

Intelligenza artificiale e sistemi di supporto decisionale

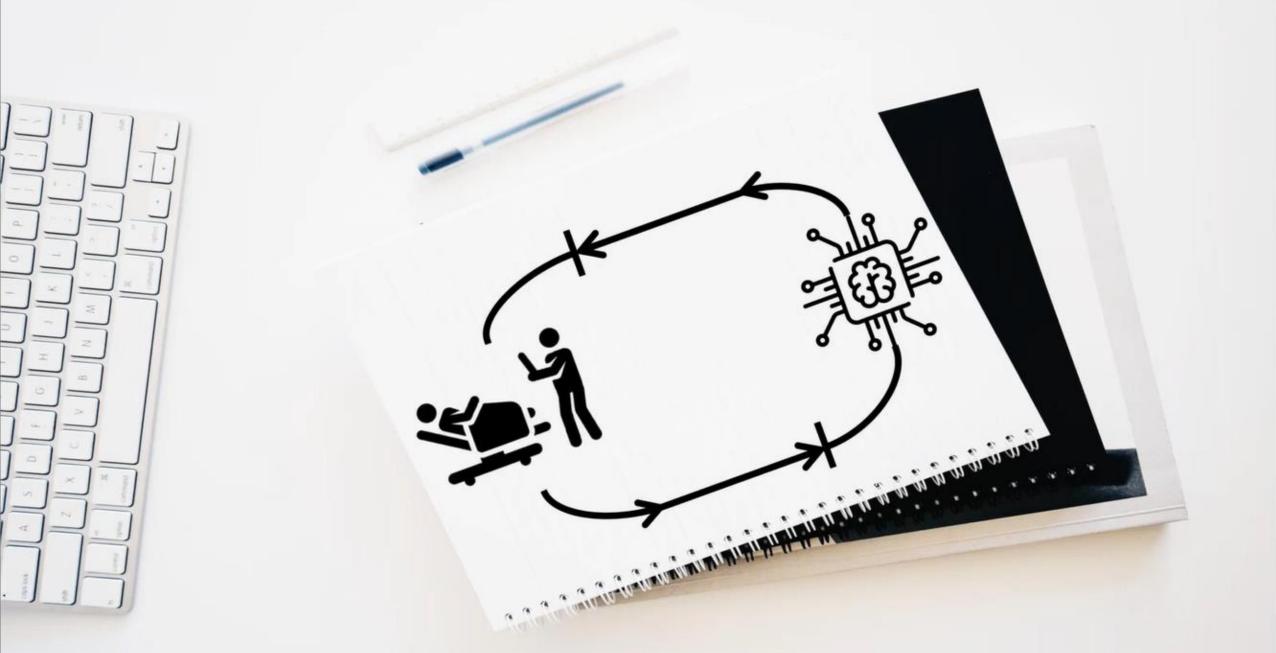




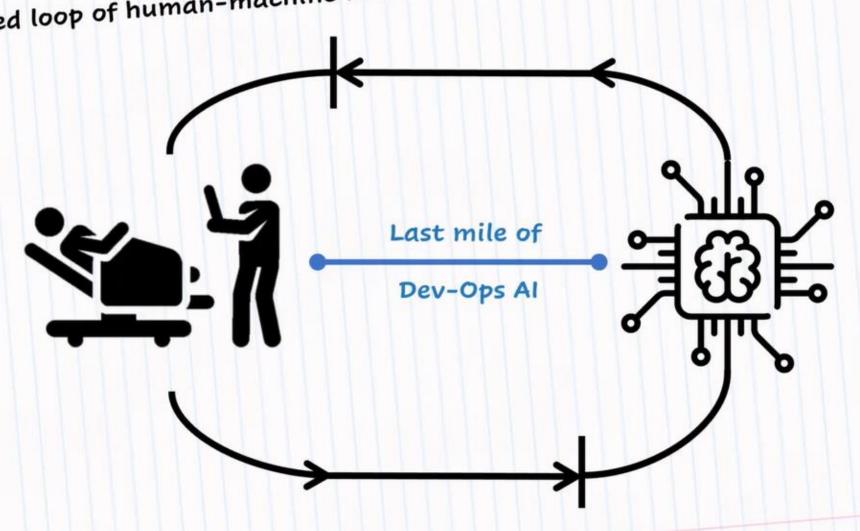
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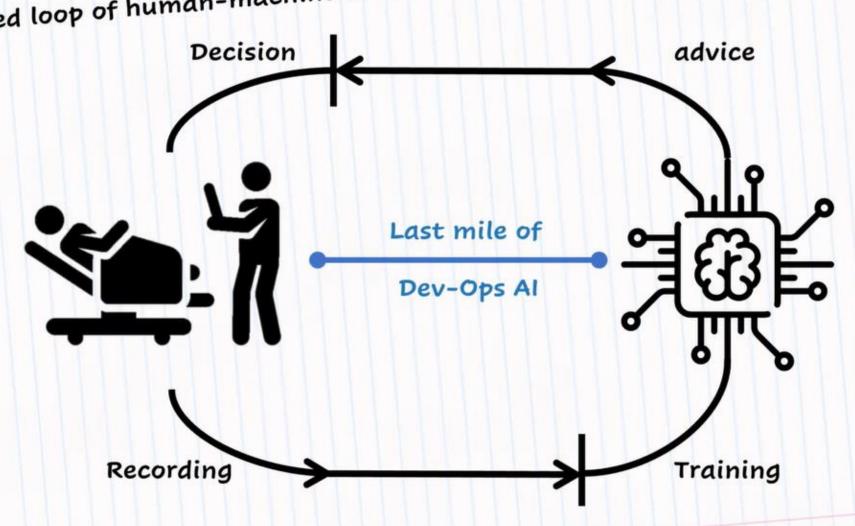




closed loop of human-machine mutual influence

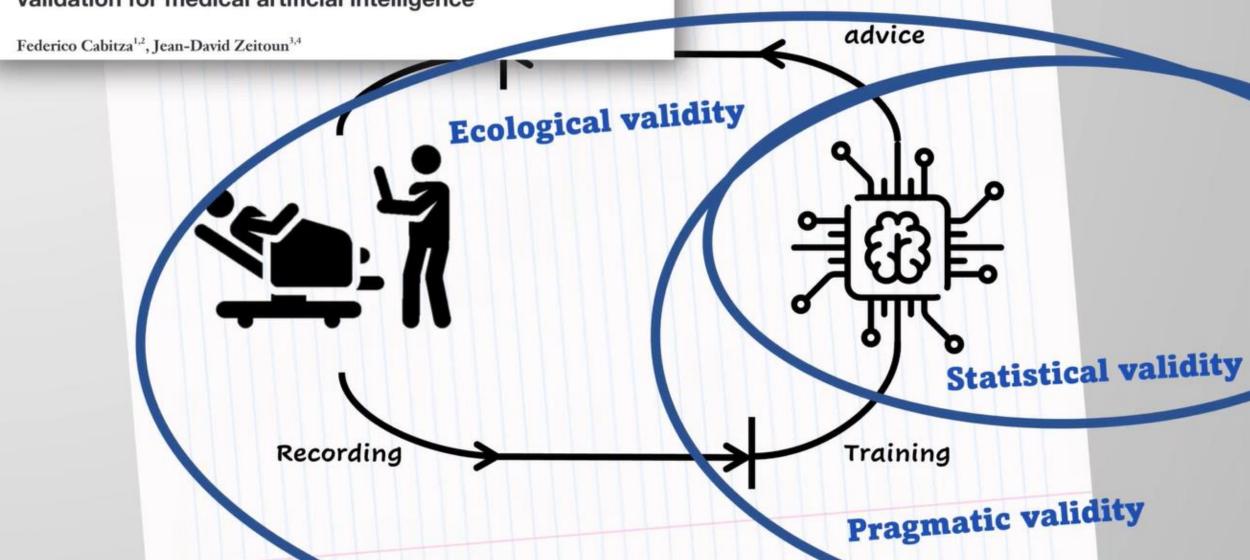


closed loop of human-machine mutual influence





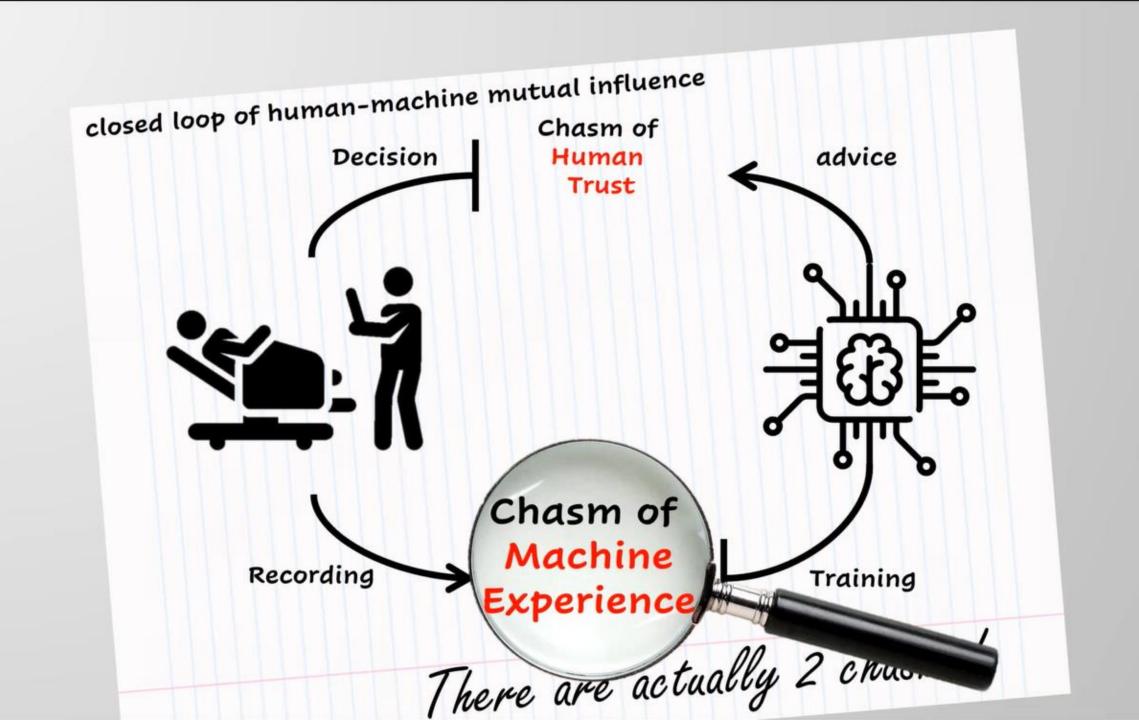
The proof of the pudding: in praise of a culture of real-world validation for medical artificial intelligence





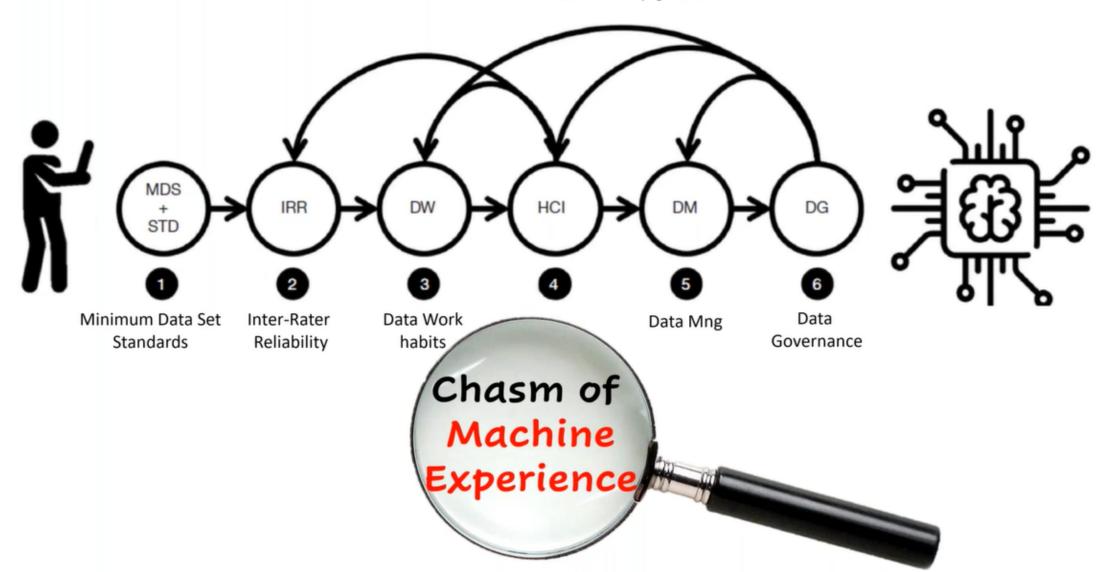
The proof of the pudding: in praise of a culture of real-world validation for medical artificial intelligence

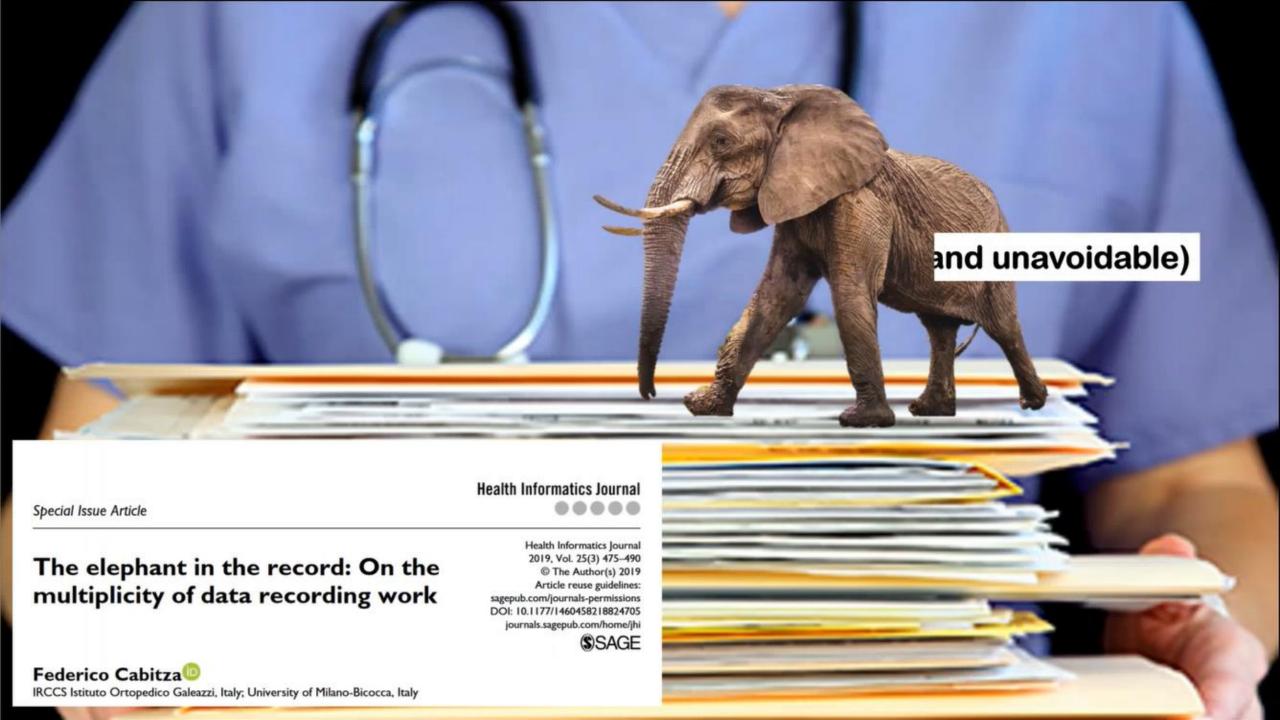
advice Federico Cabitza^{1,2}, Jean-David Zeitoun^{3,4} **Ecological validity** You can fail to achieve ecological validity even with high (statistical) accuracy, and you can have Statistical validity ecological validity without high accuracy! Training cording **Pragmatic validity**



Bridging the "last mile" gap between Al implementation and operation: "data awareness" that matters

Federico Cabitza, Andrea Campagner, Clara Balsano





Development, evaluation, and validation of machine learning models for COVID-19 detection based on routine blood tests



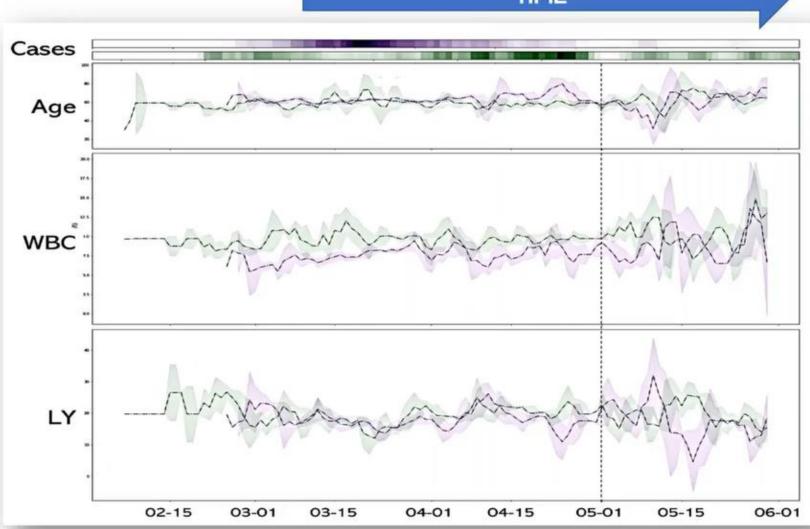
Clinical Chemistry and Laboratory Medicine (CCLM)

Volume 59 Issue 2

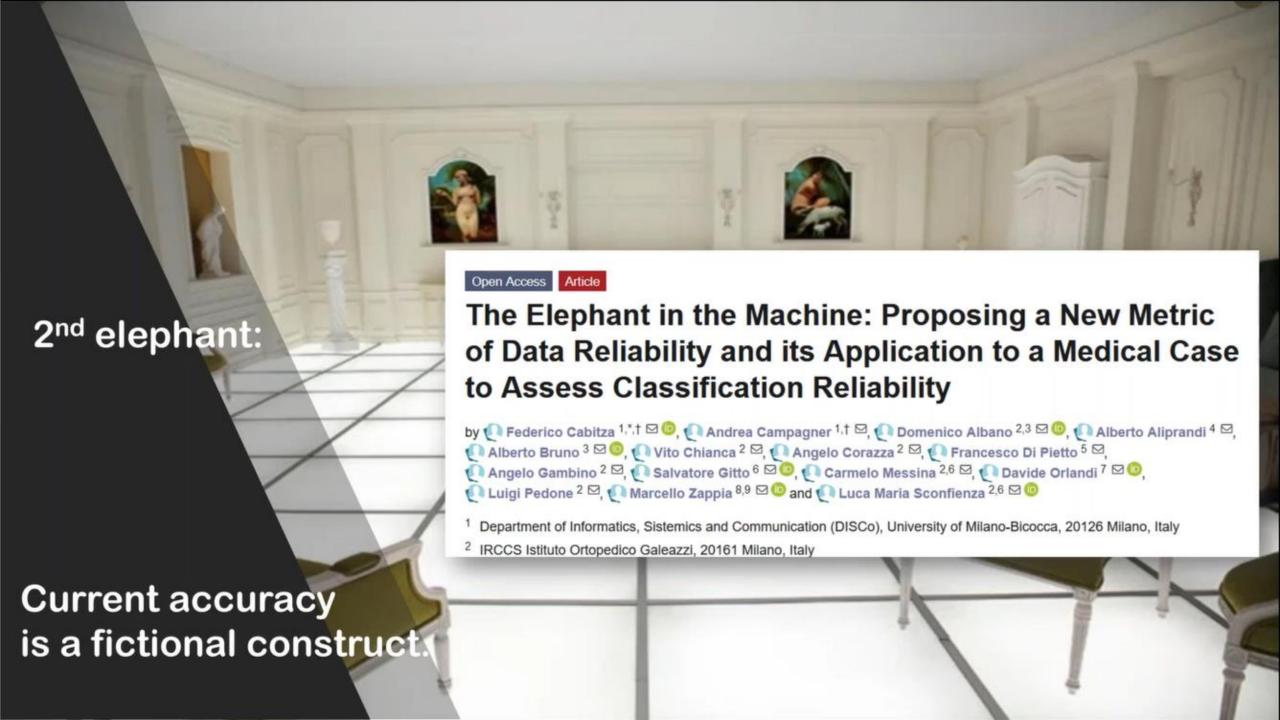
ACCURACY DECREASE

CONCEPT DRIFT



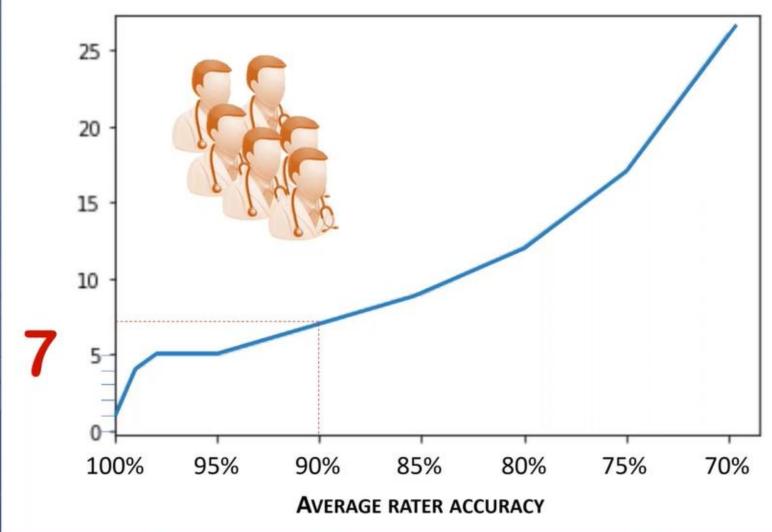






NUMBER OF RATERS TO INVOLVE TO GET A LABELLING ERROR UNDER 5%

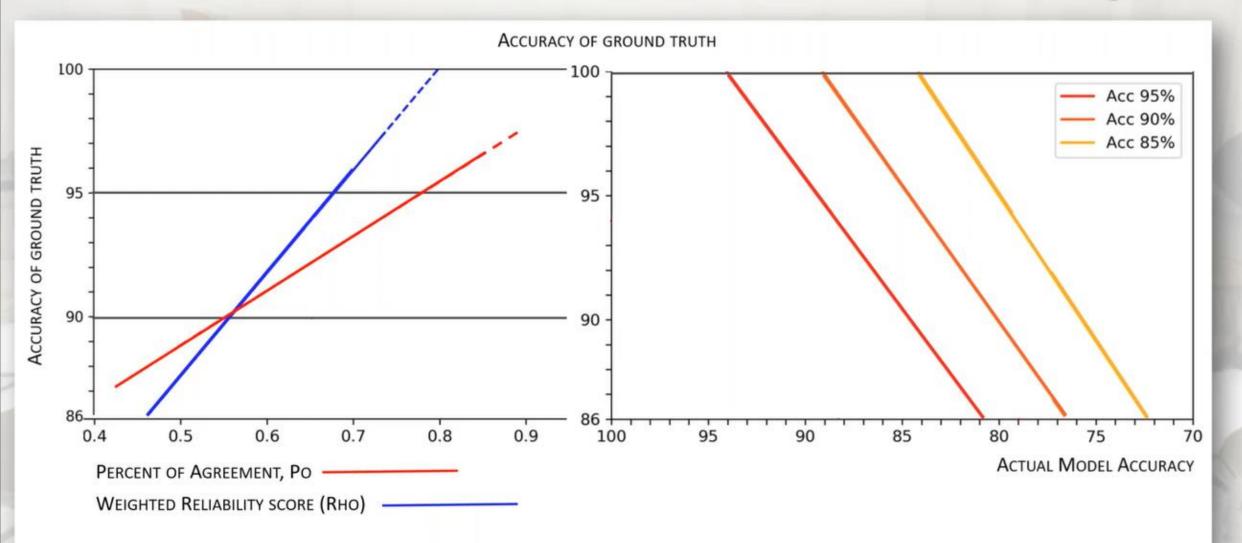
se



2nd elephant:

is a fictional constitue.

A nomogram



We proposed a new metric

7 M 6 8 the weighted Utility (wU)

$$wU(\tau,\mathbf{r},S,h) = \frac{1}{\mathbf{r}(Pos)} \sum_{x_i:y_i=1} r(x_i) \cdot \sigma_{\gamma}(\mathbf{h}(x_i)|\tau(x_i)) - \frac{1}{\mathbf{r}(Pos)} \sum_{x_i:y_i=0} r(x_i) \cdot \frac{\tau(x_i)}{1-\tau(x_i)} \cdot \sigma_{\gamma}(\mathbf{h}(x_i)|\tau(x_i)).$$

72



 $\sum_{x_i:y_i=0} r(x_i) \cdot \frac{\tau(x_i)}{1-\tau(x_i)} \cdot \sigma_{\gamma}(\mathbf{h}(x_i)|\tau(x_i)).$

Intuitively, a decision support is useful if the number of times it is right in detecting a health problem is higher than the number of times it is wrong so.



$$wU(\tau, \mathbf{r}, S, h) = \frac{1}{\mathbf{r}(Pos)} \sum_{x_i: y_i = 1} r(x_i) \cdot \sigma_{\gamma}(\mathbf{h}(x_i) | \tau(x_i)) - \frac{1}{\mathbf{r}(Pos)} \sum_{x_i: y_i = 0} r(x_i) \cdot \frac{\tau(x_i)}{1 - \tau(x_i)} \cdot \sigma_{\gamma}(\mathbf{h}(x_i) | \tau(x_i)).$$

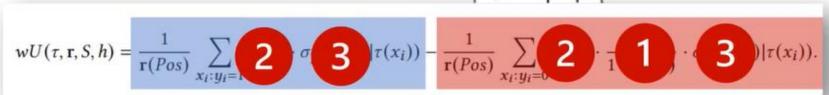
True positive rate

False positive rate

And if:

- It is optimized to avoid the most impactful kind of error (at class level)
- 2 It helps you when you need it most (i.e., most difficult/rarest cases)
- It doesn't take guesses.

Intuitively, a decision support is useful if the number of times it is right in detecting a th problem is higher than the nor of times it is wrong so.



True positive rate

False positive rate

The IJMEDI checklist for assessment of medical AI



Problem Understanding

- Is the study population described, also in terms of inclusion/exclusion criteria (e.g., patients older than 18 tested for COVID-19; all inpatients hospitalized for 24 or more hours)? §
- 2. Is the study design described? (e.g., retrospective, prospective, cross-sectional II, observational, randomized control trial II) §
- Is the study setting described? (e.g., teaching tertiary hospital; primary care ambulatory, nursing home, medical laboratory, R&D laboratory) §
- 4. Is the source of data described? (e.g., electronic specialty registry; laboratory information system; electronic health record; picture archiving and communication system) §
- 5. Is the medical task reported? (e.g., diagnostic detection, diagnostic characterization, diagnostic staging, prognosis (on which endpoint), event prediction, risk stratification, anatomical structure segmentation, treatment selection and planning, monitoring) §
- 6. Is the data collection process described, also in terms of setting-specific data collection strategies (e.g. whether body temperatures are measured only in the morning; whether some blood tests are performed only in light of a specific diagnostic hypothesis)? Any consideration about data quality is appreciated, e.g., in regard to completeness, plausibility, and robustness with respect to upcoding or downcoding practices

Data Understanding

- 7. Are the subject demographics described in terms of
 - 1. average age (mean or median);
 - age variability (standard deviation (SD) or inter-quartile range (IQR));
 - gender breakdown (e.g., 55% female, 44% male, 1% not reported); §
 - 4. main comorbidities;
 - ethnic group (e.g., Native American, Asian, South East Asian, African, African American, Hispanic, Native Hawaiian or Other Pacific Islander, European or American White

International Journal of Medical Informatics

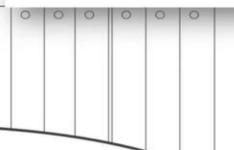
Available online 2 June 2021, 104510

In Press, Journal Pre-proof (?)



The need to separate the wheat from the chaff in medical informatics

Federico Cabitza A, Andrea Campagner



To discover more, please refer to the new IJMEDI checklist for assessment of medical AI

- Rho Div
- 5. Labelling technique te-g-r

