

Introduction to Machine Learning

 Lasse Hansen

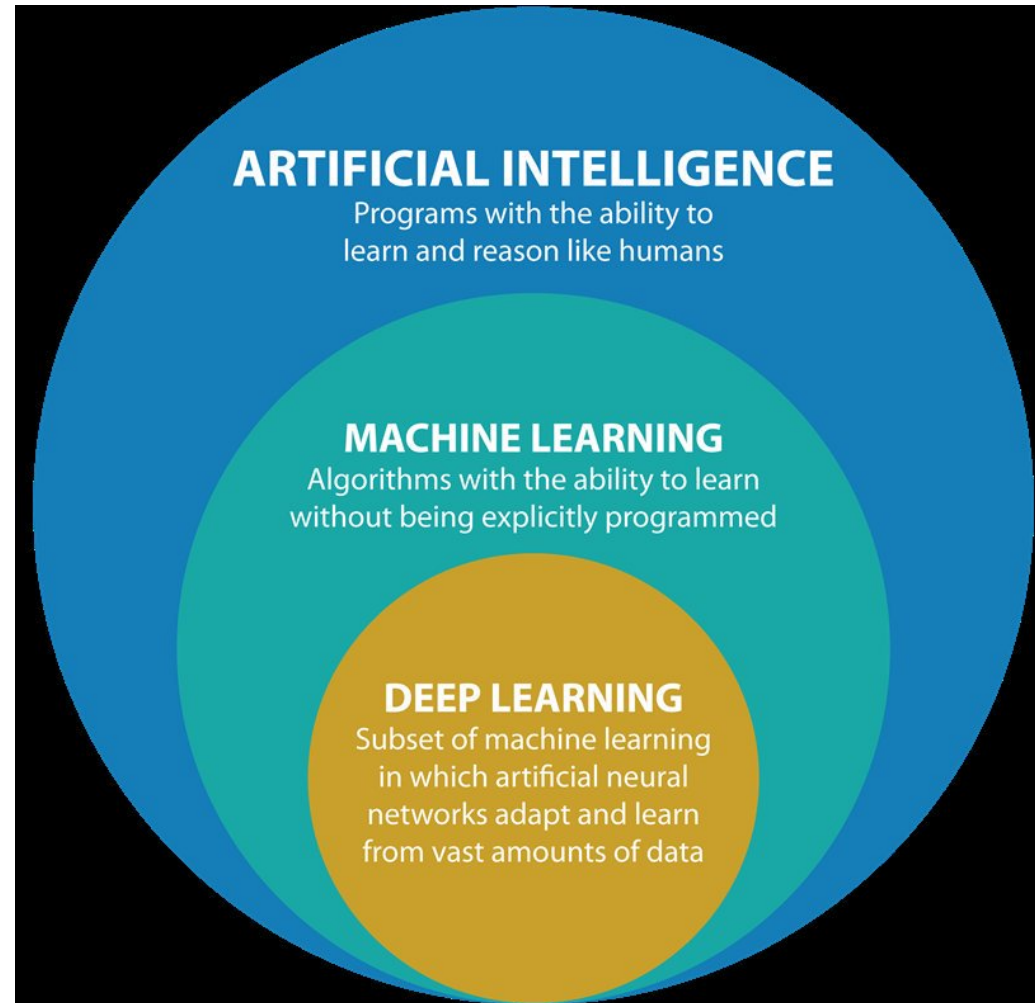
 ADA Journal Club

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Terminology



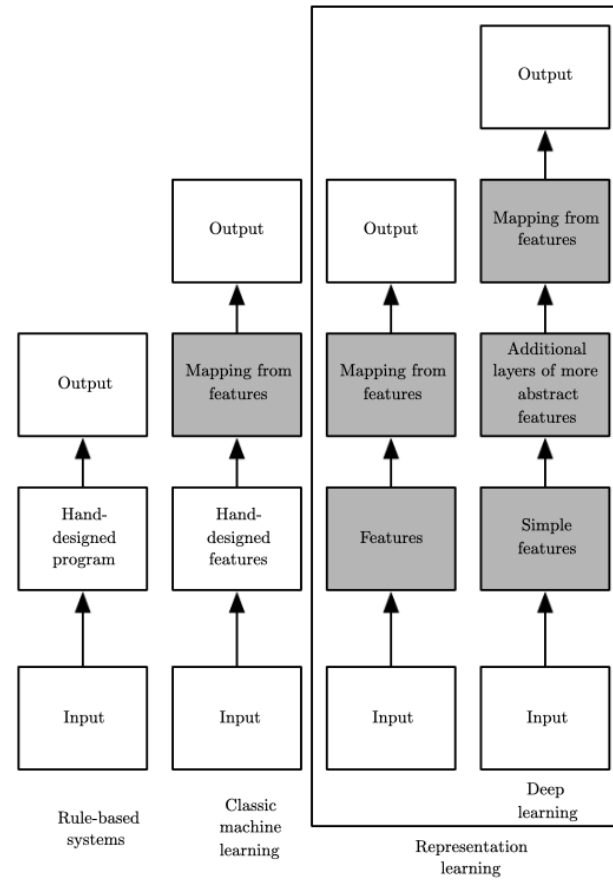
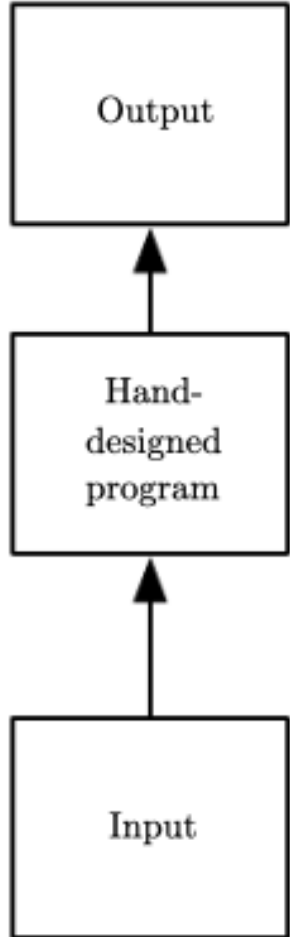


Figure 1.5: Flowcharts showing how the different parts of an AI system relate to each other within different AI disciplines. Shaded boxes indicate components that are able to learn from data.

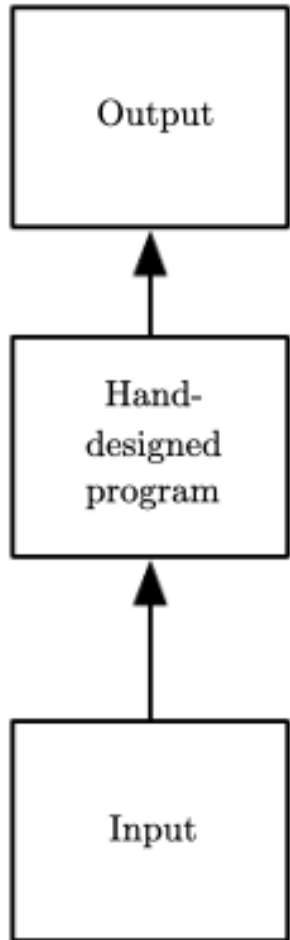
Rule-based Models

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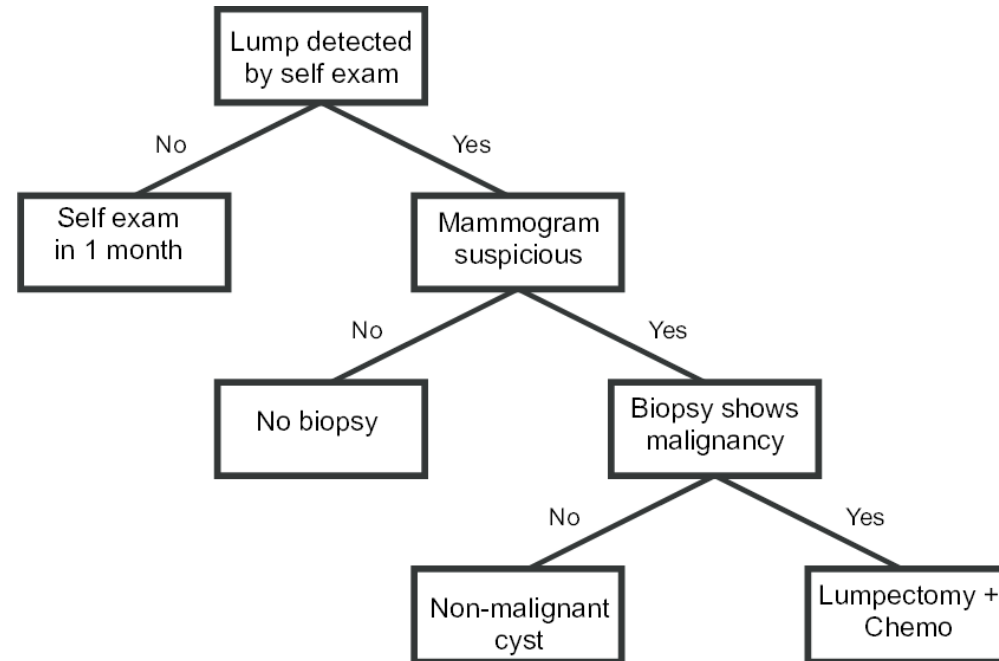


Flowcharts and decision trees based on *expert knowledge* or *learned from data*.

Rule-based Models

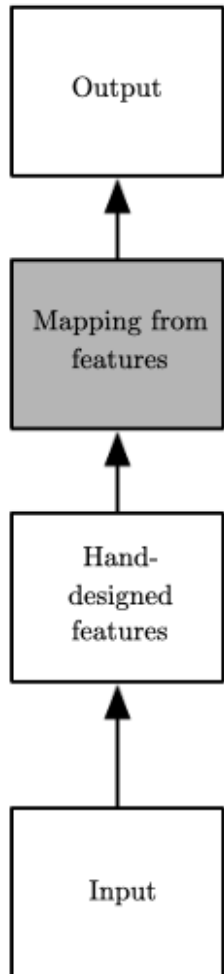


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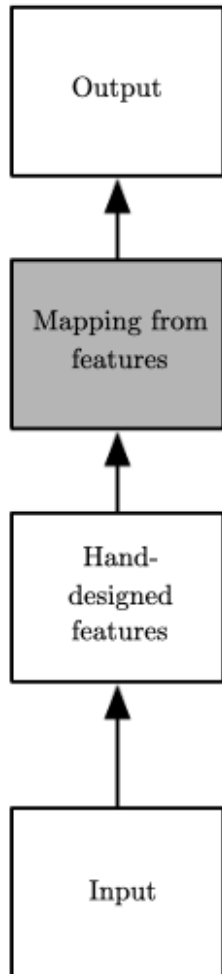
Classical Machine Learning

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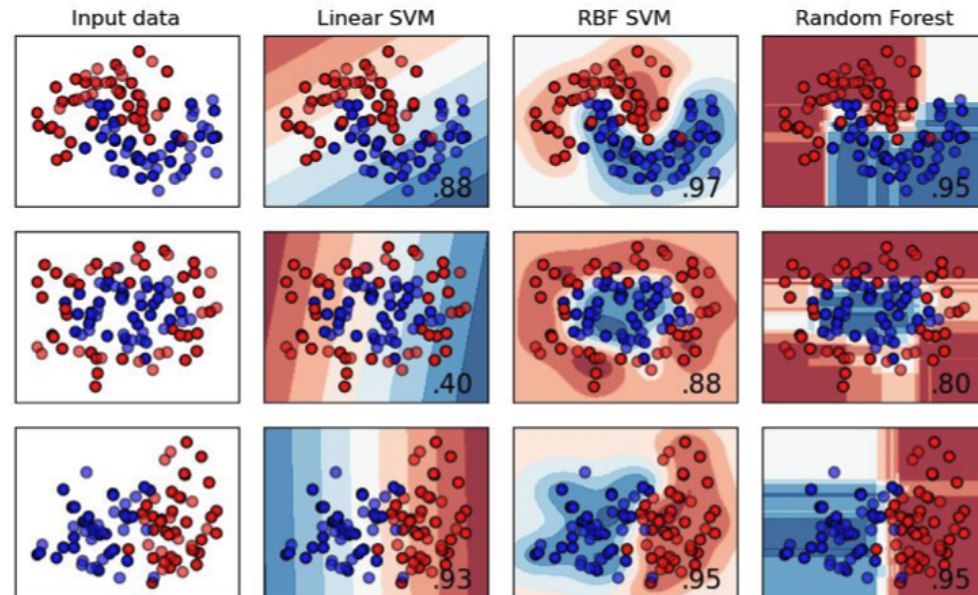


Models that learn from *structured features* such as what you find in the *SFIs* in the EHRs.

Classical Machine Learning

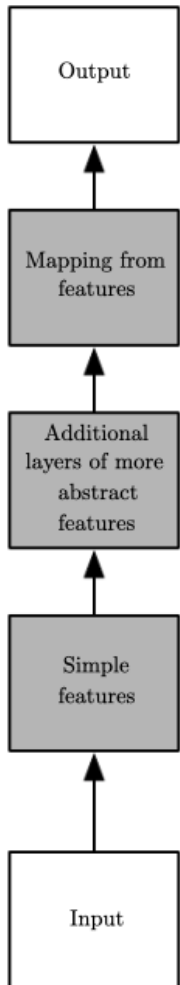


Models that learn from *structured features* such as what you find in the *SFIs* in the EHRs.



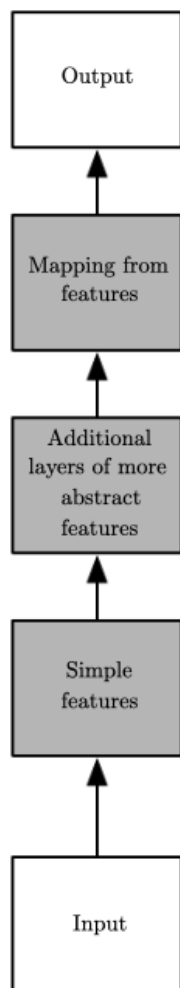
Representation/Deep Learning

Representation Learning



Models *create their own representation* from raw input data

Representation Learning



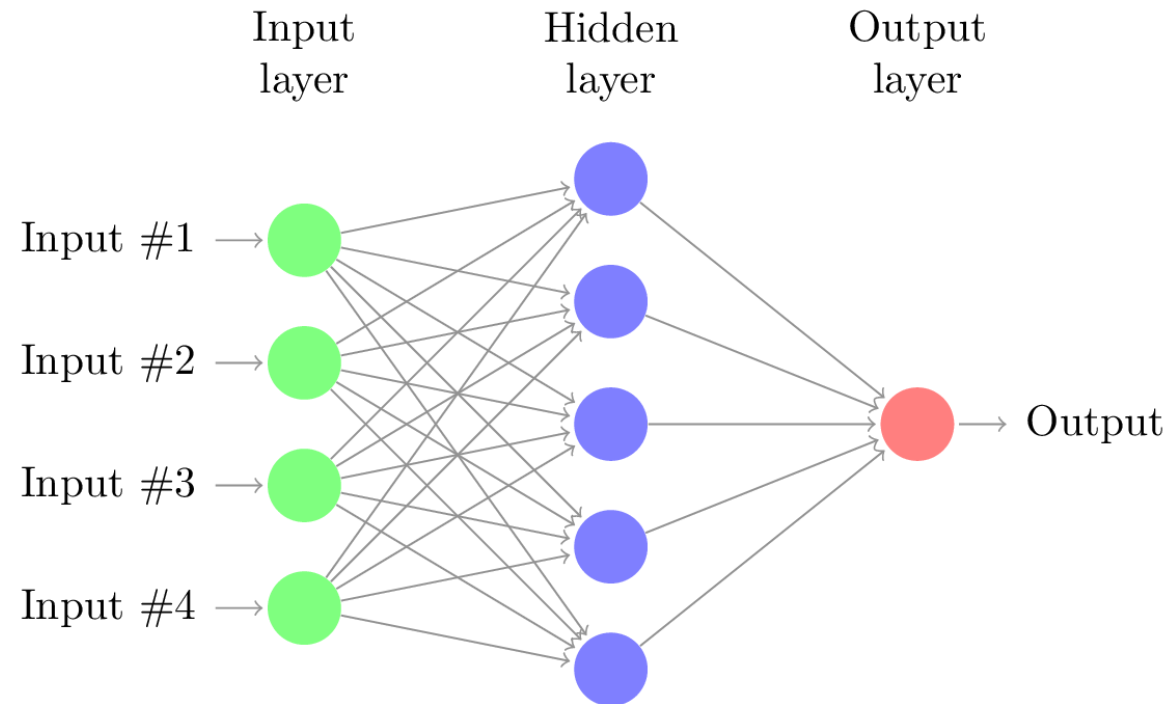
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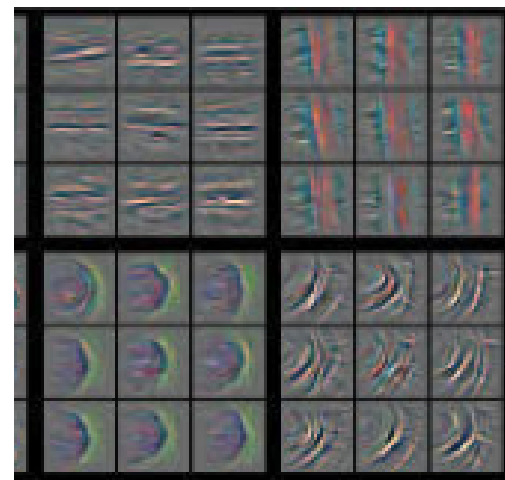
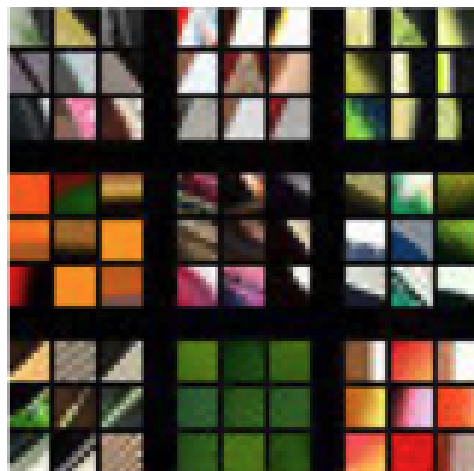
Neural Network Foundations

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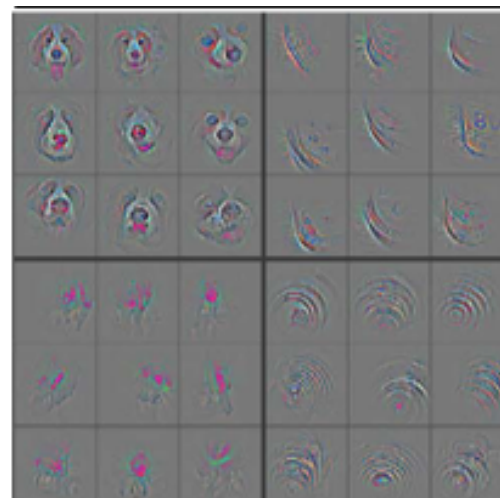
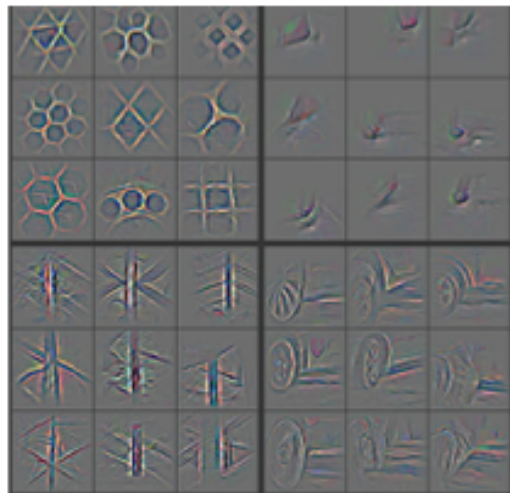
Multiple layers of *non-linear* processing



Neural Network Foundations



Neural Network Foundations



Natural Language Processing

Natural Language Processing

Computers don't like working text. We need to convert it to numbers somehow.

Solution? Word vectors.



Handcrafted Word vectors

it was the best of times
it was the worst of times
it was the age of wisdom
it was the age of foolishness

best	the	times	was	worst	age	wisdom	foolishness
1	1	1	1	0	0	0	0
0	1	1	1	1	0	0	0
0	1	0	1	0	1	1	0
0	1	0	1	0	1	0	1

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Context and semantics are completely disregarded!

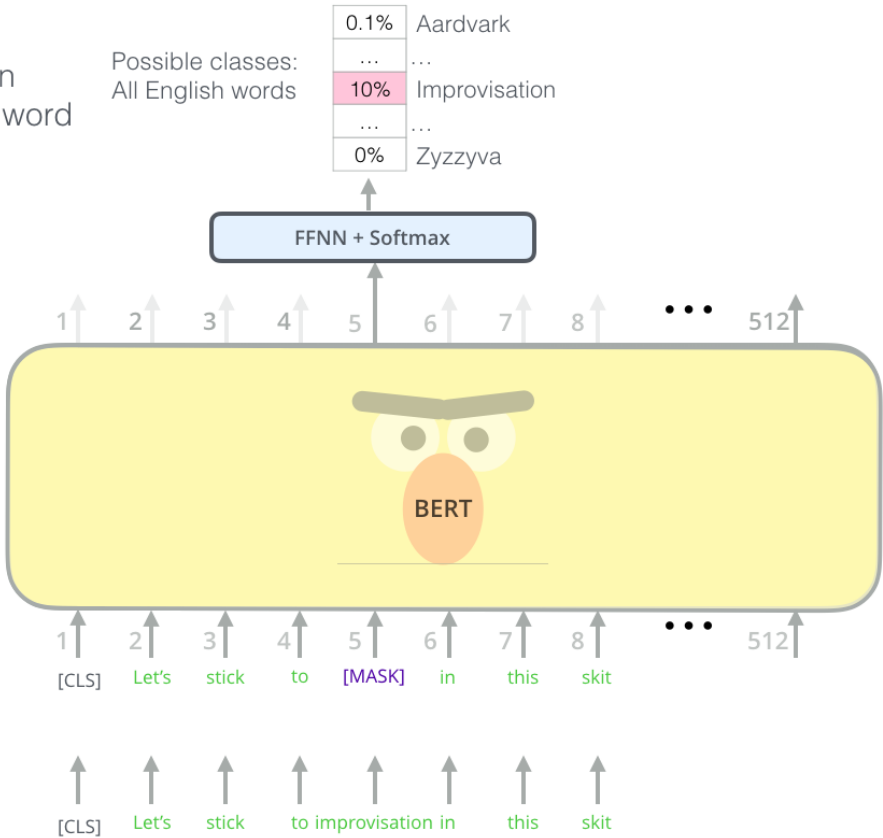
Instead, train a deep learning model to learn *contextualized word vectors*

BERT

Use the output of the masked word's position to predict the masked word

Randomly mask 15% of tokens

Input



Take Home Messages

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- Encompasses a wide range of models, and very useful for most tasks
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- Requires a lot of training data
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NLP

- How to turn text into meaningful representations for computers
- BERT learns *contextualised word representations*
- *Transfer learning* is key to the success of deep learning and NLP

Questions?