

AI IN CLINICAL TRIALS: IS THERE A ROLE FOR STATISTICS?

EFSPI Regulatory Statistics Workshop 2025

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IS THERE A ROLE FOR STATISTICS IN AI?

▷ DAGStat (German Consortium of Statistics)

- ▷ Association of 15 professional and learned societies and the Destatis (Federal Statistical Office in Germany)
- ▷ Homepage <https://www.dagstat.de/>



Advances in Data Analysis and Classification
<https://doi.org/10.1007/s11634-021-00455-6>

REGULAR ARTICLE



Is there a role for statistics in artificial intelligence?

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STATISTICAL STUDY DESIGN

- ▷ **Experiment - observational study - convenience sample**
 - ▷ AI utilizes often convenience samples (e.g. routine data) since large data sets are accessible at low costs
 - ▷ Target trial emulation
- ▷ **Naive expectation that sufficiently large data automatically lead to representativity**
 - ▷ Careful planning required
 - ▷ Example: Apple Heart Study (Perez et al (2019) NEJM) to assess Apple Smartwatch's ability to identify atrial fibrillation; Large-scale assessment including more than 400,000 participants; Caveat: Average age of participants 41 years, but AF most prevalent in older patients (>65 years of age)

STATISTICAL APPROACHES FOR VALIDATION

- ▷ **Internal vs. external validation**
- ▷ **Bench marking** (computer science) vs. **simulations** (statistics) (see also Friedrich & Friede (2023) <https://doi.org/10.1002/bimj.202200212>)
 - ▷ Experience with structuring, reporting and interpreting simulation studies in statistics
- ▷ **Sample sizes**
 - ▷ Sample size planning common task in clinical trials, but not routinely performed in AI / ML applications
 - ▷ Requirements depending on dimensionality, sparsity, non-linearity, ...
- ▷ **Challenges include fast development cycles with AI technologies** (often faster than validation studies)

“GARBAGE IN, GARBAGE OUT”

- ▶ AI need to be trained and evaluated on ‘**fit for purpose**’ **data** (includes relevance, completeness, availability, timeliness, meta-information, documentation and context-dependent expertise) (Duke-Margolis, 2018)



Fig. 3 Data relevancy and quality are equivalent components of a fit-for-purpose real-world data set. Figure according to Duke-Margolis (2018)

- ▶ **More data ≠ more information** (e.g. random-effects meta-analysis; Jackson and Turner, 2017)

- ▶ **AI / ML excellent at discovering associations**
- ▶ Important to acknowledge in the interpretation of results:
Association not necessarily due to causal relationship
- ▶ **Lessons to be learned from other fields**
 - ▶ Clinical epidemiology: Bradford Hill criteria
 - ▶ Statistics: Counterfactual framework by Rubin (1974)

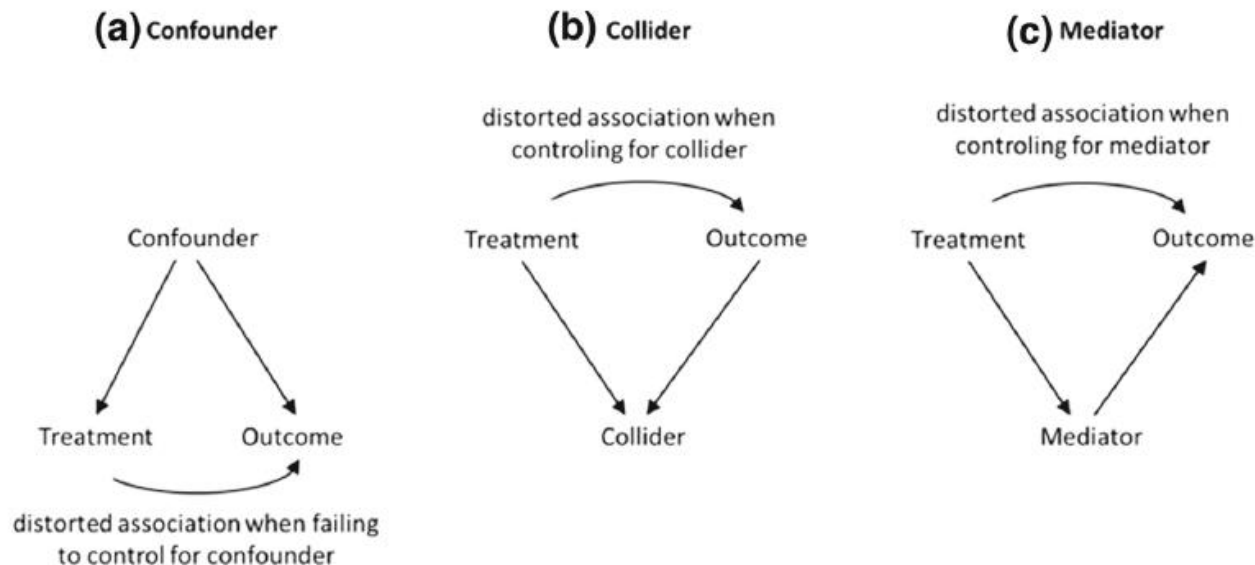


Fig. 4 Covariate effects in observational data, according to Catalogue of bias collaboration (2019)

EVALUATING UNCERTAINTY

- ▶ **Uncertainty quantification is often neglected in AI applications**
- ▶ With large data sets **sampling variation** might be small
- ▶ However: **Model uncertainty** might remain
- ▶ Derivation / computation of measures of uncertainty such as standard errors or prediction intervals can be tricky with complex analysis methods
- ▶ More recently, application of **resampling techniques** such as bootstrapping or jack knifing (theoretical properties not always established)
- ▶ Alternative approaches embed algorithmic methods in statistical models

APPLICATIONS OF AI/ML IN CARDIOVASCULAR MEDICINE: A SYSTEMATIC REVIEW

- ▶ DZHK (German Center for Cardiovascular Research)
- ▶ Project group AI / ML



DZHK
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ESC

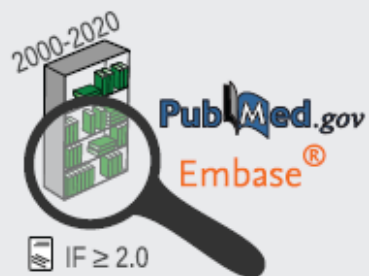
European Heart Journal Digital Health
European Society of Cardiology
doi:10.1093/ehjdh/ztab054

REVIEW

Applications of artificial intelligence/machine learning approaches in cardiovascular medicine: a systematic review with recommendations

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Marcus Kelm ^{10,11,12,13}, Andreas Leha^{1,14}, Jasmin Rühl¹, Jens Schaller^{10,13},
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Systematic review: Applications of AI/ML approaches in cardiovascular medicine



AI/ML, *cardi*, ...
original paper, clinical cohort



215 papers

- ✓ Study characteristics
disease area, #subjects, design, outcome, secondary data
- ✓ AI/ML characteristics
input types, methods, uncertainty quantification
- ✓ Reproducibility
data and code availability

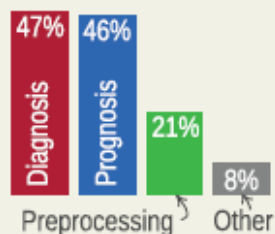


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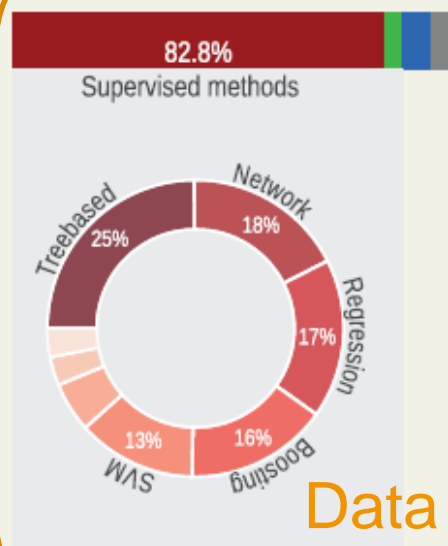
Which disease?



What for?

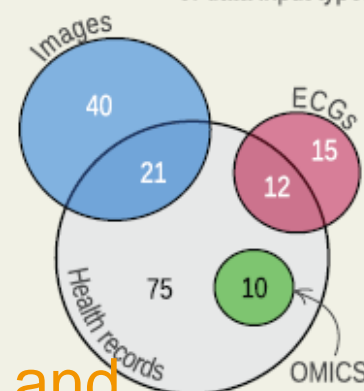


How?



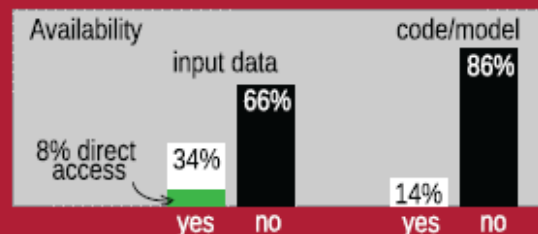
Which data?

Most frequent combinations of data input types



Data and
Methods

Deficits



Reproducibility

MV 04/2021

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WHITE PAPER



Scientific and regulatory evaluation of mechanistic *in silico* drug and disease models in drug development: Building model credibility

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In silico trials: “Class of trials for pharmacological therapies or medical devices based on modelling and simulation technologies. Such trials produce digital evidence that can serve in complement to or replacement of in vivo clinical trials for the development and regulatory evaluation of medical therapies.”

IN-SILICO CLINICAL TRIALS IN DRUG DEVELOPMENT: A SYSTEMATIC REVIEW

- ▶ Systematic review of PubMed and clinicaltrials.gov for published articles and registered clinical trials related to in-silico clinical trials
- ▶ Preprint <https://arxiv.org/abs/2503.08746>

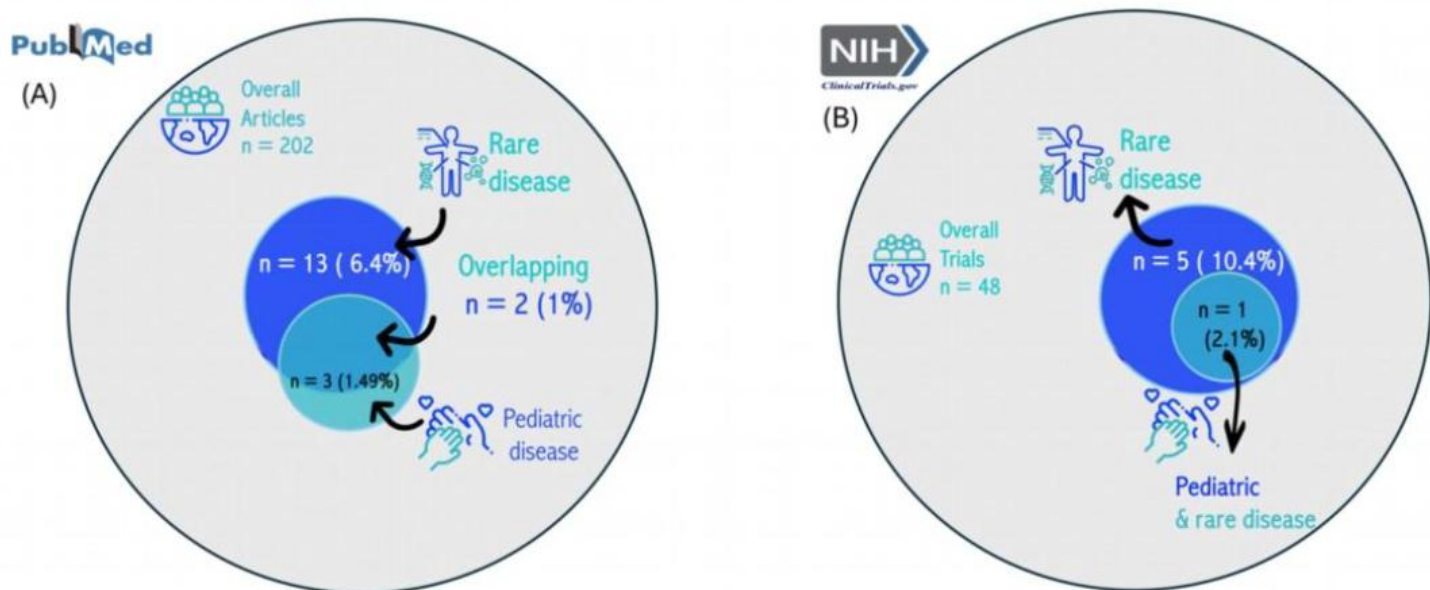


Figure 6. Panels (A) and (B) show the number of rare and pediatric diseases out of all the publications and trials.

CONCLUSIONS AND DISCUSSION

- ▷ **Increasing number of applications of AI/ML** in clinical research, e.g. AI based adjudication of clinical events, subgroup identification in clinical trials
- ▷ **Is there a role for statistics in AI / ML? Yes!**
 - ▷ Statistics contributes to **methods**, but impact on **applications** in my view even more important
- ▷ **AI fuelling innovative approaches such as in-silico clinical trials and digital twins**
- ▷ **Interdisciplinary collaboration and network**

FROM BLACK BOX TO PANDORA'S BOX

Pandora's box

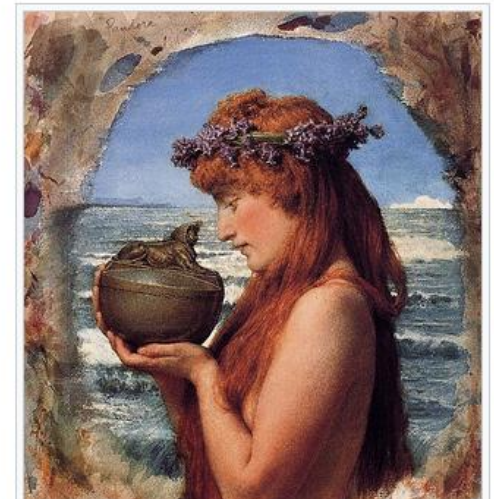
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From Wikipedia, the free encyclopedia

This article is about the mythological artifact. For other uses, see [Pandora's box \(disambiguation\)](#).

Pandora's box is an artifact in [Greek mythology](#) connected with the myth of [Pandora](#) in [Hesiod](#)'s c. 700 B.C. poem *[Works and Days](#)*.^[1] Hesiod related that curiosity led her to open a [container](#) left in the care of her husband, thus releasing [curses](#) upon mankind. Later depictions of the story have been varied, with some literary and artistic treatments focusing more on the contents than on Pandora herself.

The container mentioned in the original account was actually a large storage [jar](#), but the word was later mistranslated. In modern times an idiom has grown from the story meaning "Any source of great and unexpected troubles",^[2] or alternatively "A present which seems valuable but which in reality is a curse".^[3]



https://en.wikipedia.org/wiki/Pandora%27s_box