



人工智能技术及应用

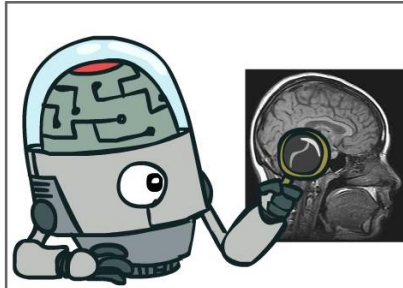
Artificial Intelligence and Application



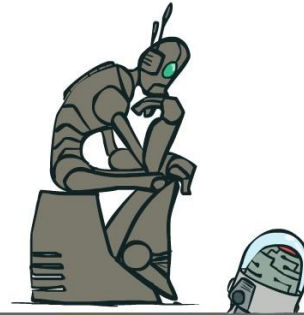
What is AI?

The science of making machines that:

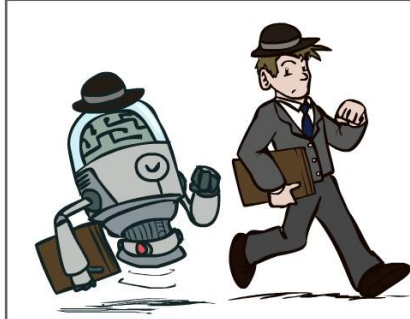
Think like people



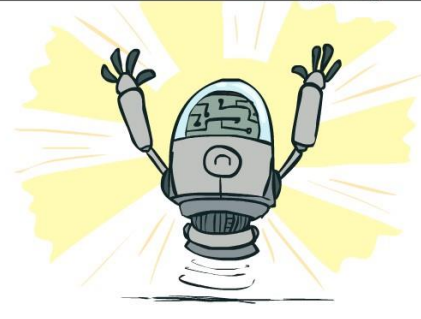
Think rationally



Act like people



Act rationally





Rational Agent

Rational Decisions

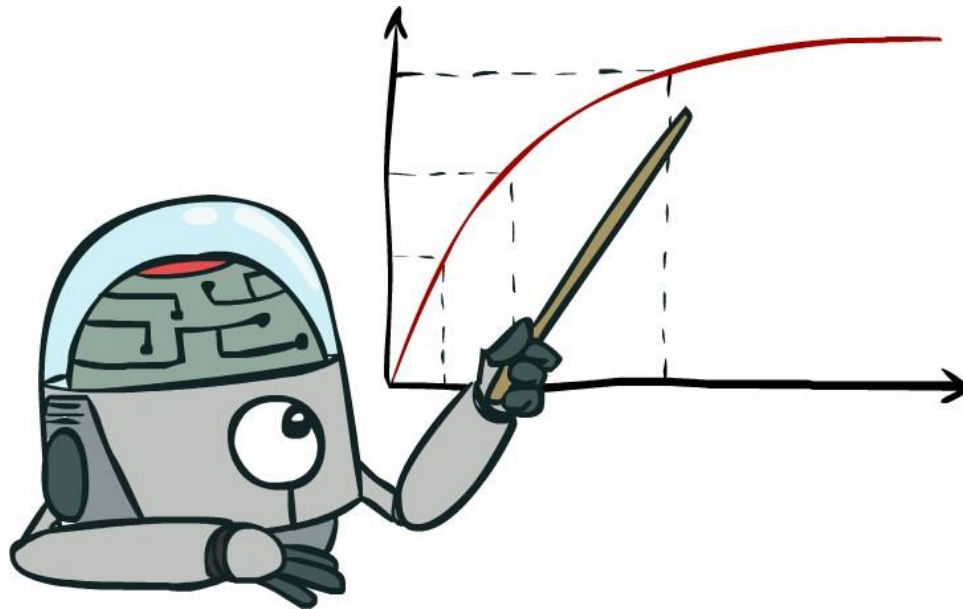
- We'll use the term **rational** in a very specific, technical way:
 - Rational: maximally achieving pre-defined goals
 - Rationality only concerns what decisions are made (not the thought process behind them)
 - Goals are expressed in terms of the **utility** of outcomes
 - Being rational means **maximizing your expected utility**

A better title for this course would be:

Computational Rationality



Maximize Your Expected Utility

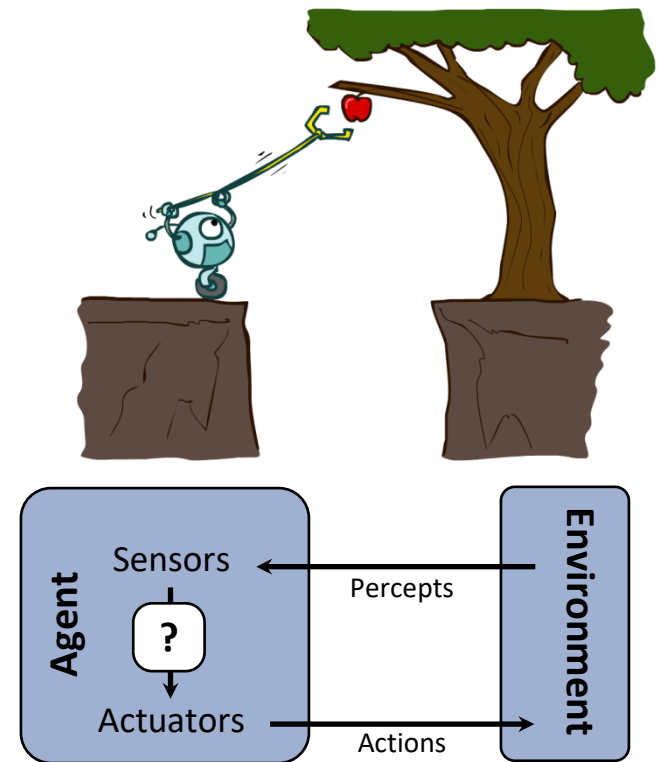




Rational Agent

Designing Rational Agents

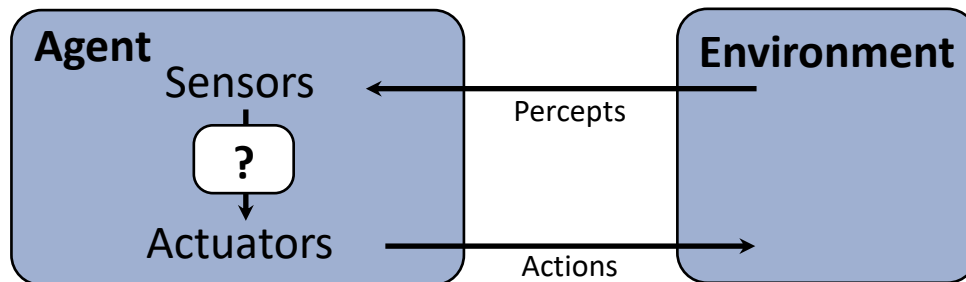
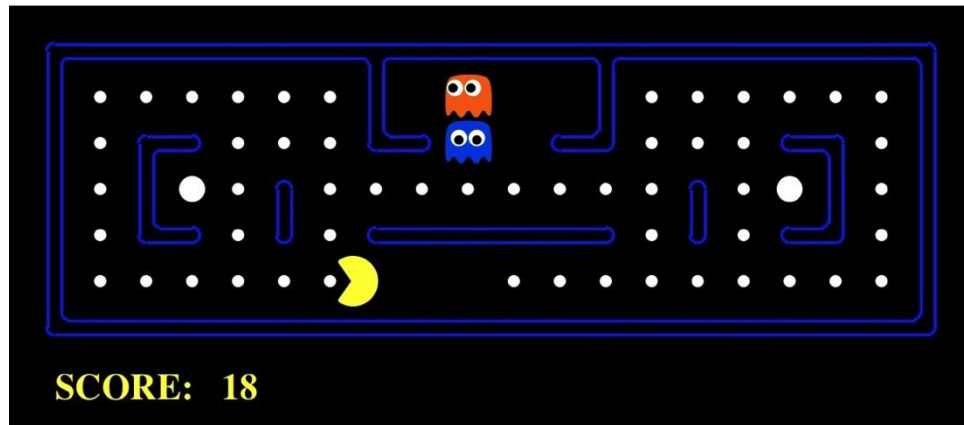
- An **agent** is an entity that *perceives* and *acts*.
- A **rational agent** selects actions that maximize its (expected) **utility**.
- Characteristics of the **percepts**, **environment**, and **action space** dictate techniques for selecting rational actions
- **This course is about:**
 - General AI techniques for a variety of problem types
 - Learning to recognize when and how a new problem can be solved with an existing technique





Rational Agent

Pac-Man as an Agent





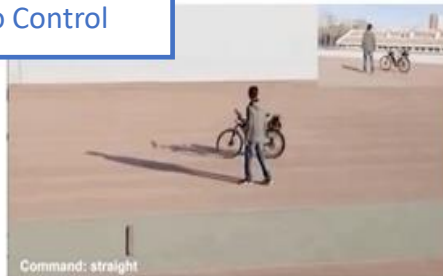
Rational Agent

Agent and Environment



Intelligent agent refers to an autonomous entity which **acts**, **learns** or uses **knowledge** towards achieving goals, upon an **environment** using observation through **sensors** and consequent **actuators**.

Audio Control



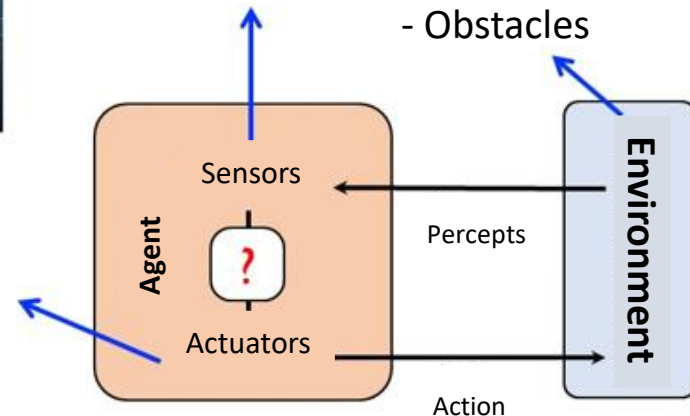
People Tracking



Actuators:
- Steering
- Chain
- Brake

Sensors:
- Camera
- Microphone

Environment:
- Streets
- Pedestrians
- Obstacles



Towards artificial general intelligence with hybrid Tianjic chip architecture. Nature. 2019.



Rational Agent

Two Views

- Intelligence Agent
 - Think like human
 - Act like human

how can we create intelligence?

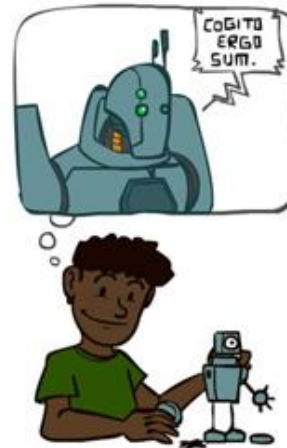
Perception Robotics Language



Knowledge Reasoning Learning

- AI Tools
 - Try to solve problems
 - Human cannot do well

how can we benefit society?



- Planning
- Scheduling
- Medical
- Image
- Topic Model
- Prediction



Rational Agent

Intelligence Agent

Keep Knowledge

- Procedural: coding, riding bike,
- Declarative: where is classroom, ...
classroom, ...

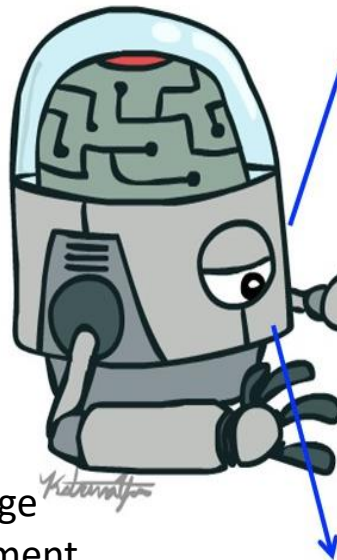
Reasoning

- Inference: exam,
- Decision: dining hall
for lunch, ...

Learning

- Learn new knowledge
- Adapt new environment

Think Like Human



Perceive the
world

Computer
vision

Perform actions

Robotics

Communicate with each
other

Natural language
Processing

Act Like Human





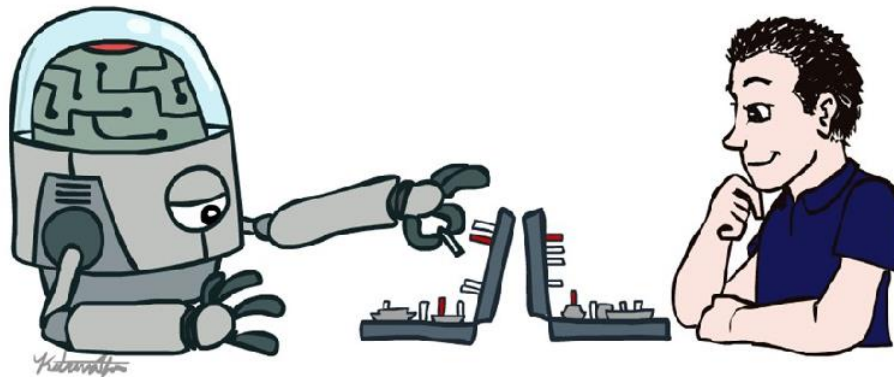
Are We There Yet

Machines

narrow tasks, millions of examples

Human

diverse tasks, very few examples



AlphaGo

- Learn from 19.6 million games
- Only play one game Go

Human

- Learn from wide set of experiences
- Do lot of tasks

We are still very far away from **Artificial General Intelligence (AGI)**.



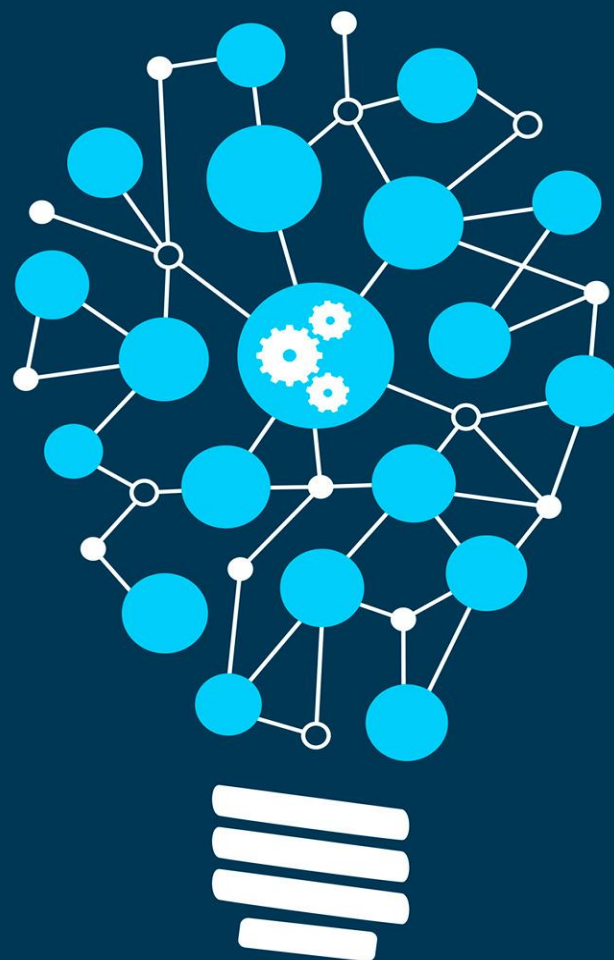
Towards AI

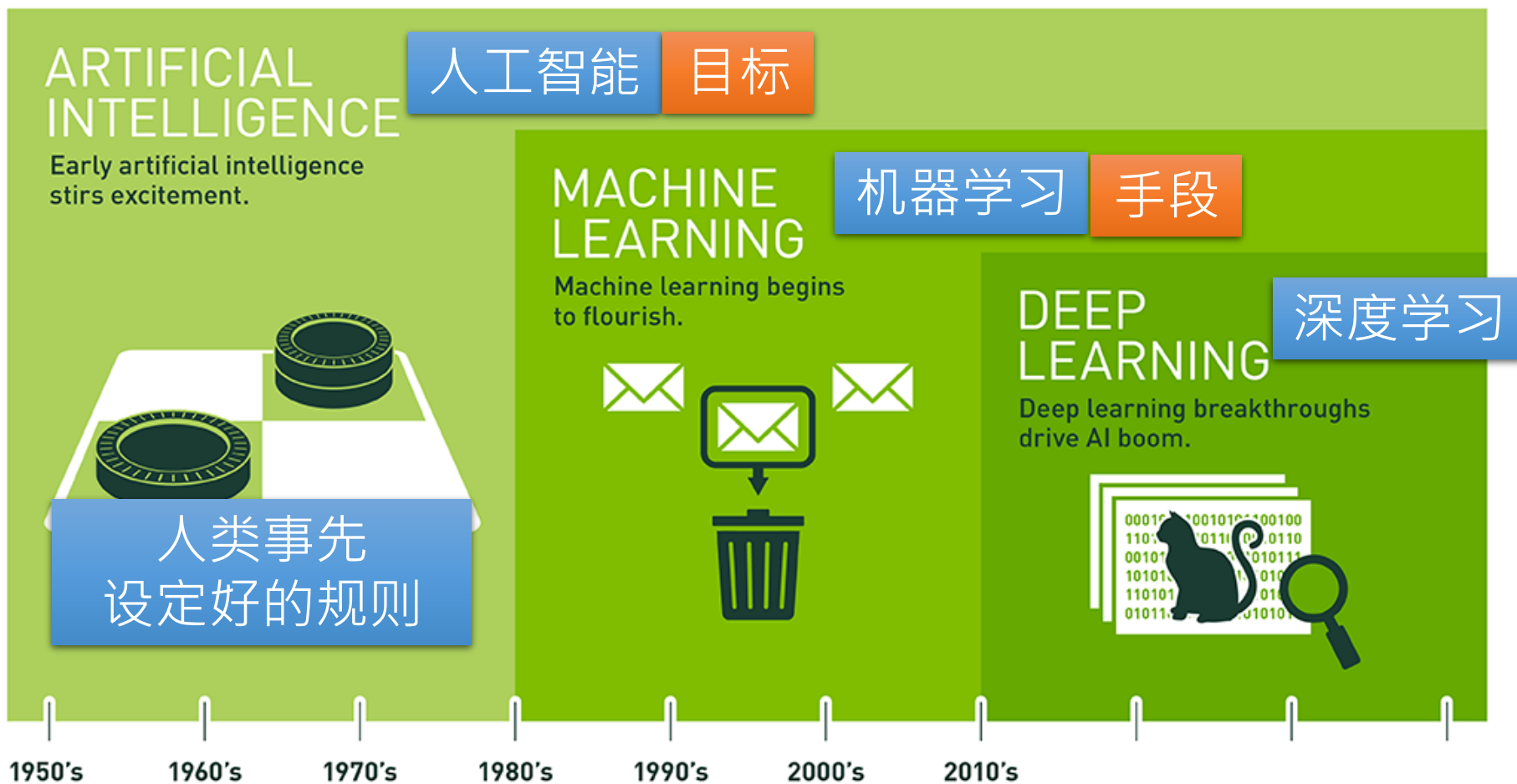
AI Methods

Applications

Search	Uniform Cost Search	CSP and Local Search	Adversarial Search	Early AI programs(1950s) Games: AlphaGo(2016)
Symbolic AI Logic	First-Order Logic	Propositional Logic		Knowledge based systems(1970-1980s)
Neural AI Learning	Statistical Learning	Deep Learning		Widely used in recent applications(2010-)
Statistical AI Uncertain Reasoning	Probabilistic Reasoning	Sequential Reasoning	Causal Reasoning	Topic Model (2000s) Weather Prediction
Communicating Perceiving Acting	NLP	CV	Robotics	

什么是
机器学习？





Since an early flush of optimism in the 1950s, smaller subsets of artificial intelligence – first machine learning, then deep learning, a subset of machine learning – have created ever larger disruptions.

Source of image: <https://blogs.nvidia.com.tw/2016/07/whats-difference-artificial-intelligence-machine-learning-deep-learning-ai/>

人类事先设定好的规则 (rule-based)

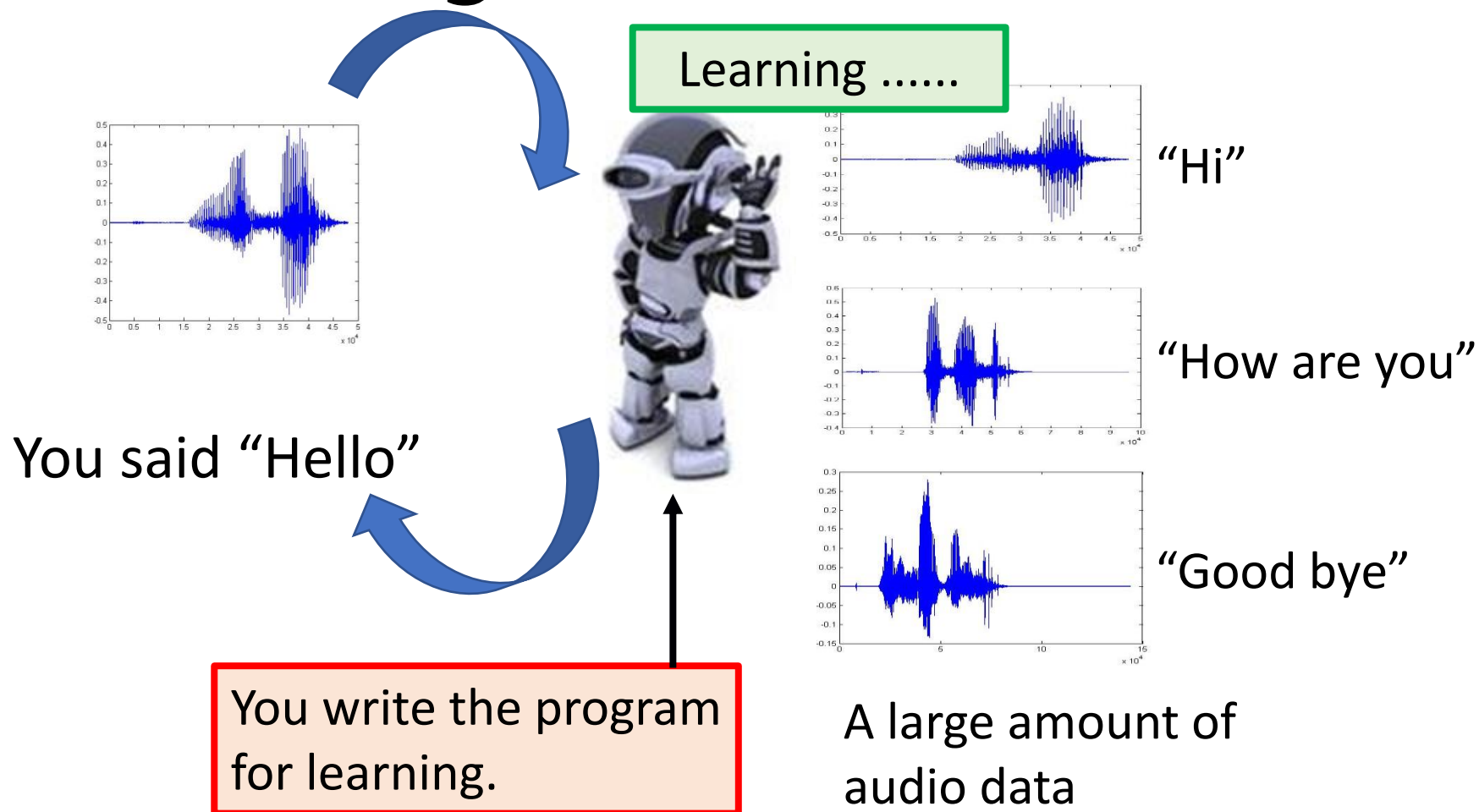
- E.g. You want to build a Chat-bot ...
 - If there is “turn off” in the input, then “turn off the music” (hand-crafted rules)
 - You can say “Please turn off the music” or “Can you turn off the music?”. Smart?
 - What if someone says “Please don’t turn off the music”
- Weakness of hand-crafted rules
 - Hard to consider all possibilities
 - 永远无法超越创造者
 - Lots of human efforts (not suitable for small industry)

人类事先设定好的规则

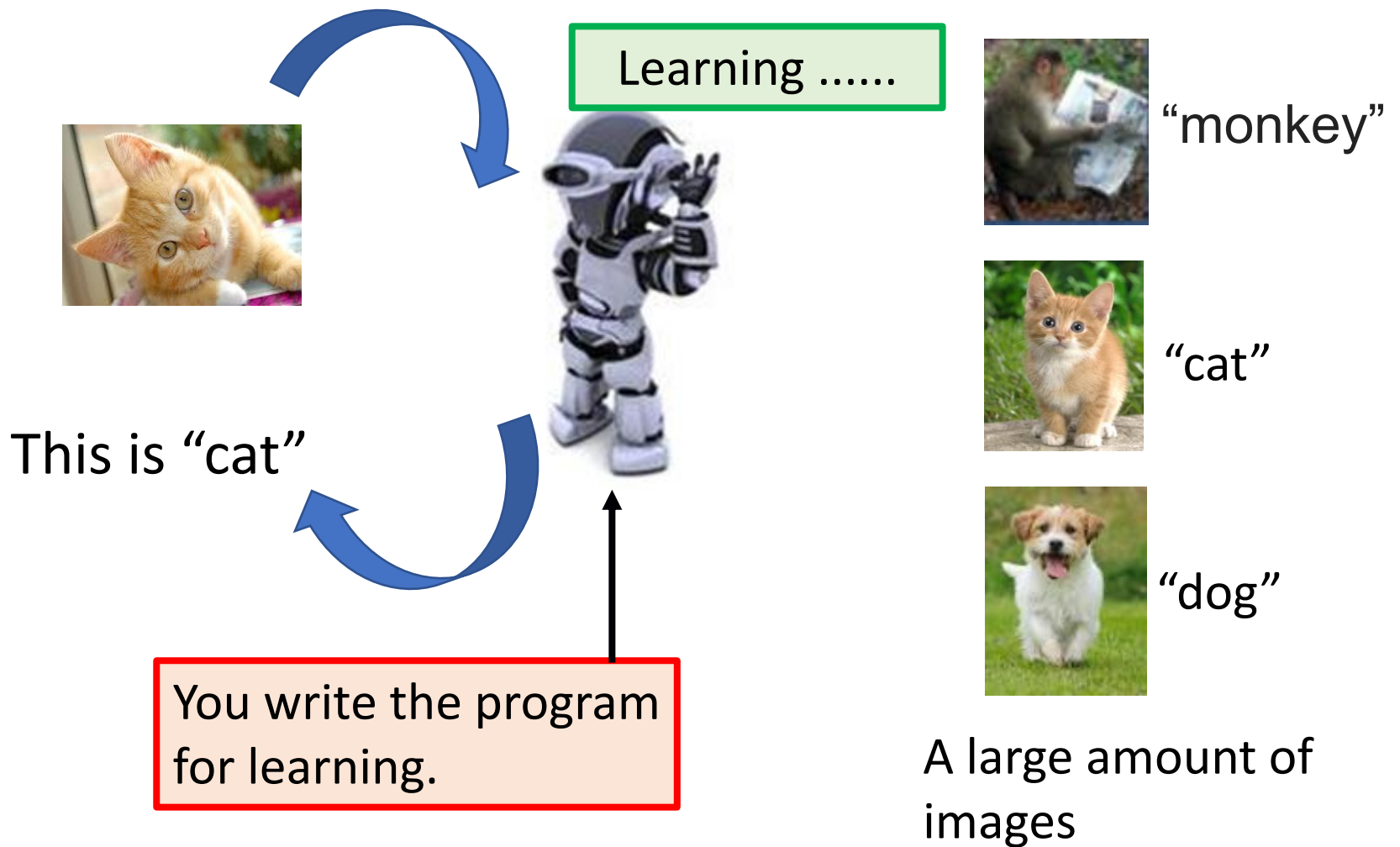
- AI?



机器学习登场！ (learning-based)



机器学习登场！



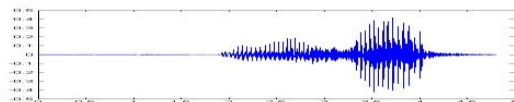
机器学习

≈ 找一个函数的能力

根据数据

- Speech Recognition

$$f(\text{audio waveform}) = \text{"How are you"}$$



- Image Recognition

$$f(\text{cat image}) = \text{"Cat"}$$



- Playing Go

$$f(\text{Go board state}) = \text{"5-5"} \quad (\text{next move})$$



- Dialogue System

$$f(\text{"How are you?"} \text{ (what the user said)}) = \text{"I am fine."} \text{ (system response)}$$

Learning Map



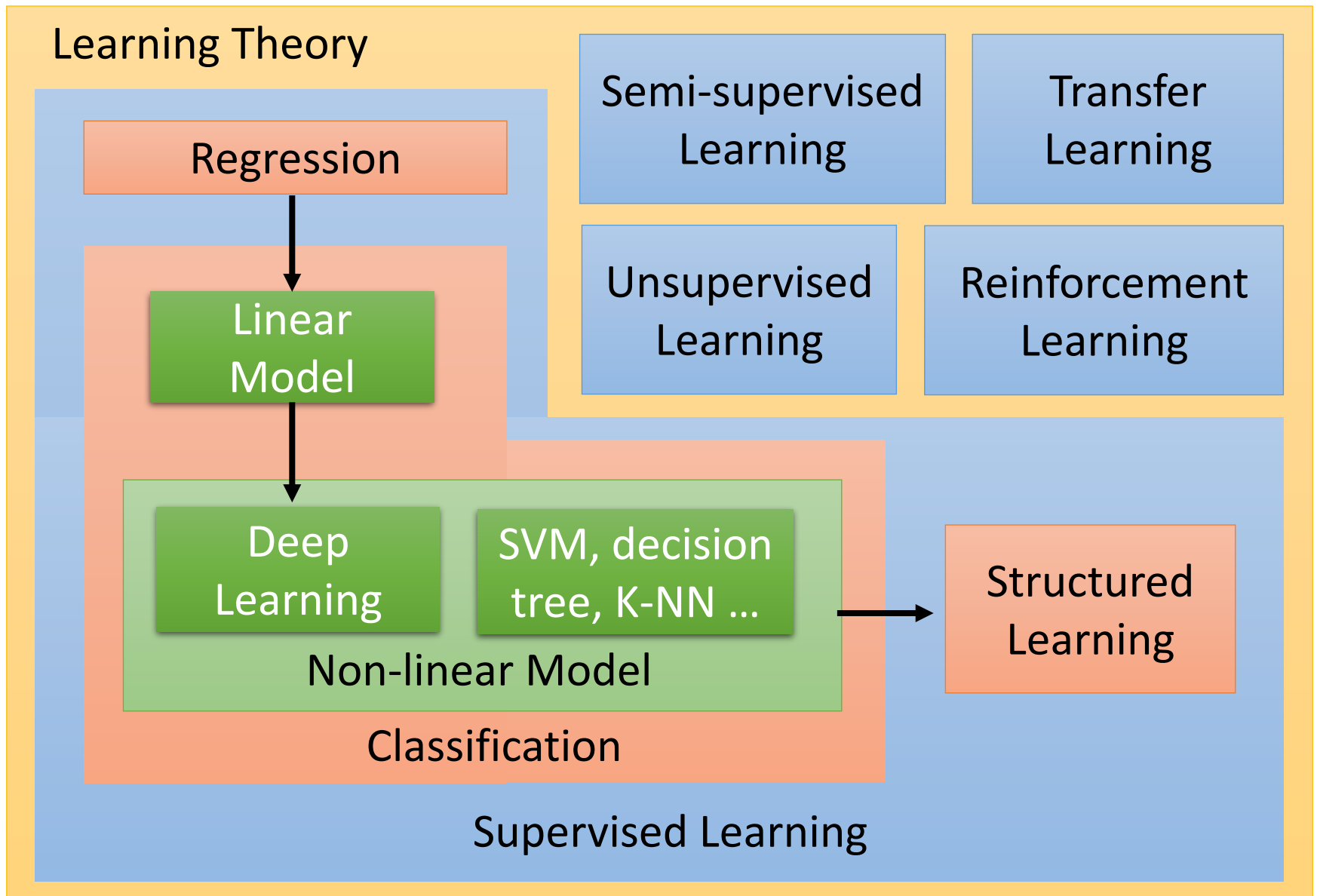
scenario



task



method



Step 0

你想找
什么样的函数？



Regression (回归)

Regression

The output of the target function f is “scalar”.

Predict
PM2.5



Training Data:

Input:

9/01 PM2.5 = 63 9/02 PM2.5 = 65

Input:

9/12 PM2.5 = 30 9/13 PM2.5 = 25

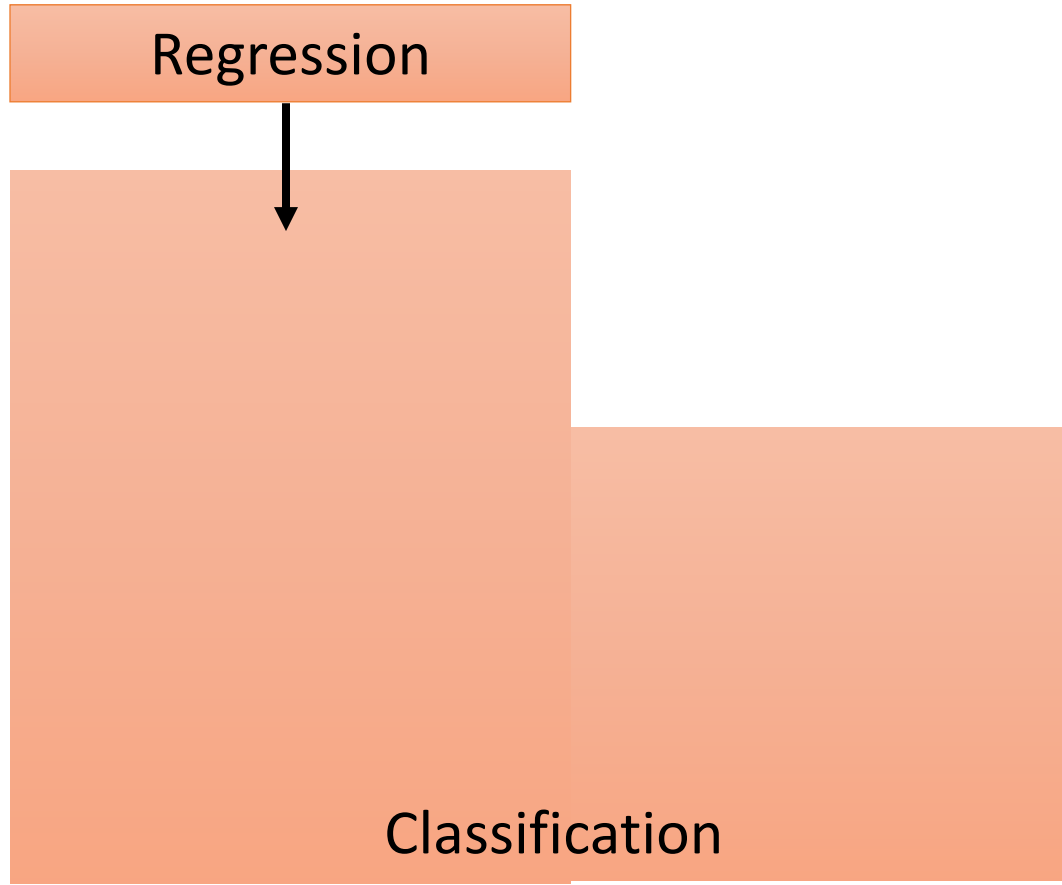
Output:

9/03 PM2.5 = 100

Output:

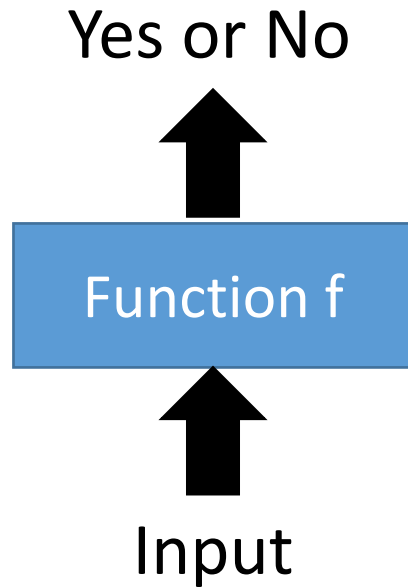
9/14 PM2.5 = 20

Learning Map

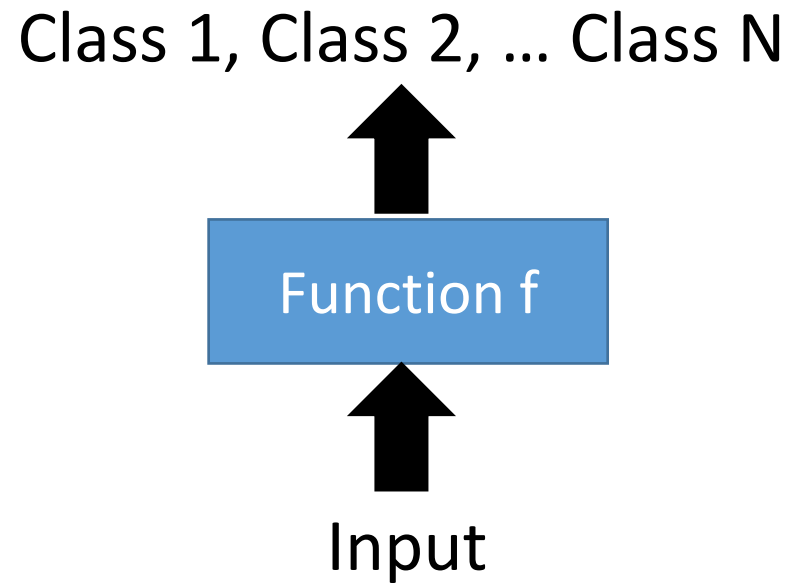


Classification (分类)

- Binary Classification
(二元分类)

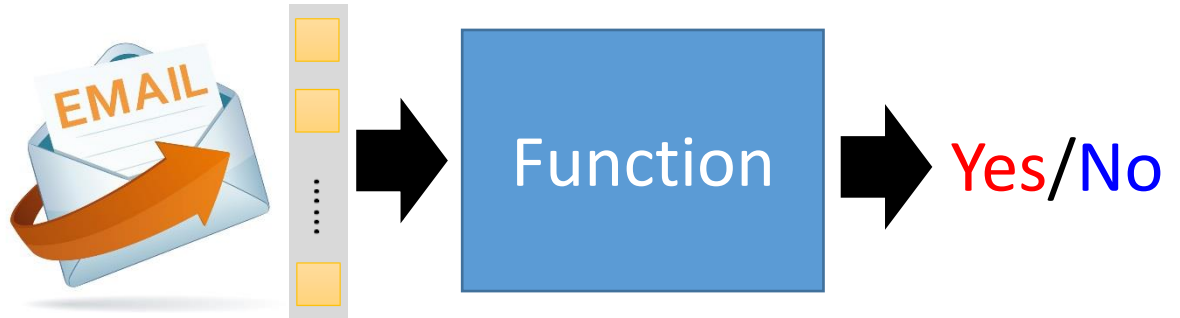


- Multi-class Classification
(多类别分类)



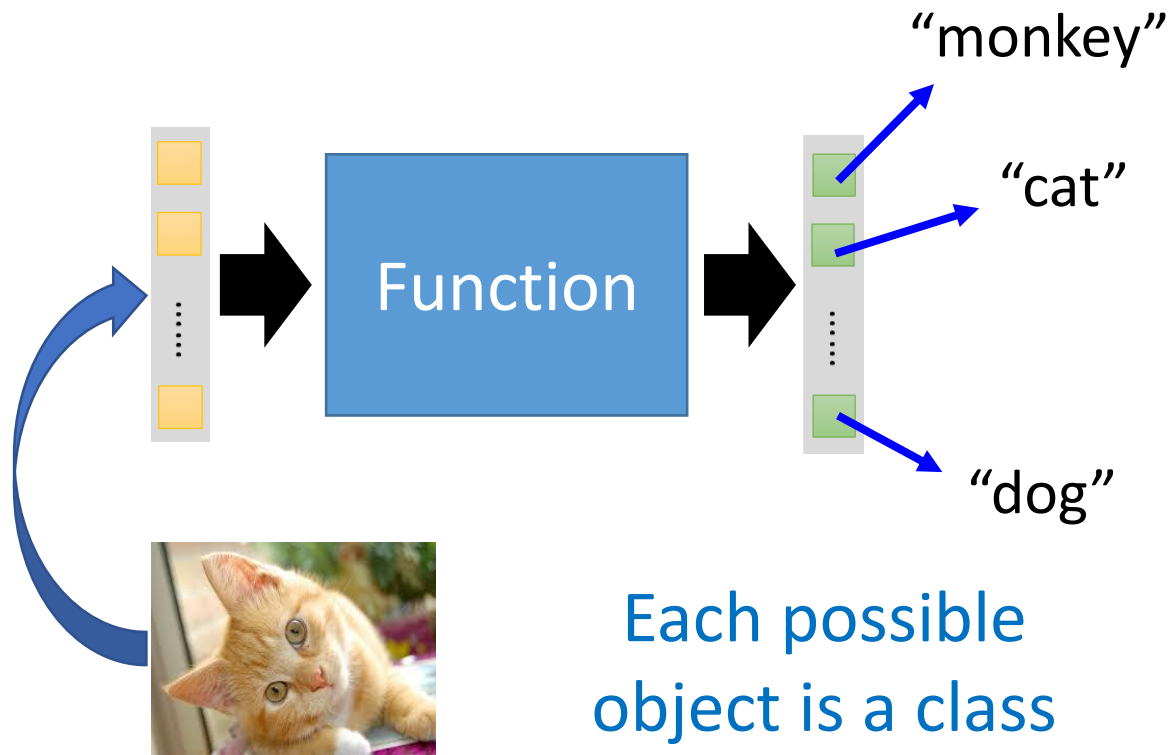
二元分类

Spam
filtering



多类别分类

Image Recognition



Training Data



"monkey"



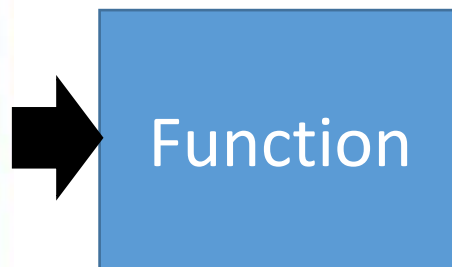
"cat"



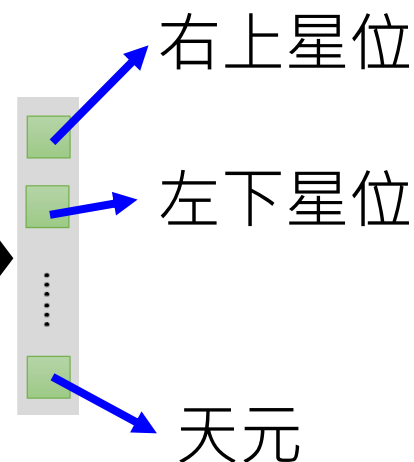
"dog"

多类别分类

Playing GO



Each position
is a class
(19 x 19 classes)



Next move

Generation (生成)

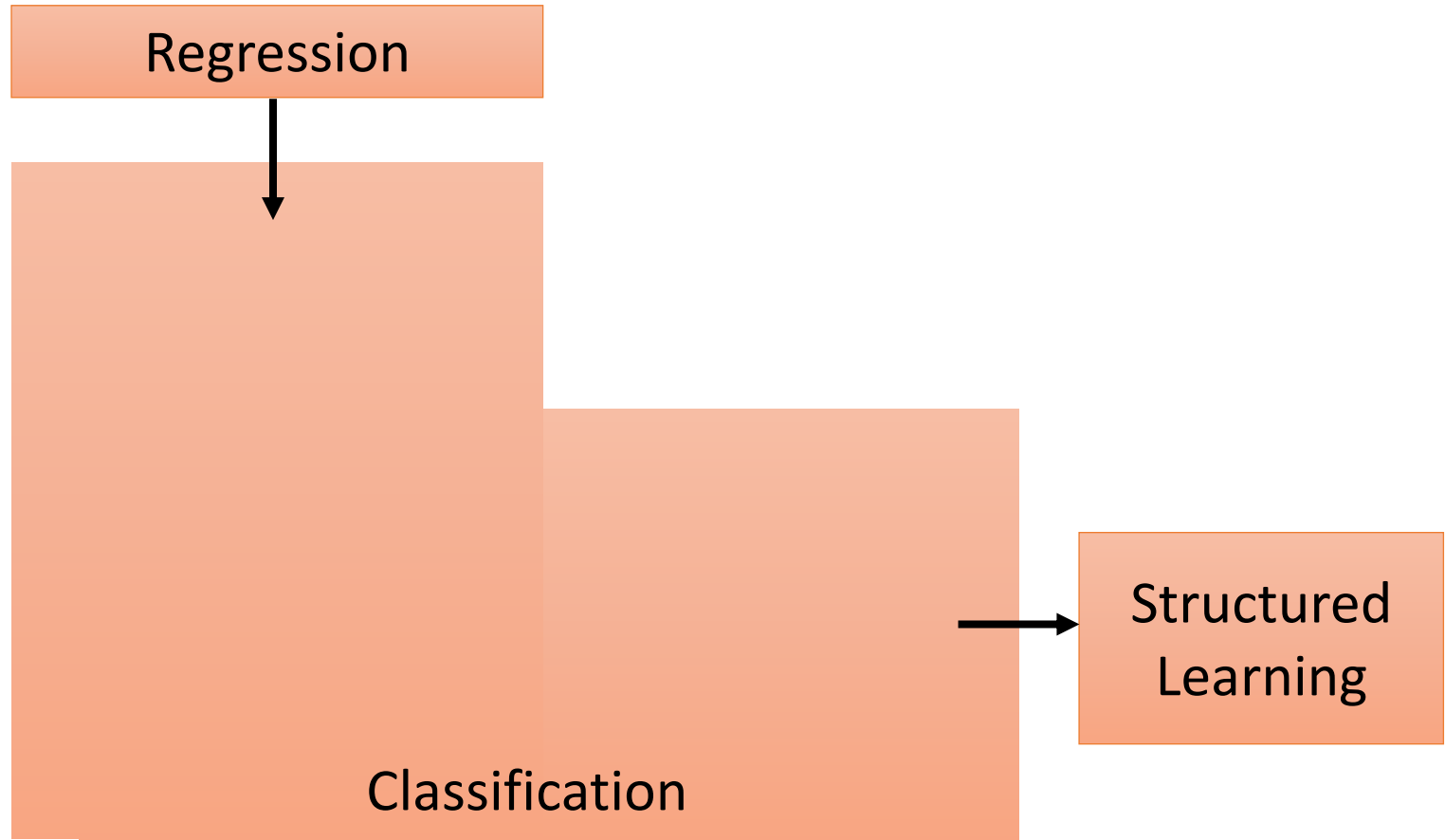
产生有结构的复杂数据
(例如：文本、图片)

拟人化的讲法—创造

Regression,
Classification



Learning Map



Structured Learning

- Beyond Classification

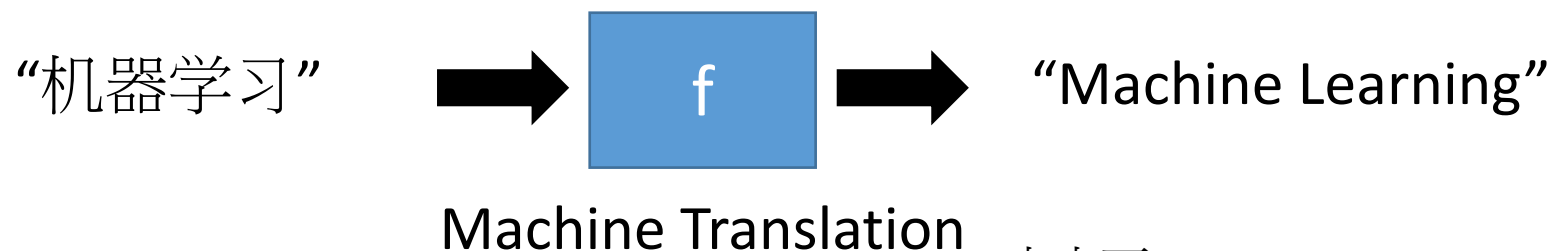
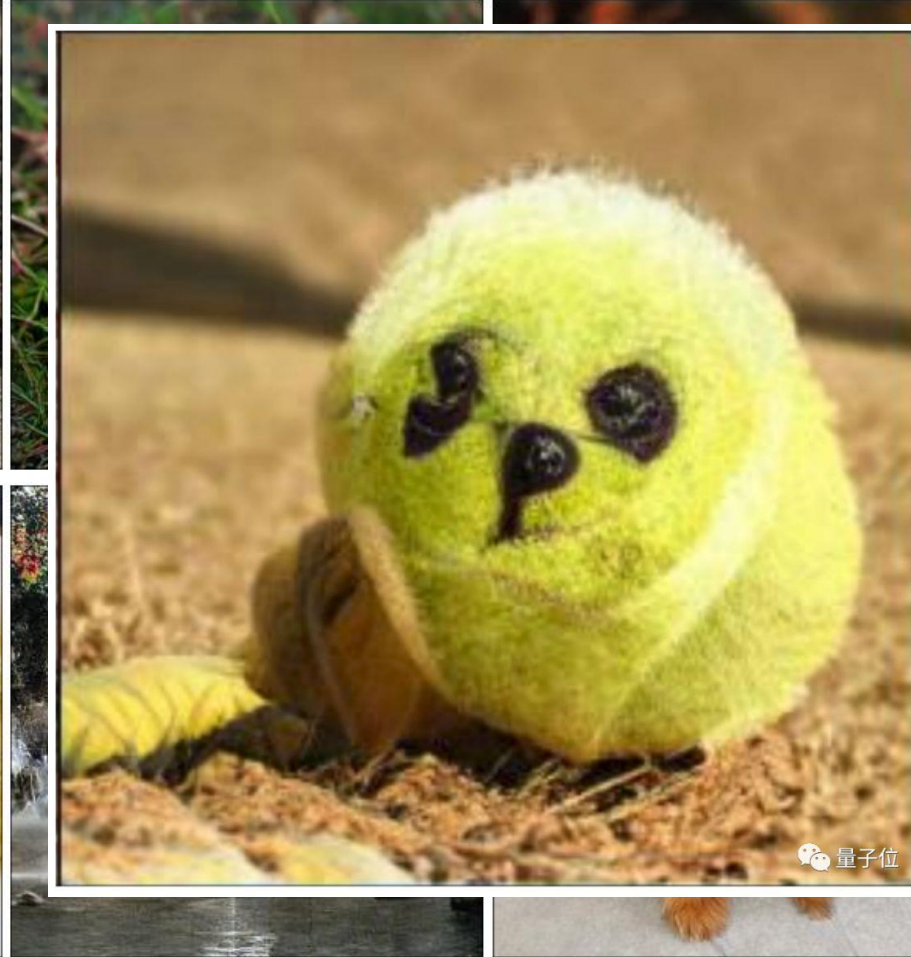


Image Generation

<https://papers.nips.cc/paper/5423-generative-adversarial-nets.pdf>

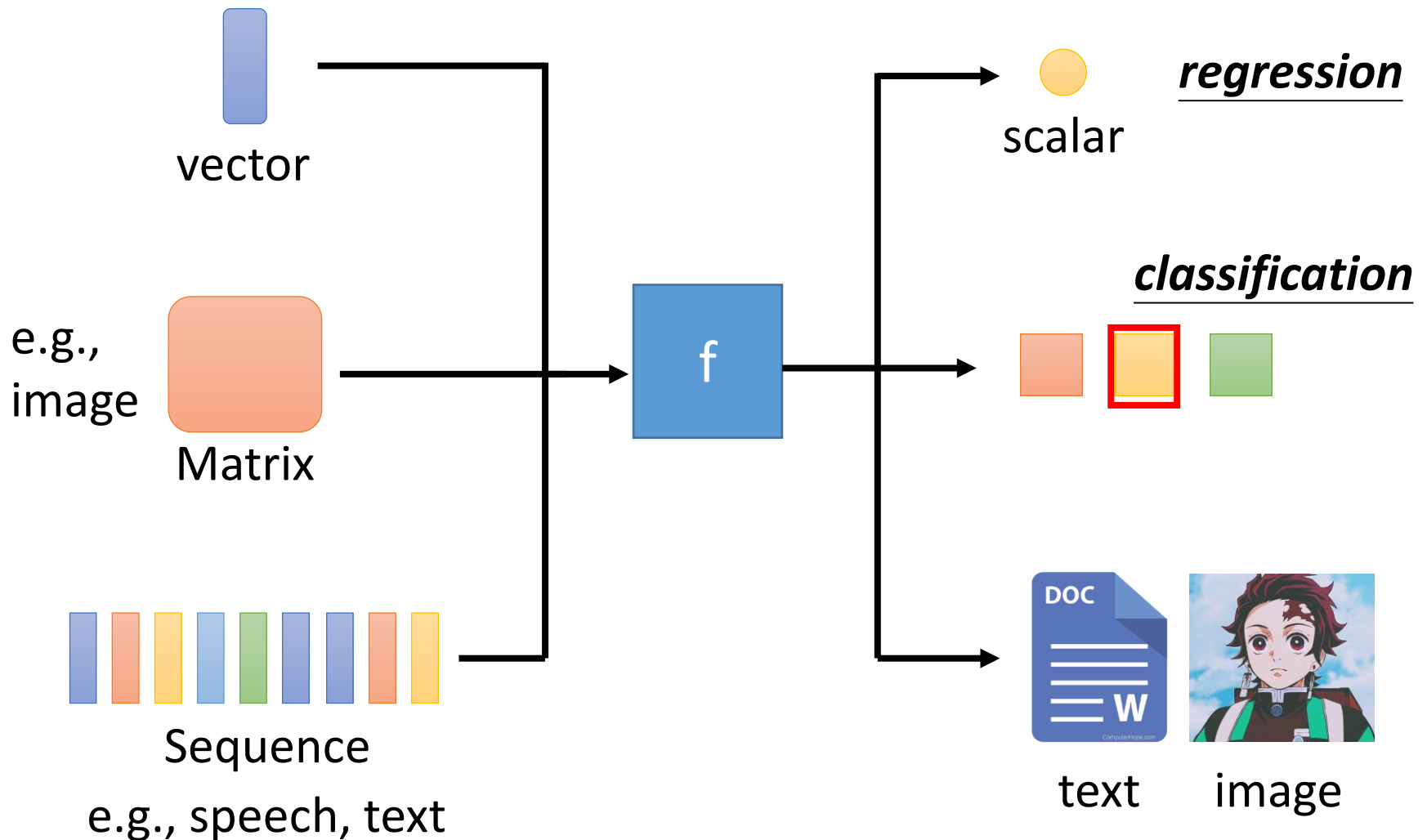


<https://arxiv.org/abs/1809.11096>

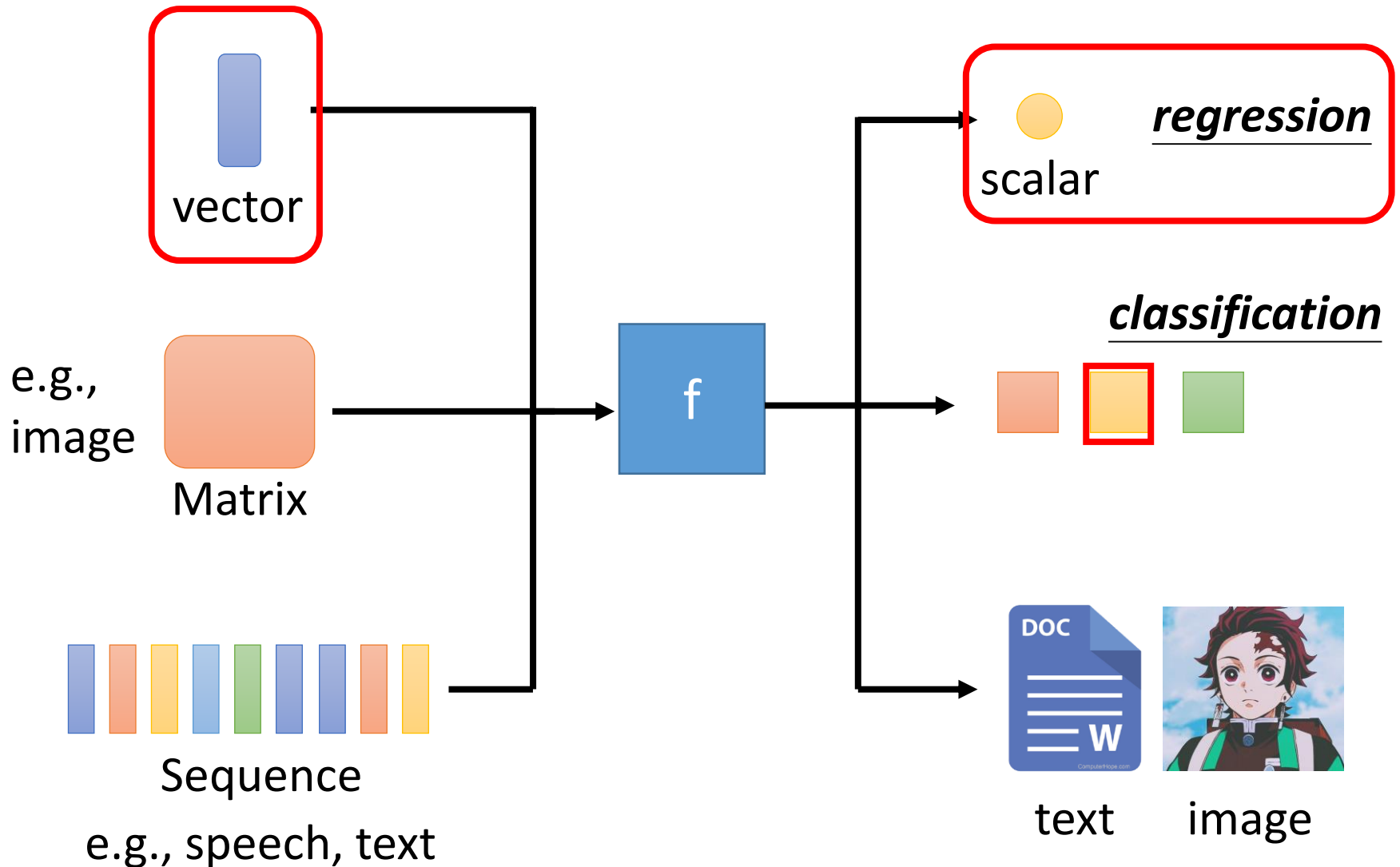


Generative Models: variational auto-encoder (VAE), generative adversarial network (GAN), Flow-based generative model, etc.

Different types of Functions



PM2.5 Prediction



Phoneme Classification

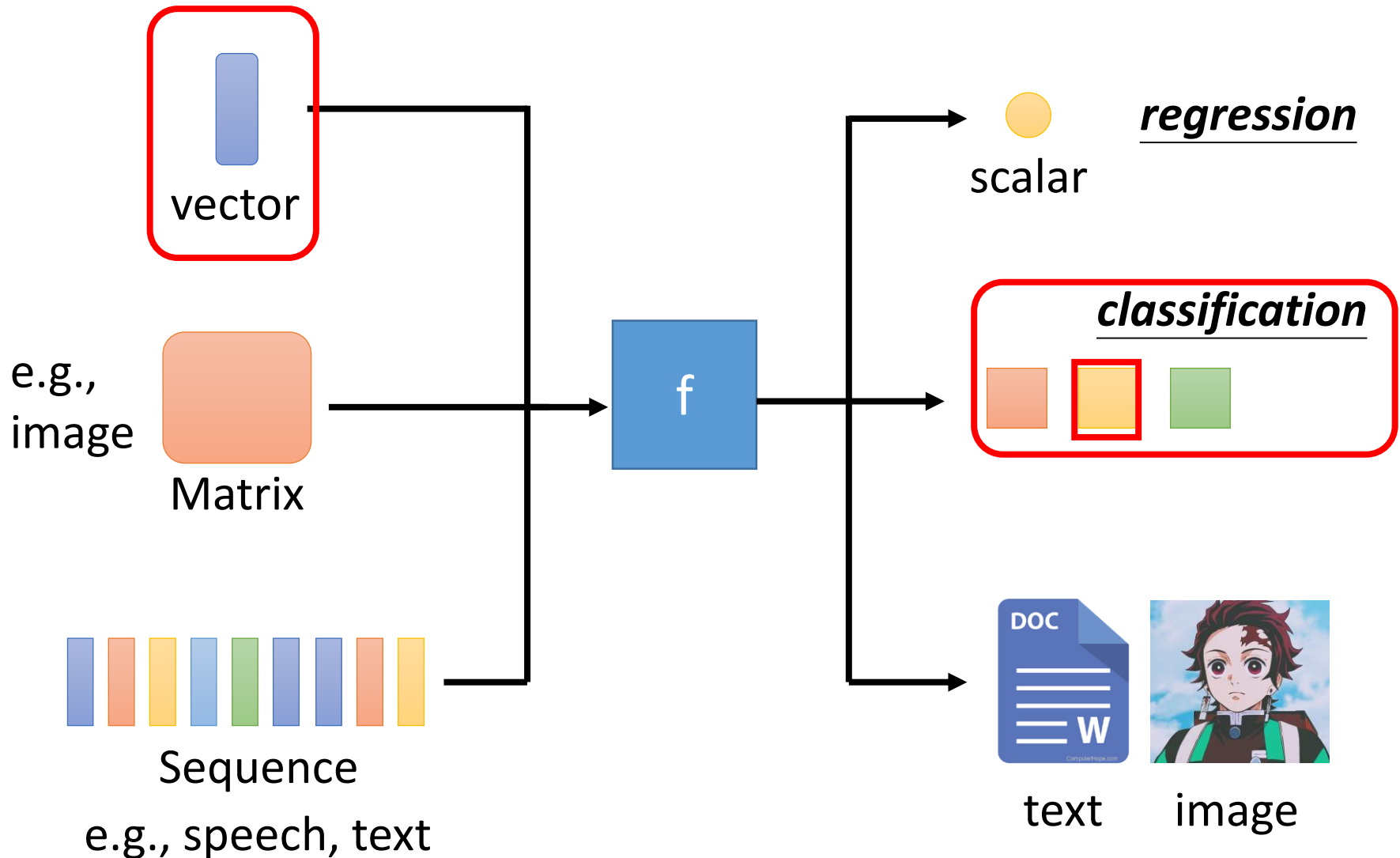
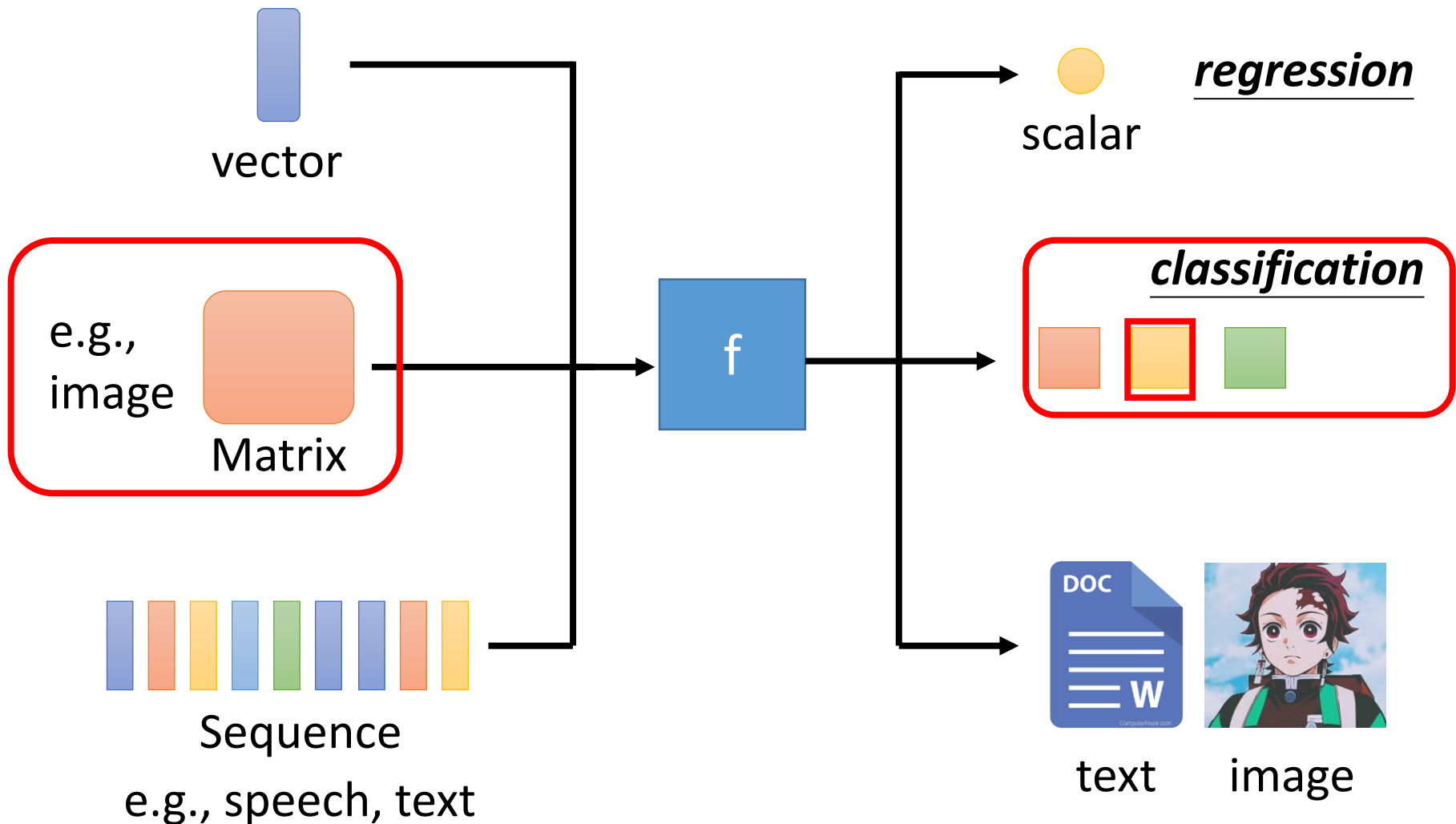
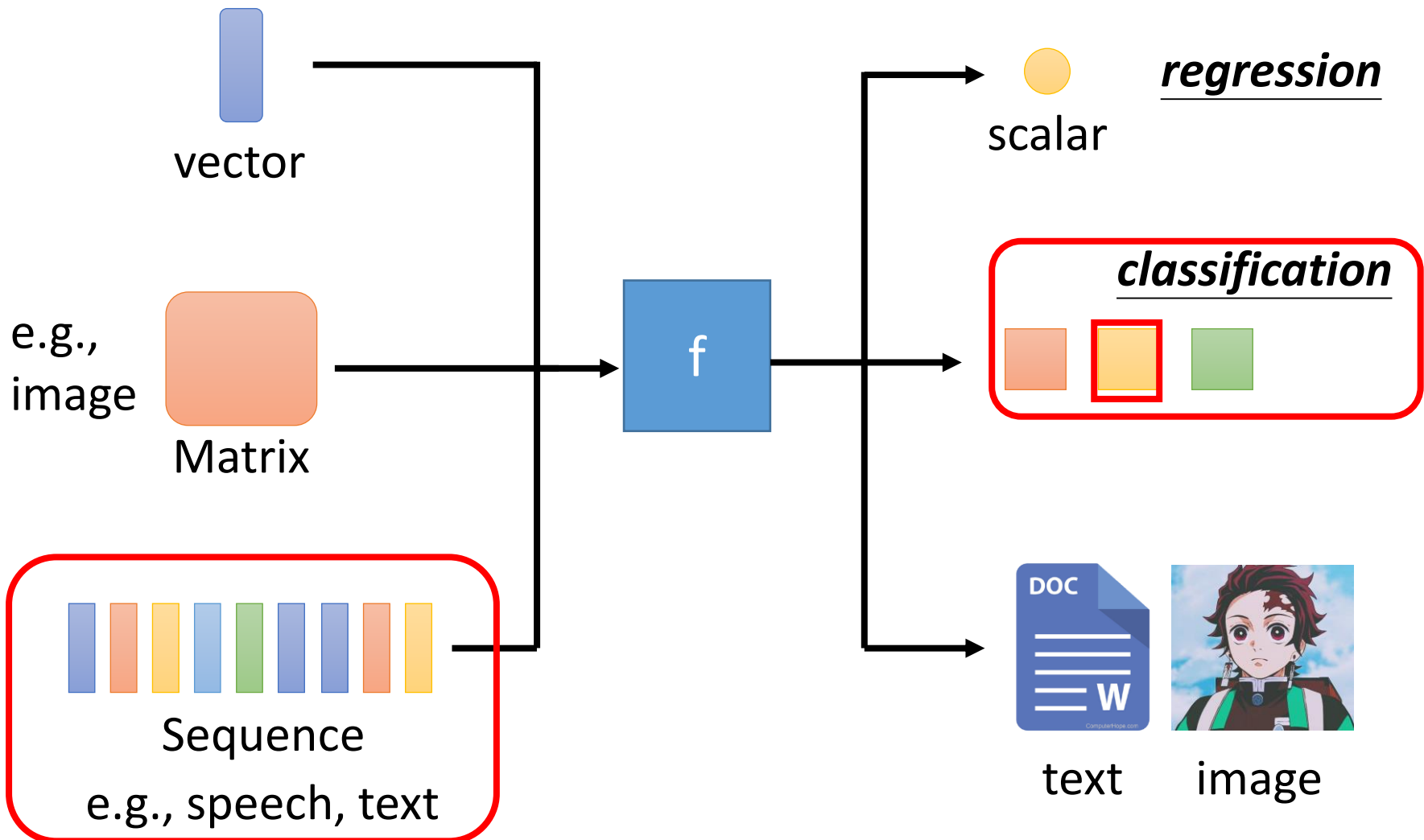


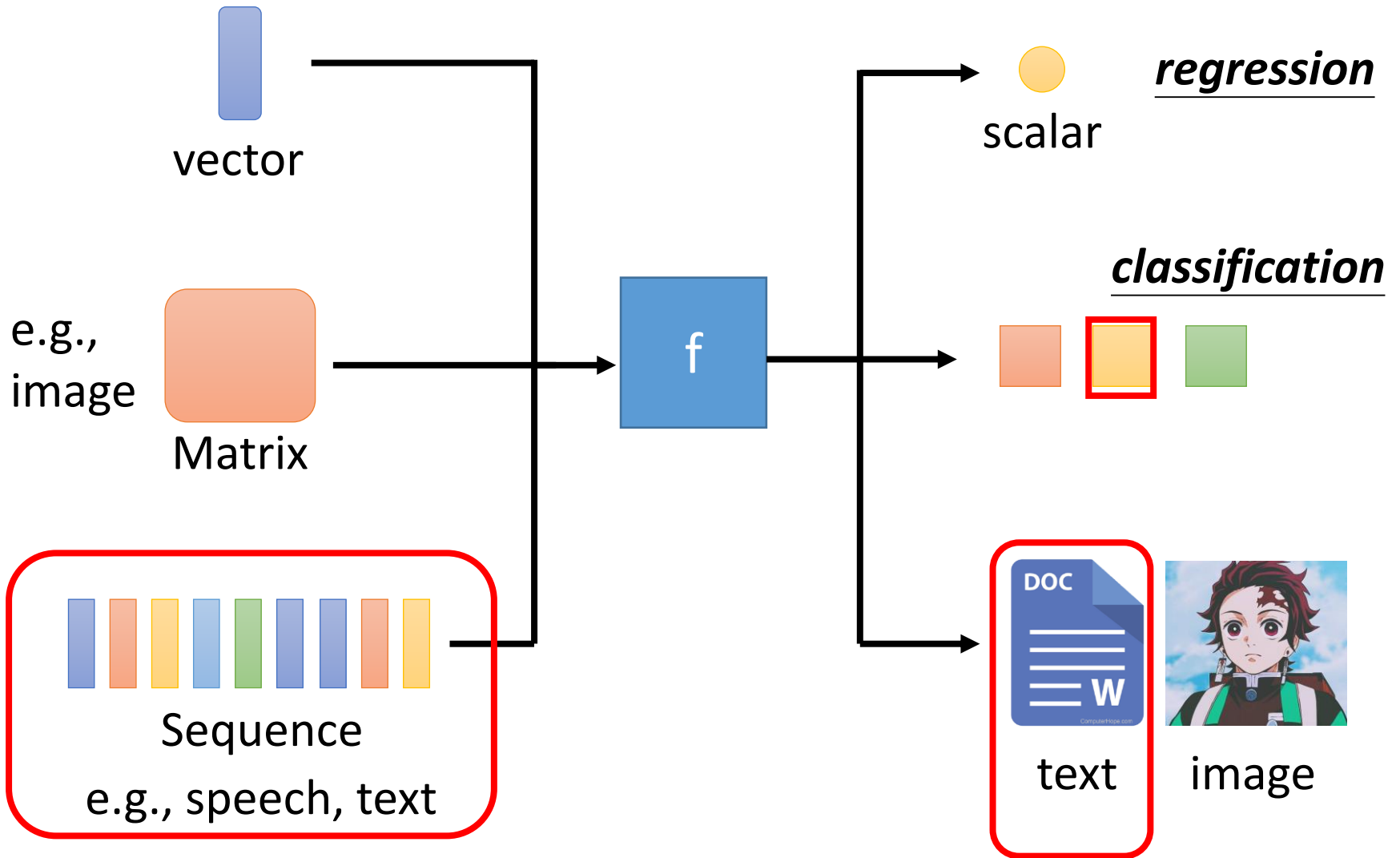
Image Classification



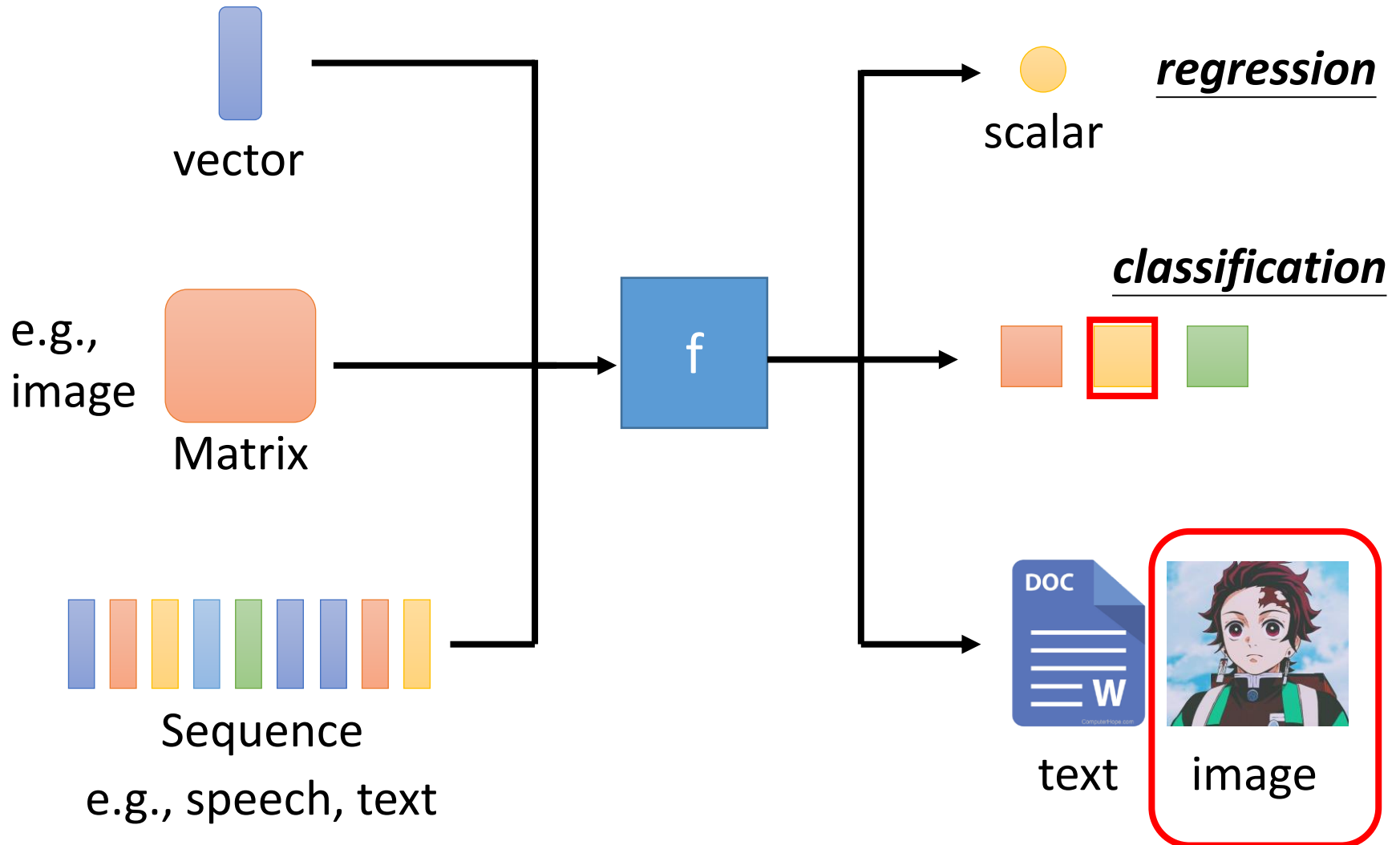
Speaker Classification



Machine Translation



Anime Face Generation



如何找到
想要的函数？

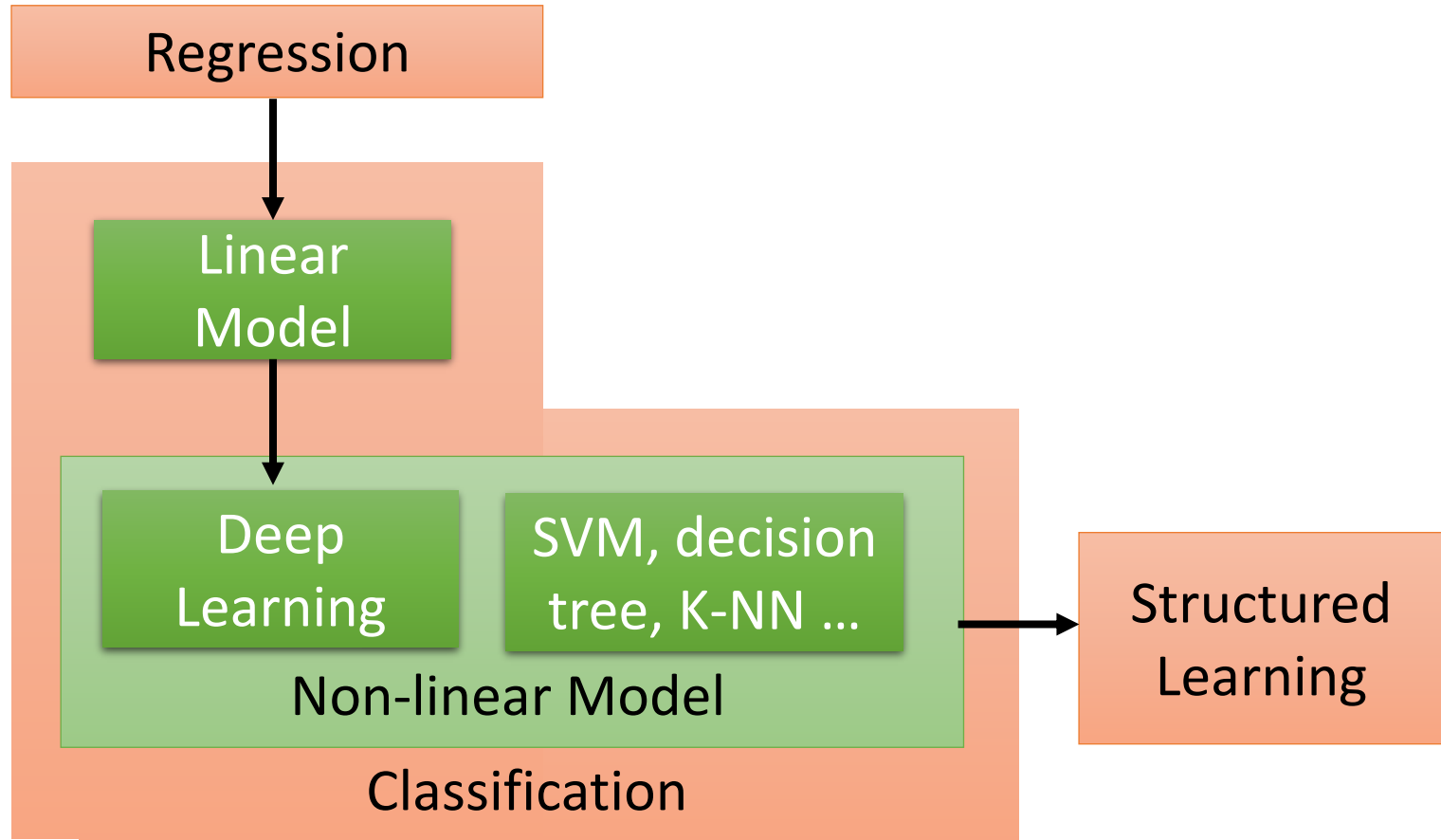


Step 1

划定一个函数集
(确定一个模型)



Learning Map



Step 2

机器如何判断找到的函数好不好？
(教会机器的几种方法)



Framework

Image Recognition:

$$f\left(\text{img}\right) = \text{"cat"}$$



$$f_1\left(\text{img}\right) = \text{"cat"}$$

$$f_2\left(\text{img}\right) = \text{"monkey"}$$

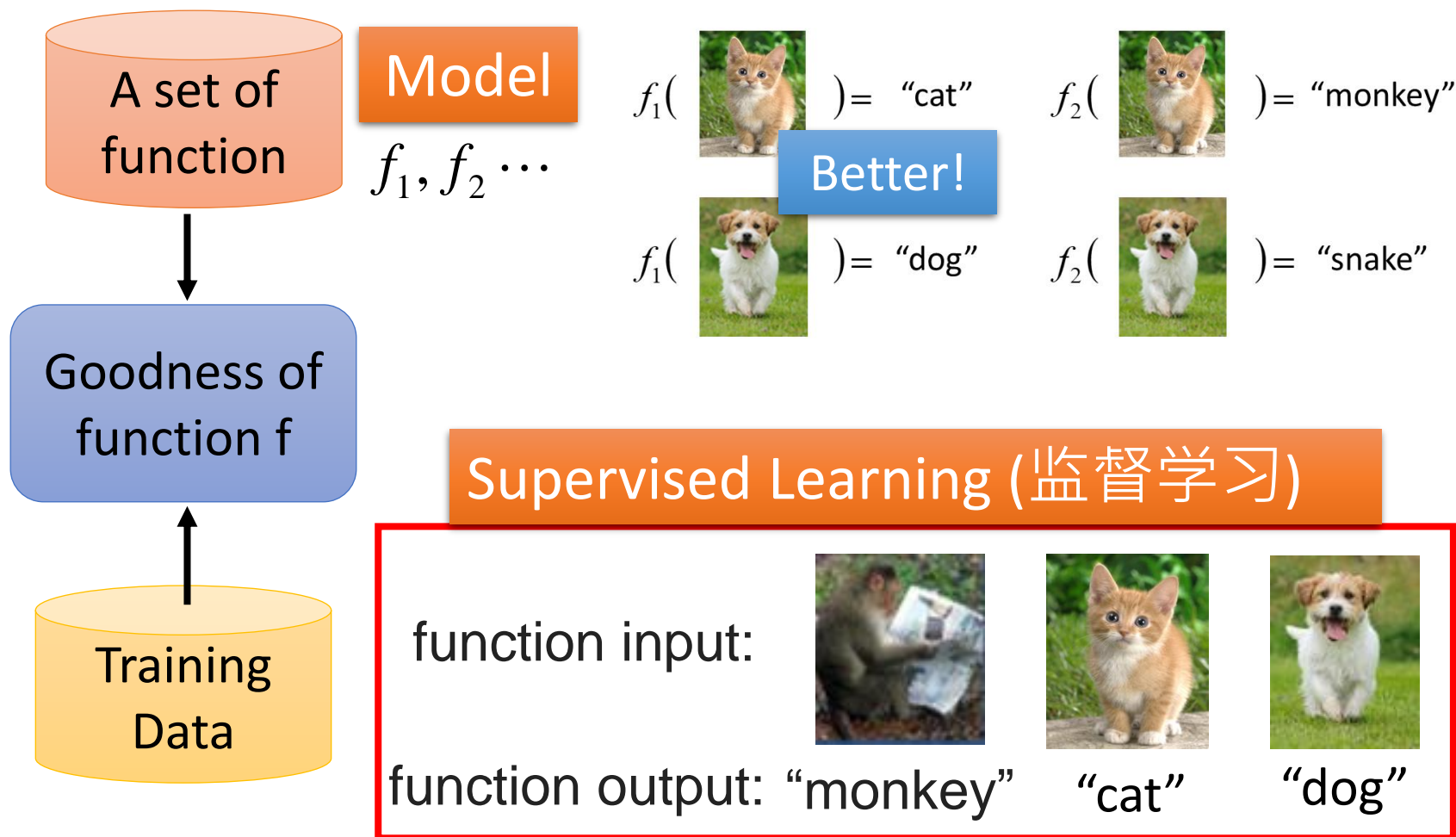
$$f_1\left(\text{img}\right) = \text{"dog"}$$

$$f_2\left(\text{img}\right) = \text{"snake"}$$

Image Recognition:

Framework

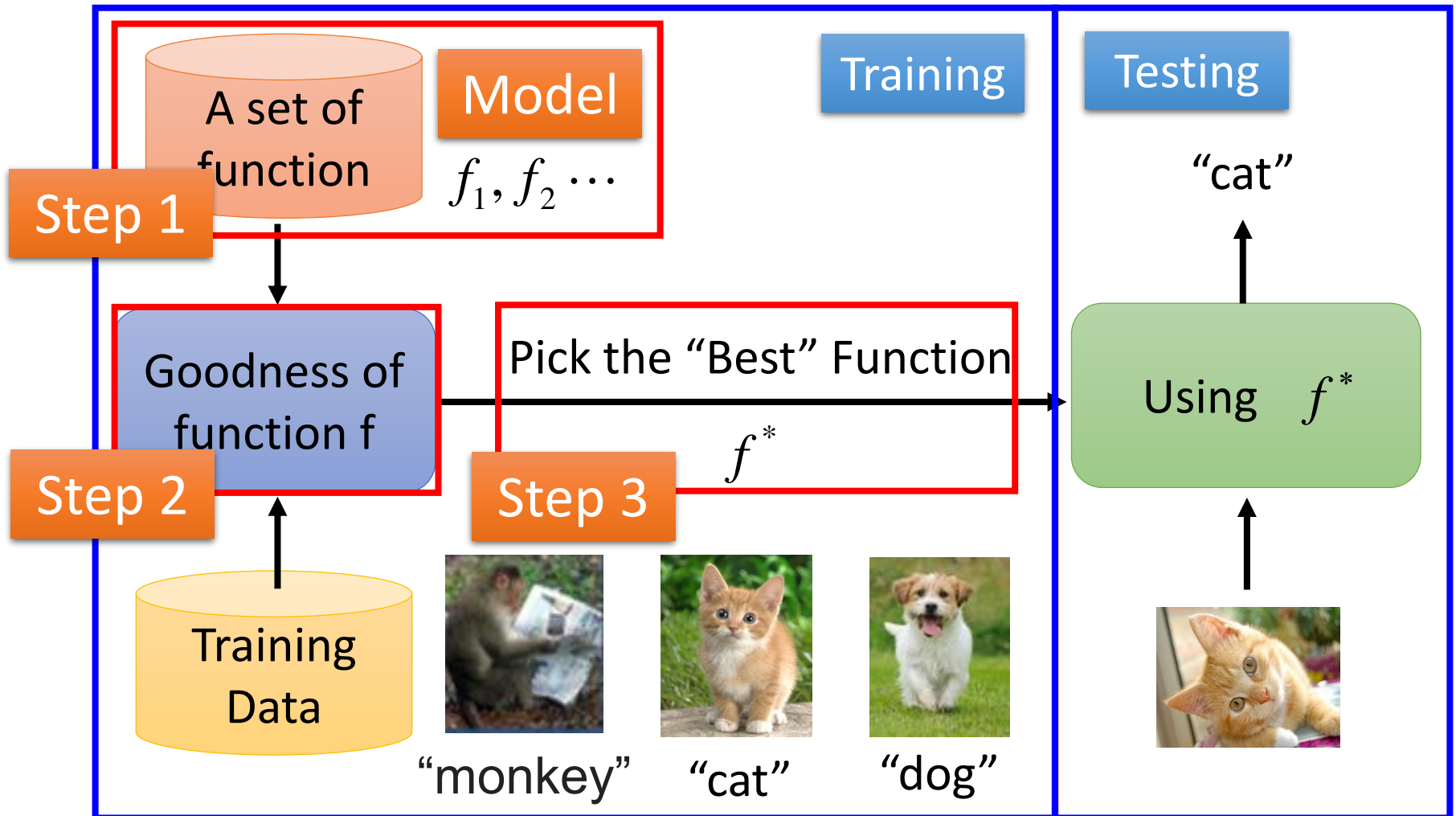
$$f(\text{img_cat}) = \text{"cat"}$$



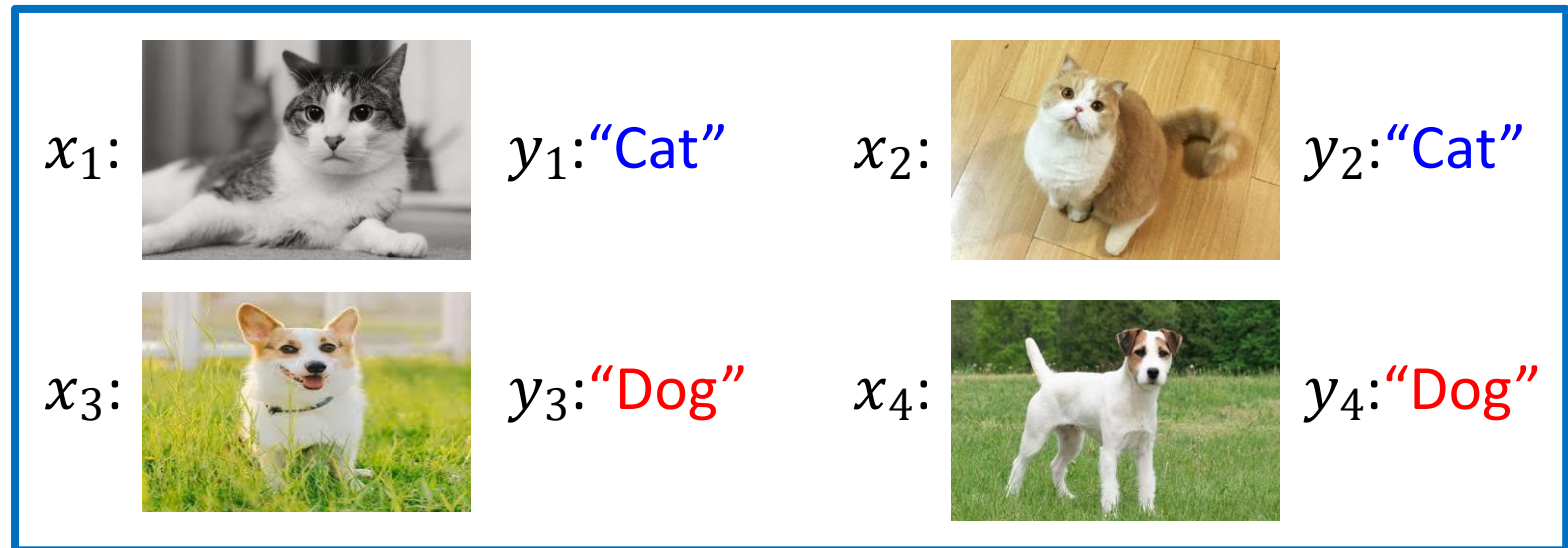
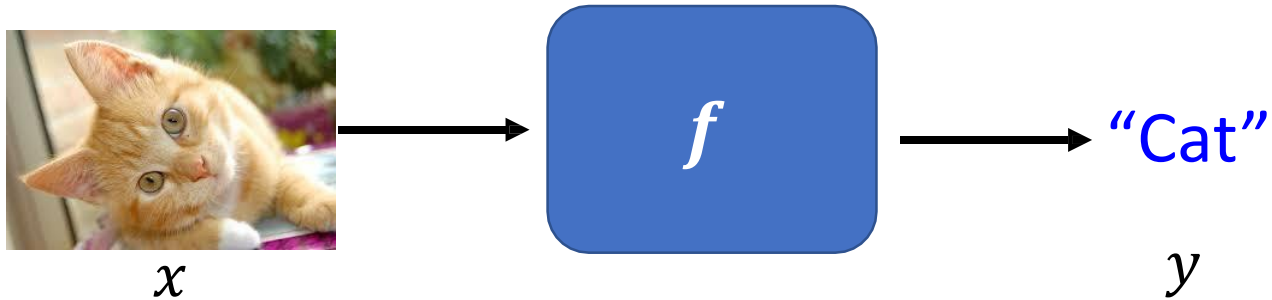
Framework

Image Recognition:

$$f\left(\text{Image of a cat}\right) = \text{"cat"}$$

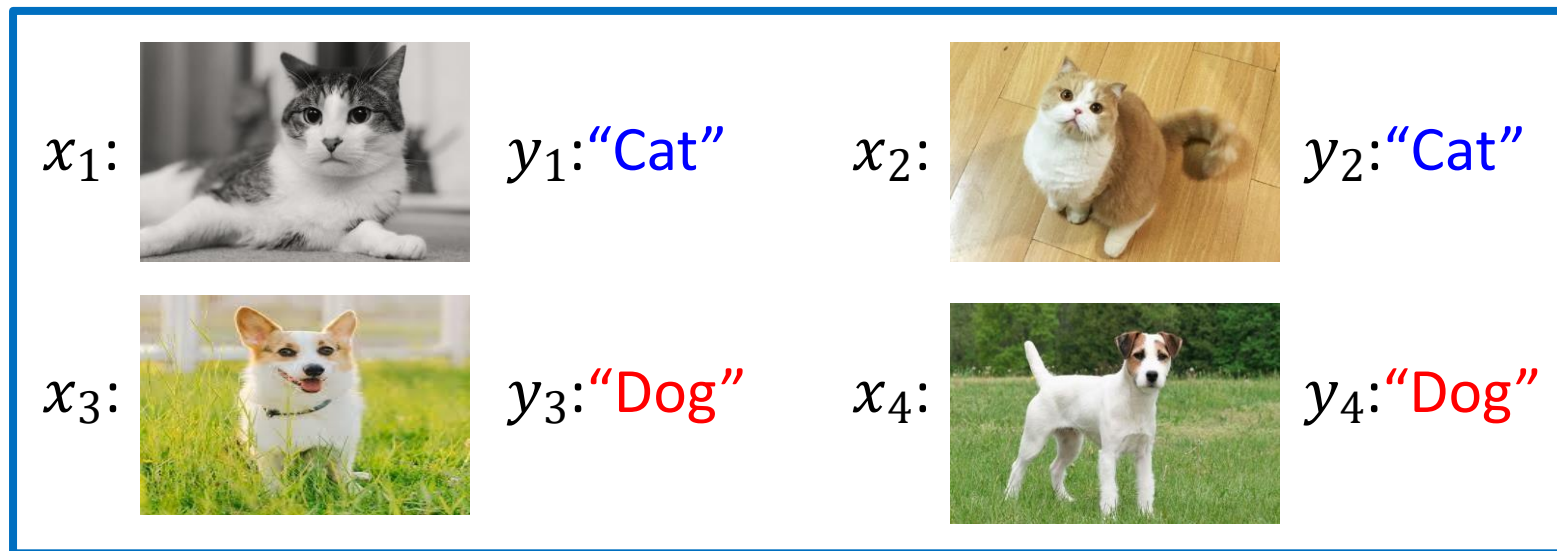
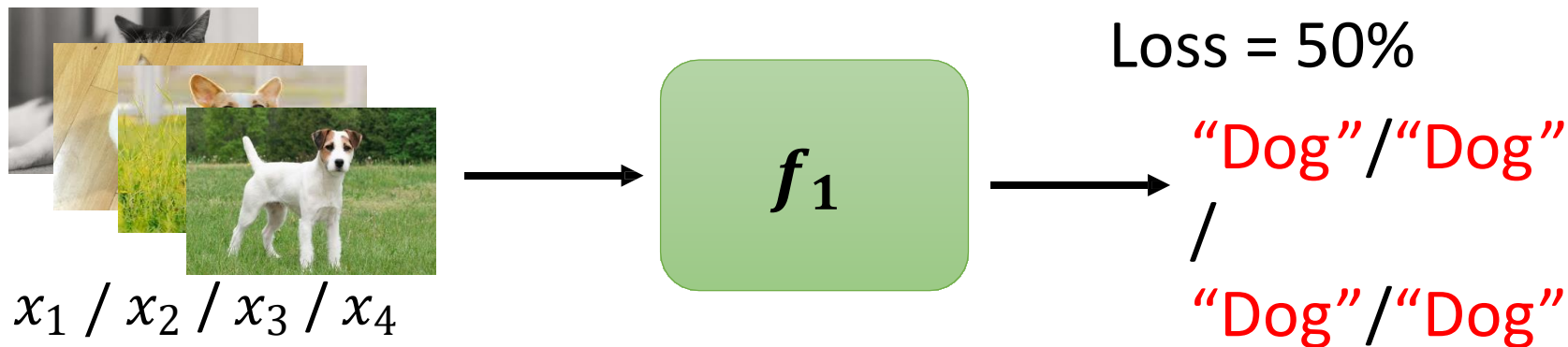


Supervised Learning



Labelled Data

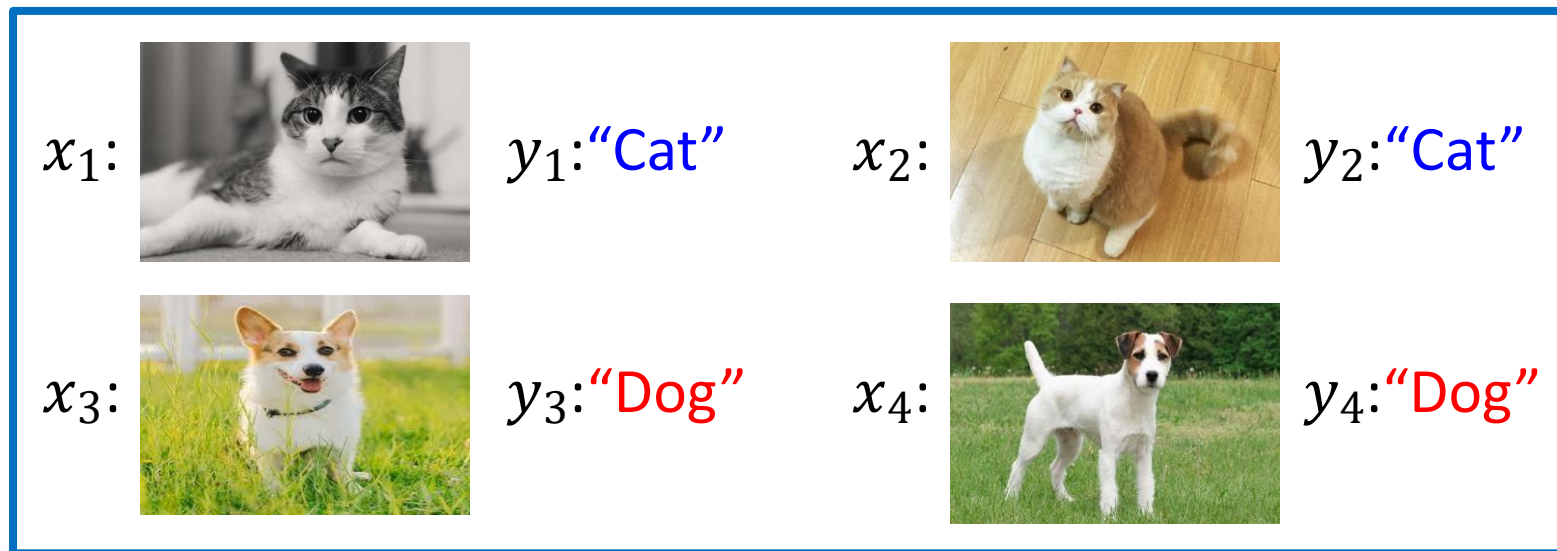
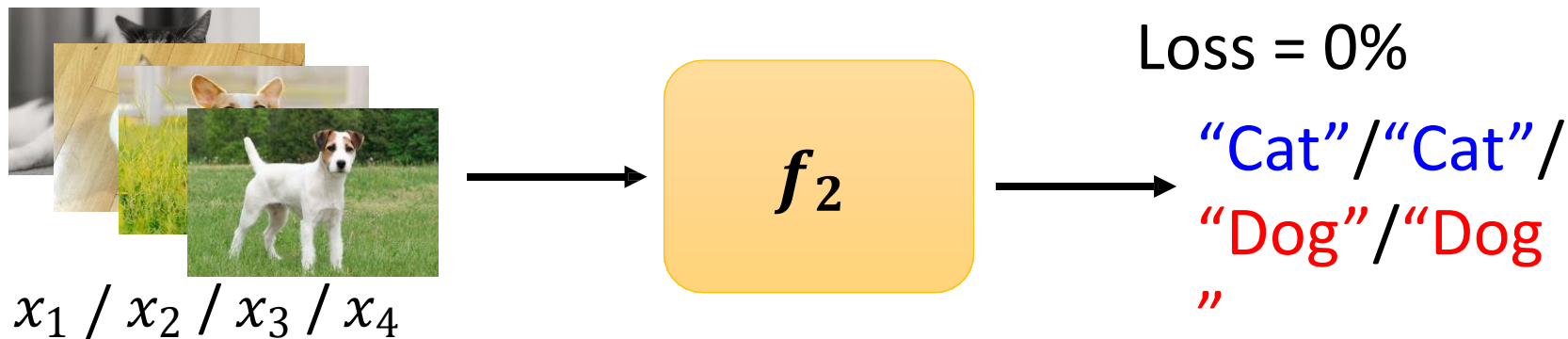
函数的 Loss



Labelled Data

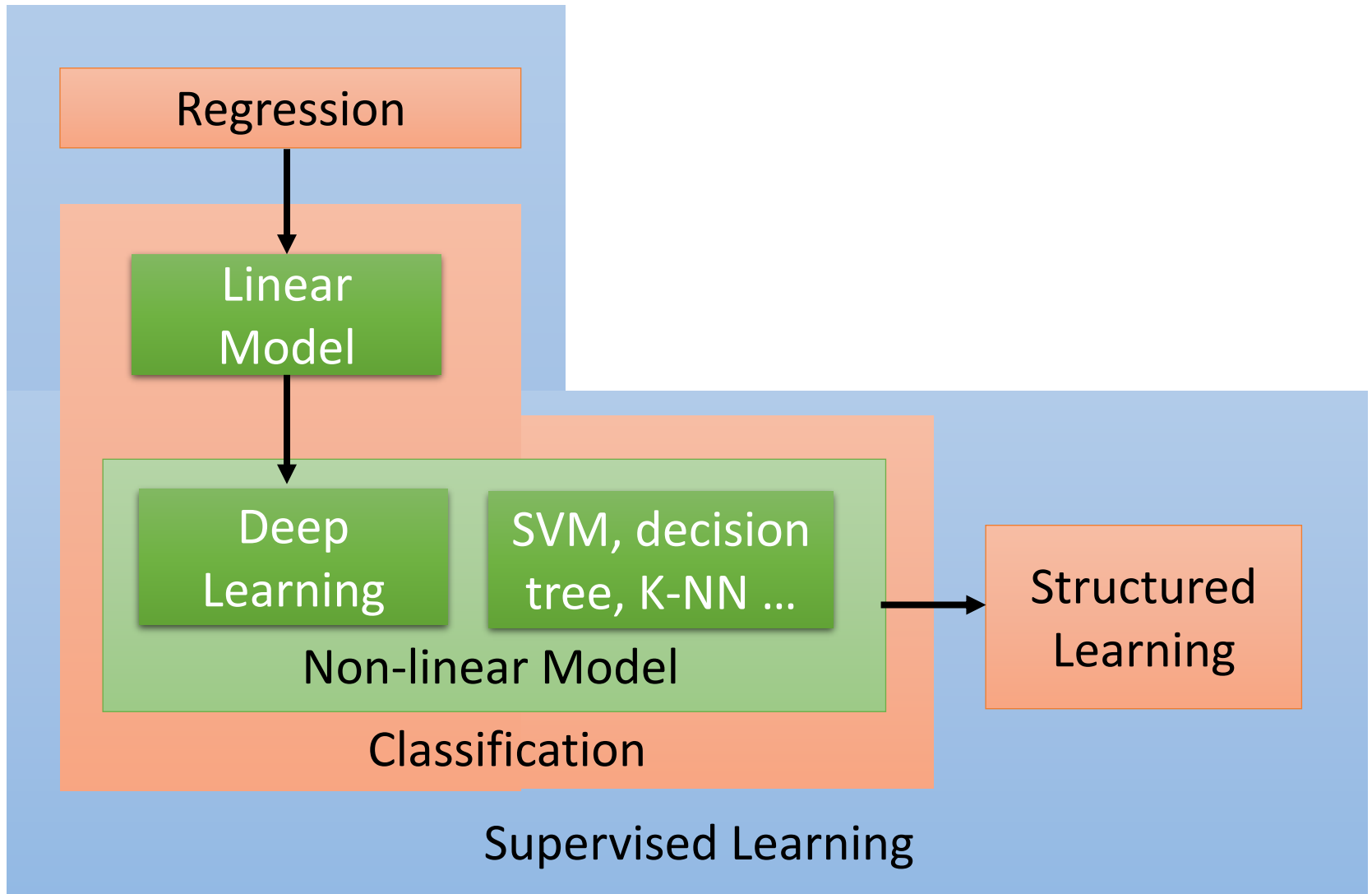
函数的 Loss

接下来机器会自动找出
Loss 最低的函数



Labeled Data

Learning Map



Learning Map

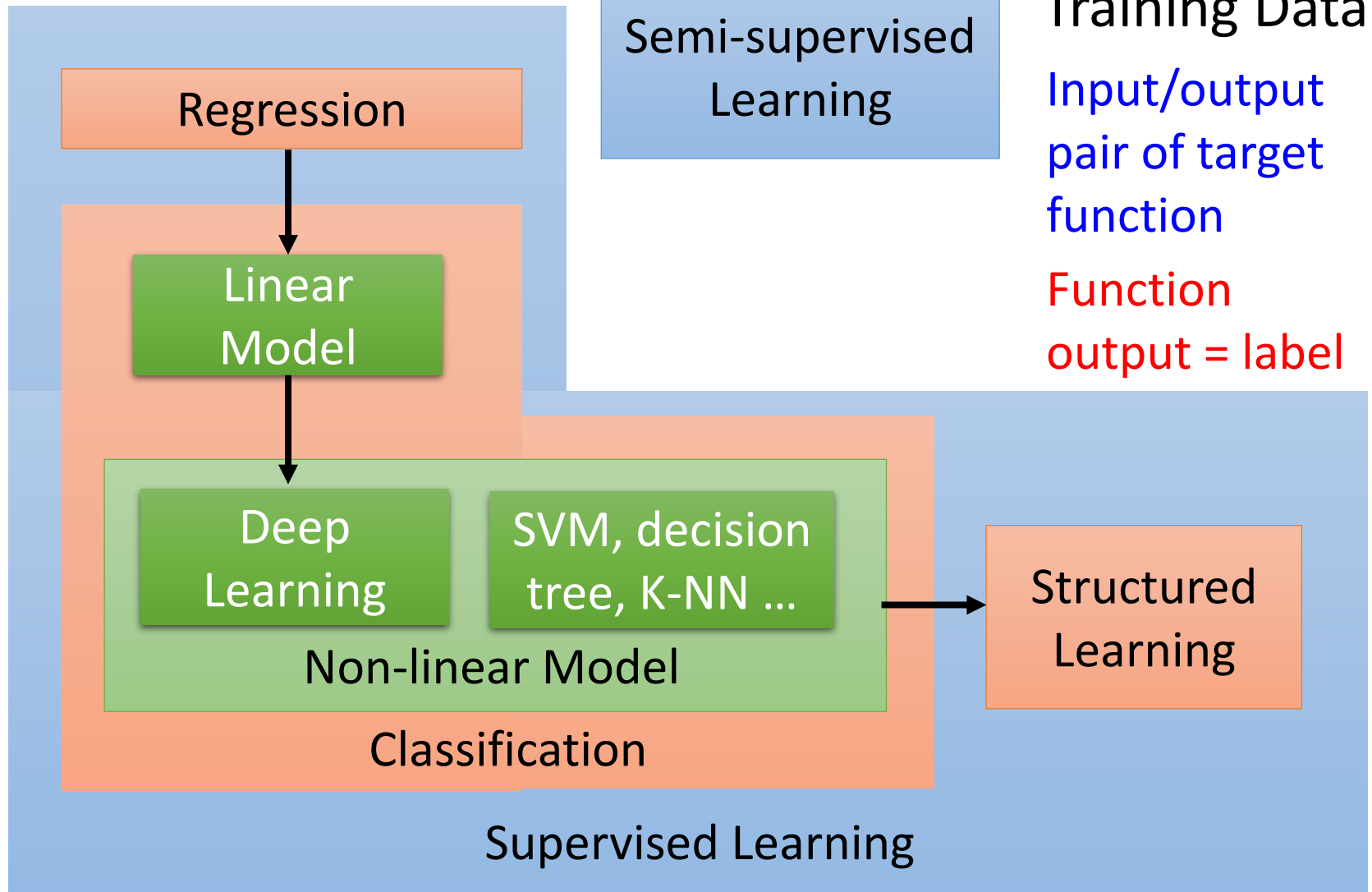
Hard to collect a large
amount of labelled data

Semi-supervised
Learning

Training Data:

Input/output
pair of target
function

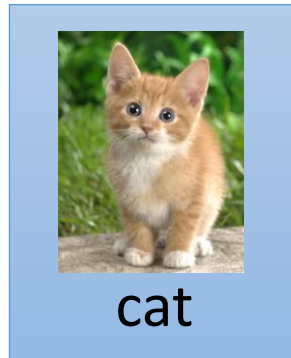
Function
output = label



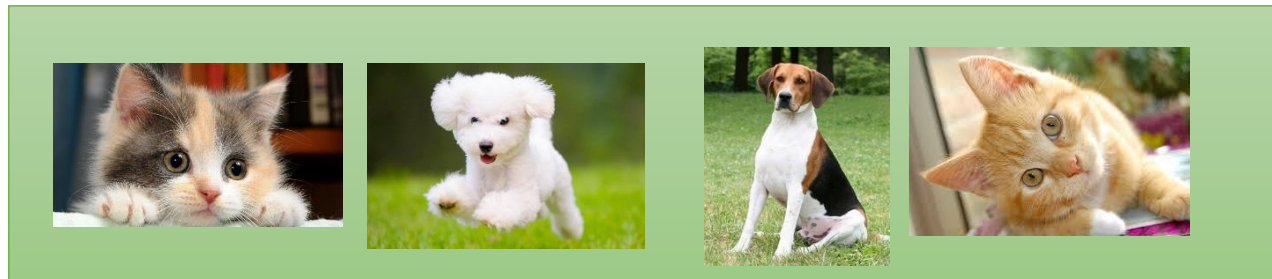
Semi-supervised (半监督学习)

For example, recognizing cats and dogs

Labelled
data

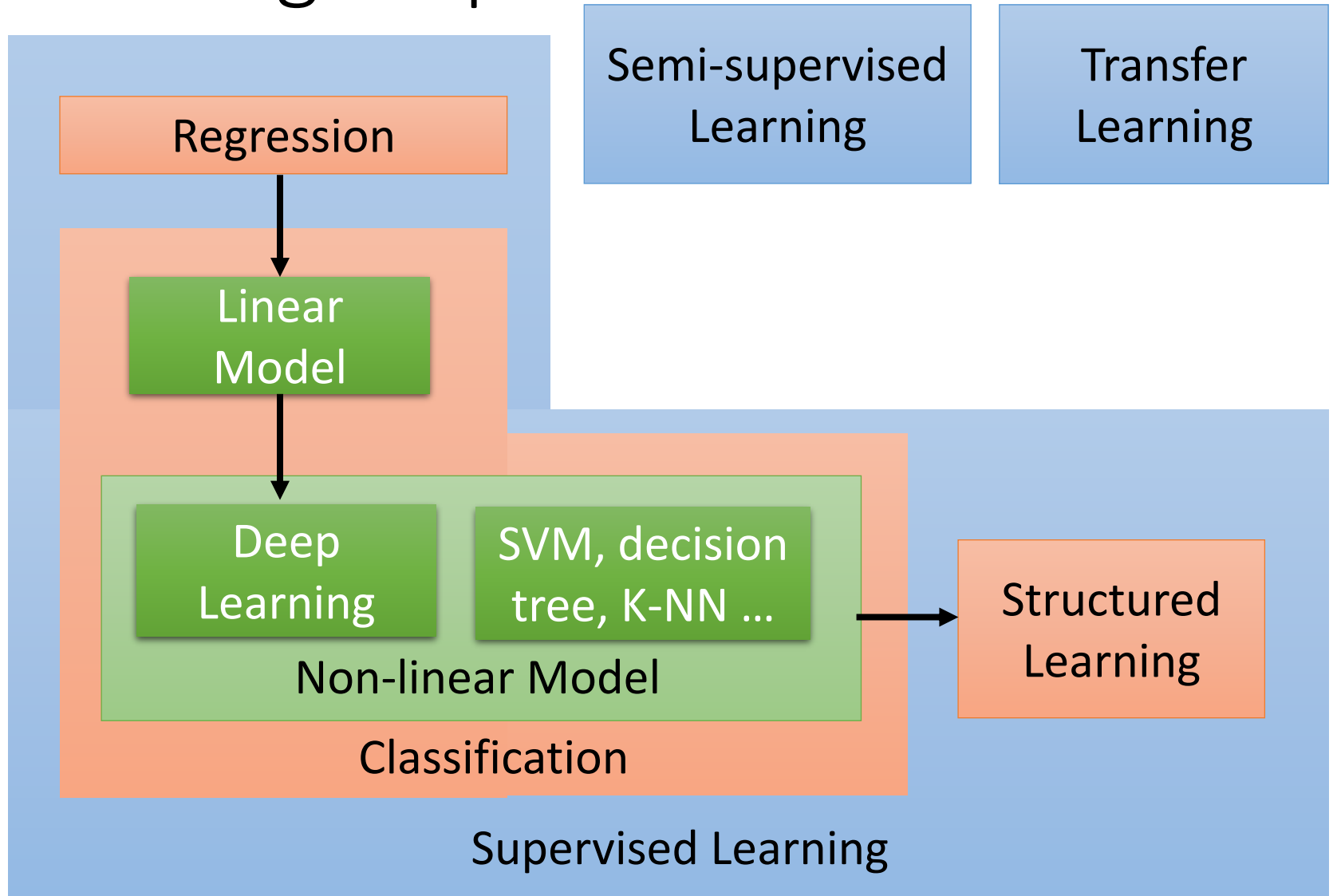


Unlabeled
data



(Images of cats and dogs)

Learning Map



Transfer Learning (迁移学习)

For example, recognizing cats and dogs

Labelled
data



cat



dog



elephant

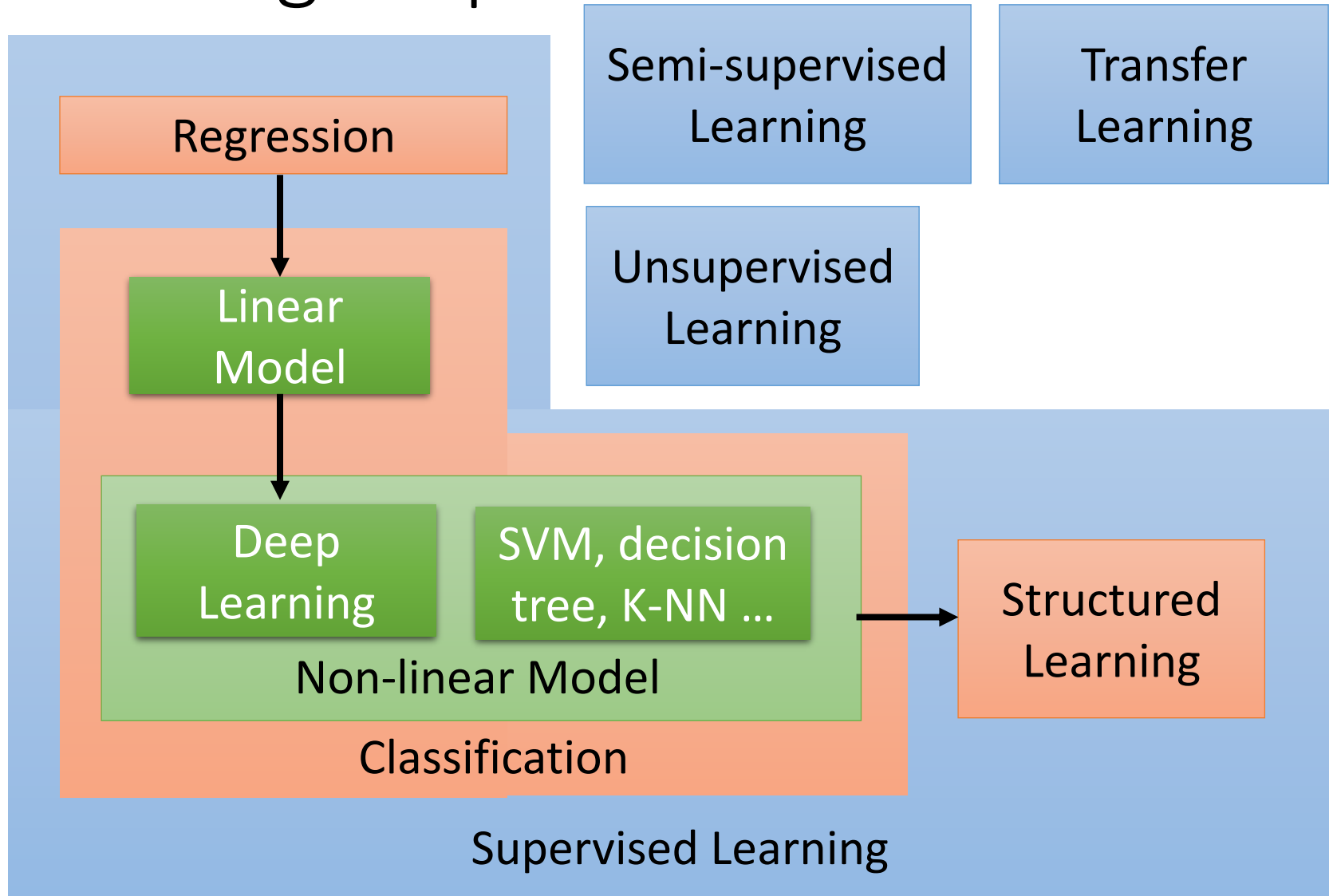


Naruto



Data not related to the task considered
(can be either labeled or unlabeled)

Learning Map



Unsupervised (无监督学习)

Training AI *without paired data*

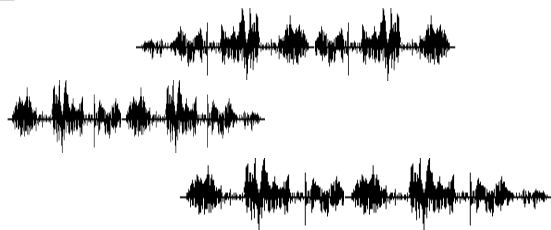


Supervised



Unsupervised

AI listening in
the environment



Audio x

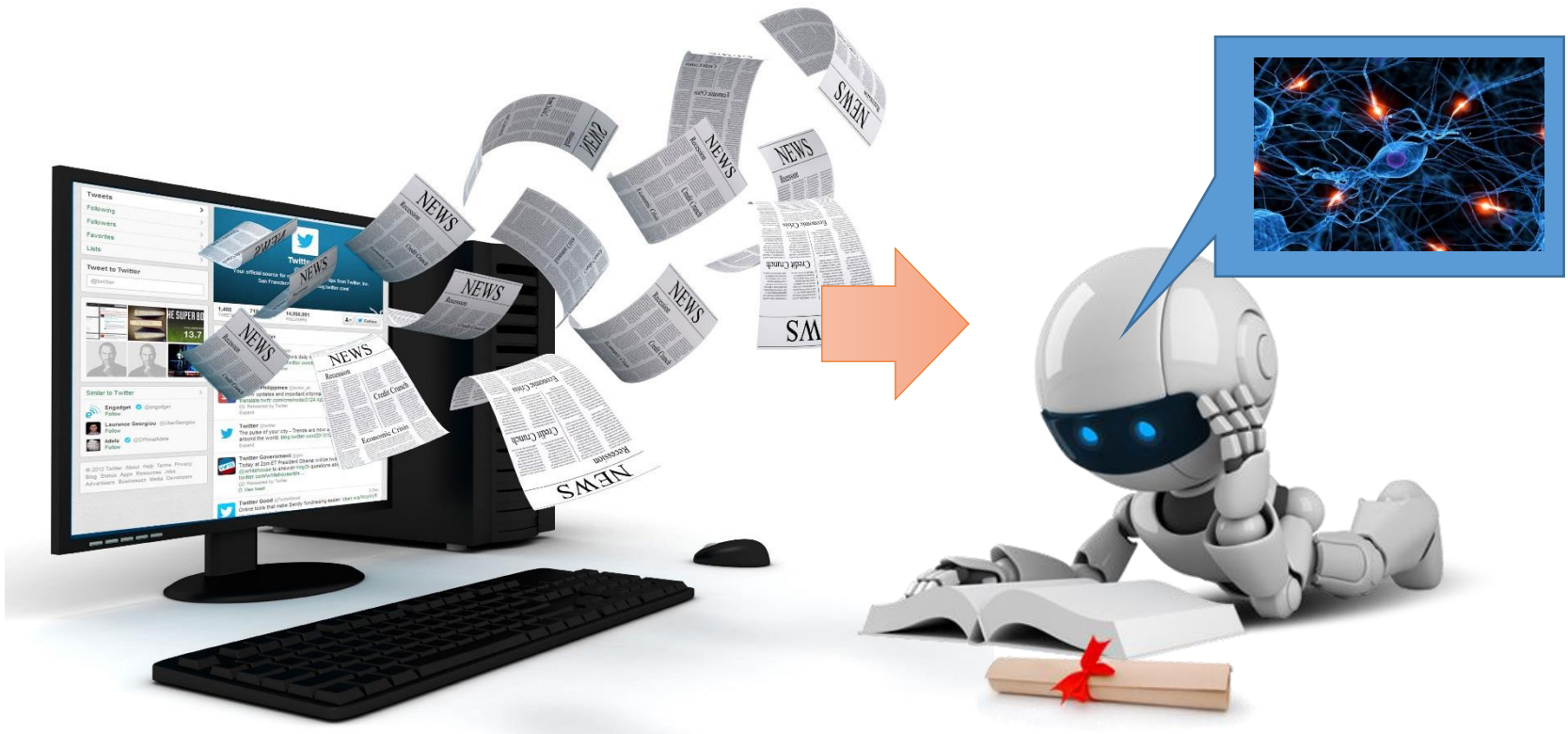


Text y

AI reading
documents on
the Internet

Unsupervised (无监督学习)

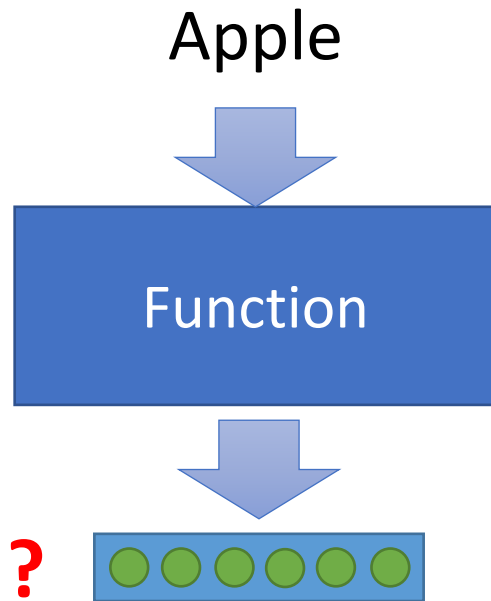
- Machine Reading: Machine learns the meaning of words from reading a lot of documents



<http://top-breaking-news.com/>

Unsupervised (无监督学习)

- Machine Reading: Machine learns the meaning of words from reading a lot of documents



Training data is a lot of text



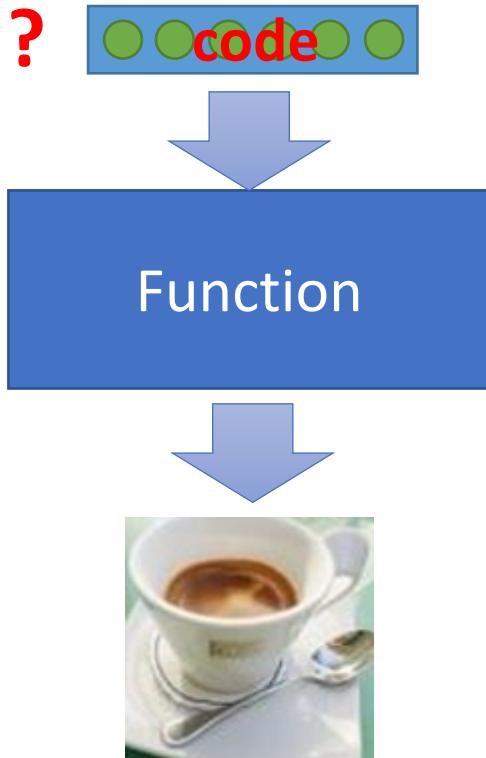
<https://garavato.files.wordpress.com/2011/11/stacksdocuments.jpg?w=490>

Unsupervised Learning

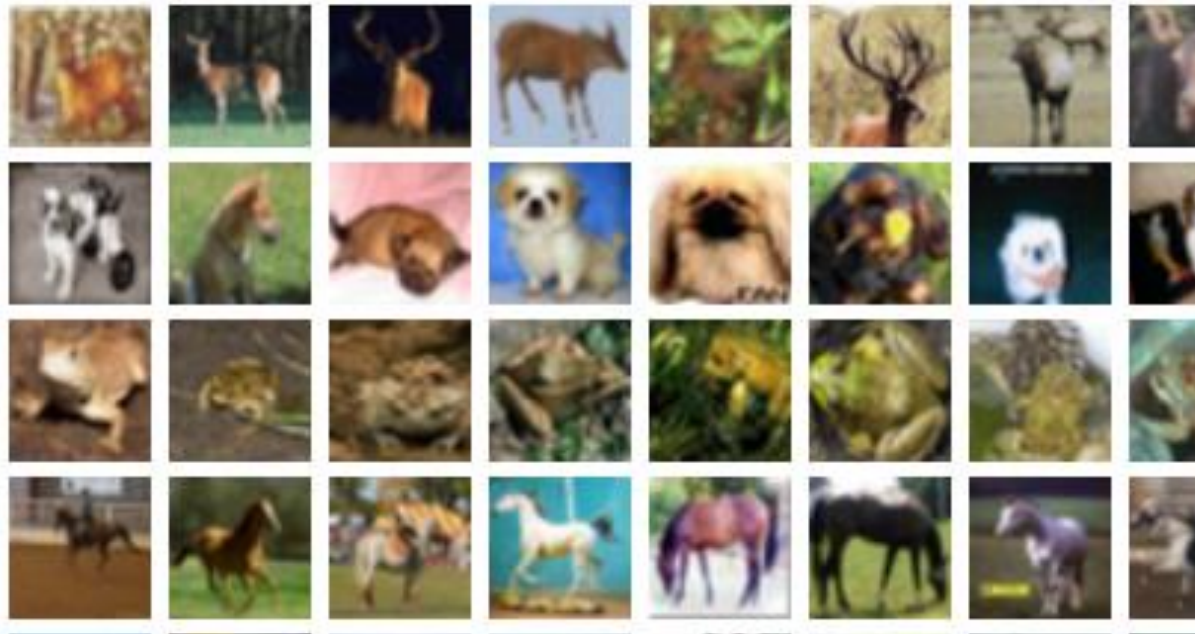


Unsupervised Learning

- Machine Drawing



Training data is a lot of images

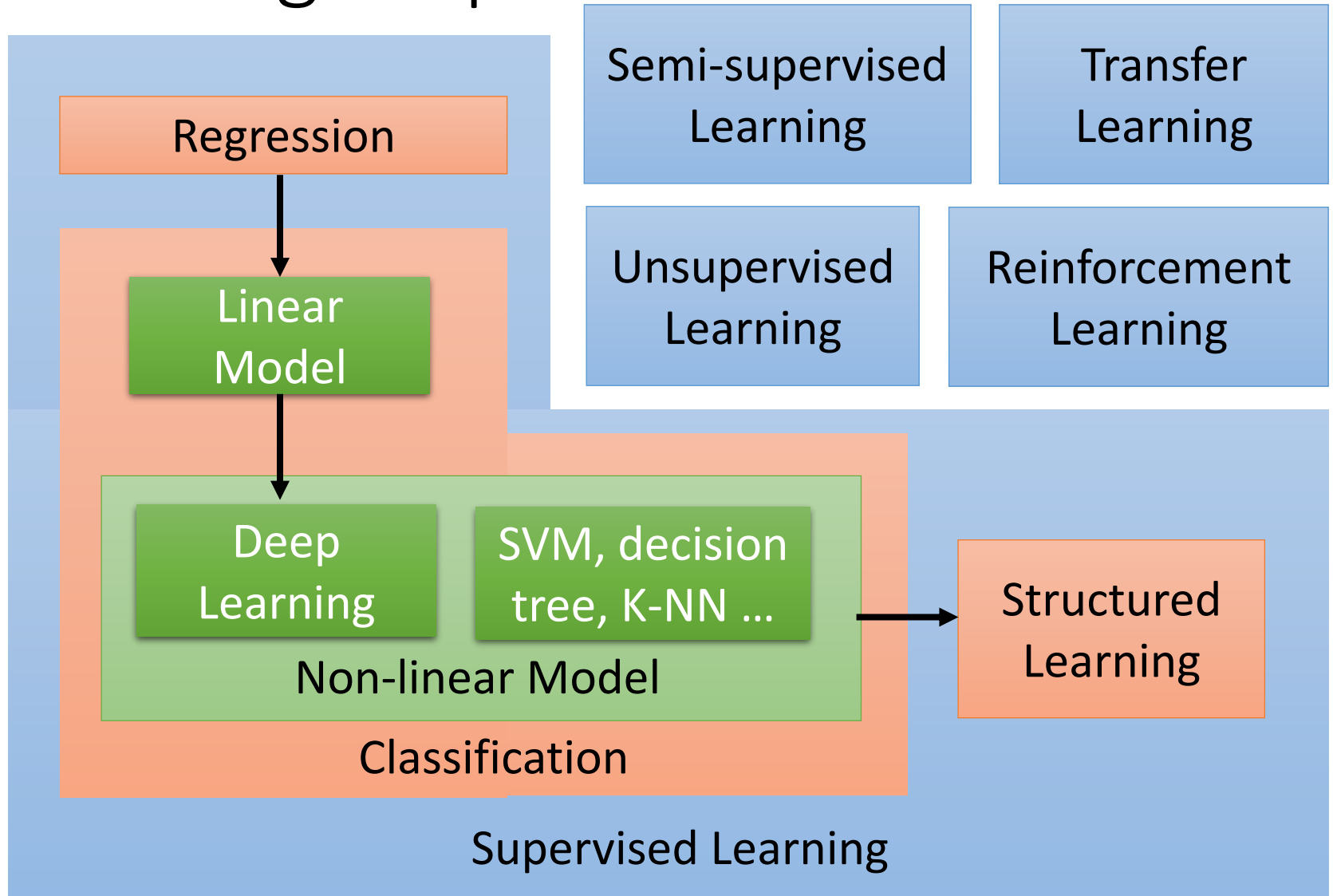


Unsupervised (无监督学习)

- Machine Reading: Machine learns the meaning of words from reading a lot of documents
- ELMO/BERT



Learning Map



Reinforcement Learning (强化学习)



Deep Reinforcement Learning: $AI = RL + DL$

Supervised v.s. Reinforcement

- Supervised

Learning from
teacher



"Hello"

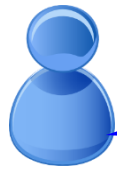
Say "Hi"



"Bye bye"

Say "Good bye"

- Reinforcement



.....



.....

.....



Bad

Learning from
critics

Hello 😊

Agent

.....

Agent

Supervised v.s. Reinforcement

- Supervised:

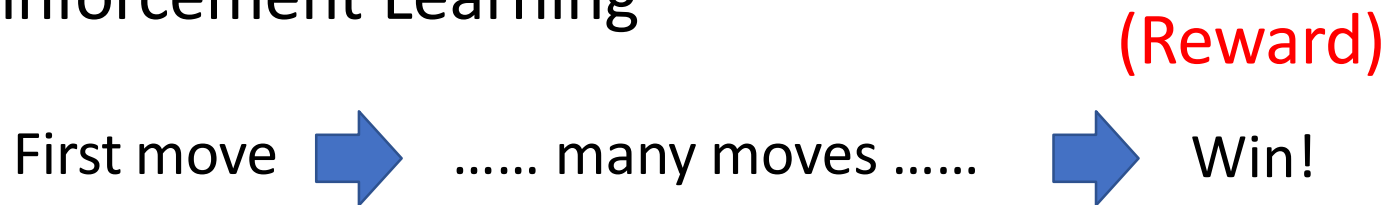


Next move:
"5-5"



Next move:
"3-3"

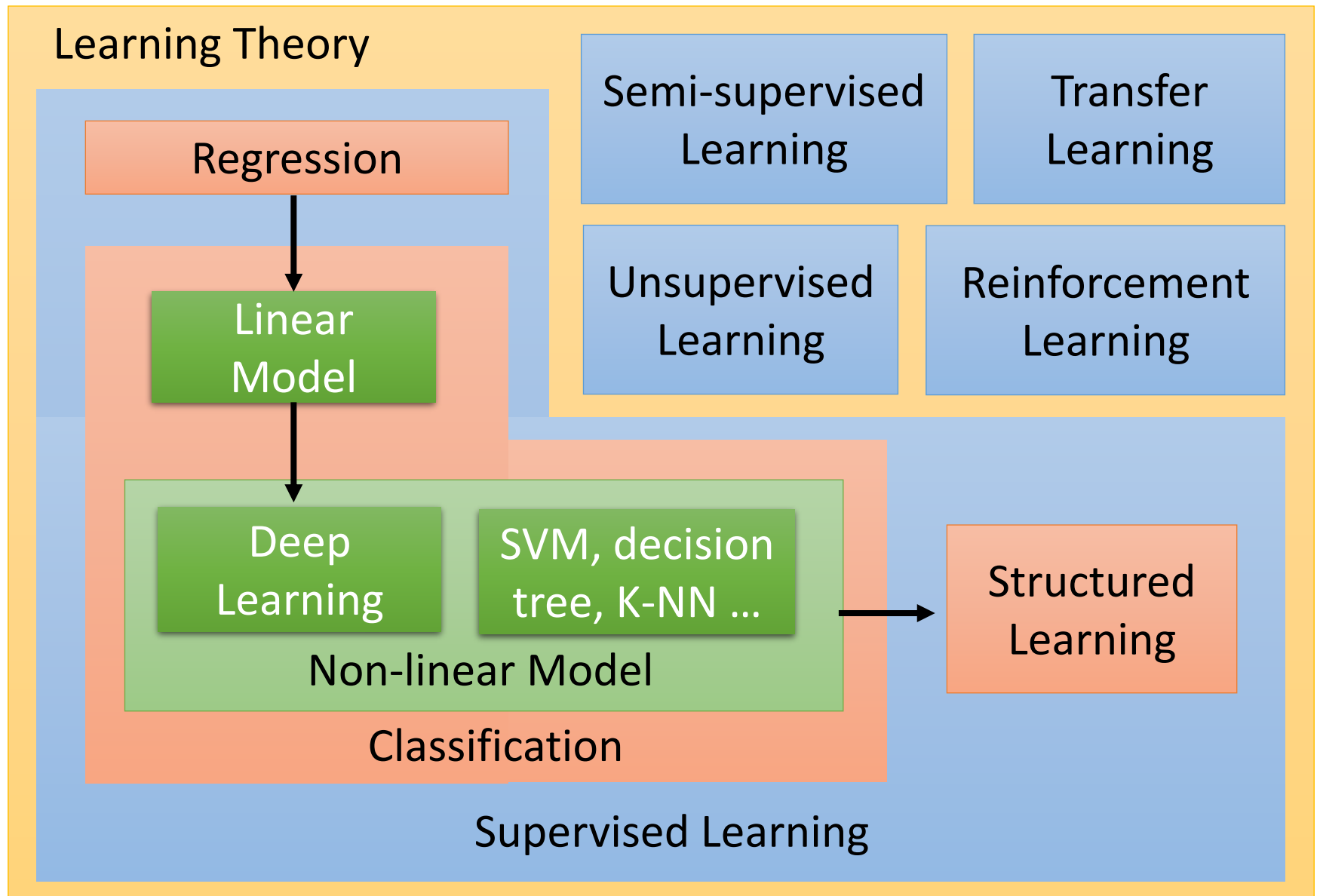
- Reinforcement Learning



Alpha Go is supervised learning + reinforcement learning.

Learning Map

scenario task method



Step 3

如何找到 最好的函数？



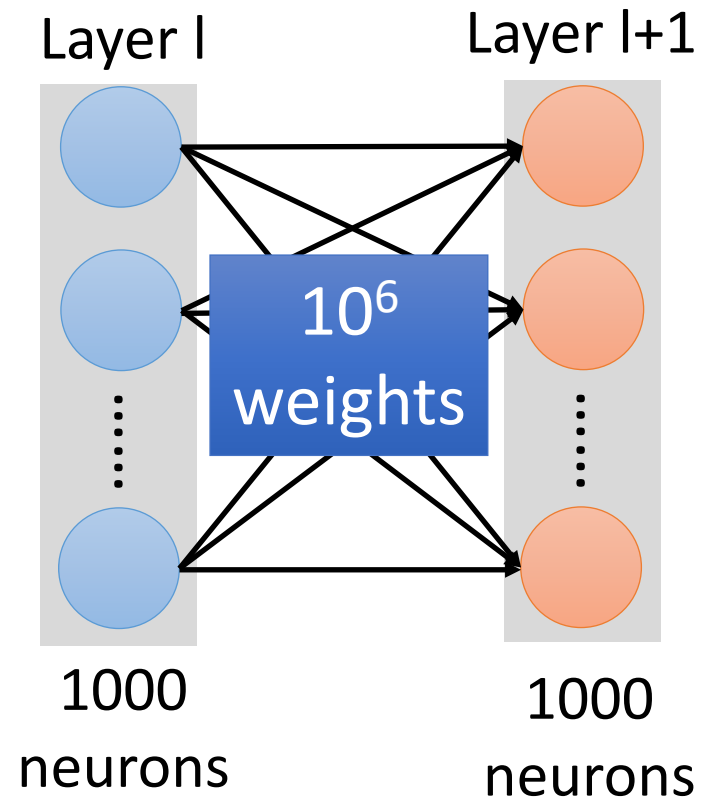
如何找出最好的 function ?

Enumerate all possible values

Network parameters $\theta =$
 $\{w_1, w_2, w_3, \dots, b_1, b_2, b_3, \dots\}$

Millions of parameters

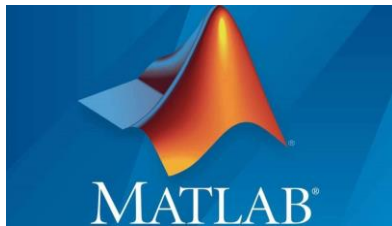
Today a network can have
more than 100M parameters.



Gradient Descent



PYTORCH



theano

Caffe



Deep Learning library produced by Amazon

DSSTNE



机器学习很简单

Step 0: What kind of function do you want to find?

Step 1:
define a set
of function

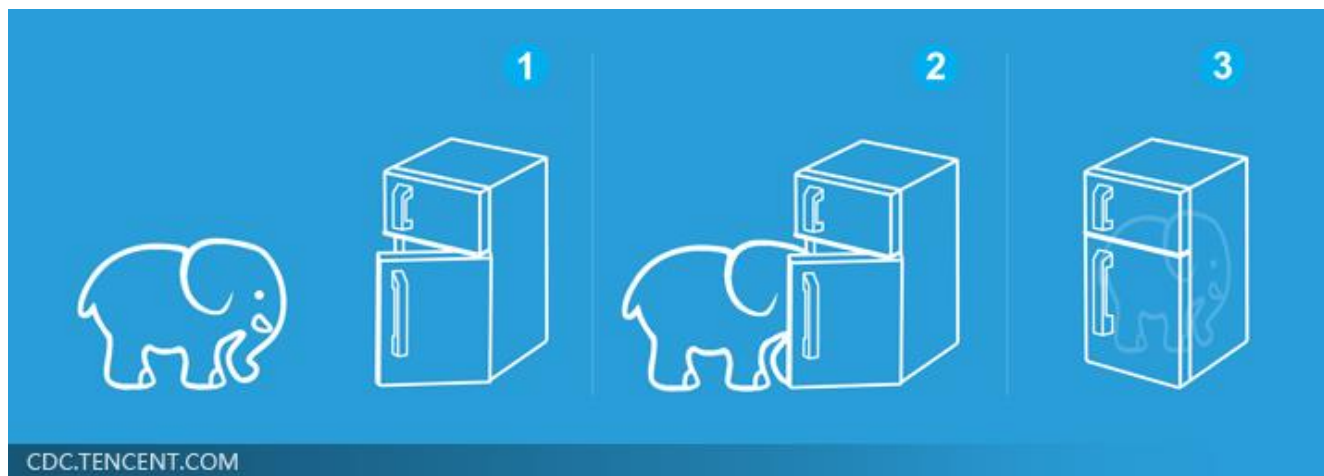


Step 2:
goodness of
function



Step 3: pick
the best
function

就好像把大象装进冰箱



机器学习很简单

Step 0: What kind of function do you want to find?

Regression, Classification, Generation

Step 1:
define a set
of function

Deep Learning
SVM
Decision Tree
.....



Step 2:
goodness of
function

Supervised
Transfer
Reinforcement
.....



Step 3: pick
the best
function

Gradient Descent
.....