



สำนักงานส่งเสริมเศรษฐกิจดิจิทัล
Digital Economy Promotion Agency



UWค.

หน่วยบริการและจัดการงานด้านการพัฒนาทางสังคม
และกึ่งด้านการพัฒนา สถาบันอุดมศึกษา
การวิจัยและการสร้างนวัตกรรม

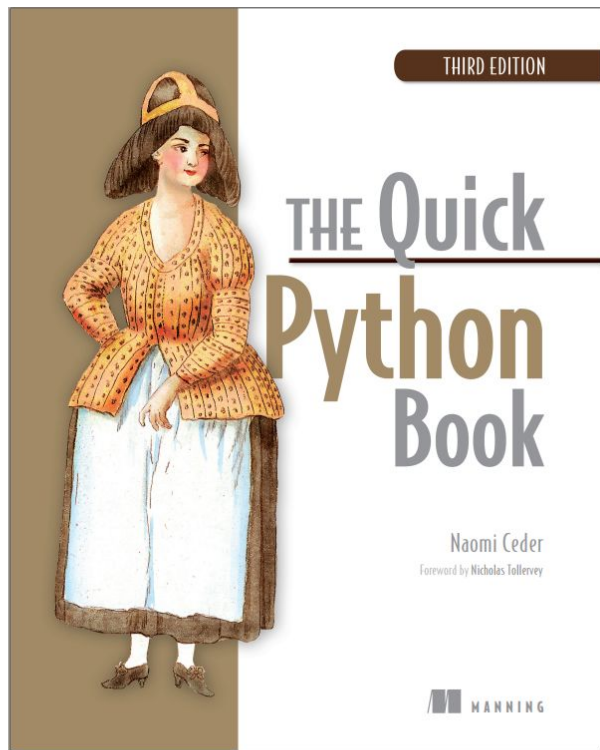
Introduction to PyTorch

An open source machine learning framework

25 September 2022

Department of Mathematics, Chiang Mai University

Why Python?



- Python is a modern programming language developed by Guido van Rossum in the 1990s
- *What Python does well*
 - *Python is easy to use*
 - *Python is expressive*
 - *Python is readable*
 - *Python is complete—“batteries included”*
 - *Python is cross-platform*
 - *Python is free*
- *What Python doesn't do as well*
 - *Python isn't the fastest language*
 - *Python doesn't have the most libraries*
 - *Python doesn't check variable types at compile time*
 - *Python doesn't have much mobile support*
 - *Python doesn't use multiple processors well*



Notepad++ : <https://notepad-plus-plus.org/downloads/>

Visual Studio Code : <https://code.visualstudio.com/>

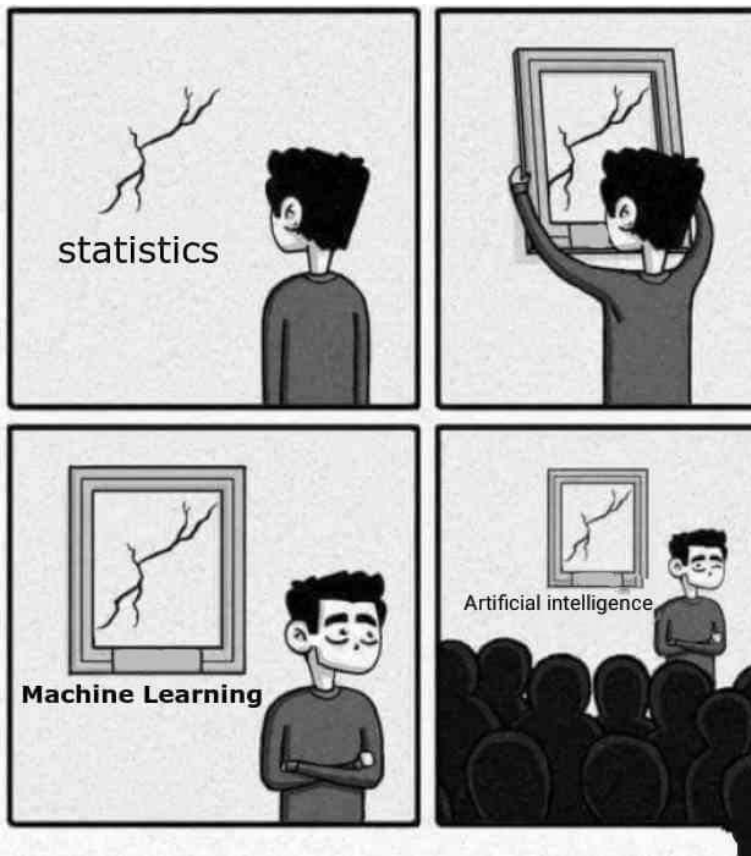




A meme featuring Woody and Buzz Lightyear from the movie Toy Story. Buzz is in the foreground, wearing his green and purple space suit, with his right arm raised and fingers spread. Woody is behind him, looking on with a neutral expression. The background is a blurred indoor setting.

MACHINE LEARNING

**MACHINE LEARNING
EVERYWHERE!**



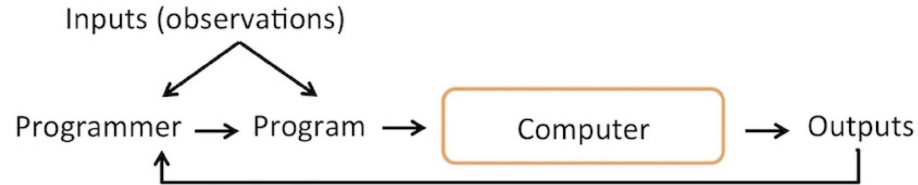
What is Machine Learning?

- Machine learning (ML) is the scientific study of algorithms and statistical models that computer systems use to perform a specific task without using explicit instructions, relying on patterns and inference instead.
- https://en.wikipedia.org/wiki/Machine_learning

Comparison between traditional programming (A) and machine learning (B)

A

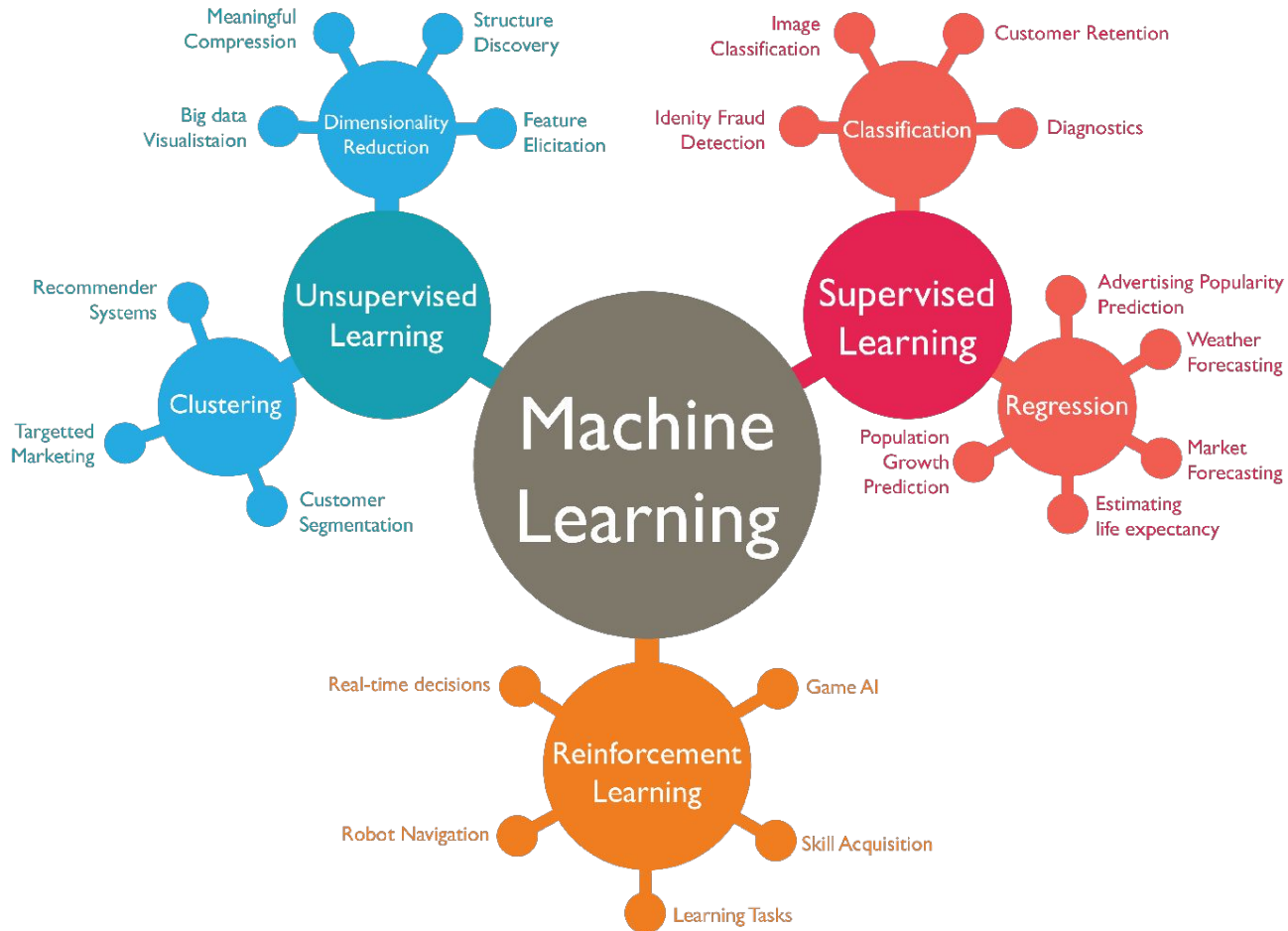
The Traditional Programming Paradigm



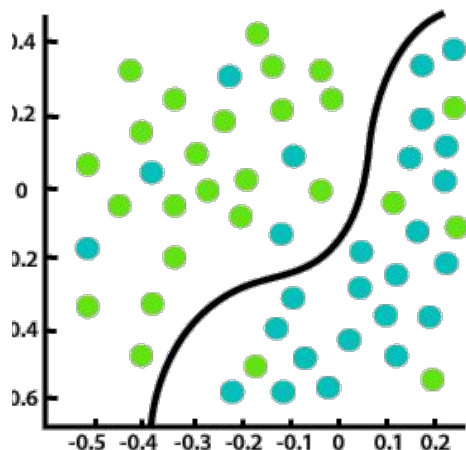
B

Machine Learning

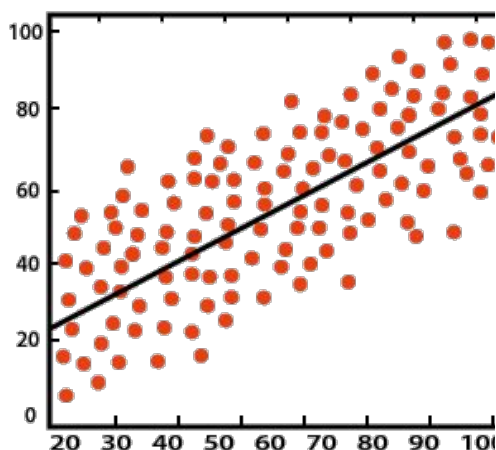




Regression vs Classification



Classification



Regression

	Underfitting	Just right	Overfitting
Symptoms	<ul style="list-style-type: none"> - High training error - Training error close to test error - High bias 	<ul style="list-style-type: none"> - Training error slightly lower than test error 	<ul style="list-style-type: none"> - Low training error - Training error much lower than test error - High variance
Regression			
Classification			
Deep learning			
Remedies	<ul style="list-style-type: none"> - Complexify model - Add more features - Train longer 		<ul style="list-style-type: none"> - Regularize - Get more data

Introduction to Pytorch

Neural Networks in brief

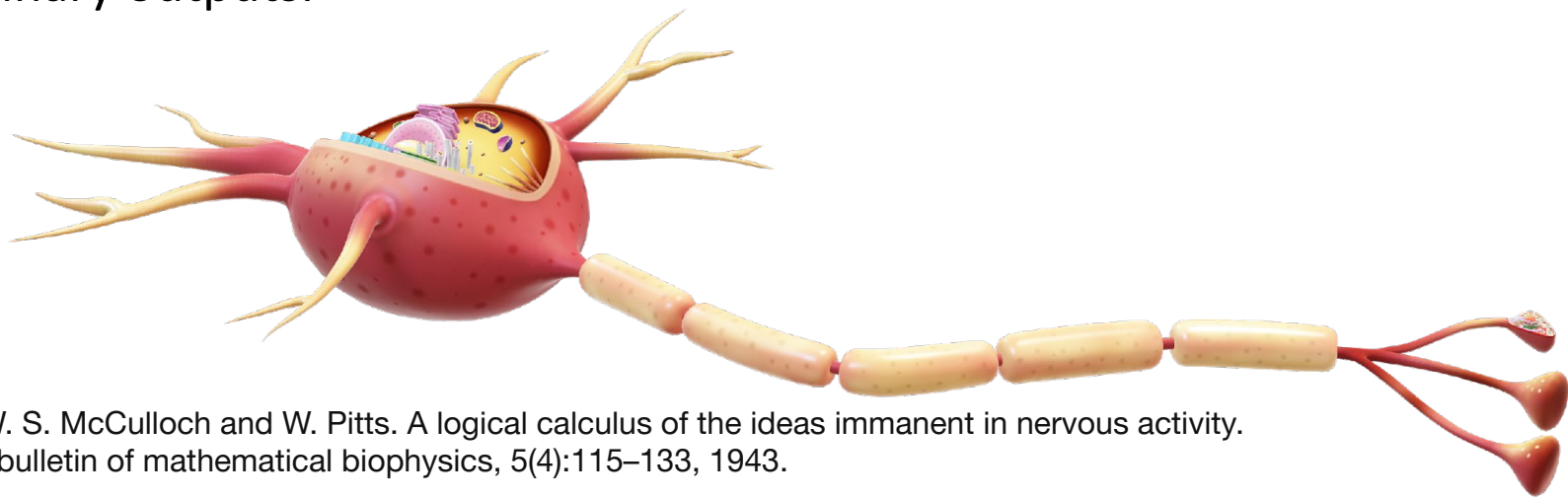
PyTorch package : Deep Learning Tools

Pytorch Linear Regression



1943 – The first mathematical model of a neural network

ANNs began with Warren McCulloch and Walter Pitts [1] who drew an analogy between biological neurons and simple logic gates with binary outputs.

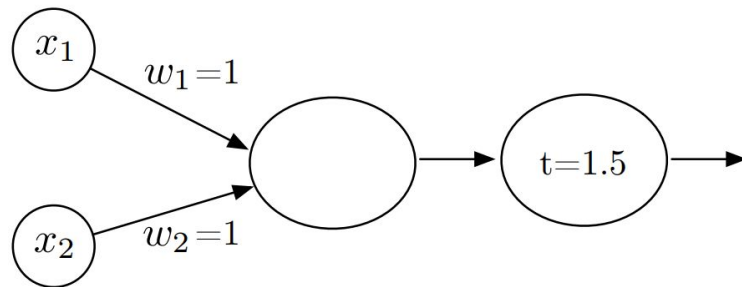


[1] W. S. McCulloch and W. Pitts. A logical calculus of the ideas immanent in nervous activity. The bulletin of mathematical biophysics, 5(4):115–133, 1943.

1943 – The first mathematical model of a neural network

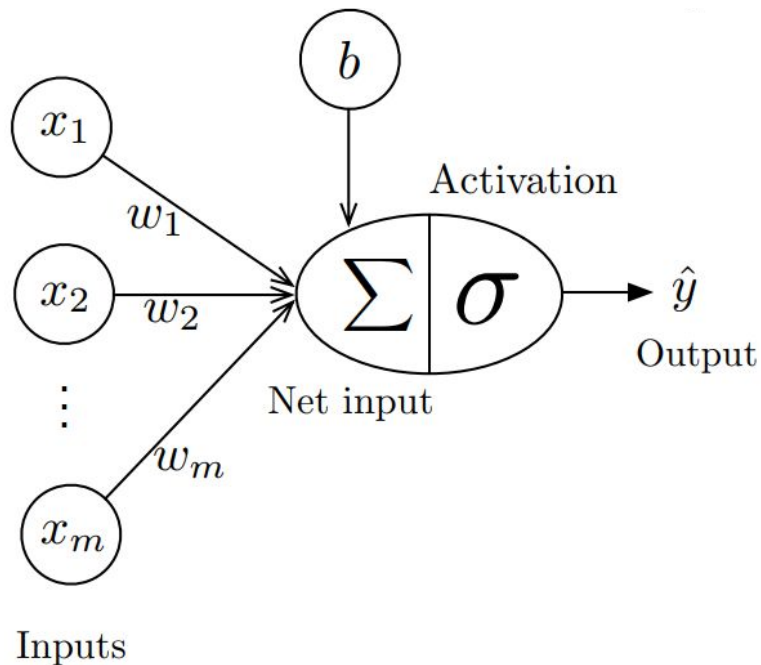
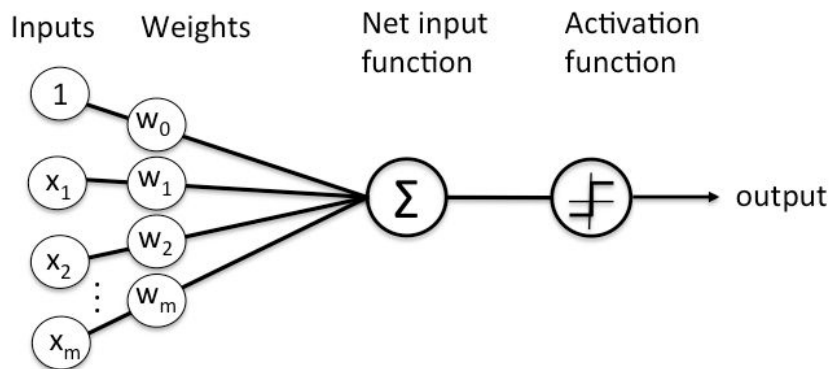
ANNs began with [Warren McCulloch](#) and [Walter Pitts](#) [1] who drew an analogy between biological neurons and simple logic gates with binary outputs.

x_1	x_2	Out
0	0	0
0	1	0
1	0	0
1	1	1



[1] W. S. McCulloch and W. Pitts. A logical calculus of the ideas immanent in nervous activity. The bulletin of mathematical biophysics, 5(4):115–133, 1943.

Neural Networks Mathematical Model



Linear Algebra inside NNs model

- Output $\hat{y} := \sigma(z)$ where $z = x^T \mathbf{w} + b, \mathbf{w} \in \mathbb{R}^{m \times 1}$

- For 1 example : $x \in \mathbb{R}^{m \times 1}$,

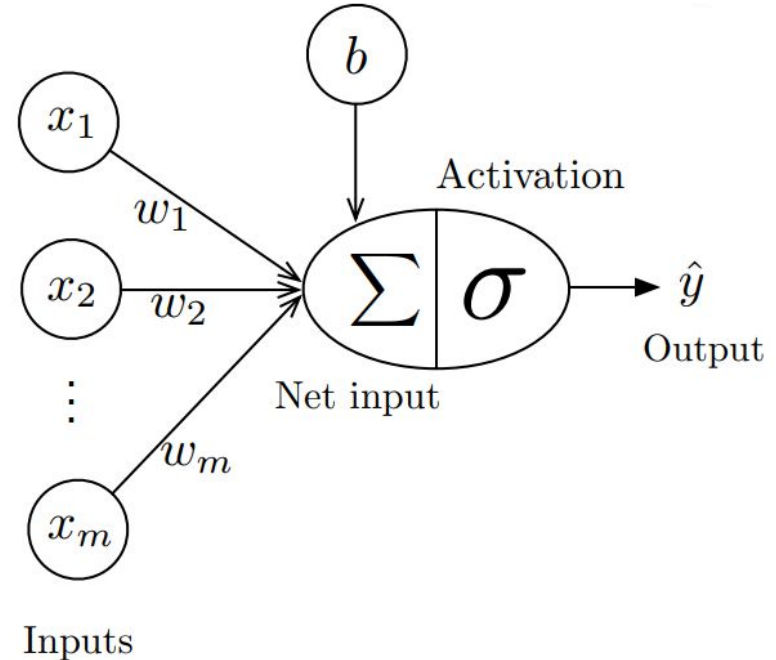
- $z = x^T \mathbf{w} + b$

- $\hat{y} = \sigma(z)$

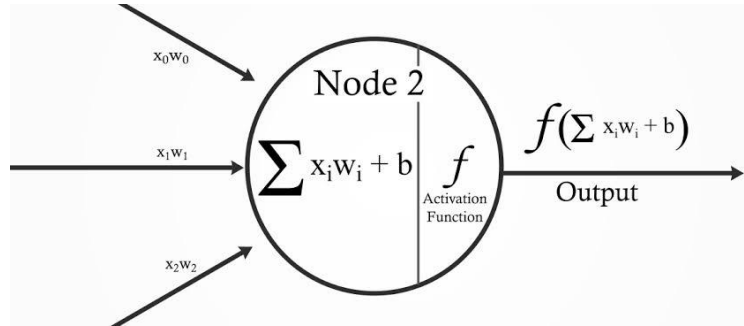
- For n examples : $X \in \mathbb{R}^{n \times m}$

- $X\mathbf{w} + b = \begin{bmatrix} (x^{[1]})^T \mathbf{w} + b \\ \vdots \\ (x^{[n]})^T \mathbf{w} + b \end{bmatrix} = \begin{bmatrix} z_1 \\ \vdots \\ z_n \end{bmatrix} = \mathbf{z} \in \mathbb{R}^{n \times 1}$

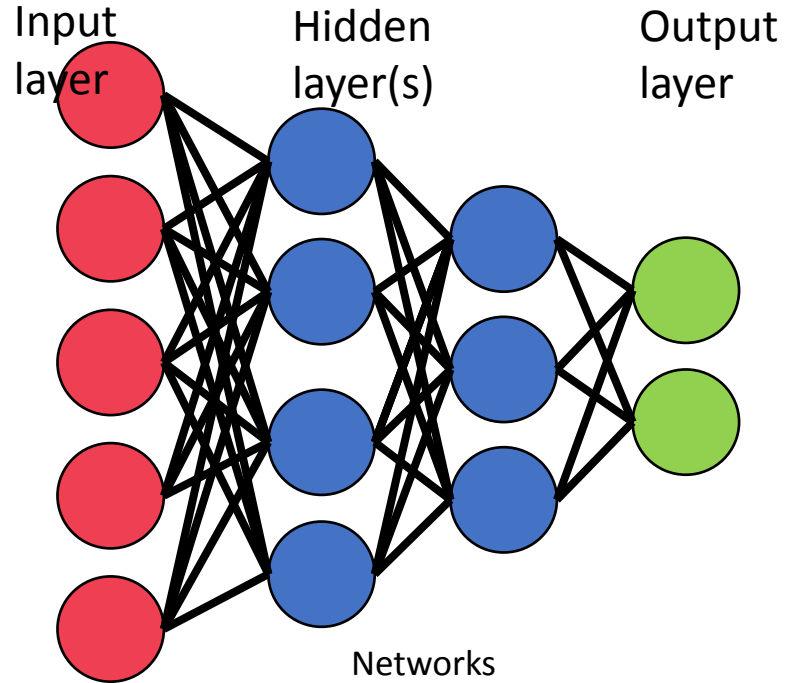
- $\hat{\mathbf{y}} = \begin{bmatrix} \sigma(z_1) \\ \vdots \\ \sigma(z_n) \end{bmatrix} = \sigma(\mathbf{z})$



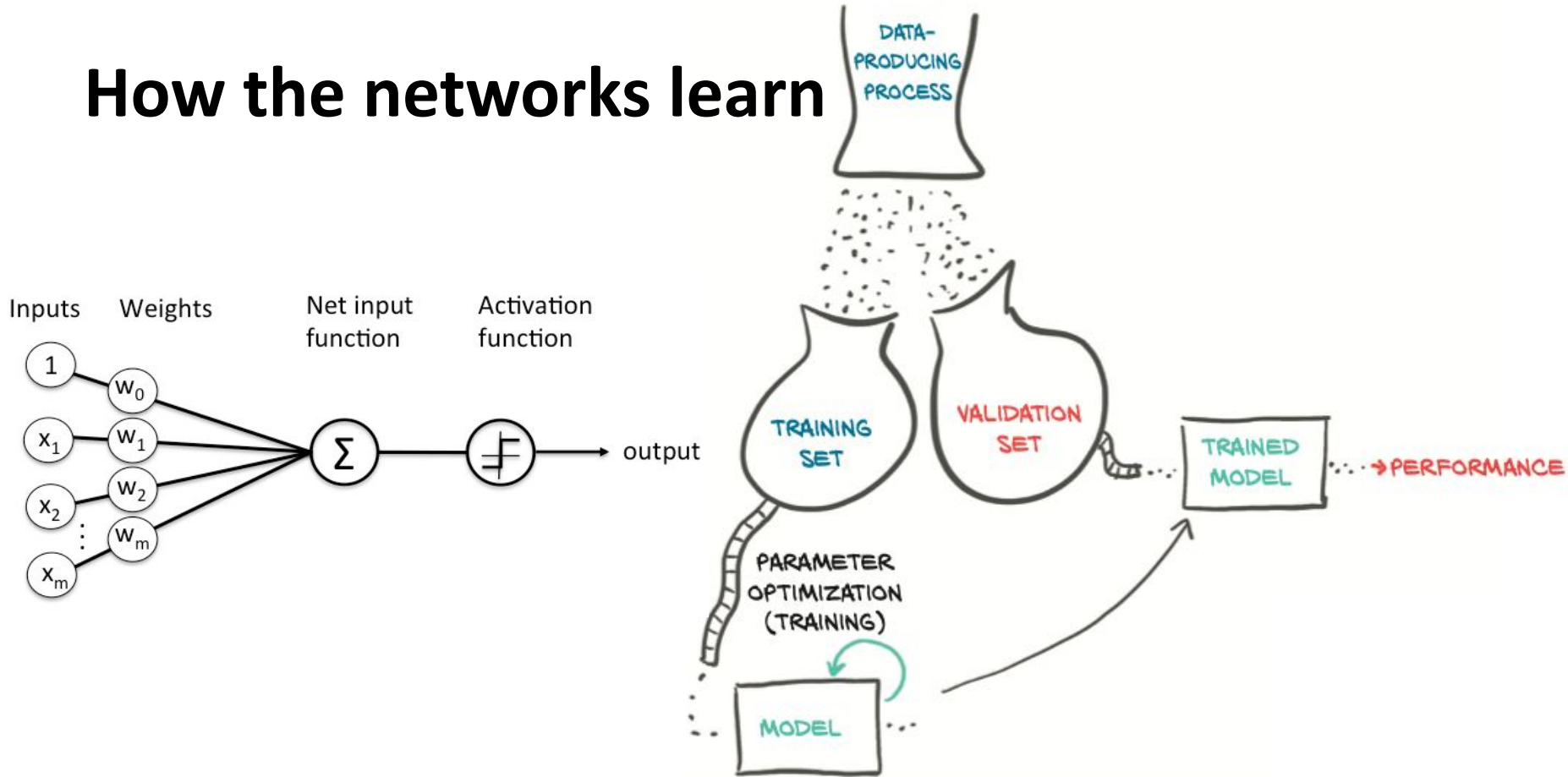
Artificial Neural Networks



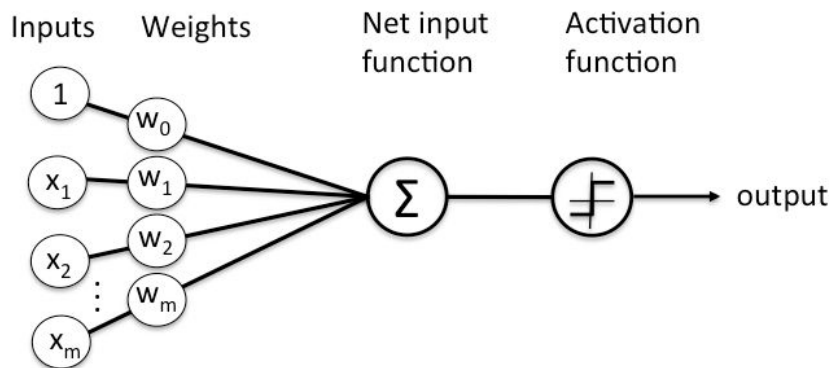
Neural



How the networks learn



How the networks learn



THE "NEURON"

$$o = \tanh(w x + b)$$

Diagram illustrating the neuron's internal calculation:

- LEARNED PARAMETERS:** w (weight) and b (bias).
- INPUT:** x .
- LINEAR TRANSFORMATION:** The input x is transformed into $w x + b$.
- NONLINEAR FUNCTION (ACTIVATION):** The result of the linear transformation is passed through the \tanh activation function to produce the **OUTPUT** o .

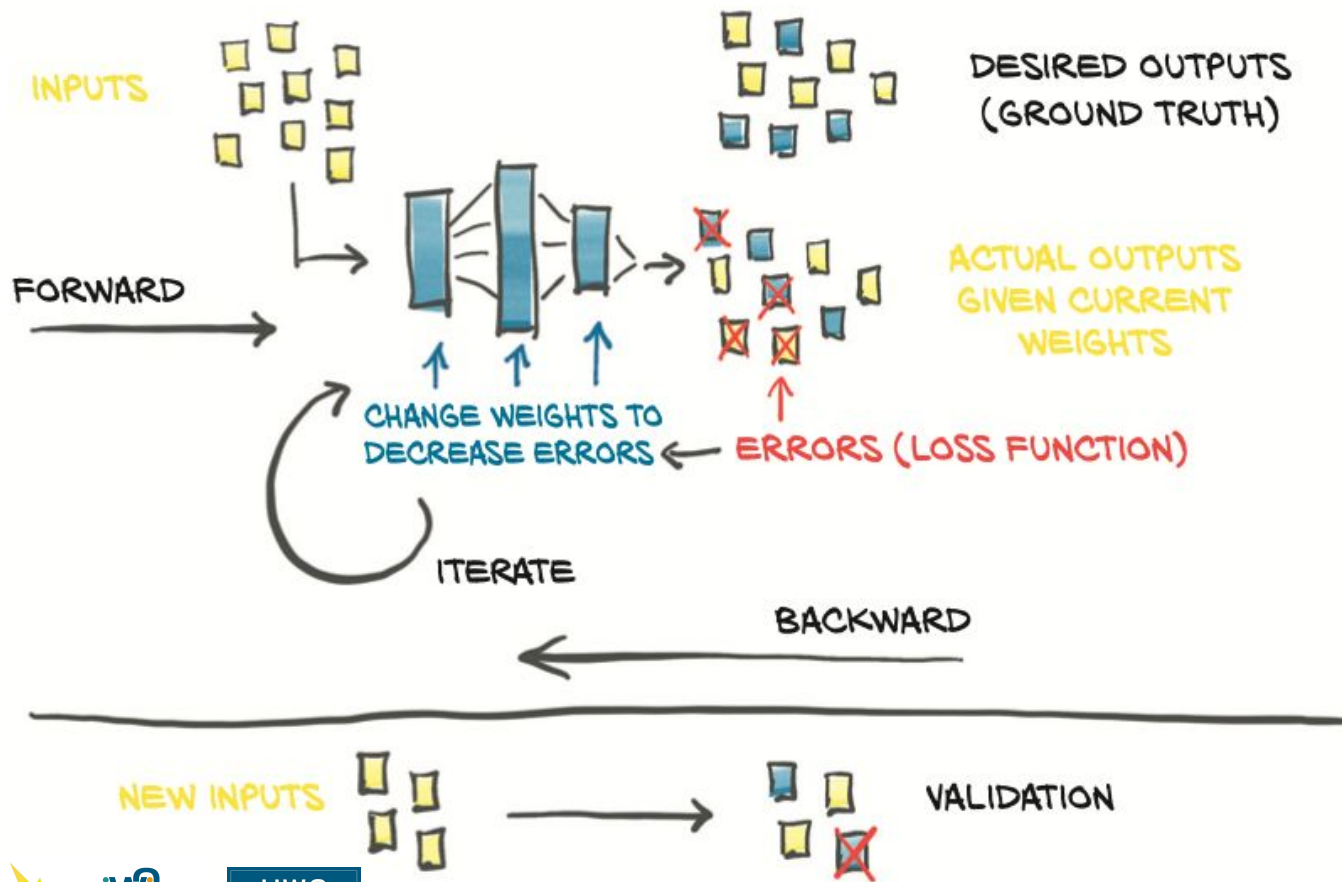
LEARNED

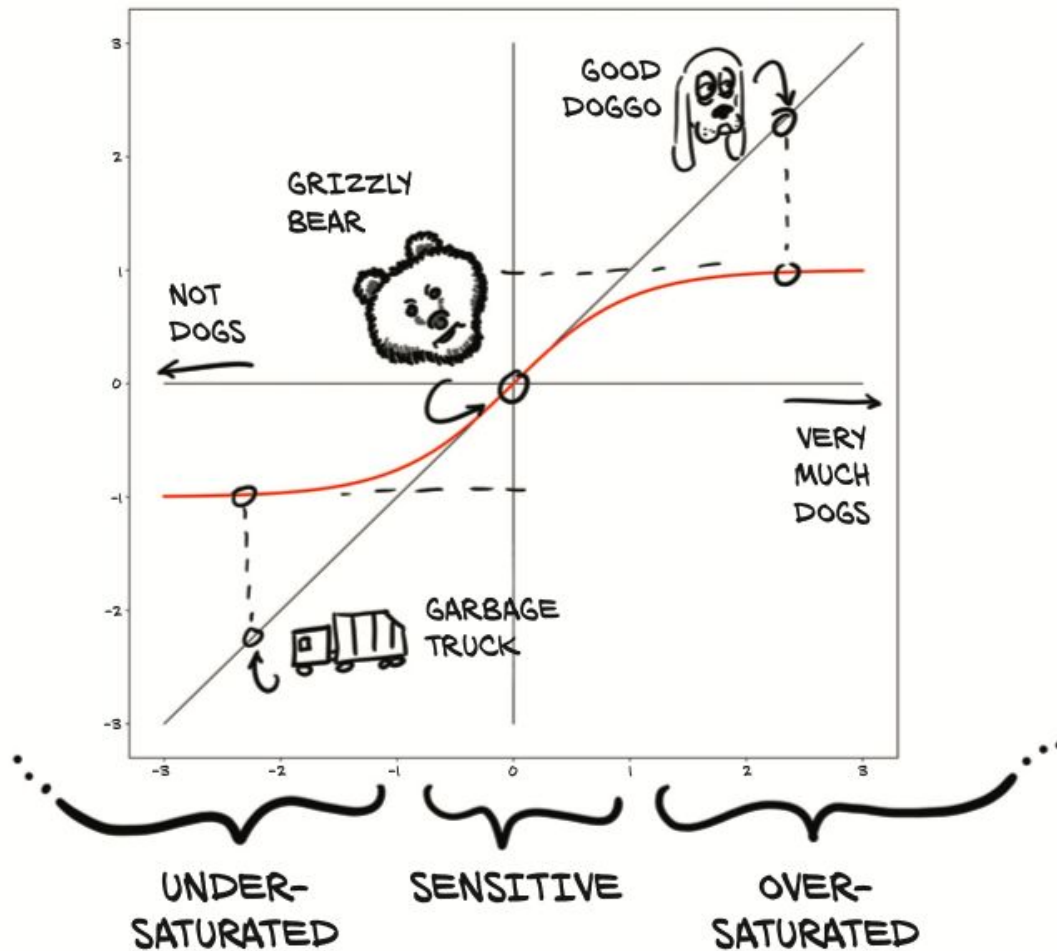
$w = 2$
 $b = 6$

Graphs showing the linear transformation $y = wx + b$ and the \tanh activation function $o = \tanh(y)$.

18	\rightarrow	$2 \times 18 + 6 = 42$	\rightarrow	$\tanh(42) = 1$
-2.79	\rightarrow	$2 \times (-2.79) + 6 = .042$	\rightarrow	$\tanh(.042) = 0.3969$
-10	\rightarrow	$2 \times (-10) + 6 = -14$	\rightarrow	$\tanh(-14) = -1$

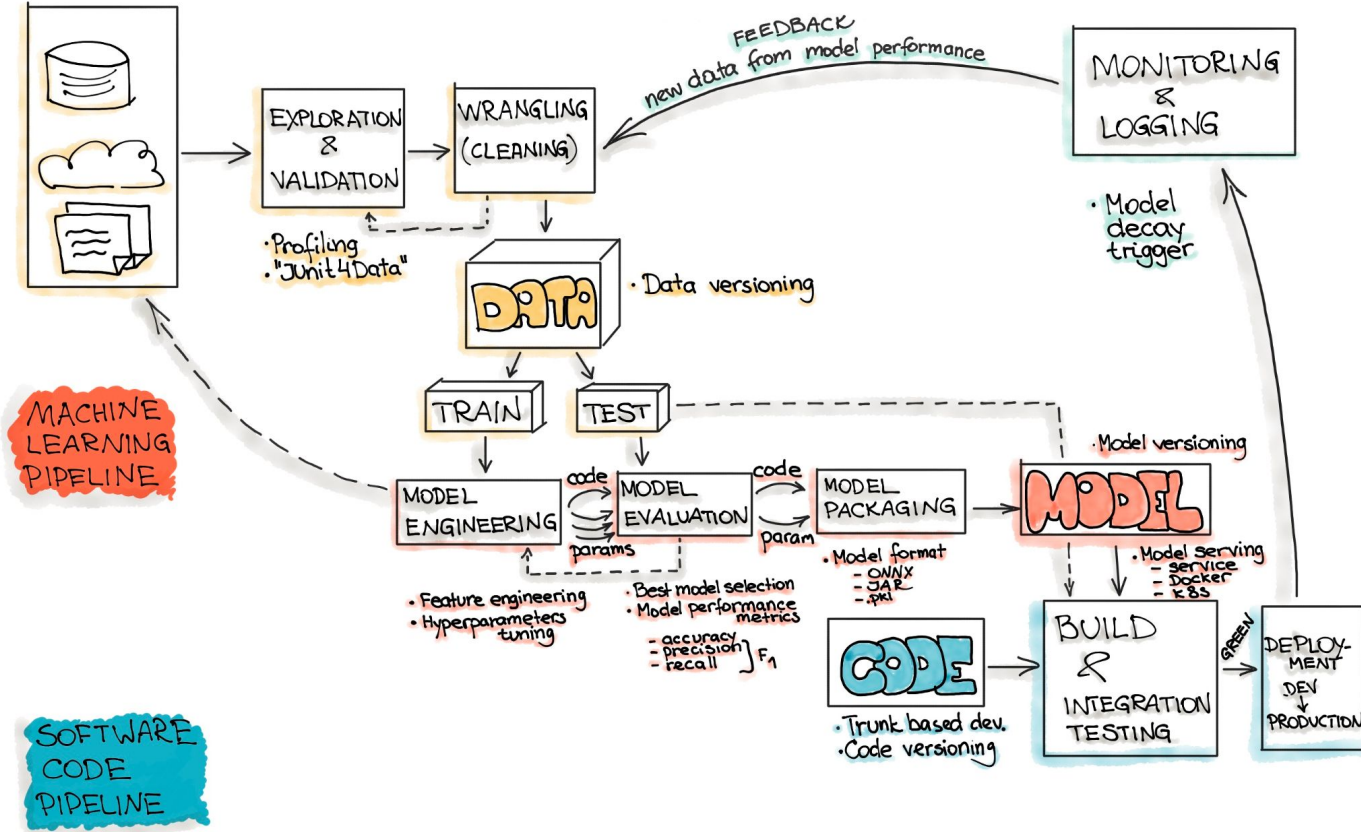
THE LEARNING PROCESS





DATA PIPELINE

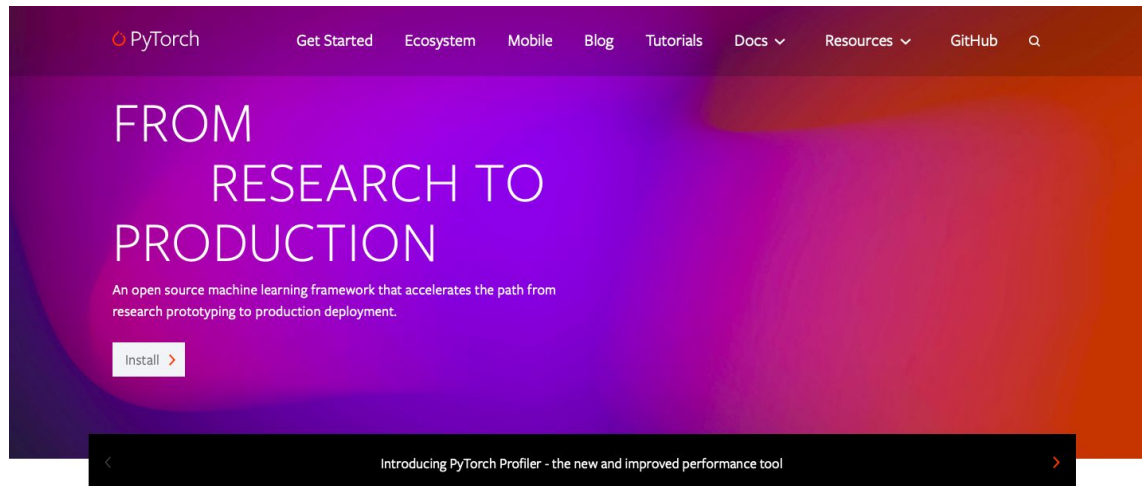
MACHINE LEARNING ENGINEERING



Introduction to PyTorch



- torch.tensor
- torch.autograd
- torch.nn
- torch.optim



KEY FEATURES & CAPABILITIES

[See all Features >](#)

Production Ready

Distributed Training

Robust Ecosystem

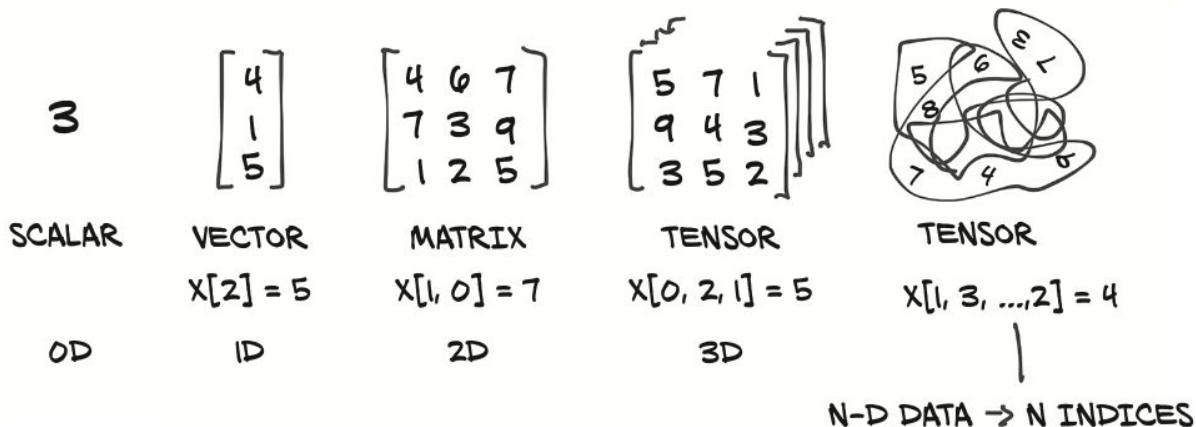
Cloud Support

<https://pytorch.org>

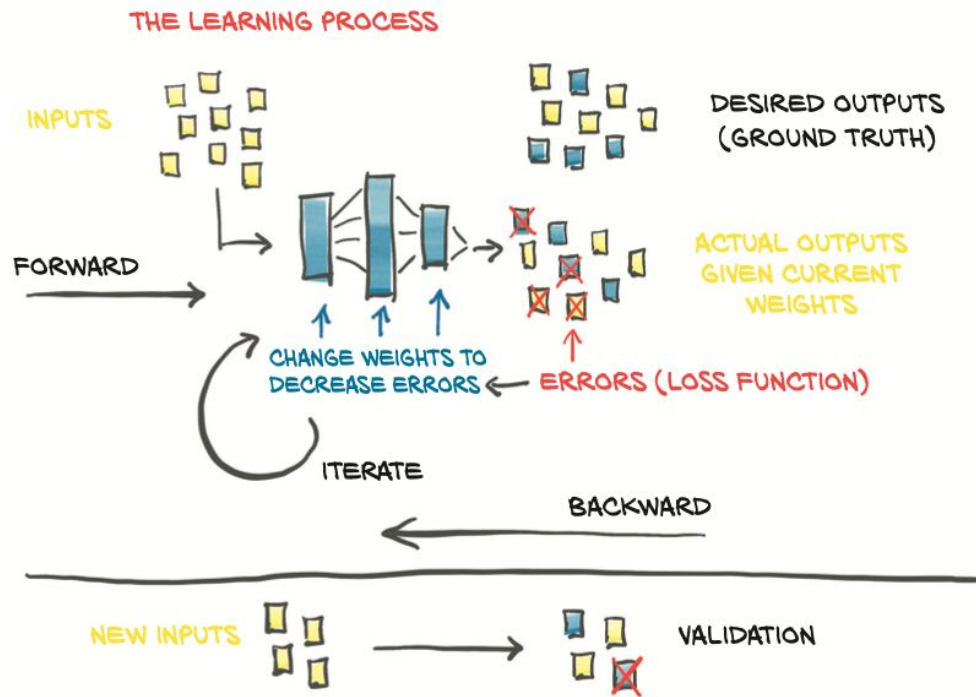


torch.tensor

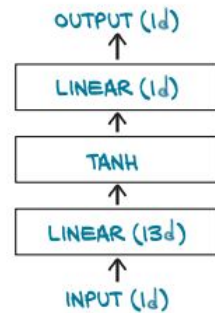
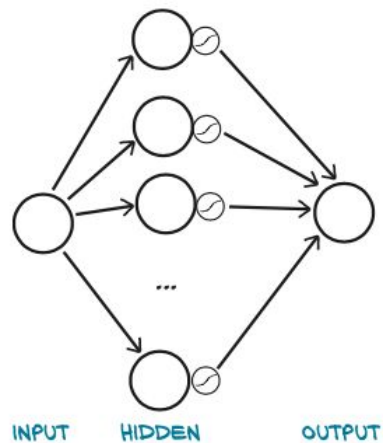
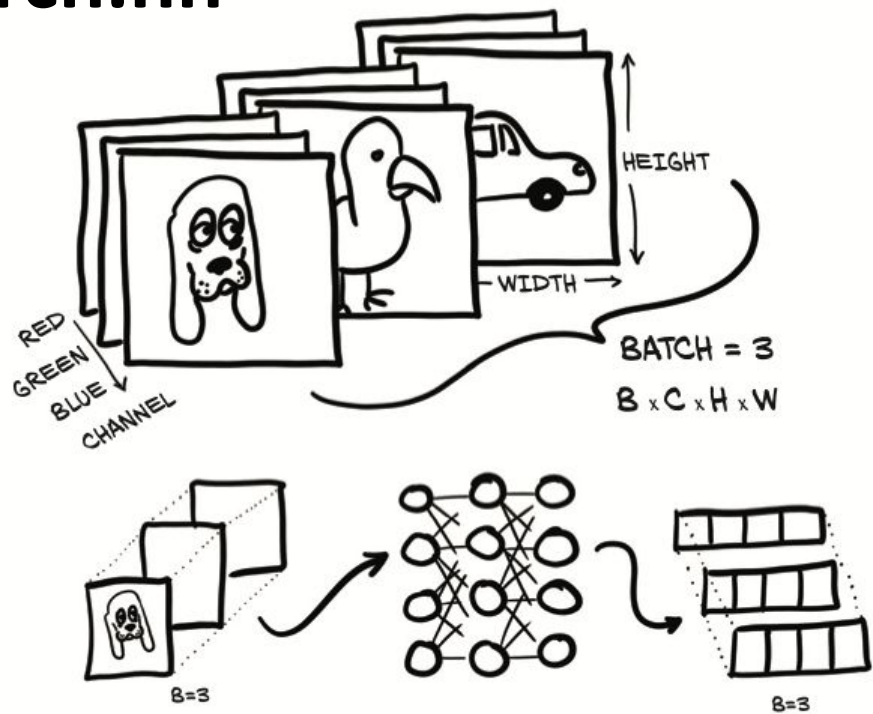
- A torch.tensor is a multi-dimensional matrix containing elements of a single data type.
- Tensors are similar to NumPy's ndarrays, with the addition being that Tensors can also be used on a GPU to accelerate computing.



torch.autograd



torch.nn

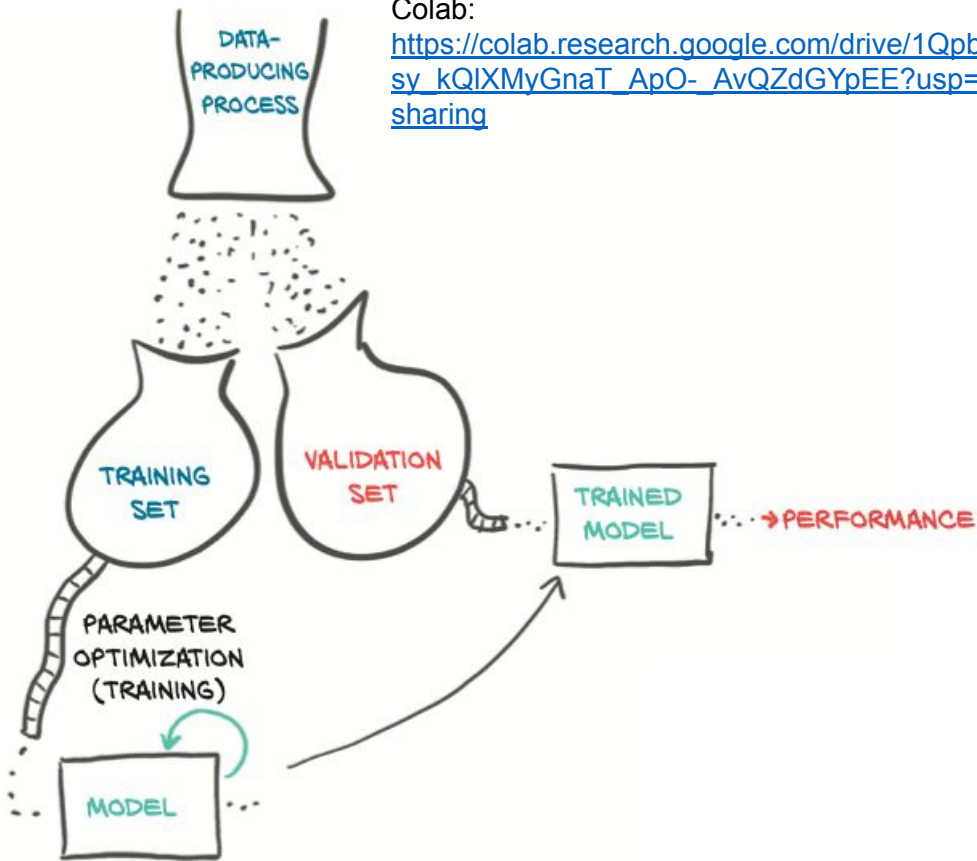
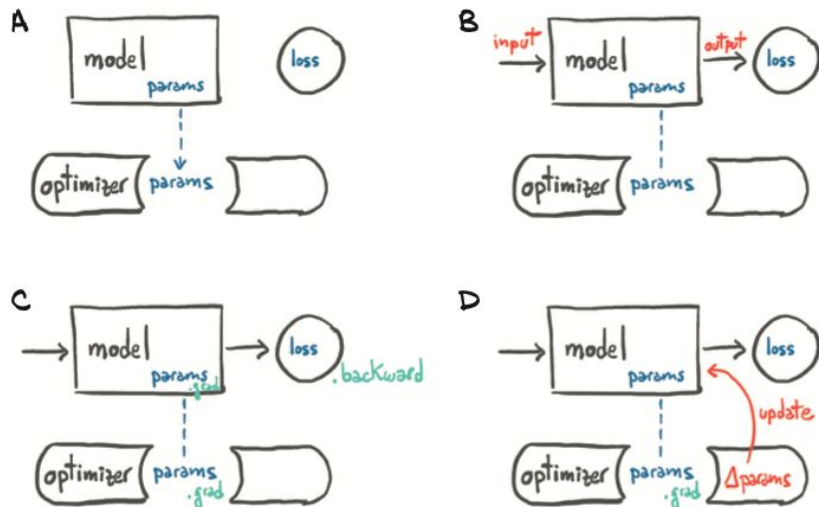


torch.optim

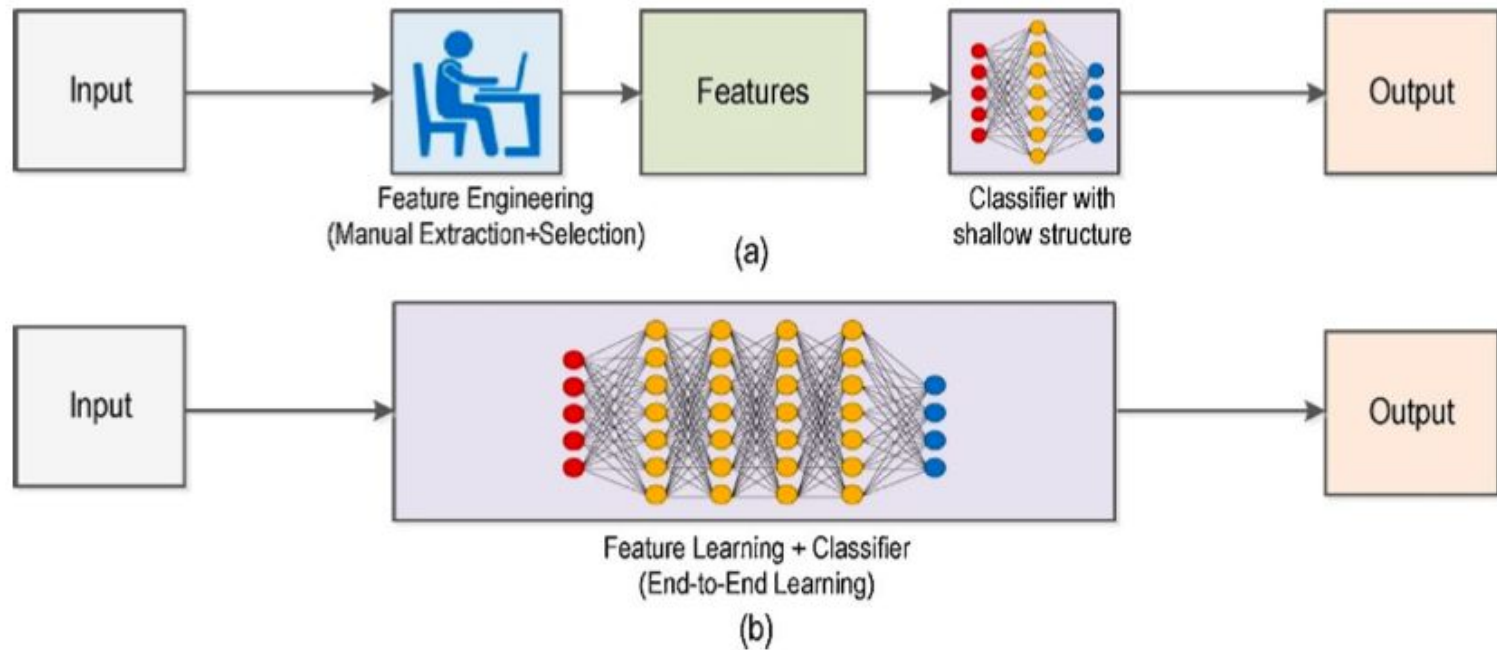
[Review] Gradient Descent

Colab:

https://colab.research.google.com/drive/1Qpb_sy_kQIXMyGnaT_ApO-AvQZdGYpEE?usp=sharing



Traditional computer vision workflow vs Deep learning workflow



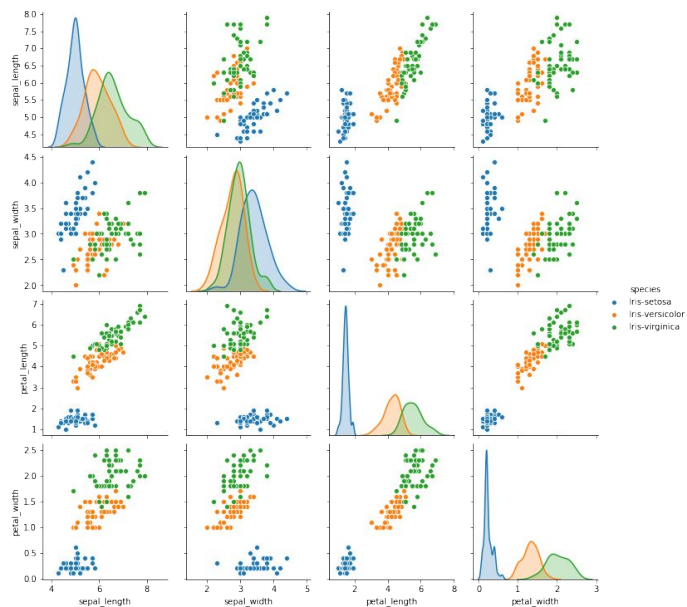
Paper source : <https://arxiv.org/pdf/1910.13796.pdf>



Iris Versicolor

Iris Setosa

Iris Virginica



Iris dataset

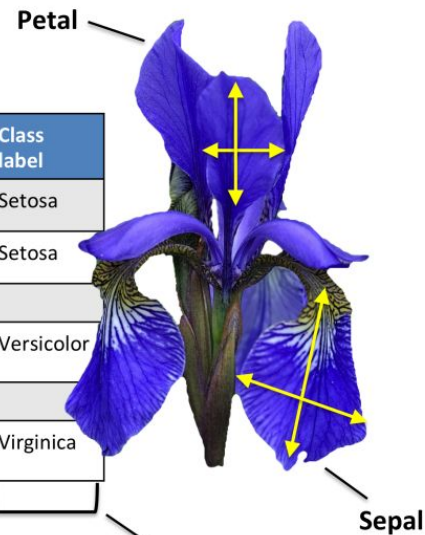
Samples

(instances, observations)

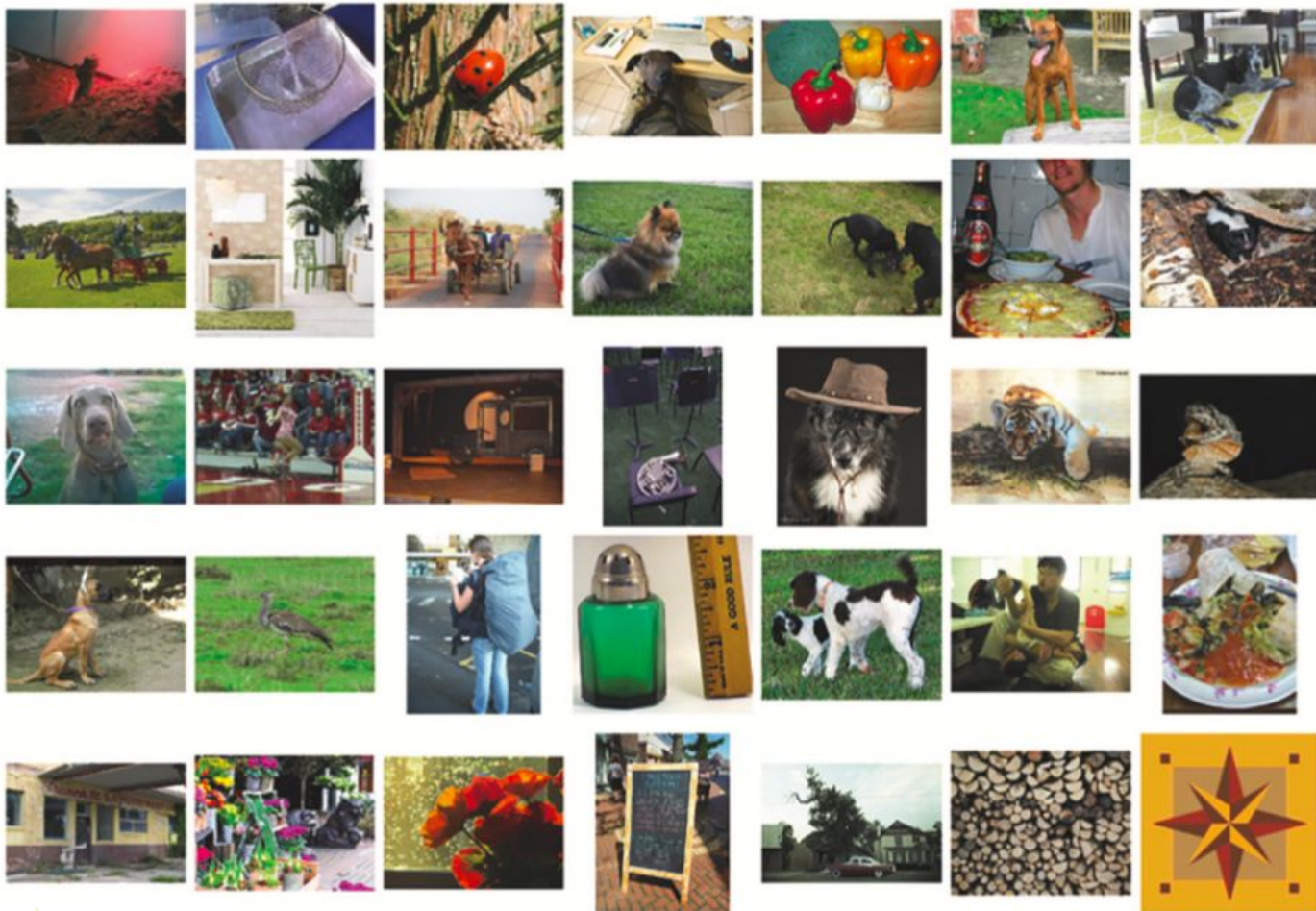
	Sepal length	Sepal width	Petal length	Petal width	Class label
1	5.1	3.5	1.4	0.2	Setosa
2	4.9	3.0	1.4	0.2	Setosa
...					
50	6.4	3.5	4.5	1.2	Versicolor
...					
150	5.9	3.0	5.0	1.8	Virginica

Features

(attributes, measurements, dimensions)



Class labels
(targets)





Machines try to see a dog as...

