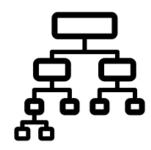
Challenges in adopting AI in Healthcare and ways to address them — seen by a data scientist

While AI is making significant progress, challenges refraining its adoption in Healthcare persist

Presence of bias in the data

Trade-off between Accuracy & Interpretability

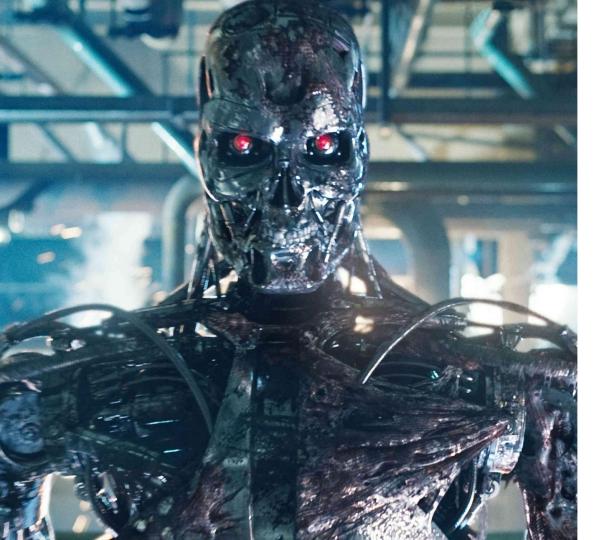






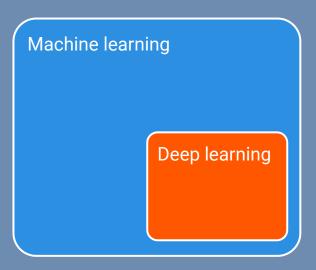


defining Al



Not Al.

Definitions



AI

Systems able to perform tasks that ordinarily require human intelligence

Machine learning

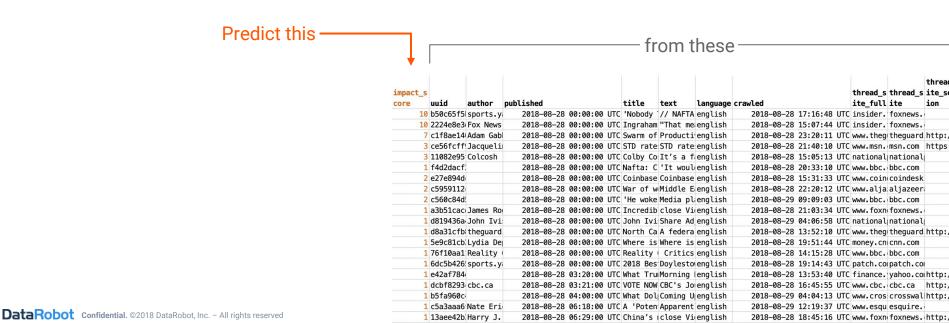
Learning from the past to predict the future

Deep learning

A specialised kind of machine learning which uses multi-layer neural networks to imitate the workings of the human brain to process data and creating patterns for use in decision making

Supervised machine learning

Ground truth (outcomes) known



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Unsupervised machine learning

No ground truth

what is similar? what is different?

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Machine learning is a two-stage process





addressing challenges in ML adoption



Presence of bias in the data



Trade-off between Accuracy & Interpretability





Presence of bias in the data







Trade-off between Accuracy & Interpretability





Training the workforce



1. Online courses

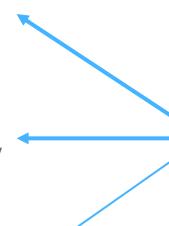
2. Learning from working on a real use case

3. Learning with a partner

4. Developing your own learning material



Presence of bias in the data



Training the workforce

practices

Adopting the best





Trade-off between Accuracy & Interpretability





Adopting the best practices





Methodology

Data partitioning: Cross-validation, out-of-time validation, group partitioning

Evaluation: measures of accuracy

Model building

Accuracy vs Interpretability: you don't have to give up on accuracy to get interpretable models

- Regularized Logistic Regression, GA2M, Eureqa...
- Favouring small number of attributes

Bias in data: monotonicity constraints



Model consumption

Prediction explanation: each prediction should be explained in terms of impact of each attribute (sign and intensity)

What-if scenario builder: model should be understood for different real-life scenarios

Presenting the most similar data points: raw historical data should be shown as explanation when similar new data points are scored



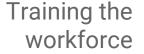
Data scientist



Expert (doctor, nurse...)



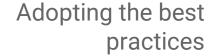
Presence of bias in the data







Trade-off between Accuracy & Interpretability











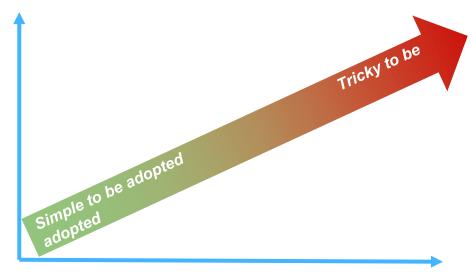
Identifying early on potential risks



Multiple risks for each use case

- Model's performance on new data not as good as on training data
- Model's mistakes can be handled: false negatives, false positives
- Model's prediction explanation doesn't make sense
- Bias from training data
- Data drift
- Model's returning an error
- Fairness

Time required for model adoption



Criticity of risks identified



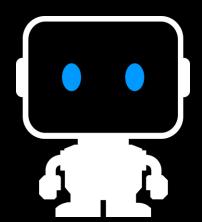
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