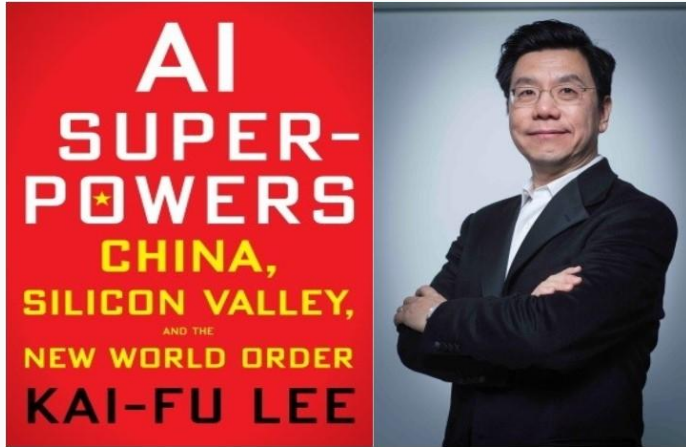


How Neural Networks See the World

Building Intuition for a Broader AI Discussion



Book was great!



Broader AI conversation.

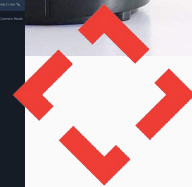
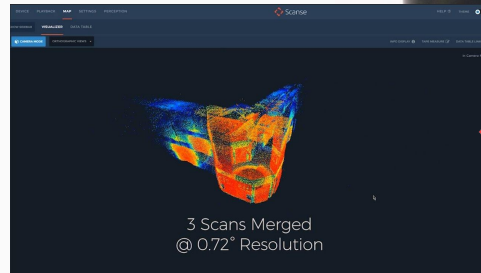
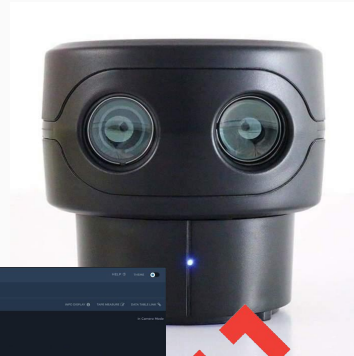
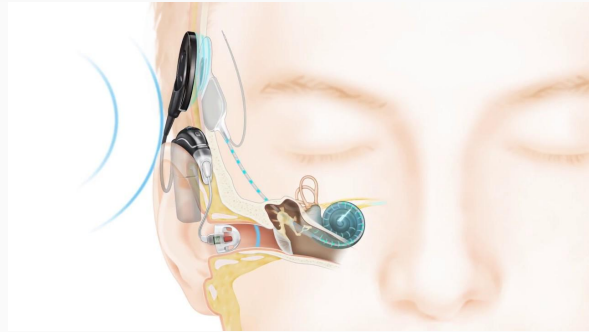
But how does all this “AI” stuff actually work?

What is deep learning?

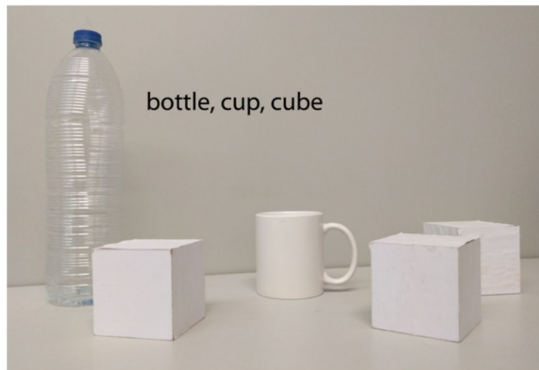
How do neural networks learn?

Engineering Perception

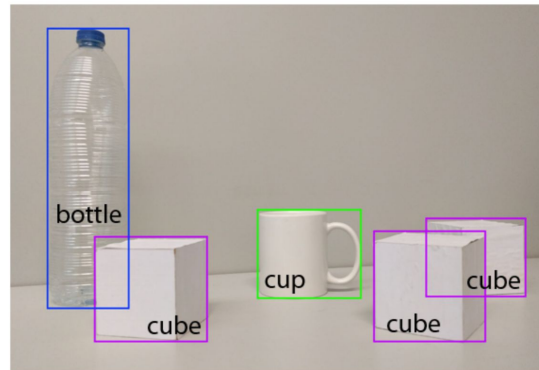
I offer a developer perspective



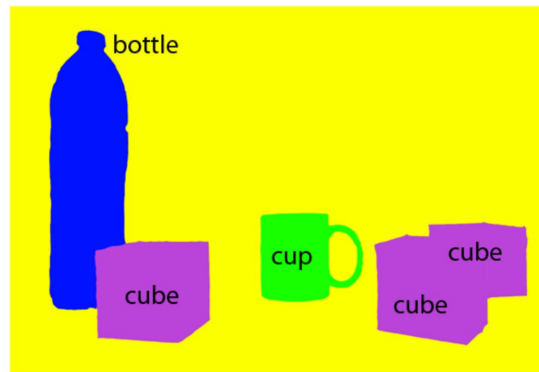
RIPCORD



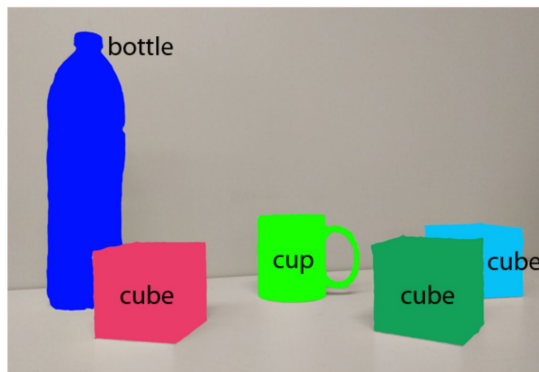
(a) Image classification



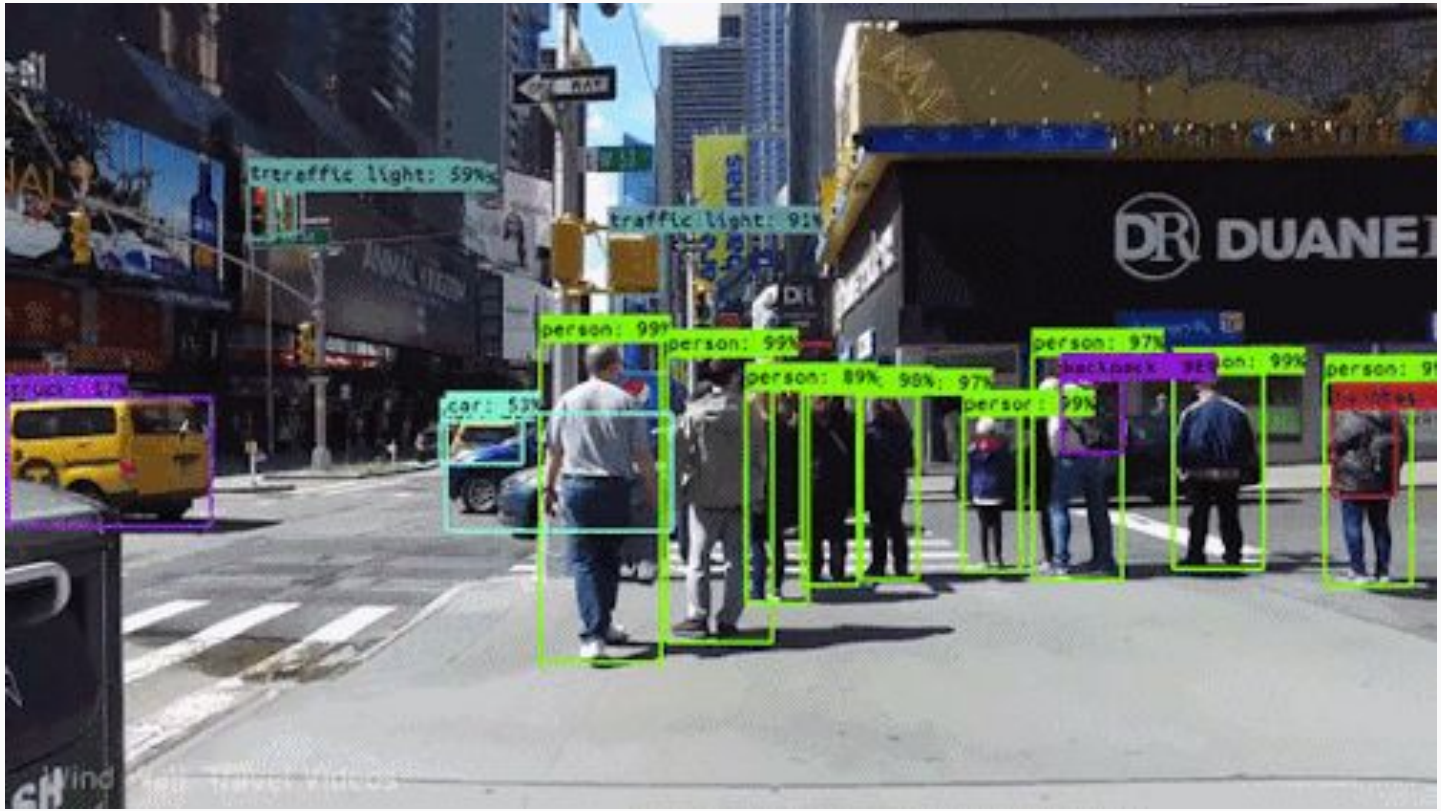
(b) Object localization

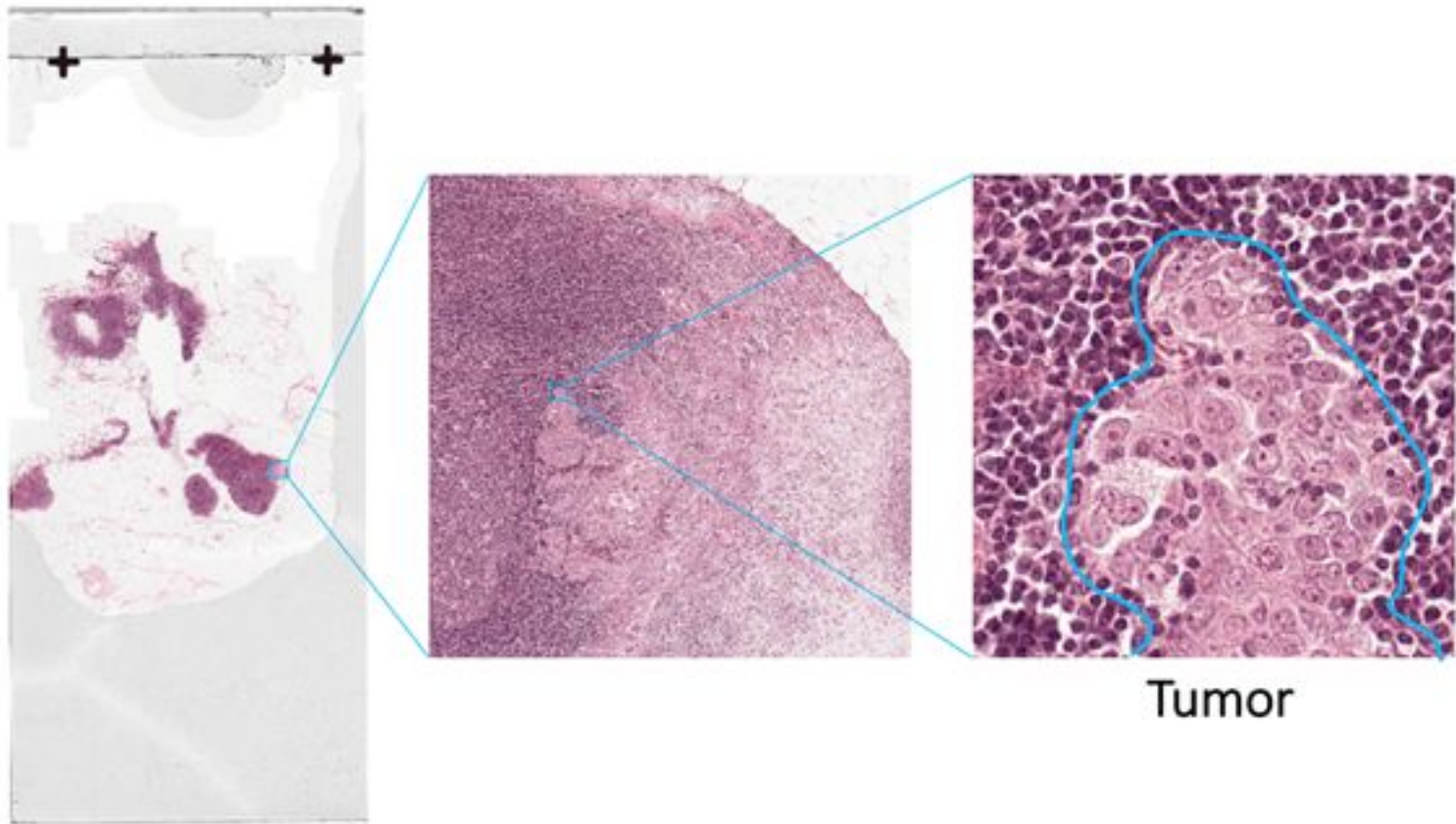


(c) Semantic segmentation



(d) Instance segmentation

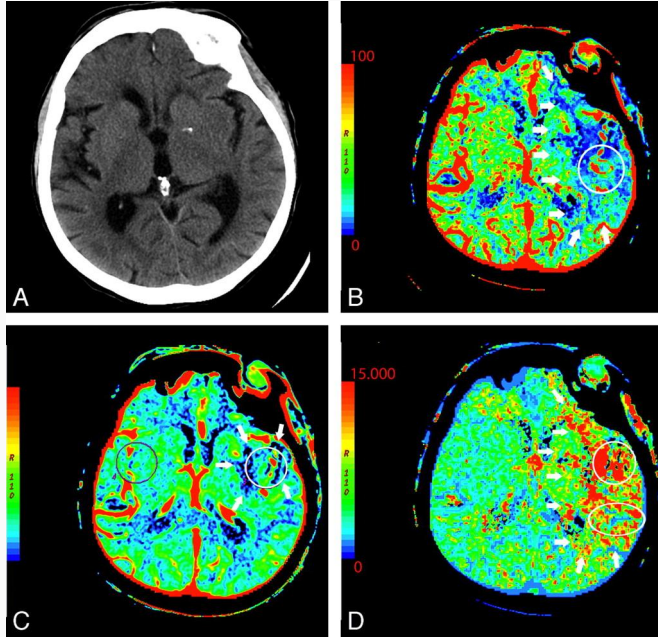




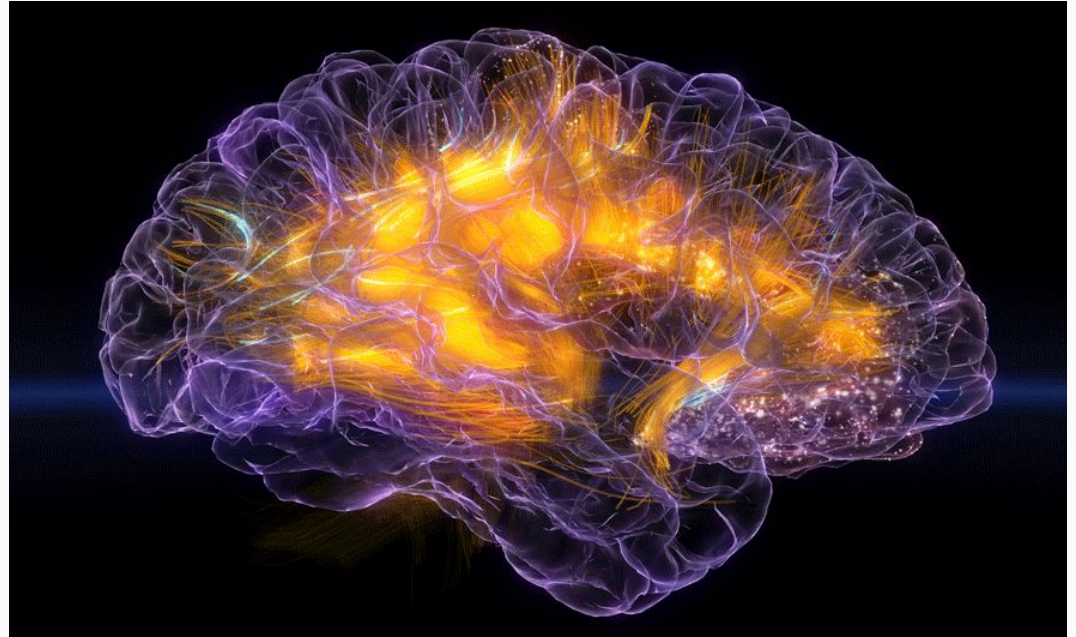
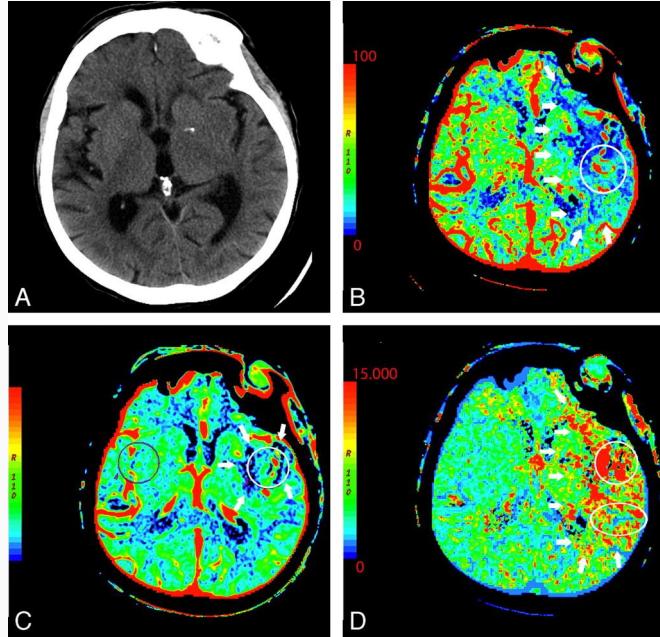
Tumor

How do **we (humans)**
see the world?

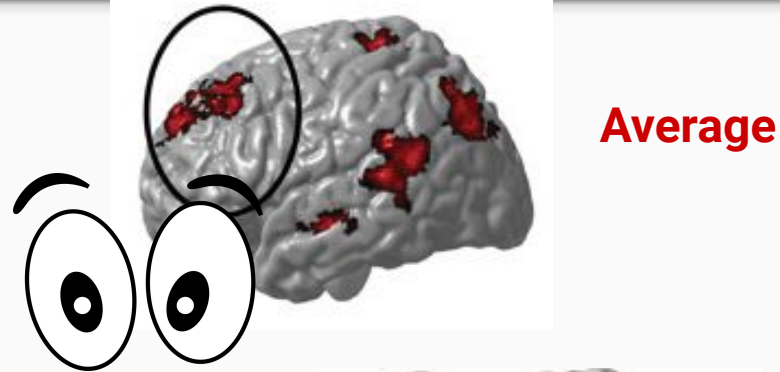
Our brains get **excited**



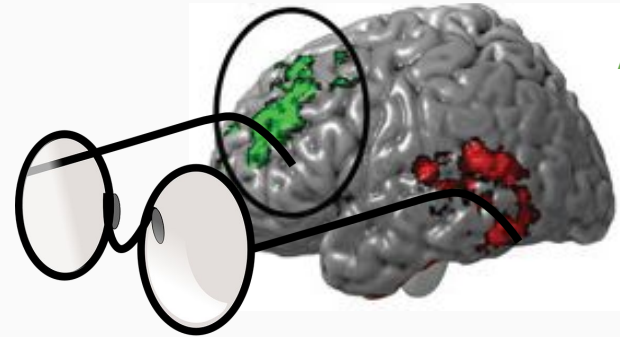
Our brains get **excited**



Experience leverages **different pathways**



Average



Attorney

Question:

What does it mean to **understand** something?

What does it mean to **learn** something?

Neural **pathways** are prepared and ready.

Reinforce, recruit, adapt, or create new neural pathways.

Question:

What does it mean to **understand** something?

What does it mean to **learn** something?

You are prepared to interpret **new examples**.

Studying or observing more examples, in preparation.

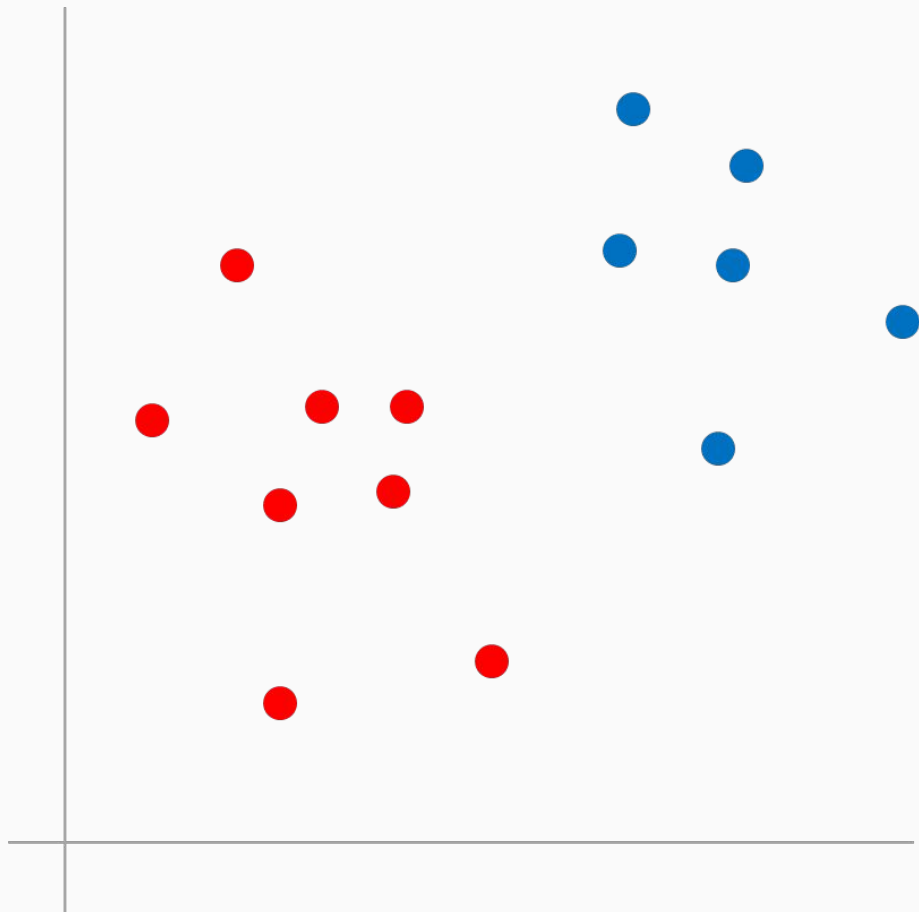
In deep learning,
neural networks **study existing data**
in order to interpret new data.

Classification is a simple way to interpret

- Managing variability
- Recognizing patterns
- This vs. That
- Concise communication about complex groups

Task:

Describe this
simple data



Rules are concise

RULE #1:

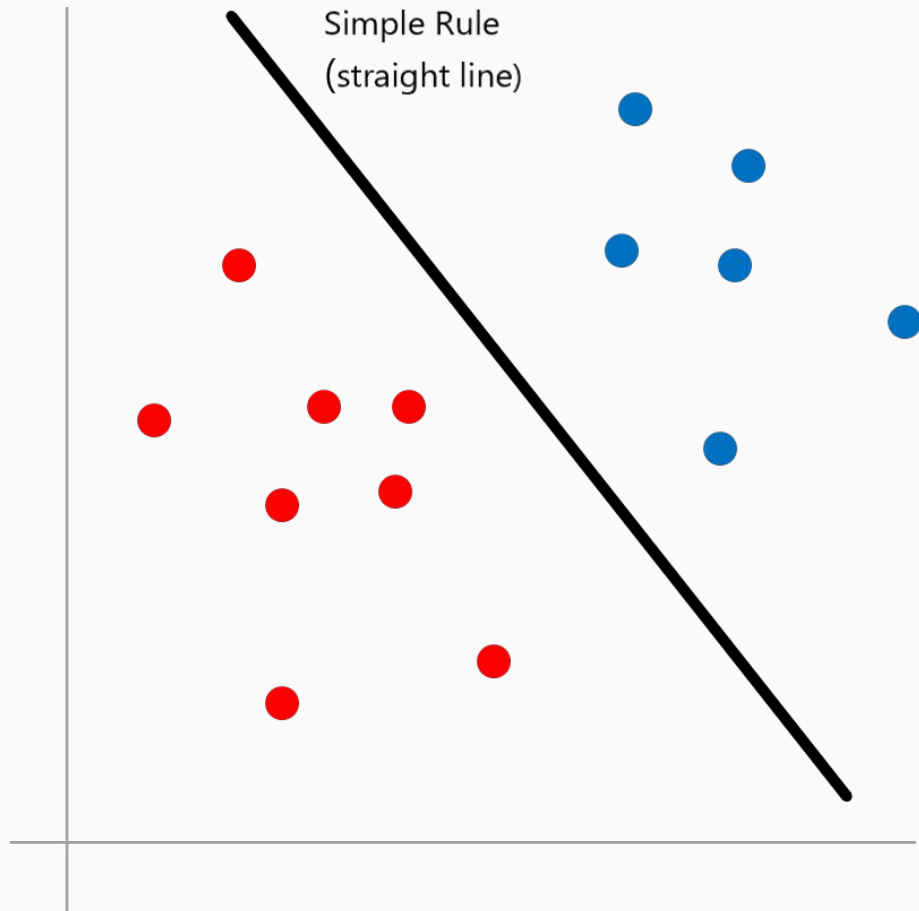
Points ABOVE the line are BLUE

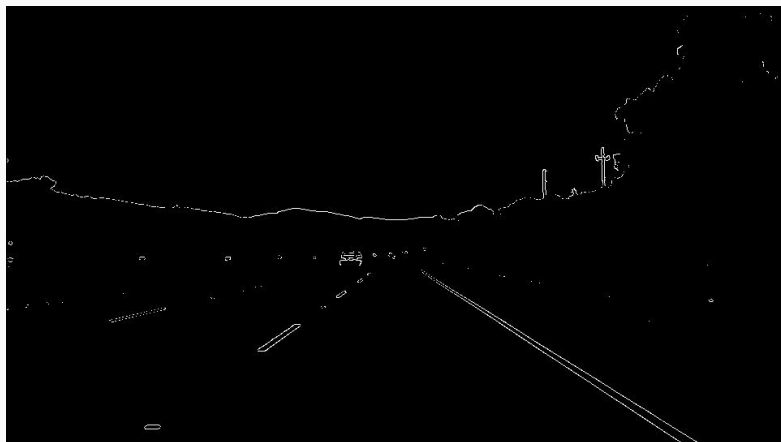
RULE #2:

Points BELOW the line are RED

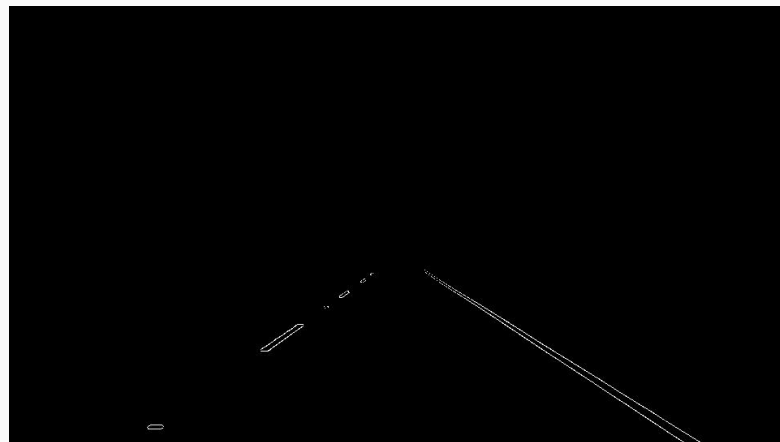
Leverage math to formally describe the rule:

$$y = mx + b$$

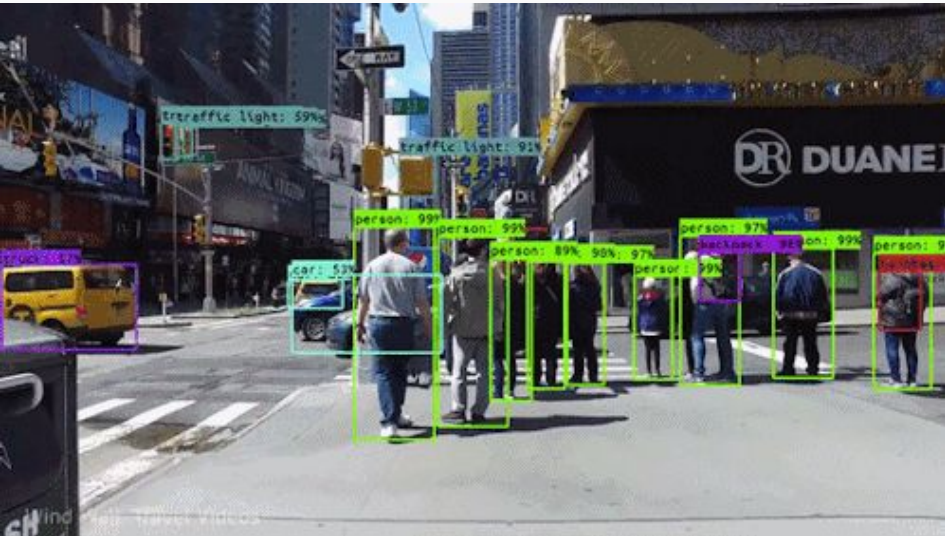




Rule: Lane Lines are WHITE

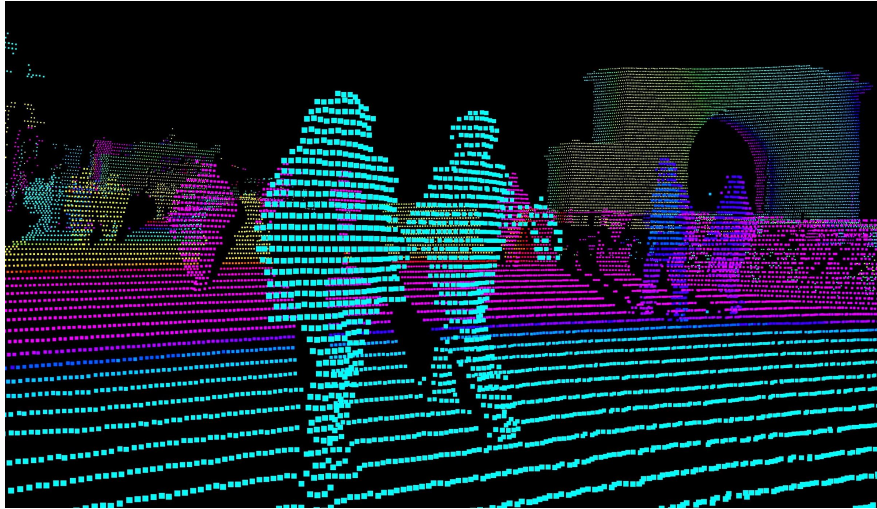
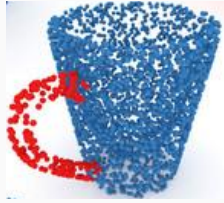
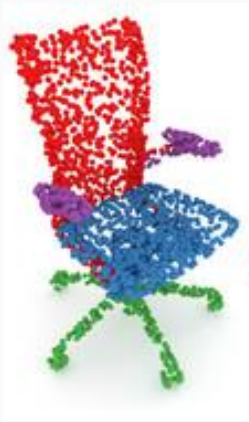


Rule: Road is below horizon



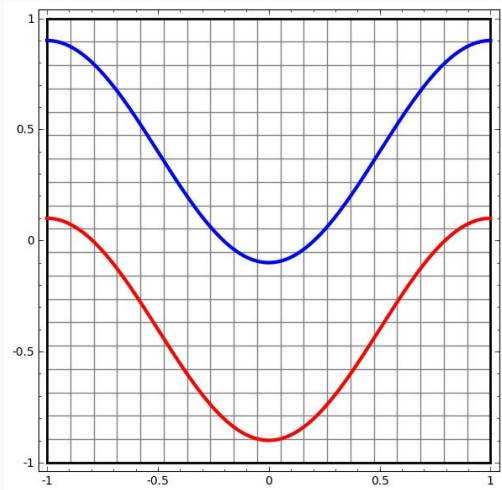
But real data is messy,
and writing complex
rules is hard.

Deep Learning makes complex
discrimination possible.

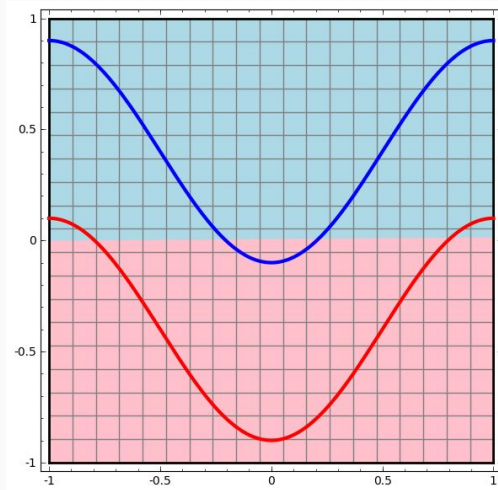


But real data is messy,
and writing complex
rules is hard.

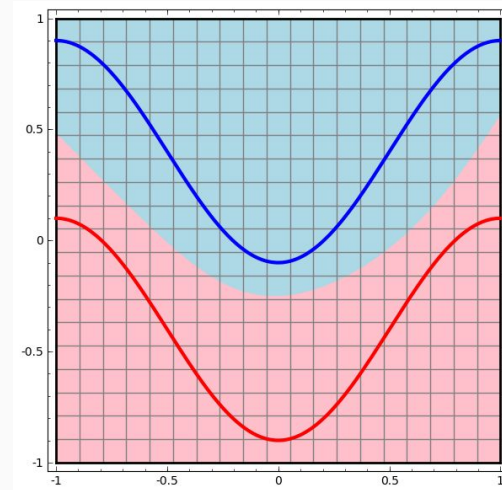
Neural networks **appear** to learn complex rules.



Complex Data



Simple Rule
(inadequate)

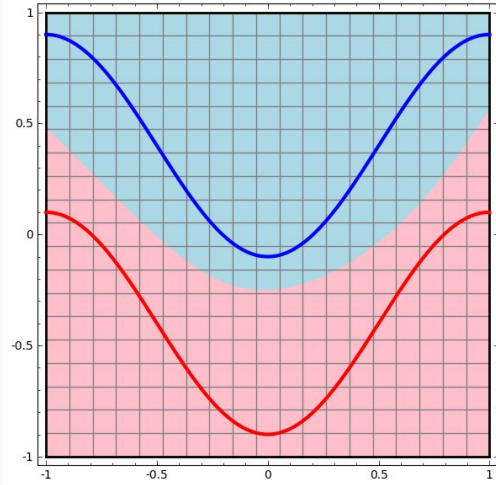


Complex Rule

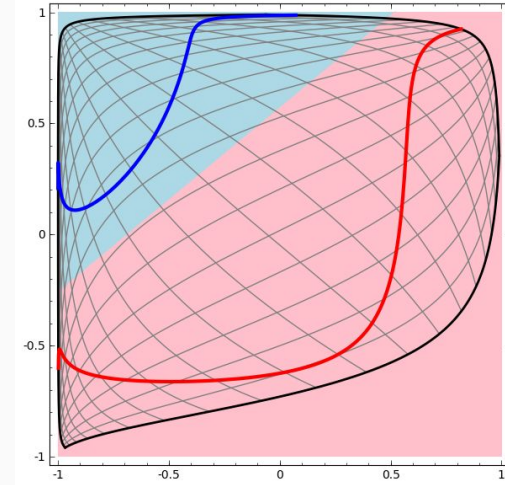
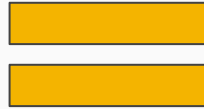
Actually:

Neural networks
apply simple rules
in a complex way.

To apply simple rules, neural networks distort space.



Complex Rule

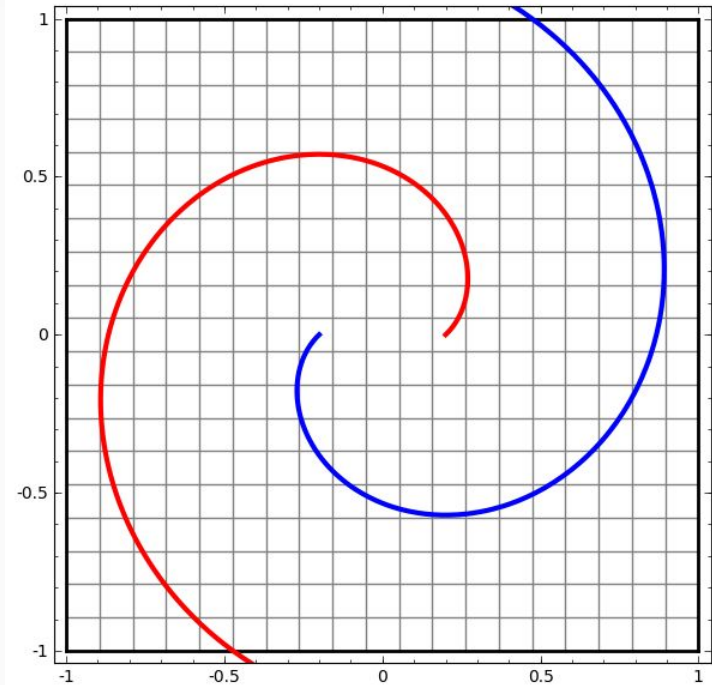
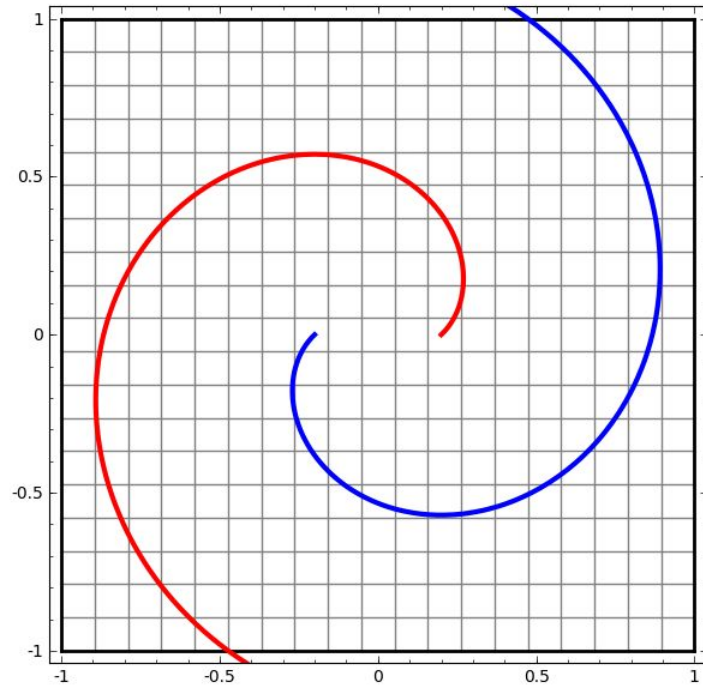


Simple Rule

Change Your Perspective



In order to understand data, neural networks distort how it is presented.

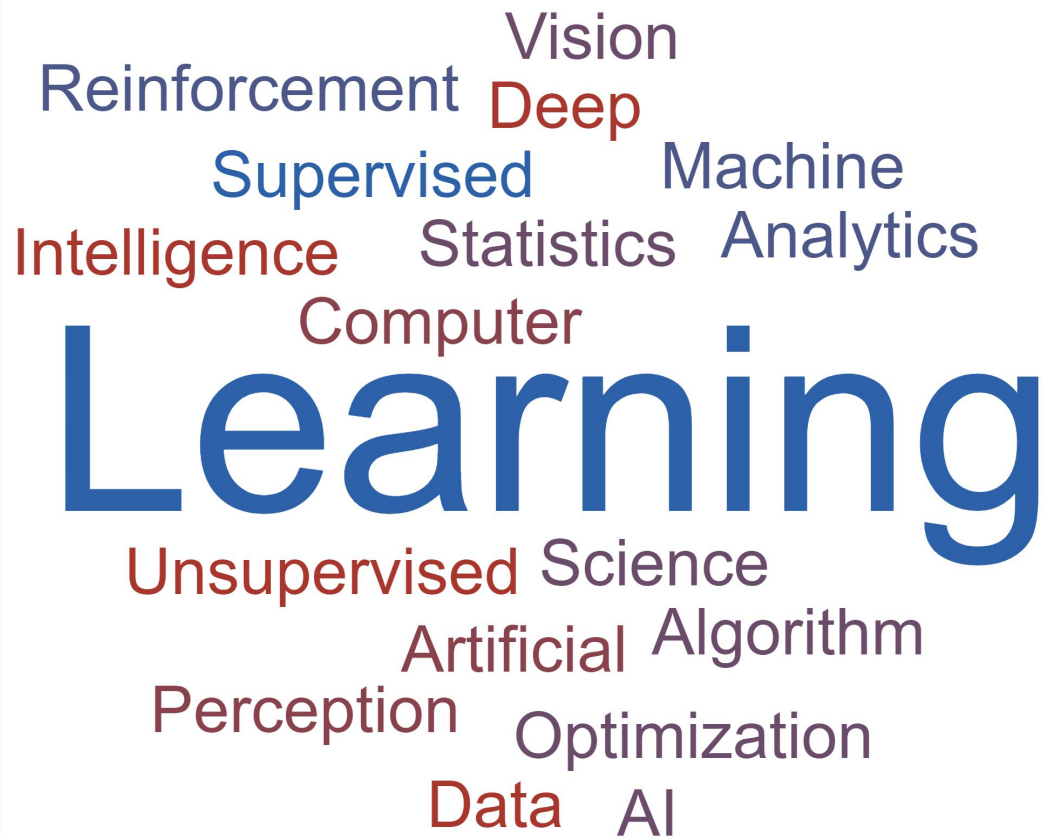


Takeaway:

An experienced neural network is prepared to interpret new examples.

AI landscape is filled with jargon.

Just know **learning** is fundamentally unique.



A word cloud centered around the word "Learning". The word "Learning" is the largest and most prominent, rendered in a large, bold, blue font. Surrounding it are various related terms in smaller fonts, some in blue and some in red. The terms include: "Reinforcement", "Vision", "Deep", "Supervised", "Machine", "Intelligence", "Statistics", "Analytics", "Computer", "Unsupervised", "Science", "Artificial", "Algorithm", "Perception", "Optimization", "Data", and "AI". The words are arranged in a somewhat circular pattern around the central "Learning" word.

Reinforcement Vision
Deep
Supervised Machine
Intelligence Statistics Analytics
Computer
Learning
Unsupervised Science
Artificial Algorithm
Perception Optimization
Data AI

Unlike traditional problem solving, learning is an indirect solution.

With **ALGORITHMS**, behavior is designed.

Algorithms are:

- Written instructions
- Rule driven
- Provable
- Guaranteed Behavior

With **Learning**, environment and curriculum are designed. Behavior emerges indirectly.

Unlike traditional problem solving, learning is an indirect solution.

With **ALGORITHMS**, behavior is designed.

Algorithms are:

- Written
- Rule driven
- Provable
- Guaranteed Behavior

Stop
Defining Rules

With **Learning**, environment and curriculum are designed. Behavior emerges indirectly.

Start
Learning Rules

Neural Networks are like students, that learn in familiar terms.

Neural networks study data.

Often under supervision that demands memorization.

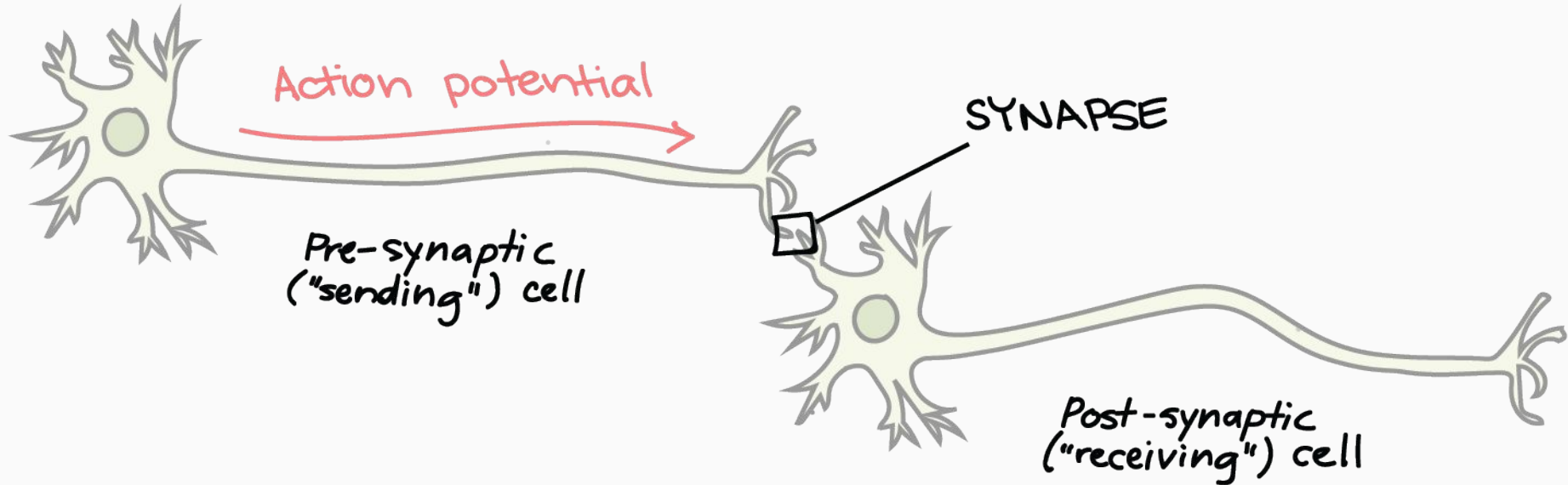
Developers are teachers, creating classroom environment to facilitate learning.



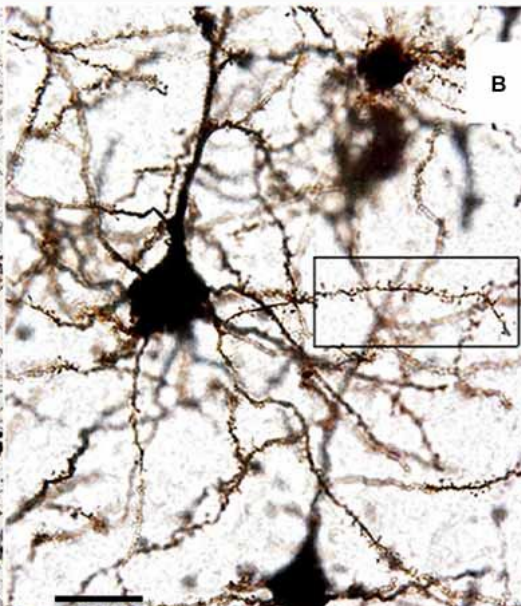
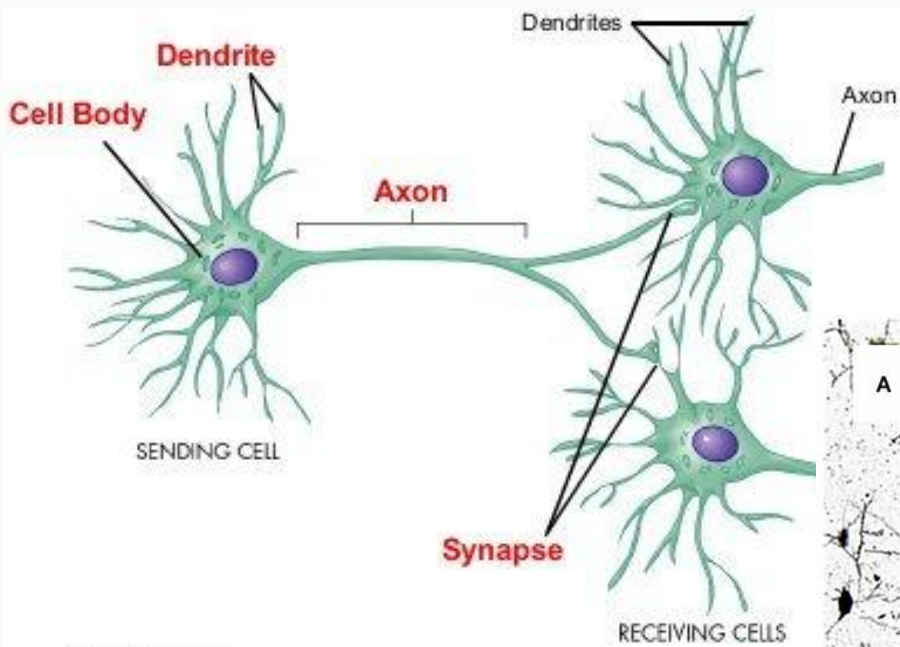
Question:

What is a network?

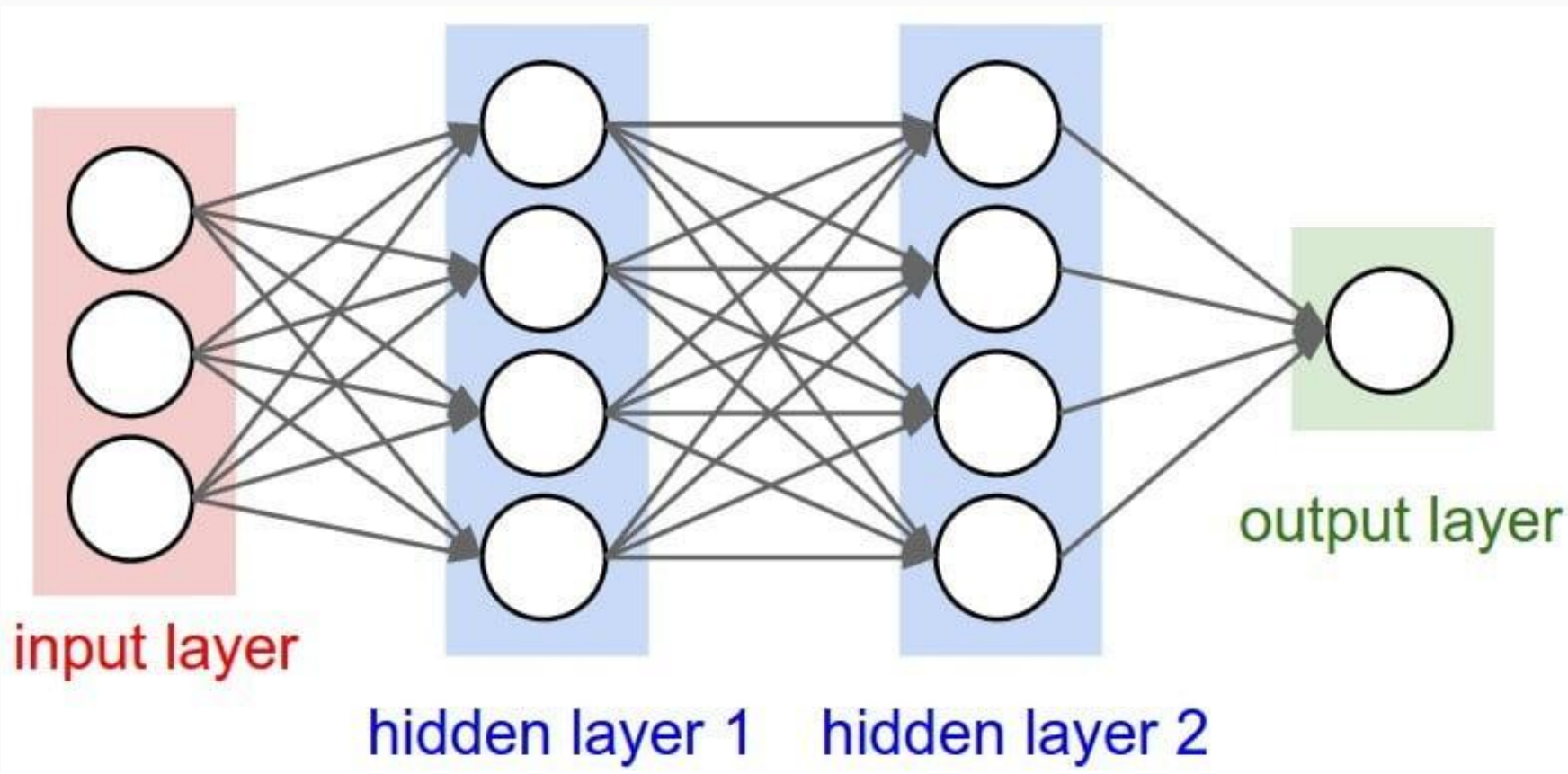
Connections are the building blocks of networks



Networks are comprised of many tangled connections

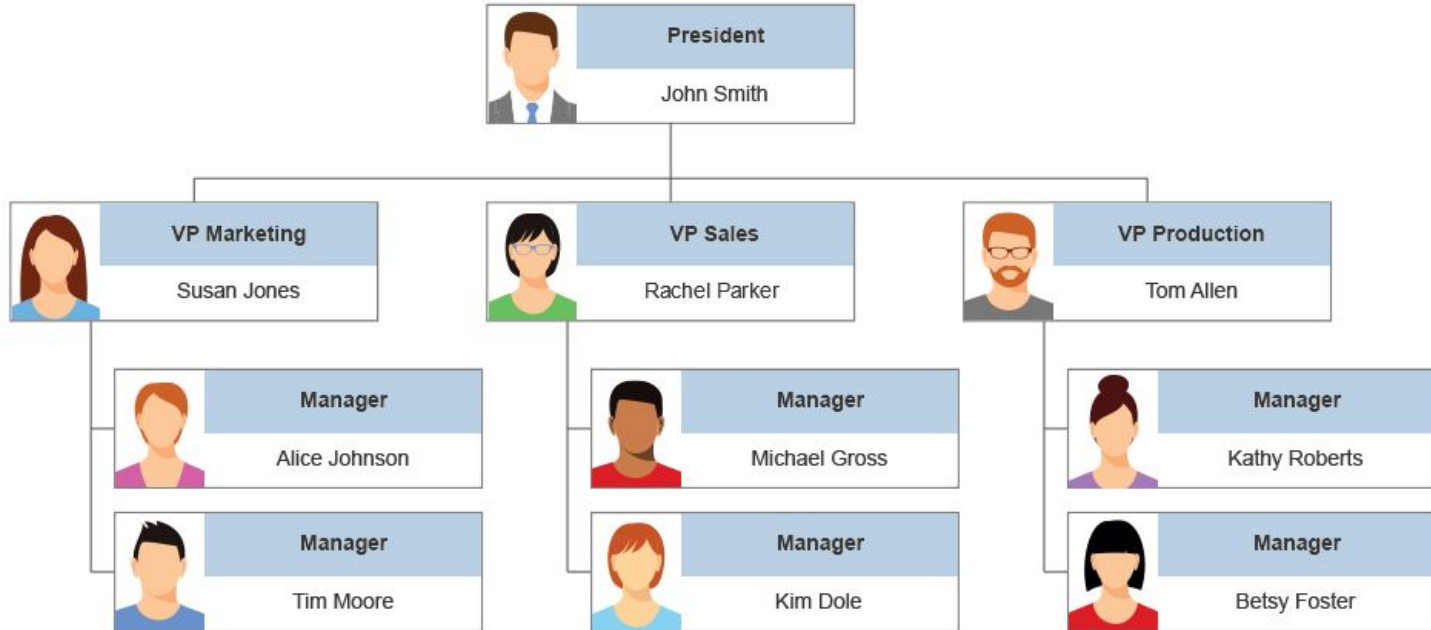


From the outside, networks have **inputs** and **outputs**

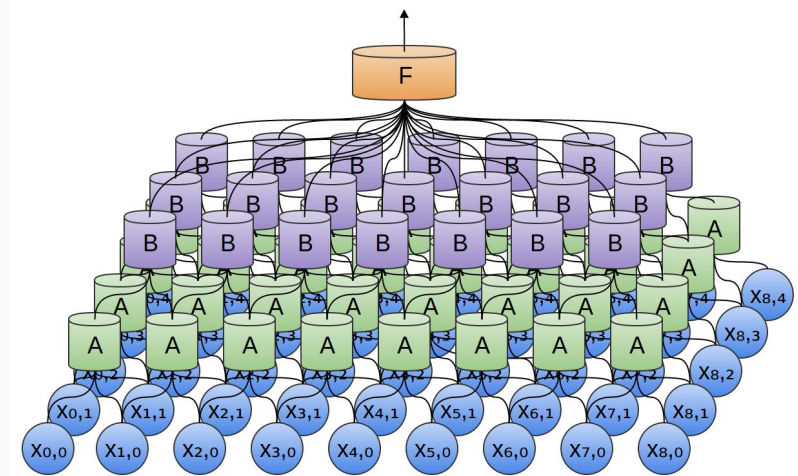
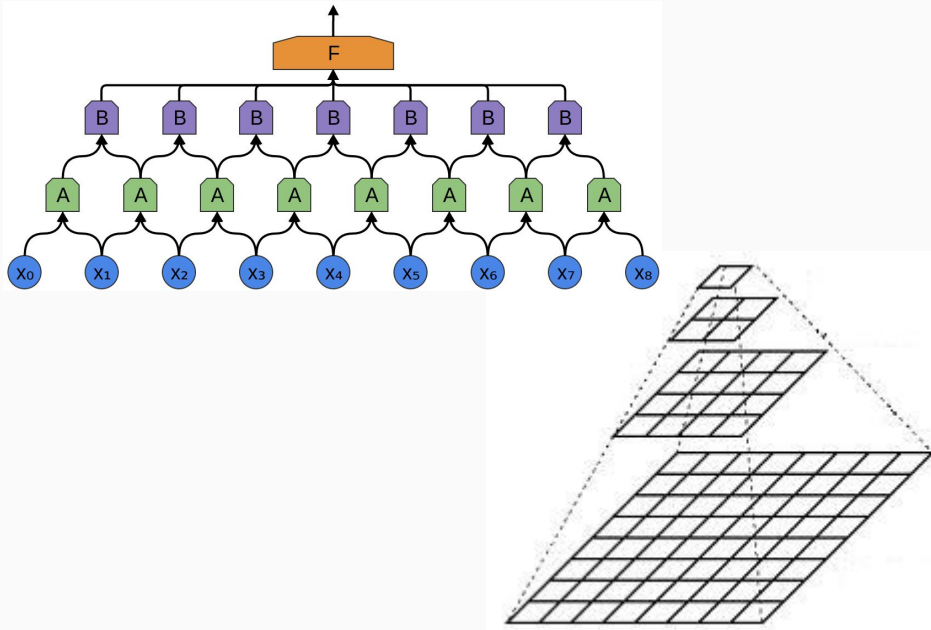


You can design
network **architectures**
for pattern recognition

Seeing the bigger picture requires abstraction and distance.



Sensing pyramids abstract details



Example: Classify handwritten digits



0
1
2
3
4
5
6
7
8
9



0
1
2
3
4
5
6
7
8
9

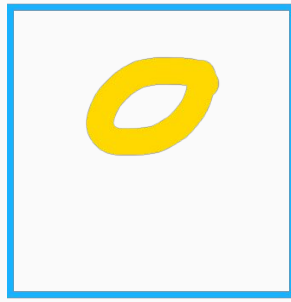


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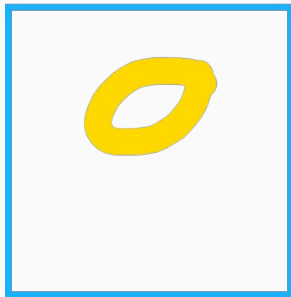


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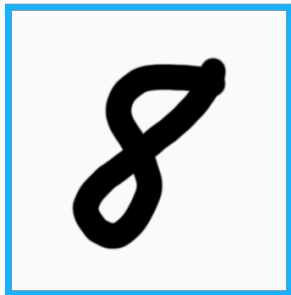




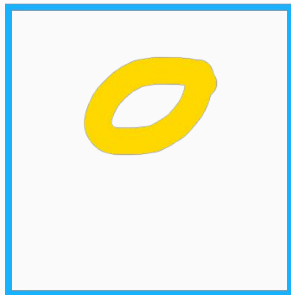
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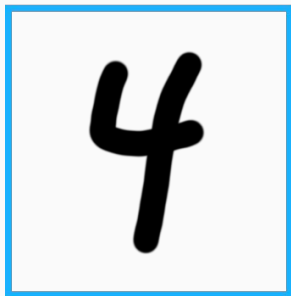
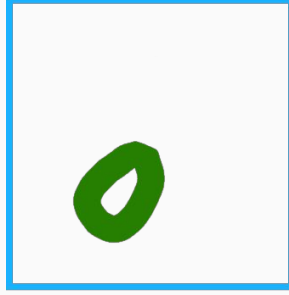
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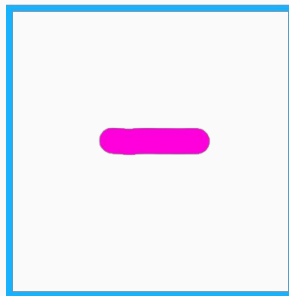
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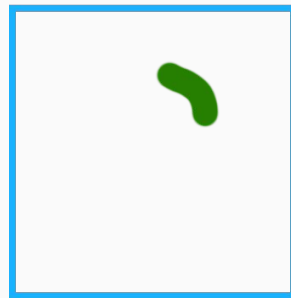
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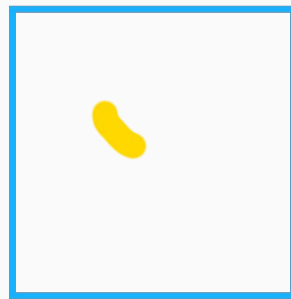
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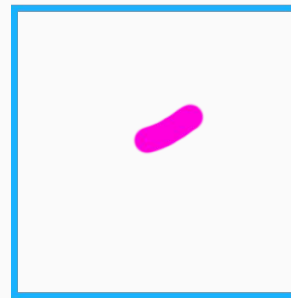
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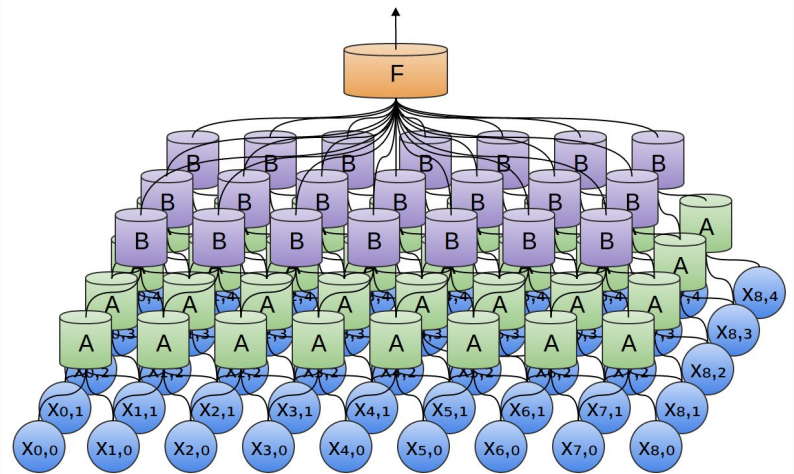
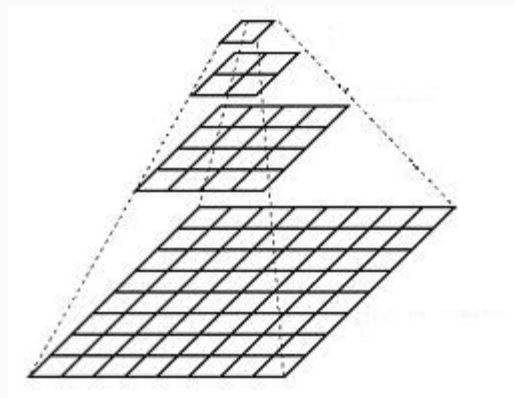
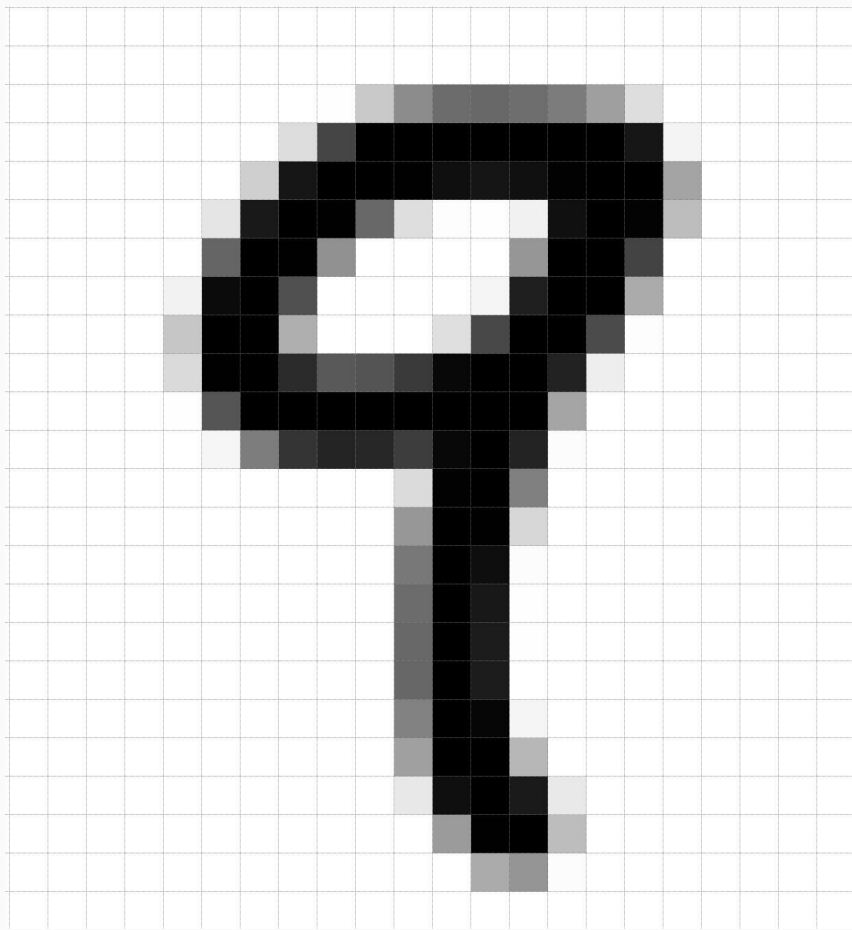


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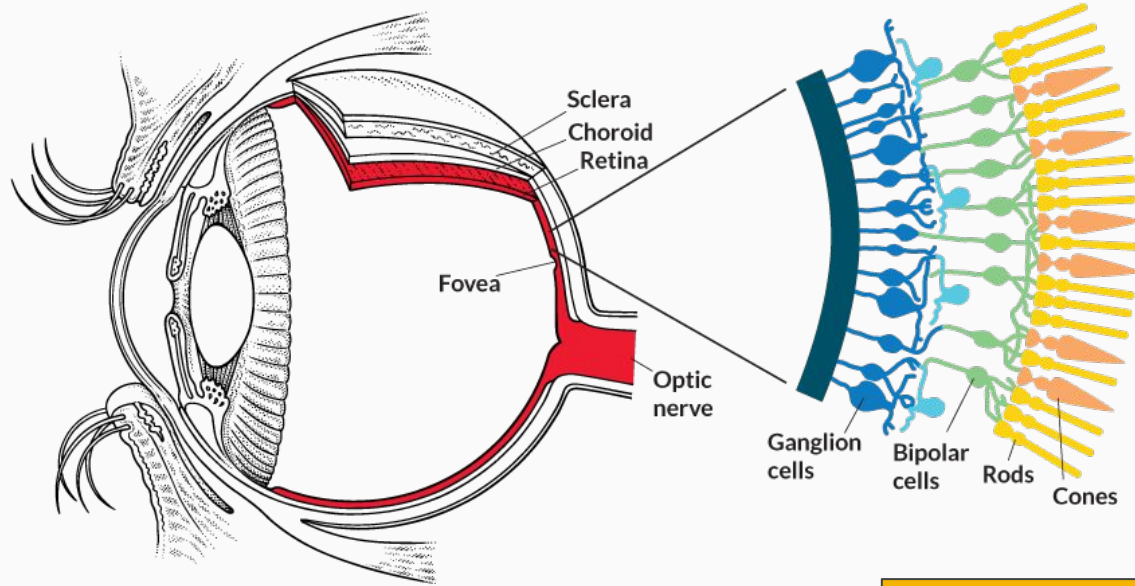
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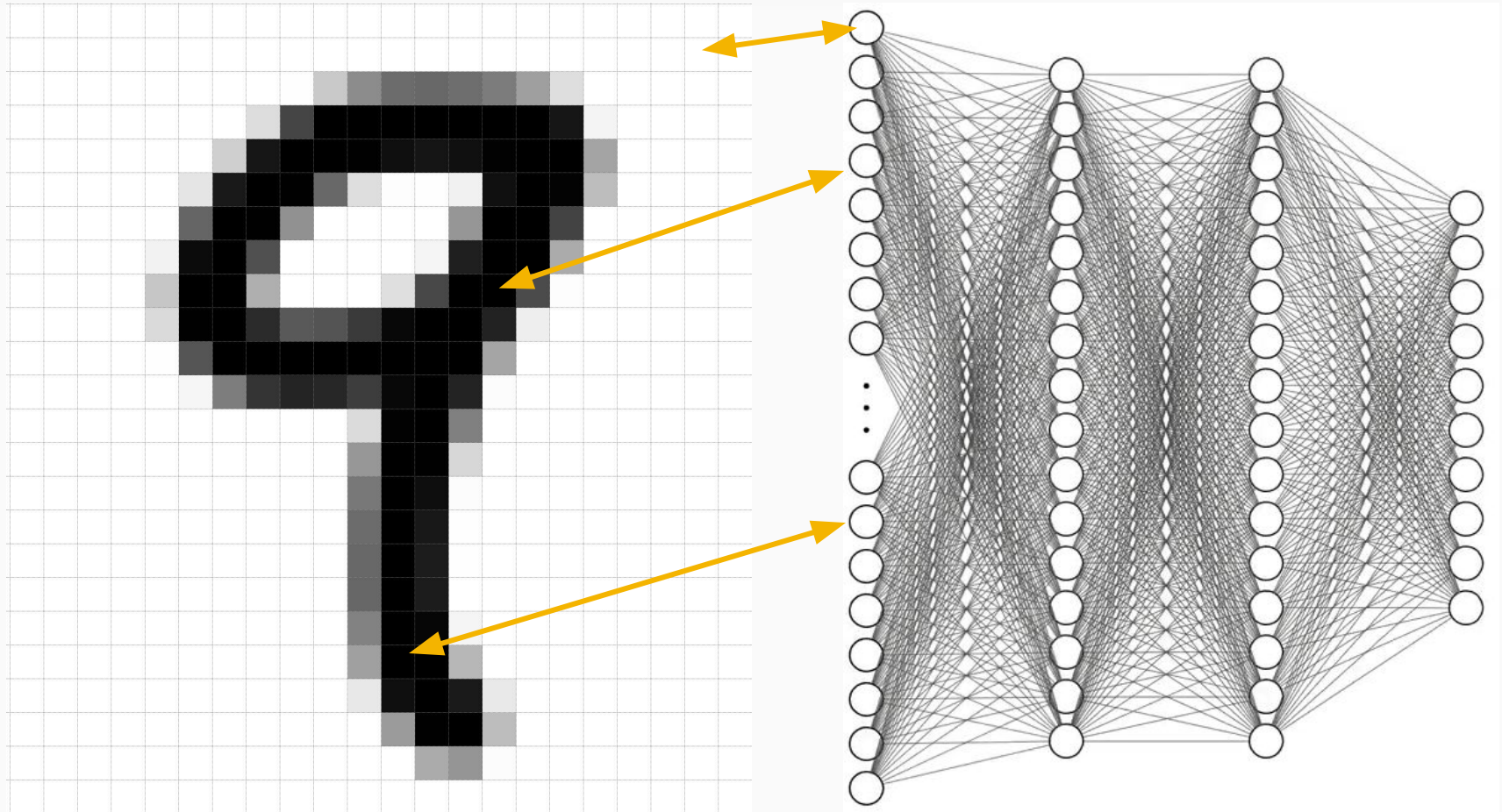


We limit detail at a certain level (pixels)

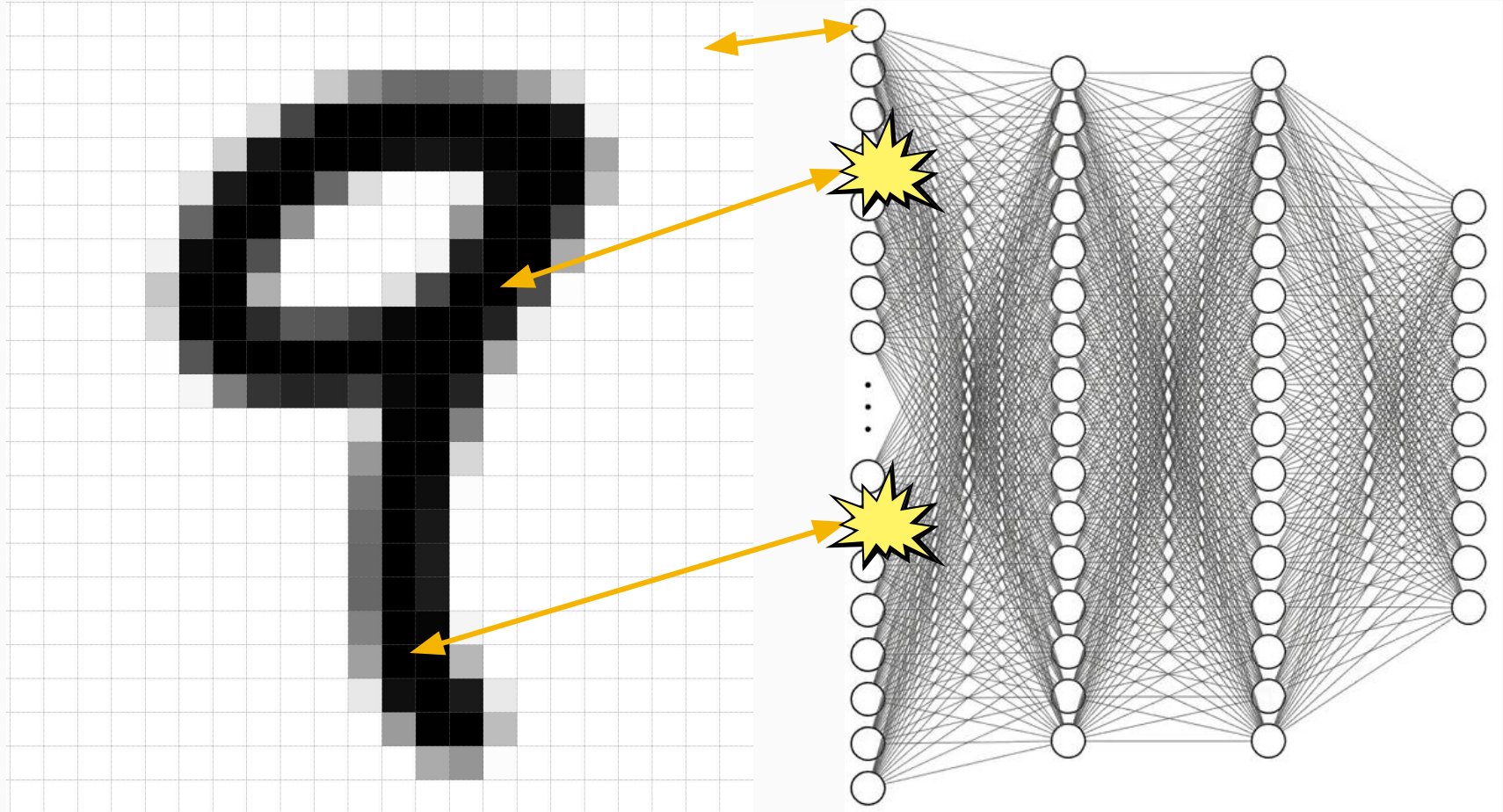
Human perception is **discrete**



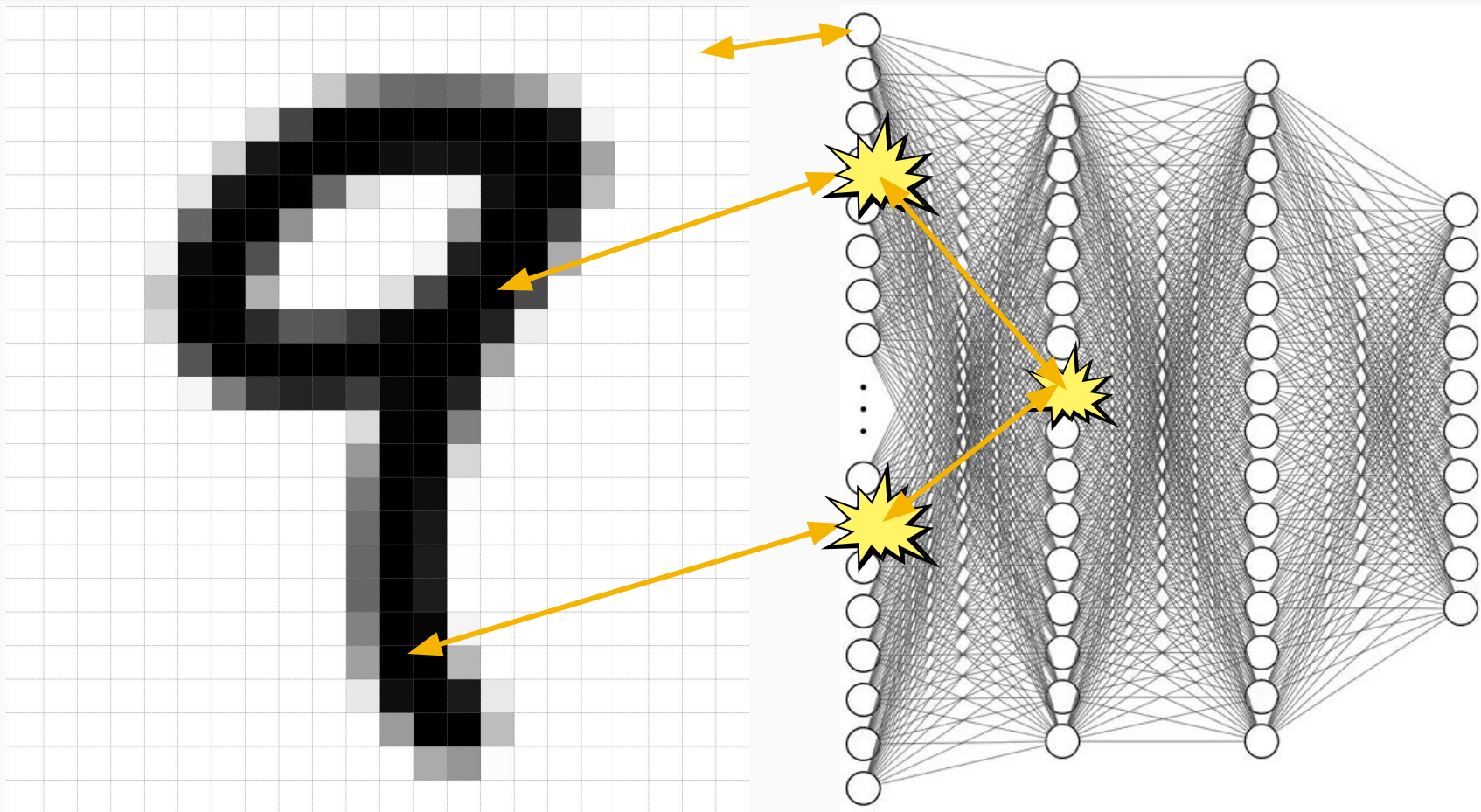
At one end, a network must **connect to data**

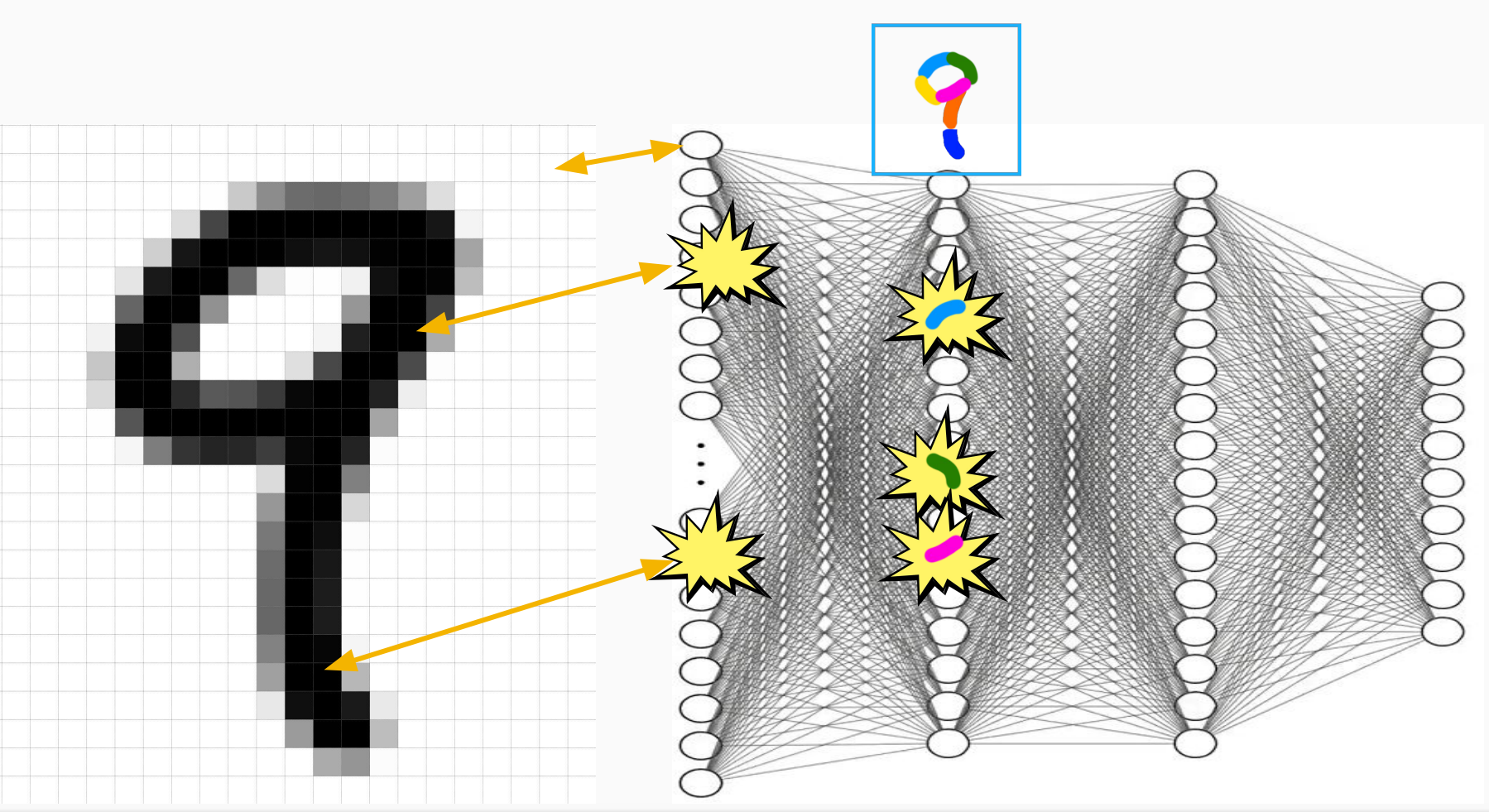


Different inputs will **excite** different pathways in the network

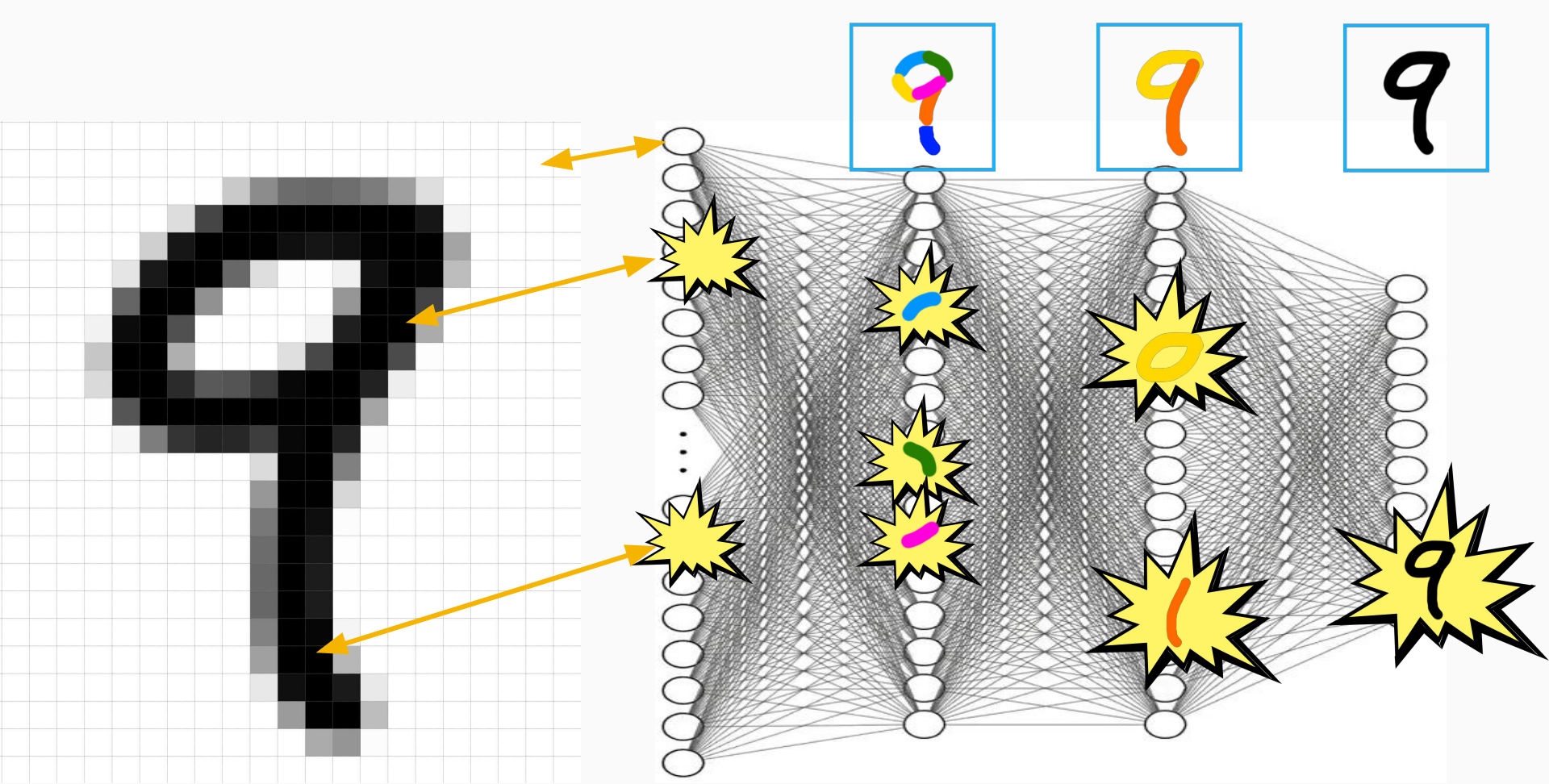


Different inputs will **excite** different pathways in the network



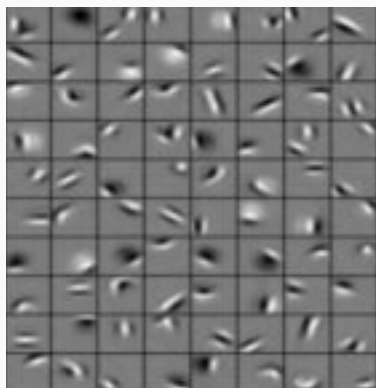
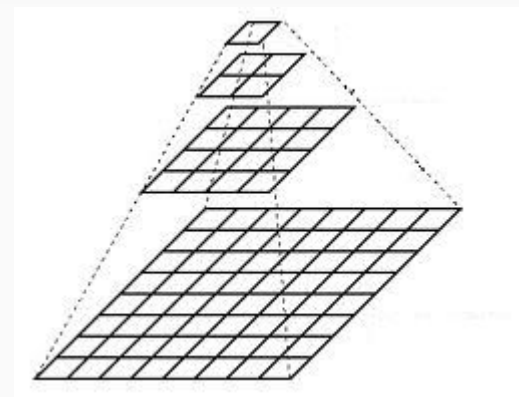
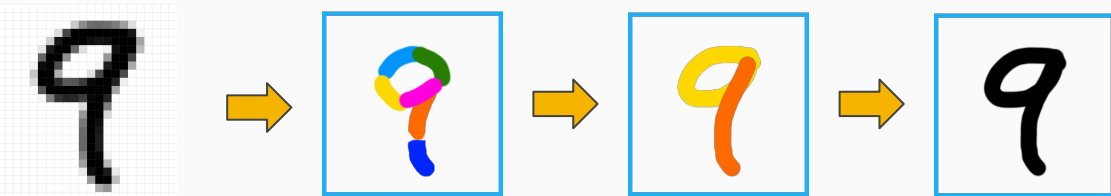


Networks consider granular details, in order to recognize **larger patterns**.



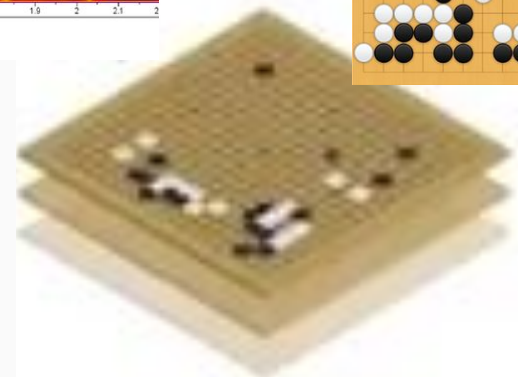
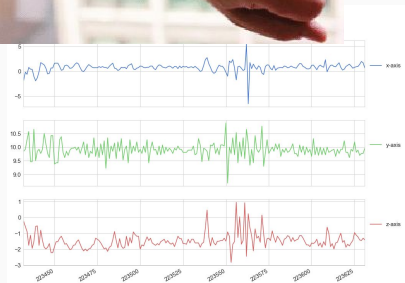
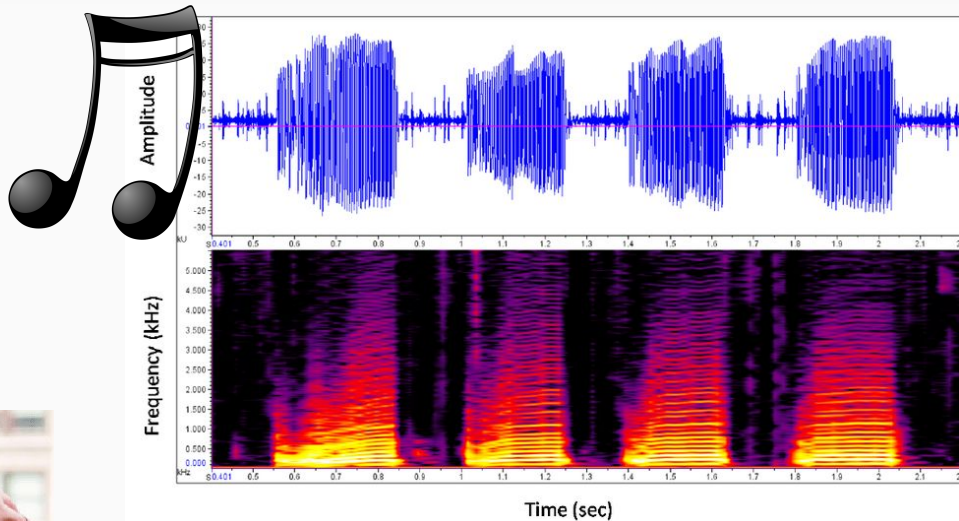
Networks consider granular details, in order to recognize **larger patterns**.

Neural networks see the world in **nested patterns**



Nested patterns
are everywhere.

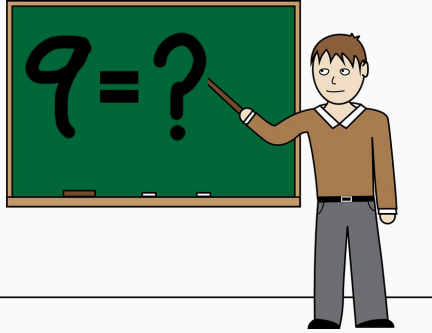
Neural Networks “see” the patterns in data like an image...



Question:

How does a neural network know which pathways are important?

Learning a rule is like repeatedly failing an exam...



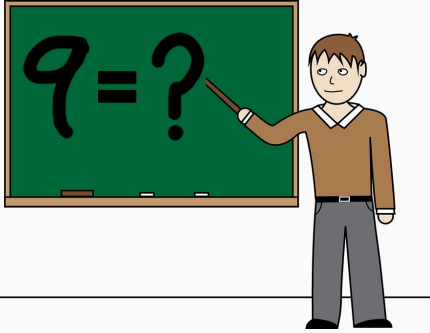
6

8

9



This is a slow, data hungry evolution



6

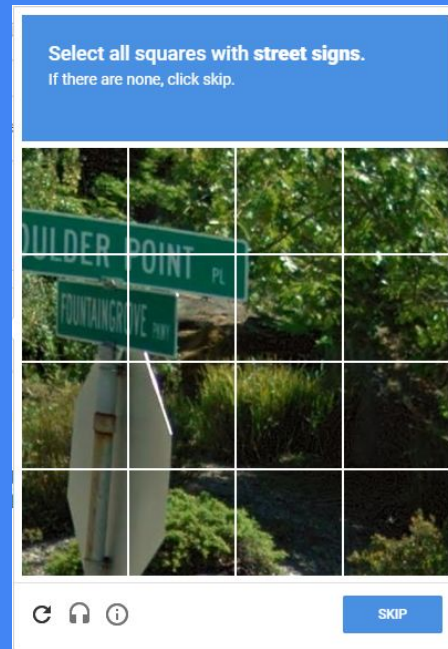
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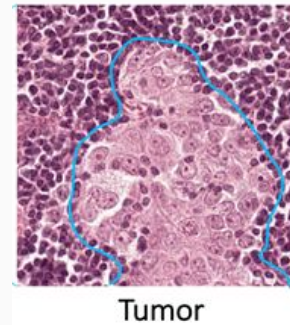
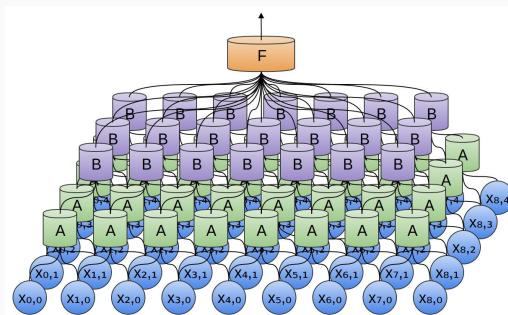
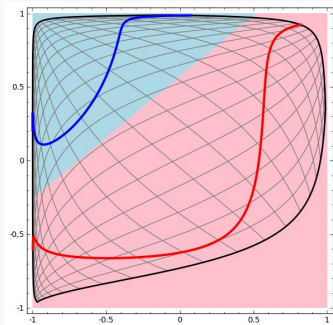
Learning by
memorization
requires many
examples.

Learning by
memorization
requires many
examples.



Current state of deep learning

- STUDY data to interpret new examples
- Build representations that exploit simple rules
- Recognize nested patterns
- Abstract details in complex inputs, present concise outputs



Reinforcement Learning
will move beyond
interpretation to **strategy**

Reinforcement learning involves exploration

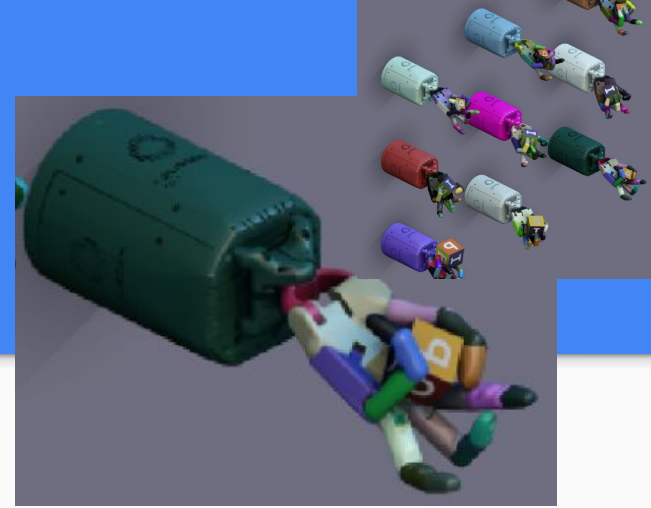
- Learning by “doing”
- Goal driven
- Attempting actions and evaluating outcomes
 - Simulated or real
- Data is created, NOT provided
 - Removes need for existing data

Reinforcement learning involves exploration

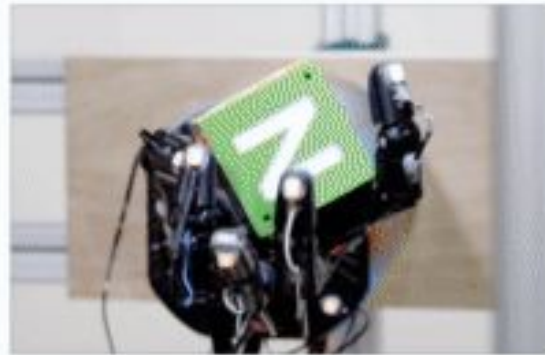
- Learning by “doing”
- Goal driven
- Attempting actions and evaluating outcomes
 - Simulated or real
- Data is created, NOT provided
 - Removes need for existing data



Explore safely in simulation, Transfer knowledge to reality



FINGER PIVOTING

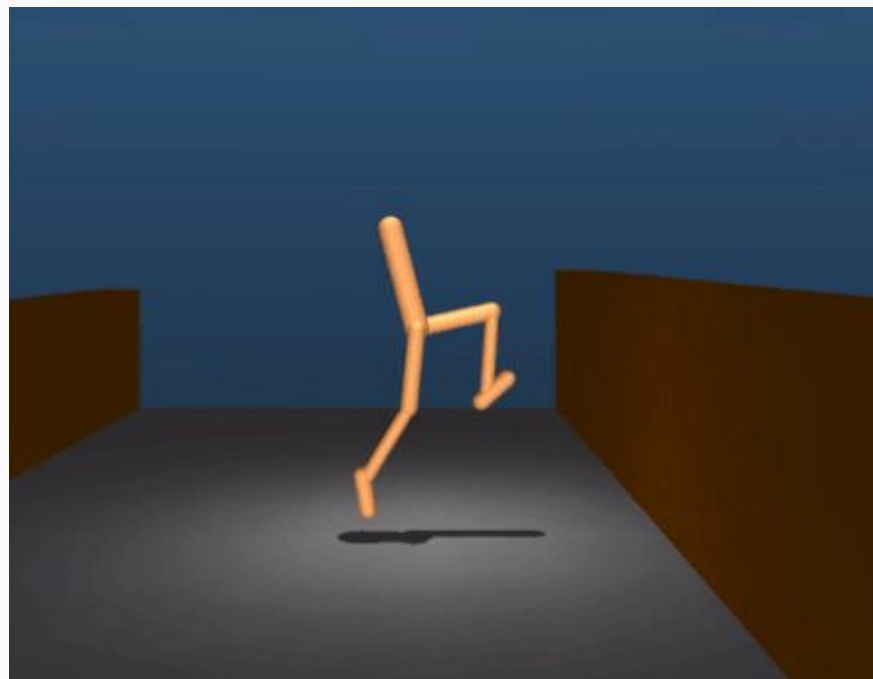


SLIDING

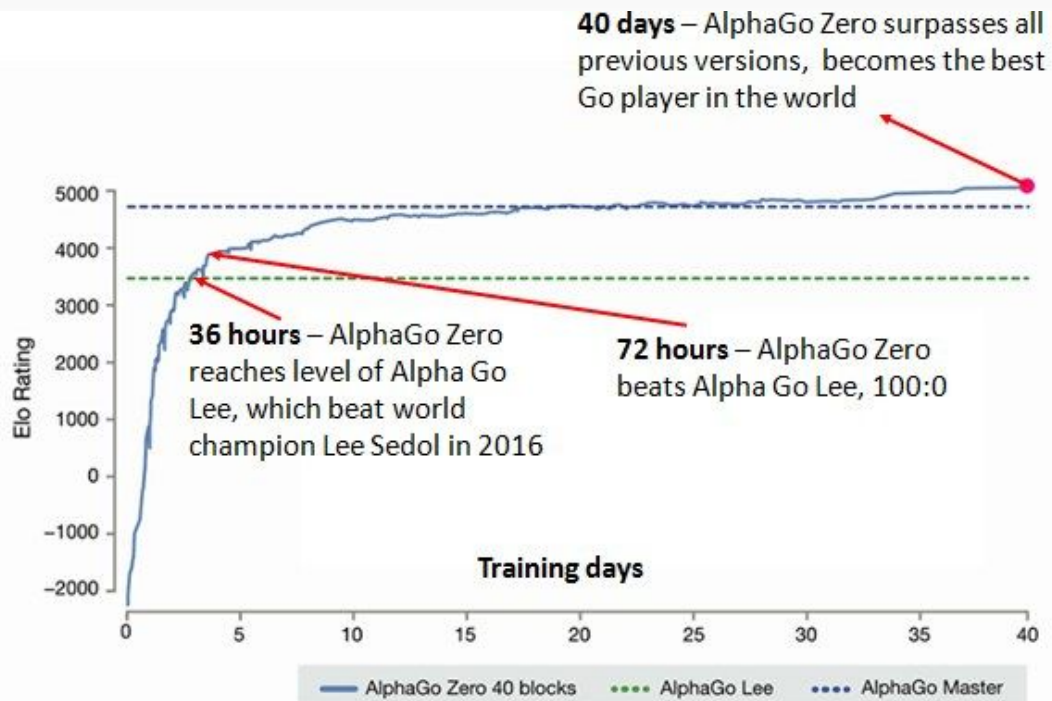


FINGER GAITING

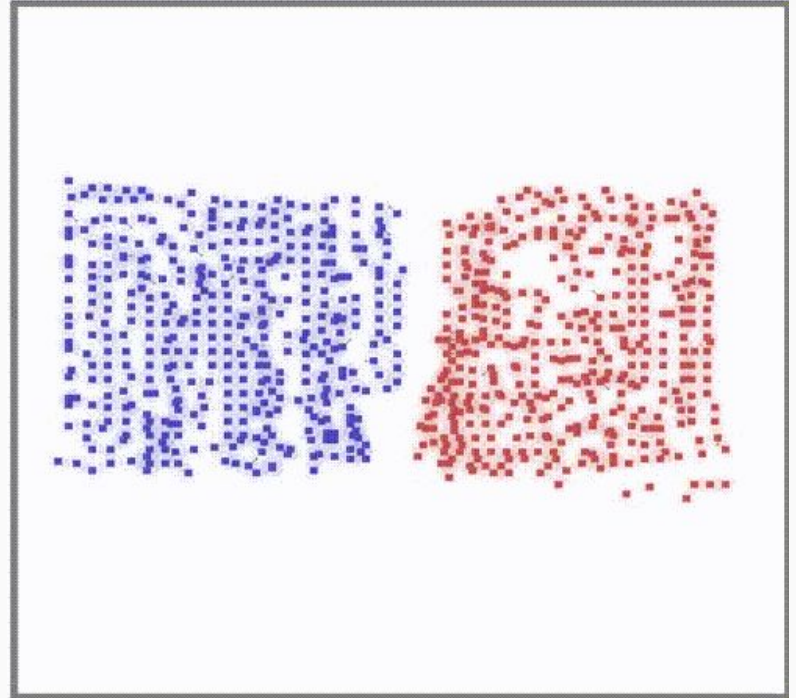
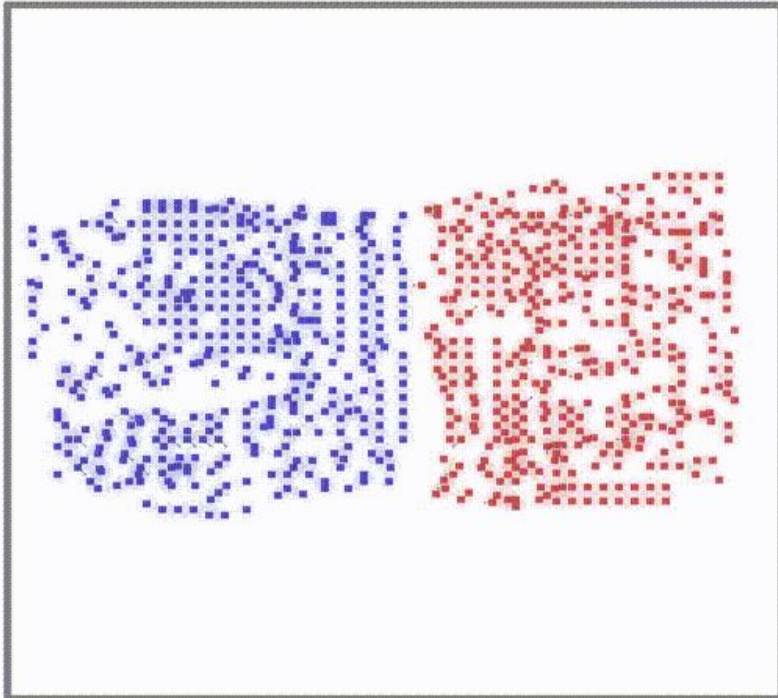
Exploration leads to **novel strategy**,
(compared to learning by example)



Self-play avoids learning plateau, ignoring limitations of teacher



Cooperative strategy is difficult to define,
but can be reinforced



Deep Learning

Sensing systems

Reinforcement Learning

Decision making systems

Deep learning will **improve our senses**

Identify things we can't see

- Anomalies, Problems and Targets
 - Tumors, leads, customers,
- Needs and Opportunities
 - Which student is struggling?
 - Which patient needs medical attention or therapy?

Summarize complex details, and simply point us to the problem.

Reinforcement Learning will **optimize**, and **solve problems**

We'll provide high level goals, or desired outcomes

- Pick up the cup
- Reduce congestion
- Increase profit
- Maximize yield
- Retain user
- **Improve health**

Neural Networks Demystified

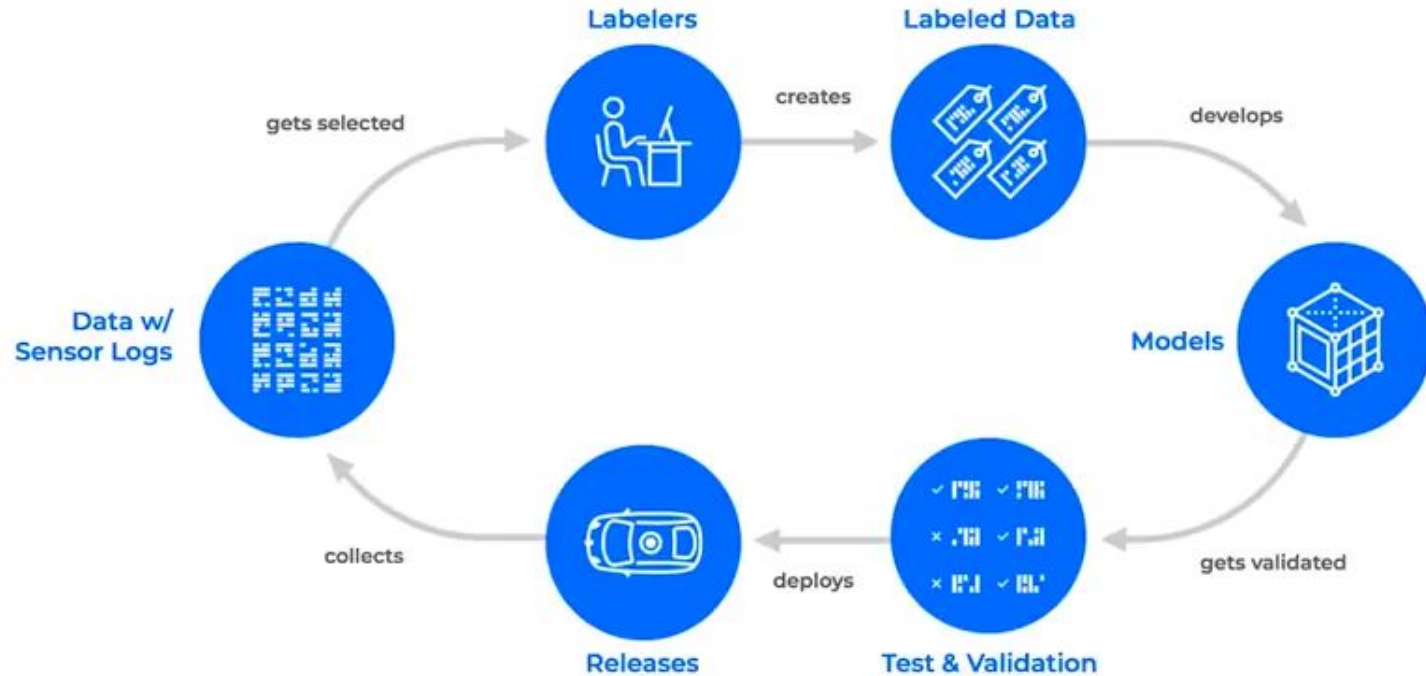
Thank you!

Contact: david@questionablyartificial.com



Bonus:
(if there is time)

Training Neural Networks is a tiny part of a Cyclical, Integrated Process



AI is the cognitive component, but the
grander problem is Sensing Infrastructure.

Cognition cannot be developed without a sensing infrastructure.

Development infrastructure is crucial

Labeling

Simulation

Gyms

Visualization

Dev Tools

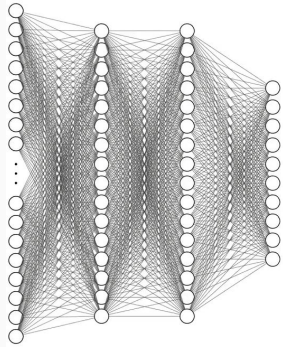
Networking

AI development requires diverse specialists, more than AI engineers.

- Developer tools
- Low level hardware optimization
- Edge compute and sensing
- Sensor fusion
- Sensor manufacturers
- Data storage, organization, and retrieval.

Translation

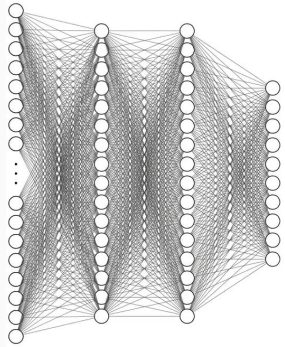
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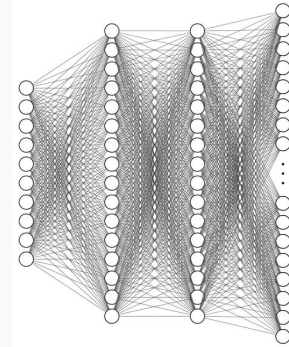
**Concise
Output**

Translation

**Complex
Input**



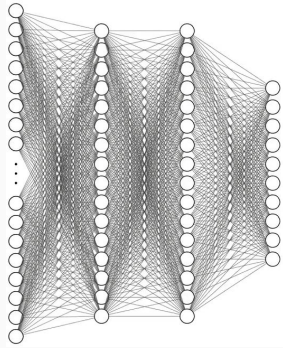
**Concise
Format**



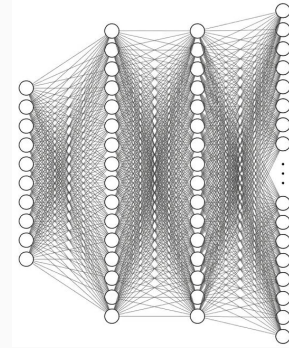
**Complex
Output**

Translation

**Complex
Input**



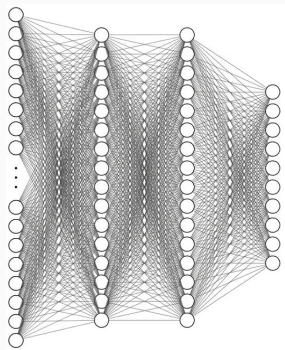
Translation



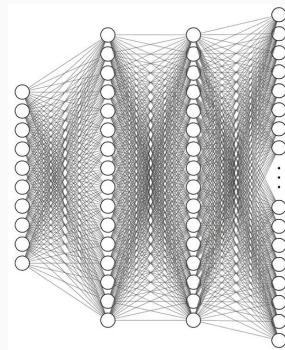
**Complex
Output**

Translation

**Complex
Input**



Translation



**Complex
Output**

hello

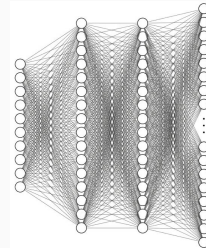
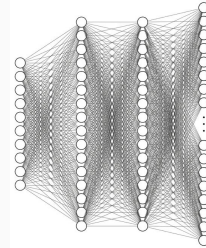
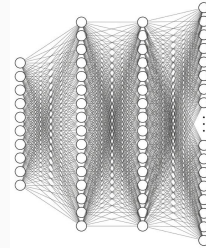
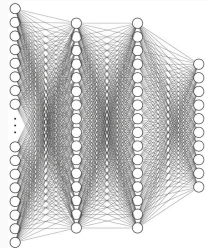
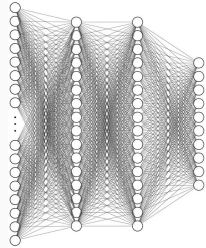
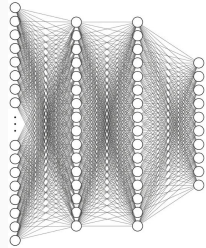
你好

Neural networks can learn to translate, even when we don't know how

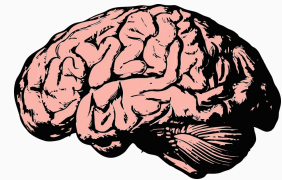
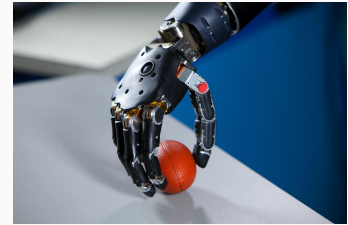
Complex Input



Translation

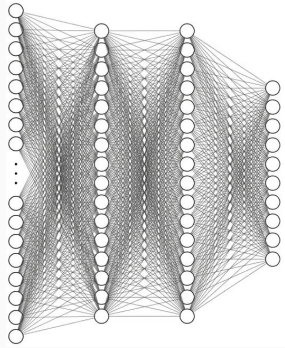


Complex Output

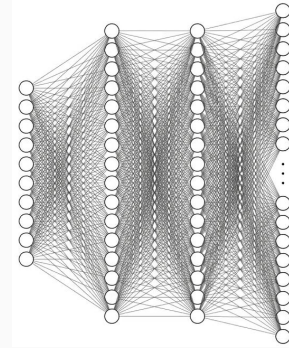


Translation

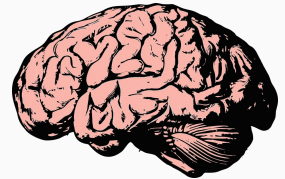
**Complex
Input**



Translation

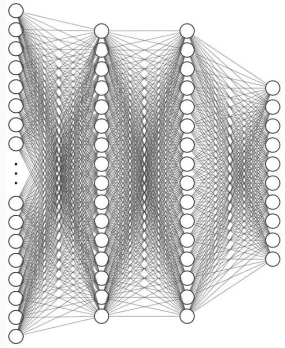
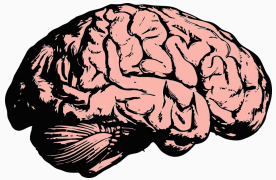


**Complex
Output**

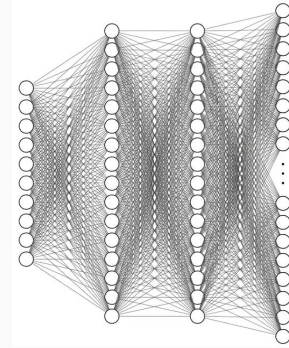


Translation

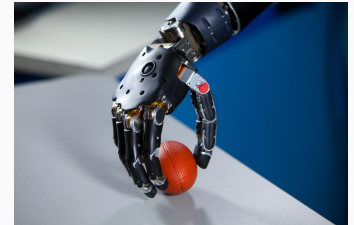
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Input**



Translation

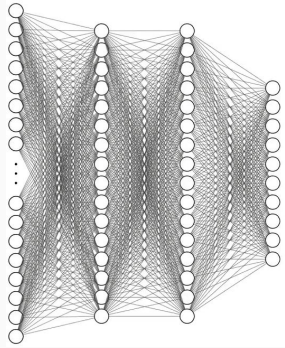
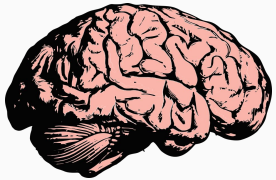


**Complex
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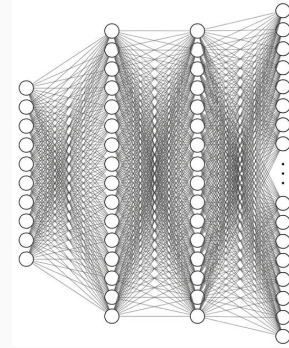


Translation

**Complex
Input**



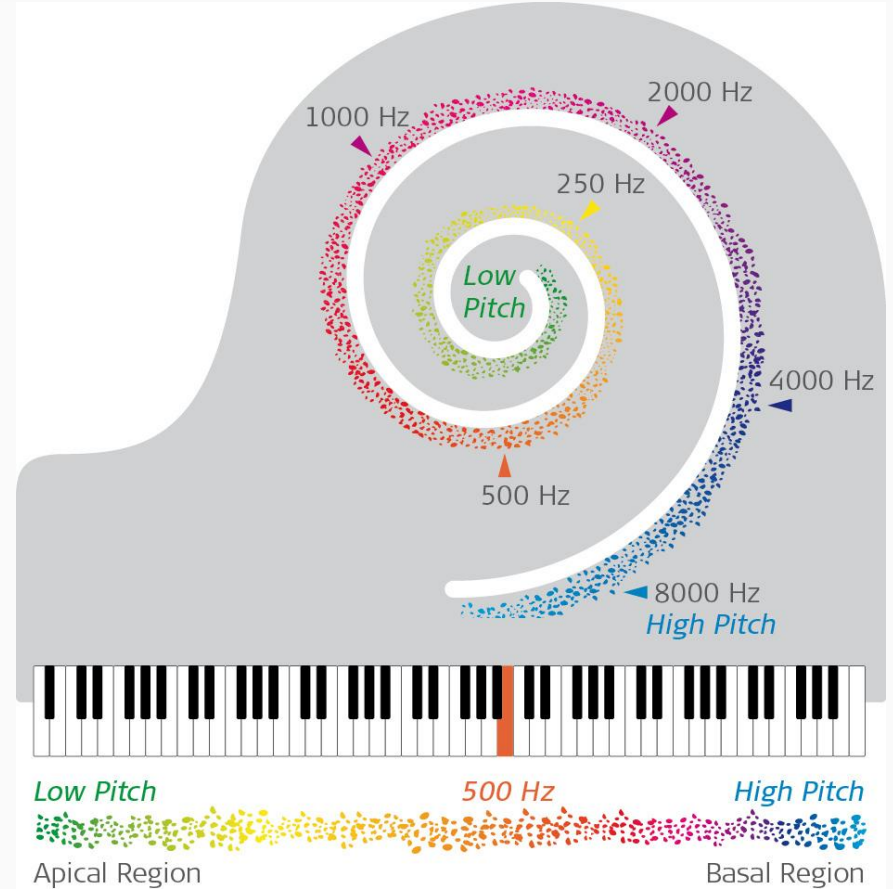
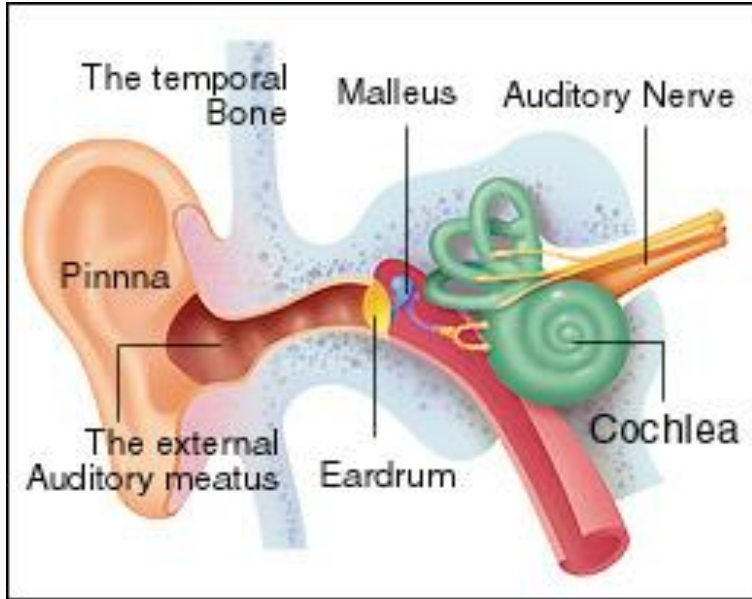
Translation



**Complex
Output**



Human perception is **discrete**



References

Some imagery pulled from various sources, most notably:

- OpenAI
- colah.github.io
- Drago Anguelov (Waymo) MIT Self-Driving Car Diagram
- 3Blue1Brown