

Deep Learning

MPHY0041 Machine Learning in Medical Imaging

Yipeng Hu
yipeng.hu@ucl.ac.uk

Introduction | Admin

Content

- Introduction
 - Admin; Background.
- Deep Neural Networks
 - Supervised learning; MLP; Activation functions; Loss functions; Backpropagation.
- Network Architecture
 - Width and depth; Branching, joining and Skipping (Shortcuts, skip layers, ResNet, DenseNet); Convolutional neural networks; Sampling; Examples (U-Net; PointNet; Attention mechanism, vision transformer); Recurrent neural networks.
- Training and Regularisation
 - Optimisation; Constrain parameters (norms, sharing); Augment data; Add randomness; Ensemble; Semi-supervised learning; Multi-task learning
- Selected Research Topics*
 - Multiple instance learning; Adversarial learning and generative models; Domain adaptation; Meta-learning.
- Medical Imaging Applications
 - Classification; segmentation; registration; synthesis.

- Hands-on tutorials

<https://weisslab.cs.ucl.ac.uk/yipenghu/mphy0041>

- Learning materials - `mphy0041/docs`

- Supported development environment
- Learning materials for TensorFlow and PyTorch
- Reading list

- Objectives

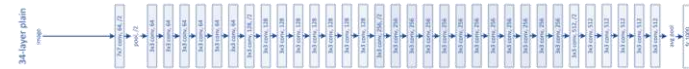
- Basic deep learning knowledge
- Hands-on experience with supervised convolutional neural networks

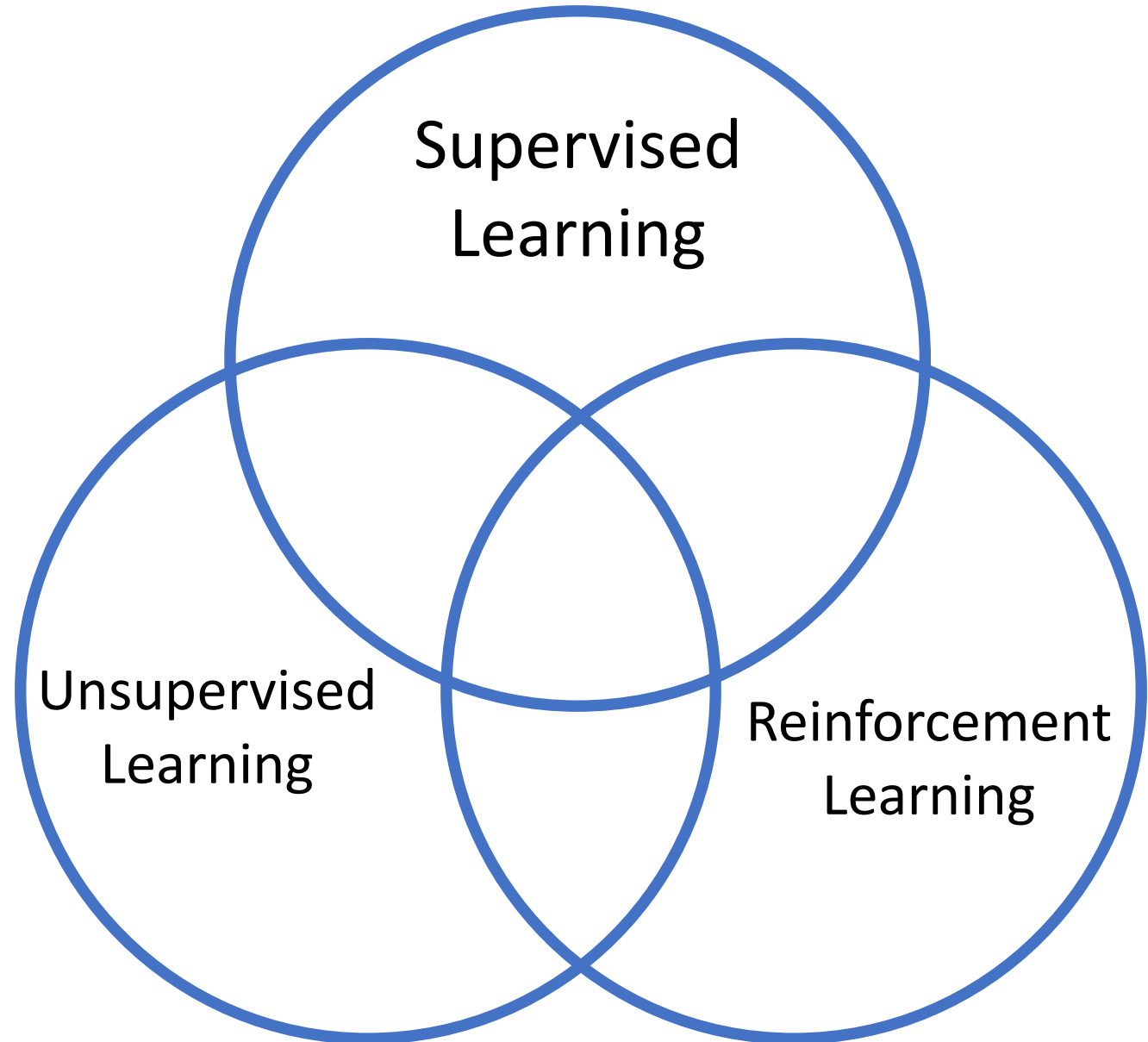
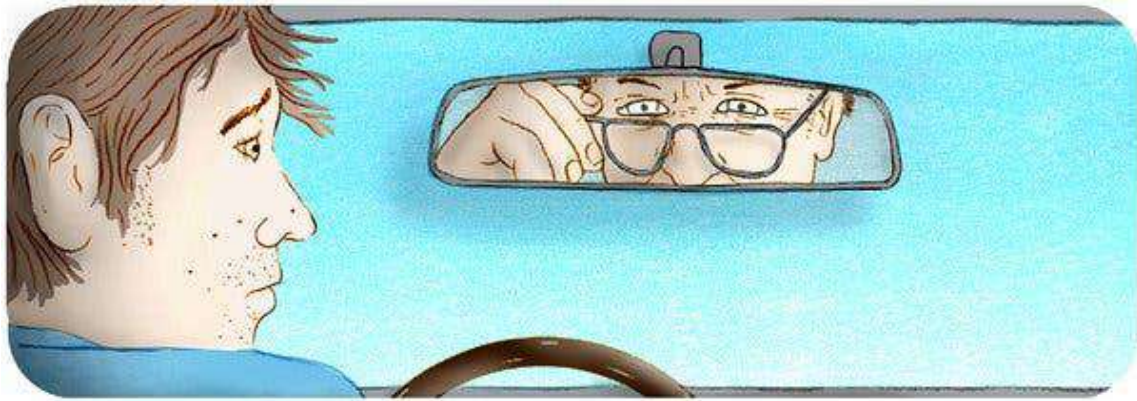
- Prerequisites

- Machine learning basics
- Maths: linear algebra, calculus, probability and statistics
- Programming: Python, TensorFlow/PyTorch

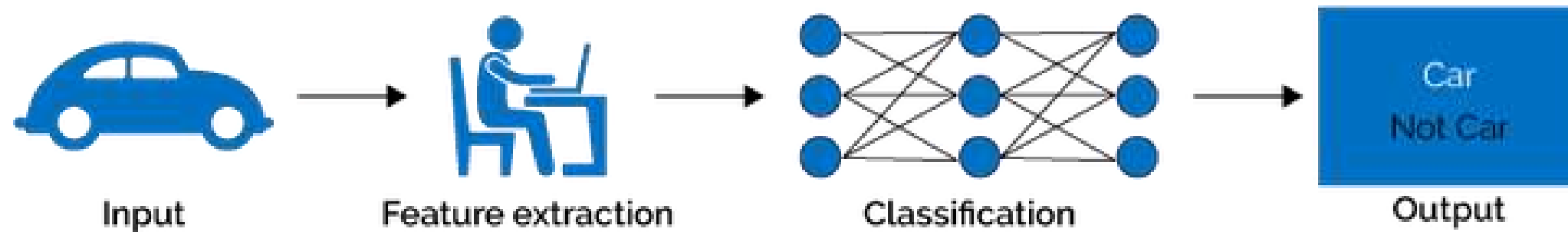
- **Live sessions**
 - 12:00-13:00 on Tuesdays
 - 10:00-11:00 on Fridays
- **Lecture videos**
 - Organised by topics, posted at the “weekly tabs”
- **Flexible office hours**
 - Emails or message teams (yipeng.hu@ucl.ac.uk)
 - Book a time (15-30 min) with me or one of the tutors within a week time
- **Formative assessment**
 - Associated with hands-on tutorials
 - Feedback are available but not compulsory
- **Coursework (groupwork)**
 - Programming - TensorFlow 2 / PyTorch
 - Solve real-world clinical problems
 - Submit code in Git repository
 - Report results in the form of a scientific paper

Introduction | Background

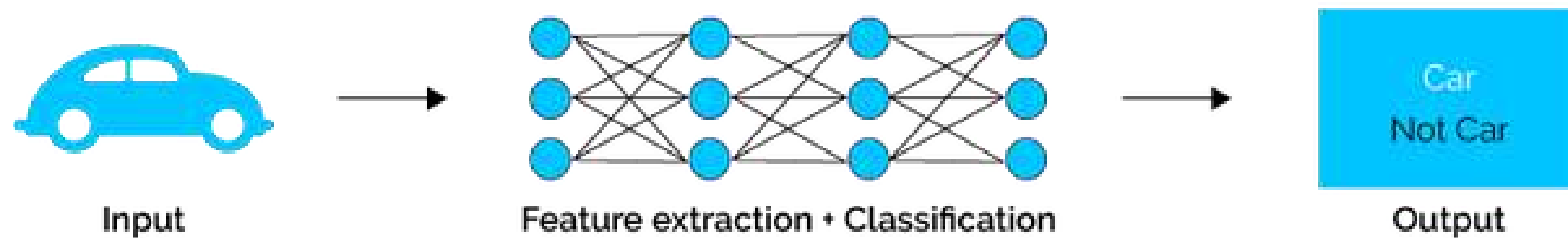


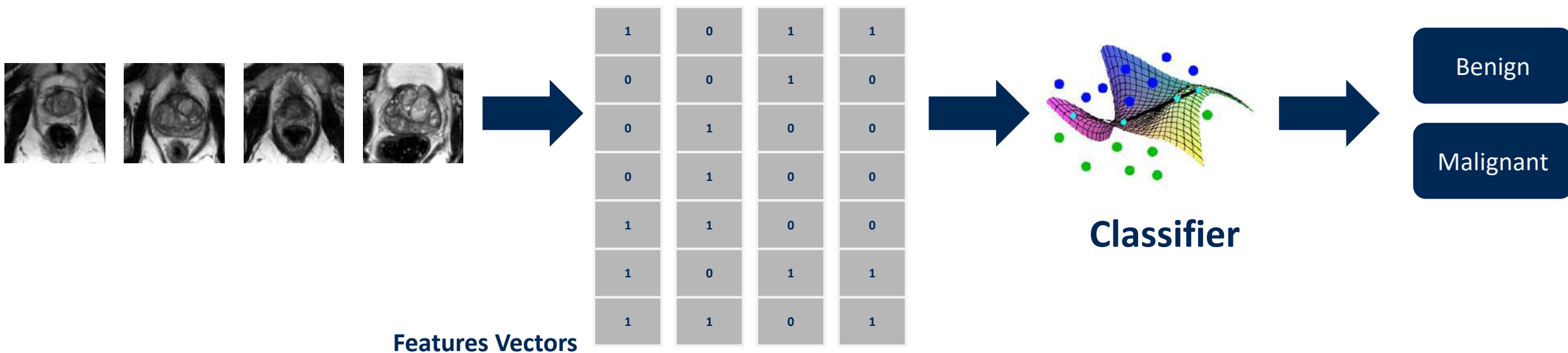


Machine Learning



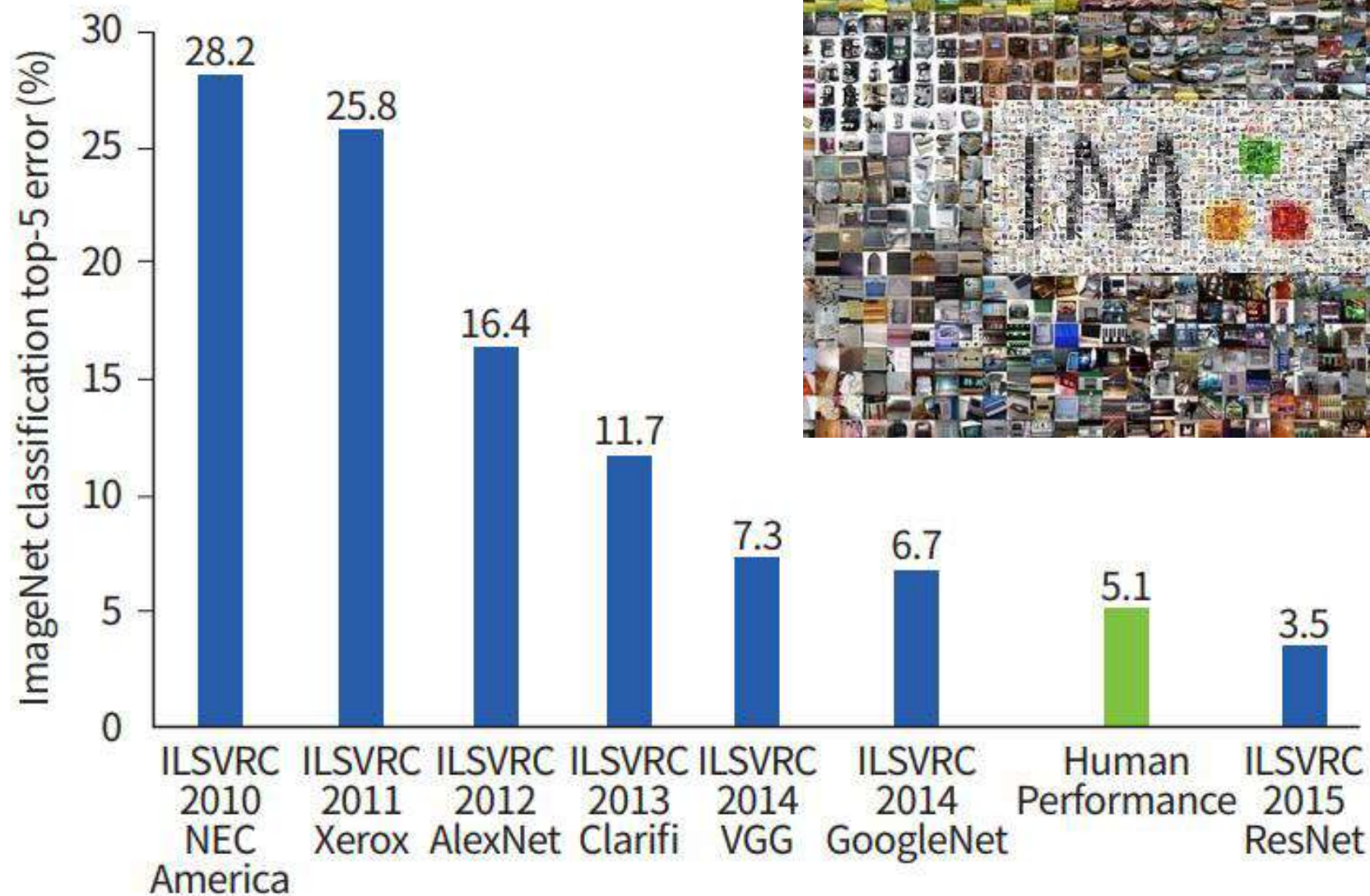
Deep Learning



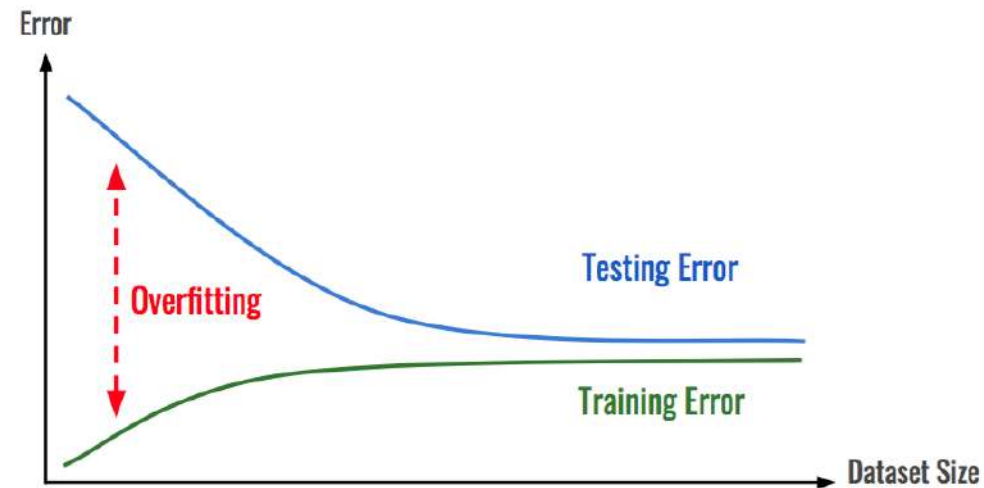
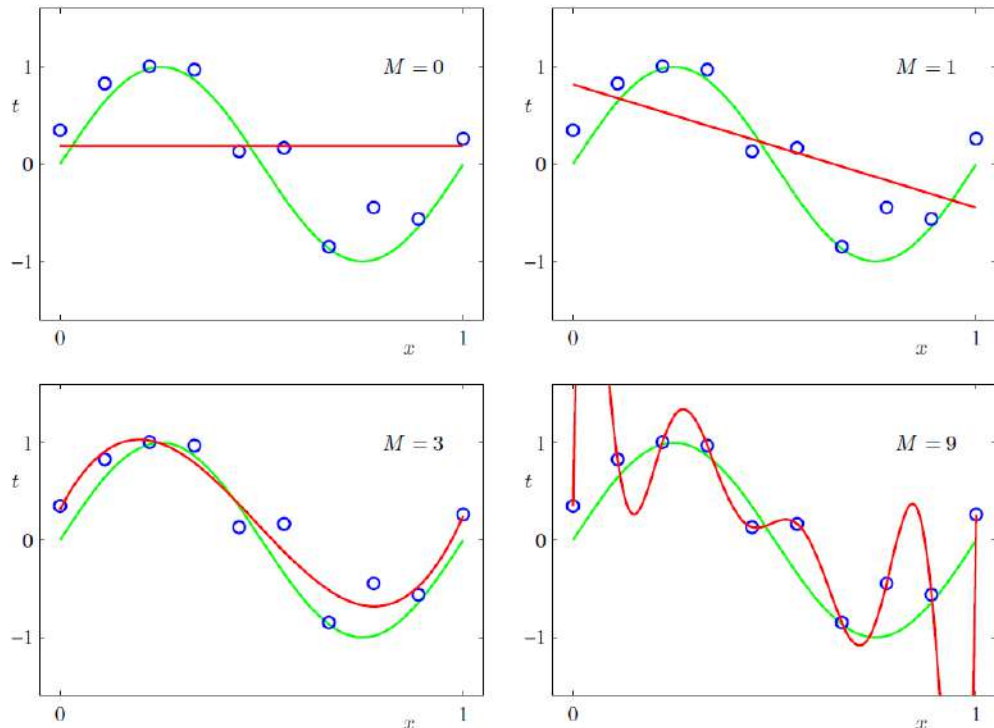


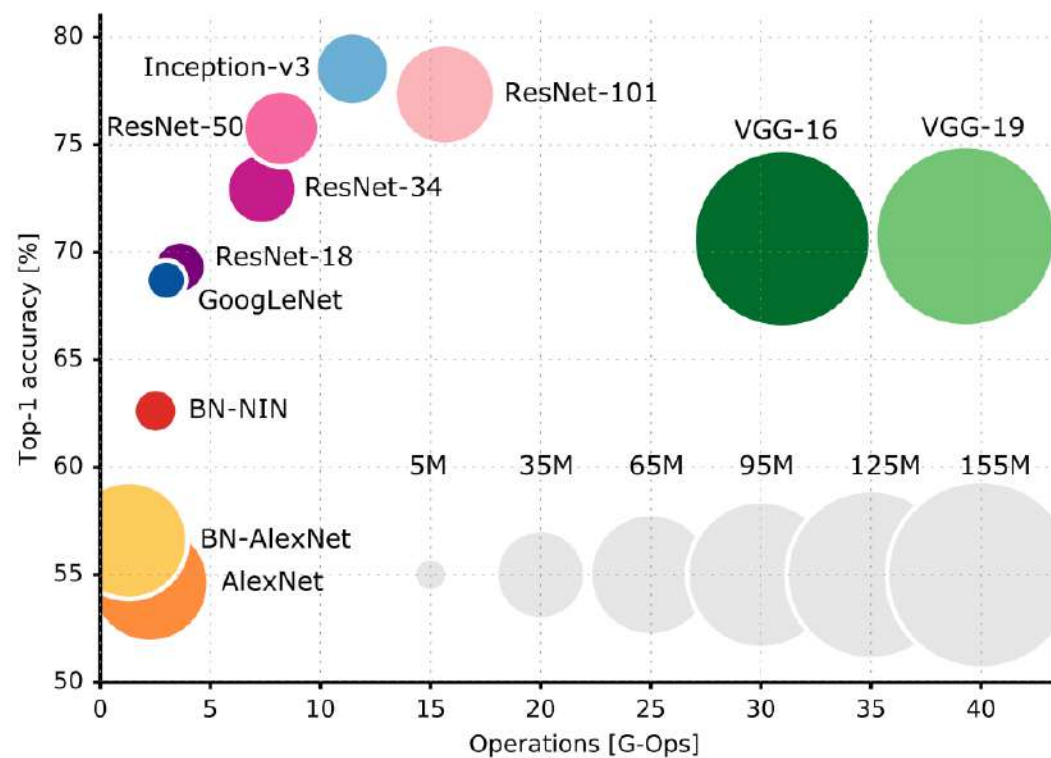
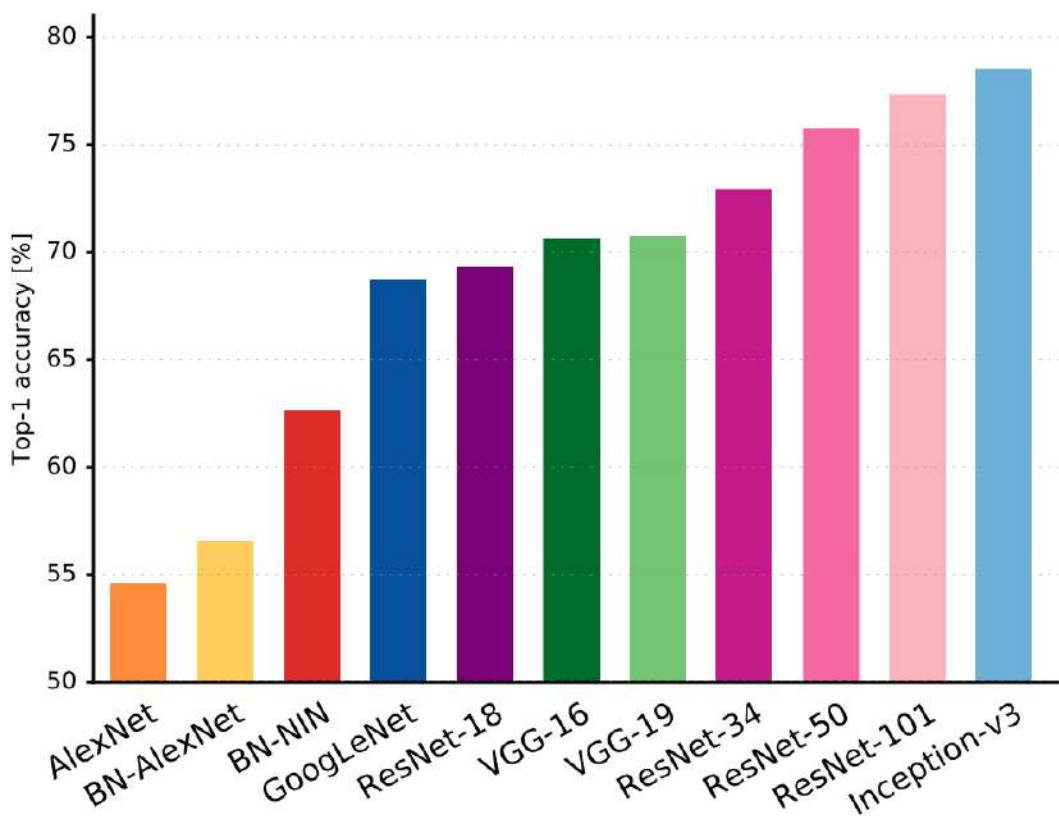
Deep Representation Learning

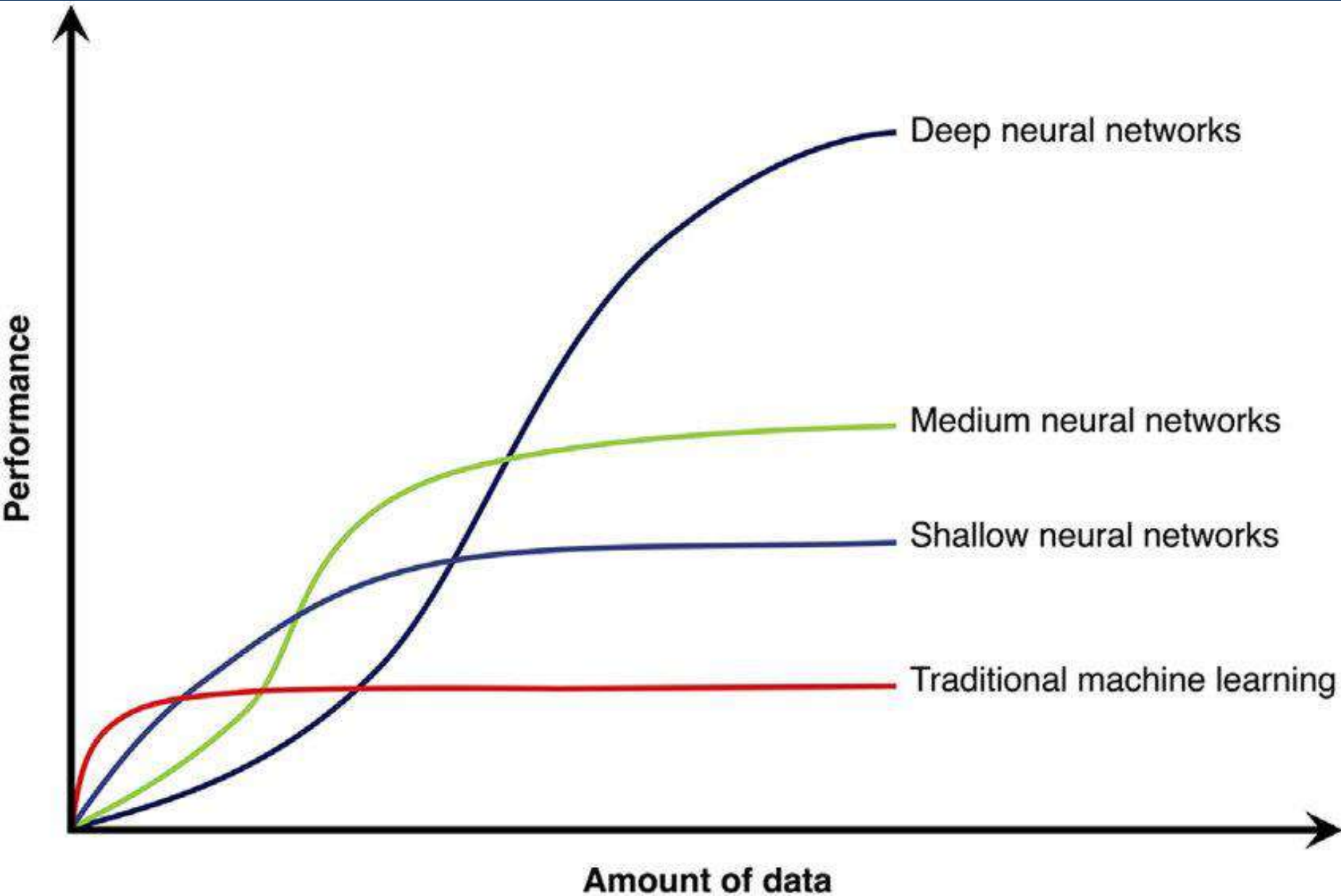




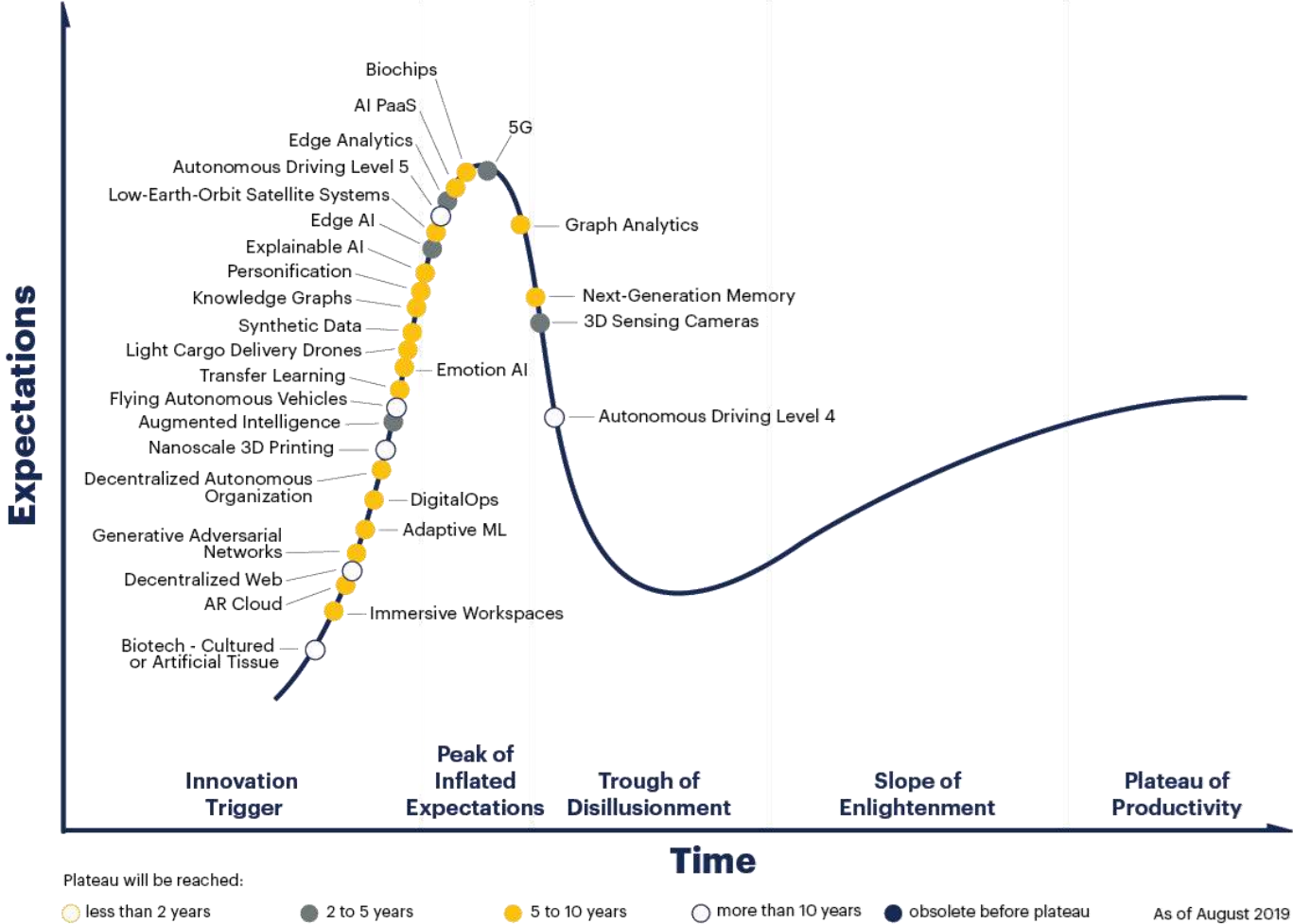
- Over-parameterisation and generalisation
- Architecture / training and regularisation / selected research topics







Gartner Hype Cycle for Emerging Technologies, 2019



Classic machine learning methods (first part of the module)

- Linear regression
- Support vector machine
- Random forest

Tools

- Deep neural networks
- Computer vision and natural language processing
- Supervised learning with quality labeled data

Research

- Advanced learning and generative modelling
- Domain adaptation and meta-learning
- Unsupervised learning, reinforcement learning, Monte Carlo, approximate inference, model interpretation, auto-ML...

Development environment

- OS
 - Ubuntu, Debian
 - MacOS
 - Windows Subsystem for Linux
 - Linux on Chrome OS
- Python and NumPy
- TensorFlow and PyTorch
- Jupyter Notebook
- Cloud computing, e.g. Google Colab
- Git
- GPU computing