medical field

A large variety of AI-based algorithms have already been developed to analyse medical images and data. Their potential to improve clinical workflows has not yet been fully exploited due to:

- the lack of model robustness or generalizability
- the lack of transparency



NEXT_AIM: stato di avanzamento e obiettivi 2023



WP3: Real-world study cases

WP conveners: P. Oliva (CA),
A. Lascialfari (PV), M. Marrale (CT)



		FE			NA	PI			Radiomics in Digital Breast Tomosynthesis (DBT)
BO		FE			NA				Super-Resolution in DBT
BO		CT							Radiomics in prostate cancer
BO		CT							Radiomics and DL in tMRgFUS
BO				GE LNS					Nuclear NeuroImaging Quantification and Radiomics
BA	CA	CT				PI		PD	Connectivity in functional MRI
	CA	CT	FE	FI	GE	MI	PI	PV	Radiomics and Deep Learning analysis of CT/RX and patient data in COVID-19
							PI	PV	Radiomics and ML-segmentation on Facio-Scapulo-Humeral dystrophy (FSHD) and lung tumor
BA								PV	ML on Imaging data of 10B uptake tracks and dose monitoring by Compton cameras
			FI				PI		Artificial intelligence for monitoring RT response in soft-tissue sarcomas
		FE						PV PD	Machine Learning techniques for cardiological applications
		FI					PI	PD	Application of NLP techniques to clinical notes towards the automated reading of instrumental data

2022			MILESTONES
31 Dec	M1.1		Identification of methodological pitfalls in case of small datasets
31 Dec	M2.1		Identification of explainability requirements for medical applications
31 Dec	M3.1		Identification of data samples for practical use cases and fist tests
30 Jun	M4.1a		Identification of available resources and usage instructions
31 Dec	M4.1b		SW package release instructions
31 Dec	M5.1		Workshop organization: "WS AI@INFN, Bologna 2-3 Maggio 2022 https://agenda.infn.it/event/29907/
2023			
31 Dec	M1.2		Definition of robust pipelines for efficient model training on small datasets
31 Dec	M2.2		Customization of explainability pipelines to AI models for medical imaging
31 Dec	M3.2		Implementation of robust pipelines and explainability algorithms in at least three different use cases
31 Dec	M4.2		Integration of at least 1 application
31 Dec	M5.2		Workshop organization: "Workshop "AI methods and applications in Medical Physics"
2024			
30 Jun	M2.3		Definition of optimal explainability methodology for medical problems
31 Dec	M3.3		Result evaluation for the practical use case and reporting
31 Dec	M4.3		Integration of all analysis pipelines trained for the use cases of WP3 in the nextAIM SW package repository
31 Dec	M5.3		Submission of at least 1 scientific publication per use case

+

Workshop "The right to explanation"

covidcxr - Hackathon
Artificial Intelligence for Covid-19 prognosis: aiming at accuracy and explainability

<https://ai4covid-hackathon.it/>
Challenge on chest X-ray and clinical data of patients with covid-19 pneumonia

The AIM-WG team achieved the 4th place with 74% accuracy in predicting patient prognosis

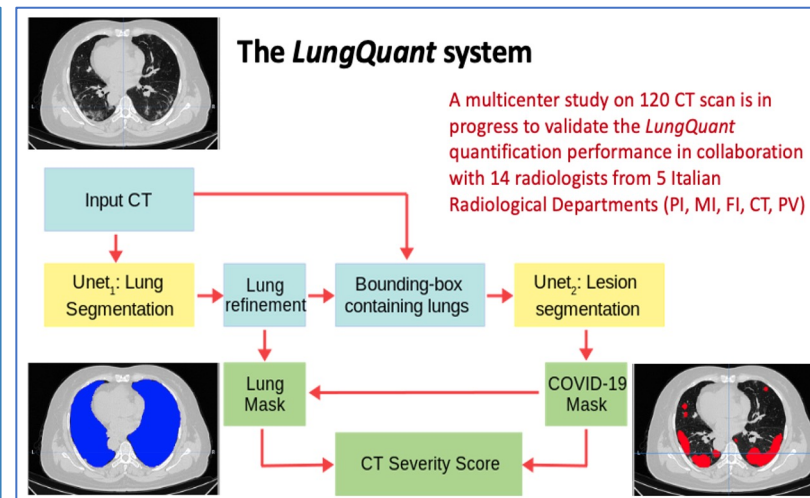
CRX image

Clinical features

- Age
- Sex
- Body Temperature (°C)
- Cough
- Dyspnea
- CRP
- Fibrinogen
- LDH
- D-dimer
- O2
- PaO2
- pH
- Cardiovascular Disease
- Respiratory Failure

Input_1: images
ResidualCNN: 6 residual blocks with 18 convolutional layers (LeakyReLU as activation) + batch normalization Global Average Pooling 2 dense layers

Input_2: features
1 dense layer + dropout
CONCATENATE
5 dense layers Output: sigmoid



rs-fMRI

segnale BOLD per ogni voxel/regione

Studio della correlazioni tra due serie temporali

Misure di connettività

Connessioni funzionali tra diverse aree cerebrali

Case-control discrimination with Deep Learning models vs traditional ones (RF)

Harmon. pipeline	AB I + II		AB I eye = open	
	DNN	RF	DNN	RF
Harmon_k-fold	71±1	66±2	73±1	70±3
Harmon_upstream	74±2	72±1	76±2	71±2
No_harmon	73±3	65±1	72±1	68±3

ABIDE
Autism Brain Imaging Data Exchange

Sinergia delle tematiche con FAIR-AIM (Regione Toscana)

NEXT_AIM: personale @ INFN PI, richieste

Fantacci Maria Evelina	UNIFI	70%
Retico Alessandra	INFN	80%
Lizzi Francesca	INFN assegnista	90%
Scapicchio Camilla	UNIFI dottoranda	100%
Cipriano Emilio	UNIFI dottorando	100%
Biagi Laura	Stella Maris	10%
Bosco Paolo	Stella Maris	20%
Tosetti Michela	Stella Maris	10%
Arezzini Silvia	INFN	10%
Giampaolo Alberto	INFN	10%
Marzoni Enrico	INFN	5%
Laruina Francesco	SNS	50%

5,55 FTE

Missioni	6 kE	Partecipazione workshop, sessioni di lavoro con altri gruppi
Consumo	2 kE	Metabolismo relativo a calcolo e storage
Pubblicazioni	1.5 kE (SJ)	Open access publication fee
Inventario	~	See next slide

NEXT_AIM: richieste sui servizi, licenze e calcolo

Servizi in Sezione

Collaborazione con il personale del centro di Calcolo di Pisa

Licenze

Nessuna richiesta di licenze SW

Calcolo

Richieste di HW da integrare nelle risorse INFN nazionali:

2 x A100 Nvidia GPU (80 GB) $\rightarrow \sim 30$ kE

oppure

2 x H100 Nvidia GPU (80 GB) $\rightarrow \sim 60$ kE

+ Storage $\rightarrow 3$ kE