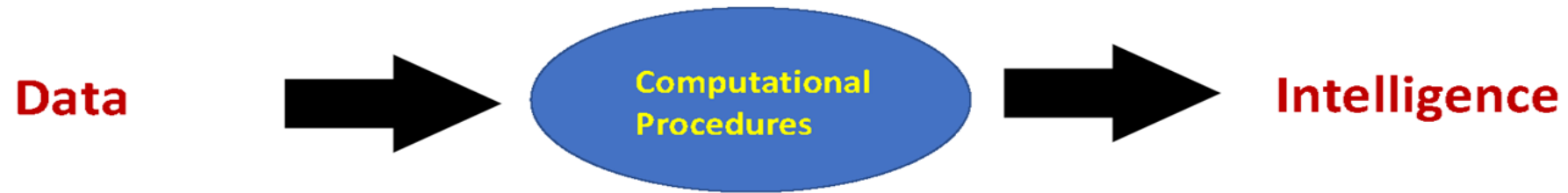


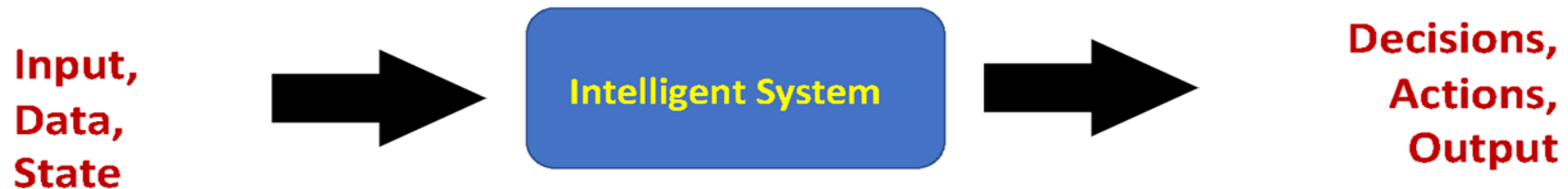
Introduction to the Modern AI/ML/DL

C. V. Jawahar
IIIT Hyderabad

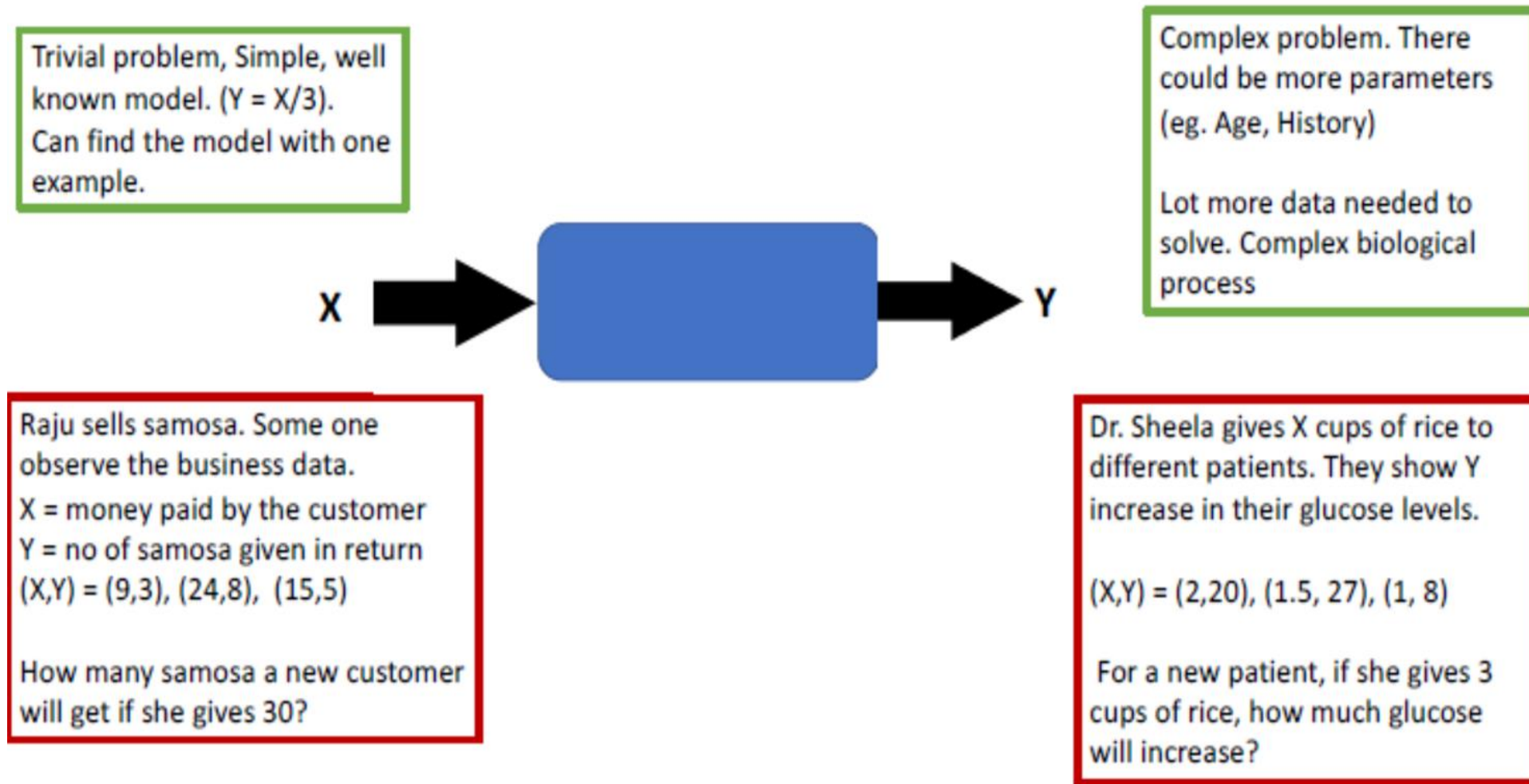
What is Modern “AI” or “ML”?



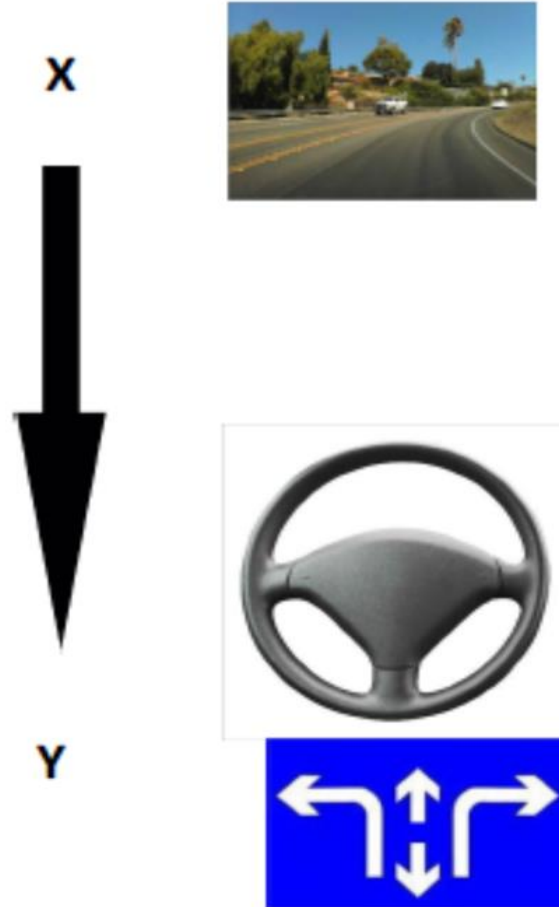
Data Driven Intelligence



Data Driven Problem Solving



More Examples



In this match , Rajasthan captain Ajinkya Rahane won the toss and decided to bowl first .



General Strategy: Given many examples of (X,Y), learn an automated solution to predict Y given a new X.

इस मैच में राजस्थान के कप्तान अजिंक्य रहाणे ने टॉस जीतकर पहले गेंदबाजी का फैसला किया ।

“Black and White Dog Jumps over Bars”

ILL Posed Problems: Why do they work?

X

Y

Can Human(experts) do this?

How do they do?



Title: Biscuits

Ingredients:

Flour, butter, sugar, egg, milk, salt.

Instructions:

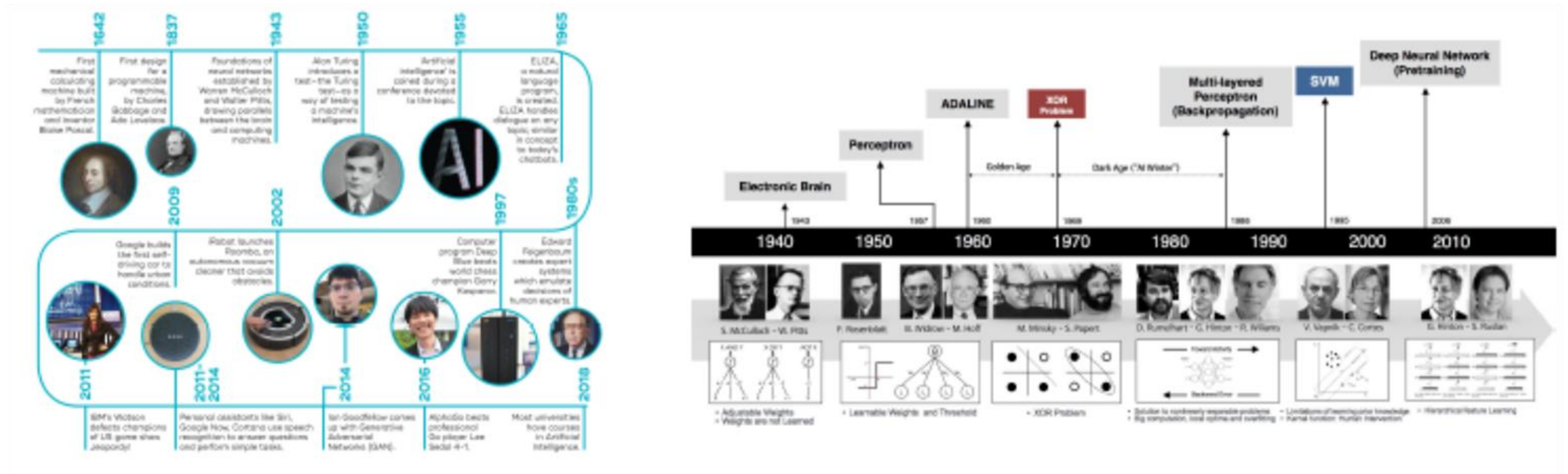
- Preheat oven to 450 degrees.
- Cream butter and sugar.
- Add egg and milk.
- Sift flour and salt together.
- Add to creamed mixture.
- Roll out on floured board to 1/4 inch thickness.
- Cut with biscuit cutter.
- Place on ungreased cookie sheet.
- Bake for 10 minutes.

Extensive use of Prior Knowledge.

Composition of parts seen in the past.

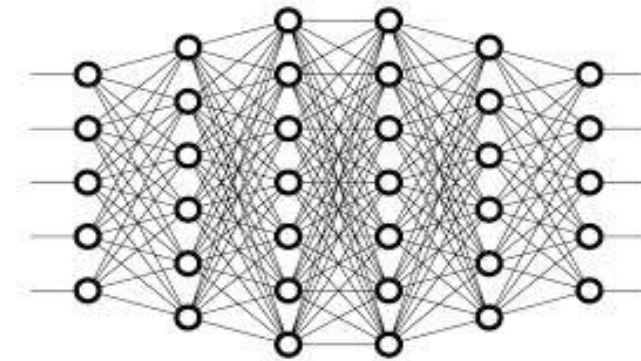
Inverse Cooking (CVPR 2019)

Is AI, ML, DL really New?



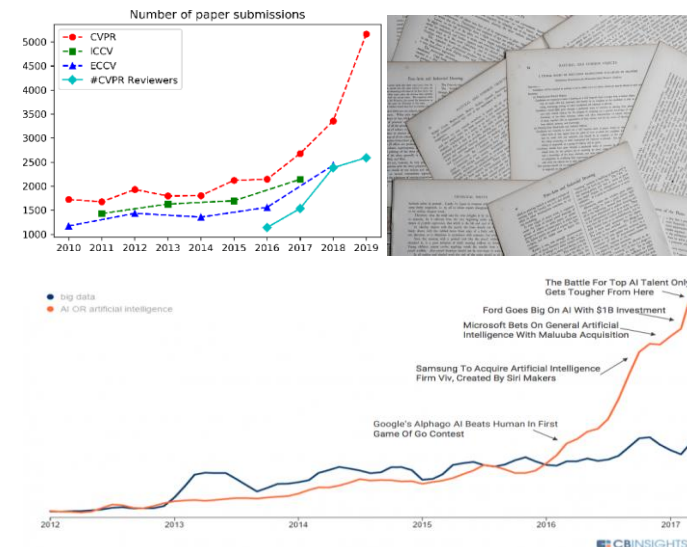
Why AI started to work now?

**Data; Internet;
Connectivity**

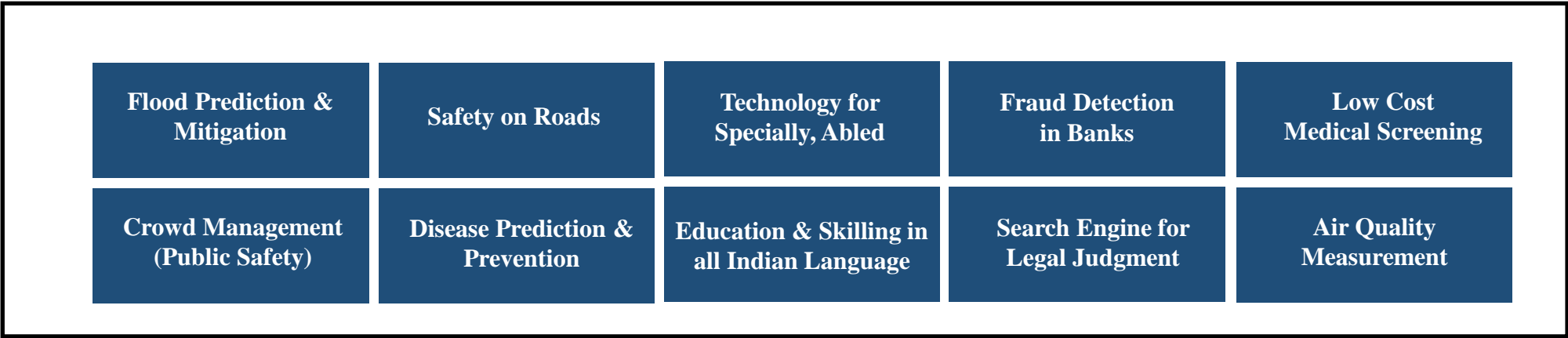


**Algorithms,
Deep Learning**

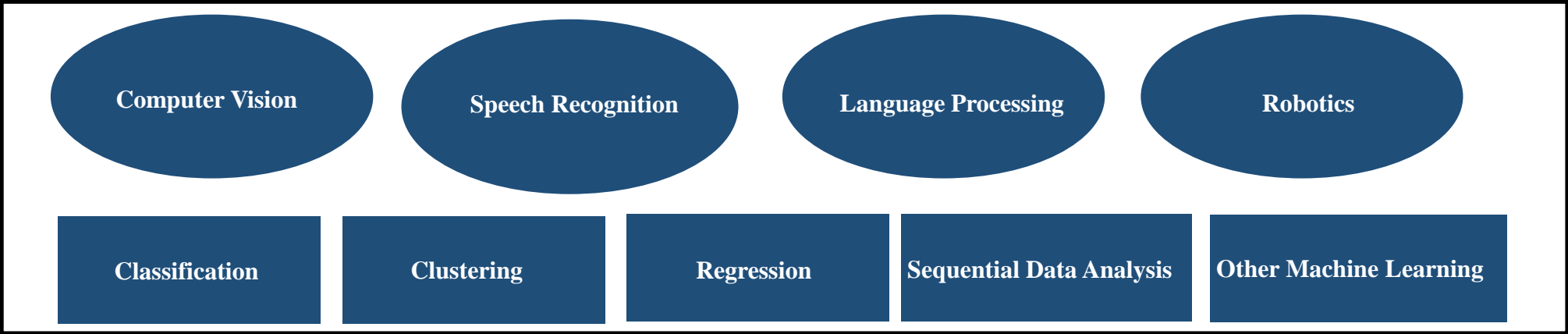
**Compute,
Cloud, APIs,
Libraries**



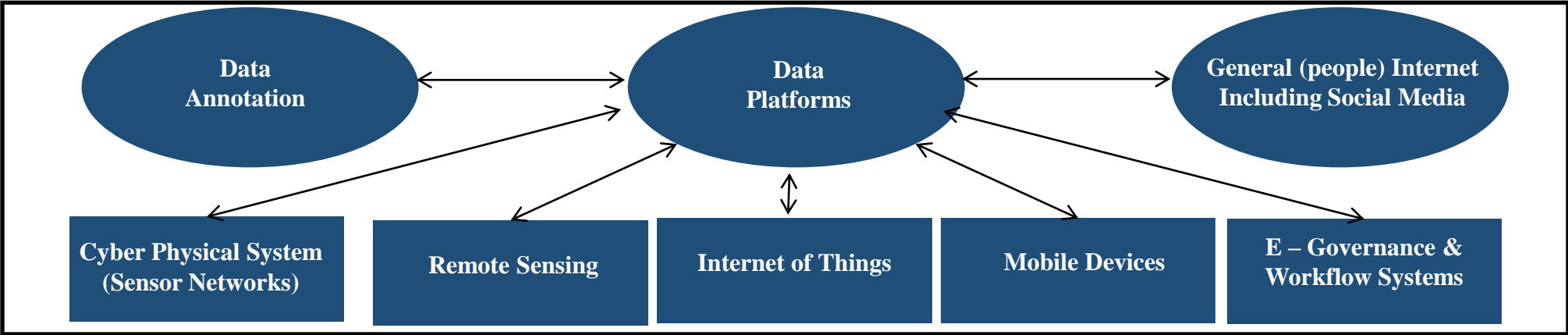
**More People,
Papers, Results,
Funding, People.
Positive Feedback.**



Domains, Applications and Problems



Algorithms and Areas



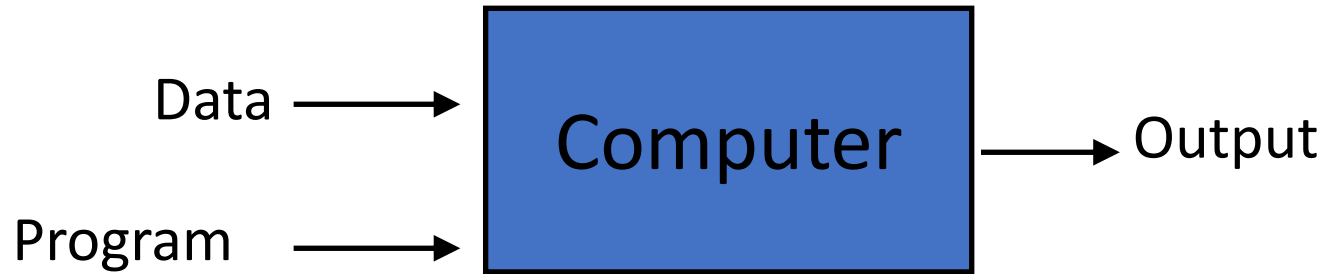
Data and Technology

What is Machine Learning?

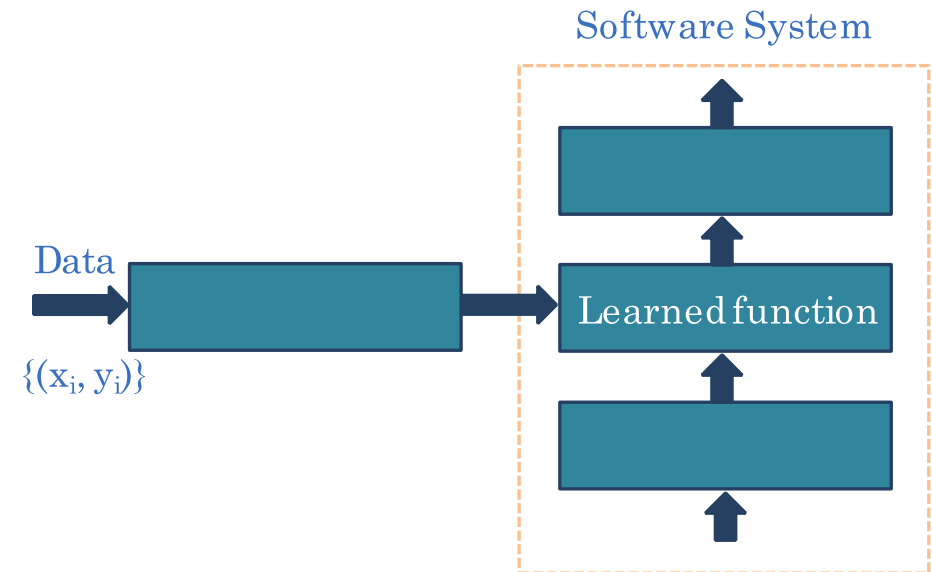
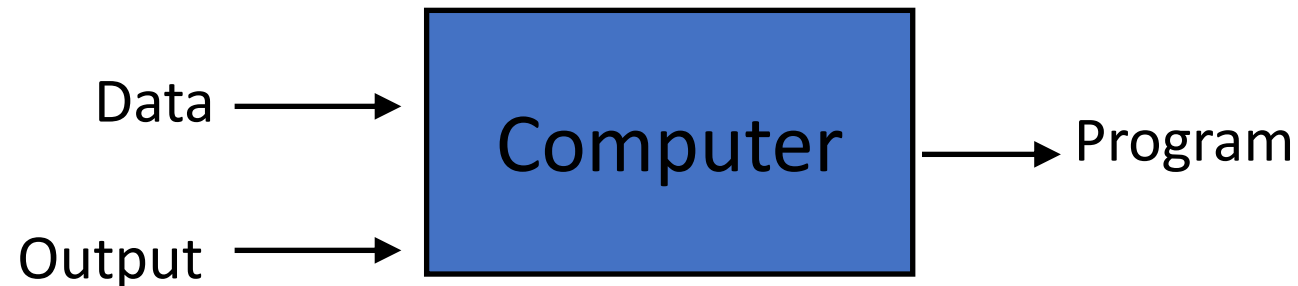


Data Driven Solutions

Traditional Programming



Machine Learning



The machine learning framework

- Apply a prediction function to a feature representation of the “sample” to get the desired output:

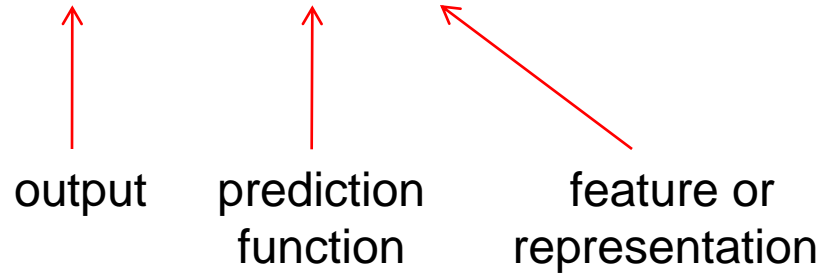
$f(\text{apple}) = \text{“apple”}$

$f(\text{tomato}) = \text{“tomato”}$

$f(\text{cow}) = \text{“cow”}$

The machine learning framework

$$y = f(\mathbf{x})$$



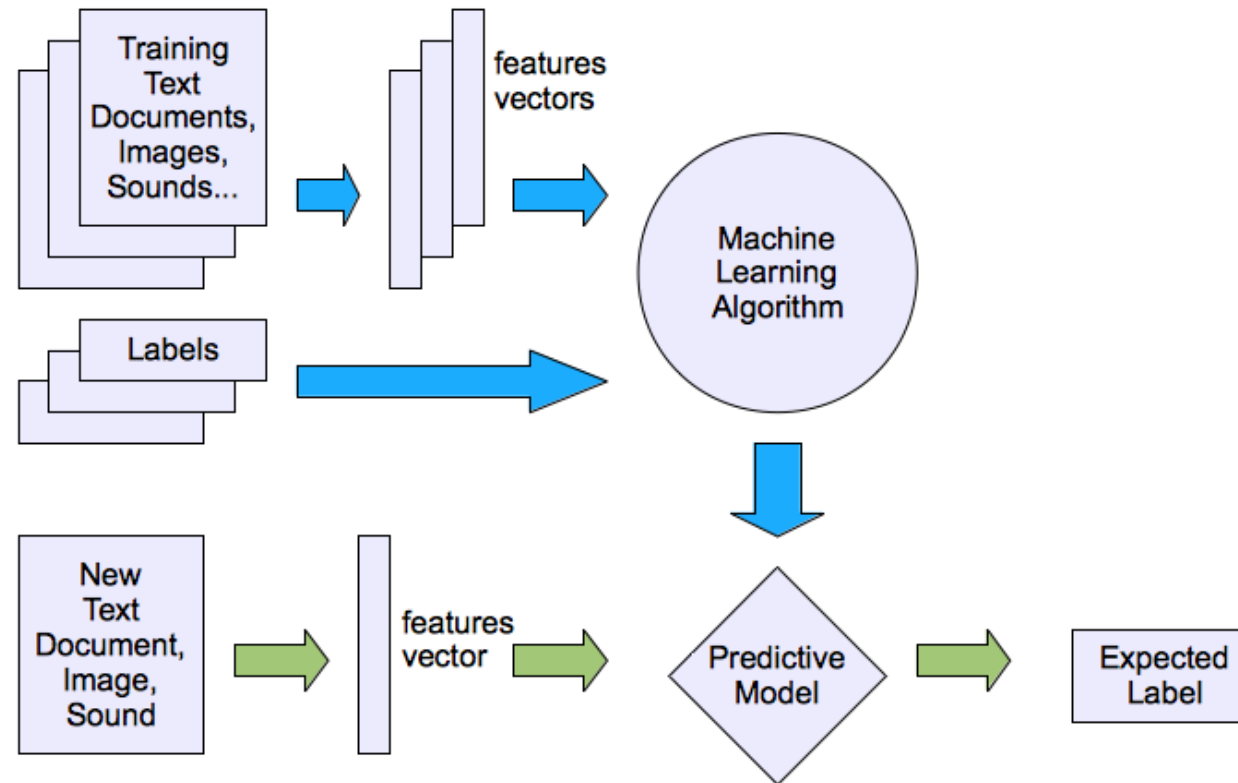
- **Training:** given a *training set* of labeled examples $\{(\mathbf{x}_1, y_1), \dots, (\mathbf{x}_N, y_N)\}$, estimate the prediction function f by minimizing the prediction error.
- **Testing:** apply f to a never-before-seen *test example* \mathbf{x} and output the predicted value $y = f(\mathbf{x})$

Supervised Learning

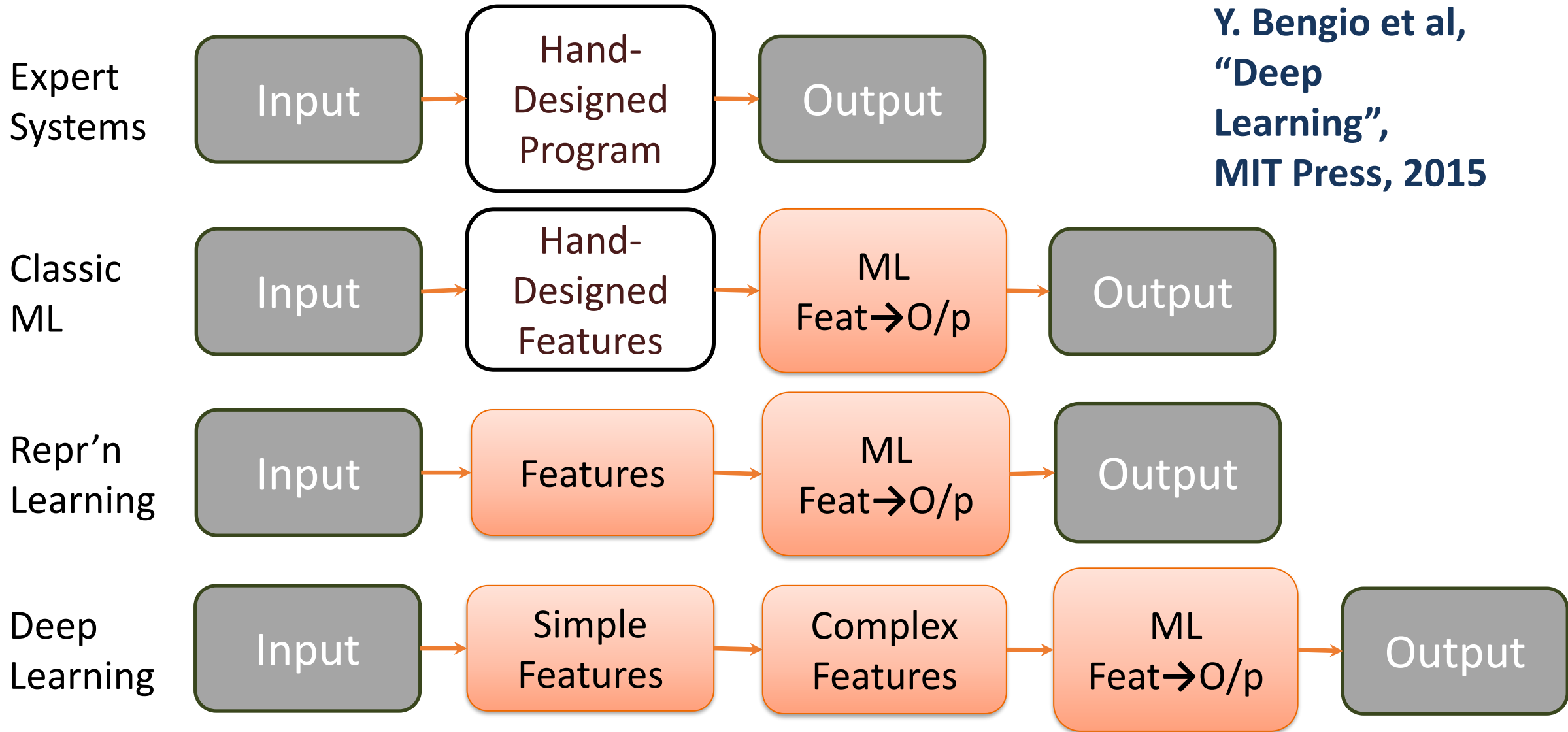
- Input: x (images, text, emails...)
- Output: y (spam or non-spam...)
- (Unknown) Target Function
 - $f: X \rightarrow Y$ (the “true” mapping / reality)
- Data
 - $(x_1, y_1), (x_2, y_2), \dots, (x_N, y_N)$
- Model / Hypothesis Class
 - $g: X \rightarrow Y$
 - $y = g(x) = \text{sign}(w^T x)$
- Learning = Search in hypothesis space
 - Find best g in model class.

Machine learning structure

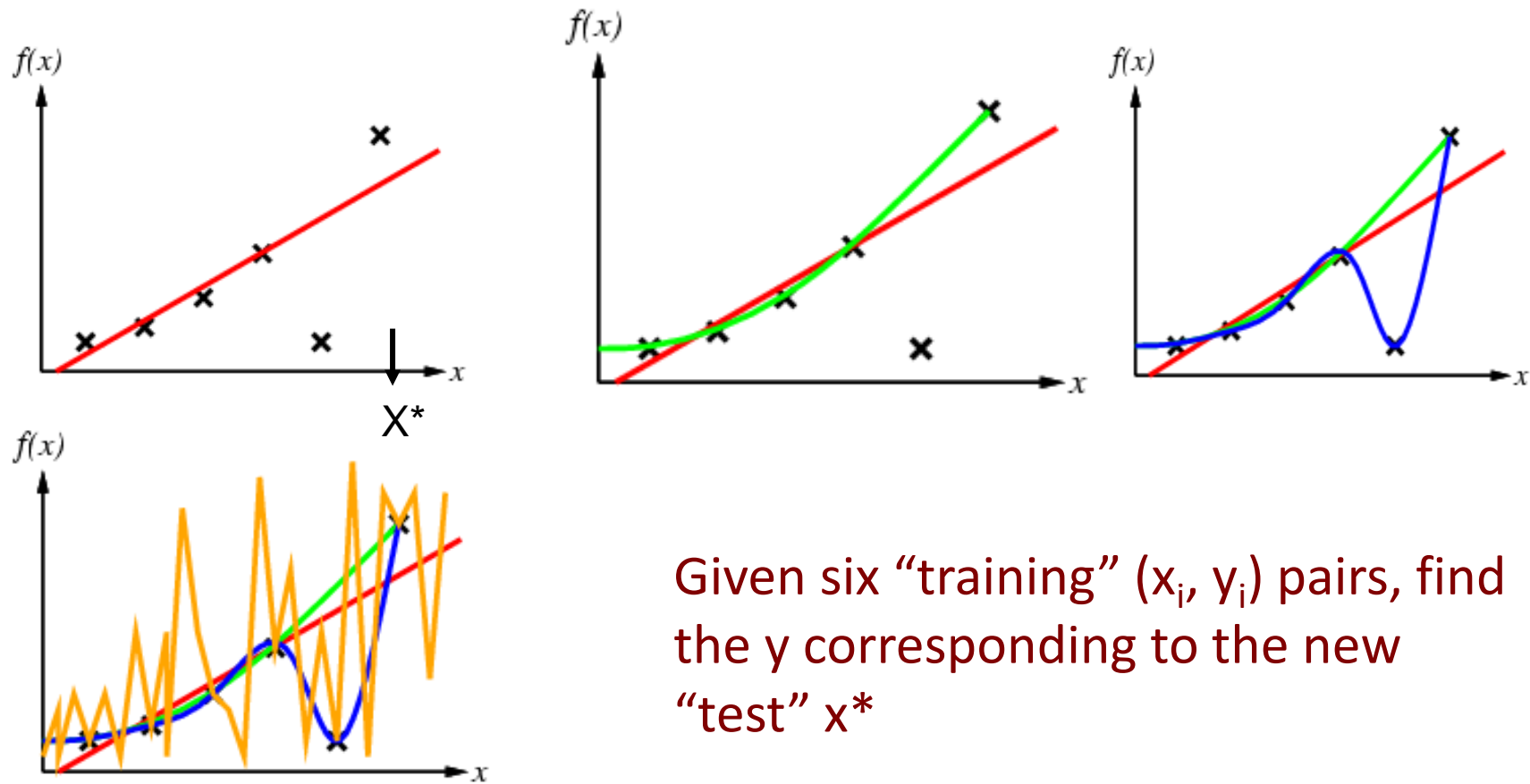
- Supervised learning



Evolution of Learning



A Simple Fitting/Predicting Problem



Given six “training” (x_i, y_i) pairs, find the y corresponding to the new “test” x^*

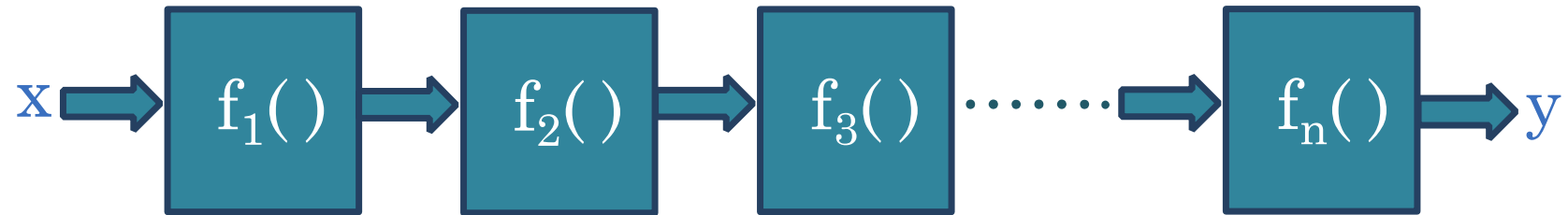
Which curve is the best?

Occam's Razor

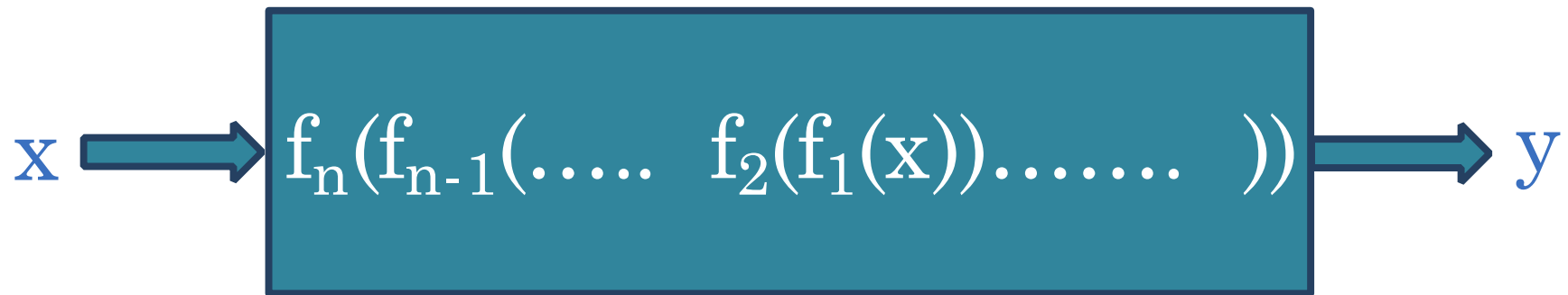
Select the simplest hypothesis (solution) that suits the data.

Eg. Minimize Sum of “fit error” and “degree of the polynomial”

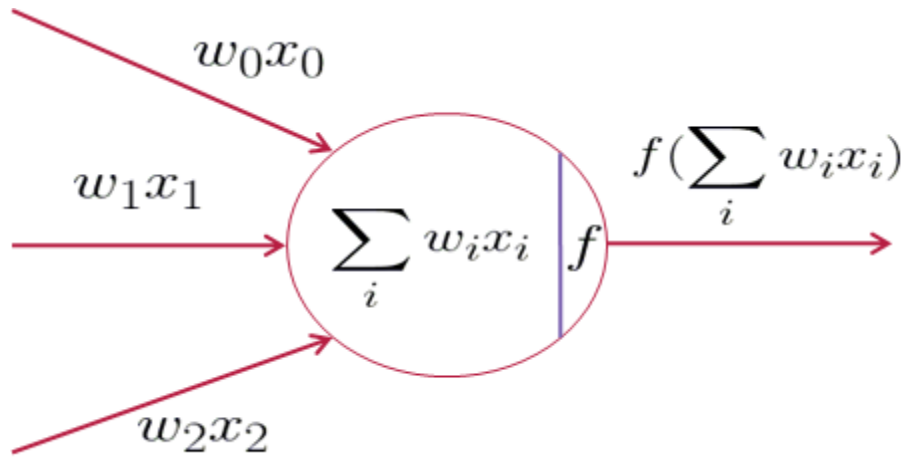
What is Deep Learning?



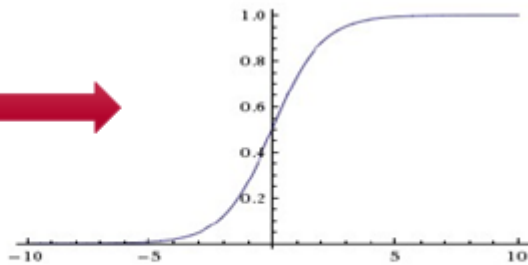
Each function could be simple.



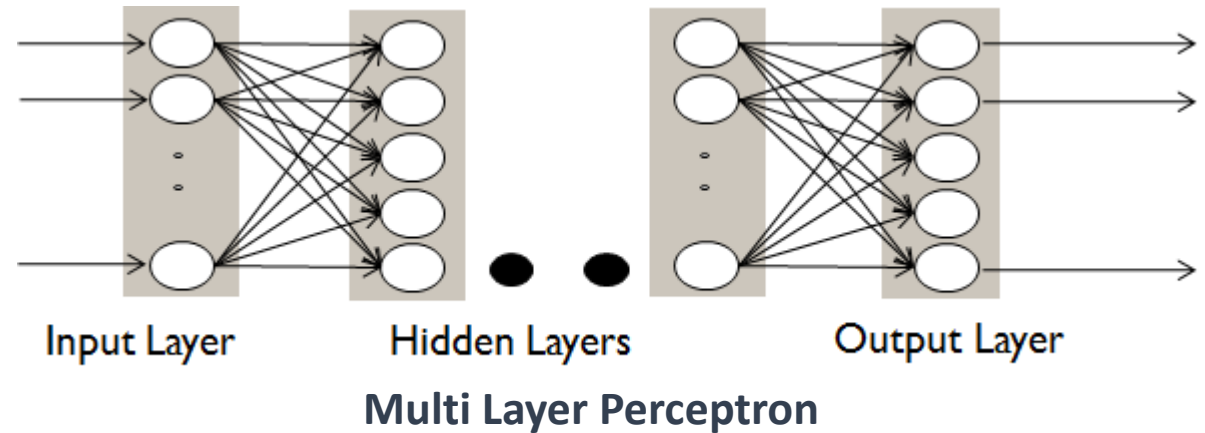
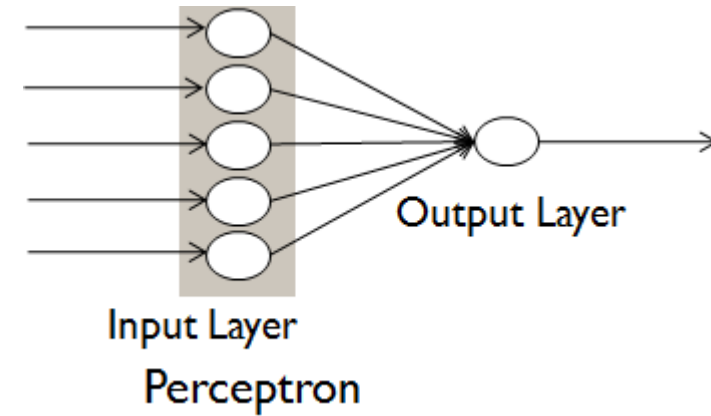
Neuron, Perceptron and MLP



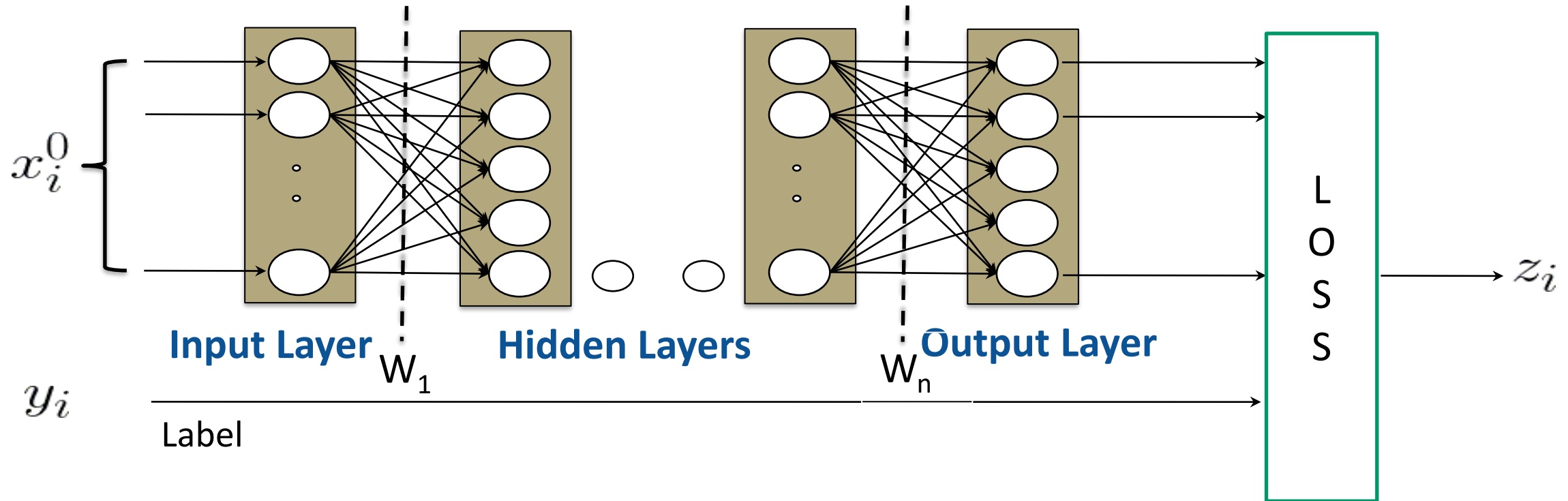
f



E.g. Sigmoid Activation Function



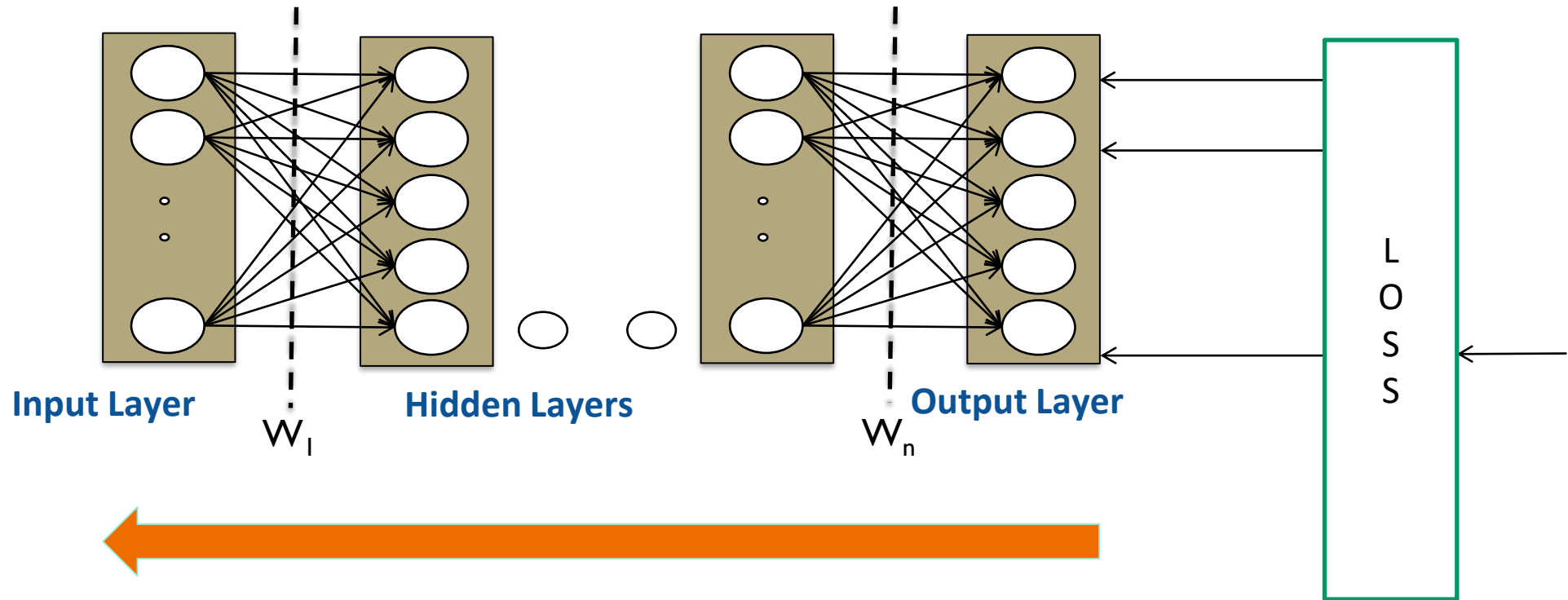
Loss or Objective



Objective: Find out the best parameters which will minimize the loss.

$$W^* = \arg \min_W \sum_{i=1}^N L(x_i^n, y_i; W) \longrightarrow \text{Weight vector}$$
$$z_i = \frac{1}{2} \| x_i^n - y_i \|_2^2 \quad \text{E.g. Squared Loss}$$

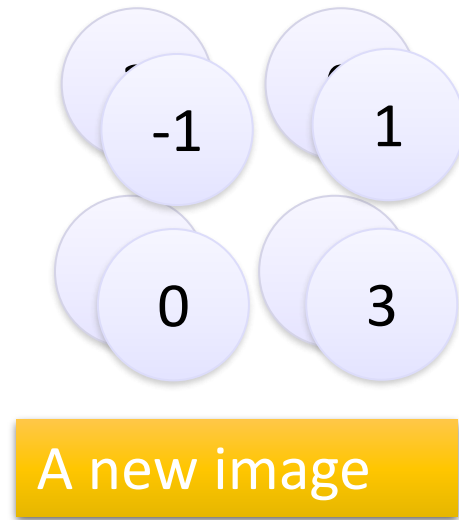
Back Propagation



Solution: Iteratively update W along the direction where loss decreases.

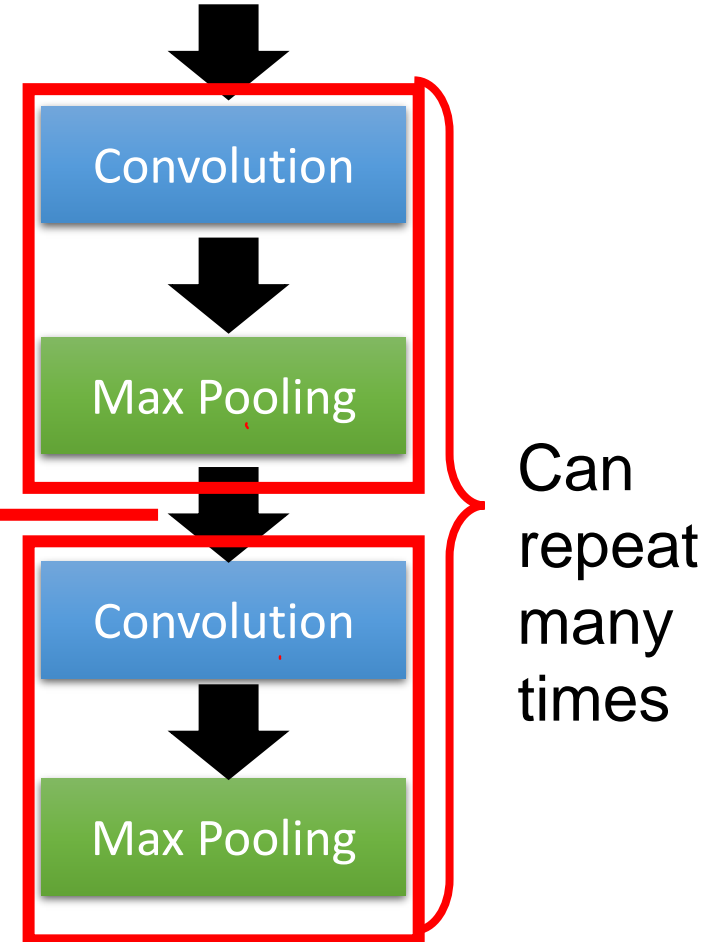
Each layer's weights are updated based on the derivative of its output w.r.t. input and weights

CNN



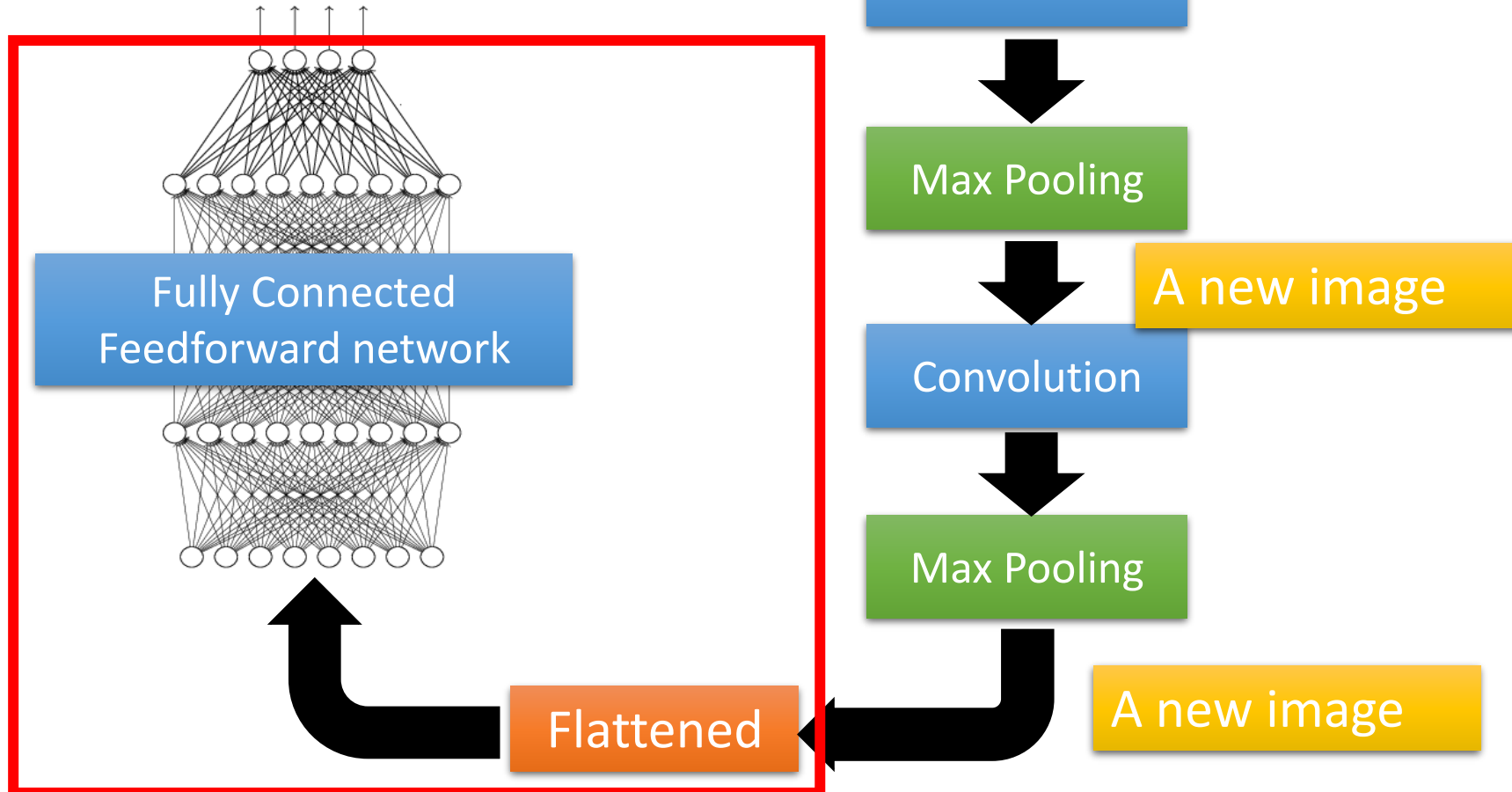
Smaller than the original image

The number of channels is the number of filters

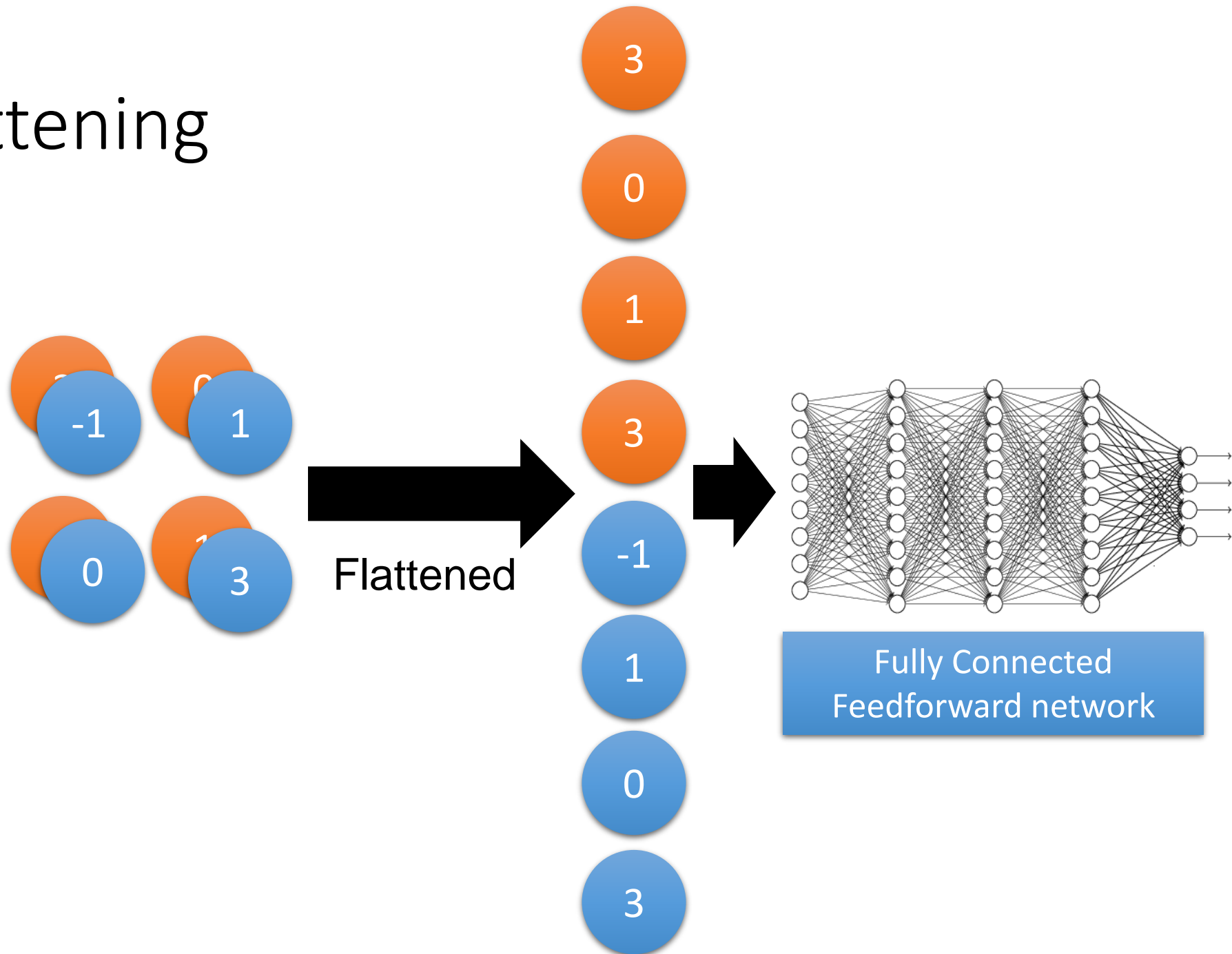


The whole CNN

cat dog

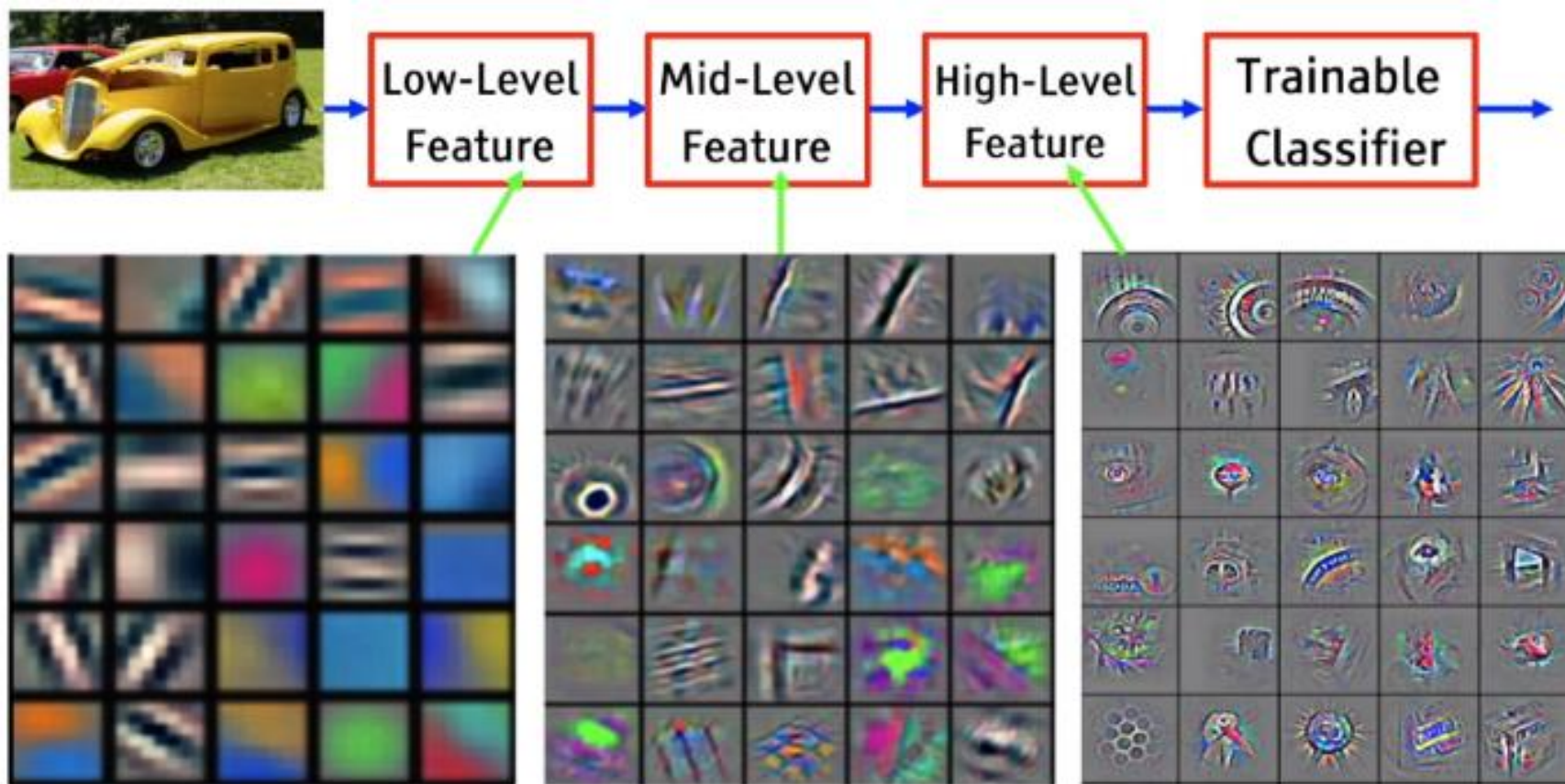


Flattening

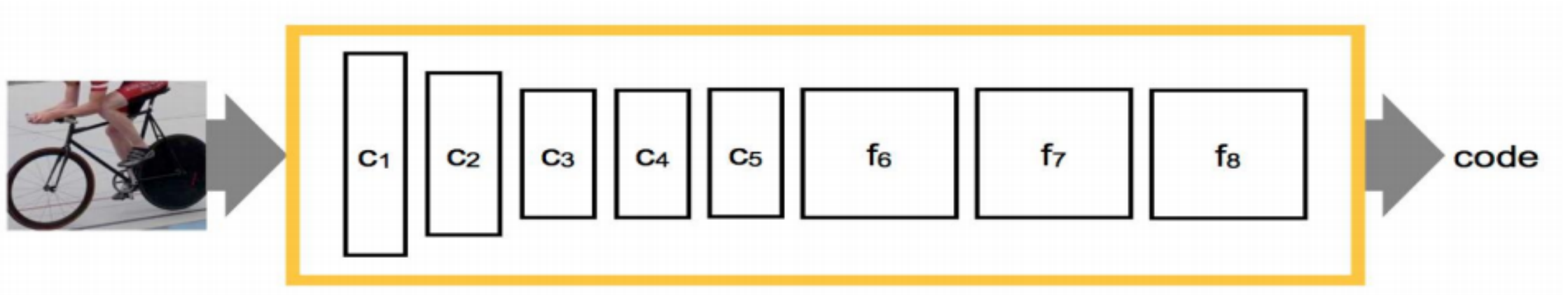
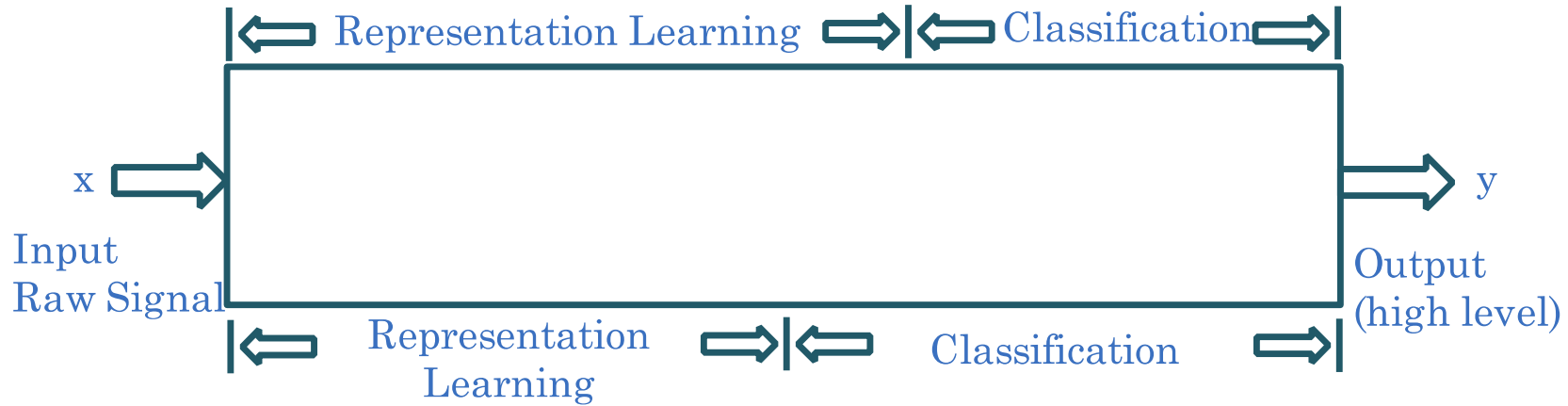


Deep Learnt Features

- It's **deep** if it has **more than one stage** of non-linear feature trans



Re-Usability



Quick Tour: Terminologies

- Training, Testing/Inference, Validation, Overfitting, Generalization
- Optimization, Regularization, Backpropagation, Initialization
- MLP, CNN, RNN, LSTM, GRU, Transformer
- Fully Connected, Convolutional, Attention, Normalization,
- Weight, Parameters, Sigmoid, Relu, Activation,
- Trained Model, Fine Tune, Deploy, Edge Implementation
- PyTorch, TensorFlow, Python, GPU, Docker..
- Supervised, Self Supervised, Reinforcement, .. Learning
- SVM, Random Forest, MLP, ..

Thank you!

Questions?