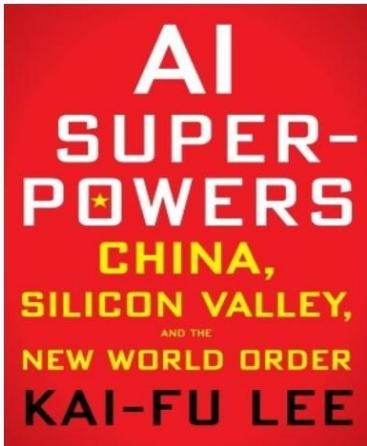


How Neural Networks See the World

Building Intuition for a Broader AI Discussion



Book was great!



Broader AI conversation.

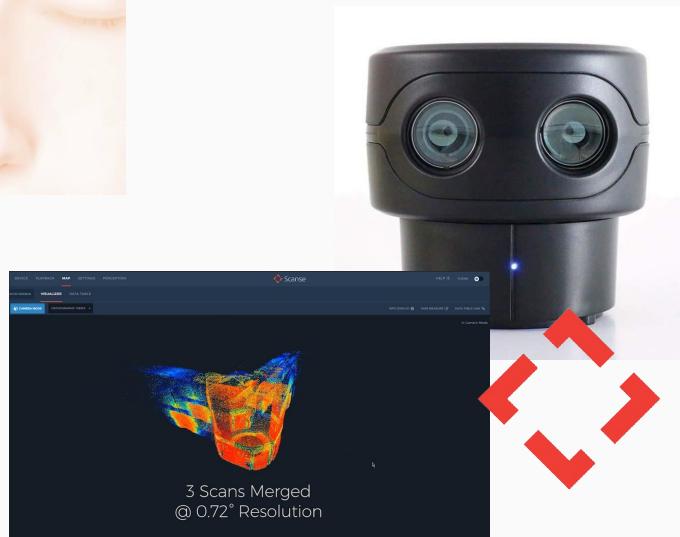
But how does all this “AI” stuff actually work?

What is deep learning?

How do neural networks learn?

Engineering Perception

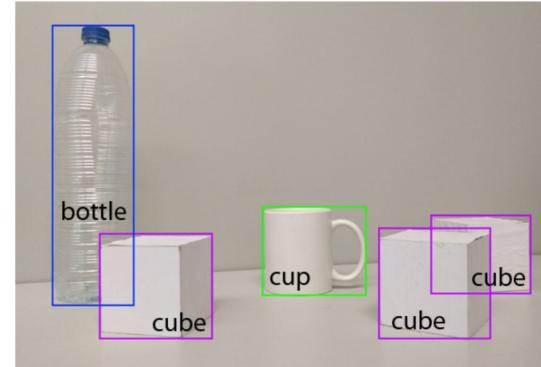
I offer a developer perspective



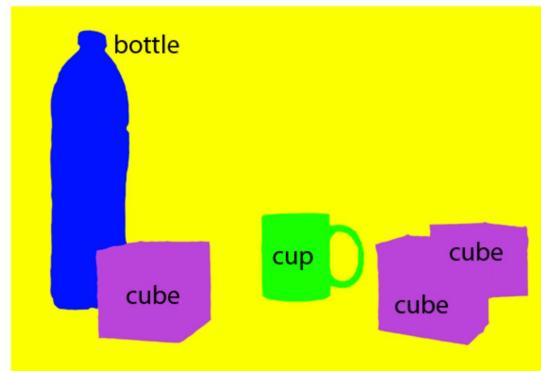
RIPCORD



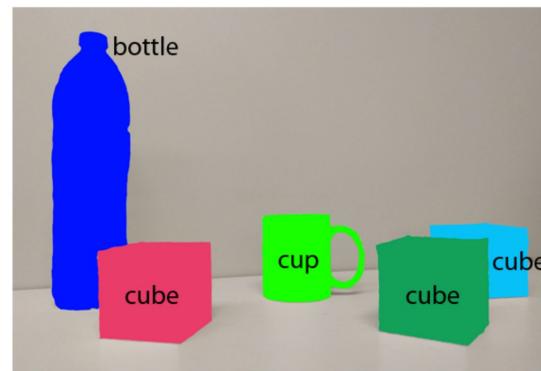
(a) Image classification



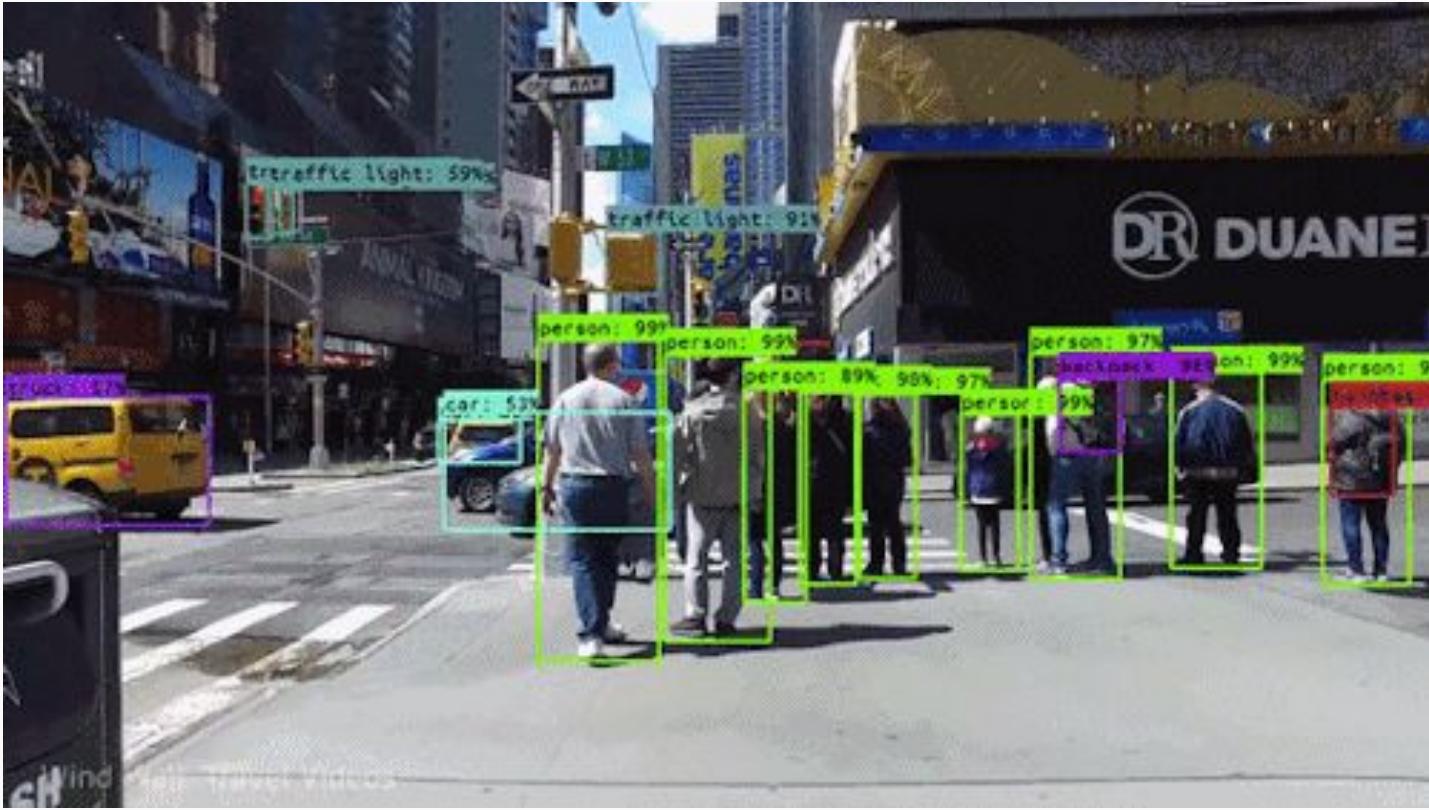
(b) Object localization

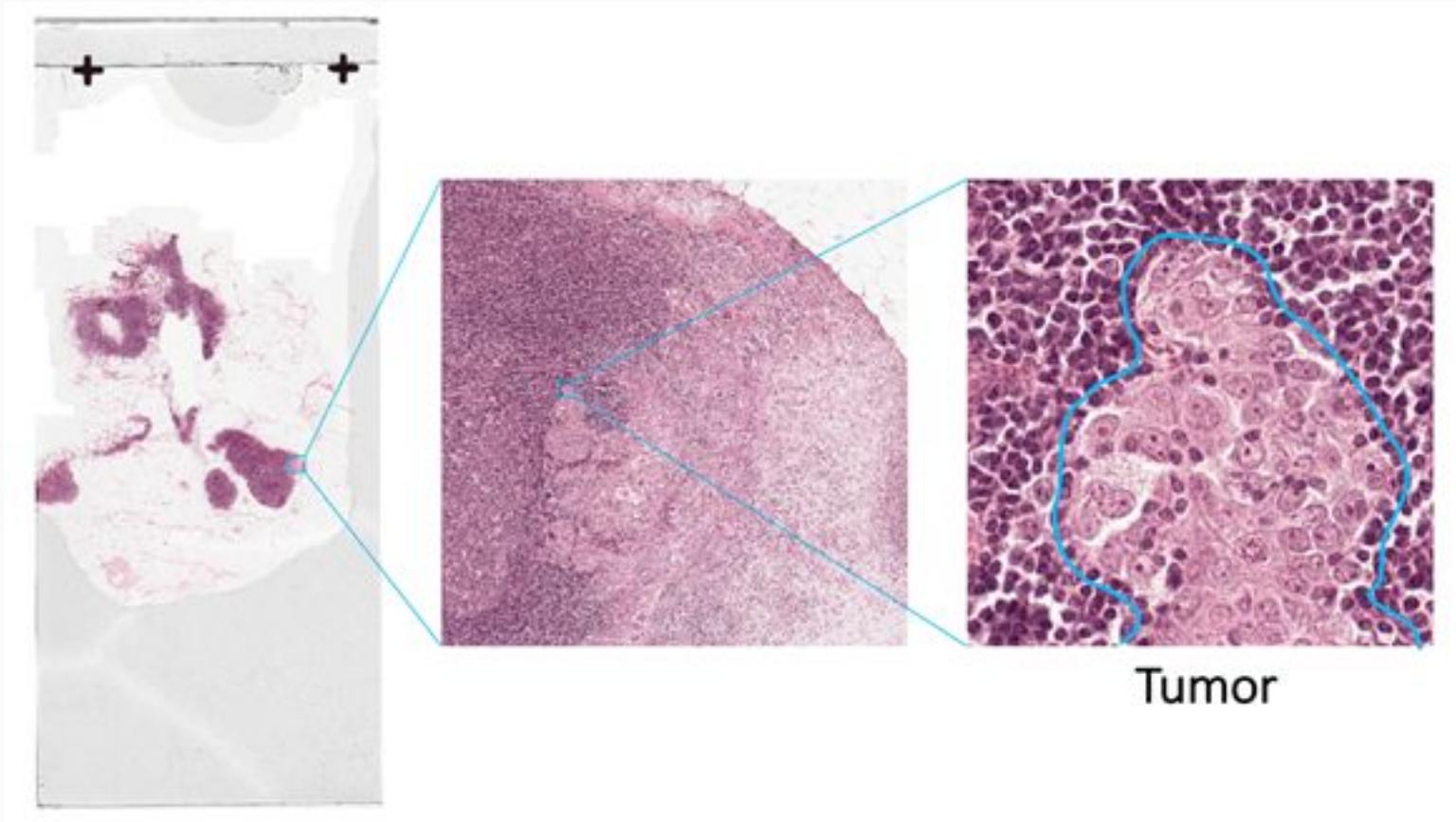


(c) Semantic segmentation



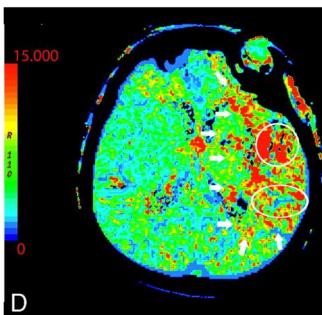
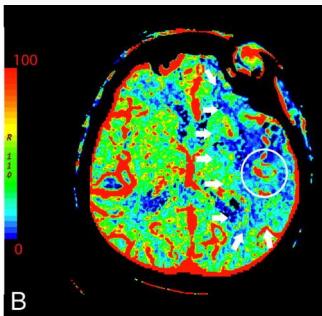
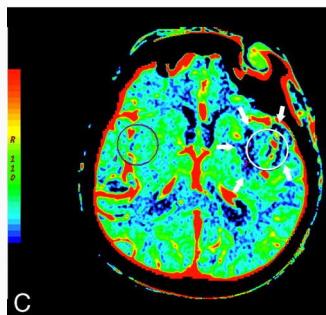
(d) Instance segmentation



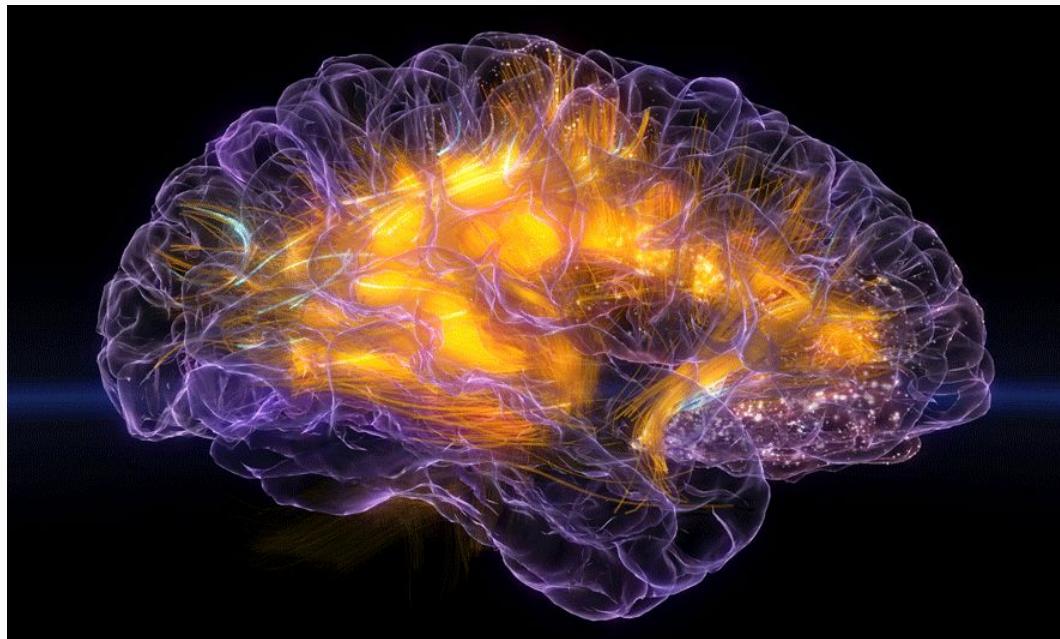
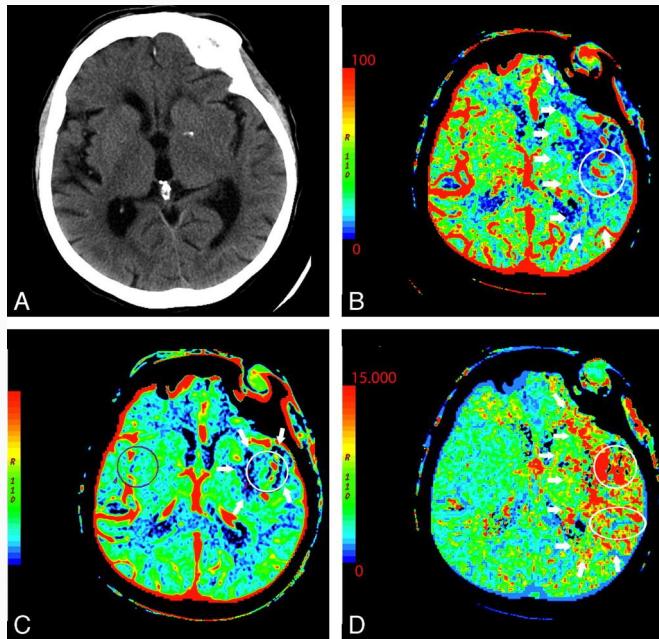


How do we (humans)
see the world?

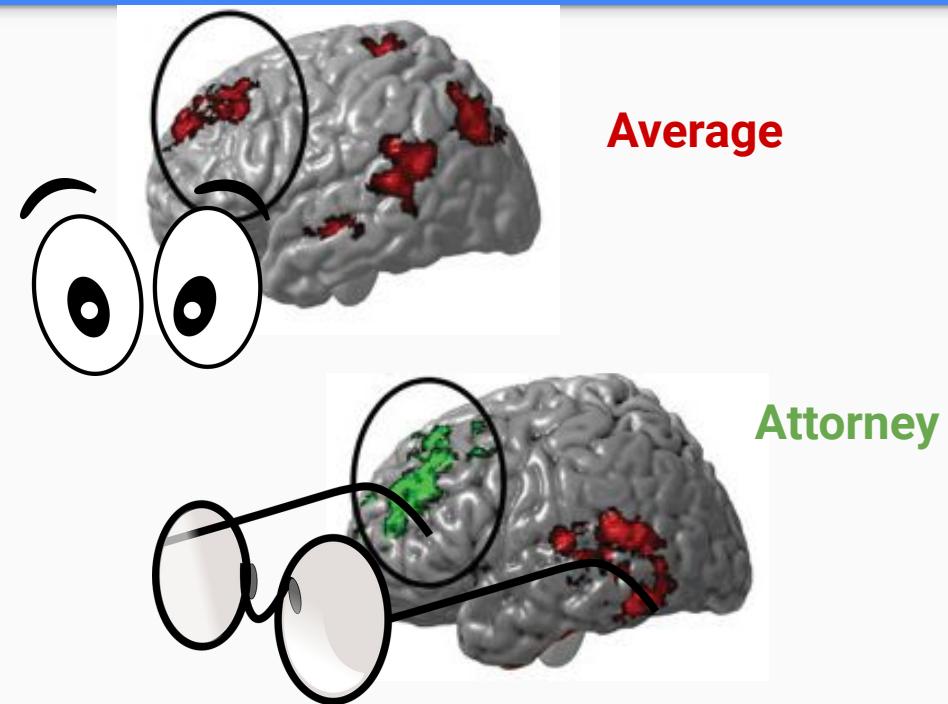
Our brains get **excited**



Our brains get **excited**



Experience leverages **different pathways**



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.

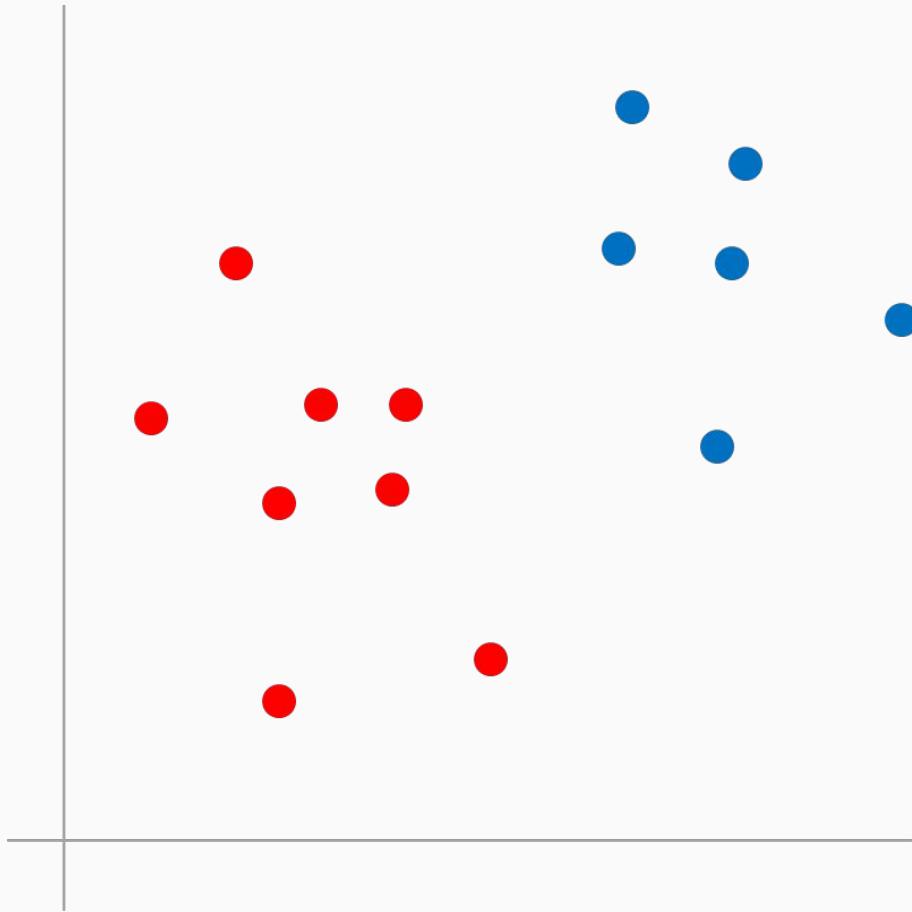
Studying or observing more
examples, in preparation.

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:

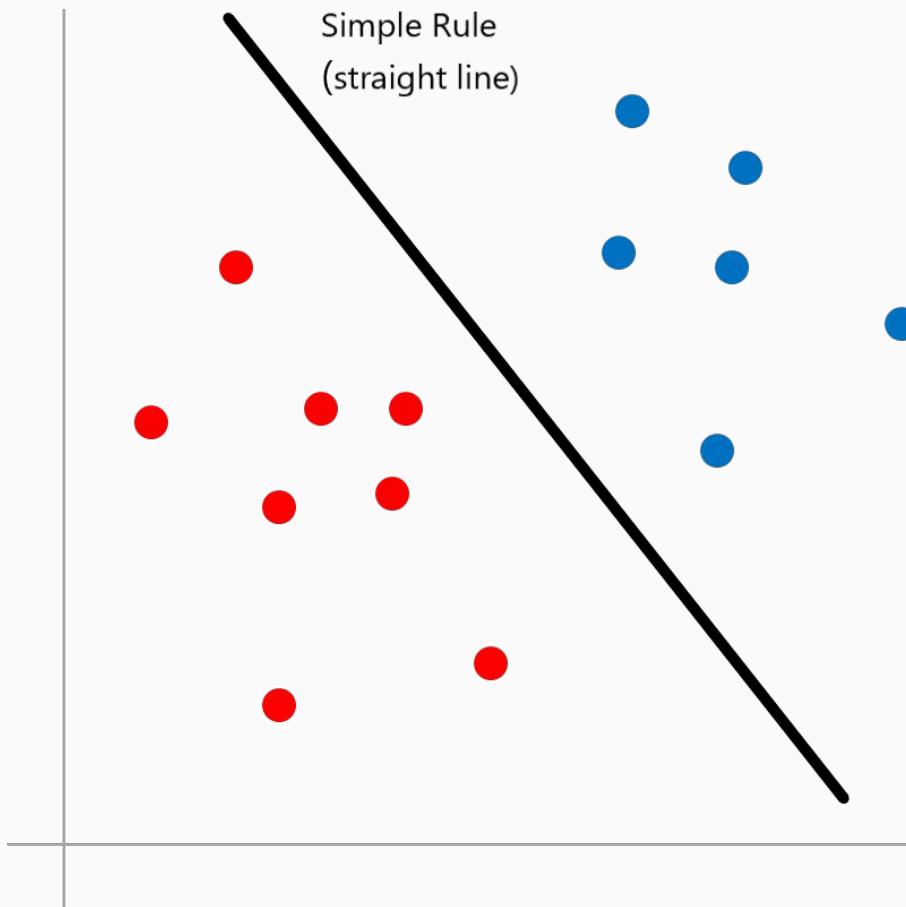
Points ABOVE the line are BLUE

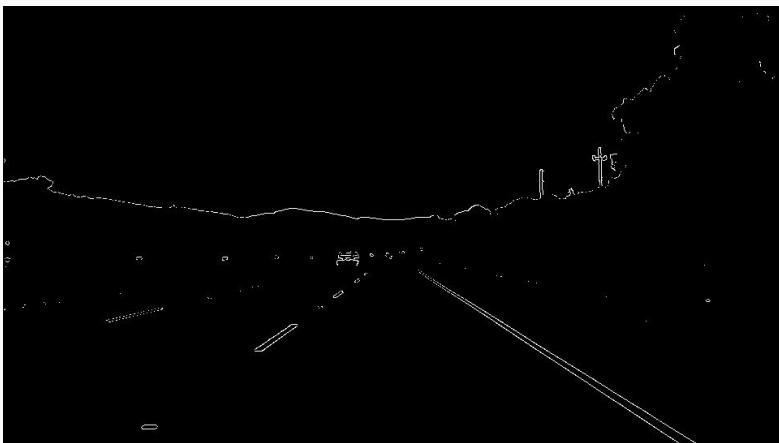
RULE #2:

Points BELOW the line are RED

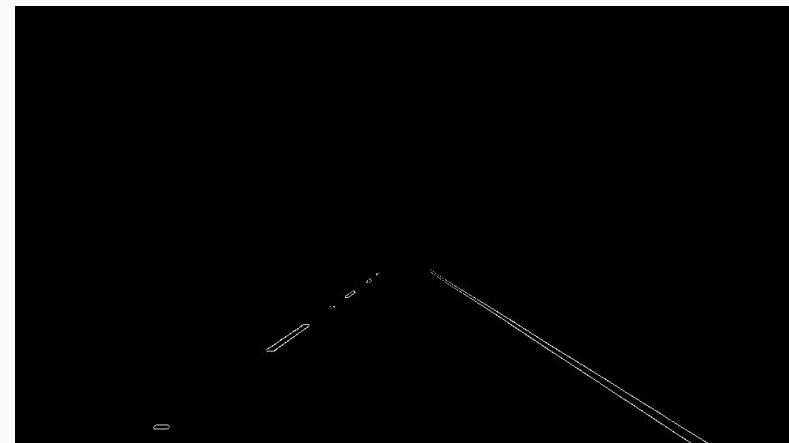
Leverage math to formally describe the rule:

$$y = mx + b$$

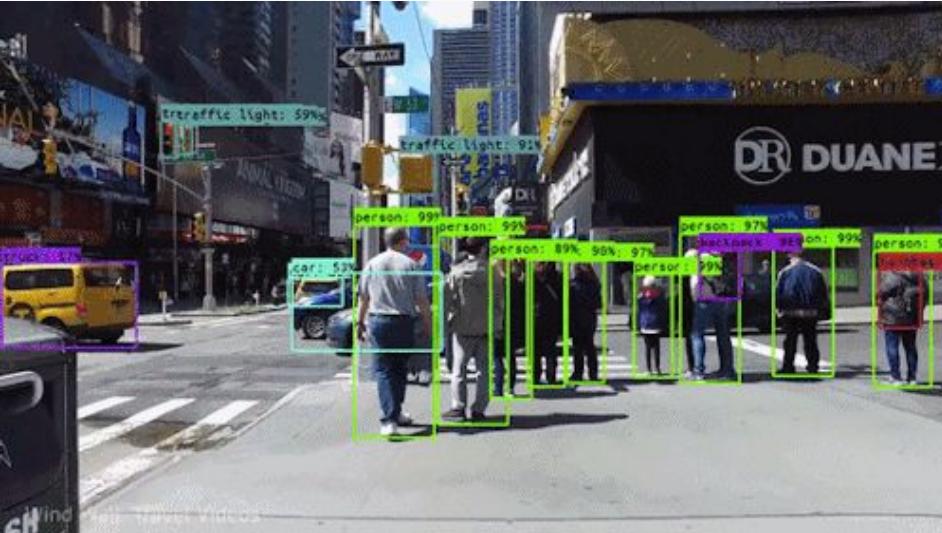




Rule: Lane Lines are WHITE

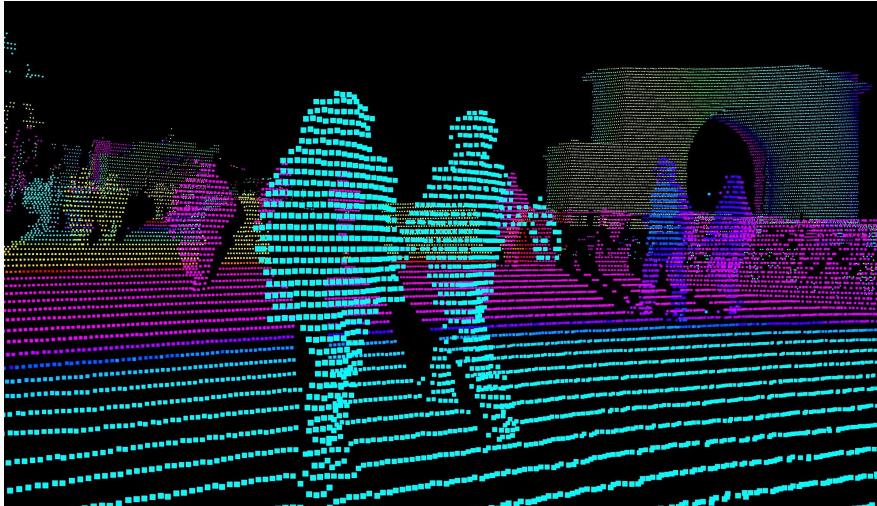
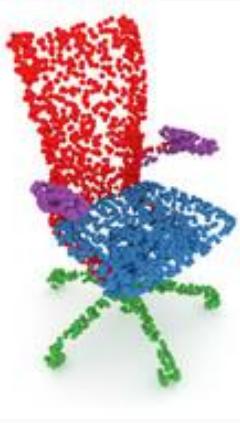


Rule: Road is below horizon



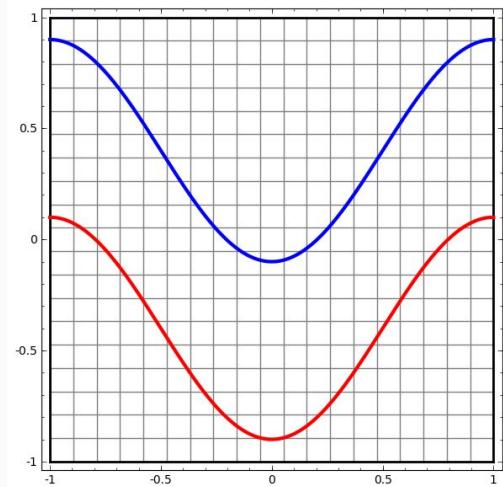
But real data is messy,
and writing complex
rules is hard.

Deep Learning makes complex
discrimination possible.

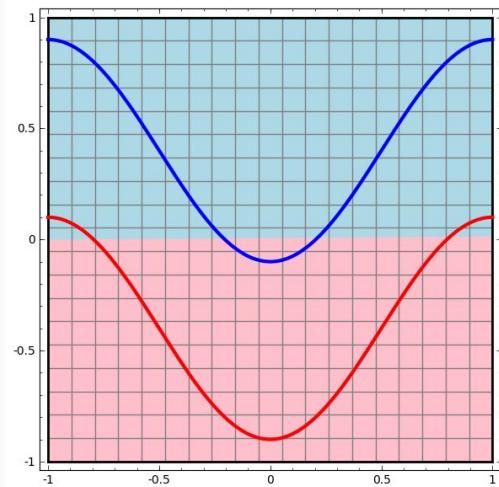


But real data is messy,
and writing complex
rules is hard.

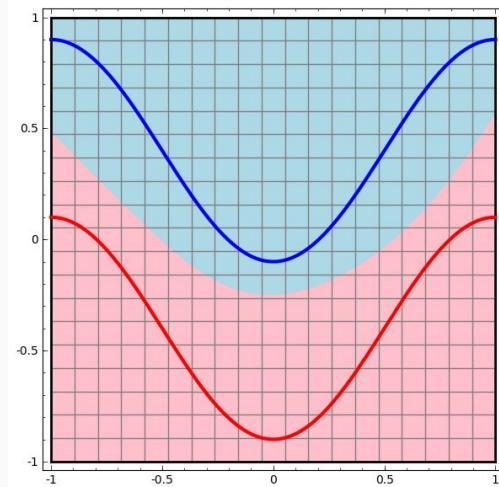
Neural networks **appear** to learn complex rules.



Complex Data



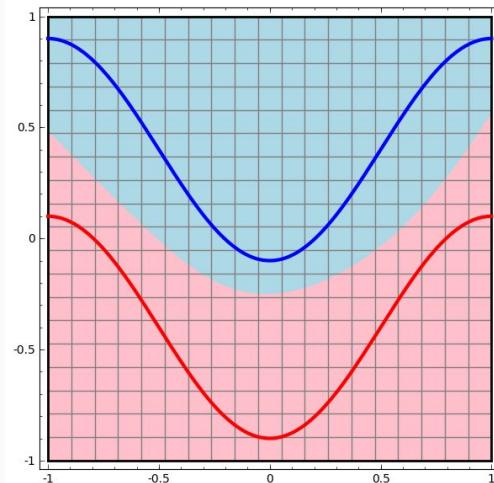
Simple Rule
(inadequate)



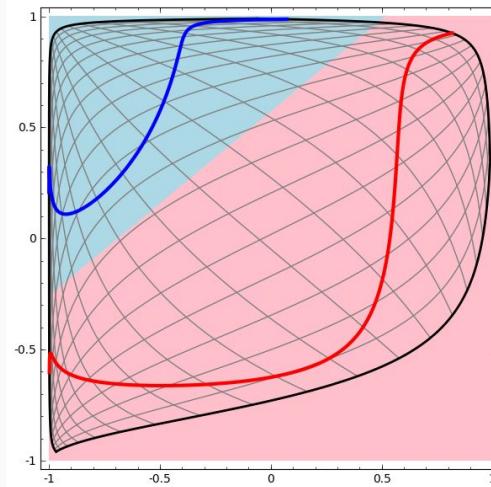
Complex Rule

Actually:
Neural networks
apply simple rules
in a complex way.

To apply simple rules, neural networks distort space.



Complex Rule

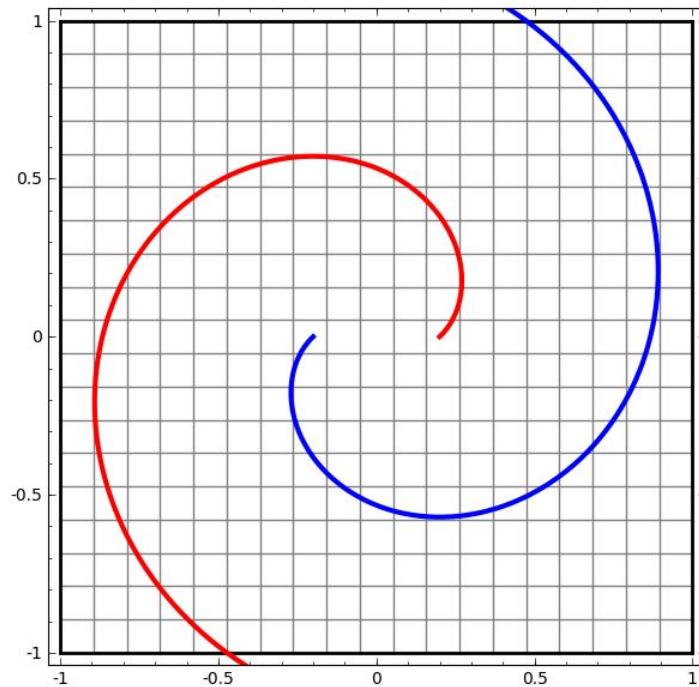
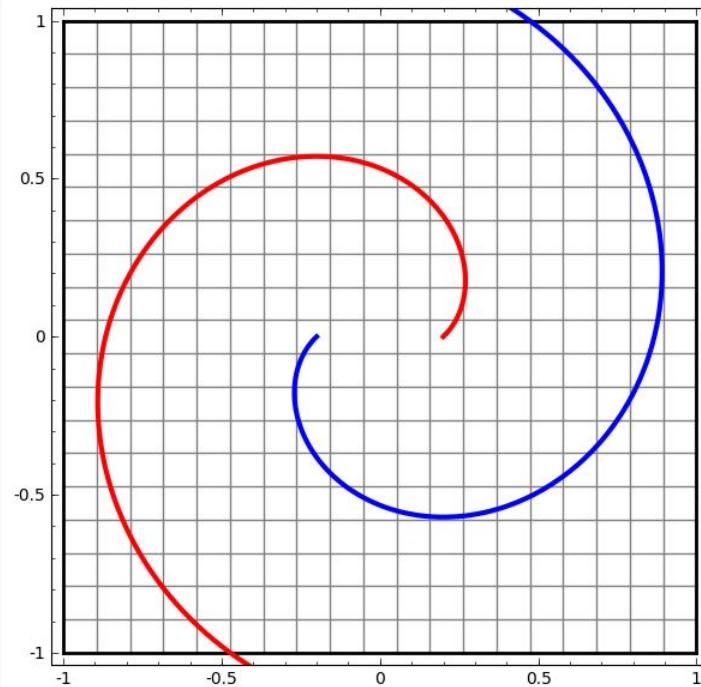


Simple Rule

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.

Vision
Reinforcement 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

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

Stop Defining Rules

Start Learning Rules

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

Neural networks study data.

Often under supervision that demands
memorization.

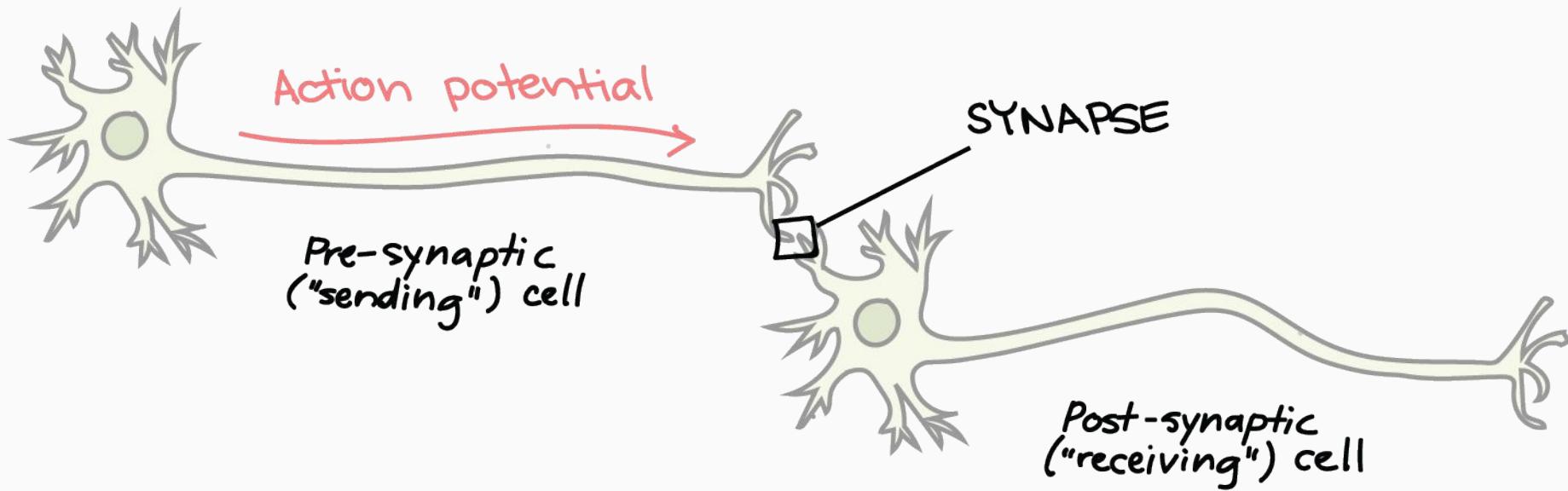
Developers are teachers, creating classroom
environment to facilitate learning.



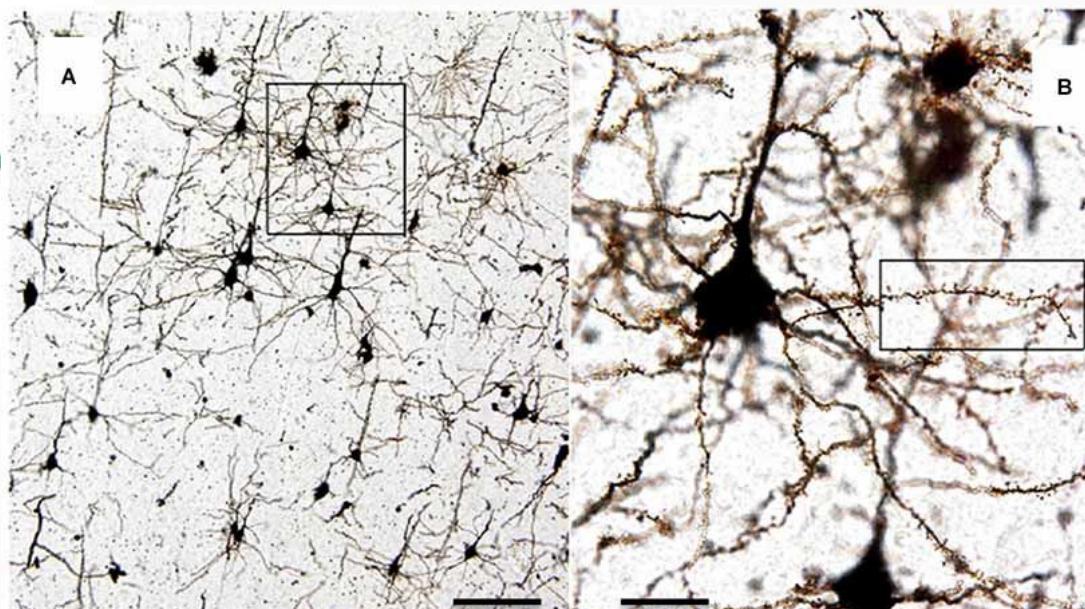
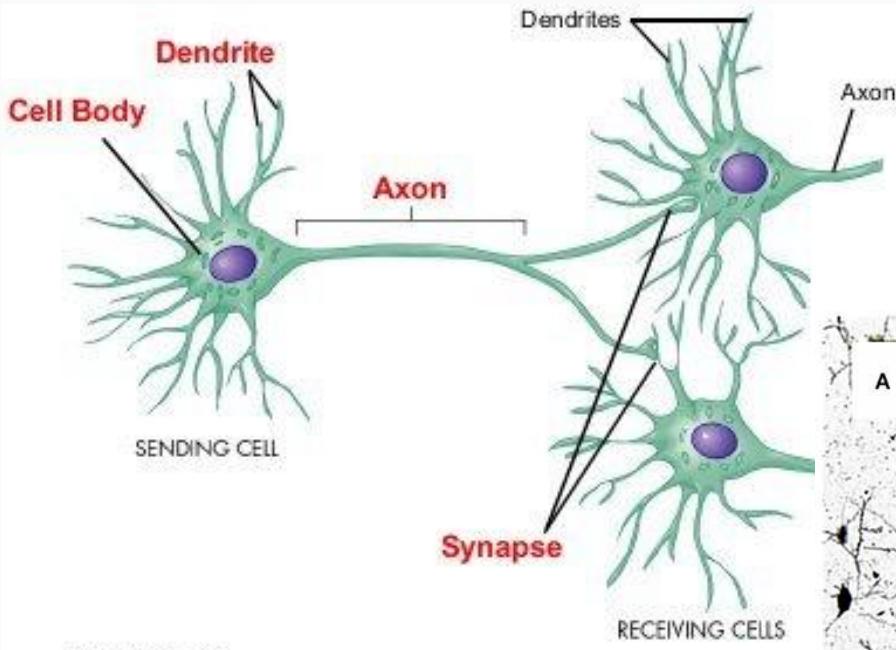
Question:

What is a network?

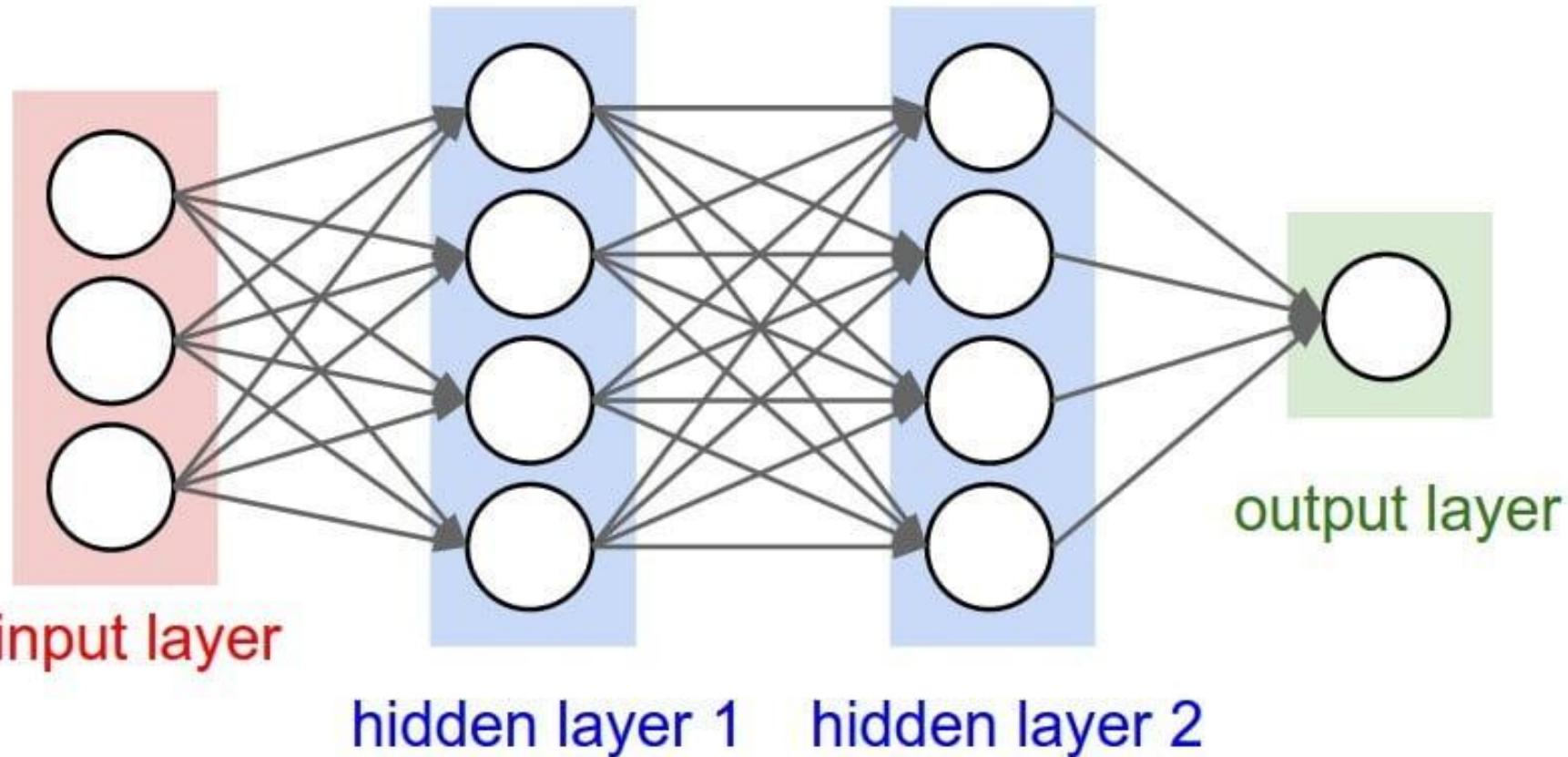
Connections are the building blocks of networks



Networks are comprised of many tangled connections

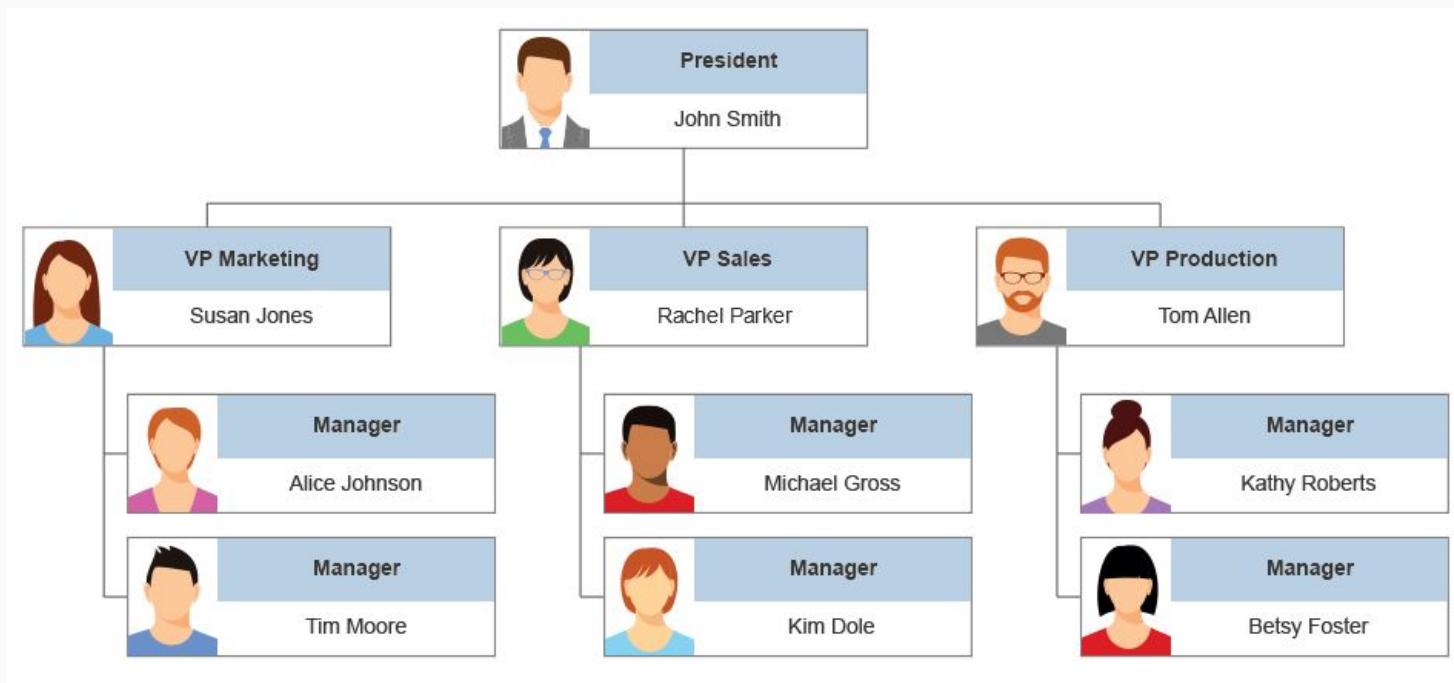


From the outside, networks have **inputs** and **outputs**

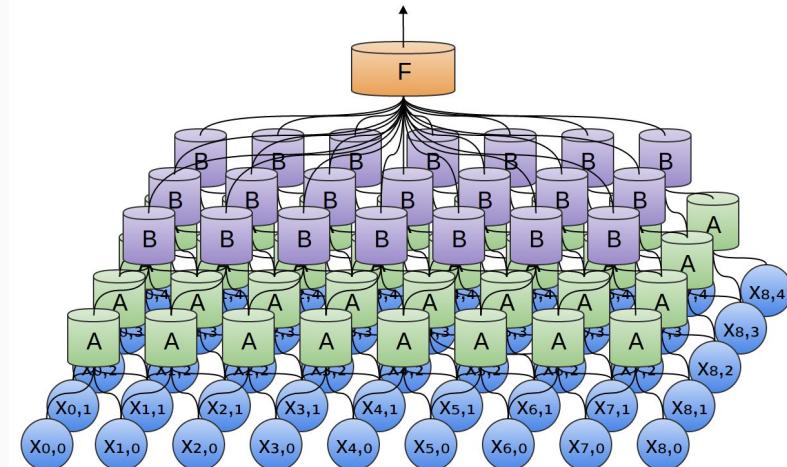
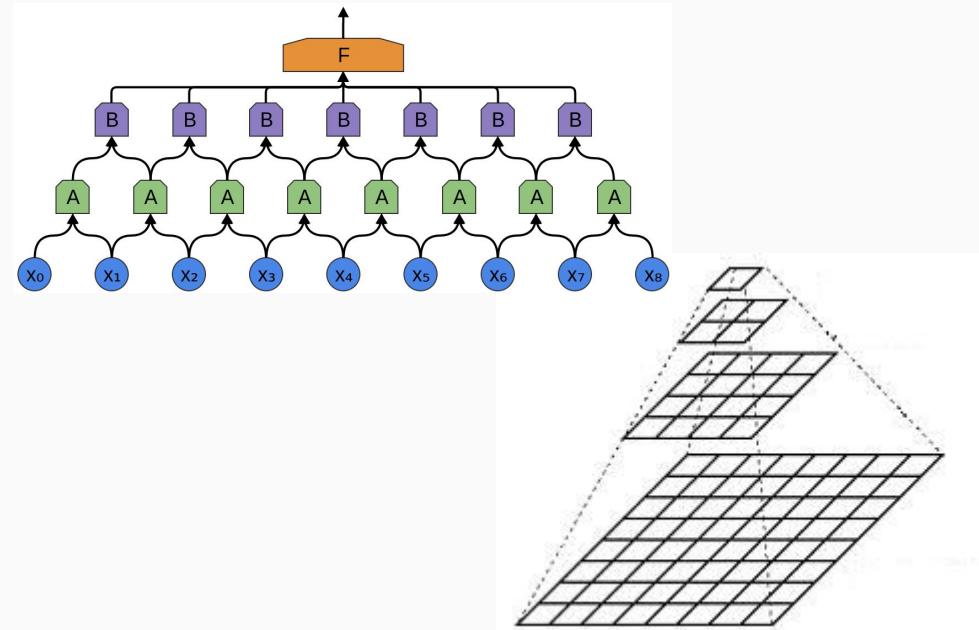


You can design
network **architectures**
for pattern recognition

Seeing the bigger picture requires abstraction and distance.



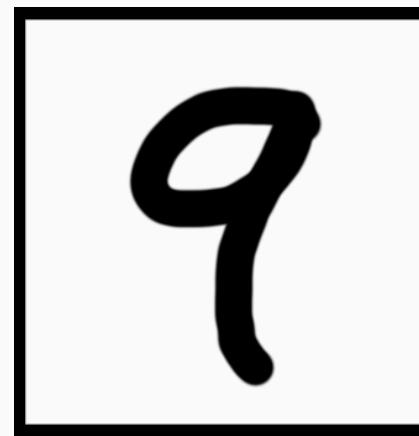
Sensing pyramids abstract details



Example: Classify handwritten digits



| |
|---|
| 0 |
| 1 |
| 2 |
| 3 |
| 4 |
| 5 |
| 6 |
| 7 |
| 8 |
| 9 |



| |
|---|
| 0 |
| 1 |
| 2 |
| 3 |
| 4 |
| 5 |
| 6 |
| 7 |
| 8 |
| 9 |



?

=



?

9

=

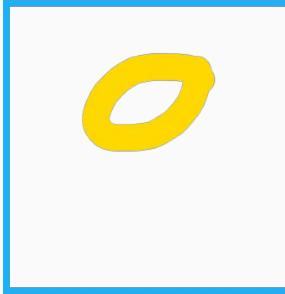


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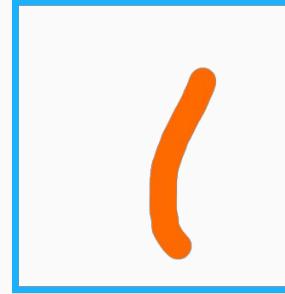


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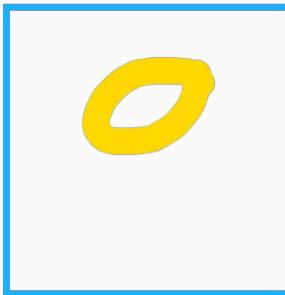


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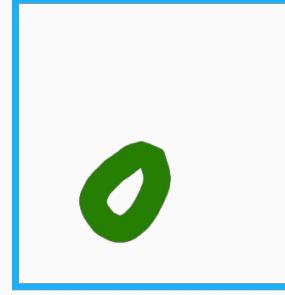


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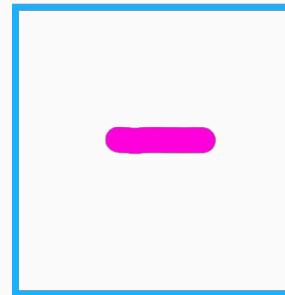


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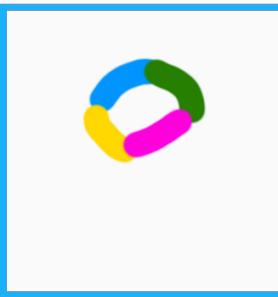


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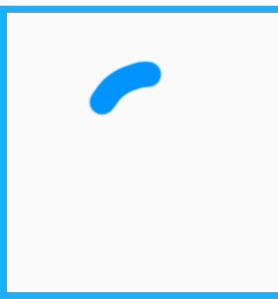




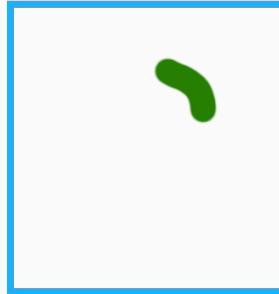
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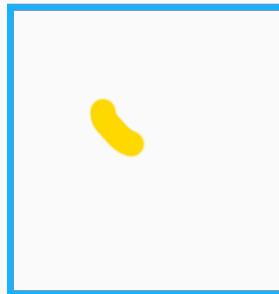
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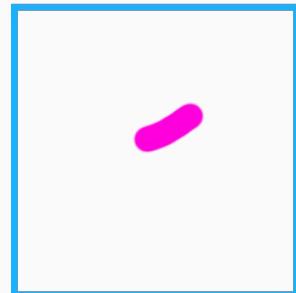
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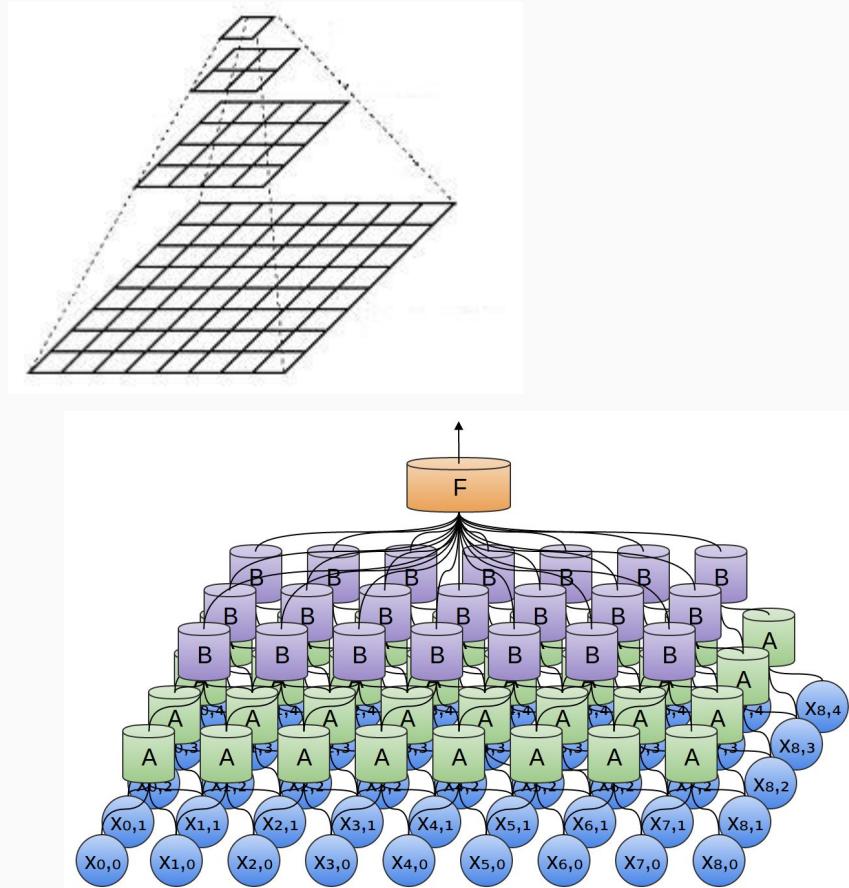
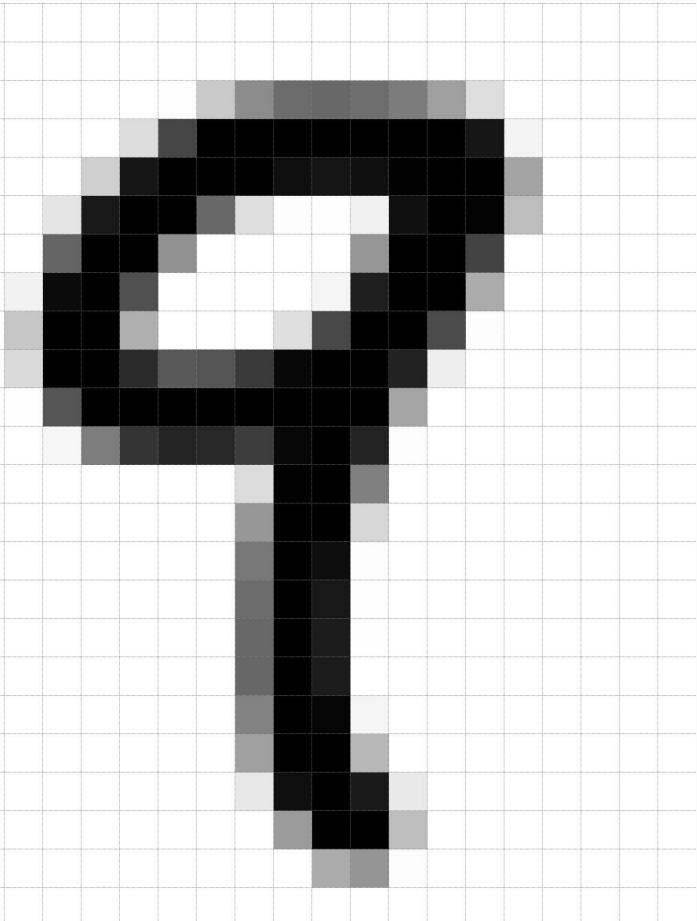


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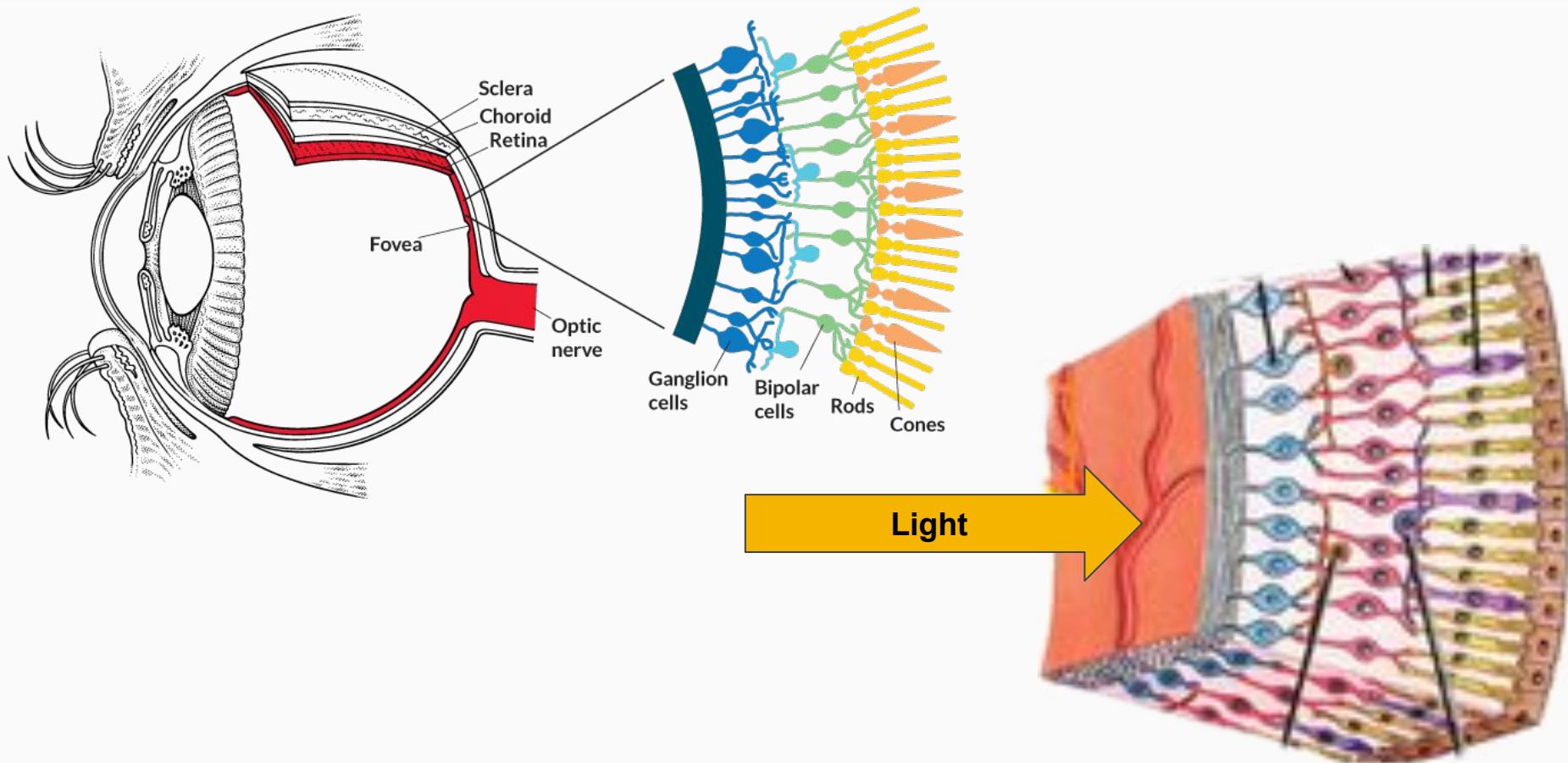
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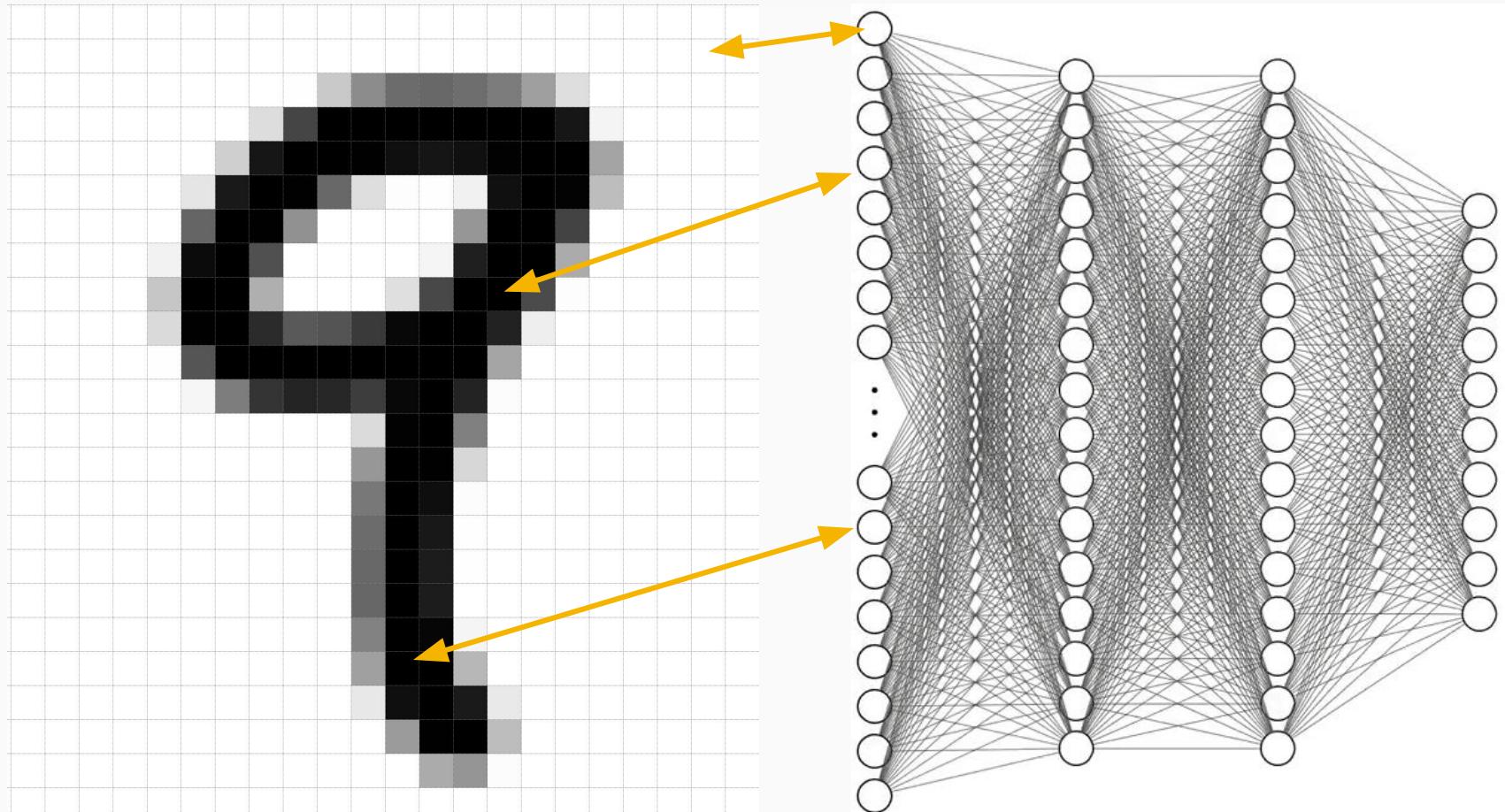


We limit detail at a certain level (pixels)

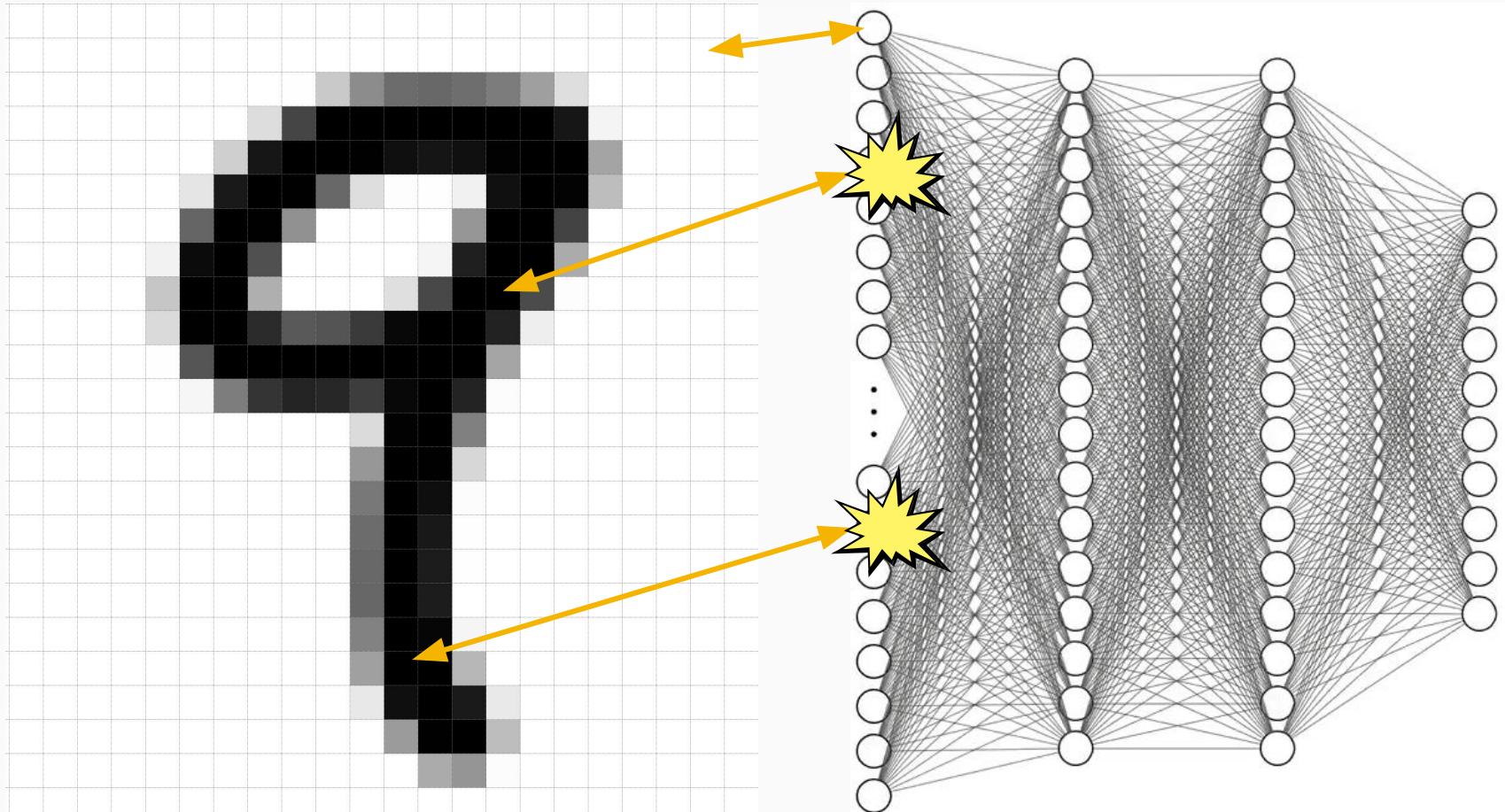
Human perception is **discrete**



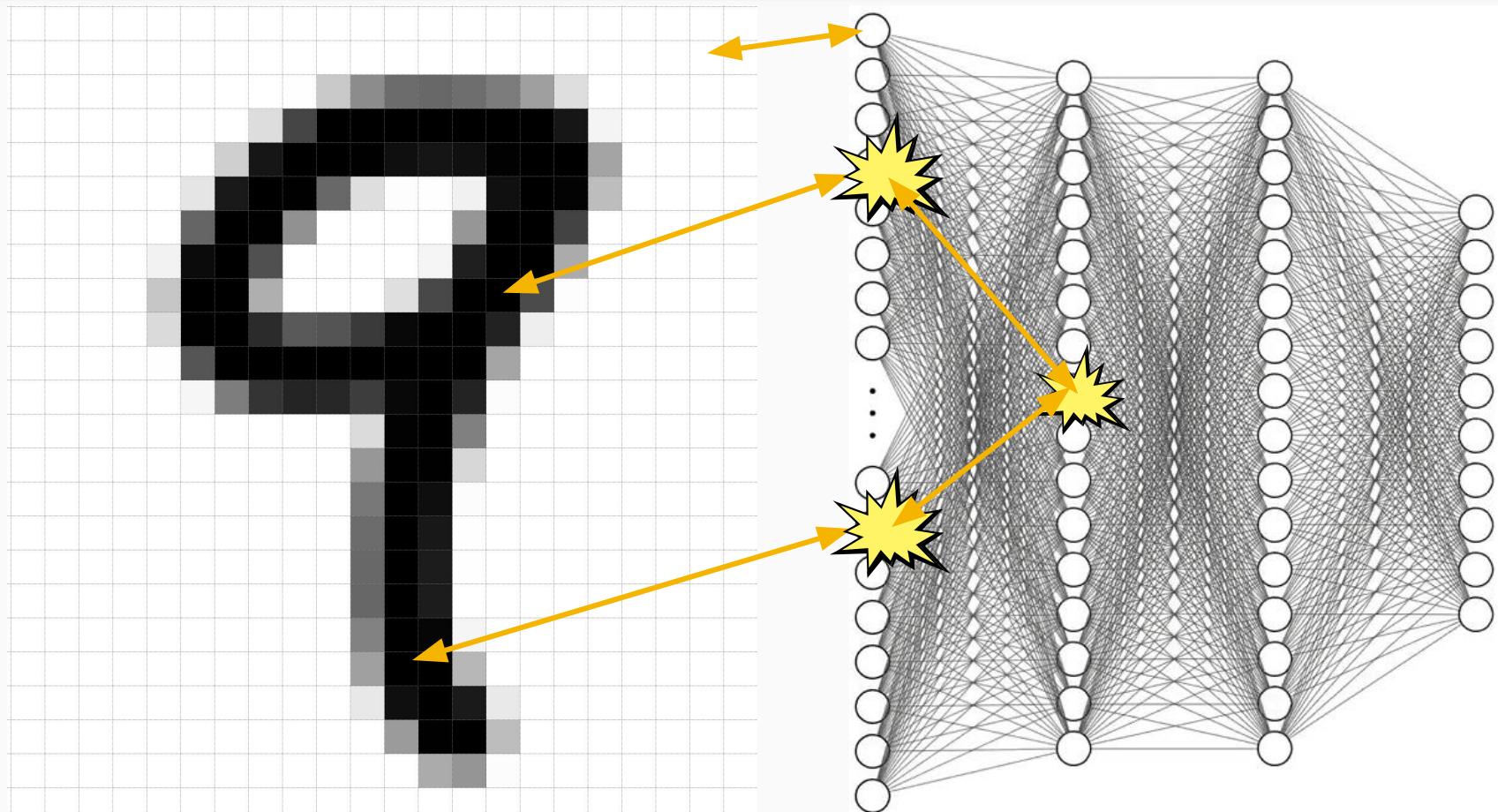
At one end, a network must **connect to data**

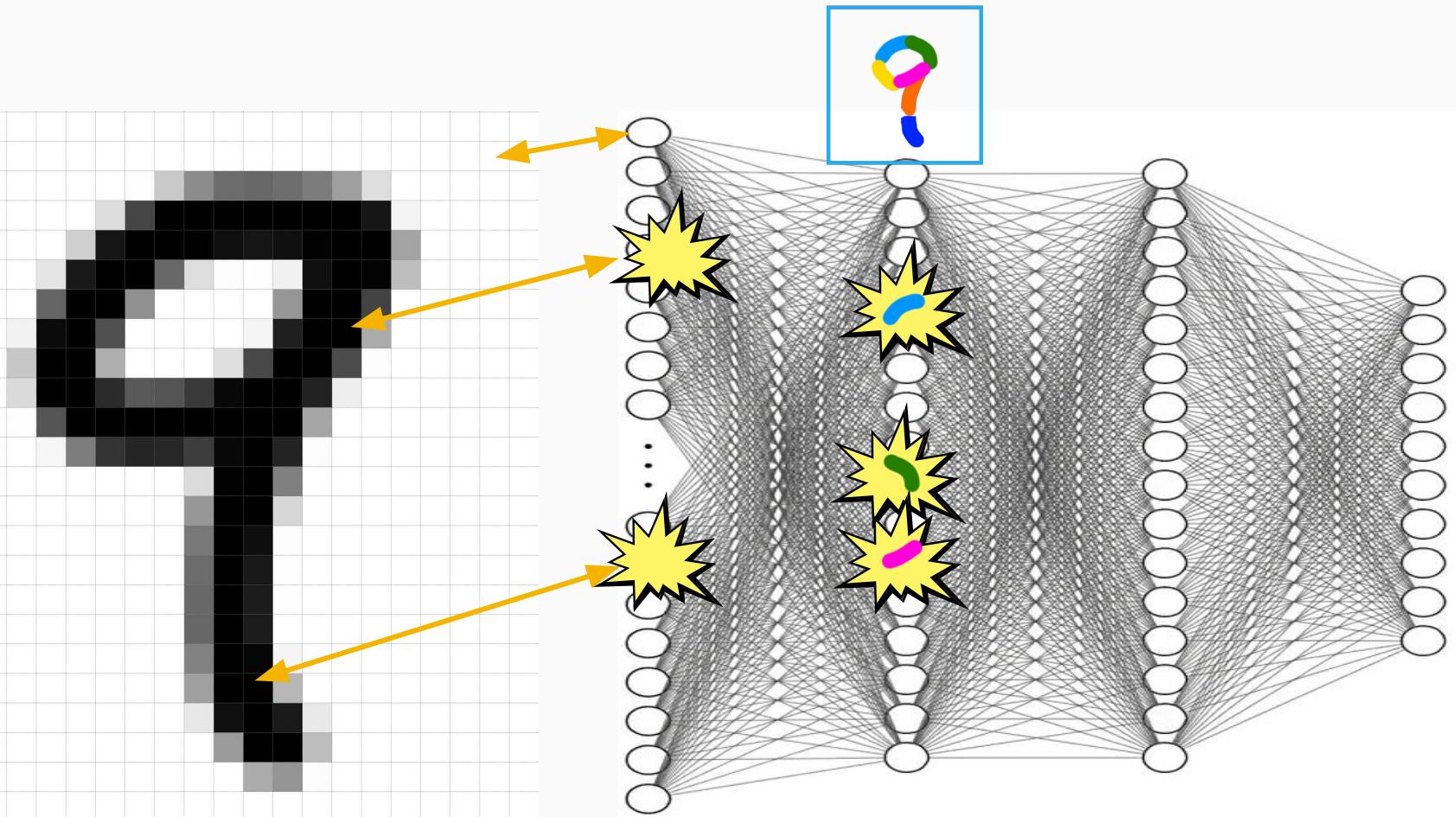


Different inputs will **excite** different pathways in the network

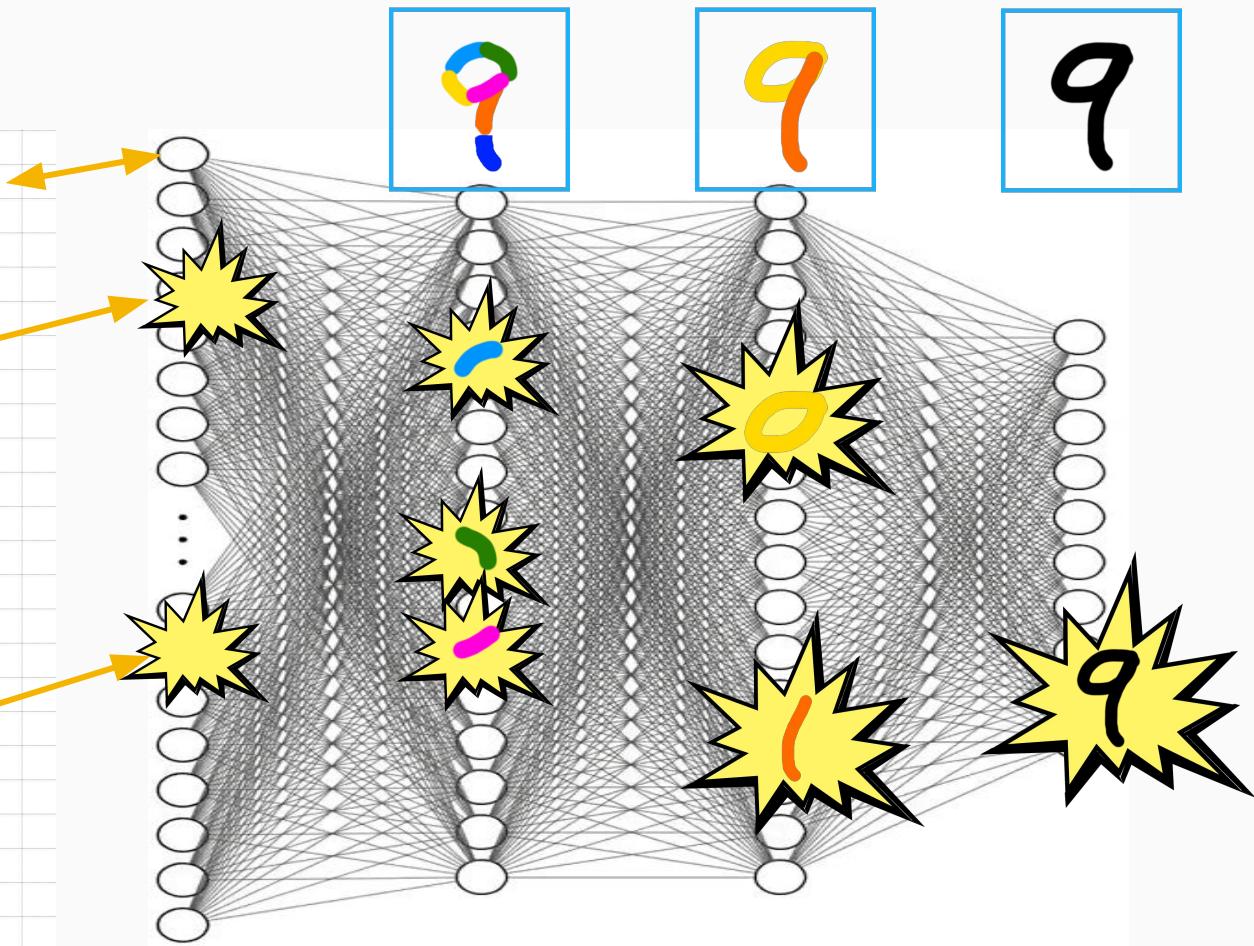
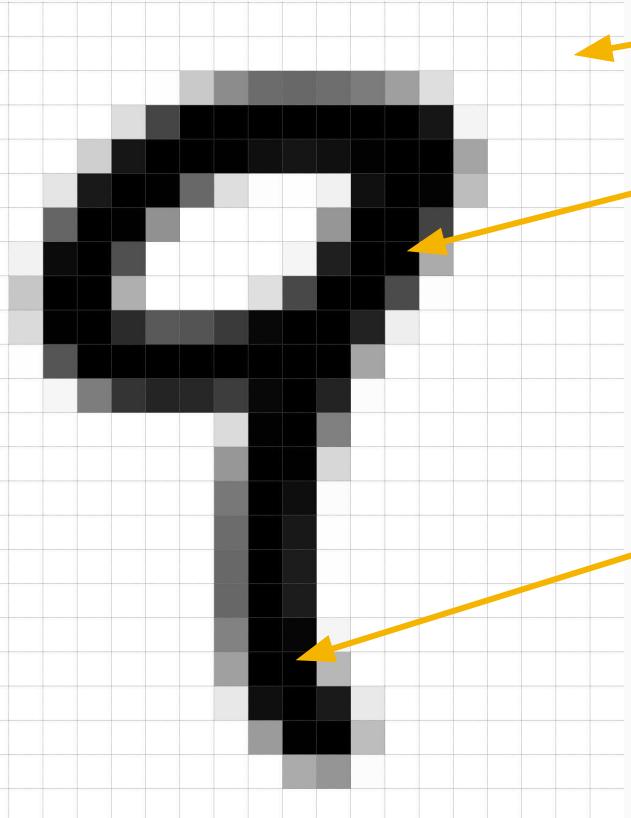


Different inputs will **excite** different pathways in the network



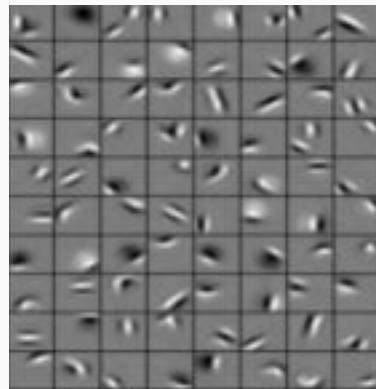
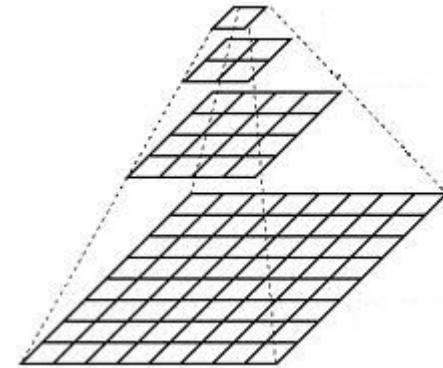
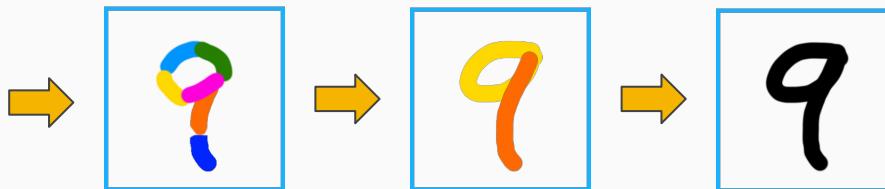
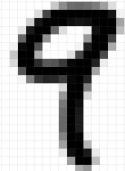


Networks consider granular details, in order to recognize **larger patterns**.



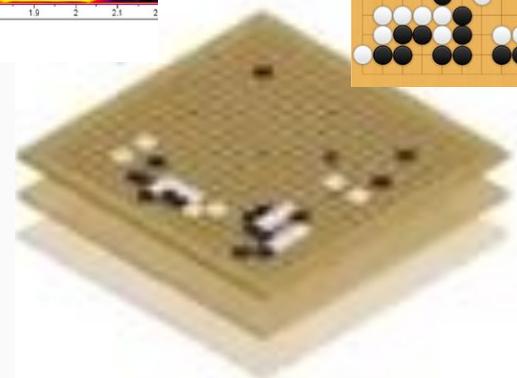
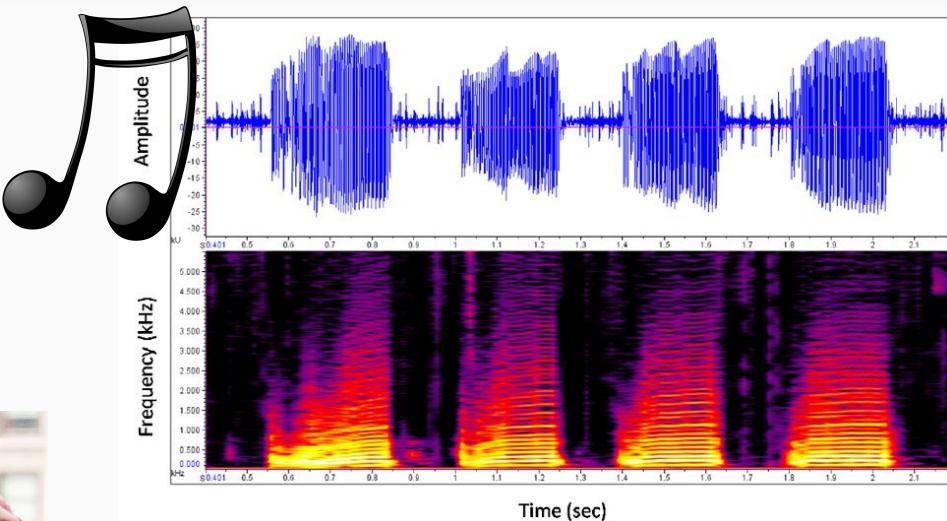
Networks consider granular details, in order to recognize **larger patterns**.

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



Nested patterns
are everywhere.

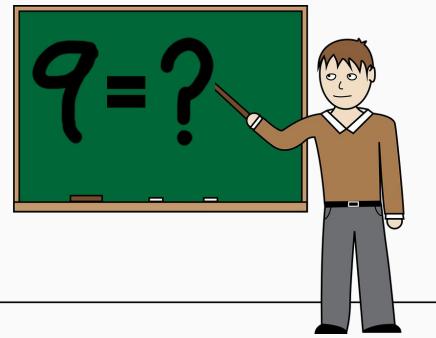
Neural Networks “see” the patterns in data like an image...



Question:

How does a neural network
know which pathways are
important?

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



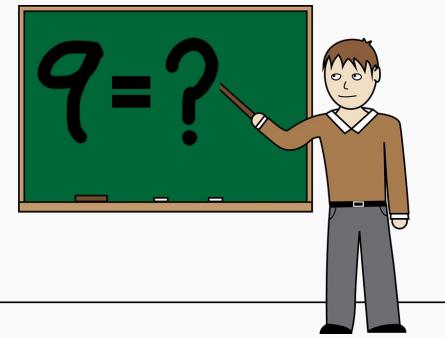
6

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9



This is a slow, data hungry evolution



6

8



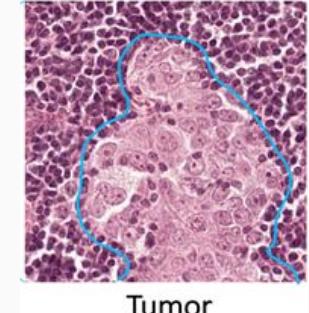
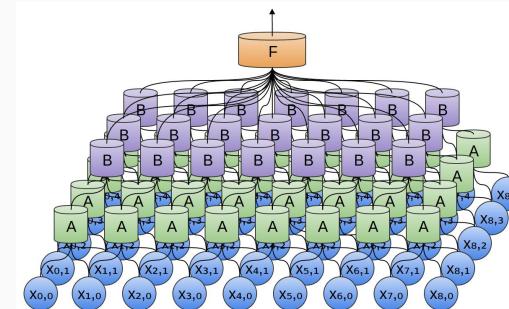
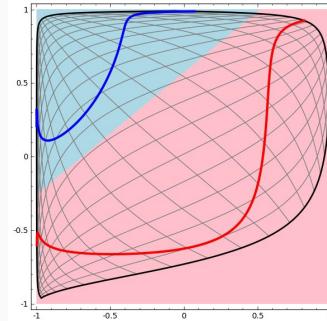
Learning by
memorization
requires many
examples.

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

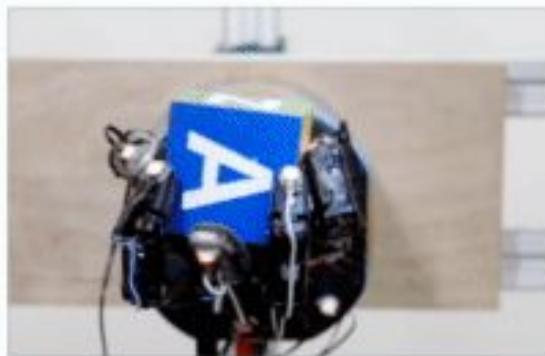
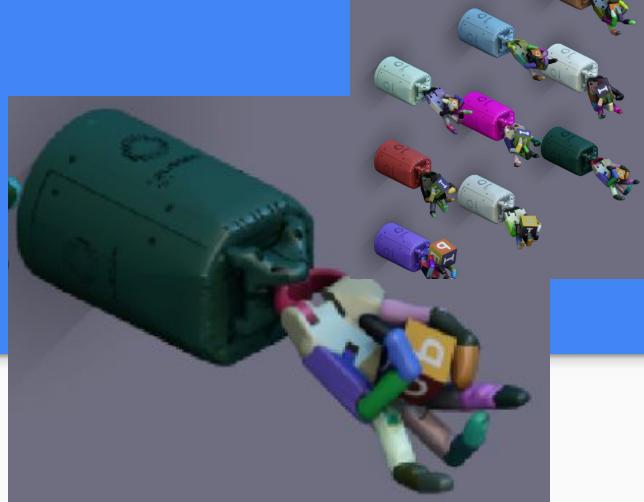
- Learning by “doing”
- Goal driven
- Attempting actions and evaluating outcomes
 - Simulated or real
- Data is created, NOT provided
 - Removes need for existing data

Reinforcement learning involves exploration

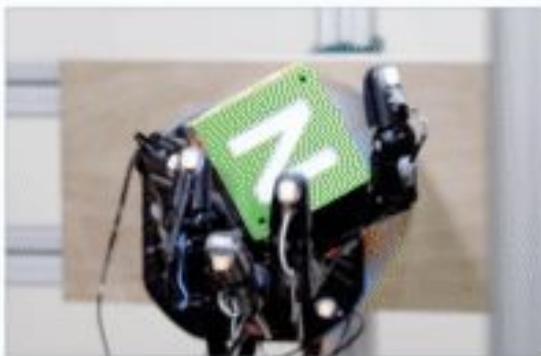
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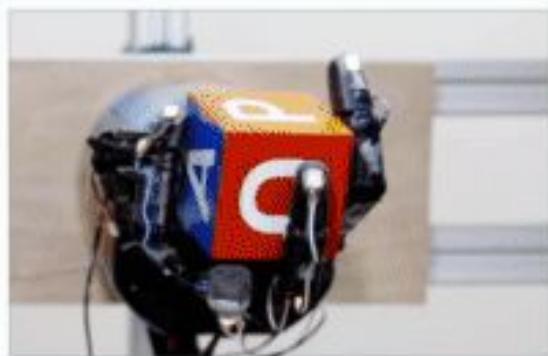
**Explore safely in simulation,
Transfer knowledge to reality**



FINGER PIVOTING



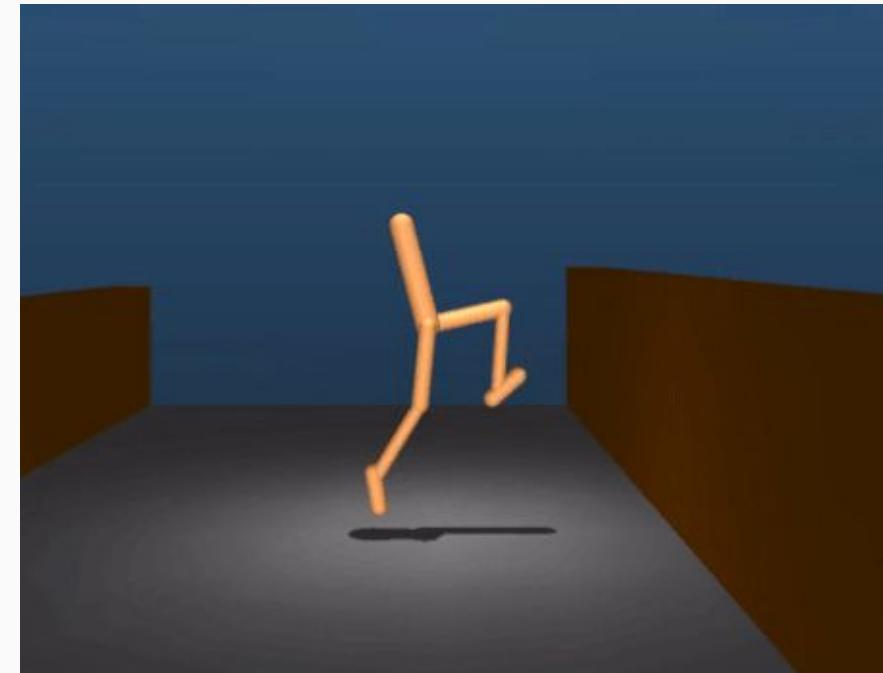
SLIDING



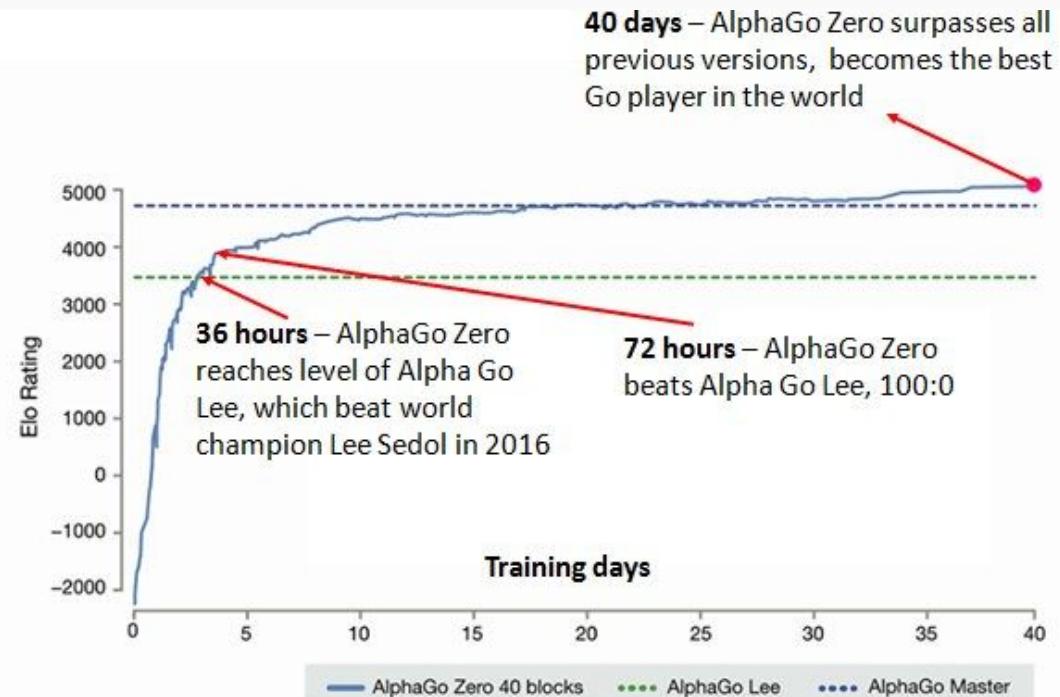
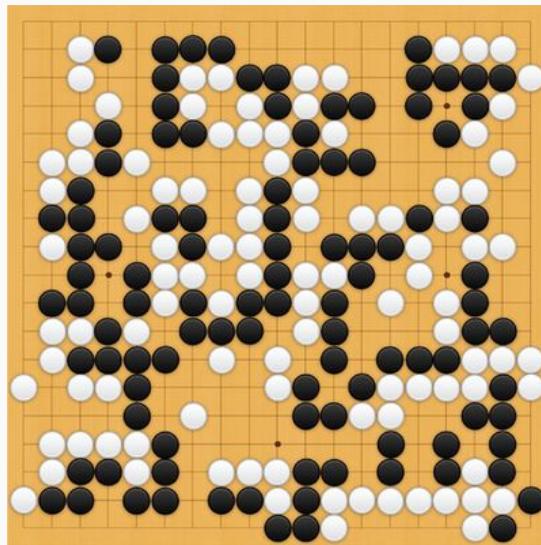
FINGER GAITING

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

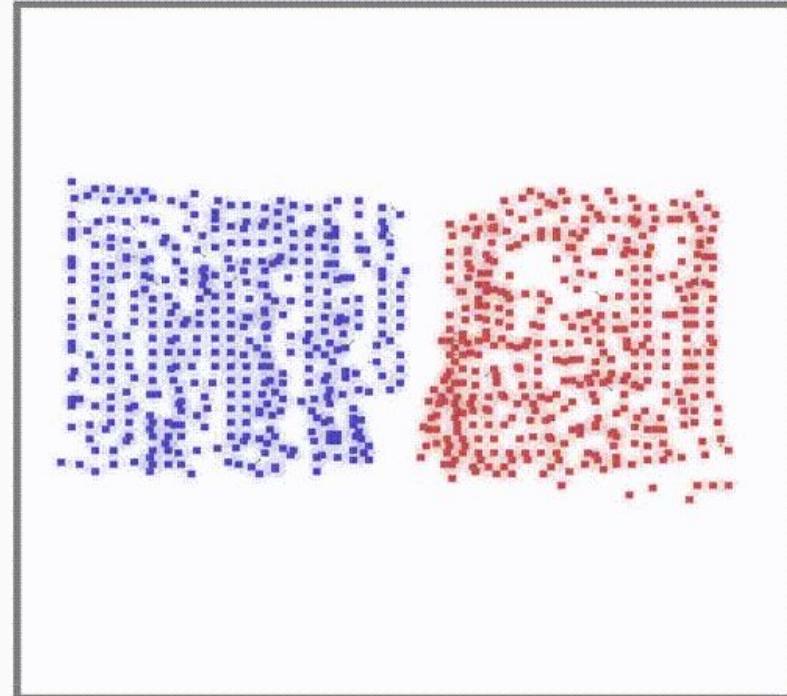
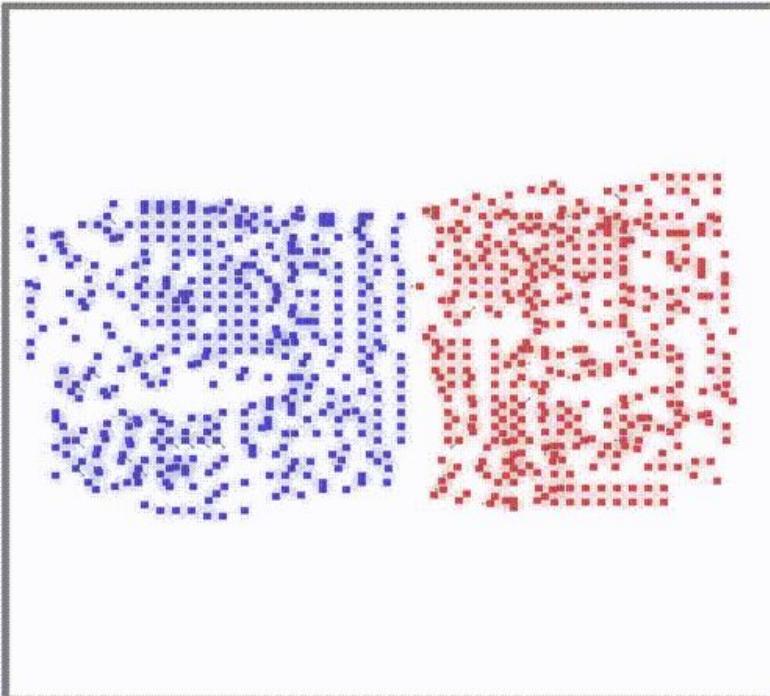
and gaps.



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**

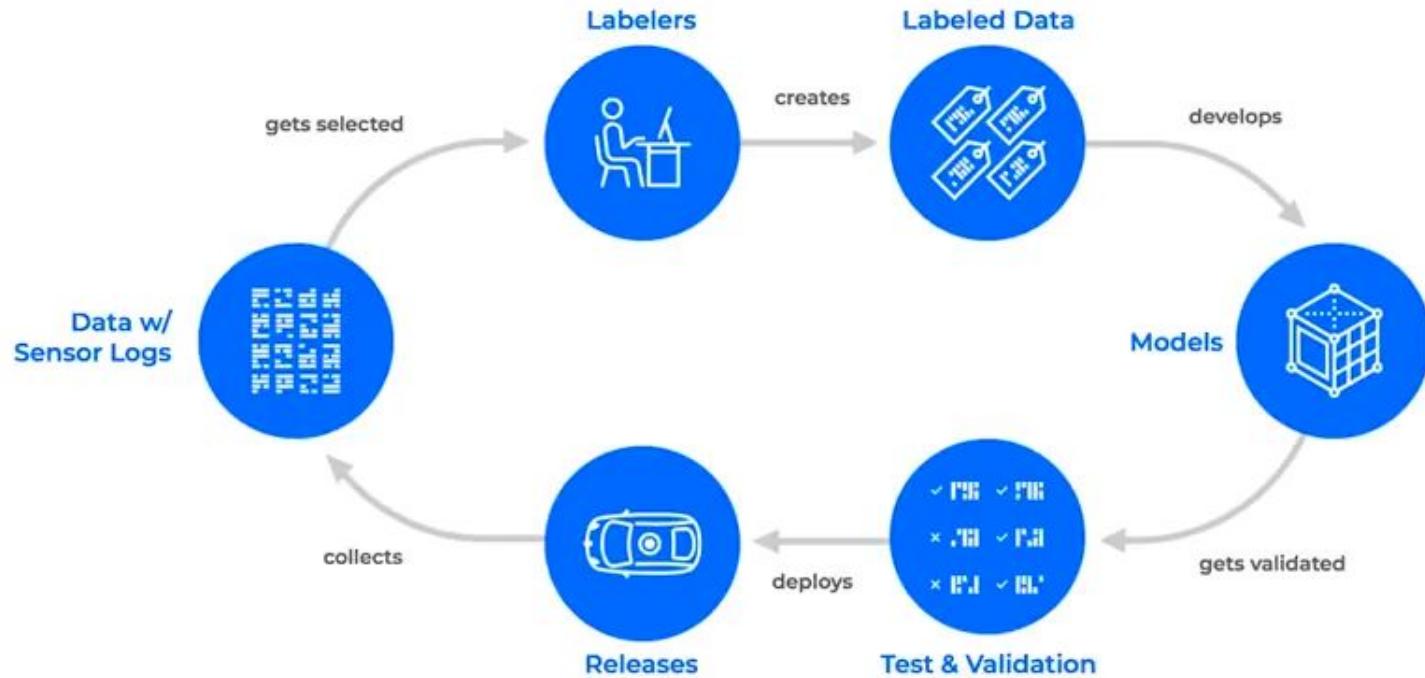
Neural Networks **Demystified**

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

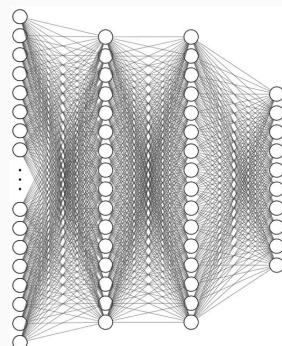
- Labeling
- Simulation
- Gyms
- Visualization
- Dev Tools
- Networking

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

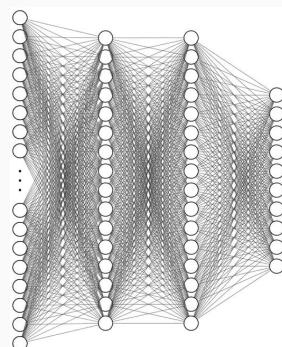
Complex
Input



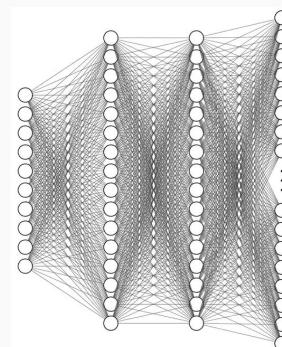
Concise
Output

Translation

Complex
Input



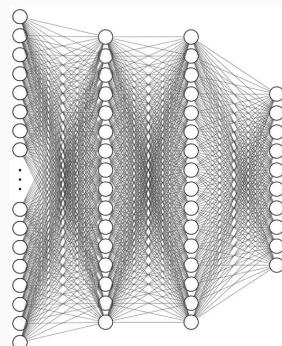
Concise
Format



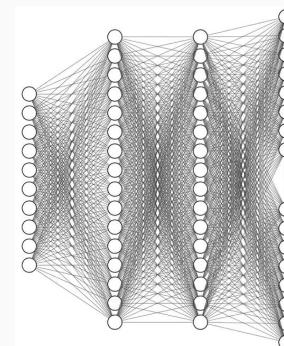
Complex
Output

Translation

Complex
Input



Translation

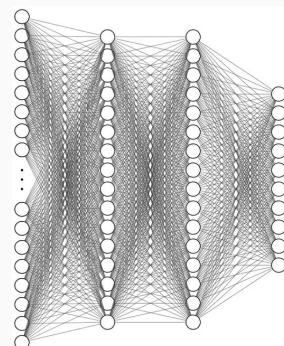


Complex
Output

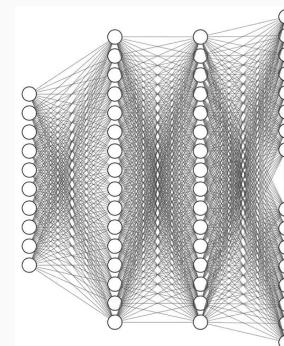
Translation

Complex
Input

hello



Translation

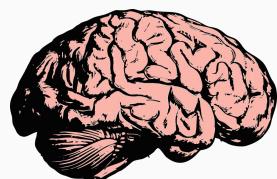
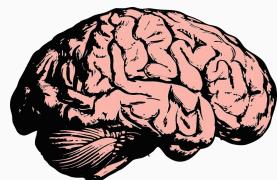


Complex
Output

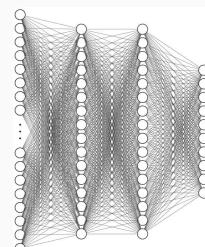
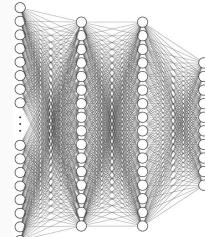
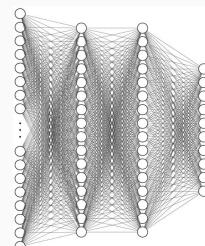
你好

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

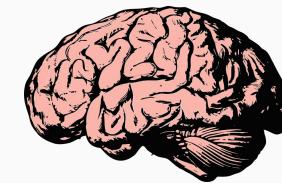
Complex Input



Translation

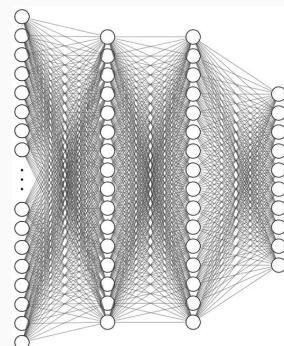


Complex Output

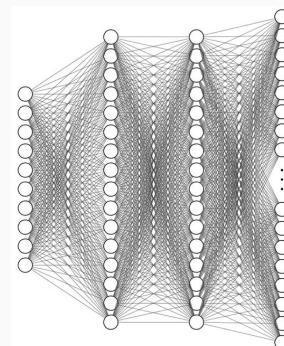


Translation

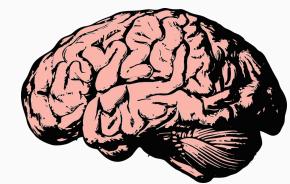
Complex
Input



Translation

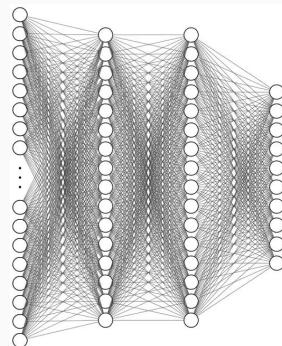


Complex
Output

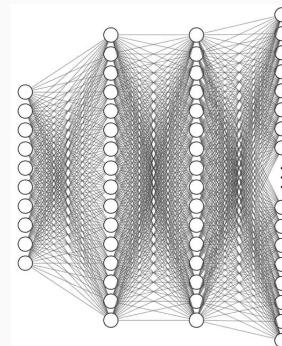


Translation

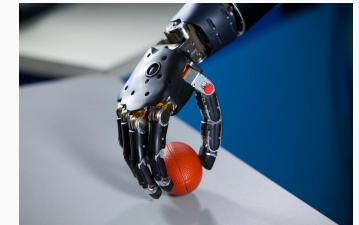
Complex
Input



Translation

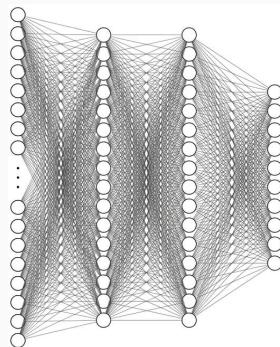
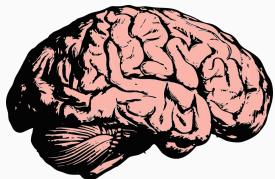


Complex
Output

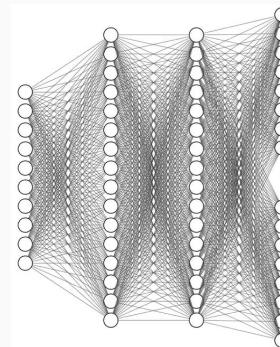


Translation

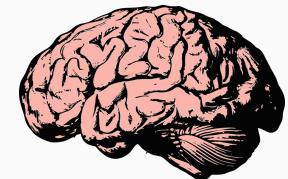
Complex
Input



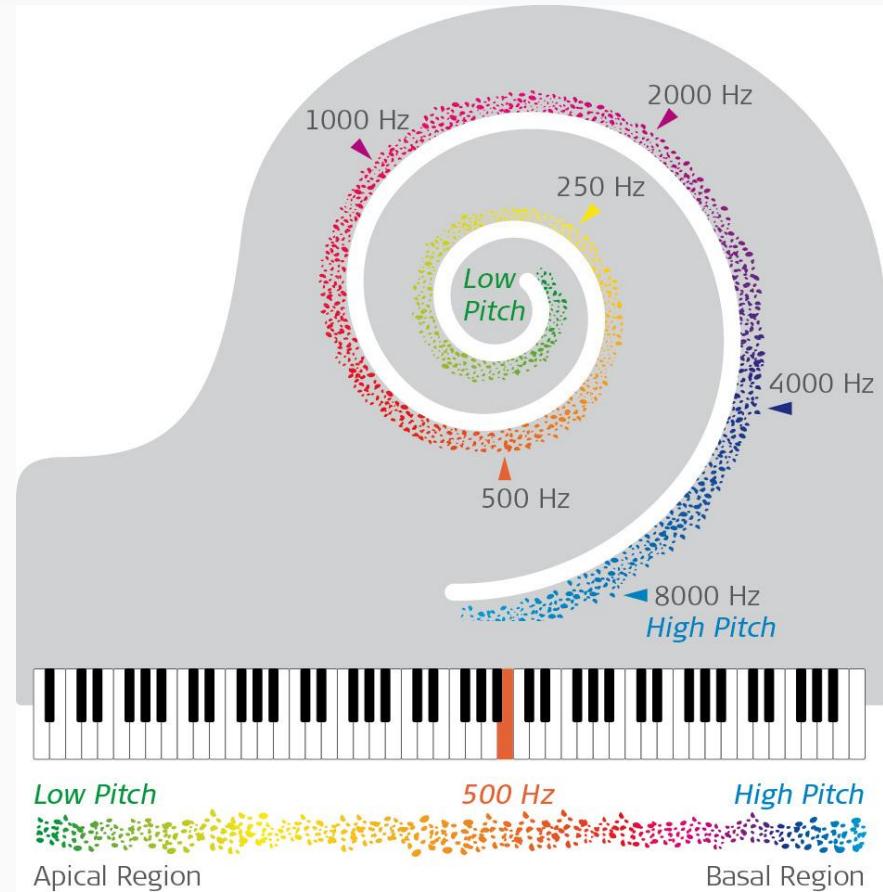
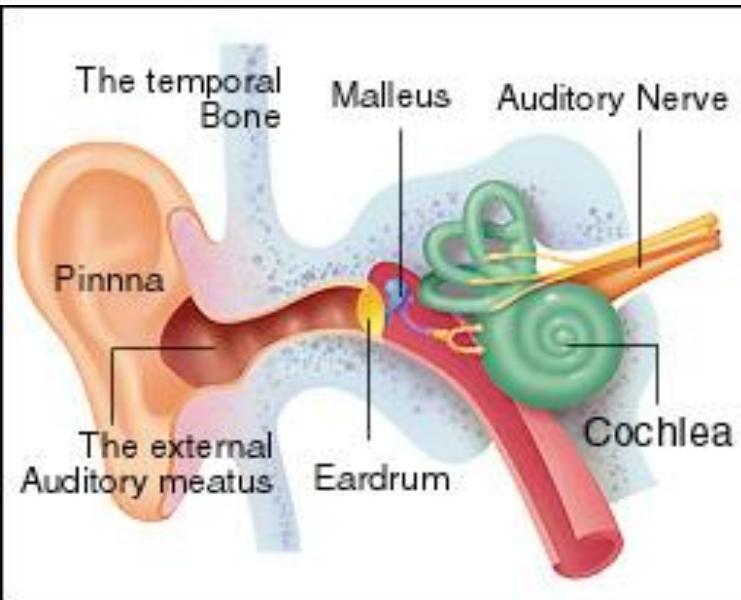
Translation



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