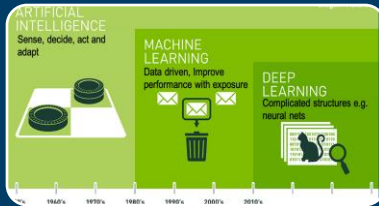


# INTRODUCTION TO MACHINE LEARNING - KEY CONCEPTS AND TERMINOLOGY

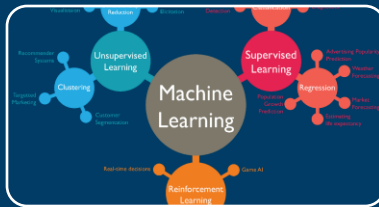
Milan De Cauwer – SINTEF Digital – Norway, Oslo

# Outline

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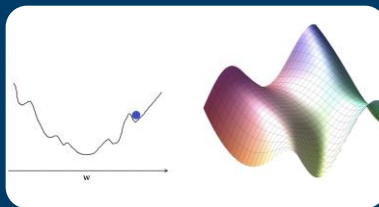


## Discussion on Machine Learning and AI



## Machine Learning – A definition

- Supervised learning
- Unsupervised learning
- Reinforcement learning



## Concepts in machine learning

- Data – Features, labels, instance, dataset and data splitting
- Models – Training and inference
- Bias and variance trade-off

# Who am I?

---



Milan  
SINTEF Digital

- Research scientist at SINTEF Digital, Norway
  - SINTEF is one of Europe's largest independent research organisations
  - Applied research in collaboration with industry



**Building and  
construction**



**Climate and  
environment**



**Digitalisation**



**Food and agriculture**



**Health and medicine**



**Materials**



**Microsystems and  
nanotechnology**



**Ocean space**



**Renewable energy**



**Society and security**



**Sustainable production**



**Transport and mobility**

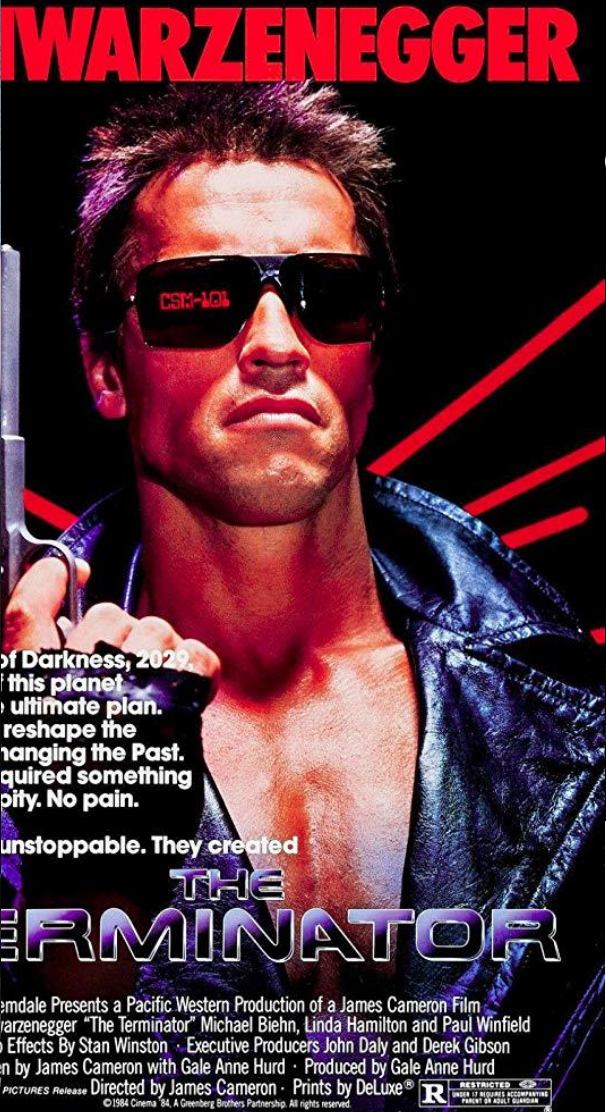
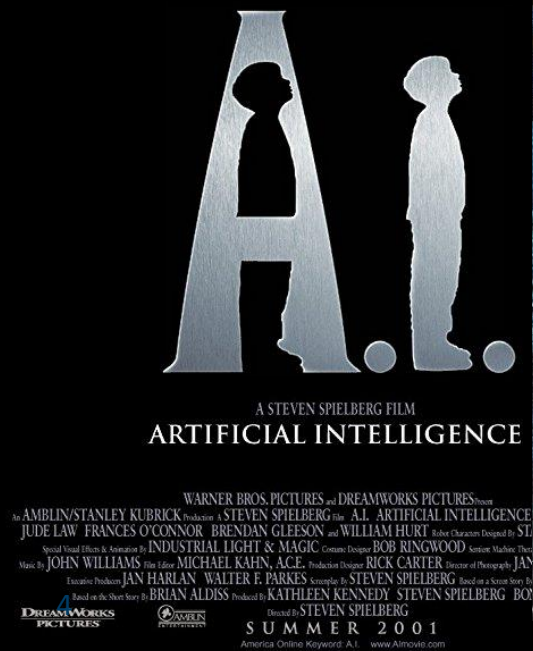
- My research interests

- Applications of **Machine Learning** for modeling and solving **combinatorial problems**



# AI in pop culture

David is 11 years old.  
He weighs 60 pounds.  
He is 4 feet, 6 inches tall.  
He has brown hair.  
  
His love is real.  
But he is not.



DOMHNALL GLEESON ALICIA VIKANDER and OSCAR ISAAC



# Examples of AI in daily life

Search  
engines



Filters



Media  
suggestions



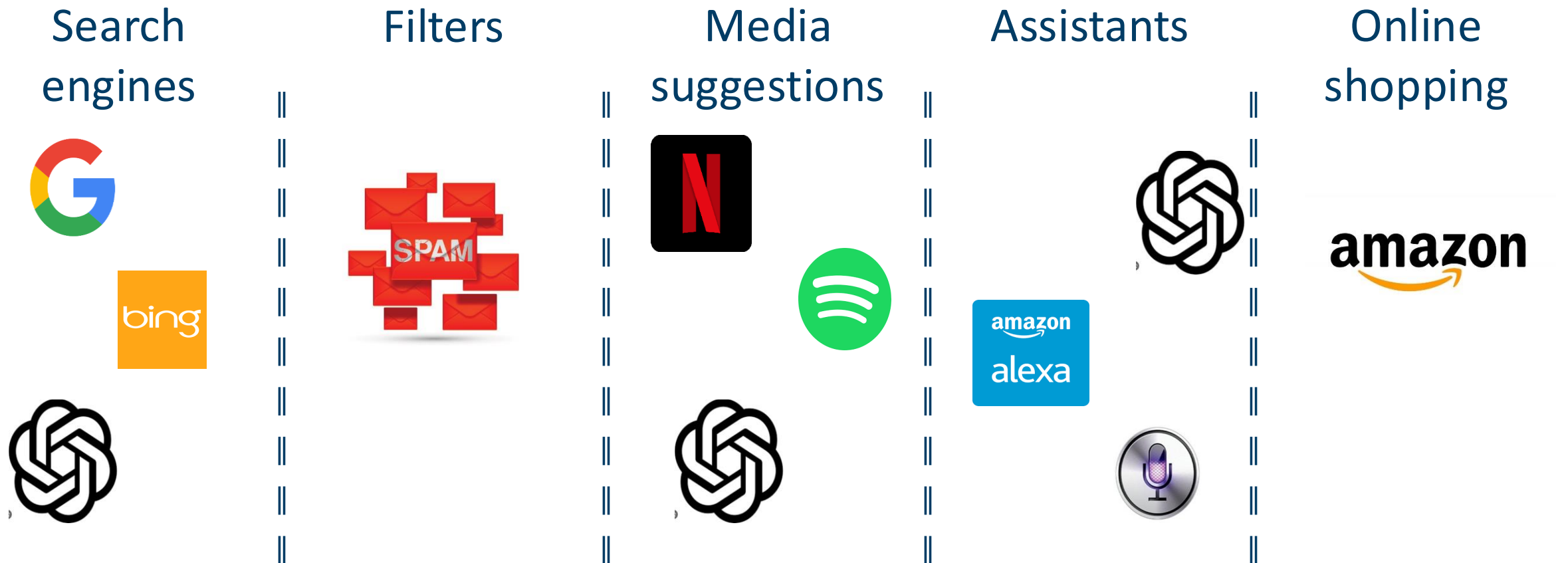
Assistants



Online  
shopping

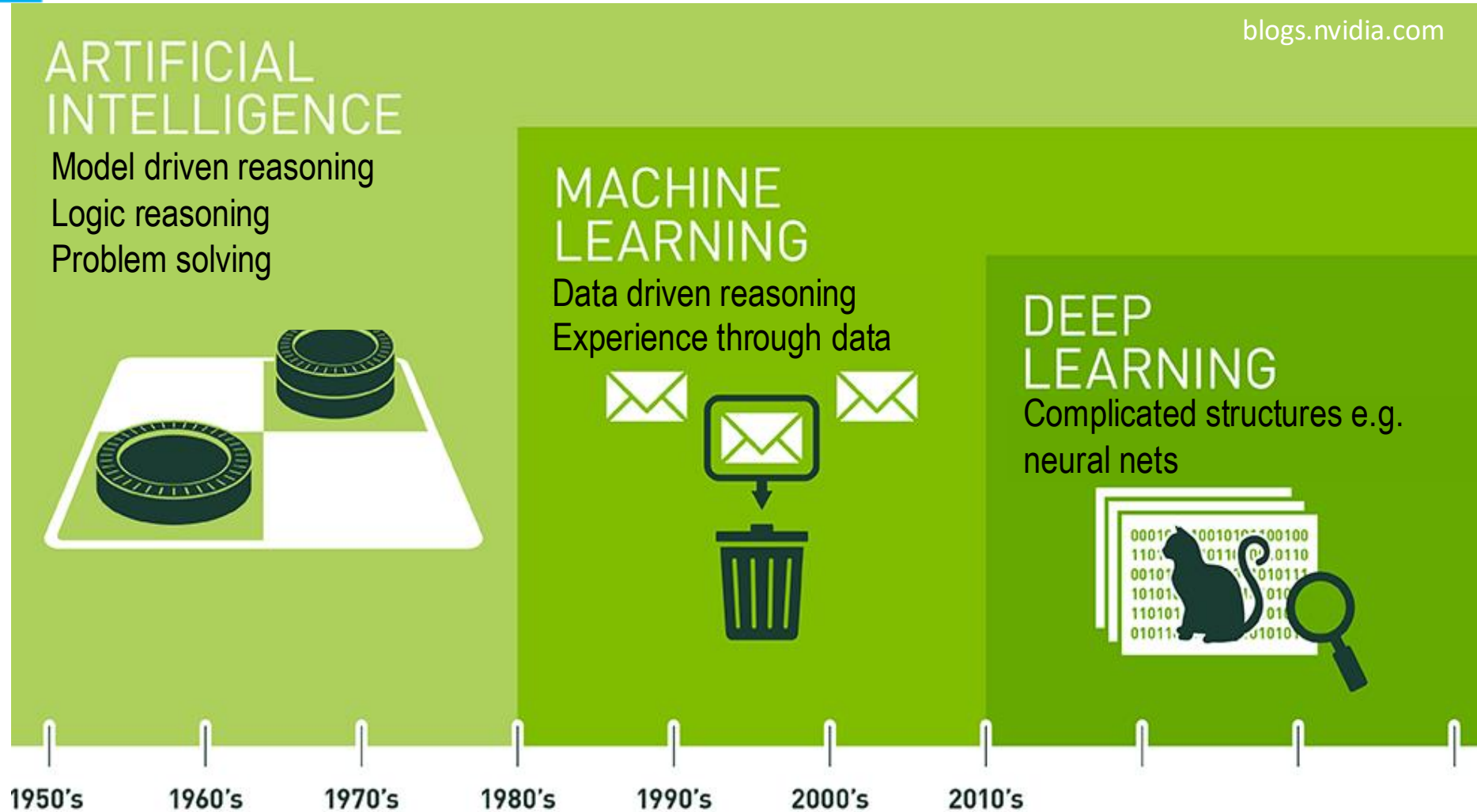


# Examples of AI in daily life – ChatGPT (Large language models)





# AI - Machine learning – Deep learning



# Why now?

Three paradigm shifts at once

Data



Technology



Analytics

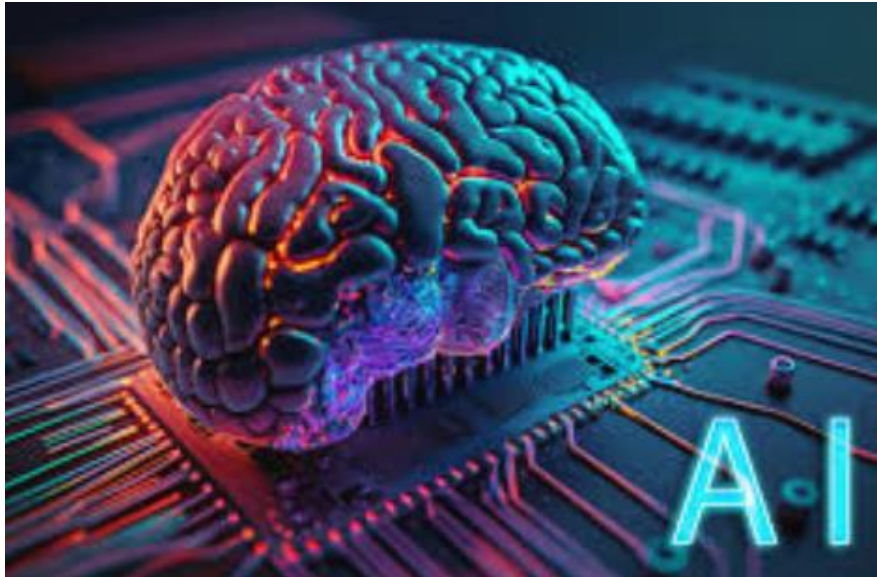


Machine learning



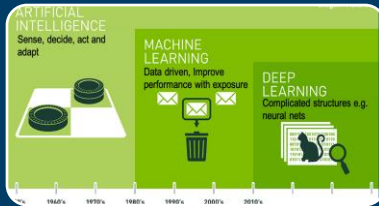
# Quiz – Do all methods in AI rely on machine learning? Why?

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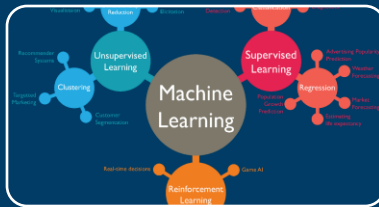


Machine Learning

# Outline

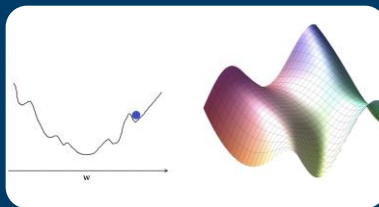


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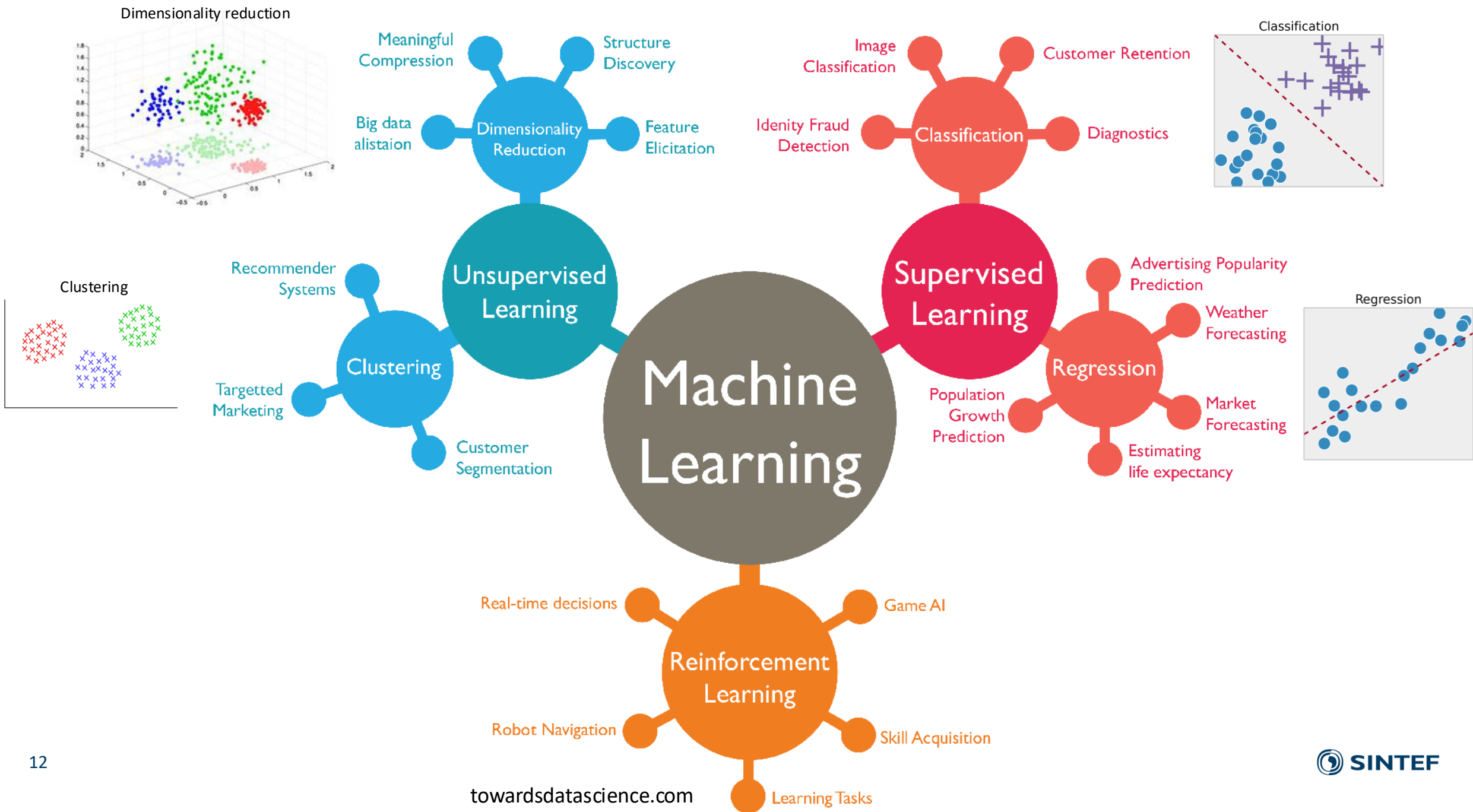
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# Machine-learning – A definition

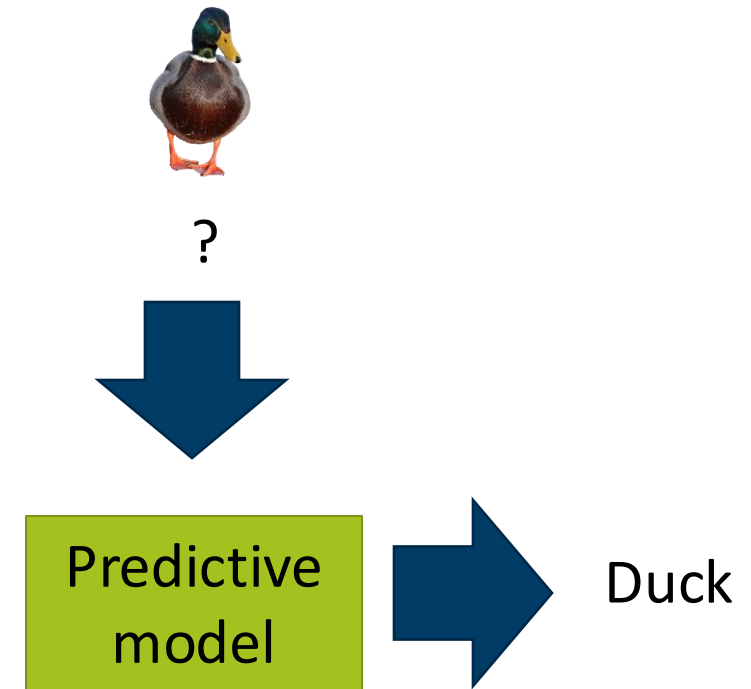
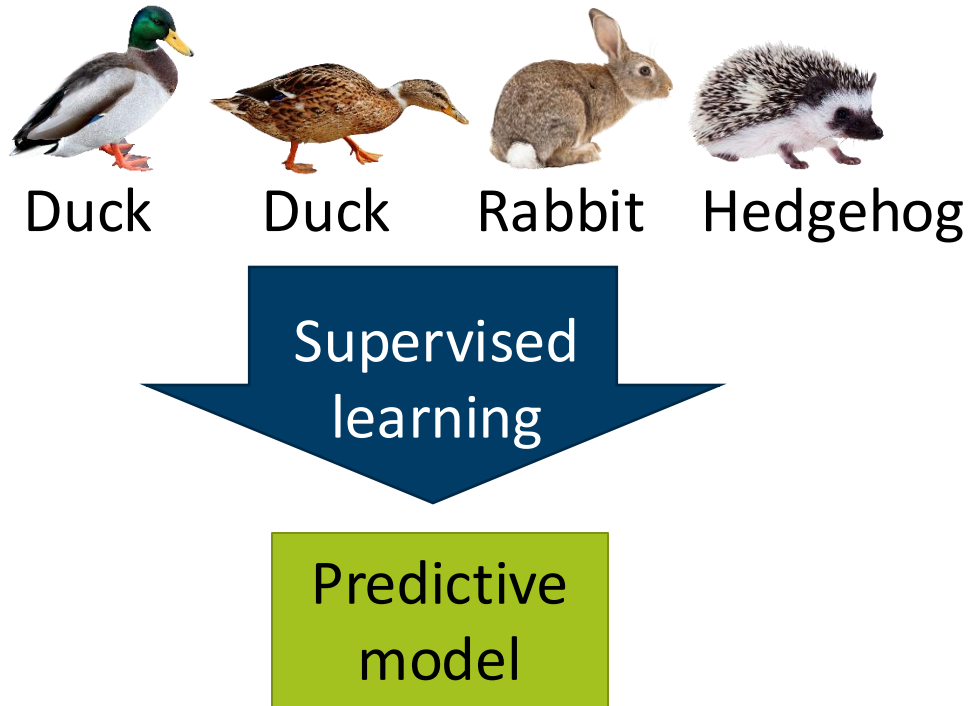
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- Definition:
  - A set of methods that enable computers to learn (something useful) from data
- Goal:
  - Build models that can make predictions or decisions based on data



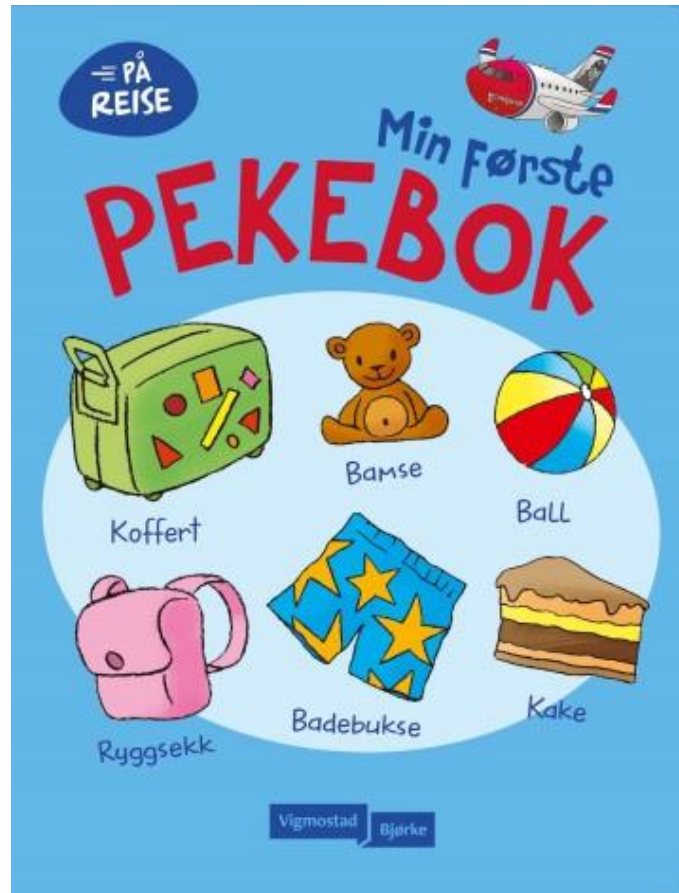


# Supervised learning



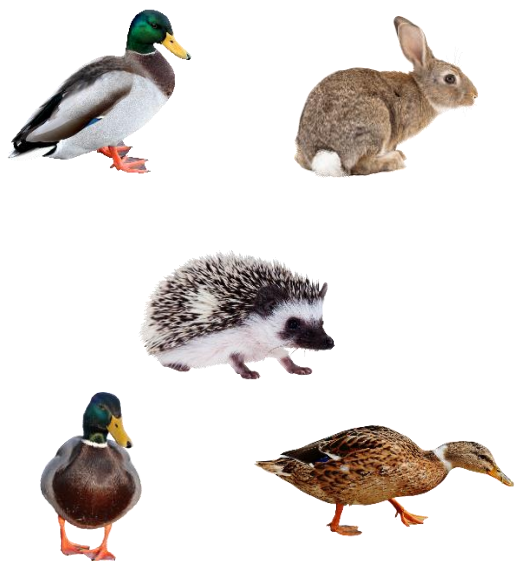
# Example of supervised learning

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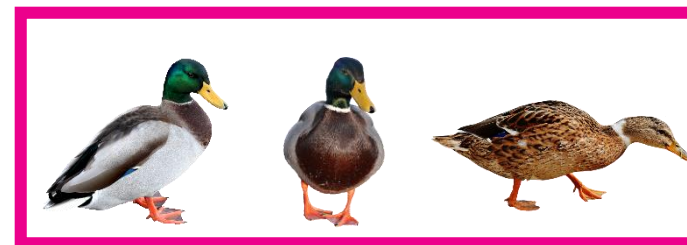


# Unsupervised learning

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Unsupervised  
learning



# Example on un-supervised learning

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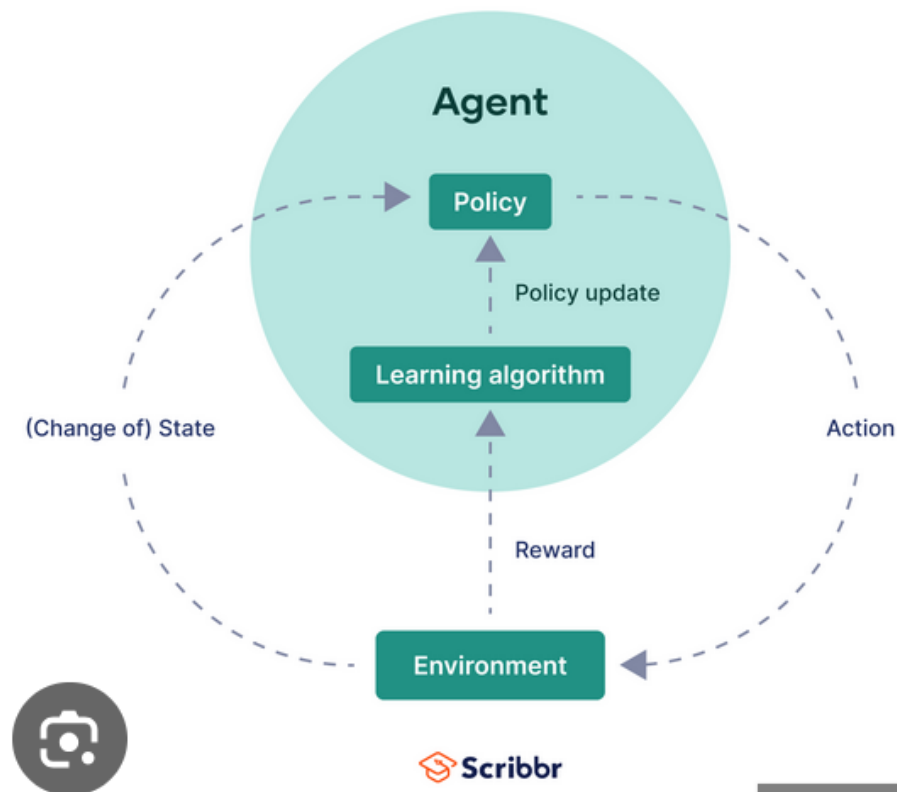
slate.com



# Reinforcement learning

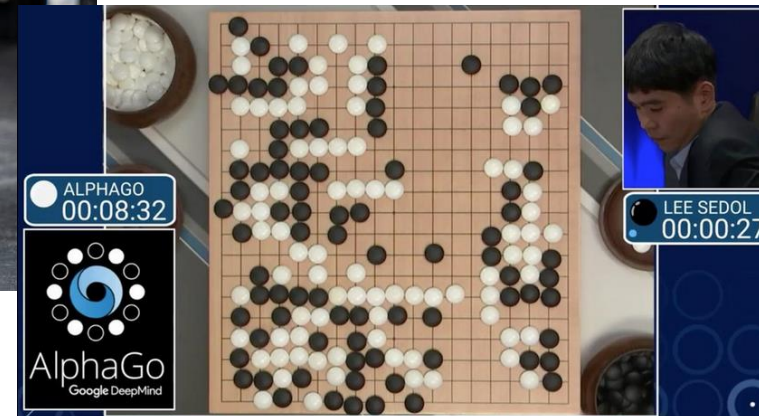
- Goal: learn how to interact with an environment to achieve a specific objective.
- The agent receives rewards (and penalties) depending on the quality of its actions

The general framework of reinforcement learning



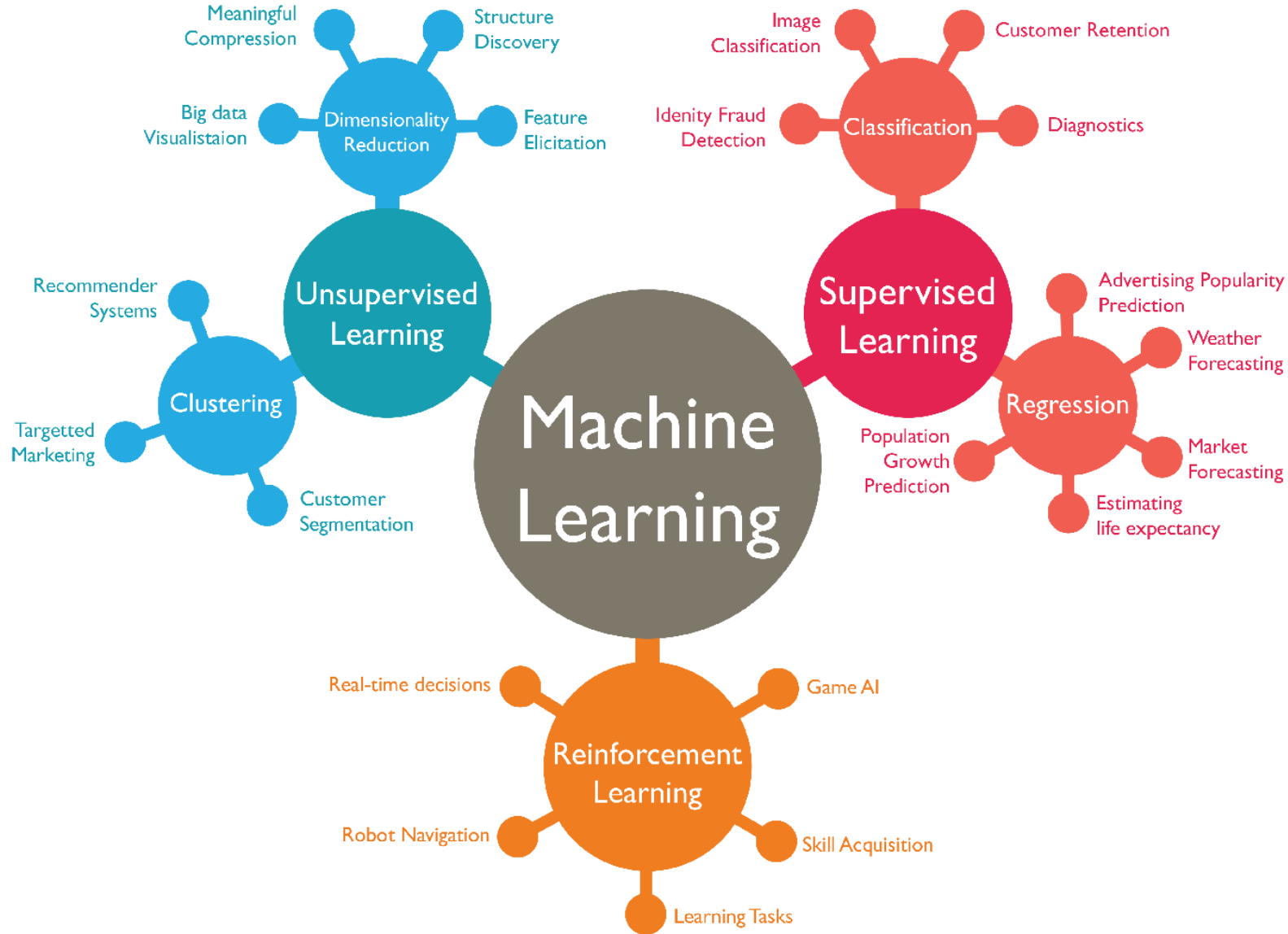
1 451 x 1 485

# Reinforcement learning examples: Self-driving cars, robotics and games



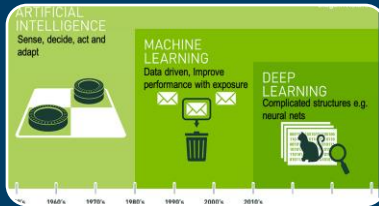
# Quiz – Use-cases

- **A:** Predicting whether the train will be late
- **B:** Teaching a computer to play angry birds
- **C:** Recommending your next dining experience.

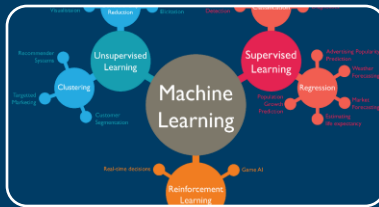


towardsdatascience.com

# Outline

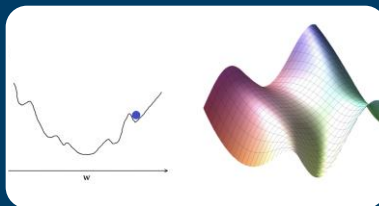


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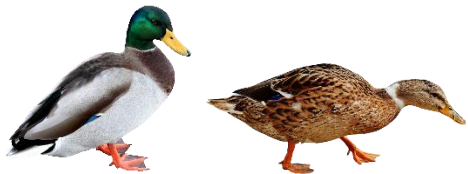
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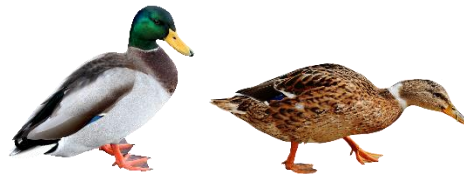


# Different kind of data

## Unlabelled data



## Labelled data



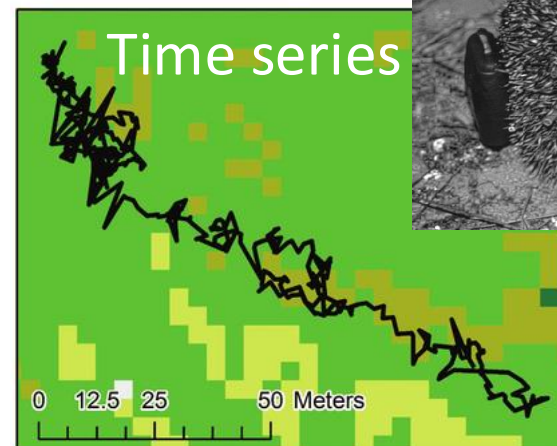
Duck Duck



Rabbit Hedgehog

## Tabular data

Image number	Animal	Size in cm	Favourite food	Lays eggs
1	Duck	43	weeds	True
2	Duck	41	seeds	True
3	Rabbit	55	lettuce	False
4	Hedgehog	33	eggs	True



# Variables and features

Observations | instances | data points

↑

Index

Variables

Categorical   Discrete   Continuous   Categorical   Binary

Sample number	Animal	Size in cm	Weight in KG	Favourite food	Lays eggs
1	Duck	43	2.1	weeds	True
2	Duck	41	2.4	seeds	True
3	Rabbit	55	1.7	lettuce	False
4	Hedgehog	33	1.1	eggs	True

↑   ↑   ↑   ↑   ↑

Class | label | response | dependent | y

Features | predictors | independent | explanatory | X

# Feature Engineering

$$\text{BMI} = \text{Weight} / \text{Size}^2$$

Observations | instances | data points

Variables

Sample number	Animal	Size in cm	Weight in KG	Favourite food	Lays eggs	BMI
1	Duck	43	2.1	weeds	True	9.75
2	Duck	41	2.4	seeds	True	7.11
3	Rabbit	55	1.7	lettuce	False	12.1
4	Hedgehog	33	1.1	eggs	True	27.2

Index

Class | label | response | dependent | y

Features | predictors | independent | explanatory | X

Categorical

Discrete

Continuous

Categorical

Binary

Continuous

# Quiz -- Variables

---

	age	income	marital_status	subscribed
0	56	5.99	divorced	0
1	69	3.72	married	0
2	41	7.30	married	1
3	58	3.53	single	1
4	21	4.53	divorced	0
5	18	7.54	divorced	1
6	64	4.53	married	0
7	68	3.54	married	0
8	68	5.31	single	1
9	47	4.02	married	0

- **Application:** predicting whether a customer will subscribe to a marketing campaign

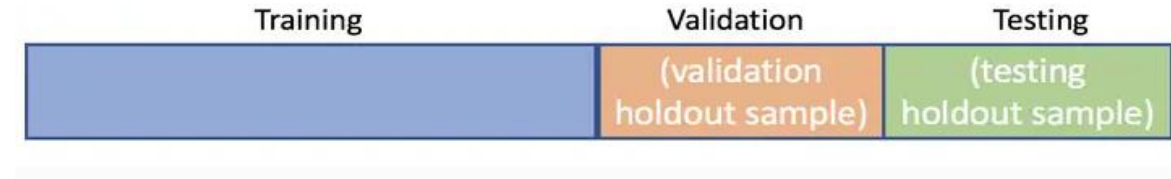
A - Find the target variable

B – What are the features?

C – Name a categorical feature.

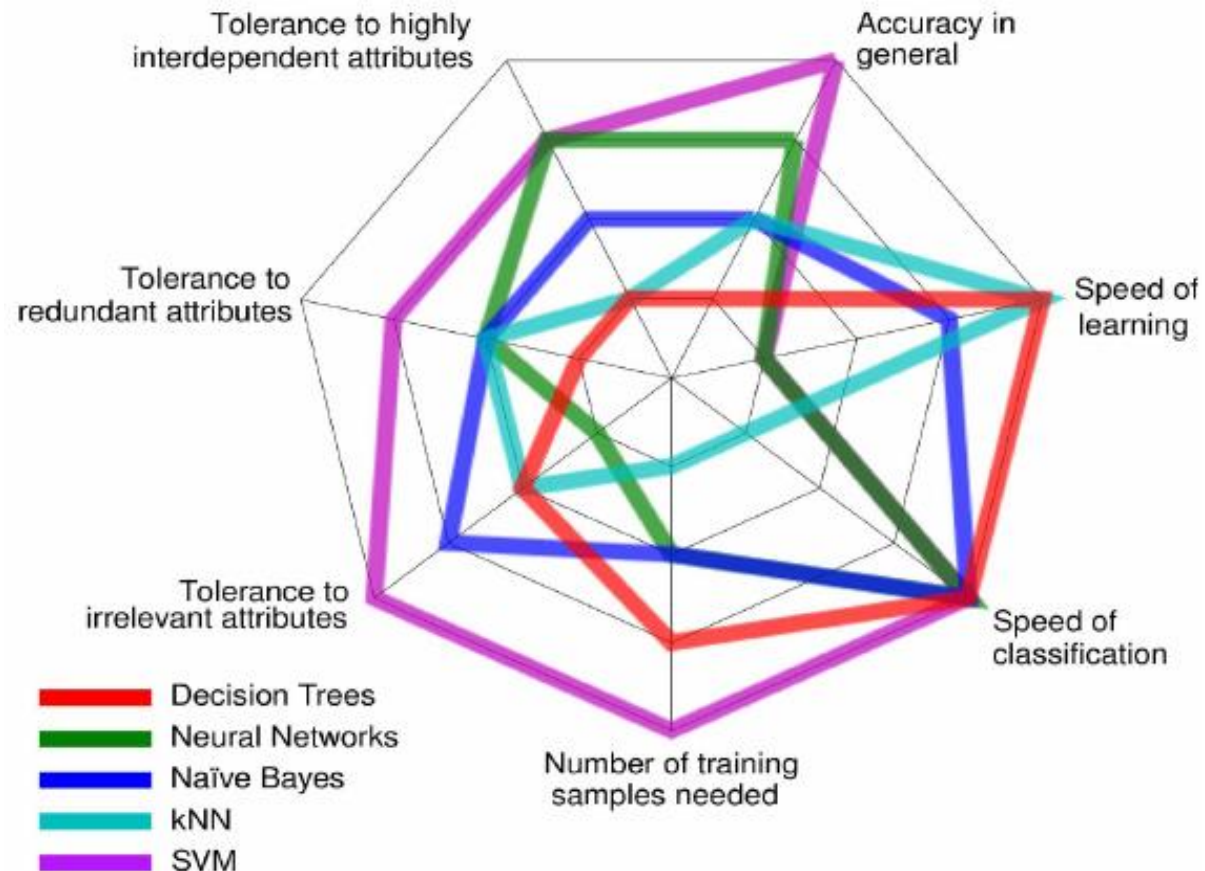
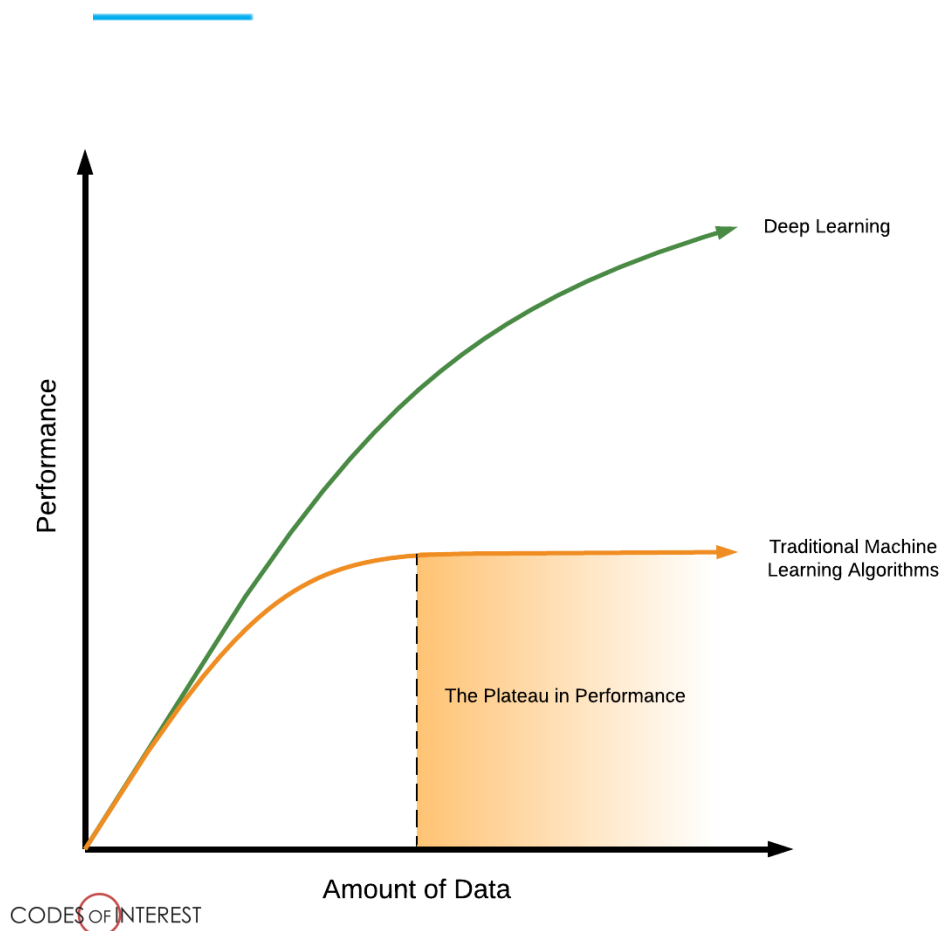


# Splitting data – Test, Train and validation data



- ideally
  - Data which we use to design our models (Training set)
  - Data which we use to refine our models (Validation set)
  - Data which we use to test our models (Testing set)
- If we do not split our data, we might test our model with the same data that we use to train our model
  - Beware of *train-test contamination*

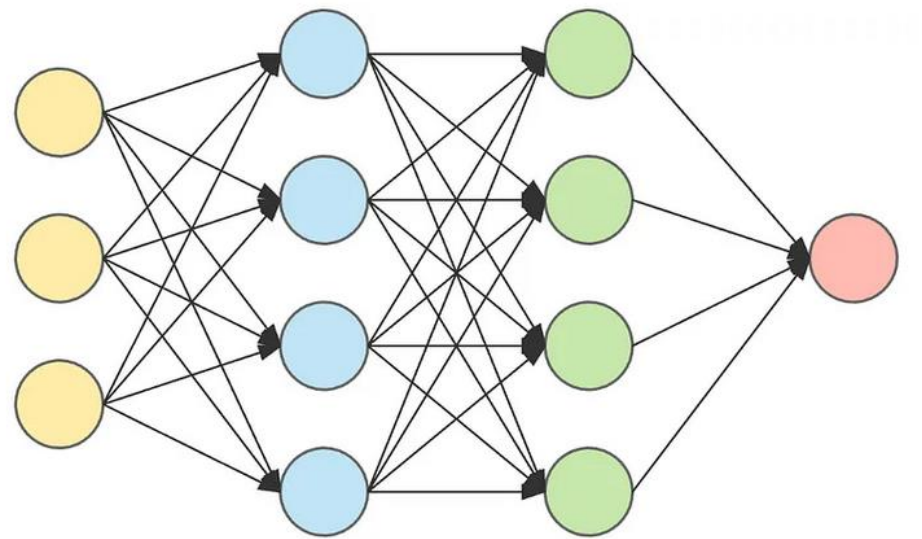
# Model Performance – No magic bullet



Widanapathirana et al. 2012

# Choosing an architecture and tuning hyperparameters

A simple building block

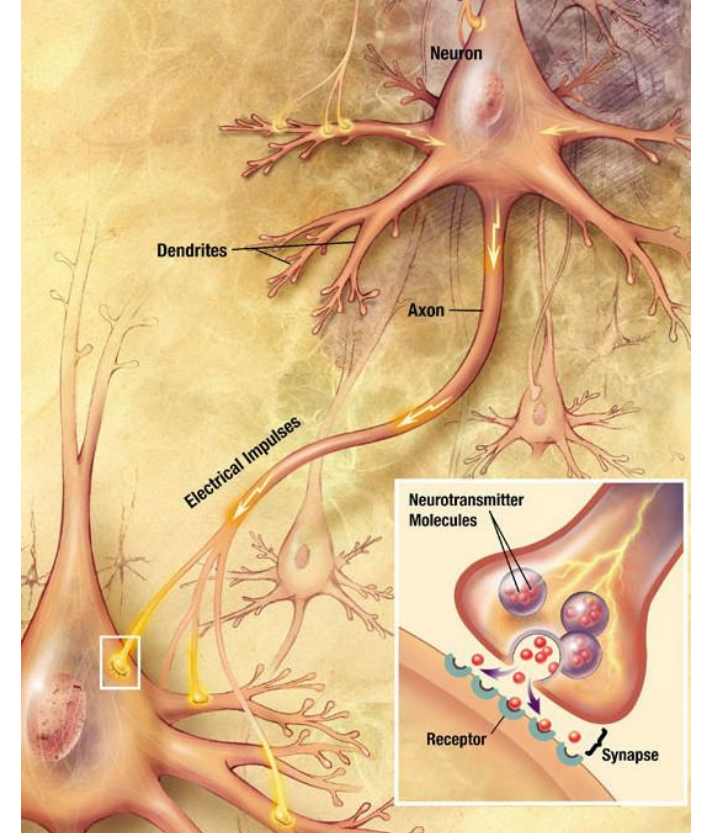


input layer

hidden layer 1

hidden layer 2

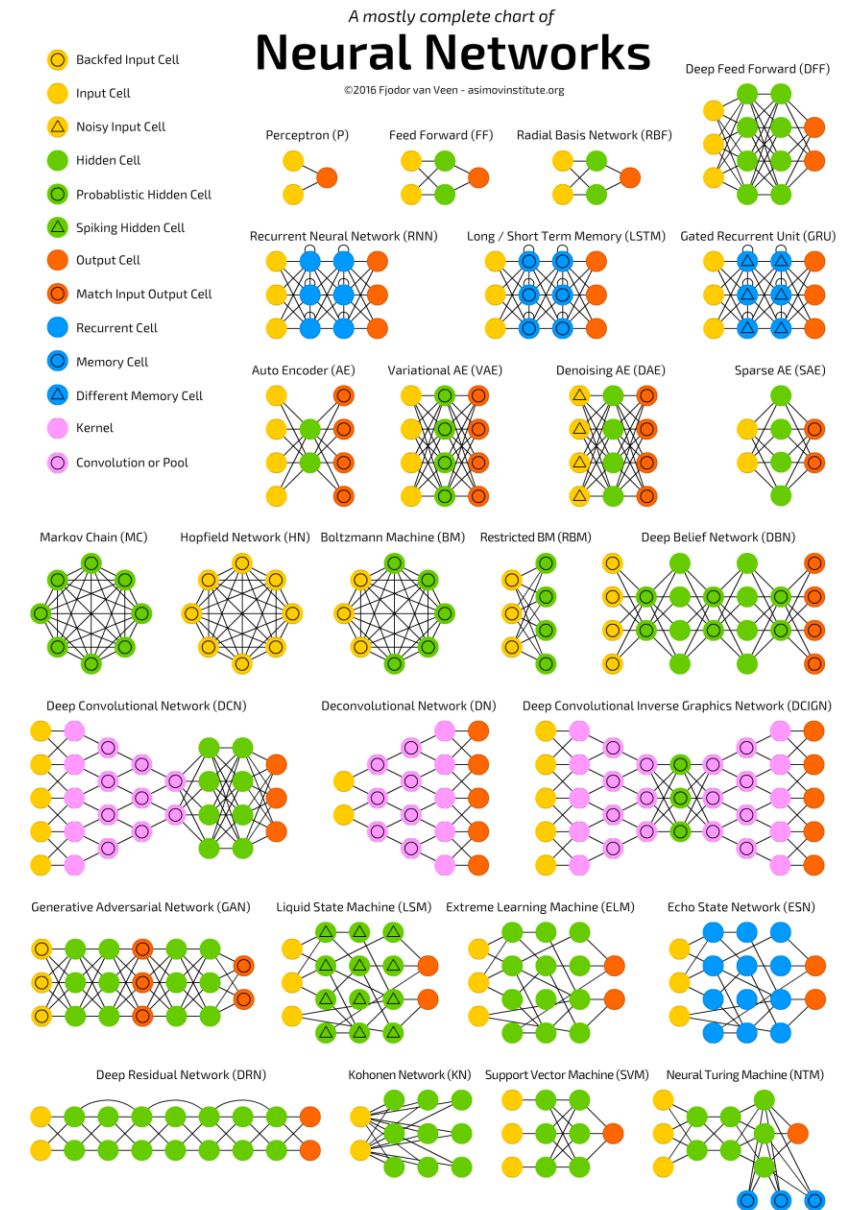
output layer



# Choosing an architecture and tuning hyperparameters

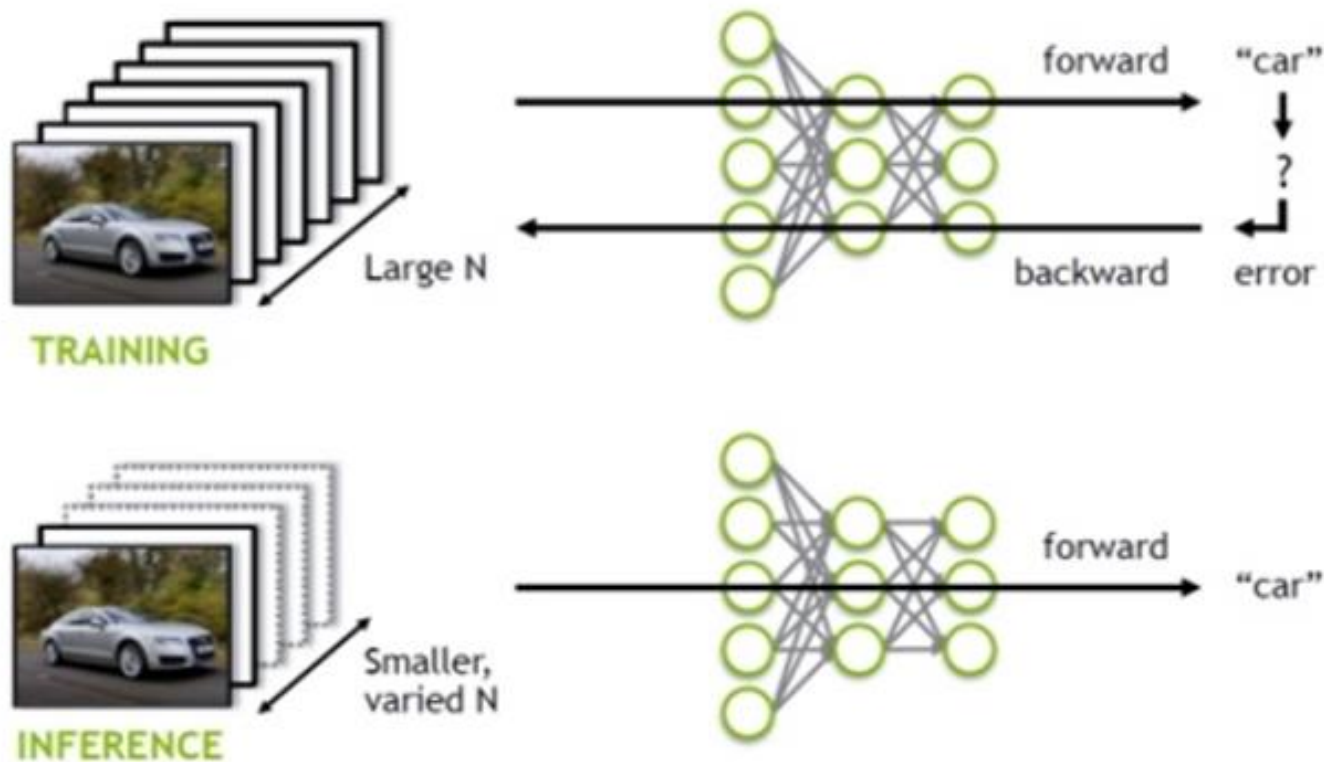
For larger systems, we need to decide:

- How many layers, how many unit?
- Activation function
- Global architecture
  - Convolutional neural nets (CNN) - Good choice for spatial data
  - Recurrent neural nets (RNN) - Good choice for temporal data
- How to connect the layers
- How to initialize weights
- Learning Rate
- Regularization Strength and technique



# Training a model VS using a model

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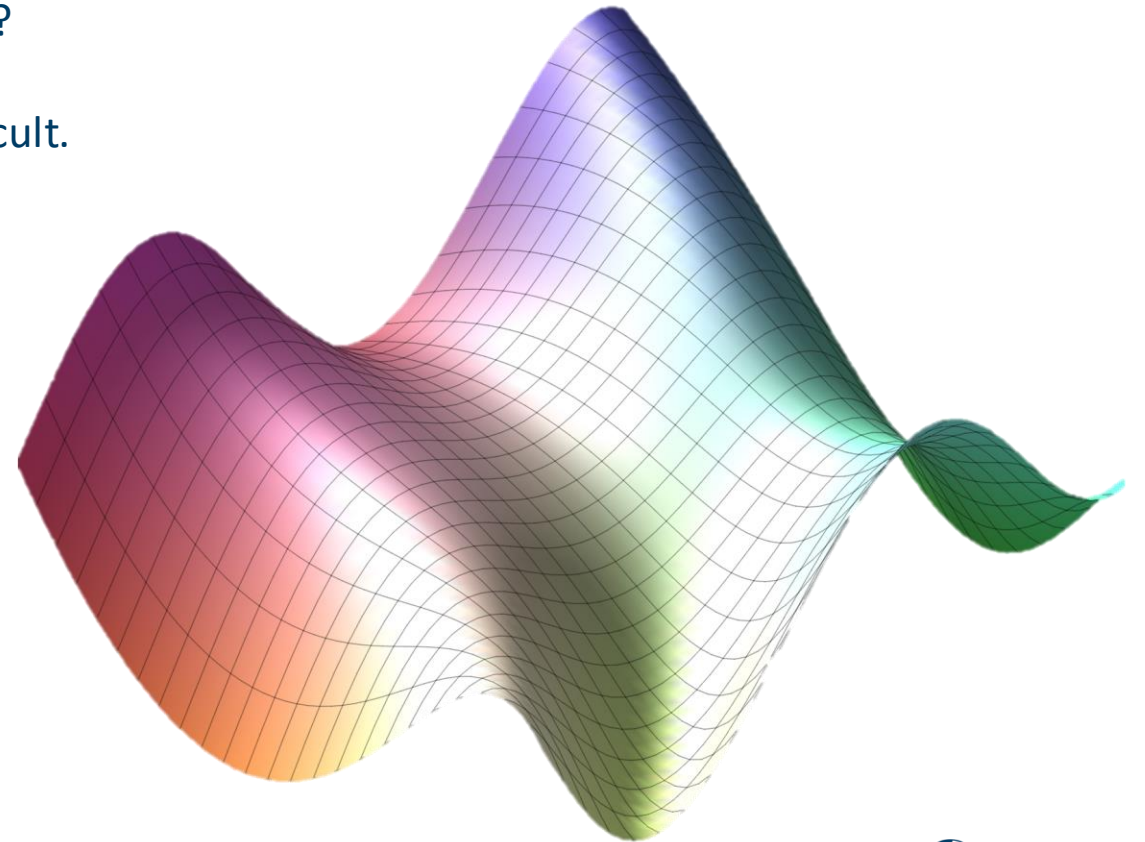
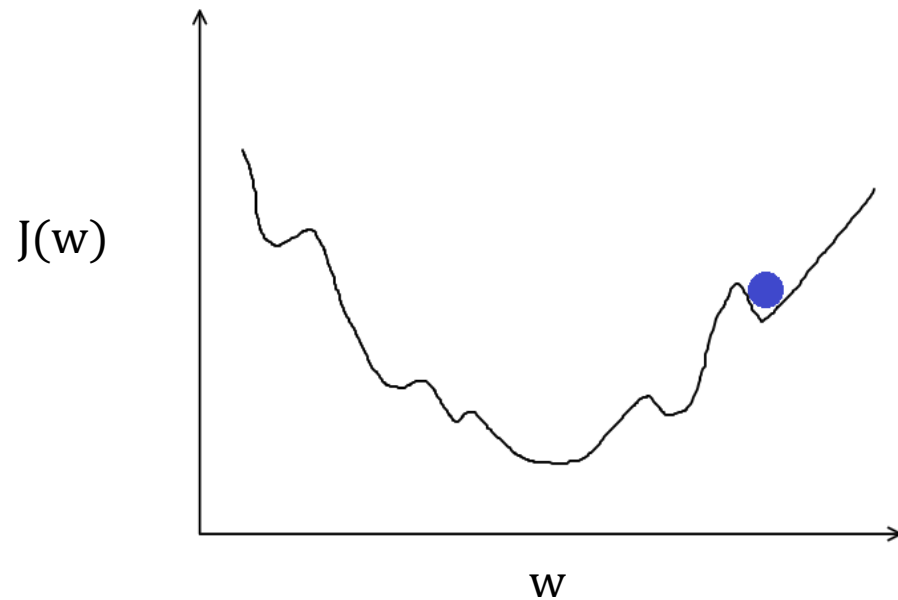


# Loss/cost/value function

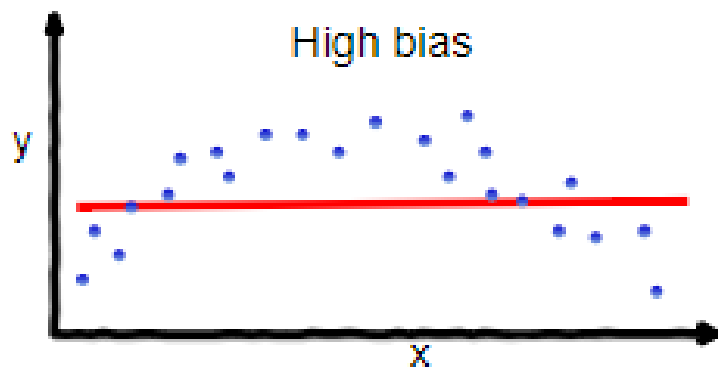
The loss function measures:

- how far is the model from the data/desired properties?

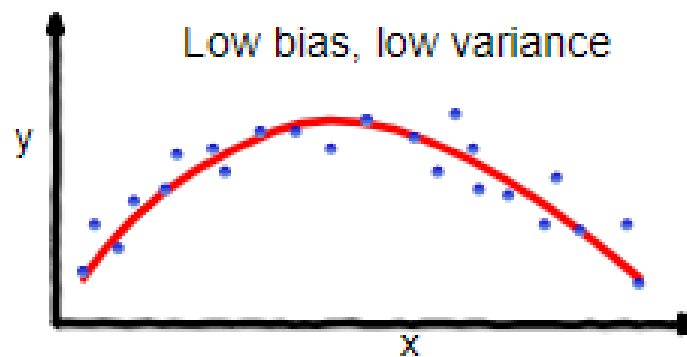
Finding the parameters leading to minimal loss can be difficult.



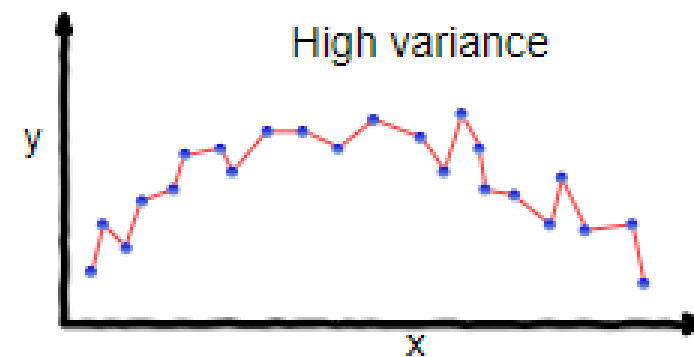
# Bias-variance tradeoff



Under fitting

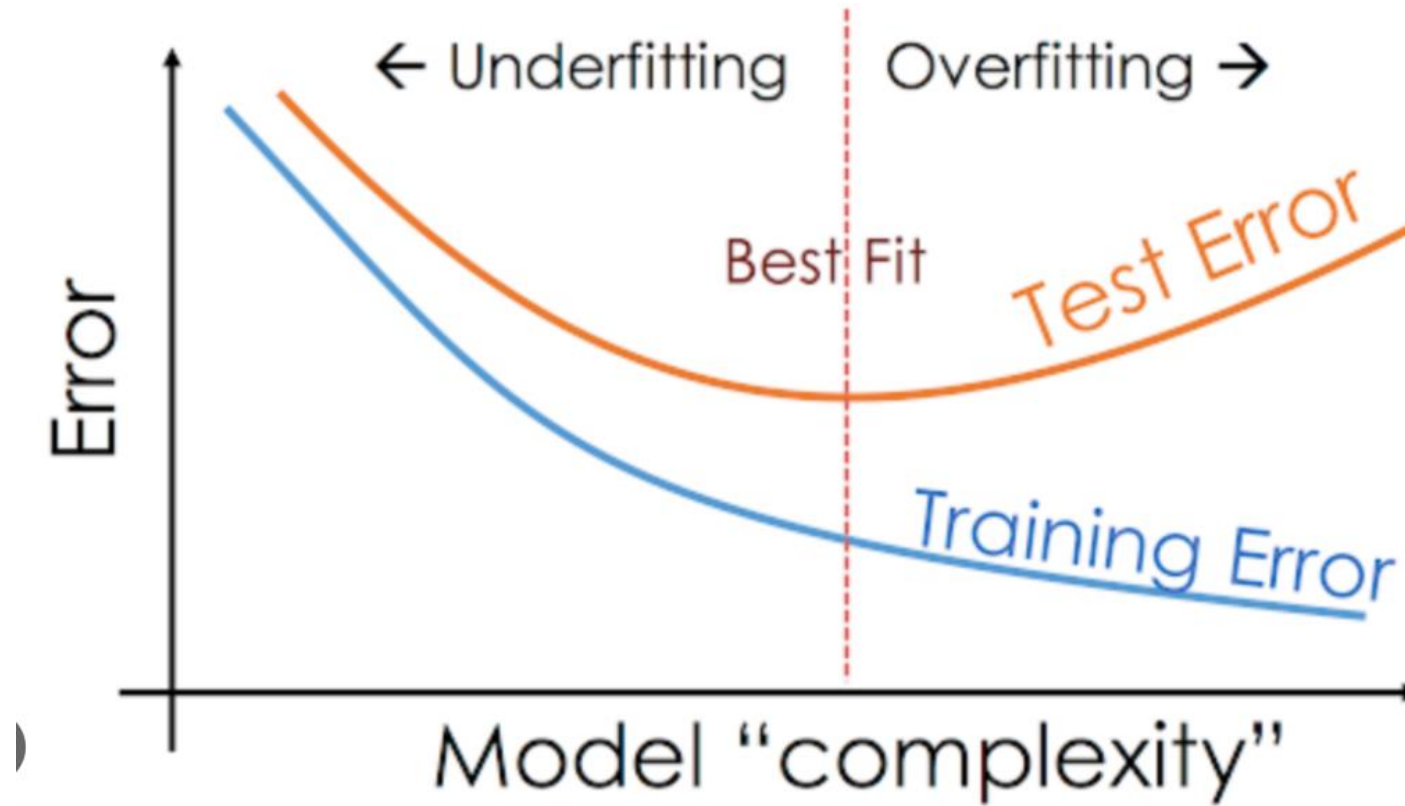


Good balance



Over fitting

# Bias-variance tradeoff

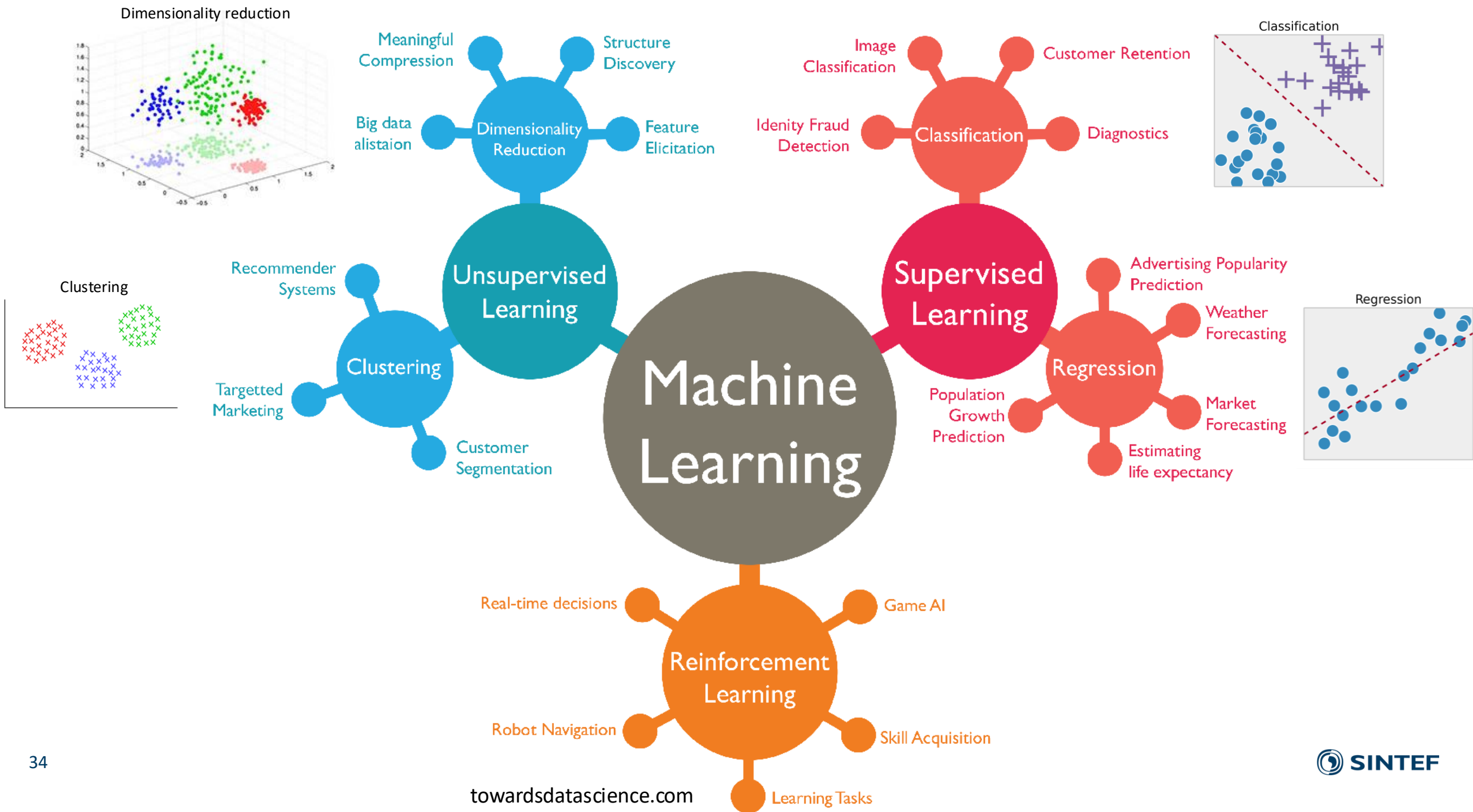


# Take home messages

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- Each context is different - many moving parts
- Experimentation is key
- Finding a method that works on your dataset is a science
  - (or perhaps it is an art)









Technology for a better society