



# THE NEW ERA: AI

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### **Artificial Intelligence**

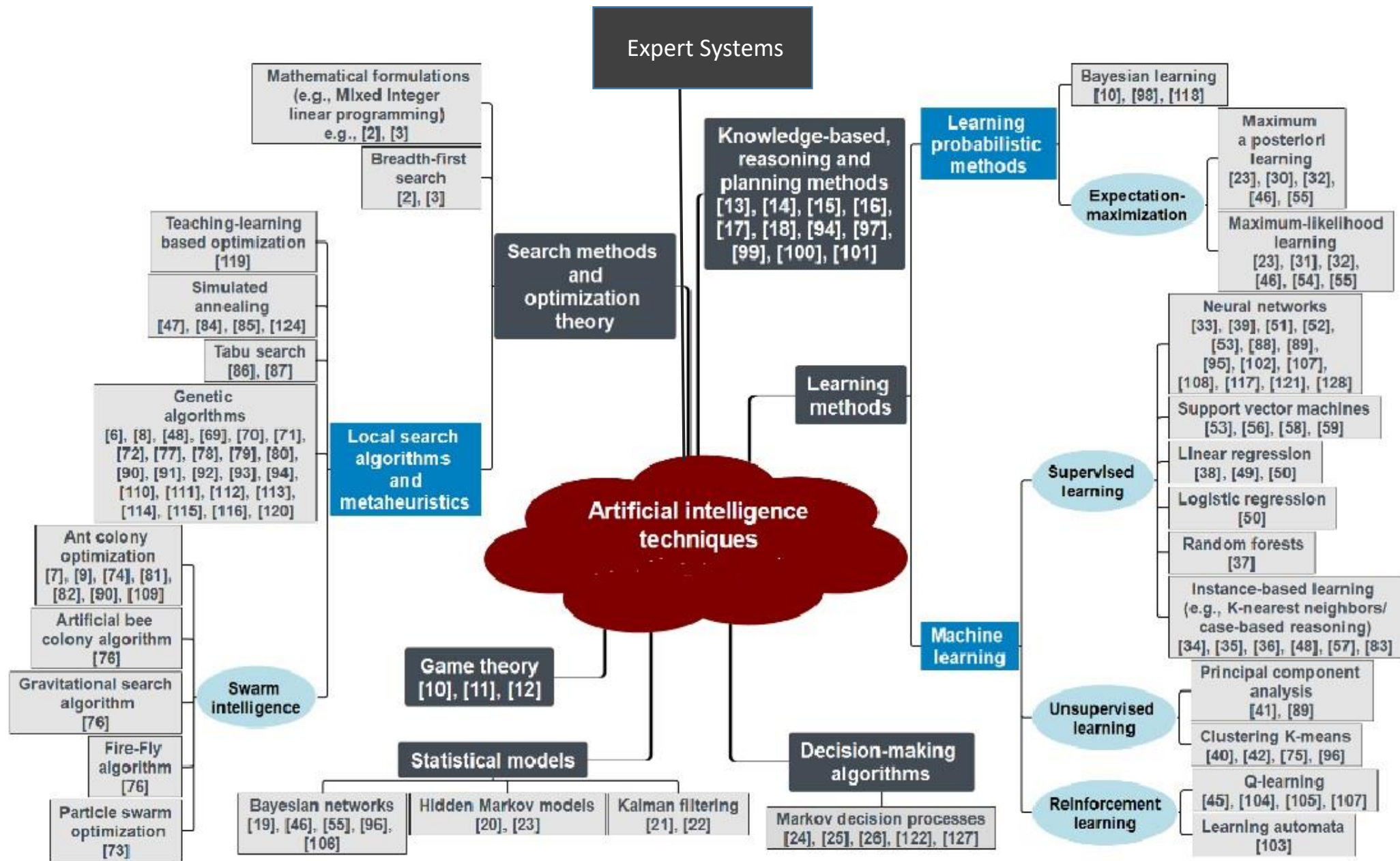
Algorithms that mimic the intelligence of humans, able to resolve problems in ways we consider “smart”. From the simplest to most complex of the algorithms.

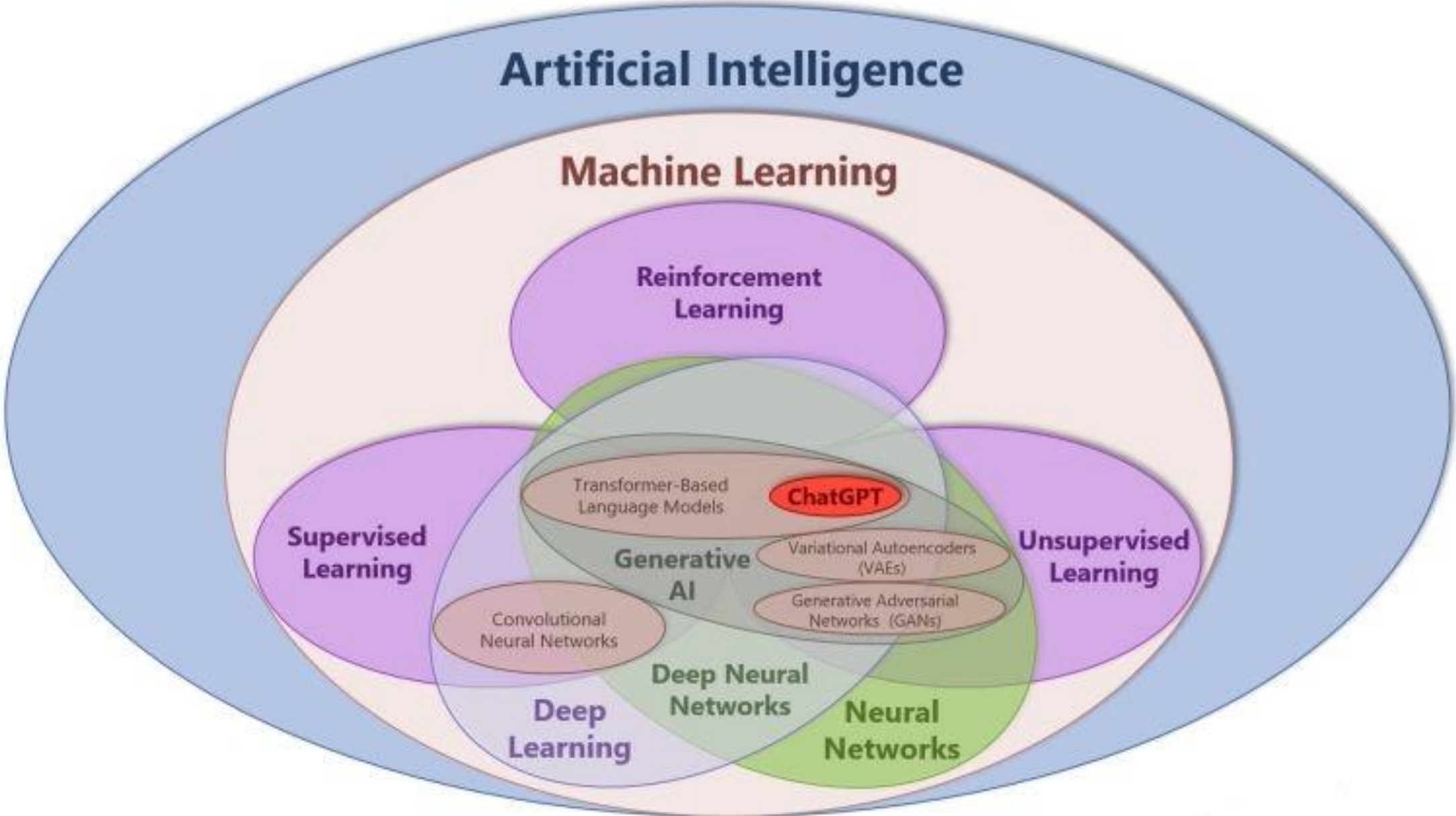
### **Machine Learning**

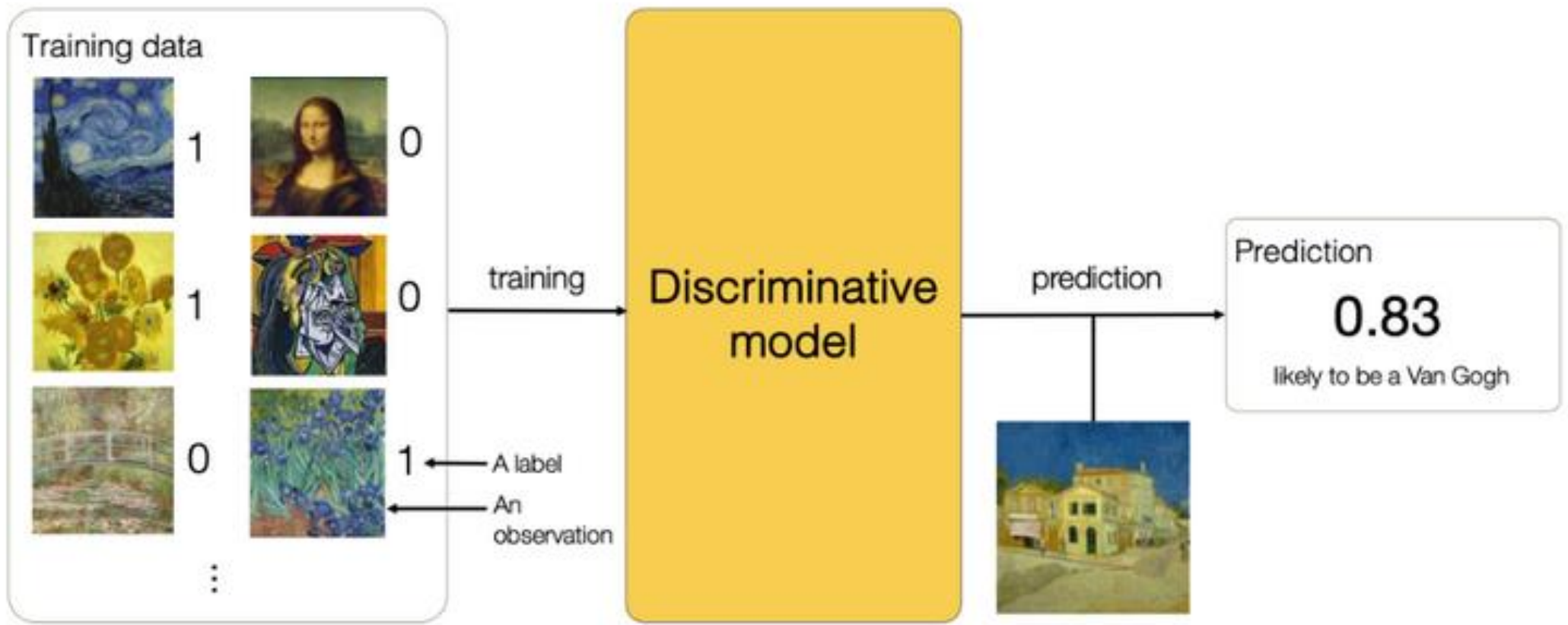
Algorithms that parse data, learn from it, and then apply what they’ve learned to make informed decisions. They use human extracted features from data and improve with experience.

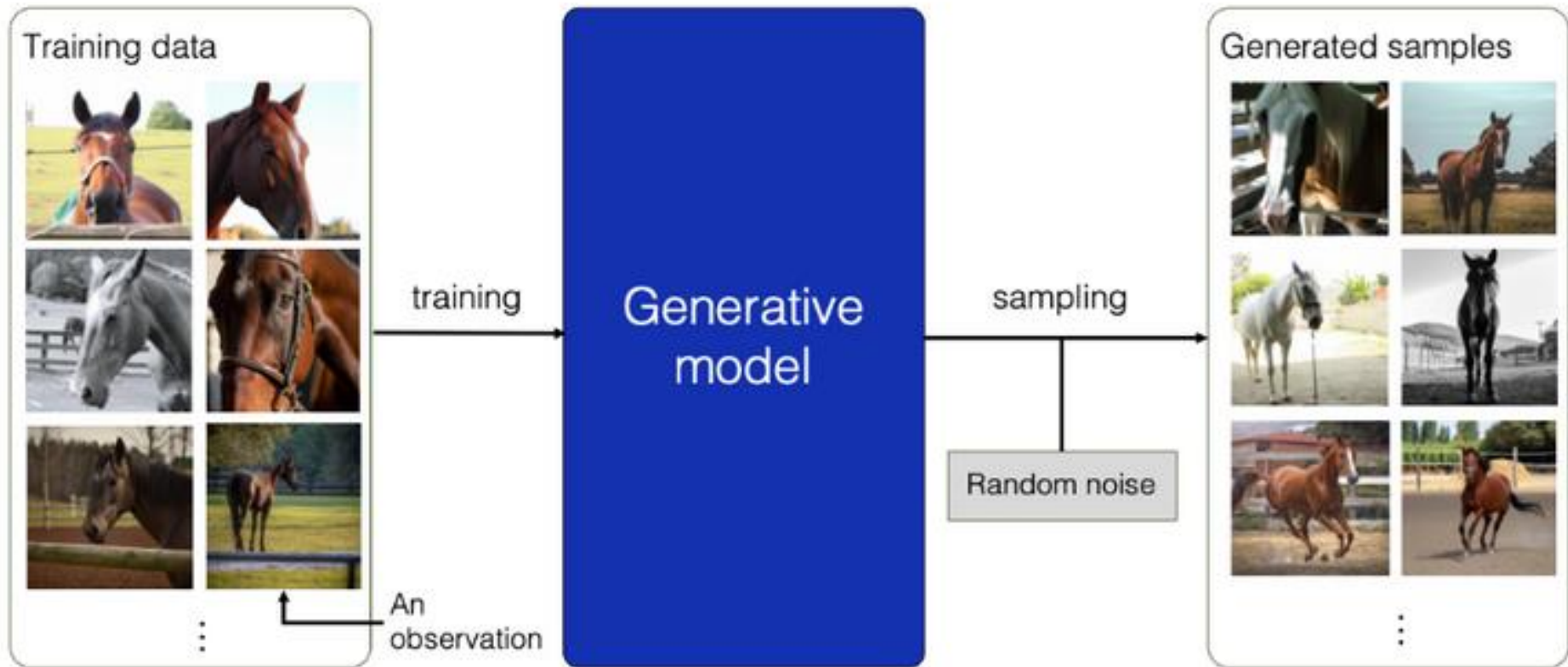
### **Deep Learning**

Neural Network algorithms that learn the important features in data by themselves. Able to adapt themselves through repetitive training to uncover hidden patterns and insights.









# Google's artificial intelligence finds two new exoplanets missed by human eyes

December 15, 2017 6.38am GMT

Artist impression of Kepler-90i, the eighth planet discovered orbiting around Kepler-90. NASA

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Two new exoplanets have been discovered thanks to NASA's collaboration with Google's artificial intelligence (AI). One of those in [today's announcement](#) is an eighth planet – Kepler-90i – found orbiting the Sun-like star Kepler-90. This makes it the first system discovered with an equal number of planets to our own Solar system.

A mere road trip away, at 2,545 light-years from Earth, Kepler-90i orbits its host star every 14.4 Earth days, with a sizzling surface temperature similar to Venus of 426°C.

The new exoplanets are added to the growing list of known worlds found orbiting other stars.





Health & Medicine

# AI Can Diagnose Heart Disease and Lung Cancer More Accurately Than Doctors

## IN BRIEF

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A pair of recently developed AI systems can diagnose lung cancer and heart disease more accurately than human doctors. These AIs have the potential to save billions of dollars and countless lives if widely adopted.

SHARE



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WRITTEN BY

Kyree Leary







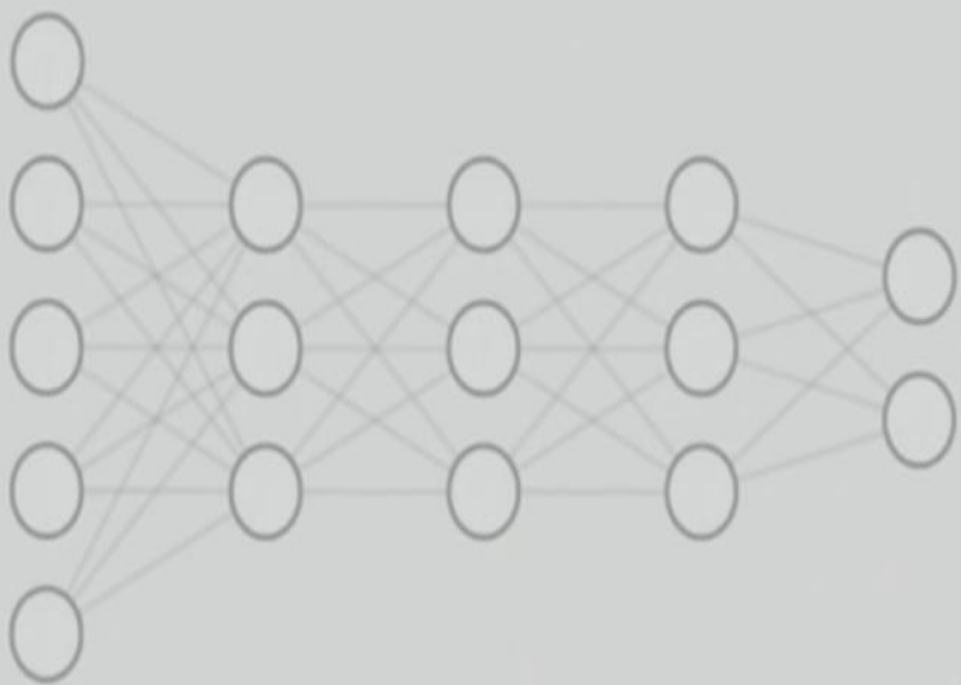
Examples of images generated by training a generative adversarial network (GAN) with portraits from the last 500 years of Western art. The distorted faces are the algorithm's attempts to imitate those inputs. Images generated at Art & Artificial Intelligence Laboratory, Rutgers.

[thispersondoesnotexist.com](http://thispersondoesnotexist.com)

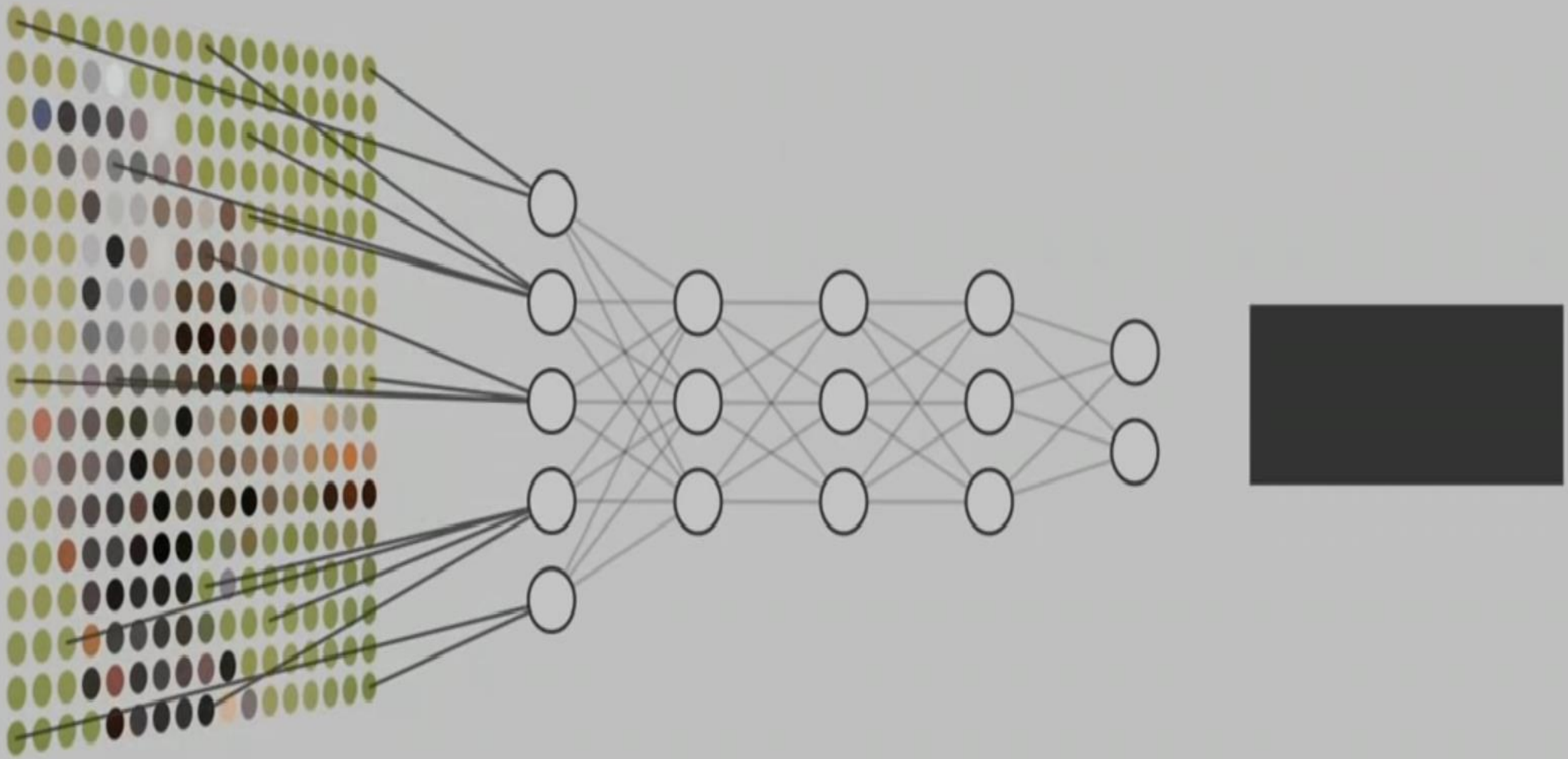


# Generative AI Applications

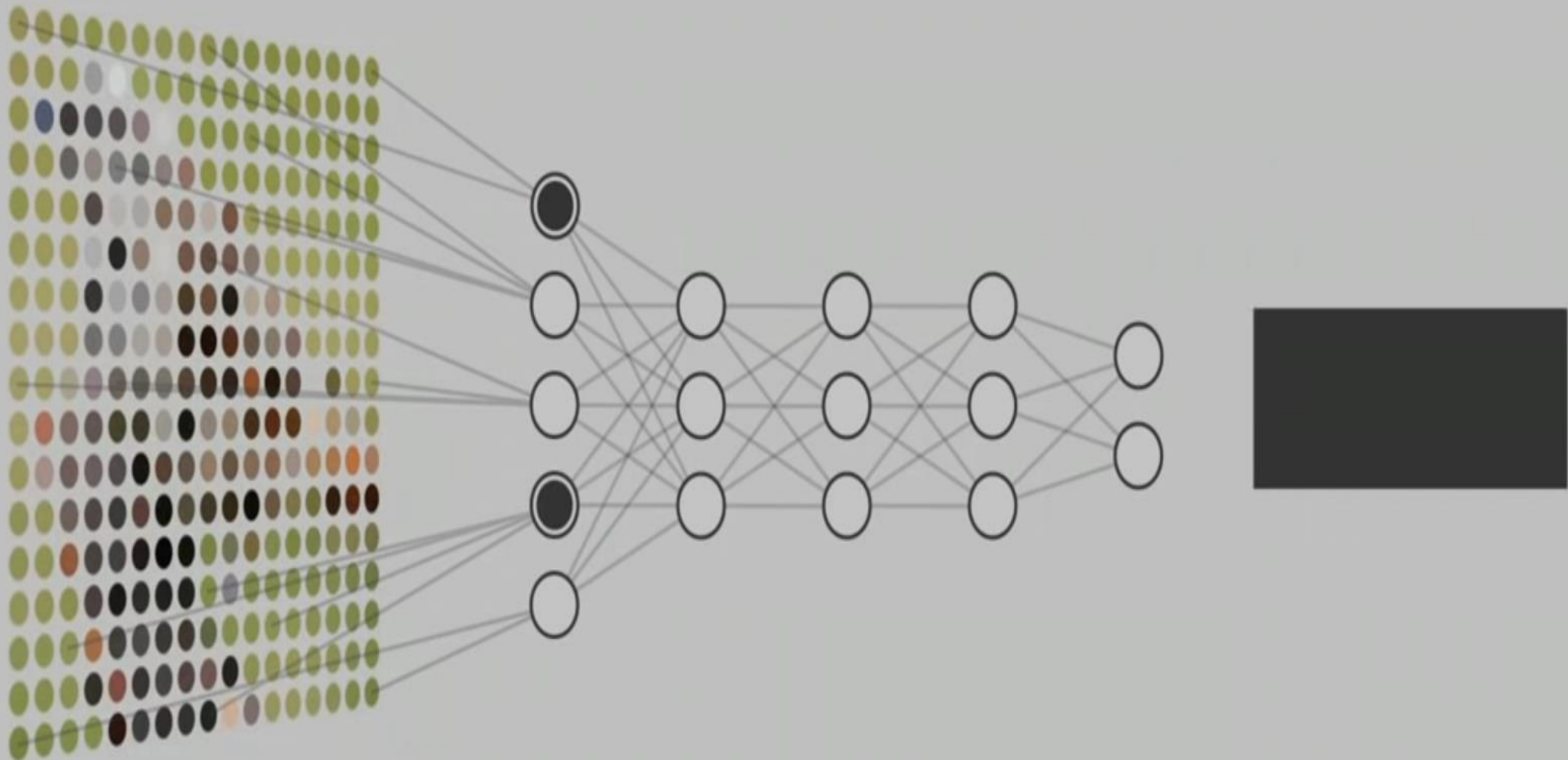


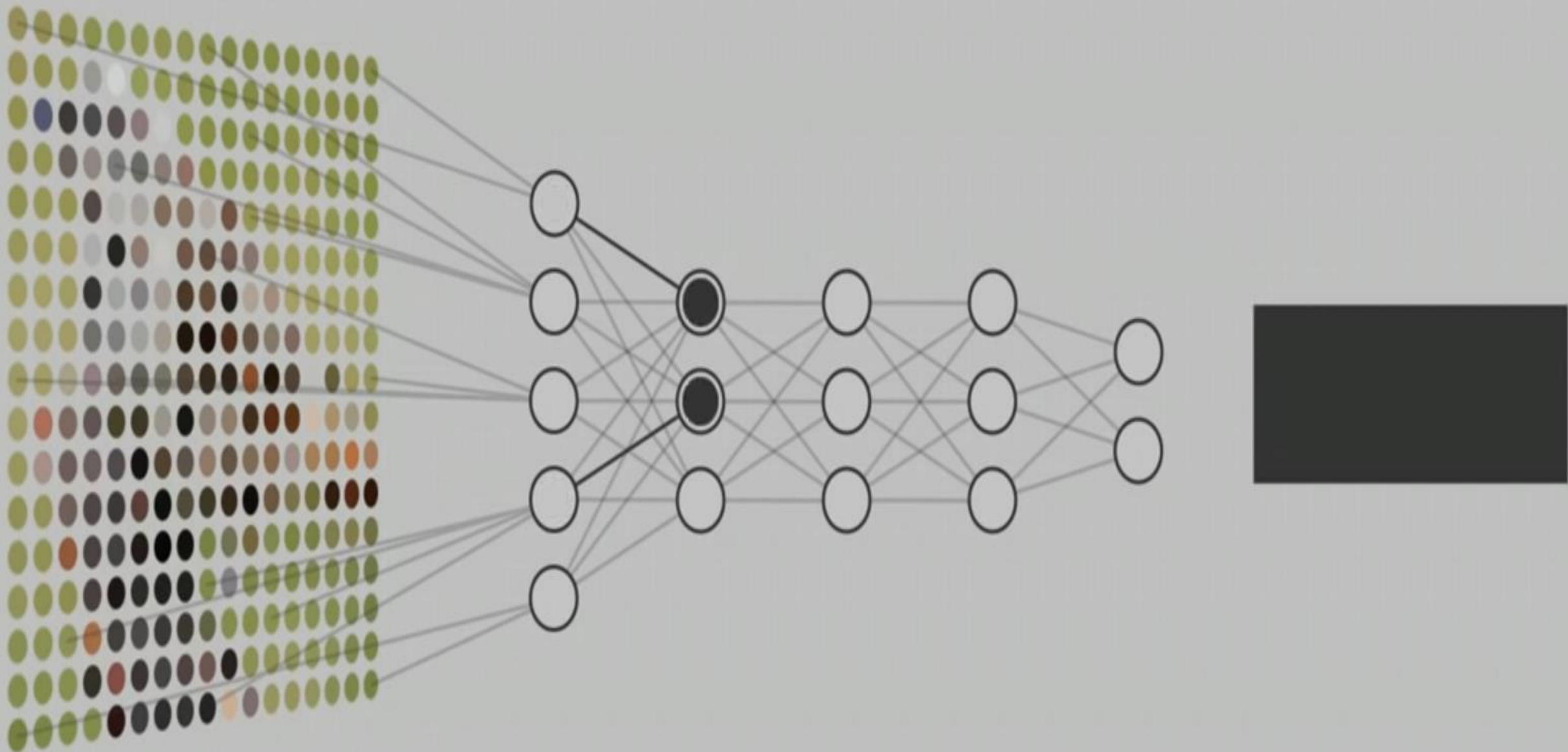


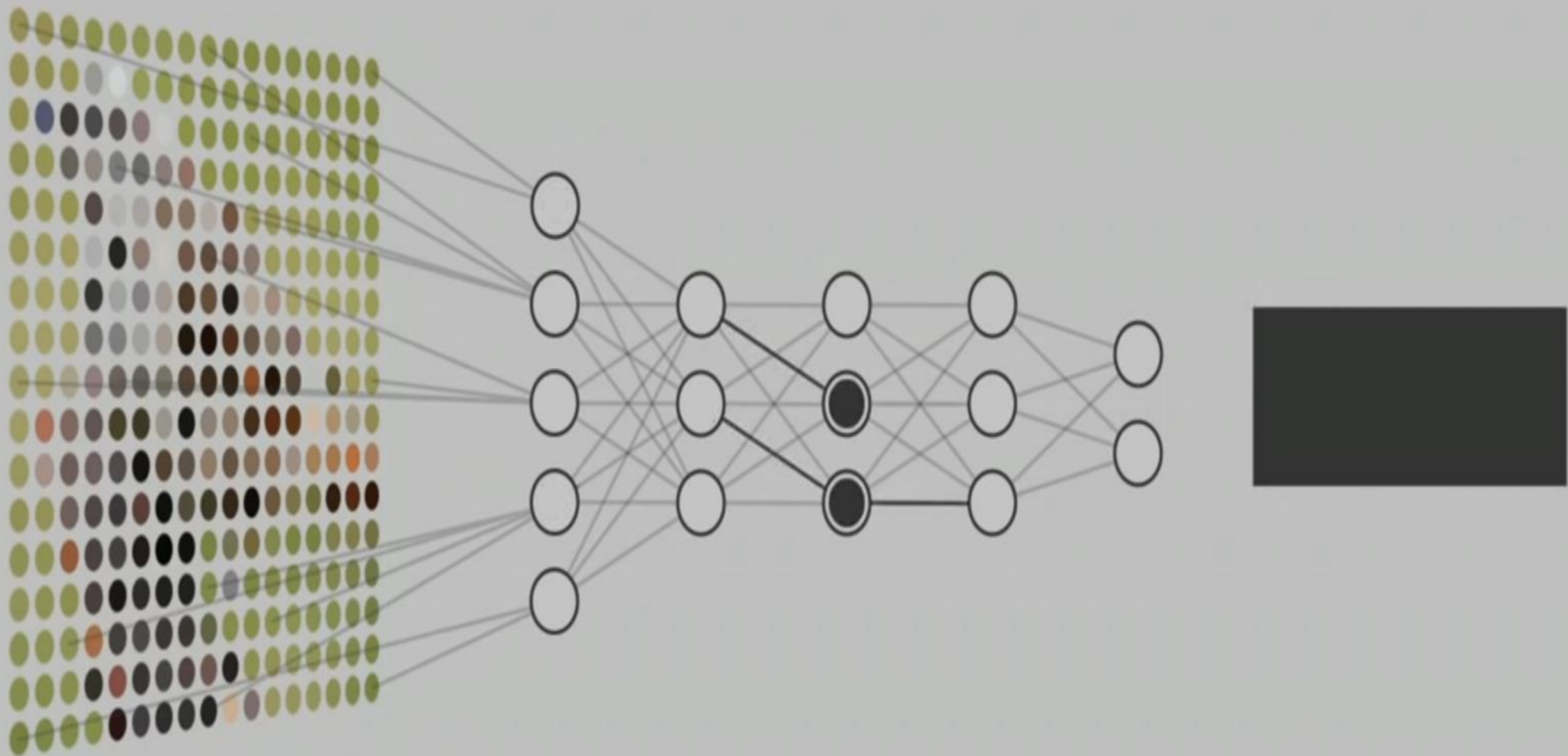
**BIRD**

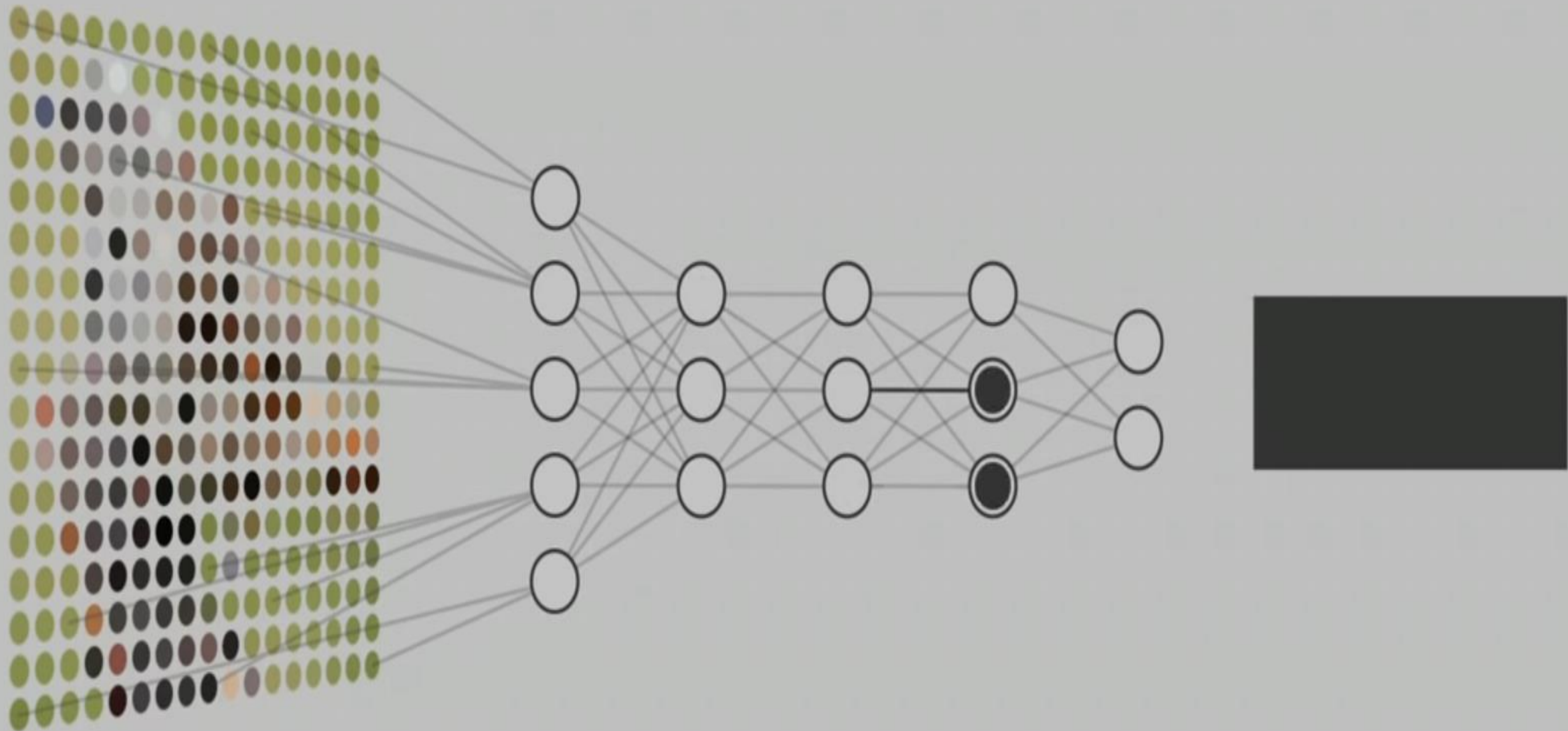


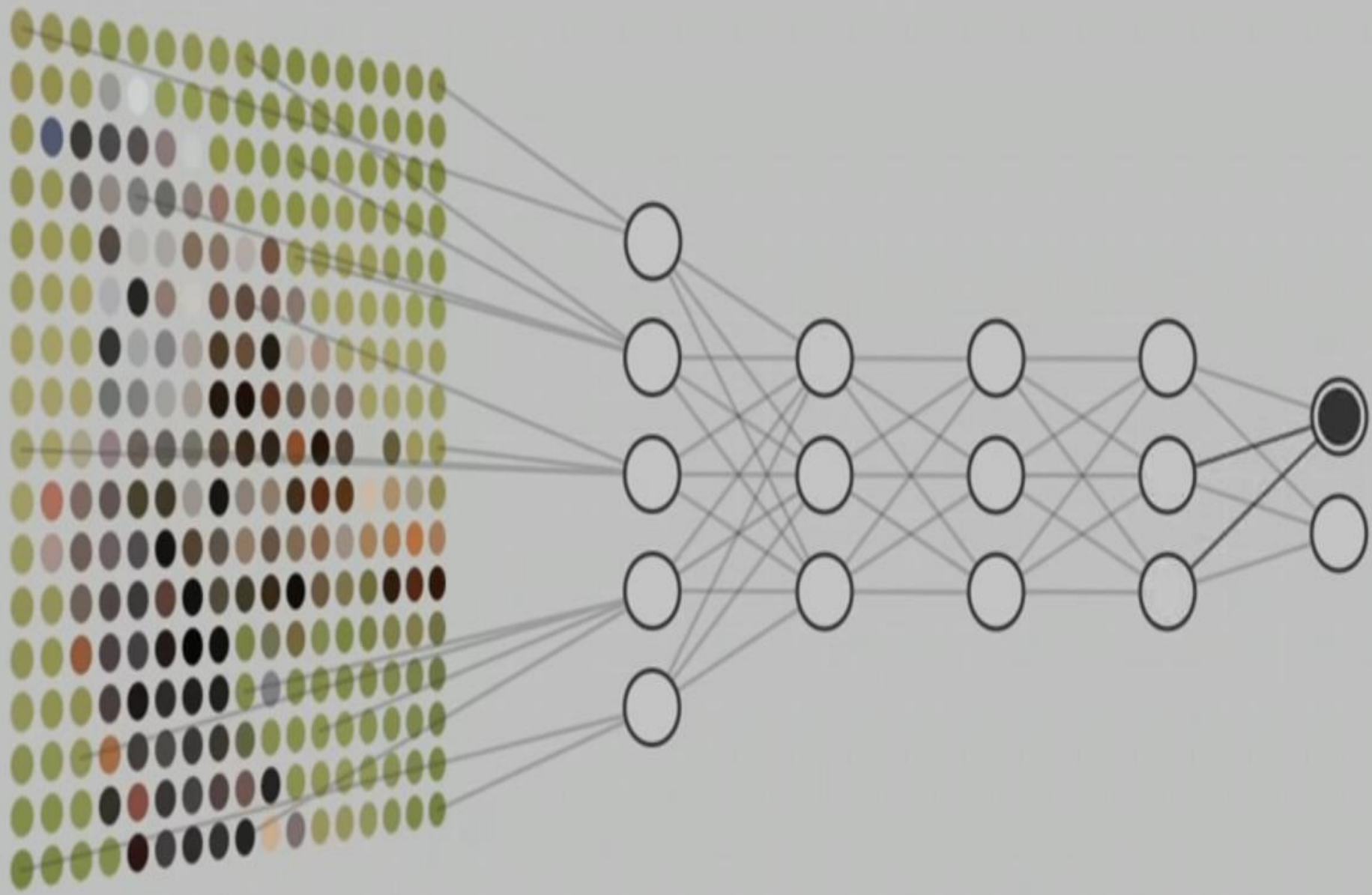






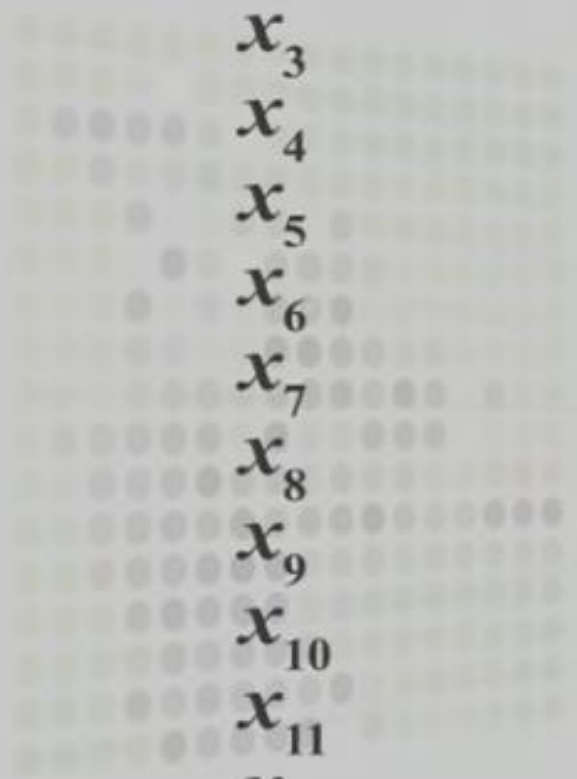






**BIRD**

$x_1$   
 $x_2$   
 $x_3$   
 $x_4$   
 $x_5$   
 $x_6$   
 $x_7$   
 $x_8$   
 $x_9$   
 $x_{10}$   
 $x_{11}$   
 $x_{12}$   
 $x_{13}$   
 $\vdots$

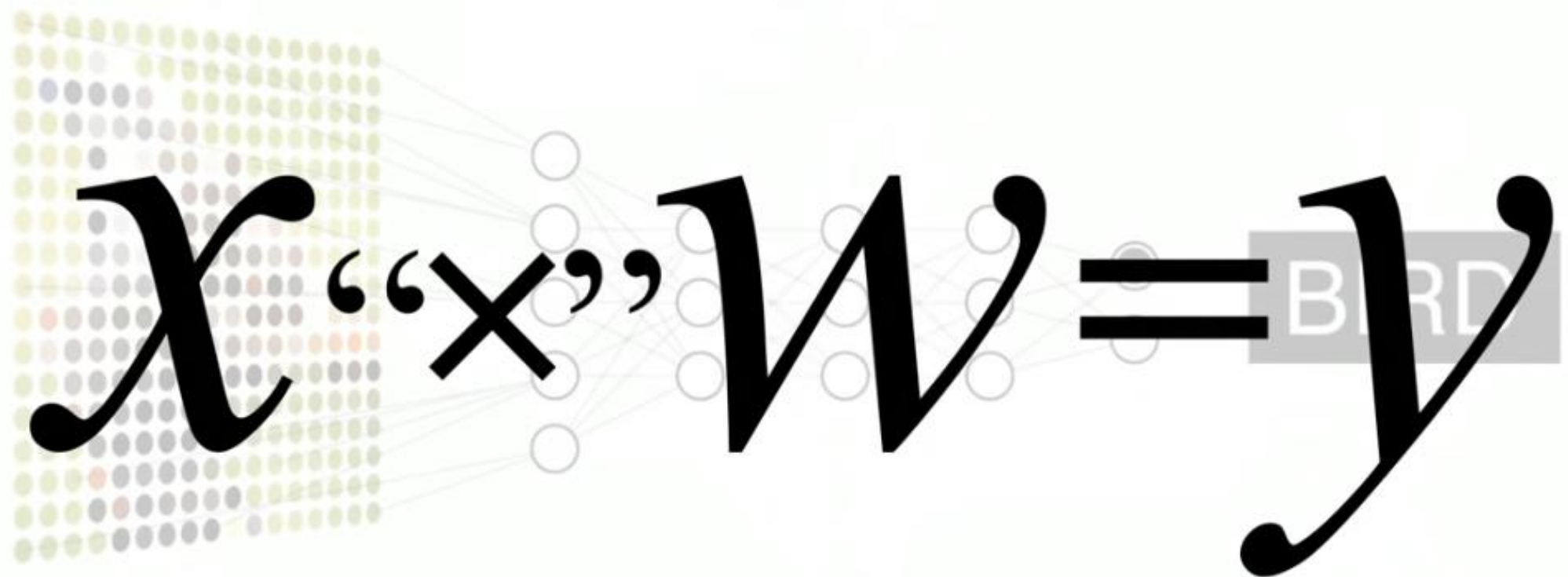


$w_1$   
 $w_2$   
 $w_3$   
 $w_4$   
 $w_5$   
 $w_6$   
 $w_7$   
 $w_8$   
 $w_9$   
 $w_{10}$   
 $w_{11}$   
 $w_{12}$   
 $w_{13}$   
 $\vdots$



$y_1$   
 $y_2$   
 $y_3$   
 $y_4$

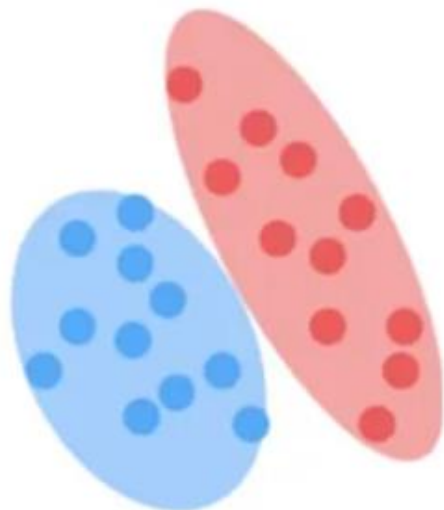




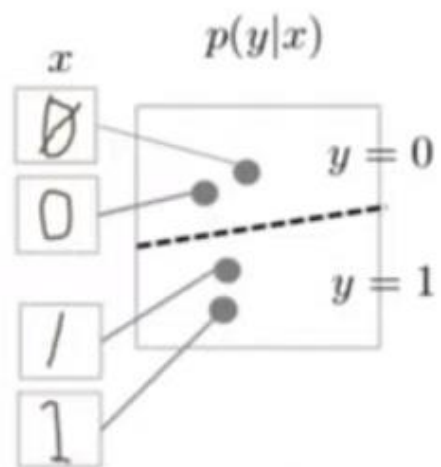
### Discriminative



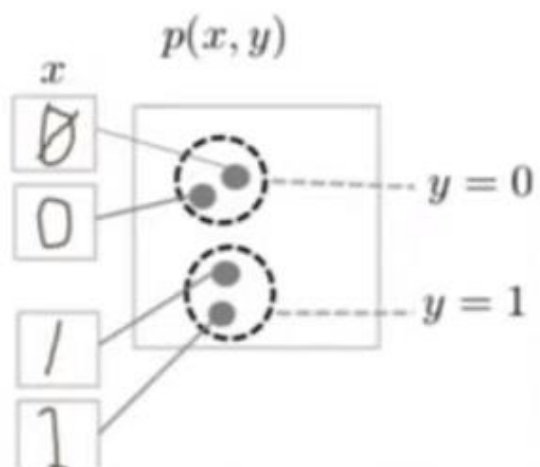
### Generative



- Discriminative Model



- Generative Model





Veri kümesi =  $\{ (1,0), (1,0), (2,0), (2,1) \}$

Ayırt edici model  $p(y|x)$ 'i öğrenir.

x verildiğinde y'nin olasılığı (probability of y given x)

	$y=0$	$y=1$
$x=1$	1	0
$x=2$	$\frac{1}{2}$	$\frac{1}{2}$

Ayırt edici modeller şartlı olasılığı hesaplar

Veri kümesi =  $\{ (1,0), (1,0), (2,0), (2,1) \}$

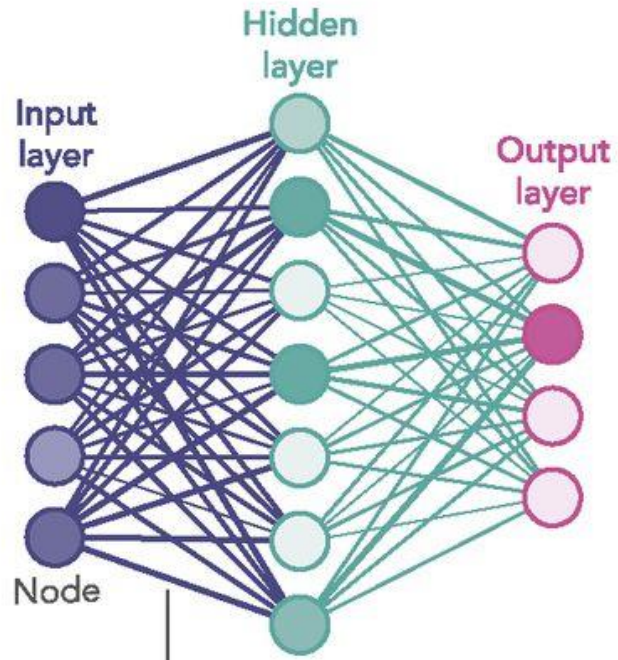
Üretici model  $p(x,y)$ 'i öğrenir.

x ve y'nin birlikte olasılığı (probability of x and y)

	y=0	y=1
x=1	$\frac{1}{2}$	0
x=2	$\frac{1}{4}$	$\frac{1}{4}$

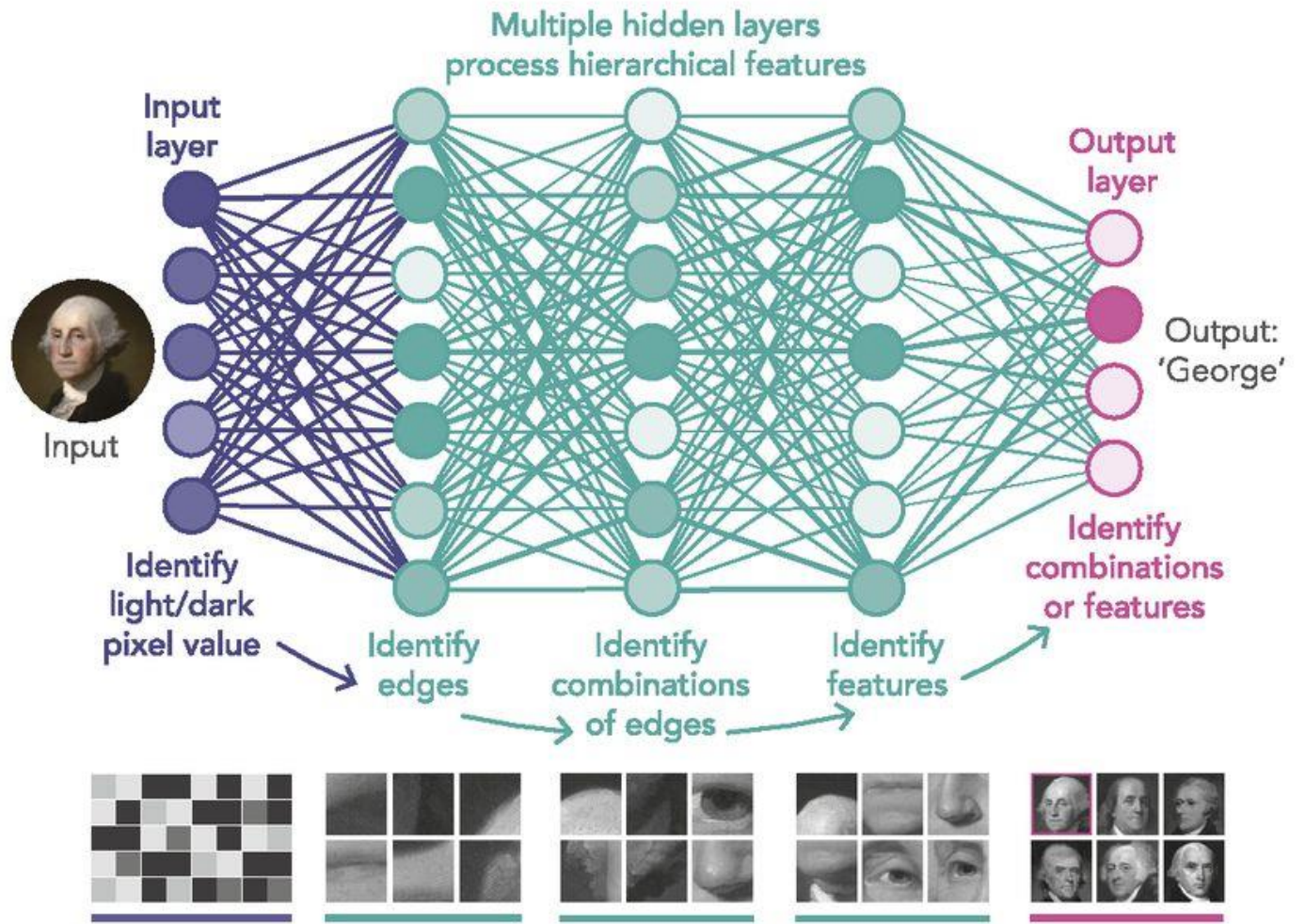
Üretici modeller birlikte gelme olasılıklarını hesaplar

### 1980S-ERA NEURAL NETWORK



Links carry signals from one node to another, boosting or damping them according to each link's 'weight'.

### DEEP LEARNING NEURAL NETWORK



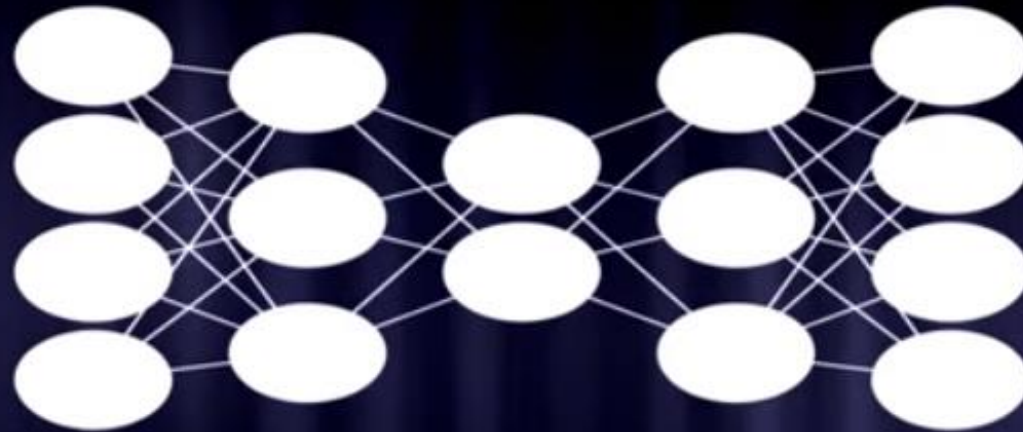
# GENERATIVE MODELS

- Autoencoders
- Adversarial Networks
- Sequence Models

# Autoencoders

ENCODE >

DECODE >



input

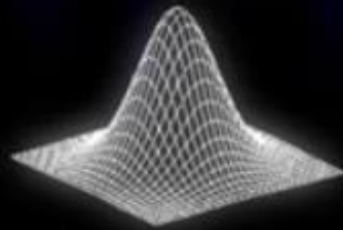
hidden

output

$$q_{\phi}(z|x)$$

$$p_{\theta}(x|z)$$

$x$



$\tilde{x}$

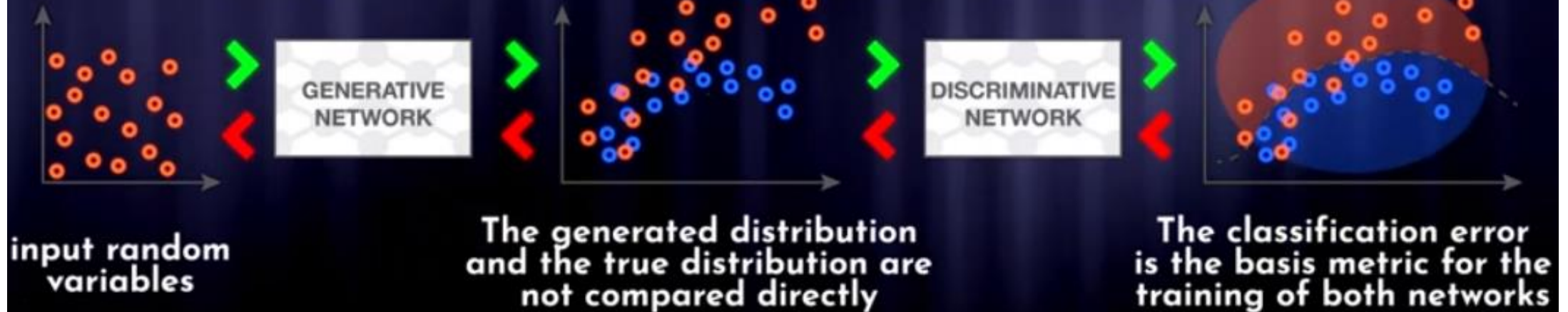
# Adversarial Networks

**forward propagation  
(generation and classification)**

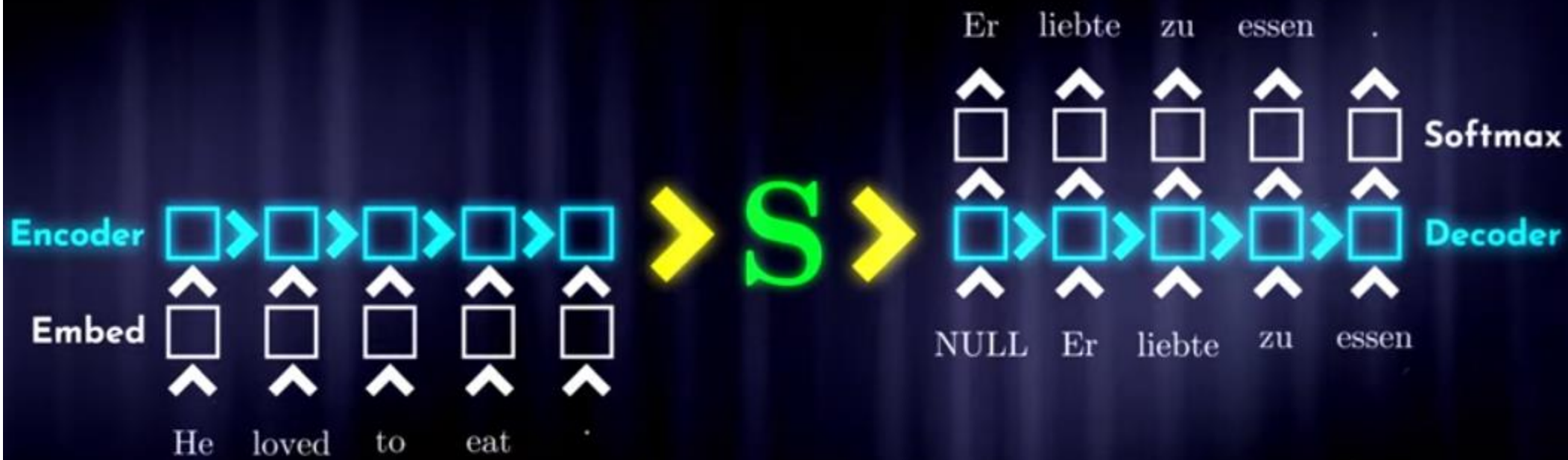
**backward propagation  
(adversarial training)**

The generative network is trained to maximise the final classification error

The discriminative network is trained to minimise the final classification error



# Sequence Models



# ChatGPT

- 1) Generative pretraining
- 2) Supervised fine-tuning
- 3) Reinforcement learning from human feedback



“

**I am learning about  
ChatGPT right now.**

”

Segmentation

Tokenization

Stop Words

Stemming

Named Entity Recognition

Speech Tagging

Lemmatization

0

1

2

3

4

**i**

**be**

**learn**

**chatgpt**

**now**

*pronoun*

*auxillary  
verb*

*verb*

*noun*

*adverb*

**Software Entity**

```
[  
  [1, 0, 0, 0],      # pronoun  
  [0, 1, 0, 0],      # auxiliary verb  
  [0, 0, 1, 0],      # verb  
  [0, 0, 0, 1],      # proper noun  
  [0, 0, 0, 1],      # proper noun  
  [0, 0, 0, 0],      # stop word  
  [0, 0, 1, 0],      # verb  
  [0, 0, 0, 0],      # stop word  
  [0, 0, 0, 1]       # noun  
]
```

**GPT**

# GENERATIVE PRETRAINED TRANSFORMER

I am learning about  
GPTChat right now

**ENCODER**

As a result, I now have a better understanding  
of how ChatGPT works and what it can do.

**DECODER**

```
[  
  [1, 0, 0, 0],  
  [0, 1, 0, 0],  
  [0, 0, 1, 0],  
  [0, 0, 0, 1],  
  [0, 0, 0, 1],  
  [0, 0, 0, 0],  
  [0, 0, 1, 0],  
  [0, 0, 0, 0],  
  [0, 0, 0, 1]  
]
```

Sequence to Sequence  
Transformation

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## Attention Is All You Need

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### Abstract

The dominant sequence transduction models are based on complex recurrent or convolutional neural networks that include an encoder and a decoder. The best performing models also connect the encoder and decoder through an attention mechanism. We propose a new simple network architecture, the Transformer, based solely on attention mechanisms, dispensing with recurrence and convolutions entirely. Experiments on two machine translation tasks show these models to be superior in quality while requiring less parallelism.

An orange, slightly tilted rectangular frame graphic on the left side of the slide, partially overlapping the text.

**INITIAL MODEL  
TRAINED WITH  
SUPERVISED  
FINE-TUNING**




**HUMAN WRITES PROMPT**

e.g. What is a car?

**HUMAN ANSWERS PROMPT**

e.g. A vehicle moving on wheels



# REWARD MODEL

Human writes prompt.  
e.g. *What is a car?*

**A**

a vehicle  
moving on  
wheels

**0.98**

**B**

a Disney  
movie

**0.43**

**C**

something  
similar to  
a bicycle

**0.56**

**D**

a tool  
typically  
found in the  
bathroom

**0.13**

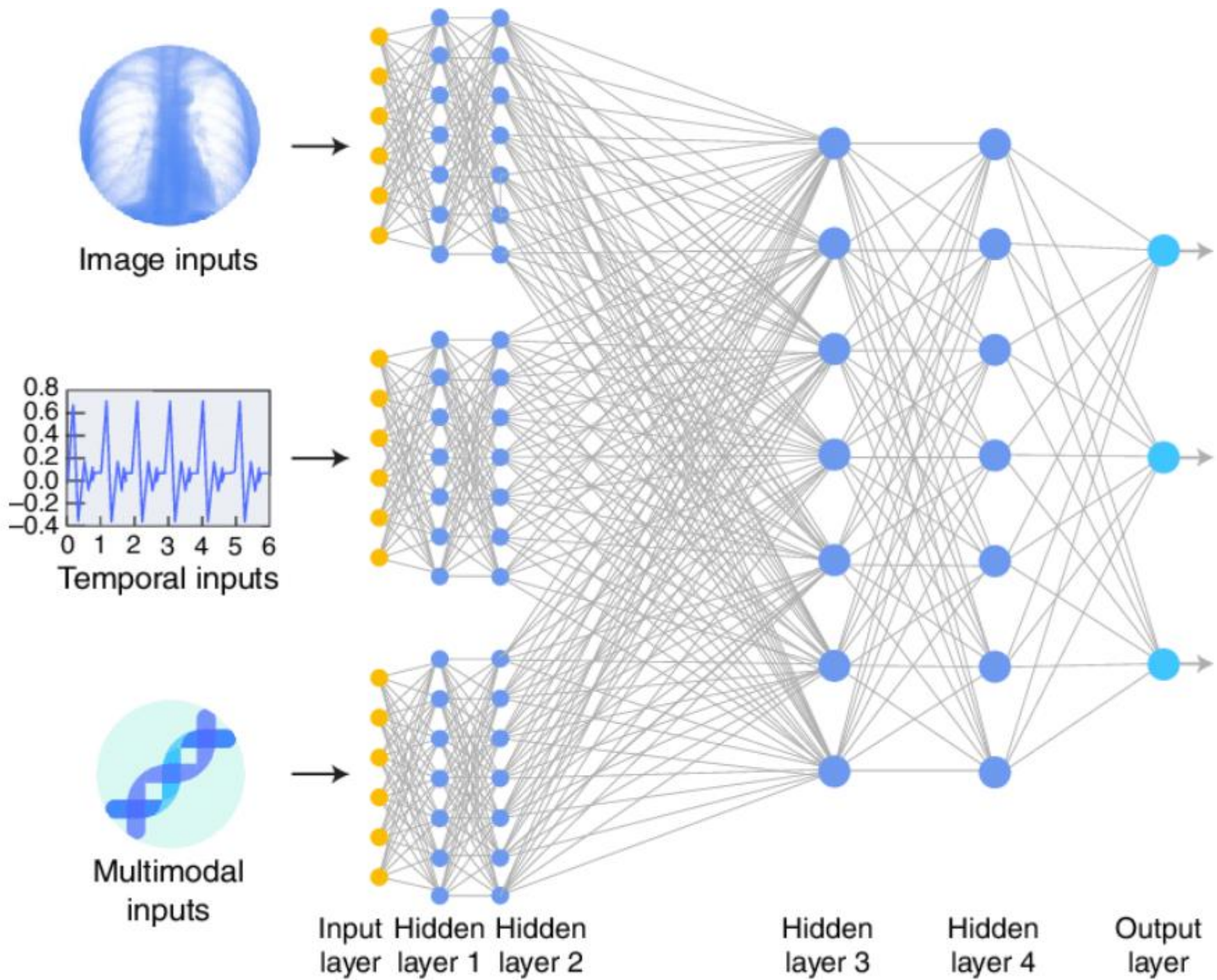
**A > C > B > D**

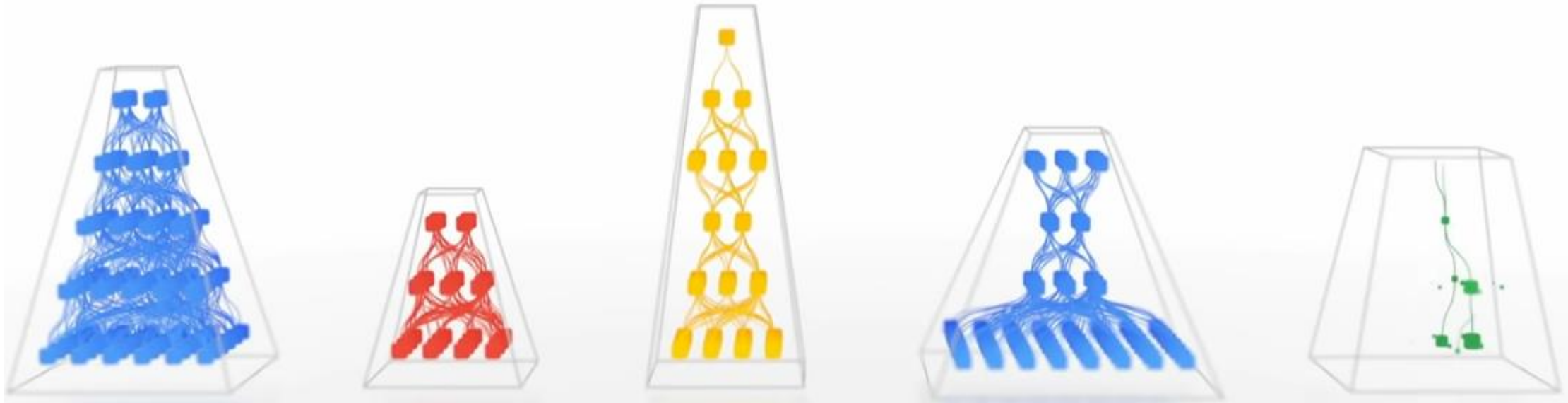




# OPTIMIZING A POLICY AGAINST THE REWARD MODEL

PPO: Proximal Policy Optimizer





Separate models for thousands of tasks

# *How Cheap Labor Drives China's A.I. Ambitions*

Give this article



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Workers at the headquarters of Ruijin Technology Company in Jiaxian, in central China's Henan Province. They identify objects in images to help artificial intelligence make sense of the world. Yan Cong for The New York Times

# Optical Adversarial Attack Can Change the Meaning of Road Signs



Published 1 year ago on August 16, 2021

By **Martin Anderson**





**"panda"**

57.7% confidence

+  $\epsilon$



=



**"gibbon"**

99.3% confidence

An image of a panda, when combined with an adversarial input, can convince a classifier that it's looking at a gibbon. IMAGE: OPENAI

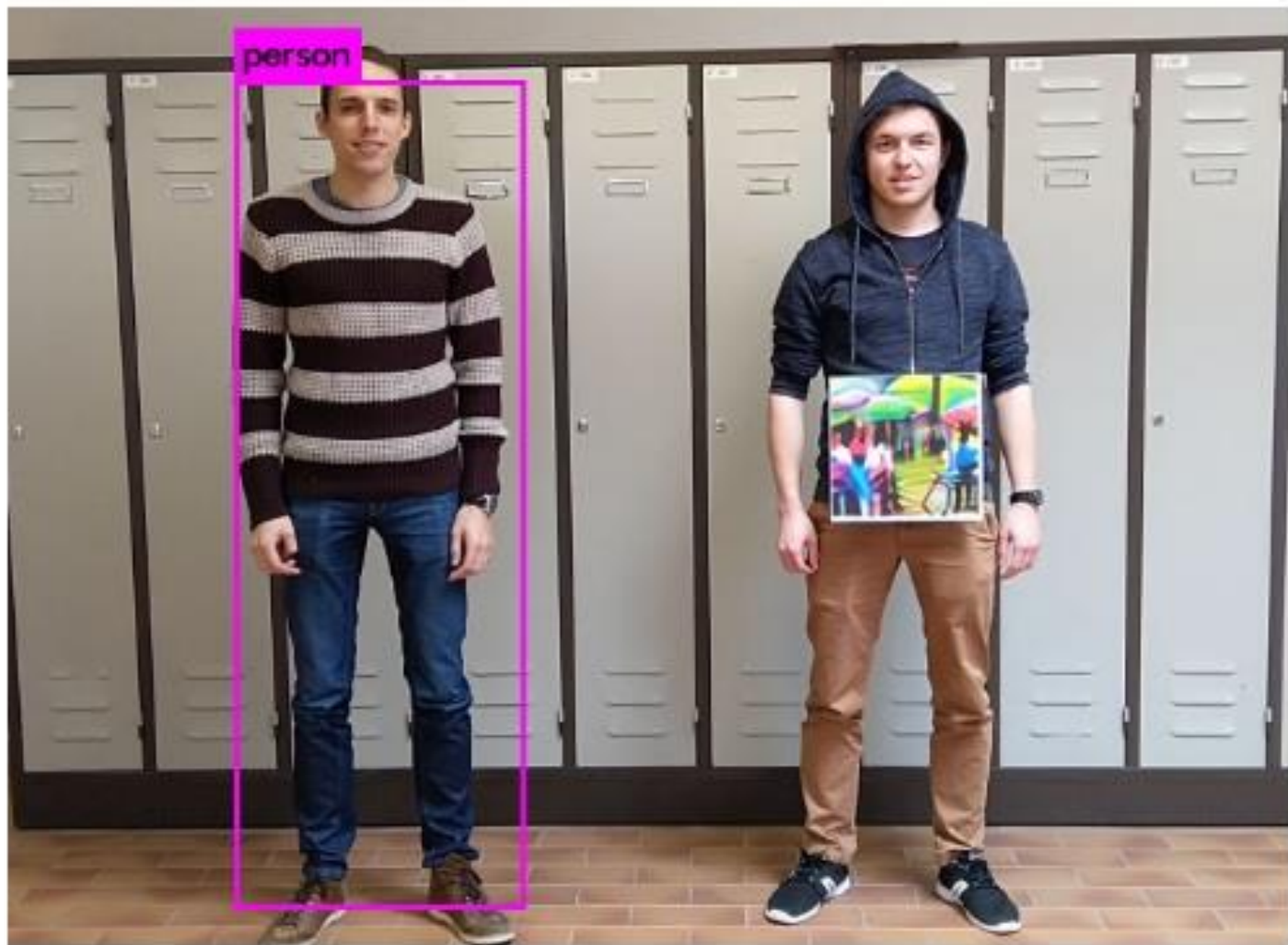


Figure 1: We create an adversarial patch that is successfully able to hide persons from a person detector. Left: The person without a patch is successfully detected. Right: The person holding the patch is ignored.