3050571 Practical Clin Data Sci

Session 16: Explainability

February 29, 2024



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- Research Affairs
- Center of Excellence in Computational Molecular Biology (CMB)
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Caution when using Al

AI (silently) makes mistakes and biases



But can you spot them?

Alkaissi, H. et al. Cureus 15:e35179 (2023)



Late onset Pompe disease (LOPD) is a rare genetic disorder characterized by the deficiency of acid alpha-glucosidase (GAA), an enzyme responsible for the breakdown of glycogen in lysosomes. The accumulation of glycogen in various tissues leads to progressive muscle weakness, primarily affecting the skeletal and respiratory muscles. However, recent studies have also reported liver involvement in LOPD, which is thought to occur as a result of the accumulation of glycogen in liver cells.

- There was no prior publication about liver involvement with LOPD
- However, the authors of this paper have <u>an unpublished manuscript</u> showing a link between liver disease and LOPD
 - Did ChatGPT just synthesized new knowledge? Or simply hallucinated?

Huge gap between development and actual use

Healthcare, Law, Regulation, and Policy, Machine Learning

"Flying in the Dark": Hospital Al Tools Aren't Well Documented

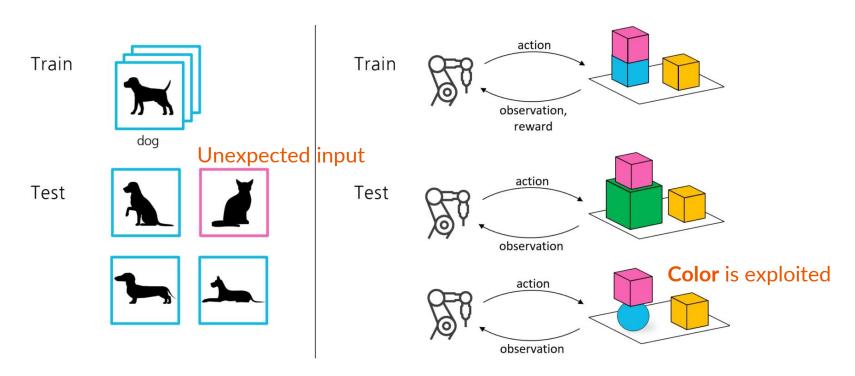
MODEL REPORTING GUIDELINES	EPIC MODEL BRIEFS											
	Deter iorati on Index	Detection of Sepsi	Risk of Unpl anne d Read missi on	Risk of Patie nt No- Show	Pediatri c Risk of Hospital Admissi on or ED Visit	Hospit al Admiss ion or ED	Inpatie nt Risk of Falls	cted Block	ning	Admiss ion of Heart	Risk of Hospital Admissi on or ED Visit for Asthma	Risk of Hyper tensio n
TRIPOD	63%	63%	61%	48%	42%	61%	47%	36%	55%	48%	44%	51%
CONSORT-AI	63%	43%	63%	60%	33%	67%	53%	47%	47%	49%	42%	51%
SPIRIT-AI	61%	55%	54%	54%	38%	61%	44%	49%	51%	41%	39%	46%
Trust and Value	46%	33%	39%	50%	29%	42%	38%	46%	46%	25%	33%	46%
ML Test Score	27%	15%	33%	24%	9%	33%	15%	6%	18%	12%	9%	15%

Evaluation of sepsis diagnosis Al

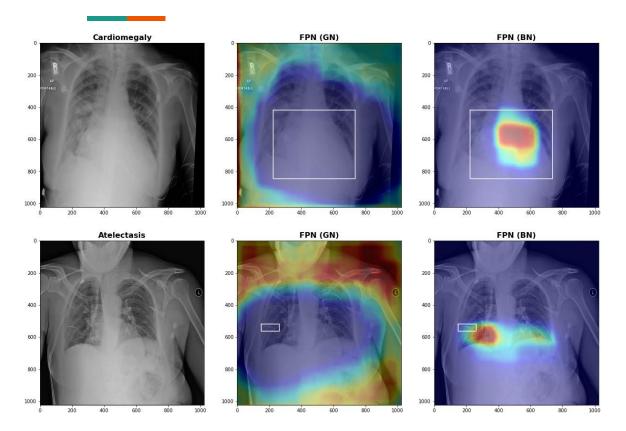
Results We identified 27697 patients who had 38455 hospitalizations (21904 women [57%]; median age, 56 years [interquartile range, 35-69 years]) meeting inclusion criteria, of whom sepsis occurred in 2552 (7%). The ESM had a hospitalization-level area under the receiver operating characteristic curve of 0.63 (95% CI, 0.62-0.64). The ESM identified 183 of 2552 patients with sepsis (7%) who did not receive timely administration of antibiotics, highlighting the low sensitivity of the ESM in comparison with contemporary clinical practice. The ESM also did not identify 1709 patients with sepsis (67%) despite generating alerts for an ESM score of 6 or higher for 6971 of all 38455 hospitalized patients (18%), thus creating a large burden of alert fatigue.

- AUC of 0.63 in practice
- Missed 67% of sepsis

(Un)expected behaviors



Correct prediction is not enough



- Two models with the same classification performance
- Both images were correctly classified
- But the explanations complete differ

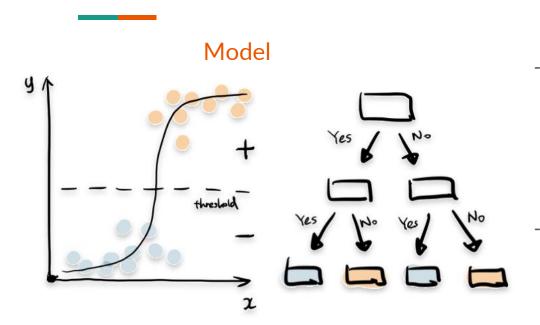
Explainability is key

Does the predicted confidence match your expectation? Output = 0.4Age = 65+0.4 Sex = F-0.3 BP = 180BMI = 40Impact of each input Base rate = 0.1on the prediction

pred: 0.57 | GT: 1 2 Parts of the input that most strongly contribute to the prediction

Inherently interpretable models

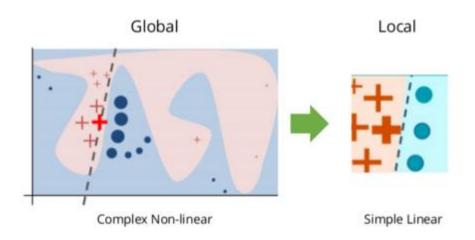
Linear and tree models



- Model decisions are immediately understandable
 - Examine coefficients
 - Trace the decision in a tree
- We can use this to approximate a more complex model

https://towardsdatascience.com/top-machine-learning-algorithms-for-classification-2197870ff501

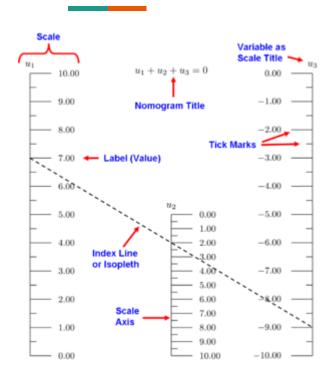
LIME



https://c3.ai/glossary/data-science/lime-local-interpretable-model-agnostic-explanations/

- Approximate decision boundary surround a data point
- Slightly perturb original data and get predictions from the model

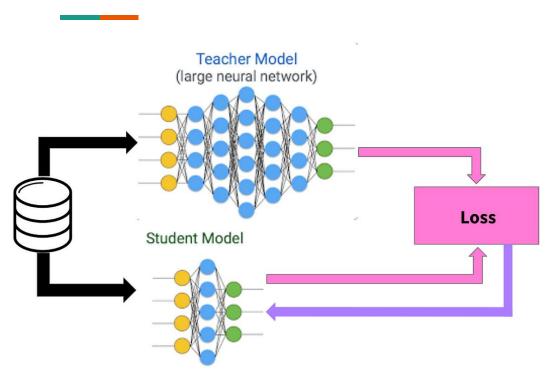
Nomogram



https://en.wikipedia.org/wiki/Nomogram

- After a complex model (C) is developed, a simpler model (S) can be fitted on the input data and the prediction made by C
- Nomogram can be fitted to mimic a random forest model
- Easy to use on site and interpretable

[Related] Distillation

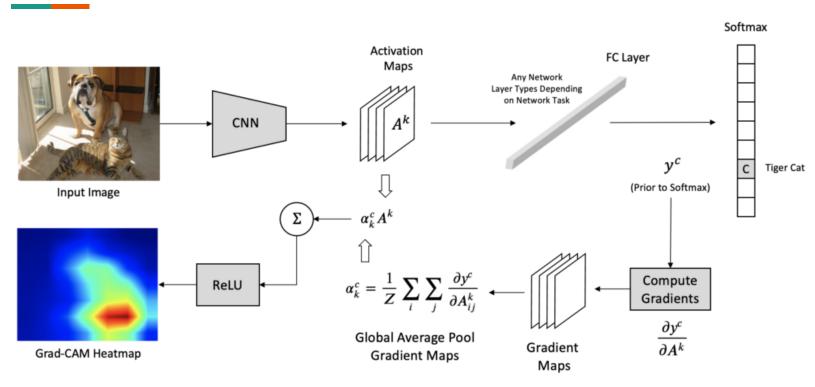


- Larger model encodes knowledge from the training data
- 0/1 label into probability and embedding

https://towardsdatascience.com/knowledge-distillation-simplified-dd4973dbc764

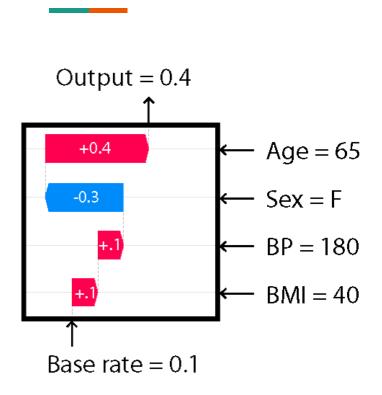
Explainability techniques

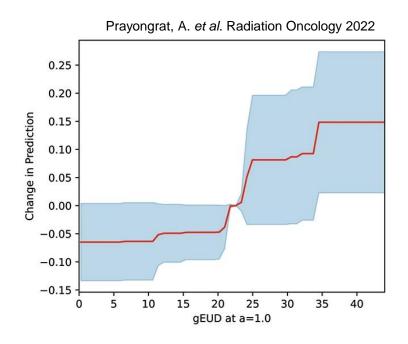
Gradient tracing & saliency map



https://learnopencv.com/intro-to-gradcam/

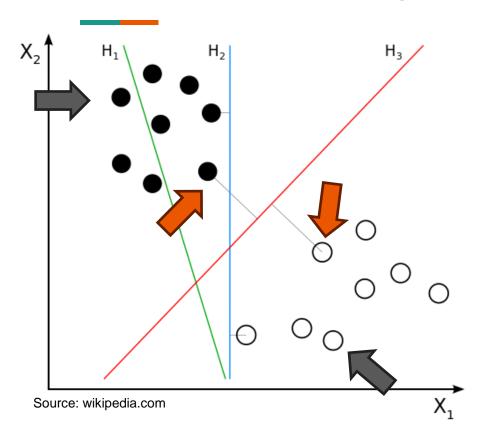
Input value perturbation





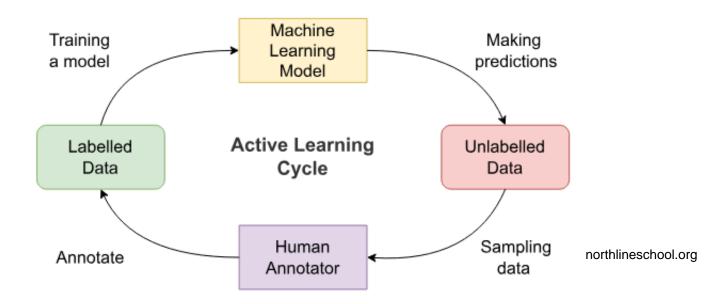
Calculate prediction changes when target feature values are perturbed

Impact of each training sample



- Some define the decision boundary
- Some are located among many other similar data points
- Drive additional data collection

[Related] Active learning



Use model predictions to determine which data to collect and which experiments to perform

Error analysis

- Identify systematic errors
- Bias in data, spurious association, mismatch between training and test set, or model limitation
- Hard examples
- Drive additional data acquisition and model improvement

Counterfactual argument

"Thinking about what did not happen but could have happened"





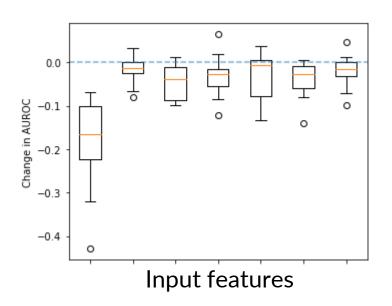
- Alter regions identified to be important to the predicted class / value
- Observe whether the **prediction changes**





Mertes, S. et al. Frontiers in Al 2022

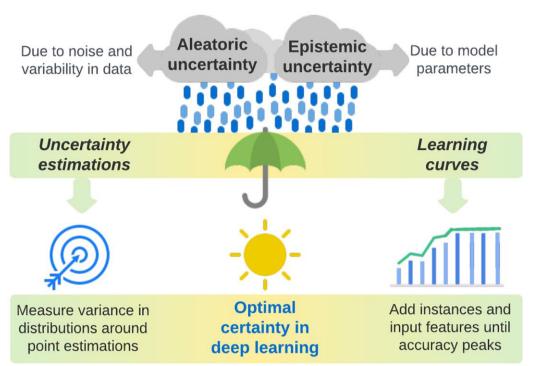
Change in performance after dropping a feature



- Compare performance with and without each input feature
- Big drop = important

Beyond accuracy

Uncertainty

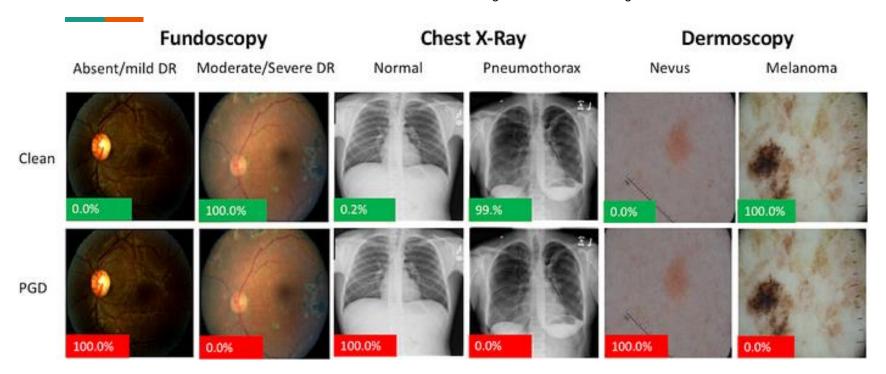


- Innate variability in data
 - Drug 3D structure
 - May be predictable
- Variability in trained model
 - Bootstrapping
 - Impact of data size
 - Ensemble approach

Loftus, T.J. et al. PLOS Digital Health 2022

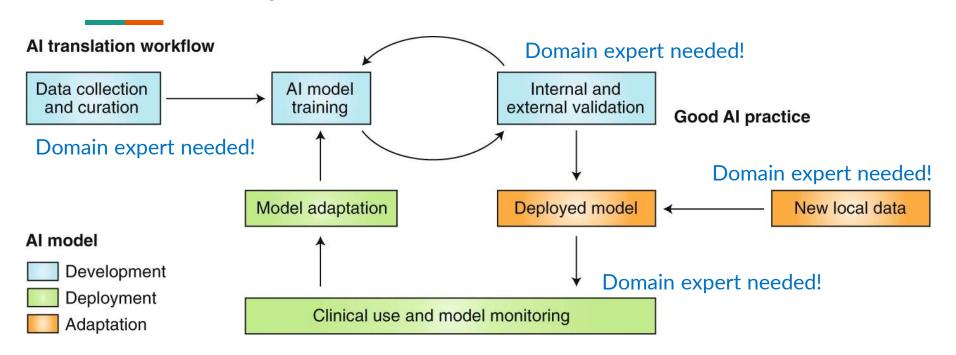
Stability

https://medium.com/pymedix/machine-learning-adversarial-attacks-its-all-fun-and-games-until-someone-gets-hurt-3f726113134d



Small input perturbation should not drastically change the prediction

Sustainability and other concerns



- Anyone can feed data through ML library
- Only domain experts can spot model weakness and find data to fix it

Any questions?

See you on March 6th