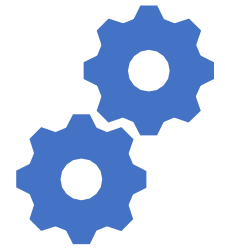


Machine Learning 2020

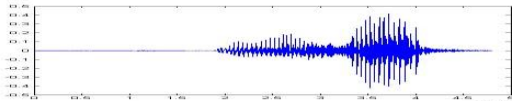


李宏毅

Hung-yi Lee

機器學習就是自動找函式


- Speech Recognition

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

- Image Recognition

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

- Playing Go

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

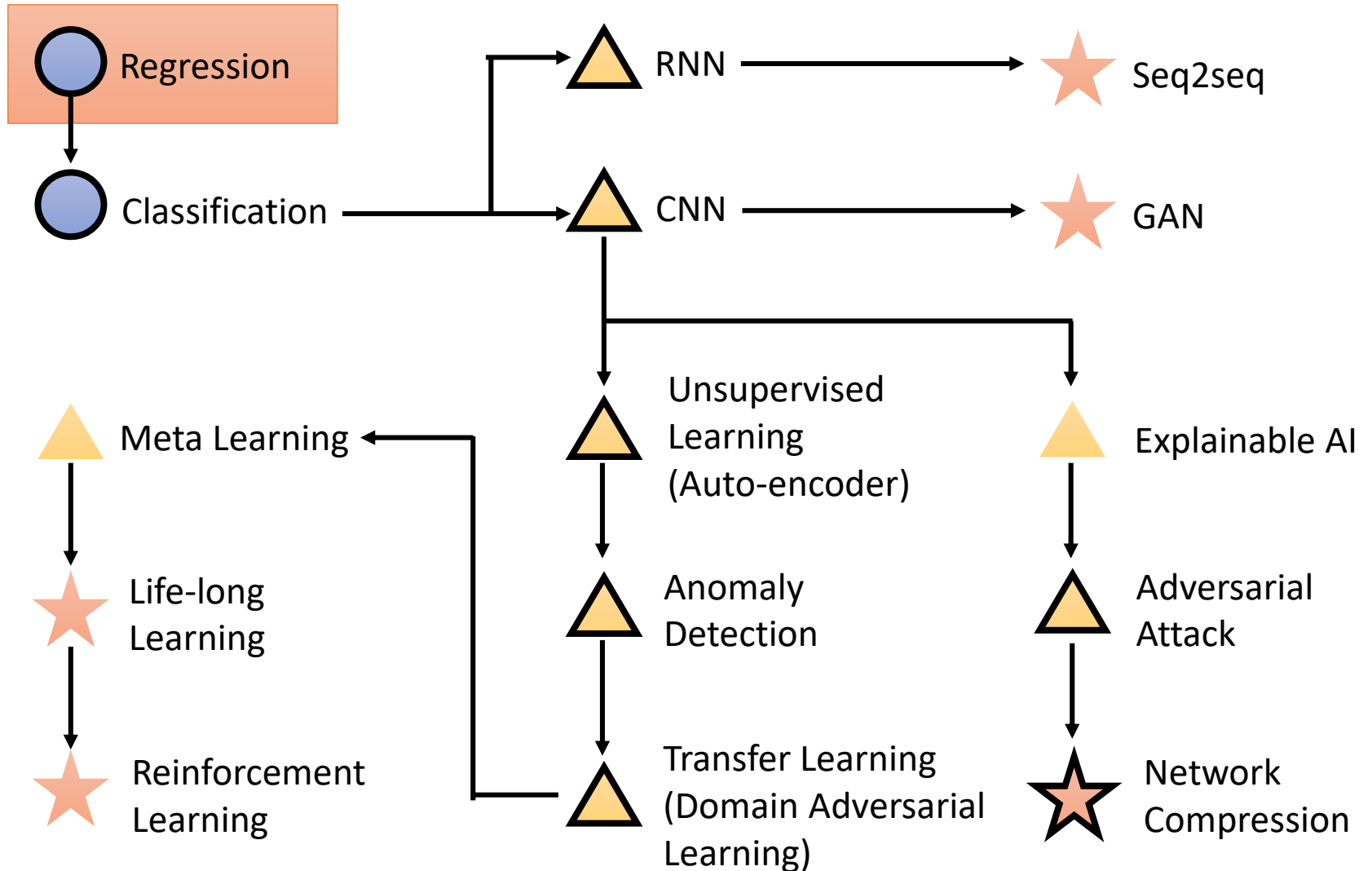
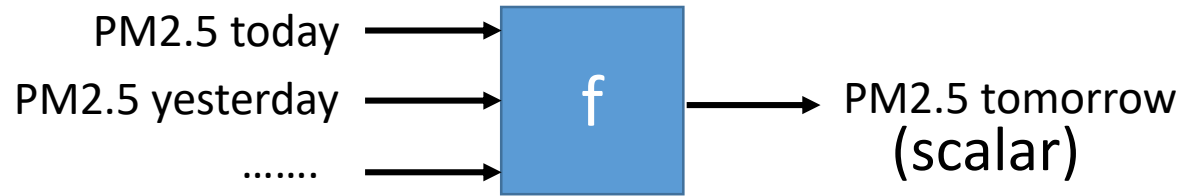
- Dialogue System

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

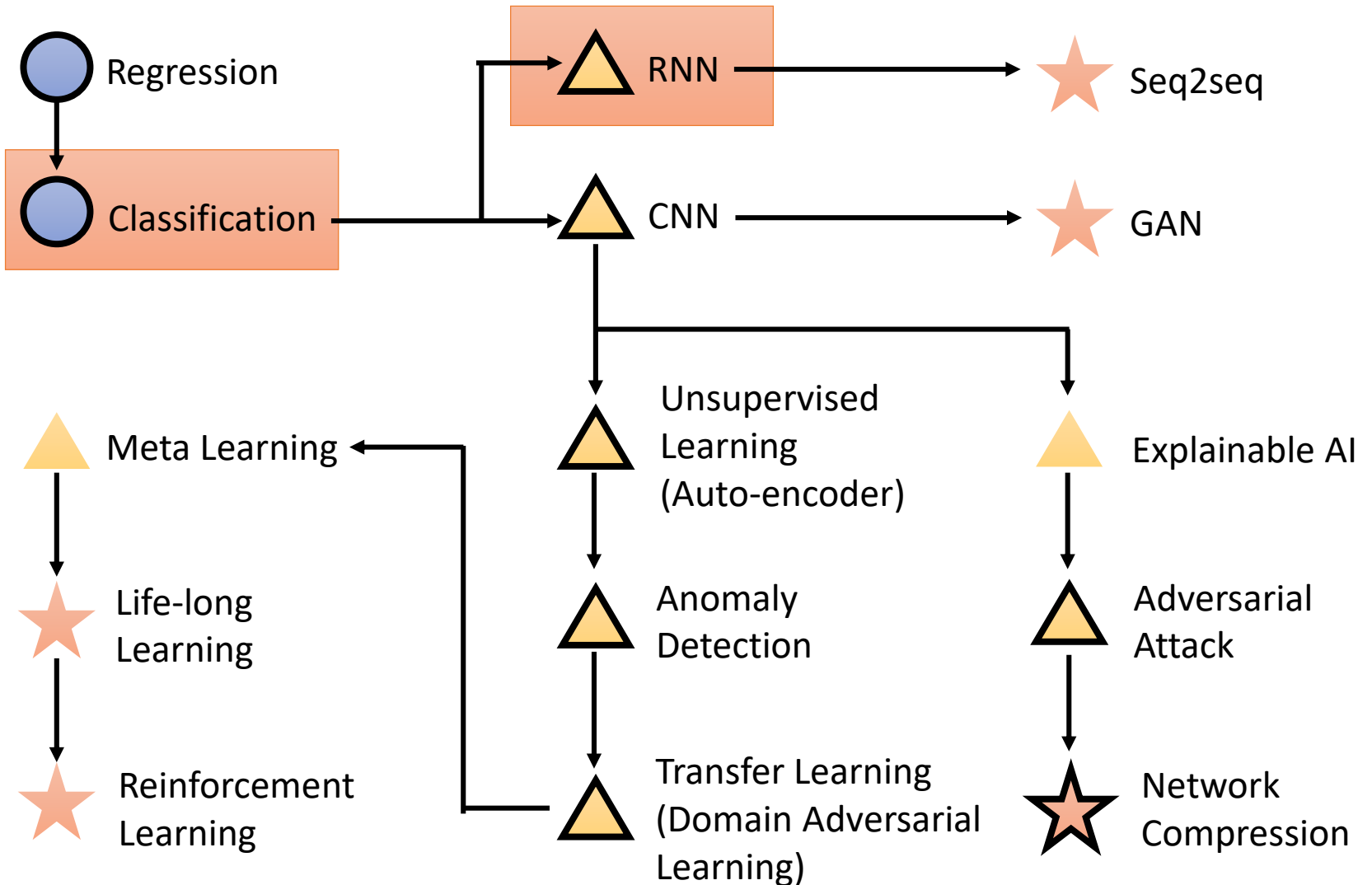
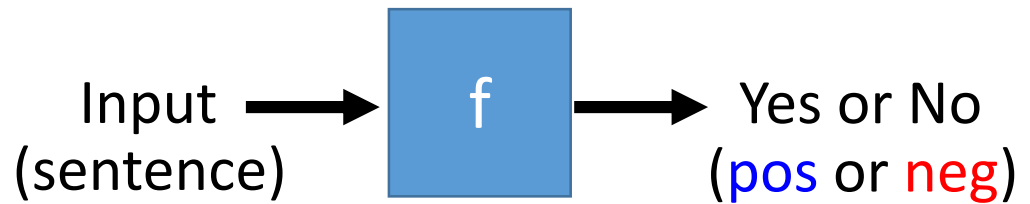
你想找什麼樣的函式？

Regression

The output of the function is a scalar.



Binary Classification



Multi-class Classification



 Regression

 Classification



RNN



Seq2seq



CNN



GAN



Meta Learning



Life-long
Learning



Reinforcement
Learning



Unsupervised
Learning
(Auto-encoder)



Anomaly
Detection



Transfer Learning
(Domain Adversarial
Learning)



Explainable AI



Adversarial
Attack



Network
Compression

Generation 的問題也是機器學習一大領域 => 讓機器具有創造性

Generation (生成)

產生有結構的複雜東西
(例如：文句、圖片)

擬人化的講法——創造

Regression,
Classification



機器學習不只有 Regression & Classification

Generation

翻譯：產生文句

Seq2seq

GAN

產生二次元
人物

Explainable AI

Adversarial
Attack

Network
Compression

RNN

CNN

Unsupervised
Learning
(Auto-encoder)

Anomaly
Detection

Transfer Learning
(Domain Adversarial
Learning)

Regression

Classification

Meta Learning

Life-long
Learning

Reinforcement
Learning

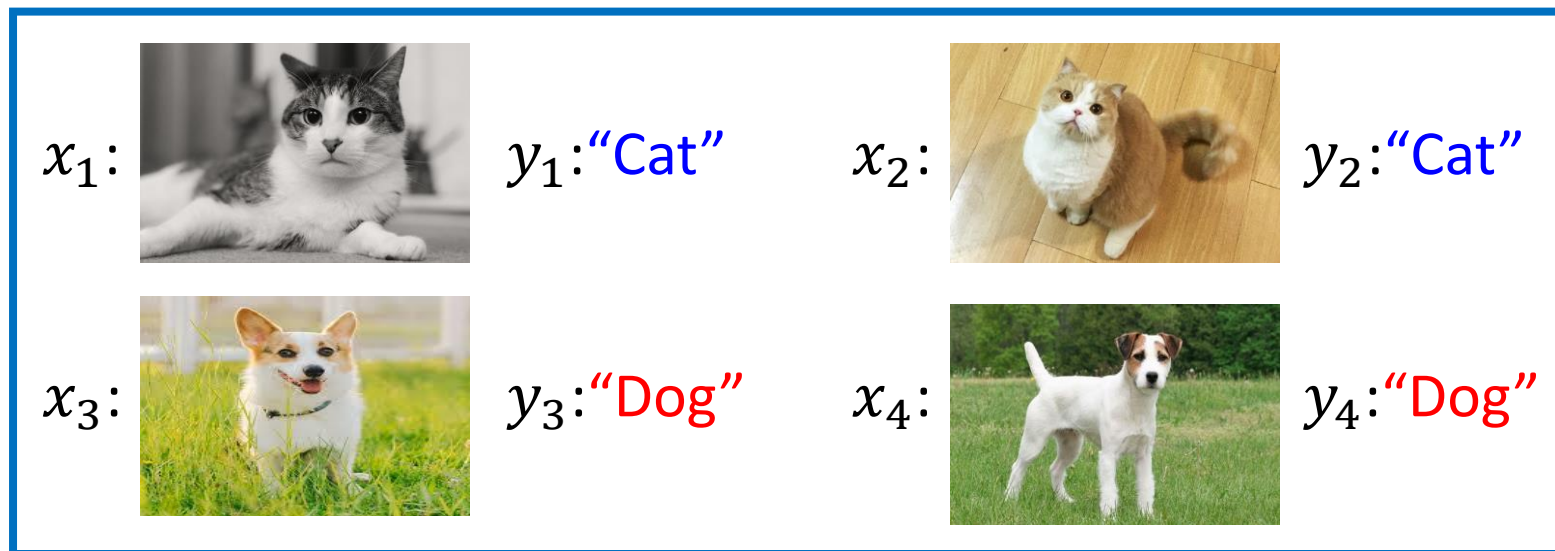
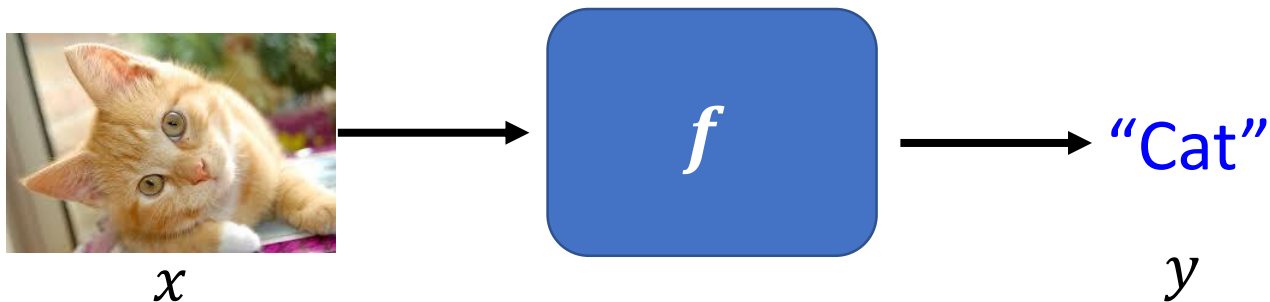


怎麼告訴機器
你想找什麼樣的函式？

幫助機器找到「函式」的方法之一

Supervised Learning

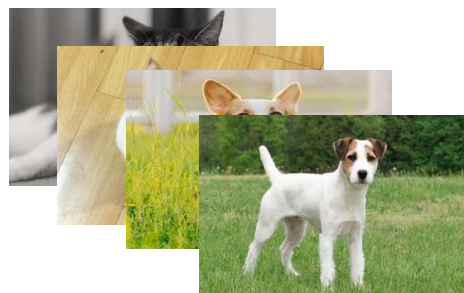
關鍵：必須準備好 Labelled Data



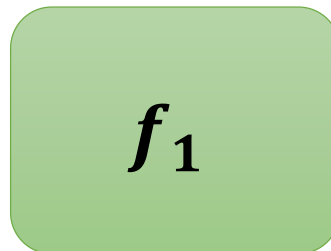
Labelled Data

將這些 Labelled Data 丟入「函式」中，機器就可以去評 Loss

函式的 Loss



$x_1 / x_2 / x_3 / x_4$



Loss = 50%

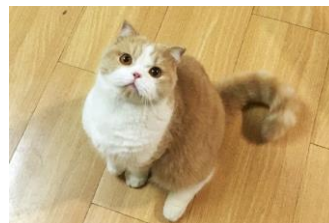
“Dog” / “Dog” /
“Dog” / “Dog”

x_1 :



y_1 : “Cat”

x_2 :



y_2 : “Cat”

x_3 :



y_3 : “Dog”

x_4 :

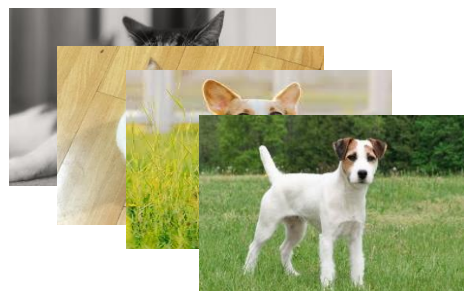


y_4 : “Dog”

Labelled Data

函式的 Loss

接下來機器會自動找出
Loss 最低的函式



$x_1 / x_2 / x_3 / x_4$



f_2



Loss = 0%

“Cat”/“Cat”/
“Dog”/“Dog”

x_1 :



y_1 : “Cat”

x_2 :



y_2 : “Cat”

x_3 :



y_3 : “Dog”

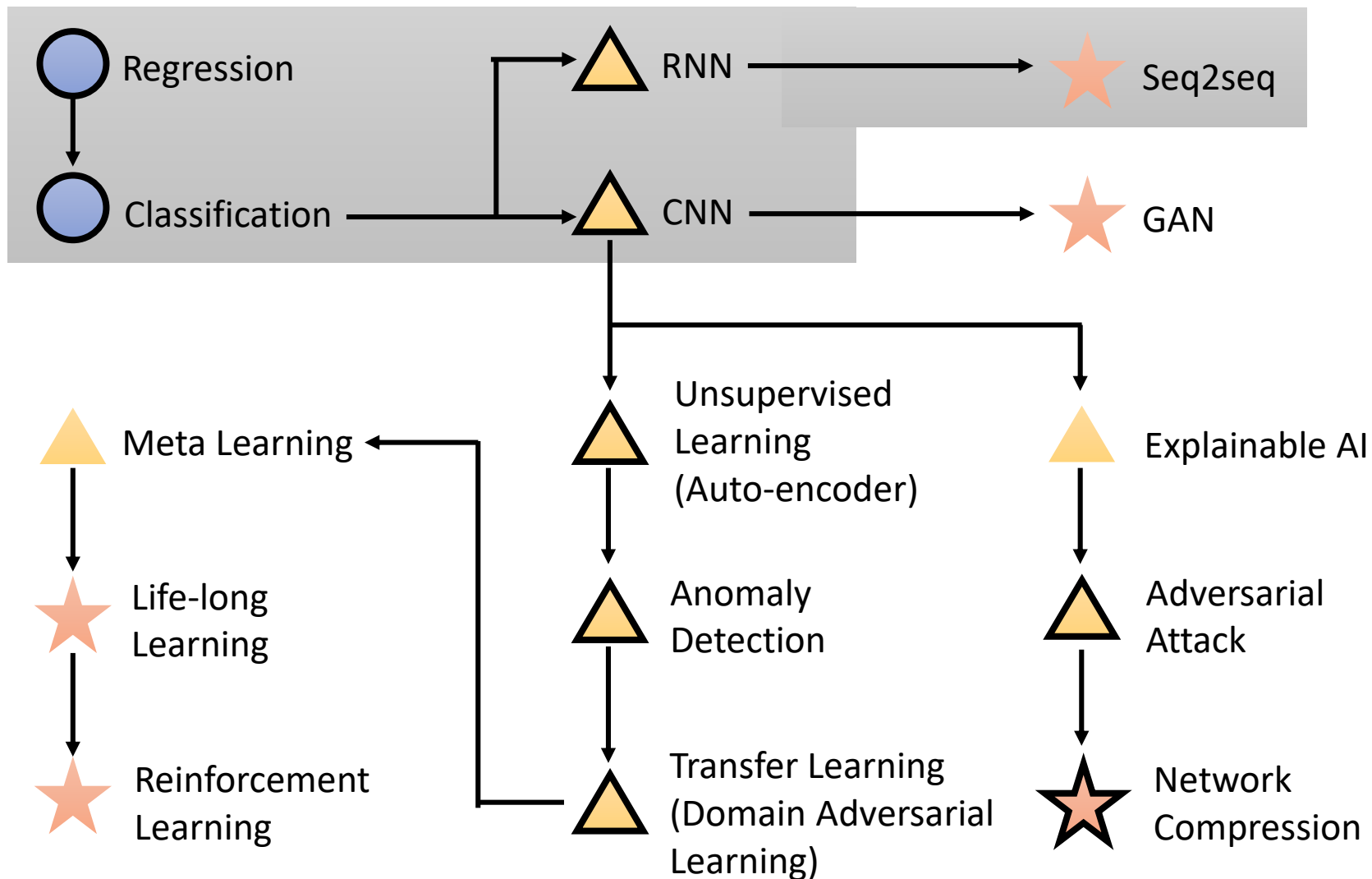
x_4 :



y_4 : “Dog”

Labeled Data

Supervised Learning



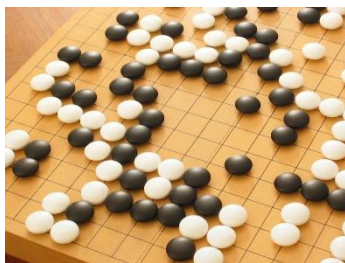
Reinforcement Learning



Supervised v.s. Reinforcement

supervised learning 的關鍵：必須提供 label data => 所以必須讓機器知道面對這樣的「盤勢」必須下「哪一步」

- Supervised:

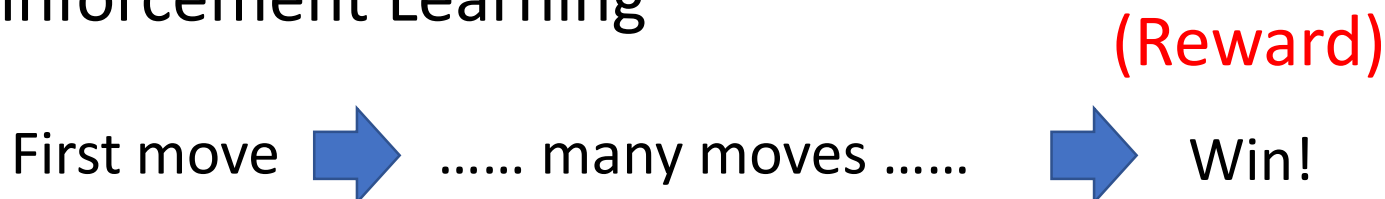


Next move:
"5-5"



Next move:
"3-3"

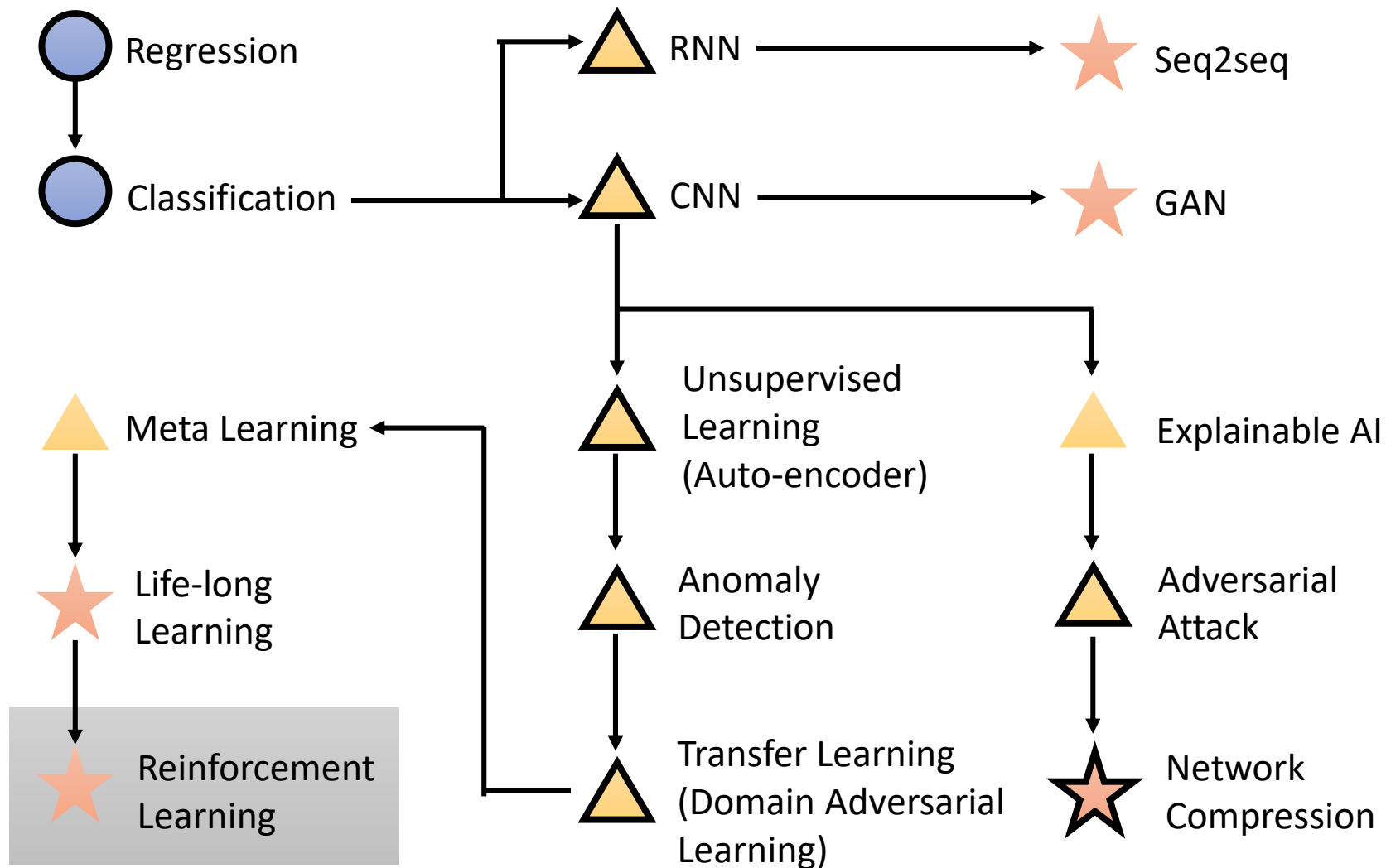
- Reinforcement Learning



reinforcement learning 的關鍵：讓機器與機器或與人不停的下棋，當機器贏了之後就會得到 reward，機器必須去學習哪些「步」是好的！

Alpha Go is supervised learning + reinforcement learning.

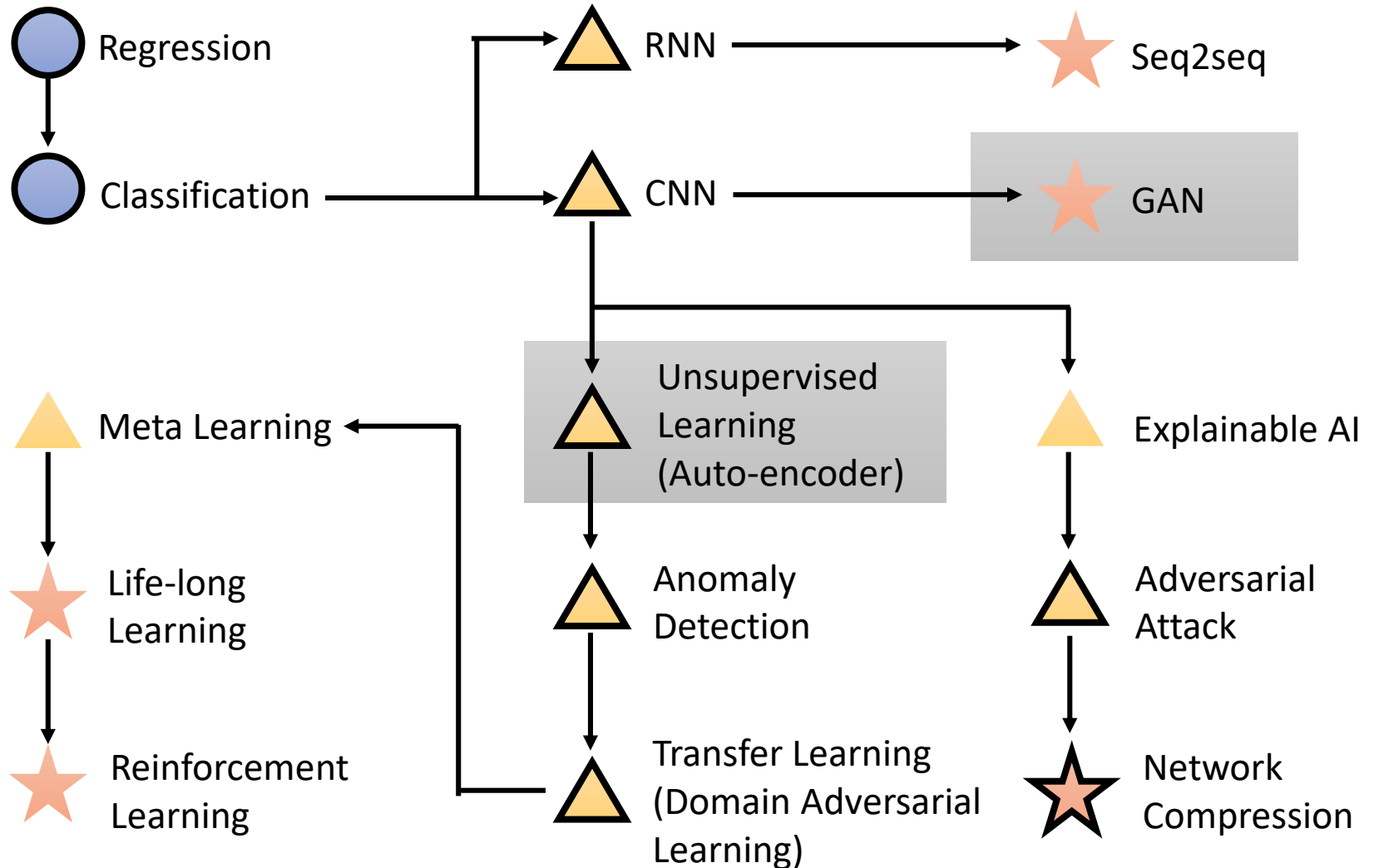
Reinforcement Learning



Unsupervised Learning



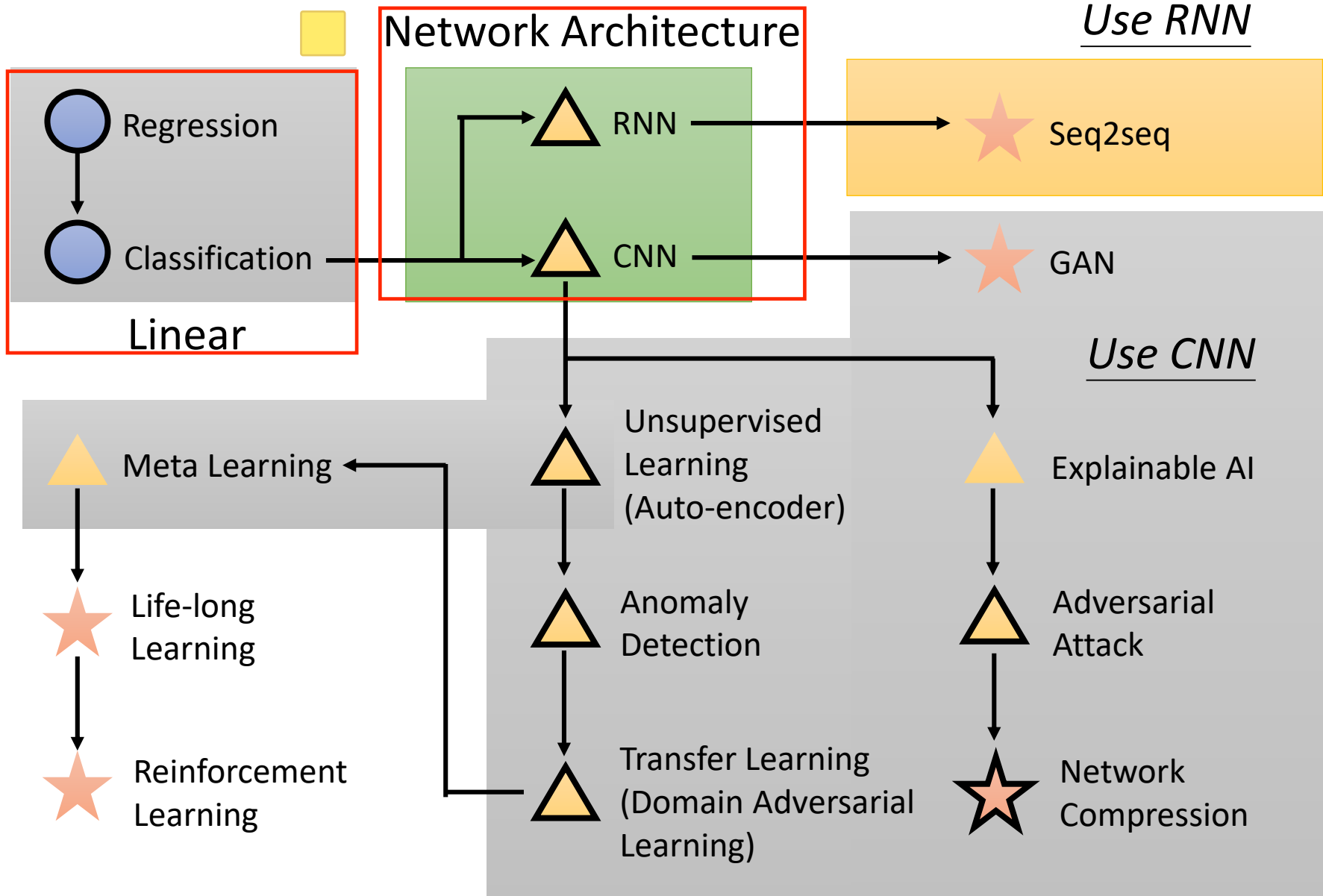
What can machine learn from unlabeled images?



機器怎麼 找出你想要的函式？

當已經知道要讓機器去找什麼「目的」的「函式」之後，接下來就要「限制函式的範圍」，告訴機器不用去考慮世界上所有的函式！

限制函式尋找範圍



函式尋找方法 – Gradient Descent



Implement the
algorithm by yourself

Regression

Classification

Deep Learning Framework

(3/26 PyTorch 教學、會錄影)



RNN



Seq2seq



CNN



GAN



Meta Learning



Life-long
Learning



Reinforcement
Learning



Unsupervised
Learning
(Auto-encoder)



Anomaly
Detection



Transfer Learning
(Domain Adversarial
Learning)



Explainable AI

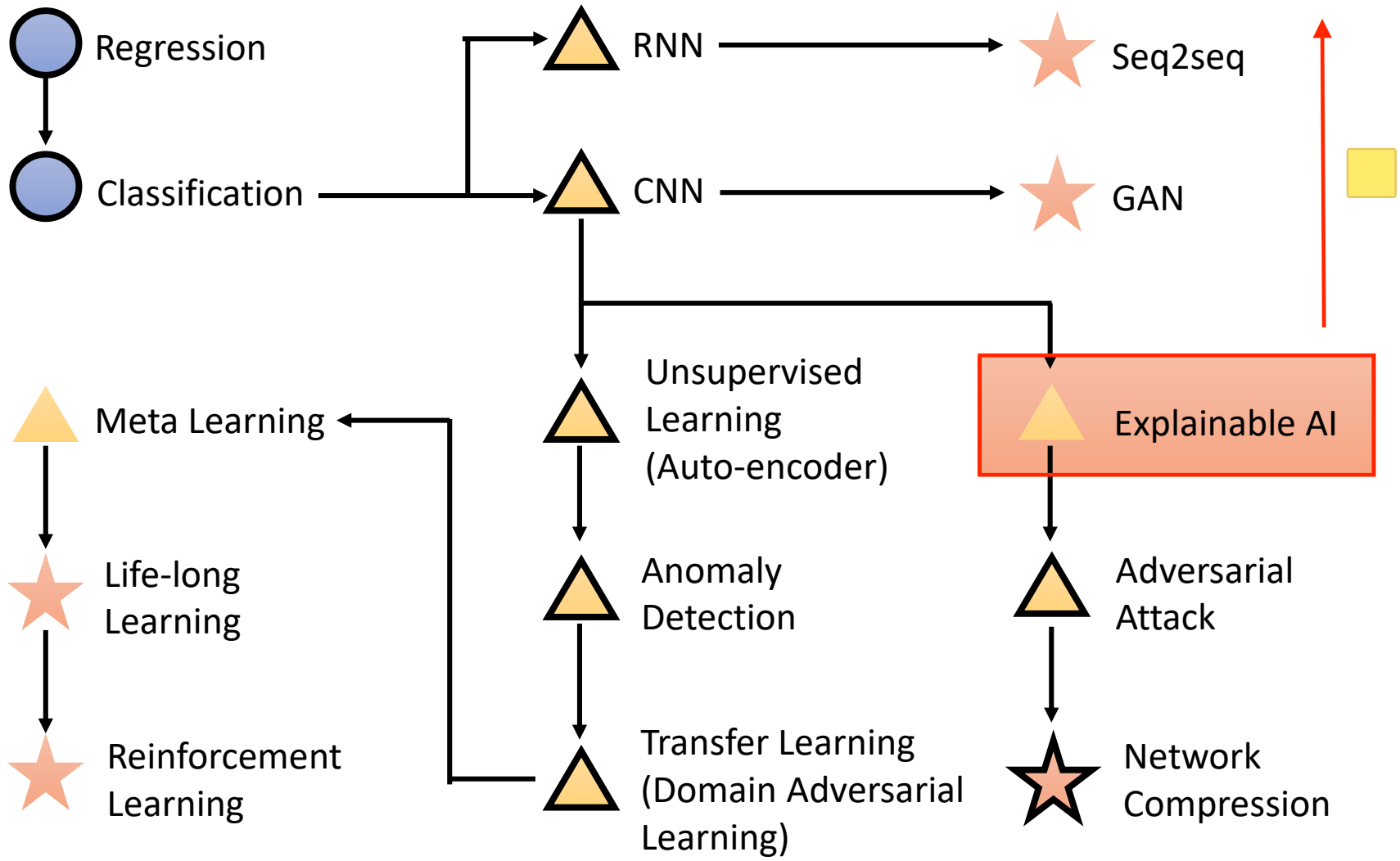
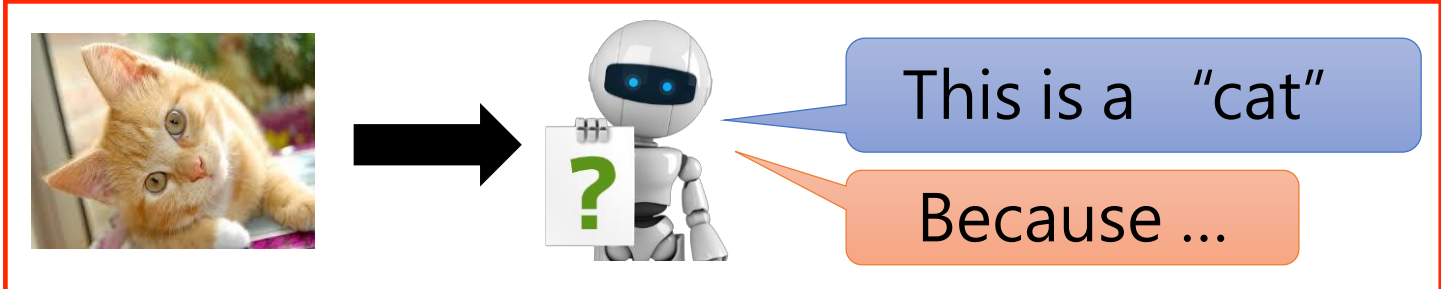


Adversarial
Attack

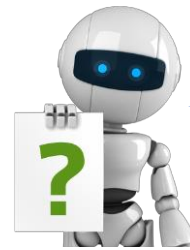


Network
Compression

前沿研究

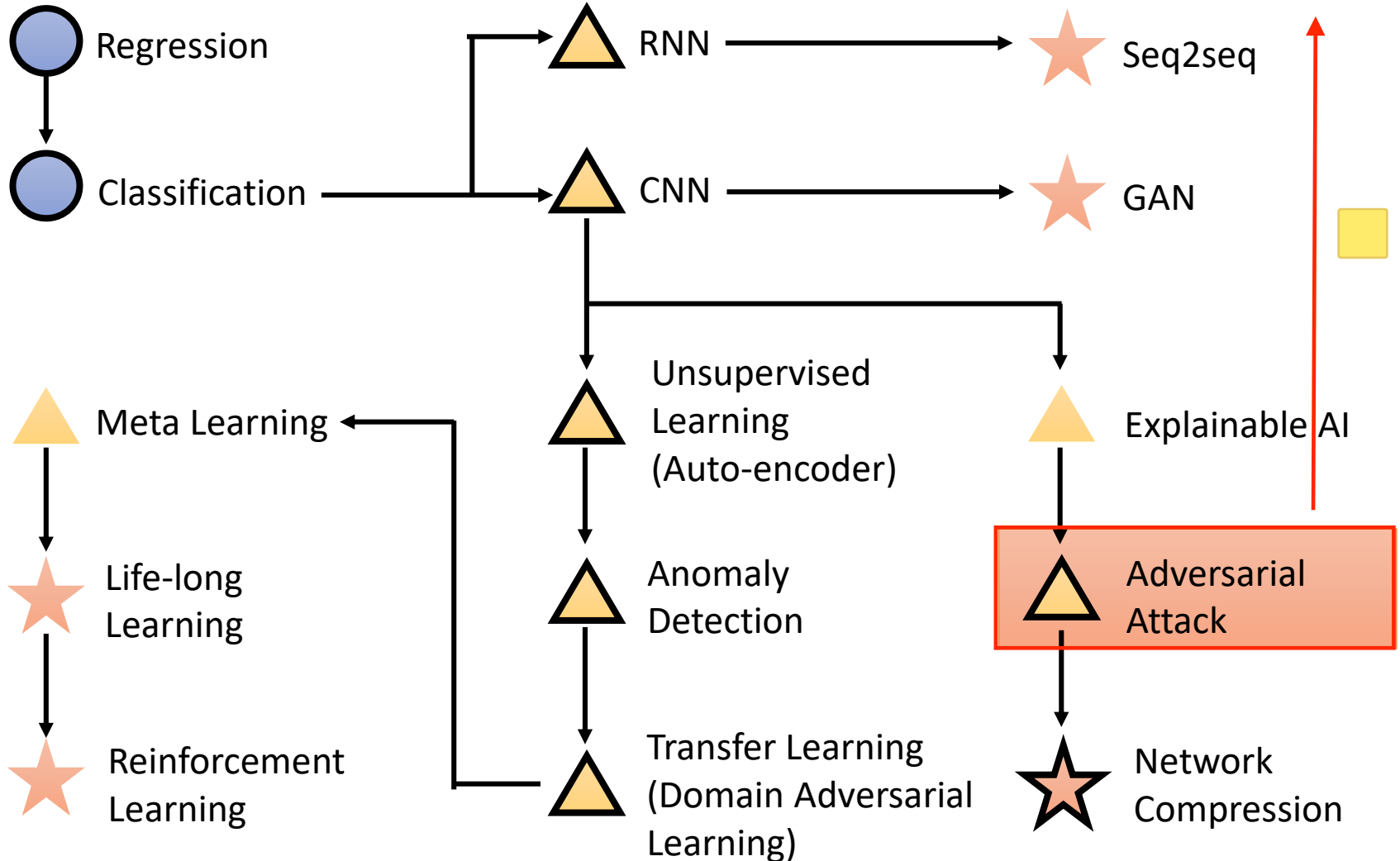


Add
noise



This is a "cat"

Star Fish ...





This is a "cat"



Regression



Classification



Meta Learning



Life-long Learning



Reinforcement Learning



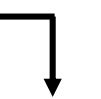
Transfer Learning
(Domain Adversarial Learning)



Seq2seq



GAN



Explainable AI

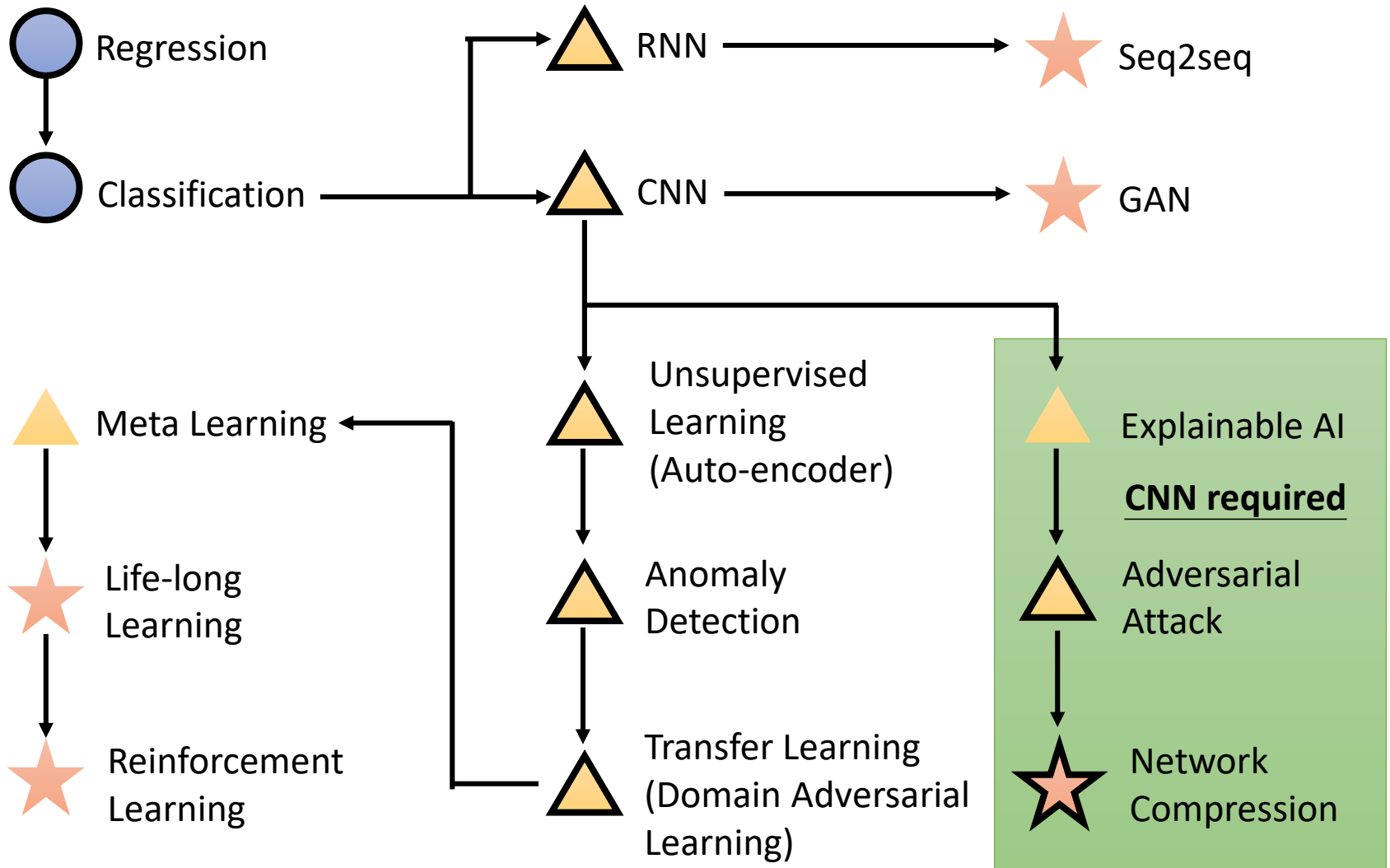


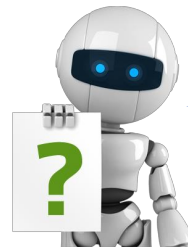
Adversarial Attack



Network Compression

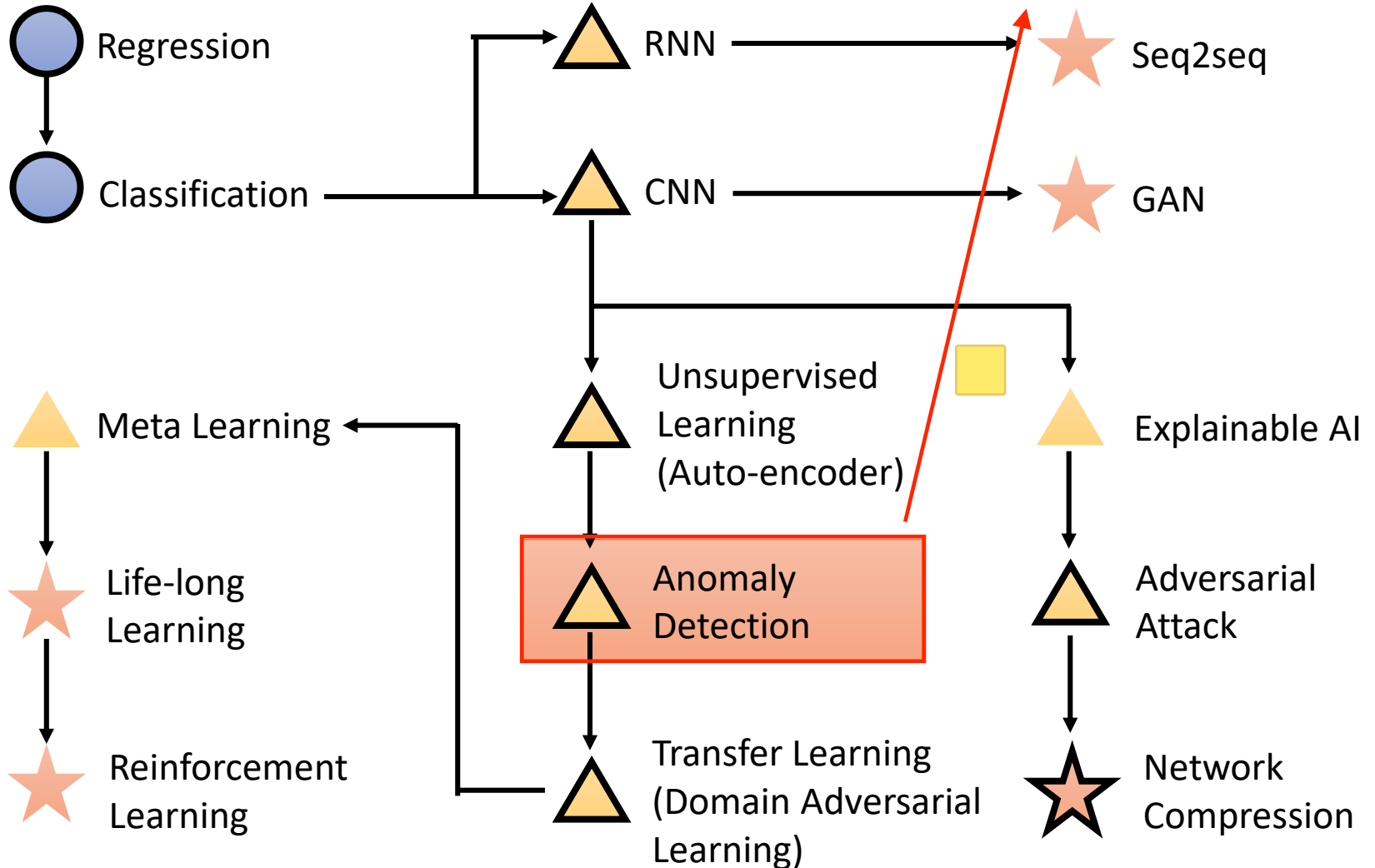






This is a "cat"

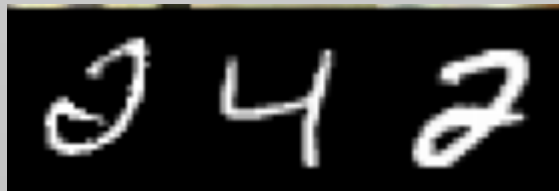
我不知道



Training
Data

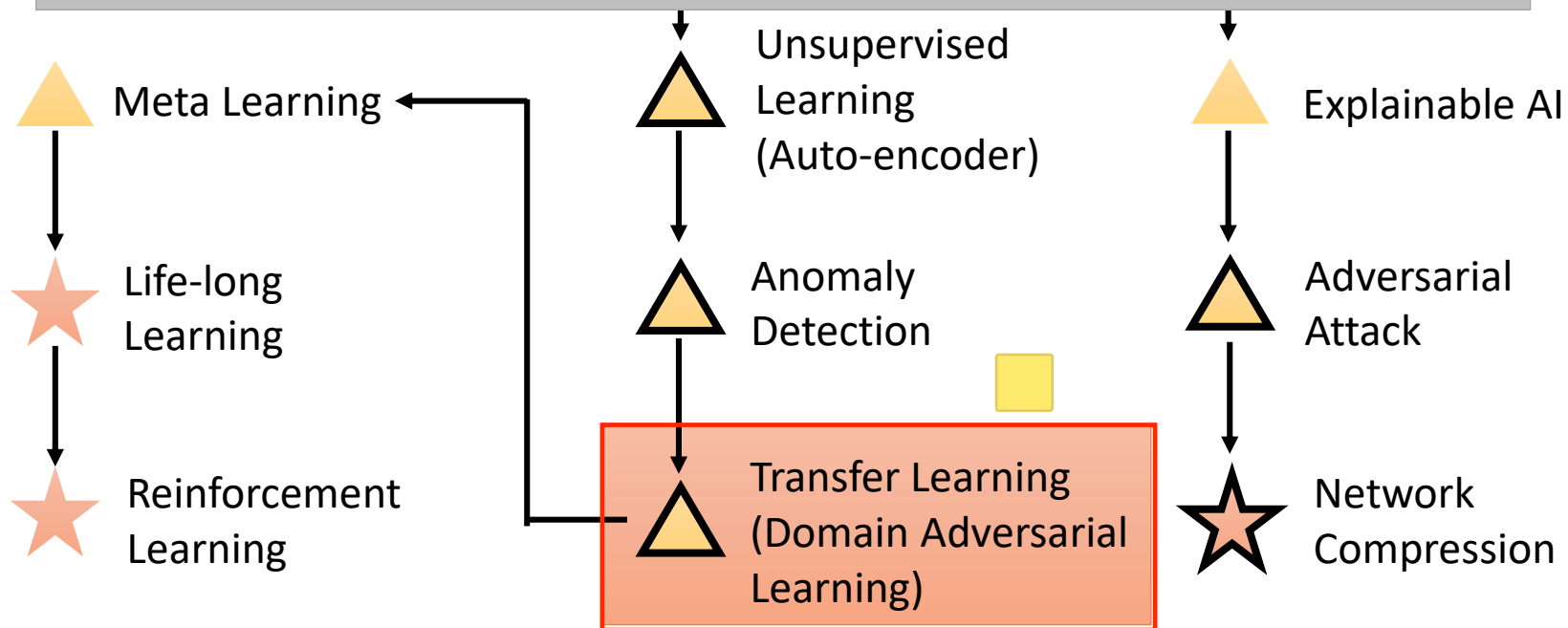


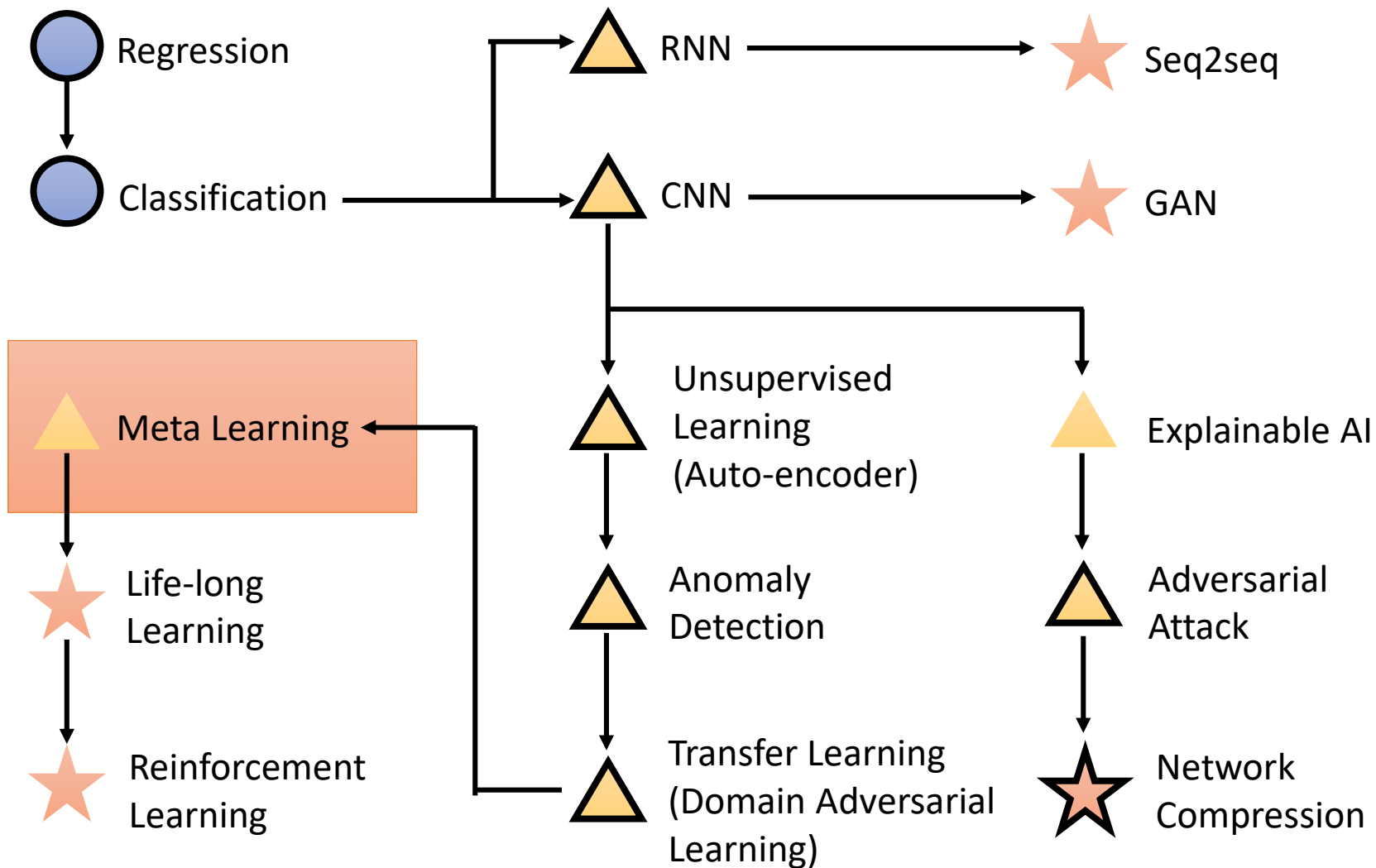
Testing
Data



99.5%

57.5%





Meta Learning = Learn to learn

- Now we design the learning algorithm



program
for learning



I can learn!

- Can machine learn the learning algorithm?



program designing
program
for learning

program
for learning



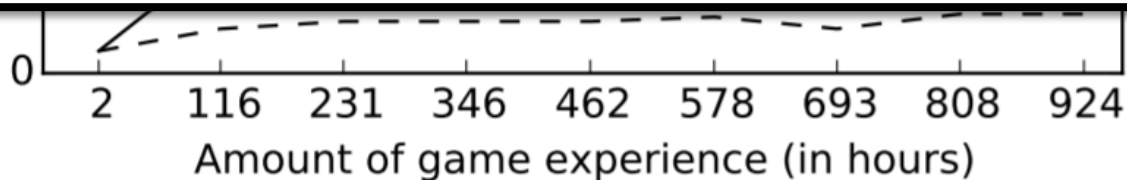
I can learn!

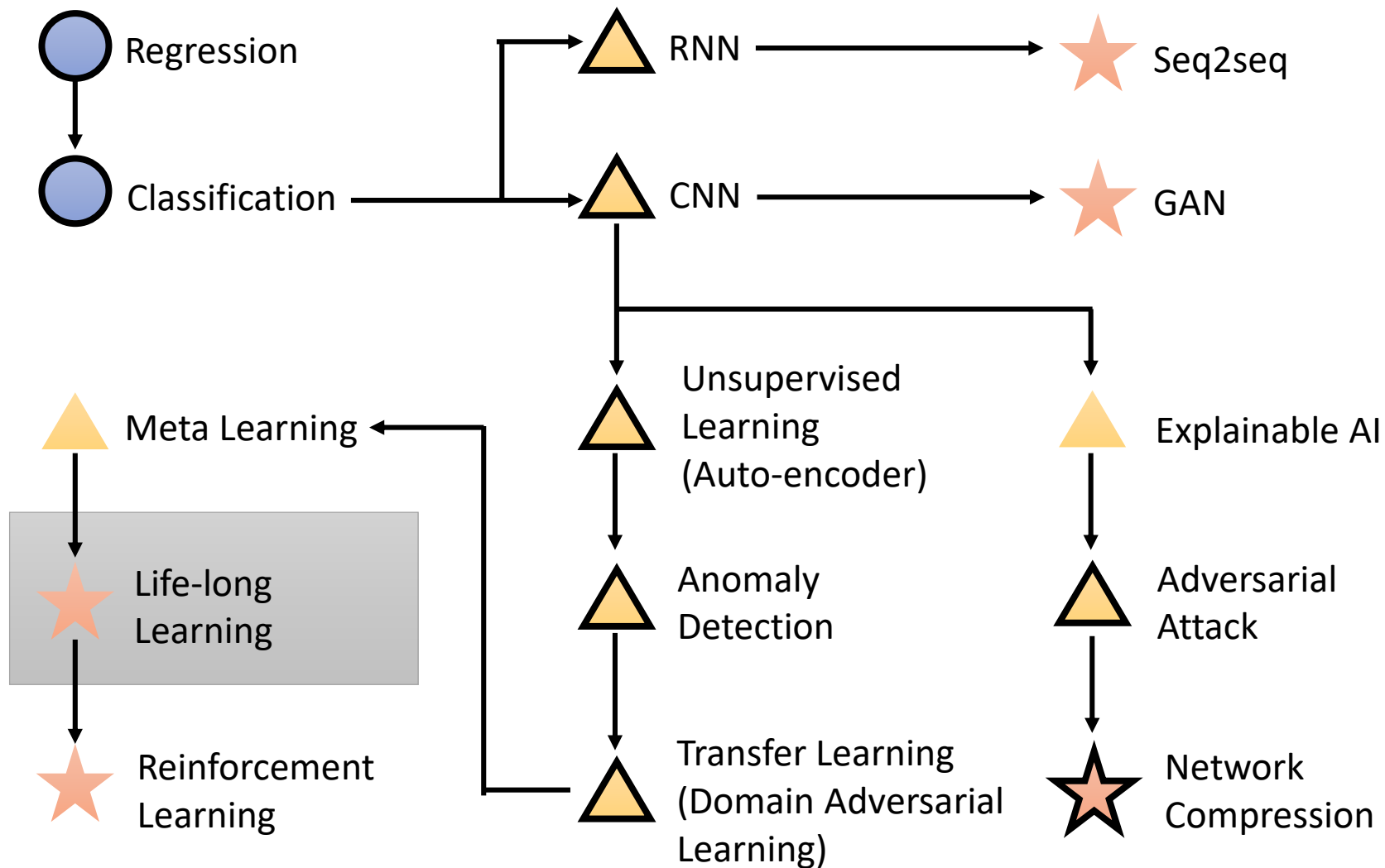
能不能讓機器聰明一點？

天資不佳卻勤奮不懈？

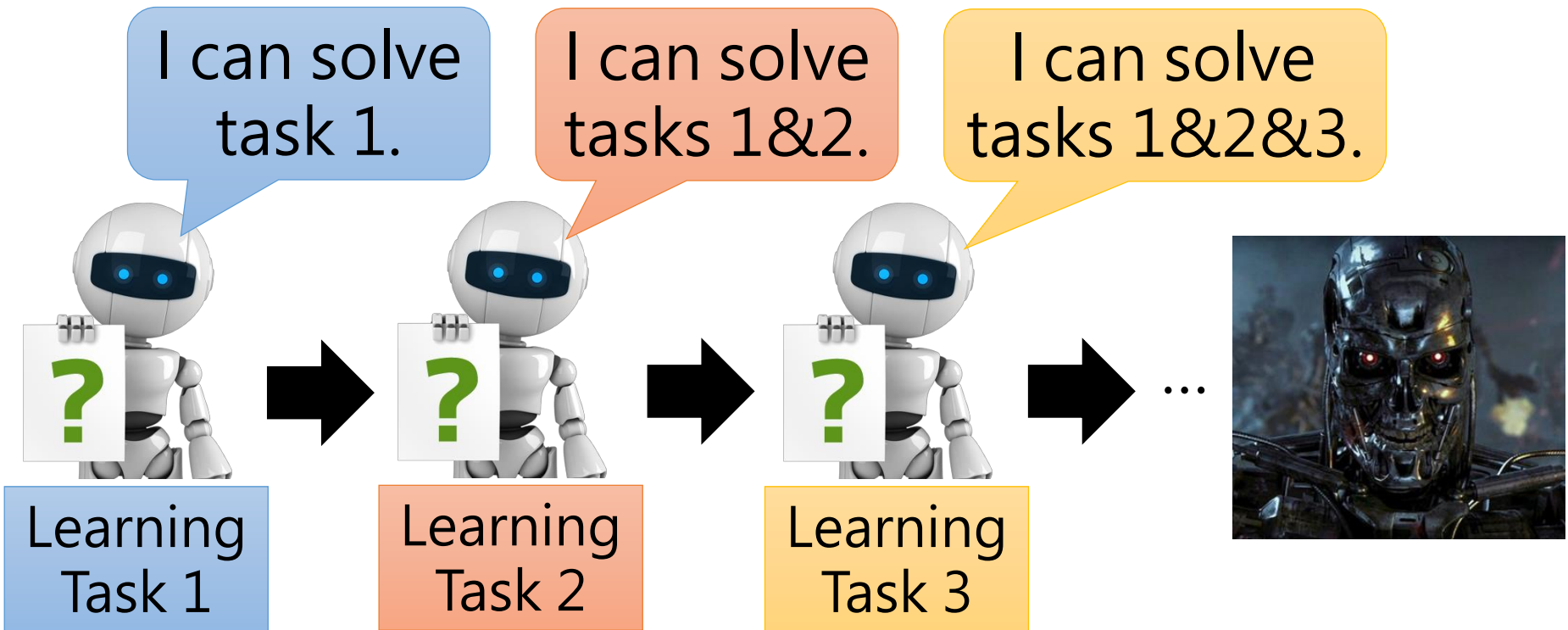
5000

In order to train AlphaStar, we built a highly scalable distributed training setup using [Google's v3 TPUs that](#) supports a population of agents learning from many thousands of parallel instances of StarCraft II. The AlphaStar league was run for 14 days, using 16 TPUs for each agent. During training, each agent experienced up to 200 years of real-time StarCraft play. The final AlphaStar agent consists of the components of the [Nash distribution of the league](#) - in other words, the most effective mixture of strategies that have been discovered - that run on a single desktop GPU.





終身學習 (Life-long Learning)



Life-Long Learning (終身學習), Continuous Learning,
Never Ending Learning, Incremental Learning