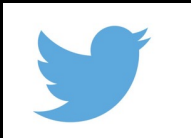


AI for health “ethics by design”

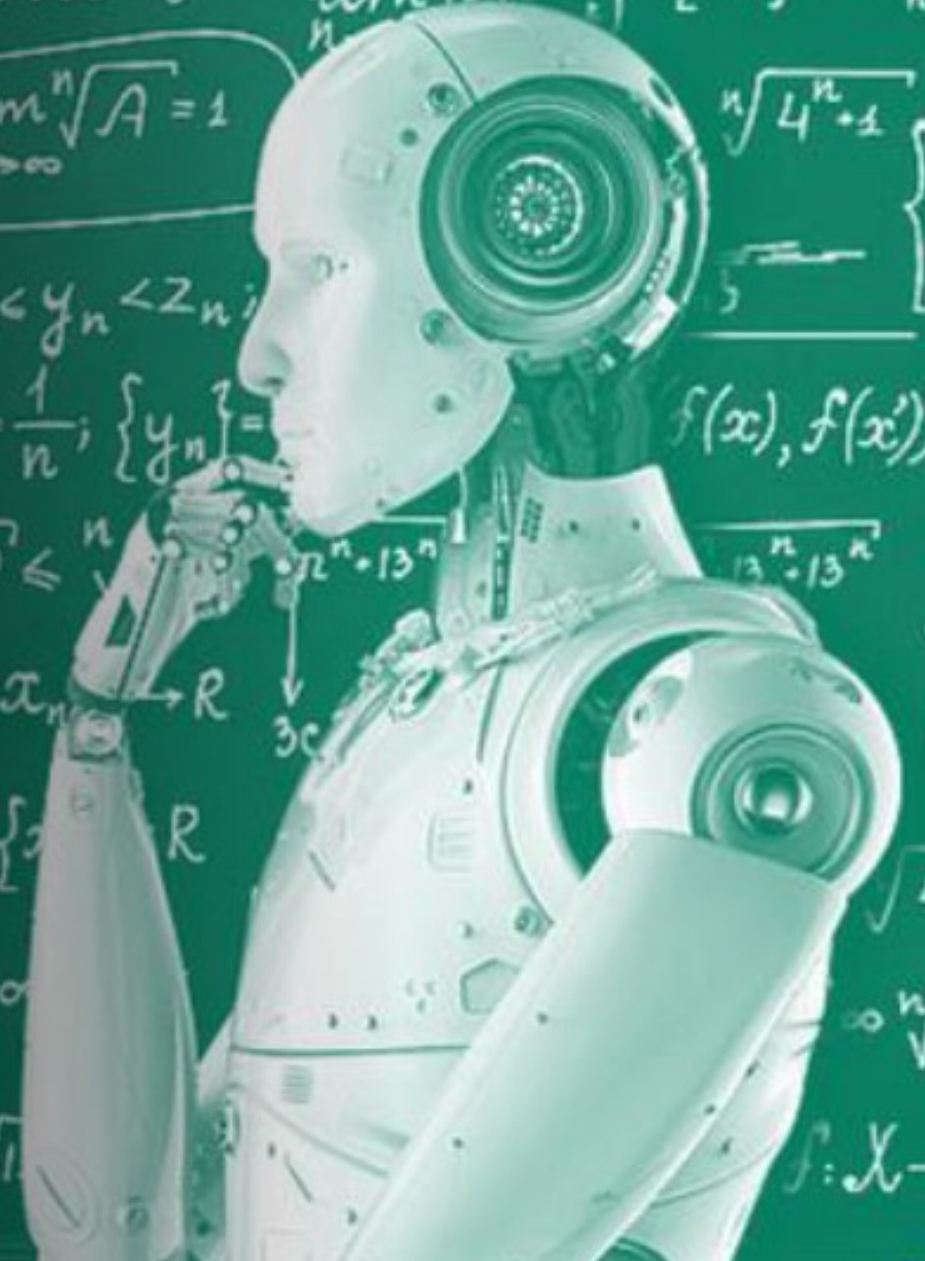
Baptiste Couvy-Duchesne
The University of Queensland



baptisteCD



@BaptisteCouv



Ethics recommendations

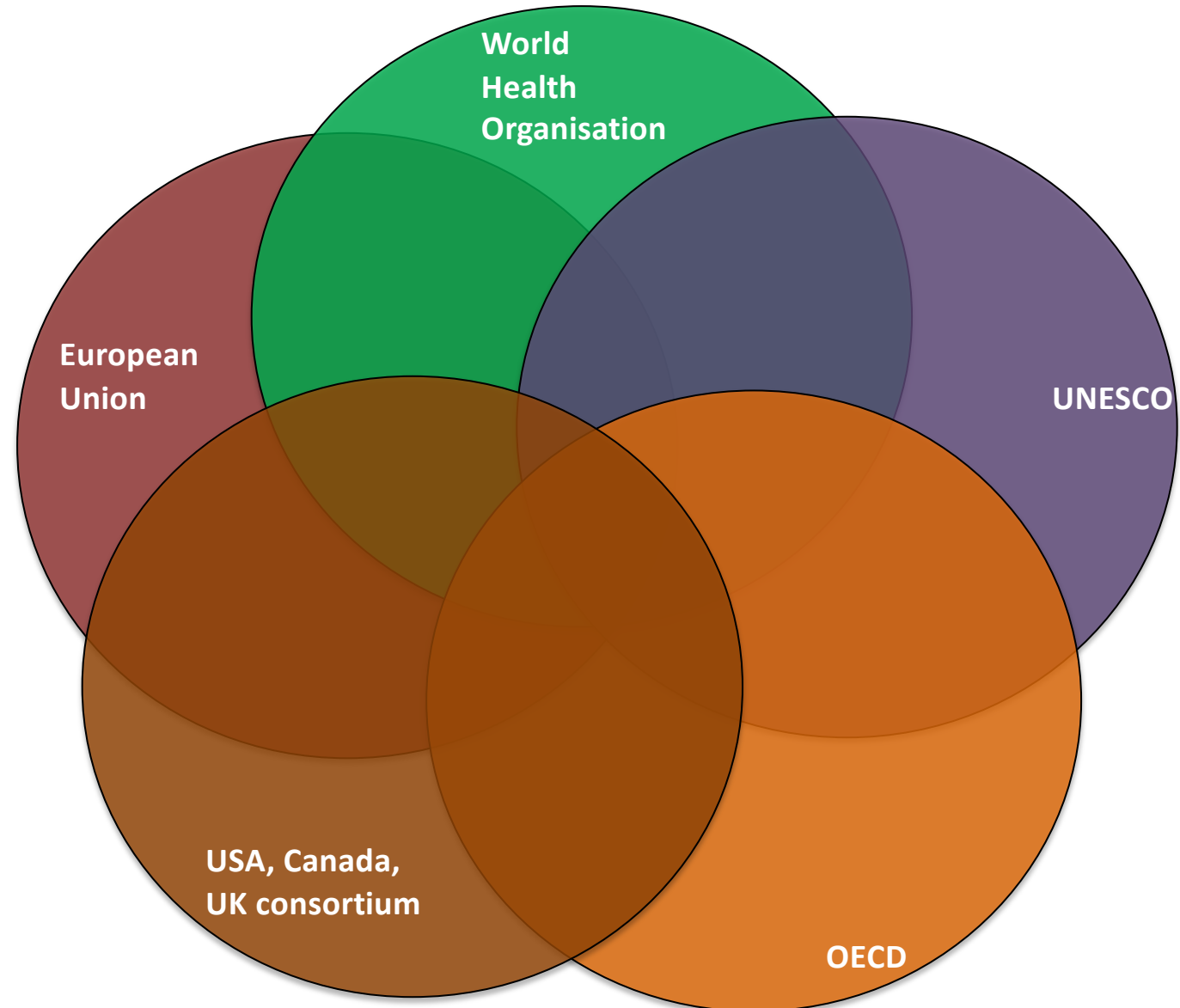
<https://op.europa.eu/fr/publication-detail/-/publication/d3988569-0434-11ea-8c1f-01aa75ed71a>

<https://apps.who.int/iris/rest/bitstreams/1352854/retrieve>

<https://fr.unesco.org/artificial-intelligence/ethics#recommandation>

<https://oecd.ai/en/ai-principles>

<https://www.fda.gov/media/153486/download>





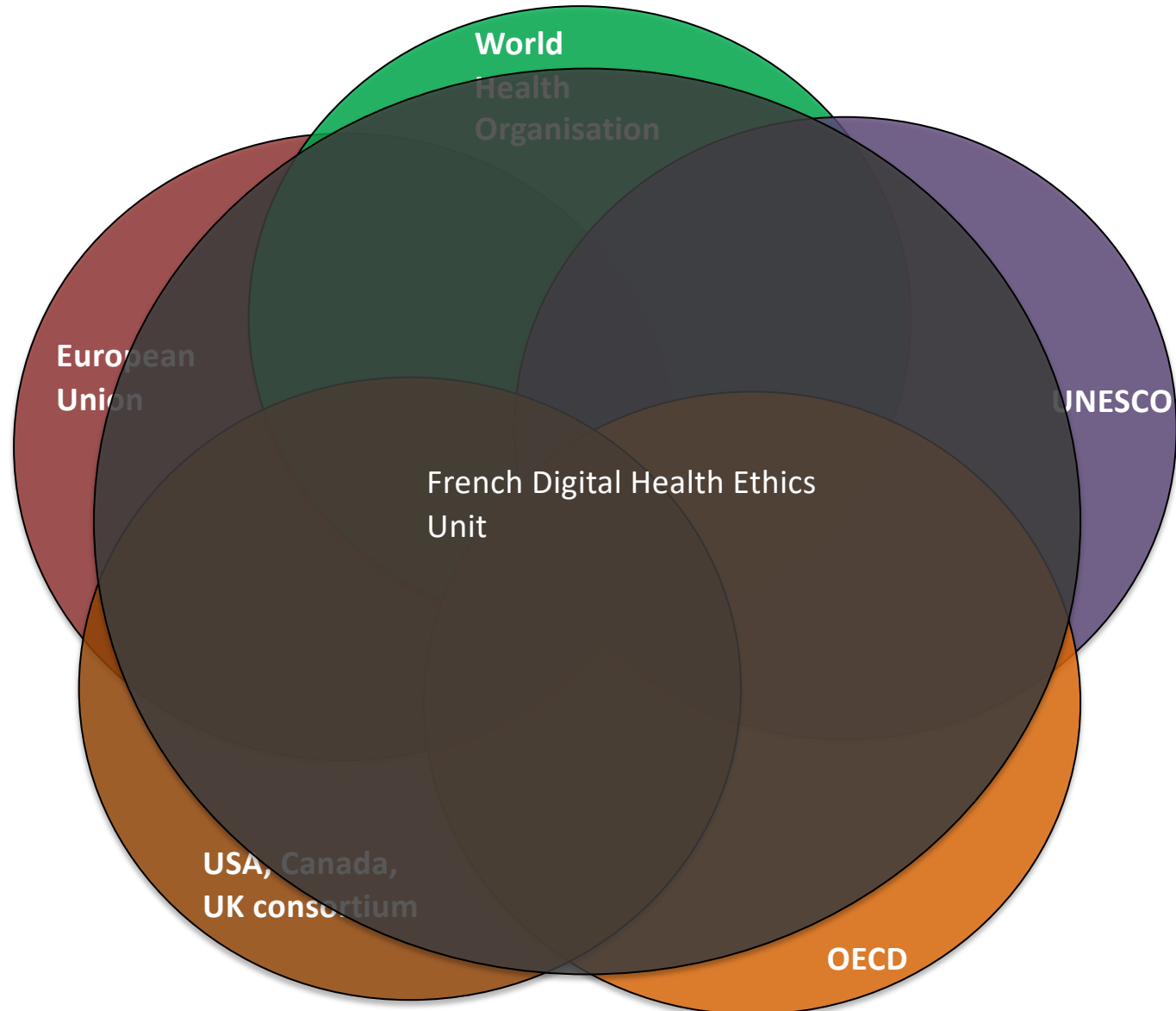
MINISTÈRE
DES SOLIDARITÉS
ET DE LA SANTÉ

*Liberté
Égalité
Fraternité*

Recommendations for good practice to integrate ethics from the development stage of Artificial Intelligence solutions in Health :
Implementing "ethics by design"

Digital Health Ethics Unit of the French Ministerial Delegation for Digital Health
– April 2022

https://esante.gouv.fr/sites/default/files/media_entity/documents/ethic_by_design_guide_vf.pdf



Definition of AI



- AI aims at developing **devices, hardware and software,**
- **capable of implementing a tool**
- **aiming at producing the same result as the one obtained by the cognitive mechanisms of a human expert engaged in a problem-solving task,**
- **for the purpose of assisting or replacing human activities.**



Ethics by design

- By Design means:
 - By and from conception
 - Intentionally, consciously and premeditated
- Core ethical values:
 - Protection of personal data
 - Human guarantee of any high-stake decision
 - Thinking about the action, its legitimacy and its consequences



Ethics by design – steps

Framing stage
Define the purpose of the AI solution and validate the ethics of the purpose

Step 1 –
Data collection

Step 2 –
Data pre-treatment

Step 3 –
Build the algorithm

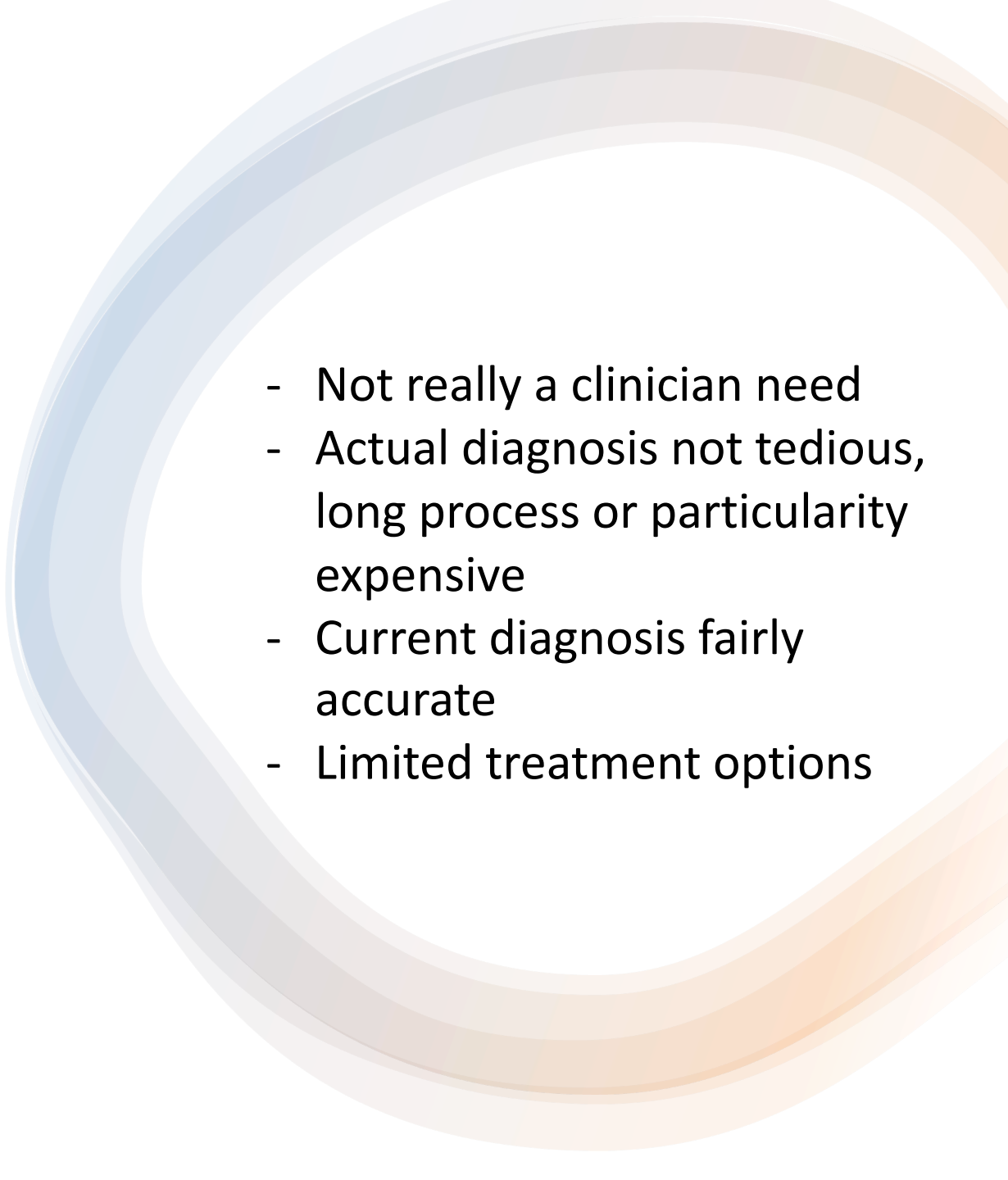
Step 4
Evaluation of the algorithm and preparation of production



Framing stage : Define the purpose of the AI solution solution and validate the ethics of the purpose

Possible use-cases of Alzheimer's disease (risk) prediction

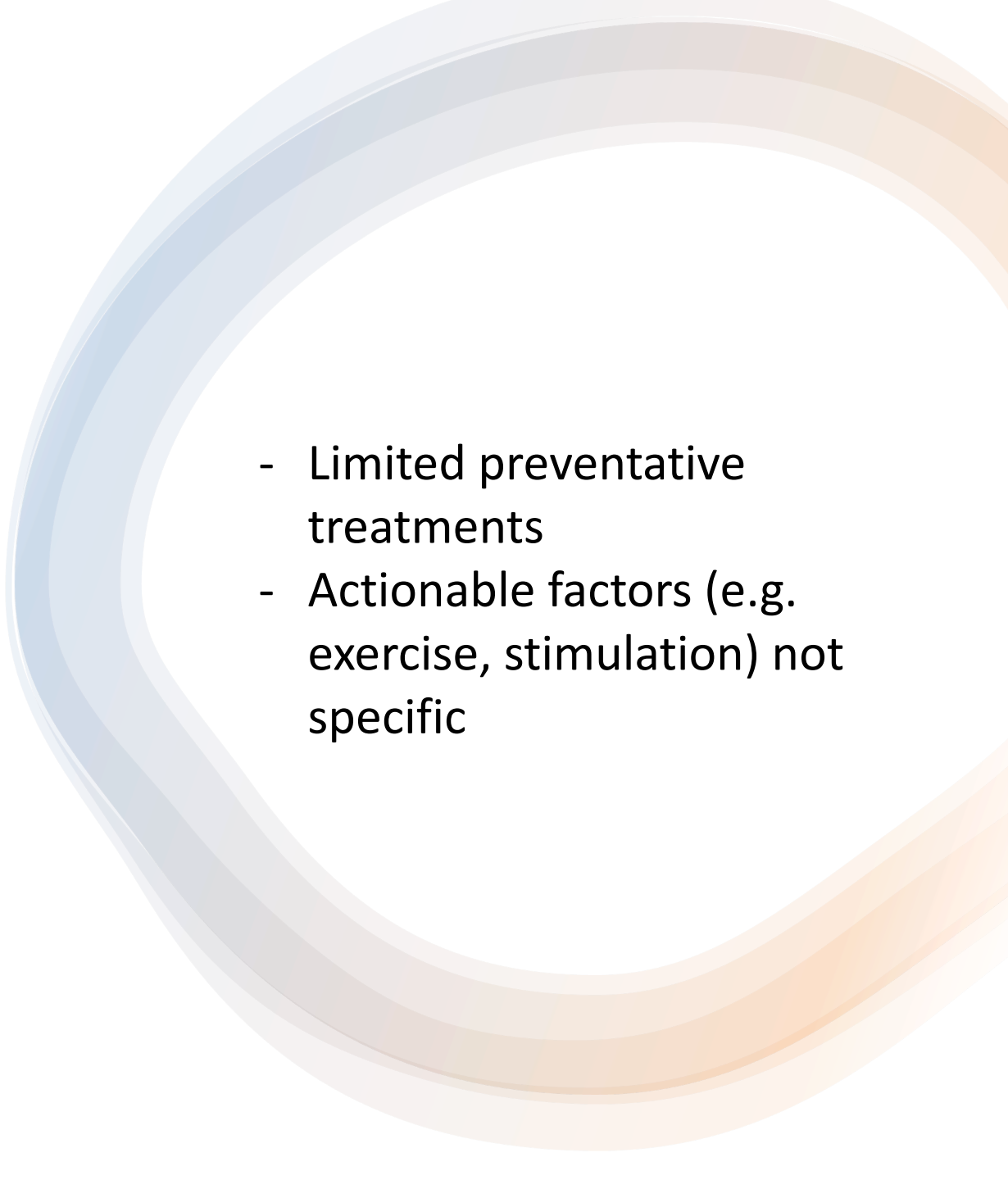
- Diagnostic

- 
- Not really a clinician need
 - Actual diagnosis not tedious, long process or particularity expensive
 - Current diagnosis fairly accurate
 - Limited treatment options

Framing stage : Define the purpose of the AI solution and validate the ethics of the purpose

Possible use-cases of Alzheimer's disease (risk) prediction

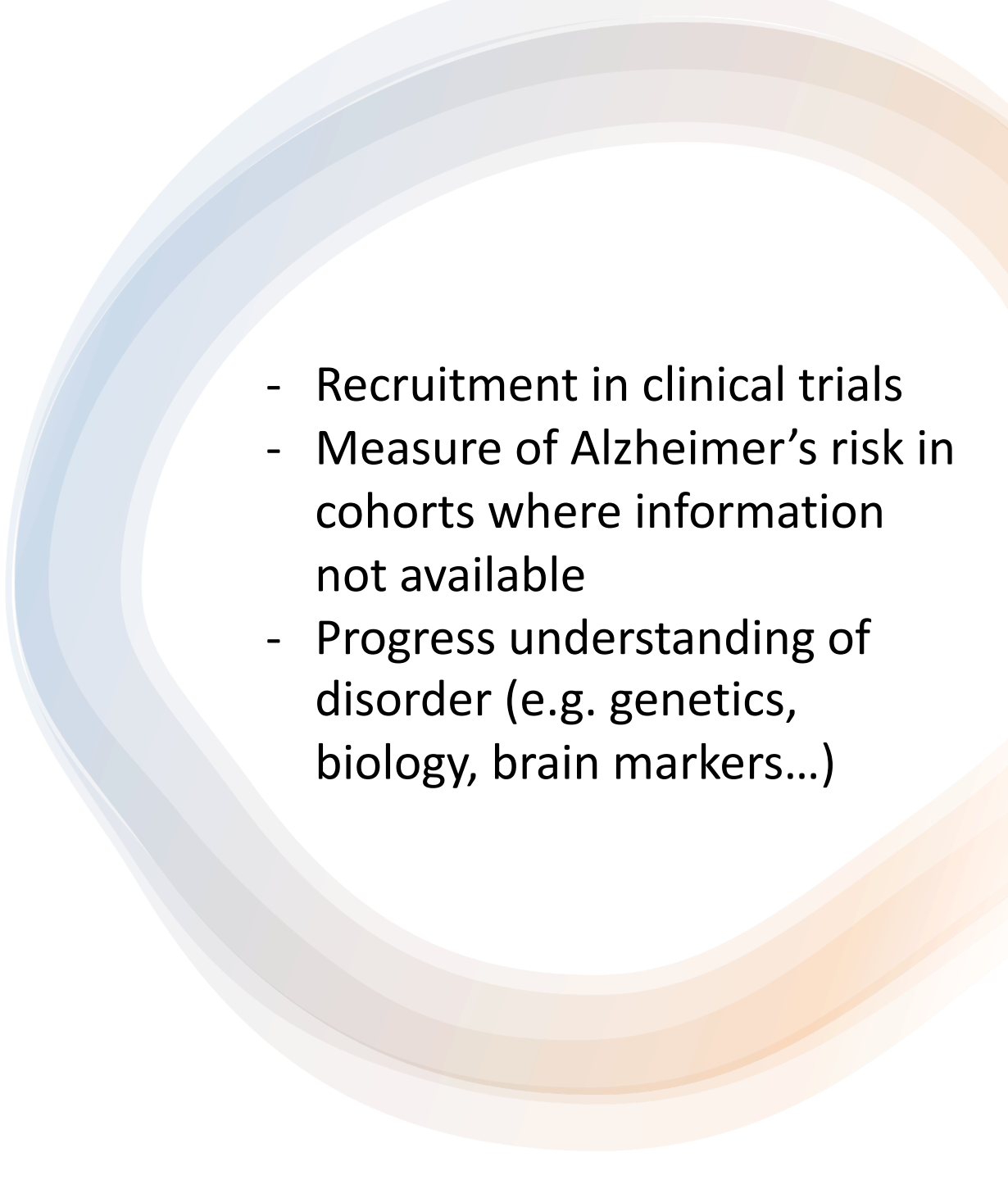
- Prognostic

- 
- Limited preventative treatments
 - Actionable factors (e.g. exercise, stimulation) not specific

Framing stage : Define the purpose of the AI solution and validate the ethics of the purpose

Possible use-cases of Alzheimer's disease (risk) prediction

- Quantify risk
- Research

- 
- Recruitment in clinical trials
 - Measure of Alzheimer's risk in cohorts where information not available
 - Progress understanding of disorder (e.g. genetics, biology, brain markers...)

Framing stage : Define the purpose of the AI solution solution and validate the ethics of the purpose

Purpose of the AI solution.

For example

- Quantify the genetic risk of Alzheimer's disease, to study the link between Alzheimer's risk and Major Depressive Disorder.

Target users

- To use in research.

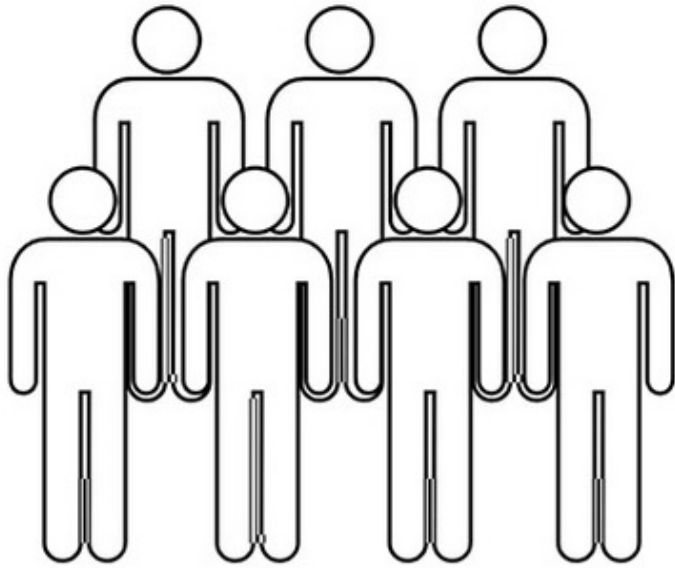
Type of learning used

- We will build a linear predictor using the latest Genome Wide Association Study to date.

The subjects for which it will be used

- We will apply the risk predictor in a local cohort of (mostly European) adults (age range 20-40) who have been screened for psychiatric disorders

Step 1 – Data collection



Informed consent

Proportionality of the data collected

Non direct re-identification of data: pseudoanonymization

Quality of the data

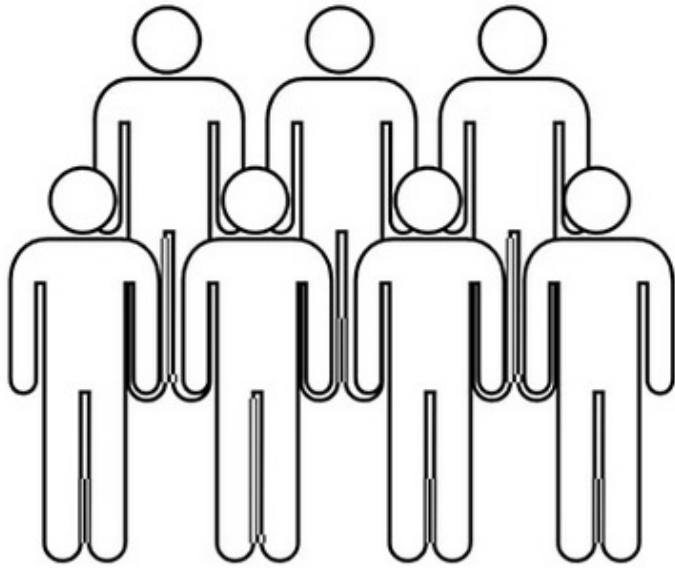
Representativeness of the analysis population/ target population + prevention of discrimination

User involvement

Secure data transfer +
Quality of data hosting +
State of the art cyber security

Measures to ensure non-ethical reuse of data

Step 1 – Data collection



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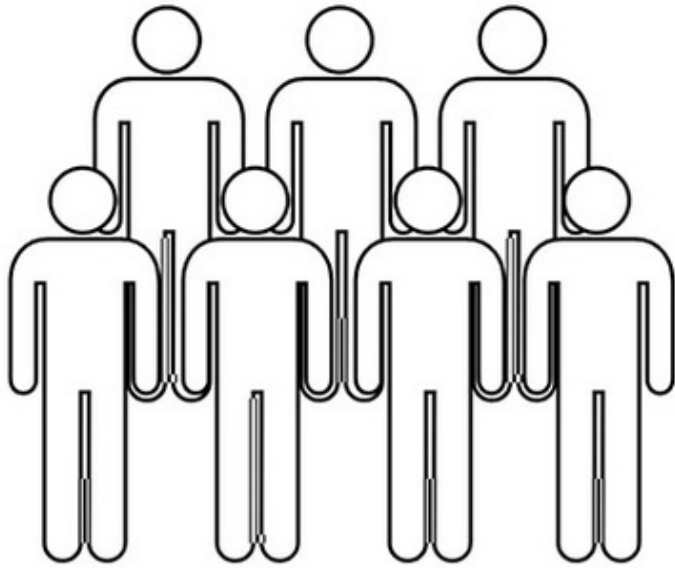
Secure data transfer + Quality of data hosting + State of the art cyber security

Measures to ensure non-ethical reuse of data

Required size of training sample

Power analysis
=> Sample size required to detect hypothesised association in local cohort

Step 1 – Data collection



Informed consent

Proportionality of the data collected

Non direct re-identification of data: pseudoanonymization

Quality of the data

Representativeness of the analysis population/ target population + prevention of discrimination

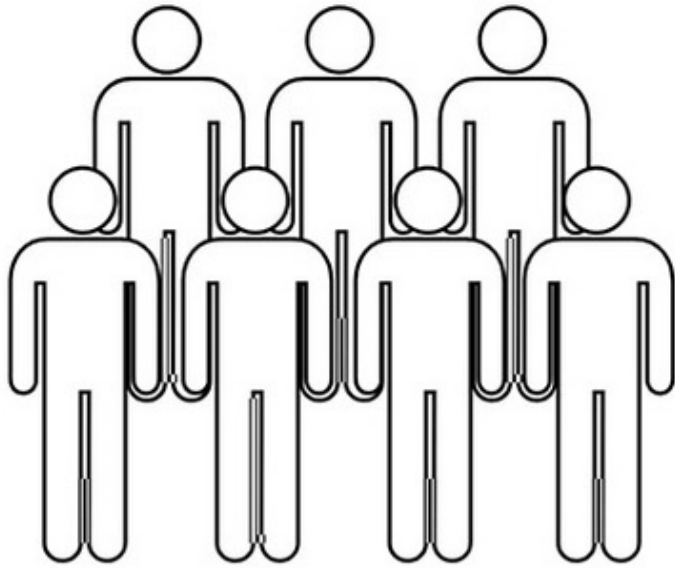
User involvement

Secure data transfer +
Quality of data hosting +
State of the art cyber security

Measures to ensure non-ethical reuse of data

- Recruit participants from **European** as well as **diverse ancestry present in Australia**
- Recruit **males and females**

Step 1 – Data collection



Informed consent

Proportionality of the data collected

Non direct re-identification of data: pseudoanonymization

Quality of the data

Representativeness of the analysis population/ target population + prevention of discrimination

User involvement

Secure data transfer +
Quality of data hosting +
State of the art cyber security

Measures to ensure non-ethical reuse of data

Meetings with people involved in project:
researchers,
geneticists,
clinicians...

Step 2 – data pre-treatment



Handling missing
data (bias
reduction)

Data segregation
(representativeness
of the learning and
evaluation sample)

Rebalancing of
minority
populations (bias
reduction)

Expert and users
involvements

Missingness in **MDD
diagnosis** (bias,
stigma?)

Step 2 – data pre-treatment



Handling missing
data (bias
reduction)

Data segregation
(representativeness
of the learning and
evaluation sample)

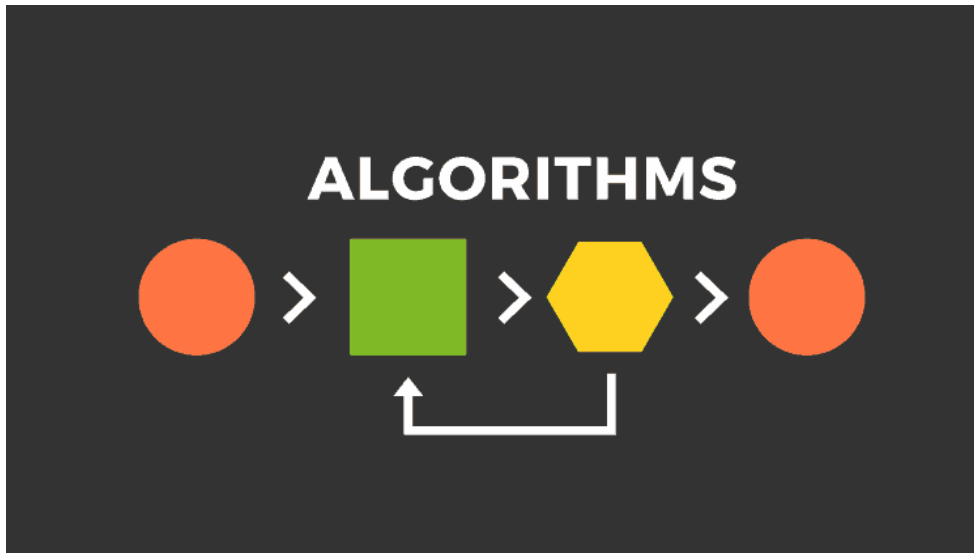
Rebalancing of
minority
populations (bias
reduction)

Expert and users
involvements

Include **covariates** in
analysis (age, sex,
site, MRI machine)

**Oversampling in
training**

Step 3 – Algorithm selection



Quality policy

Transparency
measures

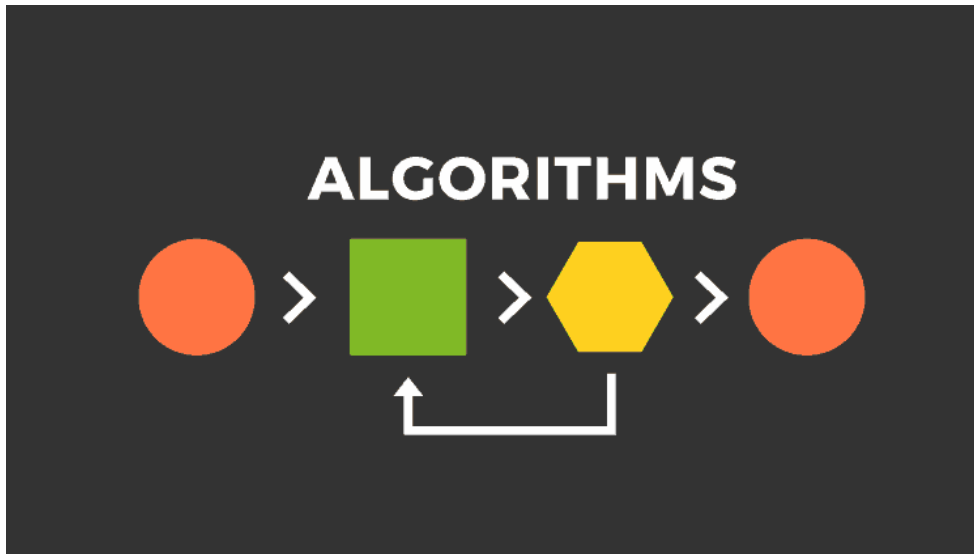
Traceability of the
algorithm
construction
process

Explainability policy
for explainable
results, auditability

**Use State of the art
algorithm OR
Pre-select **algorithms
to benchmark** (to
avoid overfitting test
data)**

**Choose a **pertinent
metric to evaluate
algorithm****

Step 3 – Algorithm selection



Quality policy

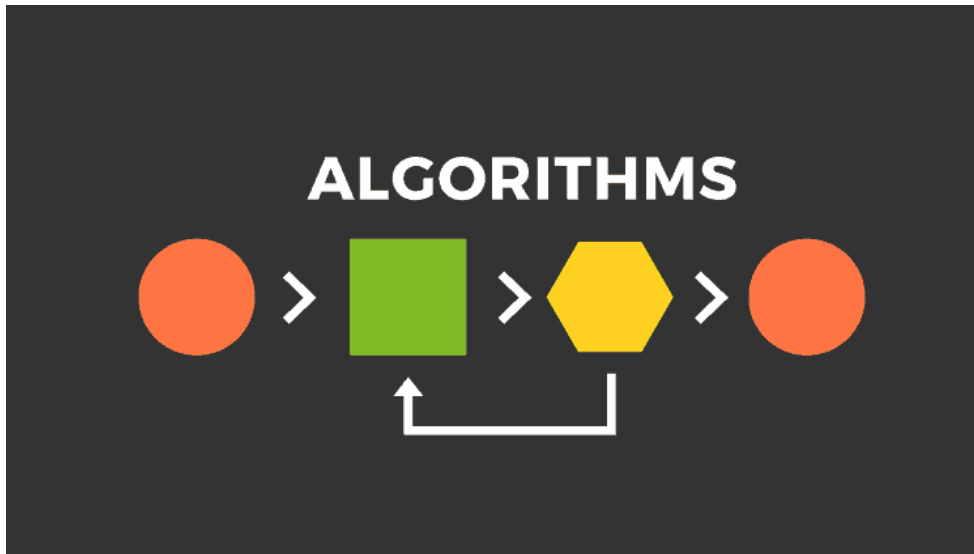
Transparency
measures

Traceability of the
algorithm
construction
process

Explainability policy
for explainable
results, auditability

**Keep track of all
decisions to help
reproducibility**

Step 3 – Algorithm selection



Quality policy

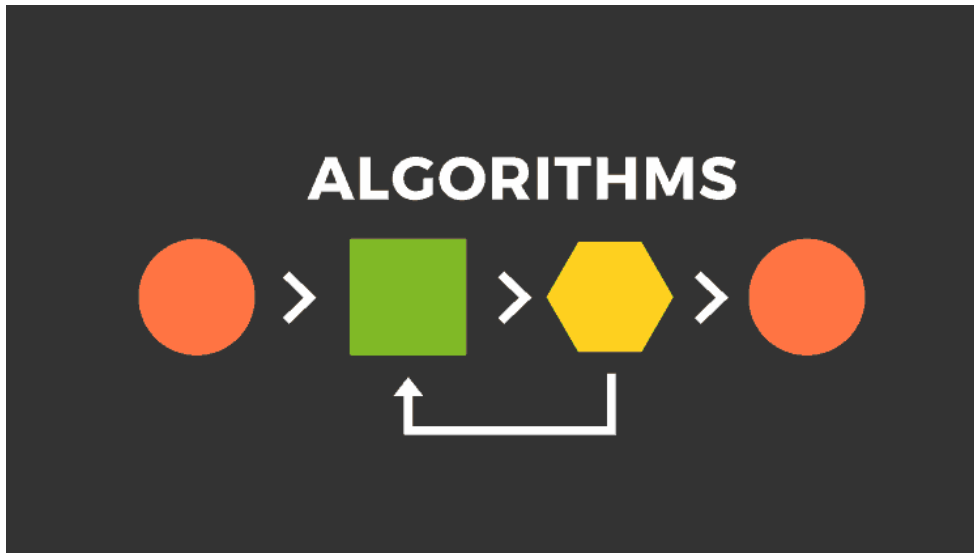
Transparency
measures

Traceability of the
algorithm
construction
process

Explainability policy
for explainable
results, auditability

**Code stored and
versioned (e.g.
Github)**

Step 3 – Algorithm selection



Quality policy

Transparency
measures

Traceability of the
algorithm
construction
process

Explainability policy
for explainable
results, auditability

Explain **why** some
individuals have a **high**
estimate of genetic
risk & how risk is
calculated

Step 4 - Evaluation of the algorithm before the production phase



Technical (bugs),
clinical (accuracy
score)

Usability

Non-discrimination

Robustness/
reproducibility

Information (fair
and equal) of the
users

Procedure in the
event of a cyber-
attack

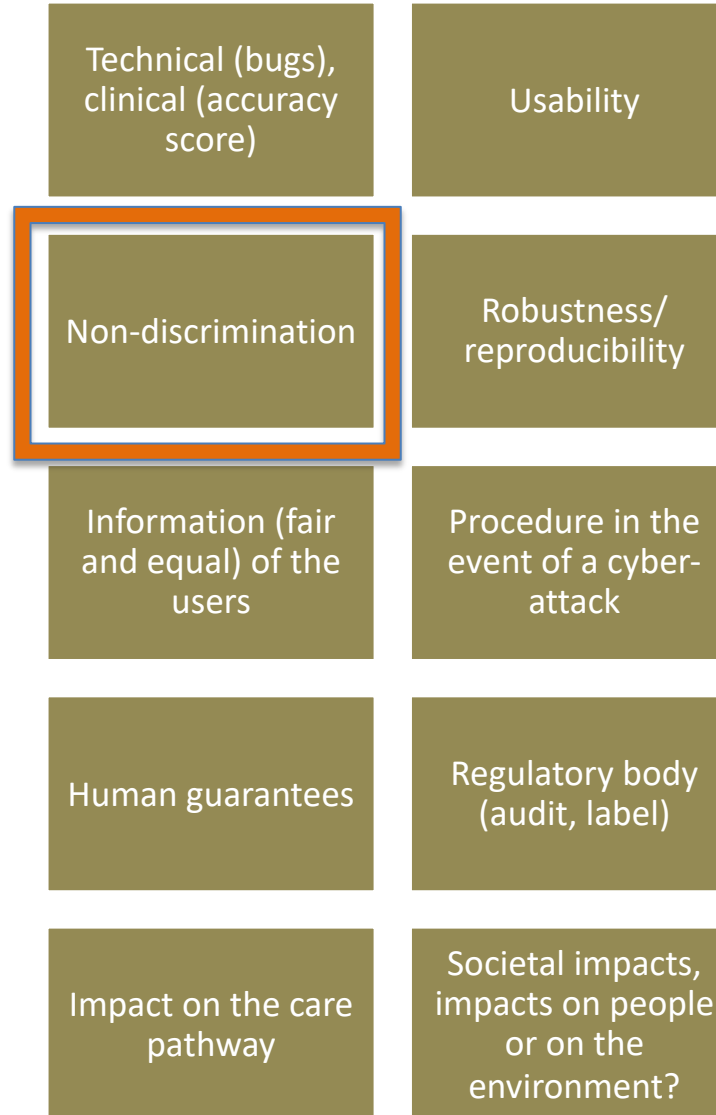
Human guarantees

Regulatory body
(audit, label)

Impact on the care
pathway

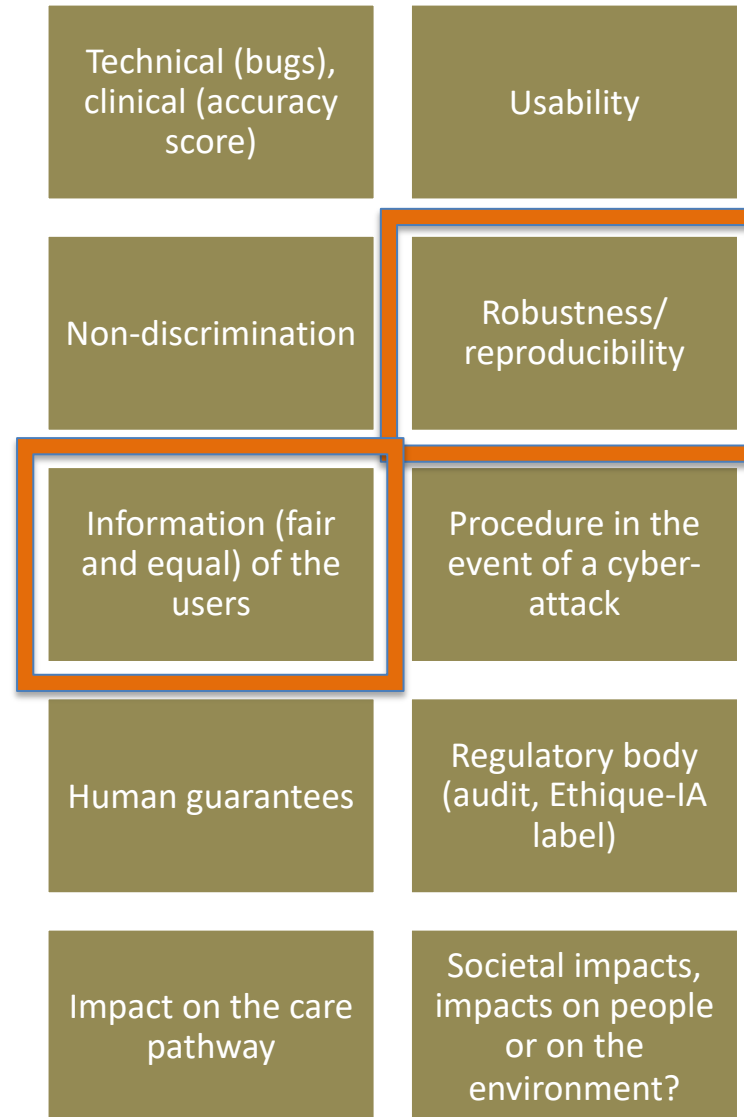
Societal impacts,
impacts on people
or on the
environment?

Step 4 - Evaluation of the algorithm before the production phase



**Evaluate algorithm
performance and
report results**
- In each sex group

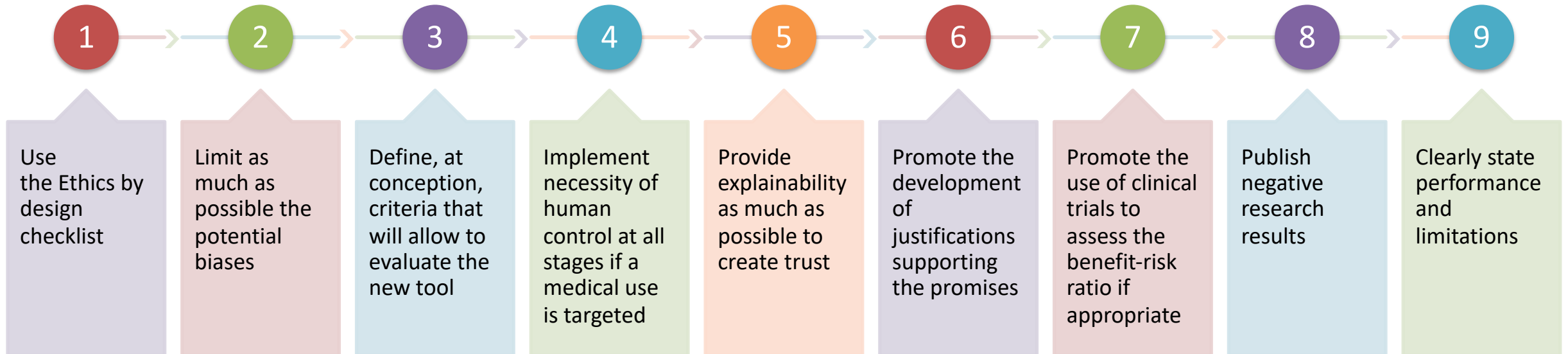
Step 4 - Evaluation of the algorithm before the production phase



All details of analysis
in **publication /
readme** that
accompanies score

Key recommendations to be implemented

Responsibilities of researchers



Thank you

Special thanks to Pr. Ségolène Aymé – head of the Ethics Committee of the Paris Brain Institute for her contributions to this presentation.

Thank you to the funding agencies : NHMRC (CJ Martin Fellowship) and INRIA.