Clinical and patient requirements for trustworthy Al

Results of co-creation workshops of Trustworthy Artificial Intelligence for Personalised Risk Assessment in Chronic Heart Failure (AI4HF) European Union program

Carina Dantas | SHINE 2Europe

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AI4HF

Trustworthy Artificial Intelligence for Personalised Risk Assessment in Chronic Heart Failure

Horizon Europe project



1 June 2023 → 31 May 2027

AI4HF harnesses the power of **Artificial Intelligence** to co-design, develop, and evaluate a new digital system for **personalized risk assessment and advice** for individuals living with **Chronic Heart Failure**.

It uses advanced AI **algorithms**, global **collaboration**, and a **patient-centered** approach to improve healthcare outcomes.



The problem to solve



The prevalence of Heart Failure (HF) is expected to increase by 46% by 2023 due to an ageing population and increases in unhealthy environments.

The challenge in Heart Failure care and management

- HF is complex as it has many risks, causes and outcomes
- Current medical models for HF follow one-size-fits-all guidelines

The challenge in current AI-based solutions

Existing AI models for predicting HF risk are promising, but far from being used in clinical settings. They often share common limitations:

- Low levels of trust between healthcare providers and patients
- Limited real-world validation of the AI models
- Though studies focus on the accuracy of prediction, few focus on aspects of Trustworthy AI

The AI4HF solution



The foreseen system developed in AI4HF integrates the following features:

<u>Patients & clinicians</u> | a <u>clinical decision support tool</u> to enable shared decision-making throughout the care journey.

<u>Citizens & patients</u> | an educational **information and communication package** for increasing transparency and digital literacy.

<u>Clinicians</u> | a <u>multi-modal AI risk assessment tool</u> for predicting personalised outcomes in HF patients and improving quality of care.

<u>Researchers</u> | an **AI traceability technology** to help researchers to effectively monitor and adjust the AI tools over time.



Partners



ABOUT US



PROJECTS

SHINE is a Portuguese SME - applied research promoting inclusive communities for all citizens.















Policy





Gamification

CHAIR COST ACTION



50 countries | +775 members

DIRECTOR





COORDINATOR



HORIZON EUROPE

Social Innovation **Responsive Environments** NETwork

What do we do at SHINE?



3 KEY AREAS

RESEARCH

+ Knowledge

New approaches – **social innovation**

Multidisciplinary networks

POLICY

Local ecosystems

Evidence-based recommendations

Translate **research - policy** priorities

PEOPLE

Reliable information

Good tools

Promote **empowerment** and engagement

A model that can be applied by design, combining:



MULTIPLE STAKEHOLDERS

PARTICIPATION, CO-CREATION, COMMITMENT & COLLABORATION





PARTICIPATORY ETHICS

ASSESSING SOCIETAL CHALLENGES AND DISCUSSING POTENTIAL "SOLUTIONS"

Trustworthy A.I.

PRODUCTS & SERVICES

PROMOTE EMPOWERMENT, INCLUSIVENESS AND PARTICIPATION





OWNERSHIP

INCREASING LITERACY, EDUCATION, NETWORKING & GOOD PRACTICES

Clustering ETHICAL CHALLENGES on Al applied to health





INDIVIDUAL

Low health and Al literacy of citizens

Lack of adequate training on AI for healthcare professionals

Personal convictions that lead to human biases

Lack of trust on AI tools

TOUCHPOINTS

Education; training; prejudice; discrimination; trust



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TECHNICAL

Lack of scientific validation of the algorithms

Lack of transparency and accuracy of algorithms and their predictions

Lack of diversity in data, namely on underserved populations

Lack of ethnicity-related data in datasets

TOUCHPOINTS

Algorithm design; training datasets; interpretation of results



ORGANISATIONAL

Lack of integration of new tools with existing clinical pathways & solutions

Low diversity in working teams – on disciplines, culture, age, gender

Lack of audit methods and tools to support organisations implementing ethical workflows

TOUCHPOINTS

Diverse teams; audit; ethics-bydesign; supporting methods and tools

SOCIETAL

Lack of clarity on professional liability with the use of Al

Absence of studies analysing longterm impact of AI use in health

Increase of inequalities due to financial reasons

Need of cohesive regulations

TOUCHPOINTS

health inequities, ethical and legal; regulatory framework; policy measures



WP1: Multi-stakeholder engagement and social innovation (SHINE)



- Develop a social innovation framework to engage relevant stakeholders, including cardiologists, patients, AI technologists, data/IT managers, social scientists, policymakers and regulatory experts.
- Leverage the social innovation framework to identify multi-disciplinary needs, requirements, obstacles and implementation pathways for AI4HF's real-world adoption.
- Translate the multi-stakeholder requirements into a set of DESIGNS, PROCEDURES AND SOLUTIONS
 based on the FUTURE-AI guidelines for subsequent trustworthy AI implementation and evaluation.

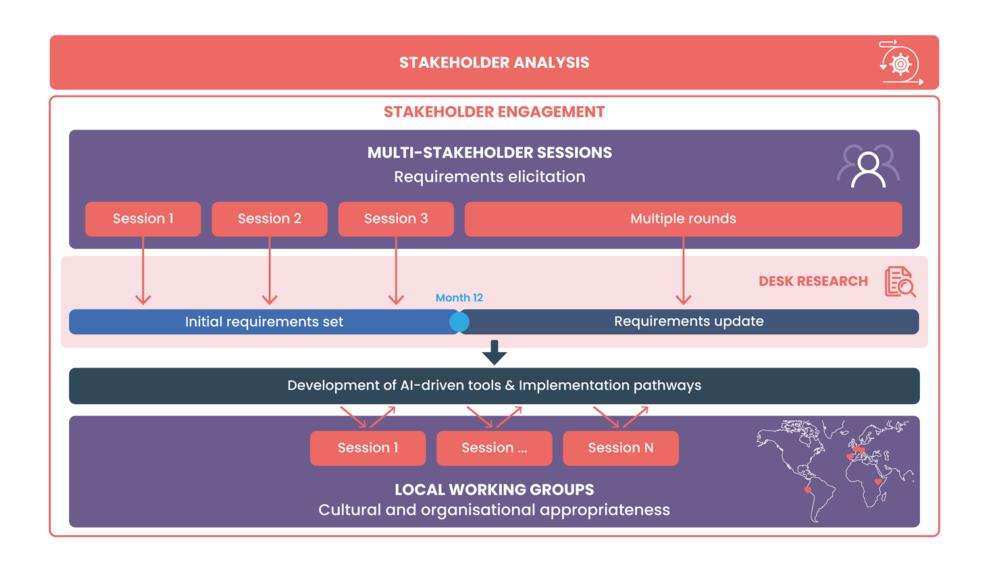






Social Innovation Framework in AI4HF





Multi-stakeholder requirements



- 1. Stakeholder mapping
- 2. Multi-stakeholder workshops
- 3. Local Working Groups Clinical/Patient
- **4. ELSI** requirements
 - 1. Literature review: alignment of AI4HF with key ethical principles
 - 2. Literature review: patient reidentification risks
 - 3. Desk research: accountability of Al-driven solutions in healthcare
- 5. Healthcare and **regulatory** requirements
 - 1. Interviews with HTA experts
 - 2. Review of relevant regulations
- 6. Requirements specification, including aligning to the FUTURE-Al guidelines

Stakeholder Analysis | an example of a Stakeholder Map





LISTENER

The person in this role is given information.



THINKER

The person in this role thinks along with the research team.



DECISION-MAKER

The person in this role takes the initiative and/or makes the necessary (final) decisions. Key player.

	Steps of Al development process according to the FUTURE-Al guidelines									
Stakeholder Ethicists, regulators	Clinical conceptualisati on	End-user requirements gathering	Technical Design	Data selection, collection and/ or preparation	Al implementation and optimisation	AI evaluation	Al deployment	Al monitoring		
1. AI Ethicists			® }	(a)	<u></u>		(a)			
2. Legal consultants	T			F 195	<u></u>	S CONTRACTOR OF THE PARTY OF TH	F TO	FINAL PROPERTY OF THE PROPERTY		
3. Human rights advocates	(4)						<u></u>	\(\frac{1}{2}\)		
4. European Health Management Association						STANKS	<i>\(\mathbb{H}\)</i>	₹ %		
5. European Committee for Standardization	©		(₹ / %	FMS.		
6. National Medicine and Medical Devices Agencies				•	<u></u>		⊠ €	88 (2)		
7. Data protection authorities		7	7 %		(a)	(a)	® C			
9. Social scientists	(P)	(P)					(B)			
9. HTA experts						<u></u>				

Multi-stakeholder sessions



M3 Workshops: Clinical Conceptualization

On-site meeting in Brussels with 14 patients

On-site meeting in Tanzania with 22 patients

On-site meeting in Peru with 15 patients

Series of online meetings with healthcare professionals from AI4HF consortium



Online meeting with patients, healthcare professionals, and social scientists (17 participants)

M12 Workshop: Trustworthy Al

On-site meeting in Lisbon (Portugal) with patients, healthcare professionals, social scientists, and technologists (21 external participants)

ELSI REQUIREMENTS

HEALTHCARE AND REGULATORY REQ



PART I		Phase I - Pre-diagnosis		Phese II - Diagnosis	Phase III - Monagement	Phase IV - Hospitalization
Patient journey of a patient with Heart failure	Primary care Entry point through primary care	Secondary Care Entry point through medical specialist	Tertiary Care Entry point through hospital ization	Diagnosis What happers at diagnosis?	Regular follow-ups What happens during regular follow-up?	Hospitalization after diagnosis What happers from the moment of troughs alberton with declarated.
Steps Write down a short story of what is happening at each step.	cases are after diagno confirm dia	hospitalisation can be entry point, most often via GP (not- academic hospital) referral chronic HF GP or consultation (internal medicine/respiratory care) v complexity referred to sis. rule-out/ agnosis and gement	hospitalisation can be entry point, most often via GP (academic hospital) sometimes second opinion or referral chronic HP GP or consultation (internal medicine) respiratory care)	Clinical work-up echocardiogram, lab (BNP/nt- pro-BNP), ECG. Signs and symptoms of heartfailure, patient history, complaints.	Cesofication dumy follow-up (phenotypimy, reduced, mid-y-reduced, preserved EF), exceleres platter overlaat inclusives; limitare guidal her directed medical therapy (EV) guidates, but more may be local variation at the doctors in guidaline advanced, same here for the heart soon and cause of HF as during hospitalization.	medication optimization, reduce volume overload (IV-drip or oral medication) and contracterization lab (intervention) healt team discussion if it is caused by volume disease, ischemic (revescularization) or other causes
People involved Write down all the actors that are involved in this step, if there is more than one circlain involved please specify the type of circusters.	General practicioner 1.5 care can (cardiologist as consultation partner fo. GP), also as video consultation	cardiologist, nurses	cardiologists, nurses	cardiologists	cardiologist, nurses, cardio-surgeons	cardiologists, nurses, cardio-surgeons
Where does this step take place? Write down of physical locations where this cop can take place. E.g., healthcare center, hospital, OP clinic, etc.	GP office, at-home video consults, GP- cardio video consults	Hospital	Hospital	Hospital (both secondary and tertiary) or 1.5 care units	hospital	hospital
Time duration How much time does this step soully takes until the patient moves to the next step?	Within the se (1.5 pert		Within six weeks	[How much time does this step usually takes until the patient moves to the next step?]	[How much time does this step usually takes until the patient moves to the next step?]	[How much time does this step usually takes until the patient moves to the next step?]
Use of Al-driven prediction tools White the name of the tool, good postes and bed pairs. Please mention tools that you currently use, have used in the part, or that you heard about.	(Produce points of the best) [Stopping points of the best] [Stopping points of the best]	SCIENCE (Pail Processed of State of Sta	(Page page of foreign Image page of foreign (Page of page of foreign Image page of foreign (Page of page of foreign Image page page of foreign (Page of foreign Image page page of foreign (Page of foreign Image page page (Page of foreign Image page page (Page of foreign Image page page page page page page page p	room and in those Plantine context Plantine context	(Section of Chick parties of Section of Sect	Flore and include Produce paint of Sequence paint of Sequenc



Summary Results Workshop: "Towards trustworthy AI-driven tools for personalized treatment of heart failure" | Lisbon, 13th of May 2024

The results of this workshop led to 22 new stakeholder requirements.

Alignment of AI HF solutions with key ethical principles

Umbrella review with 25 articles until September 2023



Right to Autonomy

 Al systems should support, not replace, human judgment. Clinicians must remain central to decisionmaking, preserving empathy and managing complex conditions. Establishing clear ethical standards and accountability mechanisms will protect patient autonomy.

Information privacy

 Tackle ethical concerns about data sharing and potential biases in Al algorithms. Develop a code of conduct and comply with GDPR guidelines to protect patient privacy. Ensure Al systems are explainable to avoid hidden biases.

Confidentiality

 Address privacy concerns in remote monitoring and data transmission. Implement robust data security measures, such as encryption and blockchain technology.

Equal treatment

 Prevent discrimination from biased training data. Implement transparent accountability for AI errors. Techniques like semisupervised and federated learning can enhance data representativeness and collaboration. More randomized controlled trials will evaluate clinical suitability of ML systems.

The results of the desk research on key ethical principles led to 15 new requirements.

Patient Reidentification Risks





- Attack models and solutions preventing reidentification found across numerous healthcare sub-fields and application domains. However,
- No solution identified fully protects patients from reidentification
- Encouraging multi-modality in privacy-enhancing frameworks can lead to decreased reidentification risks
- A large shift towards synthetic data processing algorithms is observed
 - Requires novel solutions and (evaluation) measures
- FEDERATED LEARNING is not vulnerable to patient reidentification attacks
 - \circ Aims to mitigate such attacks \rightarrow can be recommended to be used for mitigating privacy issues within AI4HF

Accountability of Al-driven solutions in healthcare



- 3 types of liability: fault-base liability, strict liability, and product liability
- Relevant legislative initiatives:
 - EU Artificial Intelligence Act (Al Act)
 - Al Liability Directive (AILD)
 - Product Liability Directive (PLD).
- Questions discussed with stakeholders:
 - How can **responsibility be clearly assigned** in the complex ecosystem of AI-driven healthcare? What legal reforms are necessary to address this issue?
 - How can regulatory frameworks be designed to promote innovation while ensuring safety and accountability?
 - What standards should be established for the transparency of AI algorithms in healthcare? How can these standards be enforced?
 - What strategies are most effective in building public trust in Al-driven treatments?
 - How can educational initiatives be designed to enhance understanding and acceptance?





Interviews with HTA experts: towards healthcare and regulatory requirements



Interviews with 6 experts in Health Technology Assessment regarding key issues such as:

- Implementation
- Reimbursement
- Potential barriers
- Ensure HTA requirements are met

The results of the interviews with HTA experts led to 15 new requirements.



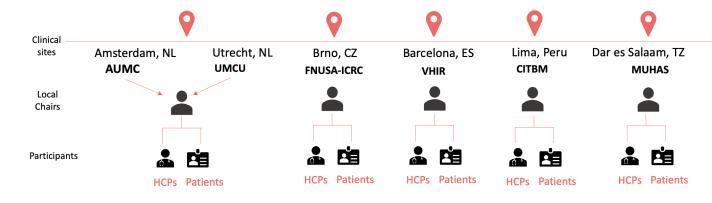
SHINE DEVELOPED

- 1. Workshop protocol with step-by-step guide
- 2. Workshop planning checklist
- 3. Reporting template

Provided in EN and translated to local languages:

- Invitation letter
- 2. Informed consent
- 3. Workshop presentation
- 4. Material for participants during the workshop

Local Clinical/Patient WGs





First Local WG workshops: Spring 2024



Highlights

- Patients and healthcare professionals involved have good access to technology.
- Neither healthcare professionals (NL), nor patients (CZ) see the need for yet another risk prediction tool.
- Patients highlight the wish for support in the daily management of the disease (CZ) and choosing best treatment (ES).
- Need for personalized solutions (no "typical" patient, NL).
- High <u>lack of trust in Al</u> from patients. Need for Al literacy (PE).
- Positive experience from patients and HCP in participating in the local groups.



Workshop on the C&I Package



Date | 5 December 2023, online meeting | 7 EHN patient consultants (from 4 European countries), 5 project partners

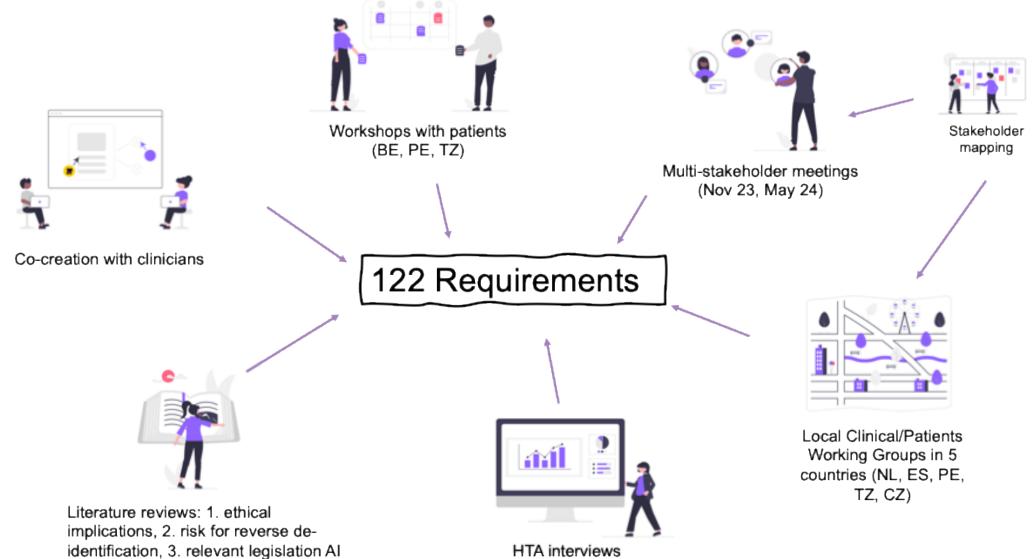
Knowledge Gap	Information Source	Trustworthiness
 5/7 did personal research right after HF diagnosis Helpful info about HF: symptoms, recommendations for selfmanagement, psychological impact and support What people should know about living with HF: it is an invisible disability, physical limitations, can still live well 6/7 did not know that AI could be used in HF care when first diagnosed Discussed what they would like to know about the use of AI in HF and exiting knowledge gaps for patients in this area. 	 7/7 would prefer information disseminated by cardiologist, followed by patient organisations (5/7) and other healthcare professions (4/7) Shared different online sources for accessing information (i.e: chat GPT, medical journals, ESC, patient organisation website) The consensus was that multiple formats for disseminating information to patients/the public is needed (i.e. video, infographics, text, podcast). 	 Majority would trust to learn this information from a healthcare professional, however, patient-led support groups and patient organisations were also described as trusted and important sources Mixed response on how much information they would like to know about if AI were integrated into their care – suggestion to have varying levels of information All respondents believed that there are differences between EU countries in access to trustworthy information on this topic.

Stakeholder engagement M1-M15



	M3 Workshops						Local WGs: workshop #1						
	International *	Brussels	Peru	Tanzania	M9 Workshop*	M12 onsite	NL	CZ	ES	PE	TZ	Interviews*	Total
Patients & caregivers	0	14	15	22	5	6	2	5	9	5	9	1	93
Healthcare professionals	9	0	0	7	3	9	4	5	4	6	7	0	54
Ethicists, regulators	0	0	0	0	9	4	0	0	0	0	0	5	18
AI developers and industry	0	0	0	0	0	2	0	0	0	0	0	0	2
Policy-makers / health authorities	0	0	0	0	0	0	0	0	0	0	0	1	1
Hospital administration	0	0	0	0	0	0	0	0	1	0	3	0	4
Payors	0	0	0	0	0	0	0	0	0	0	0	0	0
Total	9	14	15	29	17	21	6	10	14	11	19	7	172





in healthcare

Mapping of the requirements to the FUTURE-Al guidelines



ID	▼ Type	Requirement	Rationale	Corresponding FUTURE-AI princip	Corresponding (prospective) system requirement aligned with the select principle	Source
SI.17	StakeReq		The patient data can be used for different purposes. Patients must be well informed of this. In this line, we should separate informed consent (related to treatment) from data protection consent (data for research).	TRACEABILITY 6	The system shall request a data protection consent from the patient prior to processing patient's data. The system shall process the patient's data only if such a consent is obtained.	SI-Workshop-M6-bre akout-ELSI
SI.8	StakeReq	Patients do not want that Al-tools are the sole item on which clinical decision making is based.	Fear mentioned by a patient after the M6 workshop: "Implementing AI may shape and narrow the treatment options available to the patient, there will be no outside-of-the-box thinking as that will not be part of the AI programming."	USABILITY 1	Same as for SI.3: The system shall provide the given clinician and patient, as part of the output, with a notification that the system is only used as an auxiliary tool in the decision-making process and that the clinician's decision does not only rely on the system's output	SI-Workshop-M6-followup-patients
SI.19	StakeReq	Clinicians want that the clinical Al-tool is integrated within a specific clinical process based on the intended-use of the novel algorithm.	The use and application of the clinical AI tool should fit within specific work-flow and provide the necessary risk-prediction at required point of care (real-time versus non-realtime). In this sense, the risk-prediction is provided at the time-instance in the care pathway when it is required.	USABILITY 1	The system shall provide the risk prediction that is coherent with the actual state of the patient.	SI-Workshop-M6-bre akout-clinicians
LR.13	StakeReq	The Al tool shall undergo rigorous validation and refinement processes to enhance its predictive performance in comparison to statistical models.	ML models did not achieve a significant advantage in predicting events, and their clinical feasibility and reliability were worse when compared with statistical models	TRACEABILITY 4	The system's accuracy shall be compared with that of the state-of-the-art applications that only make use of statistical predictive models.	Desk research

Horizon Europe EU funding for Trustworthy AI



Trustworthy AI is lawful, ethical, and technically robust. It is when trust in AI models can be established in each stage of its lifecycle, from design to development, deployment and use.

AI4HF uses the **FUTURE-AI guidelines**, which were developed based on <u>6 guiding principles</u>:









FAIRNESS

Keeps the same quality of performance for different individuals and populations developed to minimise potential bias

UNIVERSALITY

Can be successfully used in settings outsi de the study environment

TRACEABILITY

Developed with measures for documenting and monitoring the tool from development to use

USABILITY

End users should be able to use the tool efficiently, easily and safely in real-world settings

ROBUSTNESS

The tool can maintain the same performance and accuracy when there are unexpected changes in the data it receives.

EXPLAINABILITY

Should provide clinically useful information about the logic behind the Al decisions it makes (i.e. no hidden processes)

Analysing challenges in AI for health



FUTURE-AI guidelines

Fairness

Universality

Traceability

Usability

Robustness

Explainability

ALTAI GDPR

Human agency and oversight

Data privacy

SOCIAL INNOVATION

Pathways to implementation



THANK YOU!

