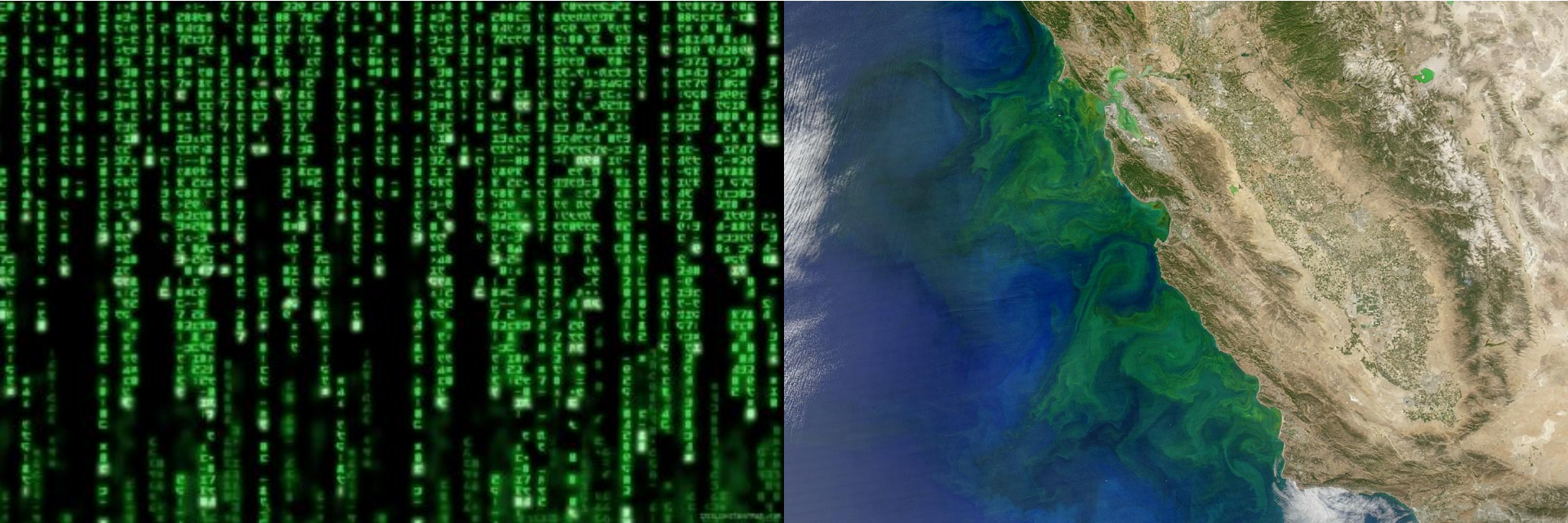


Machine Learning and Ocean Optics

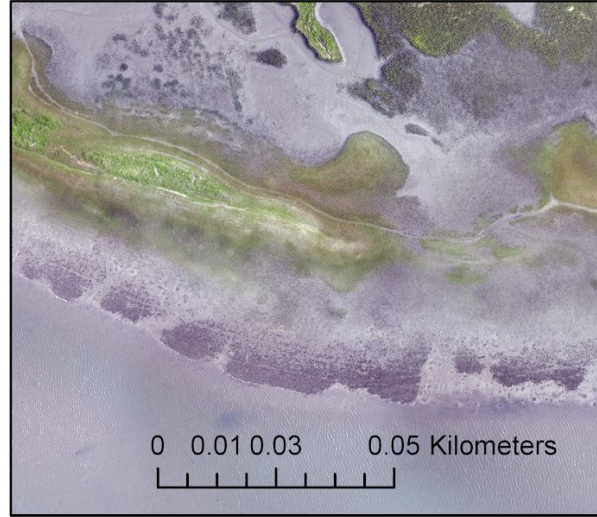


Patrick Gray
Ocean Optics Class 2023

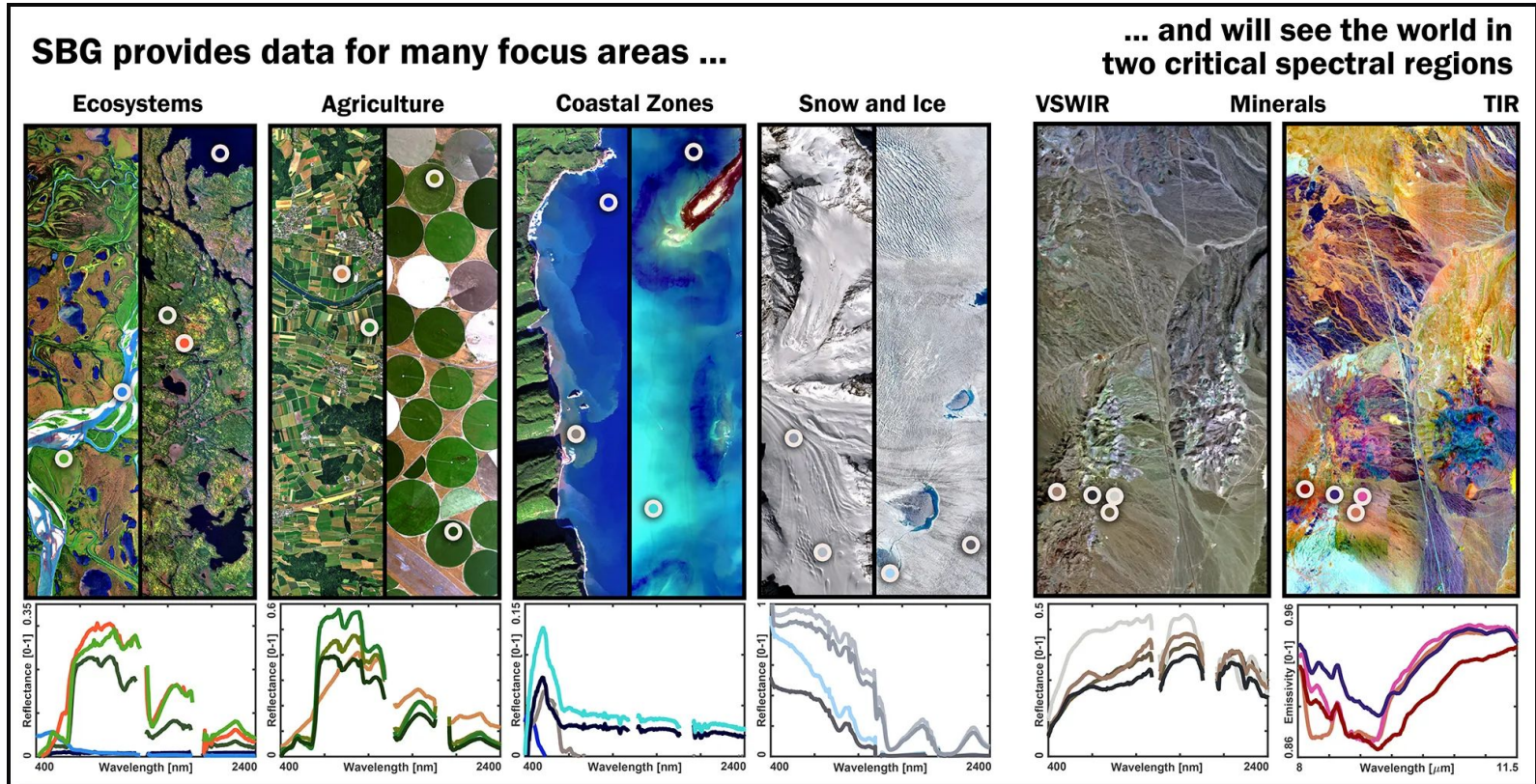
Machine Learning

ML algorithms perform a specific task without using explicit instructions, relying on patterns and inference instead.

Why Machine Learning for Remote Sensing

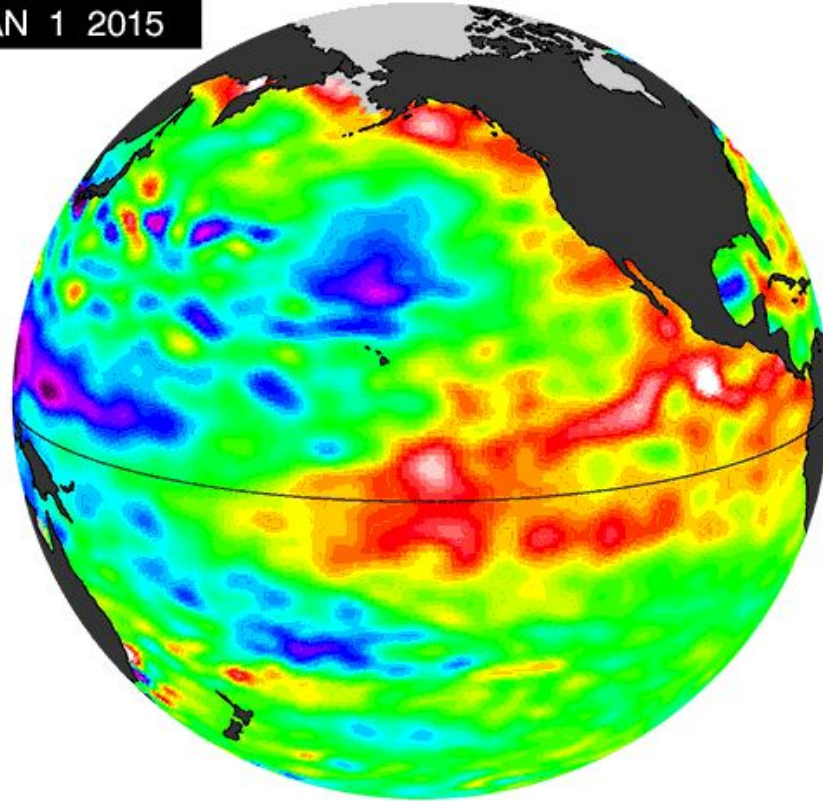


Why Machine Learning for Remote Sensing

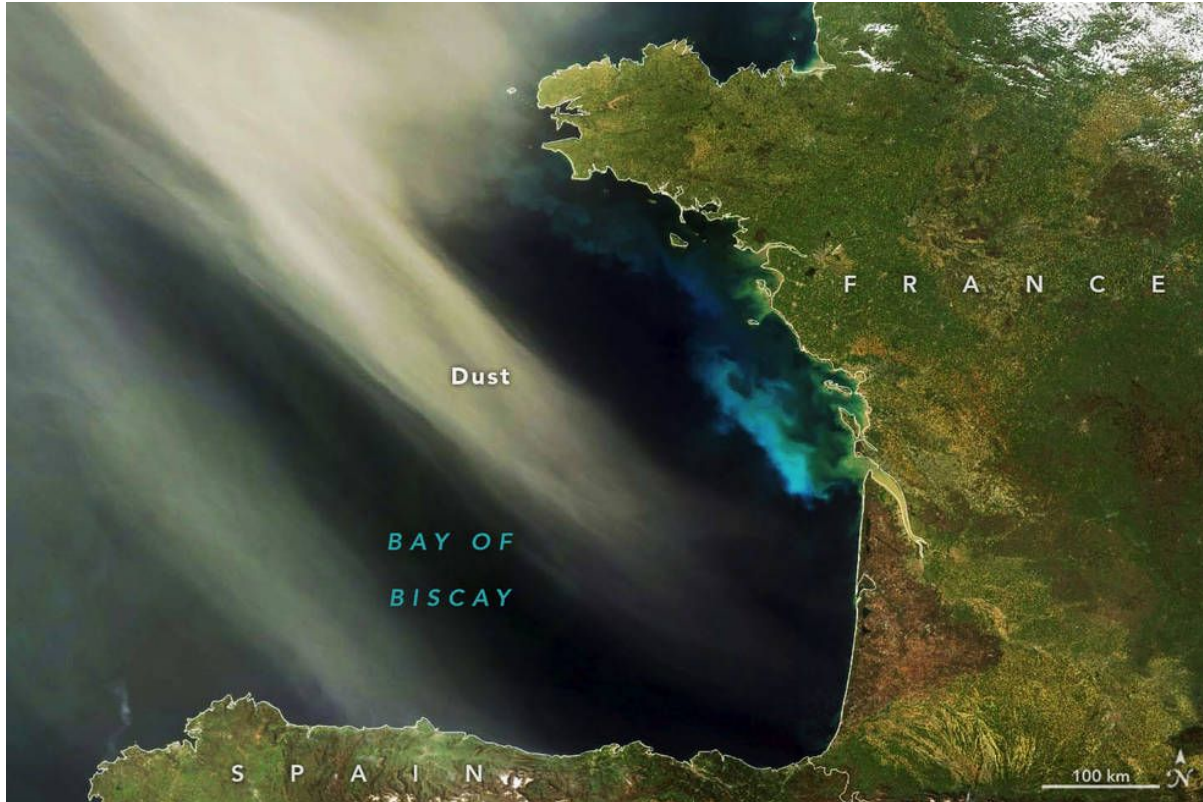


Why Machine Learning for Remote Sensing

JAN 1 2015



Why Machine Learning for Remote Sensing



What does machine learning do?

At its most basic ML systems classify input data, cluster data, recognize patterns, and regress data into models

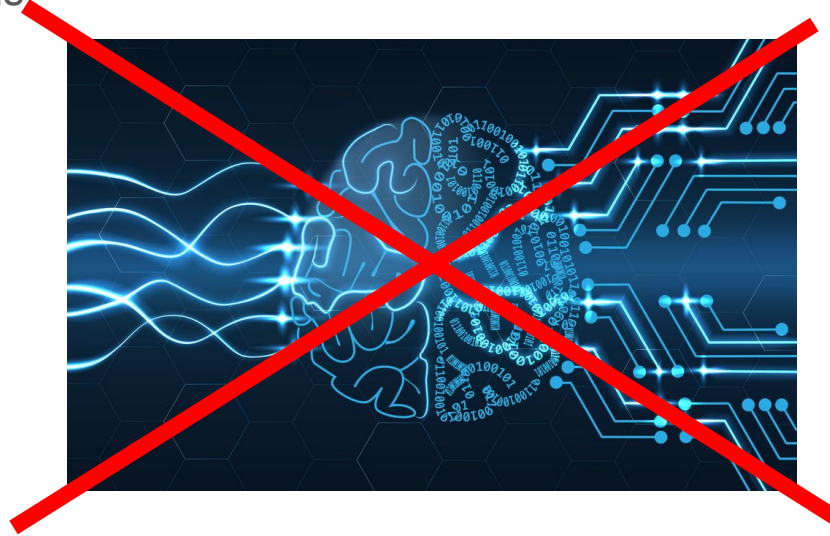
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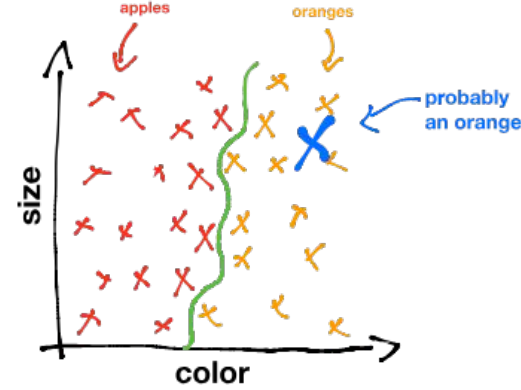
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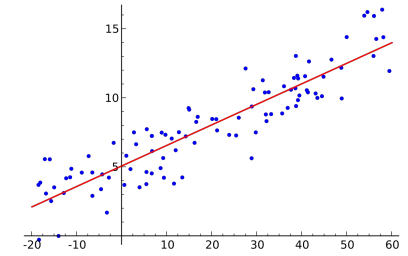
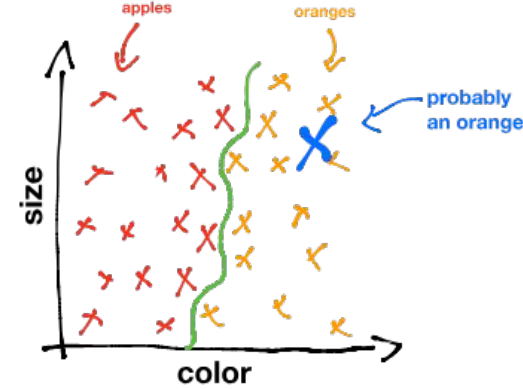
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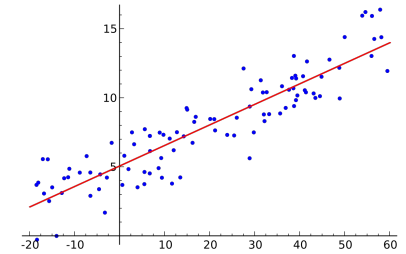
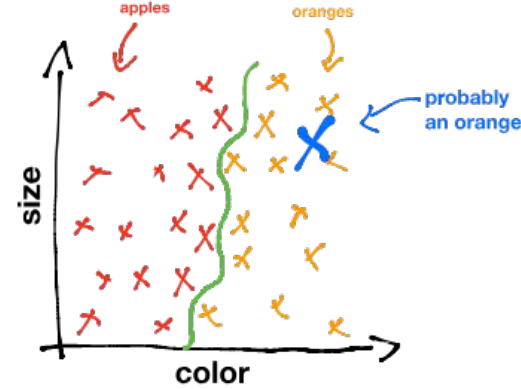
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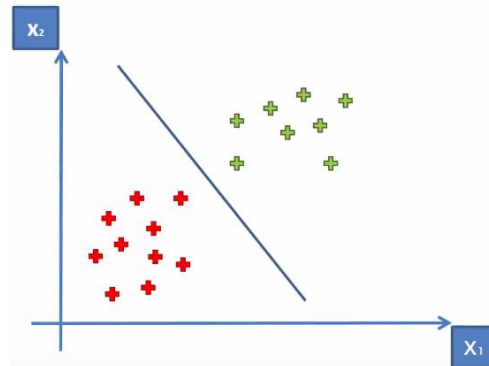


What does machine learning do?

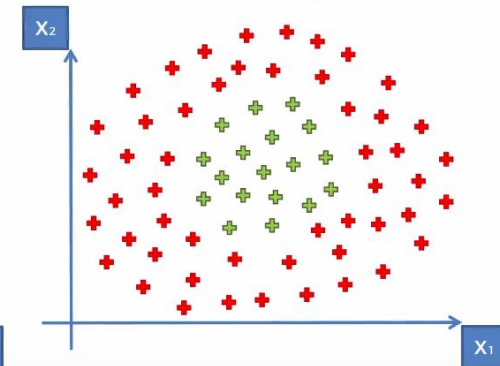
At its most basic ML systems classify input data, cluster data, recognize patterns, and regress data into models



Linearly Separable



Not Linearly Separable

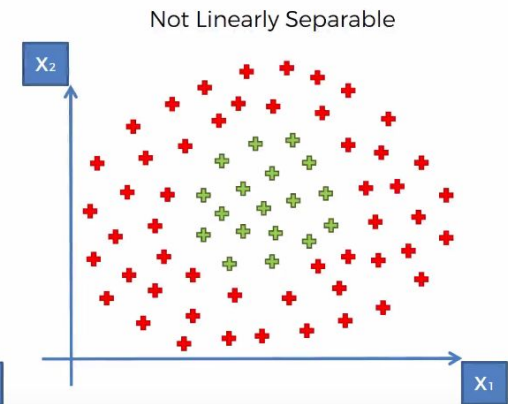
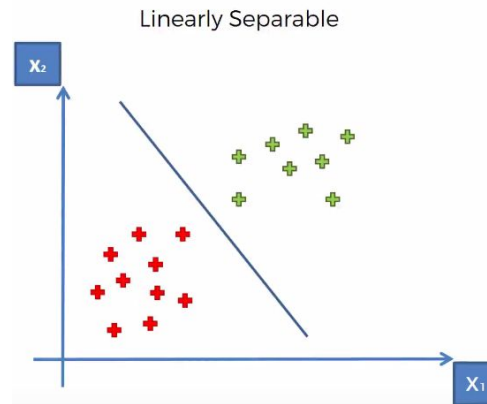
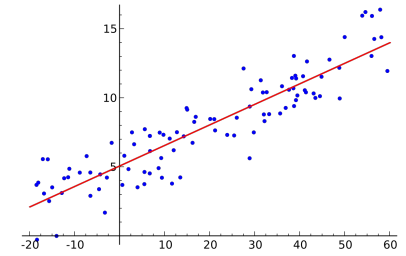
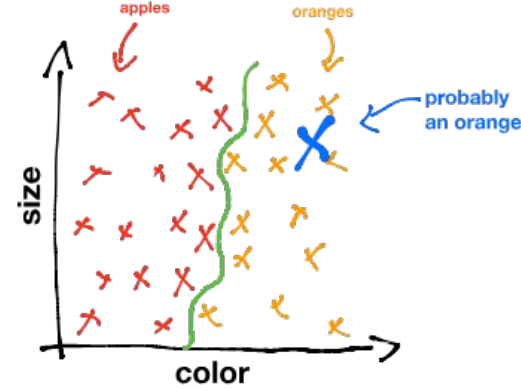


What does machine learning do?

At its most basic ML systems classify input data, cluster data, recognize patterns, and regress data into models

This could mean:

- face recognition
- predicting home prices
- modeling animal habitat
- **estimating chl-a**
- predict what you will buy
- **filling cloud gaps**
- predict netflix preferences
- **predicting PCC**



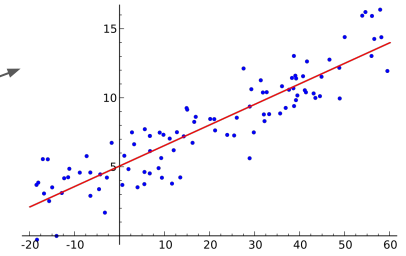
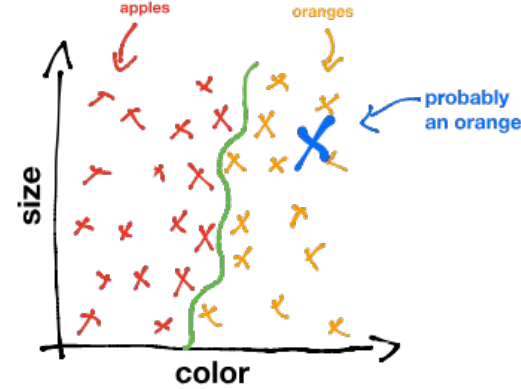
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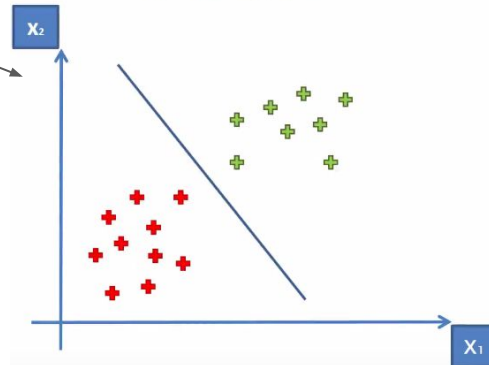
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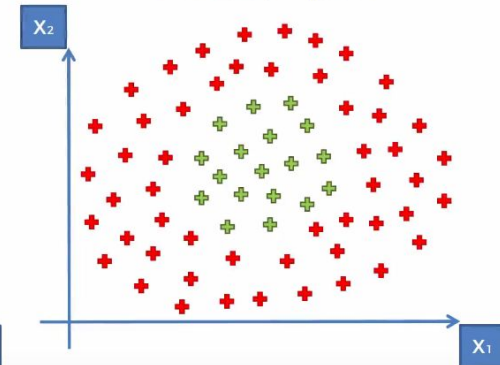
Just high-dimensional

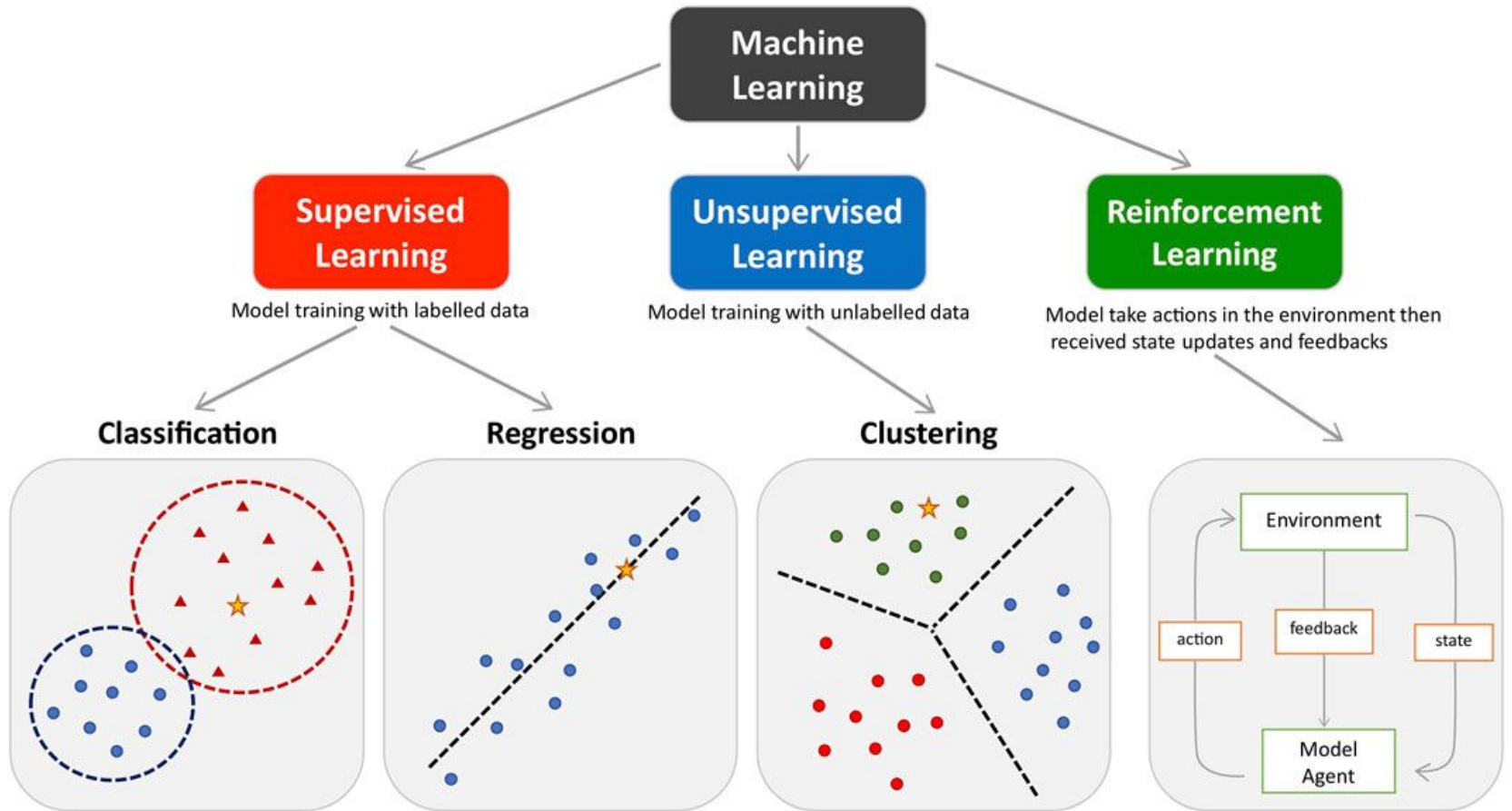


Linearly Separable

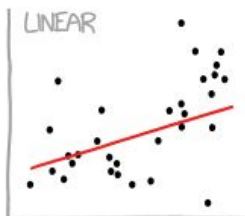


Not Linearly Separable





CURVE-FITTING METHODS AND THE MESSAGES THEY SEND



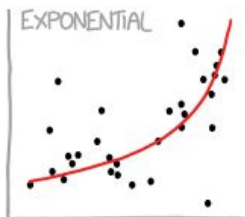
"HEY, I DID A
REGRESSION."



"I WANTED A CURVED
LINE, SO I MADE ONE
WITH MATH."



"LOOK, IT'S
TAPERING OFF!"



"LOOK, IT'S GROWING
UNCONTROLLABLY!"



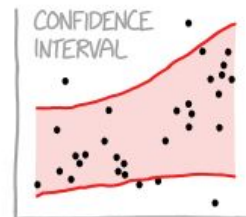
"I'M SOPHISTICATED, NOT
LIKE THOSE BUMBLING
POLYNOMIAL PEOPLE."



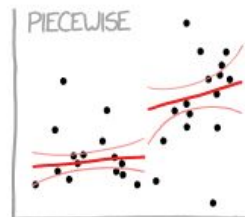
"I'M MAKING A
SCATTER PLOT BUT
I DON'T WANT TO."



"I NEED TO CONNECT THESE
TWO LINES, BUT MY FIRST IDEA
DIDN'T HAVE ENOUGH MATH."



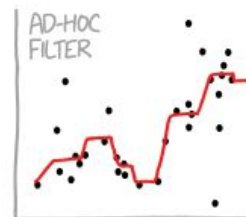
"LISTEN, SCIENCE IS HARD.
BUT I'M A SERIOUS
PERSON DOING MY BEST."



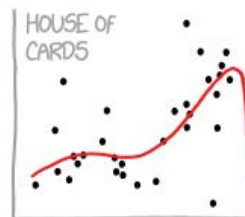
"I HAVE A THEORY,
AND THIS IS THE ONLY
DATA I COULD FIND."



"I CLICKED 'SMOOTH
LINES' IN EXCEL."

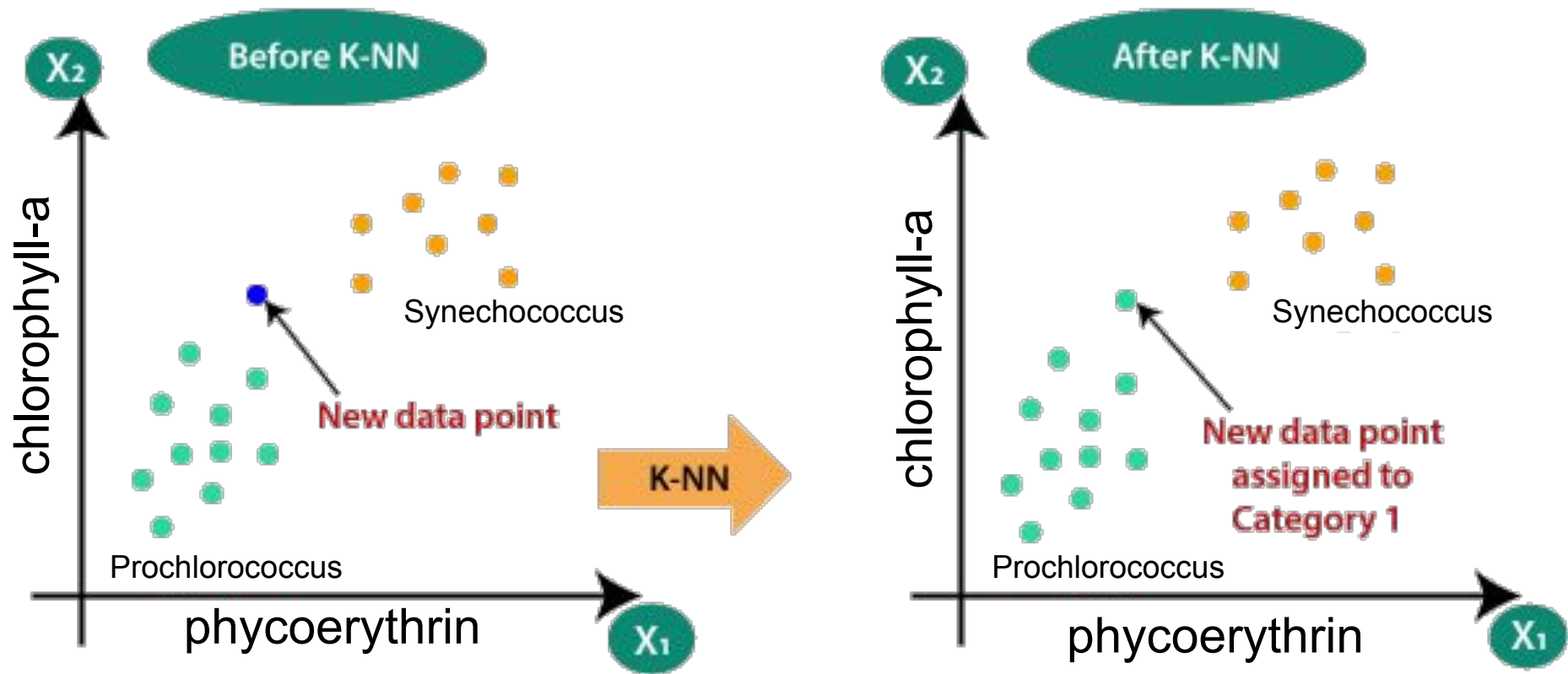


"I HAD AN IDEA FOR HOW
TO CLEAN UP THE DATA.
WHAT DO YOU THINK?"

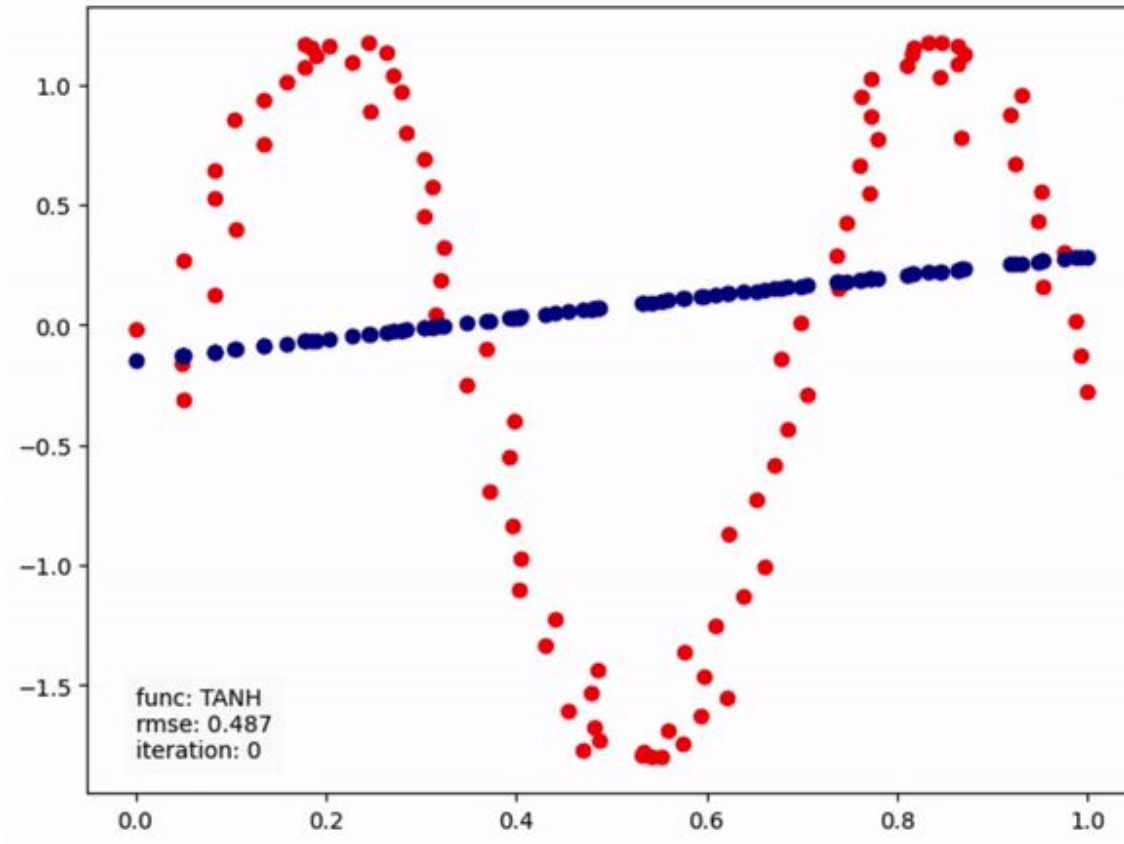


"AS YOU CAN SEE, THIS
MODEL SMOOTHLY FITS
THE- WAIT NO NO DON'T
EXTEND IT AAAAAA!!!"

Classify phytoplankton types

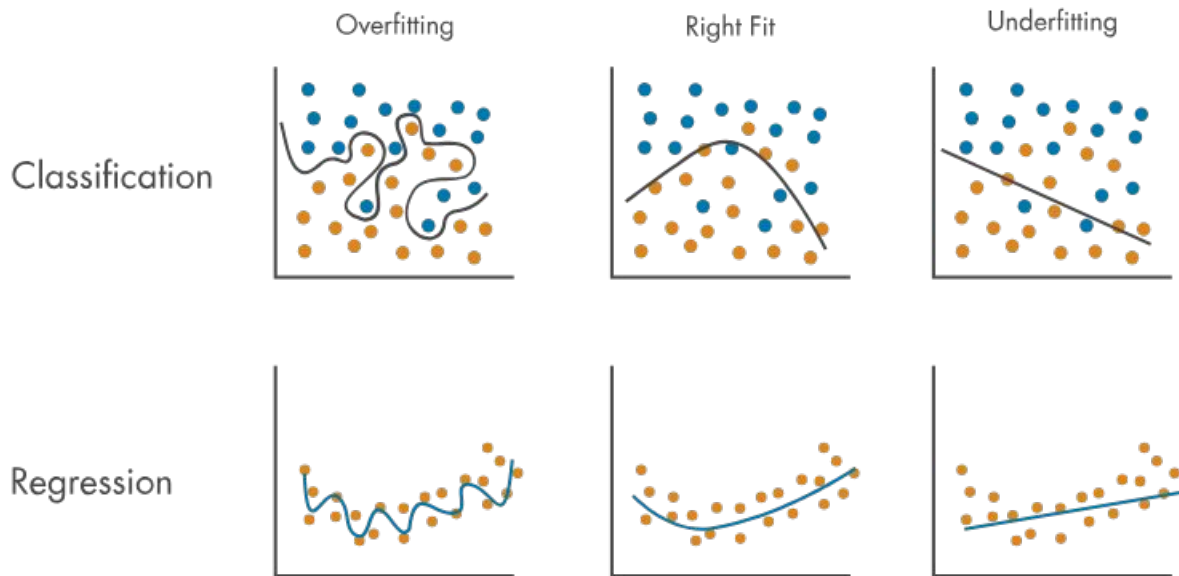


Regress chlorophyll-a



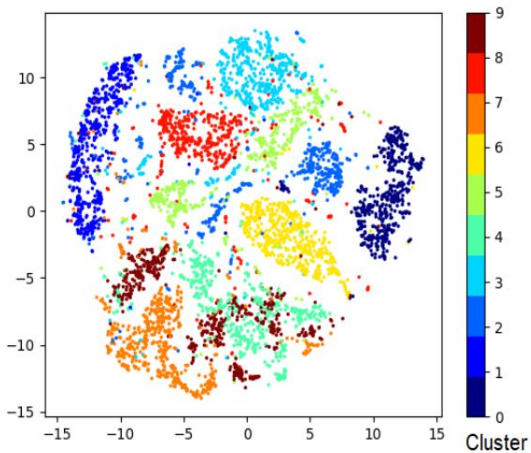
“All models are wrong, some are useful”

- some ML models can be wrong in wildly unpredictable ways
- must aggressively test and convince yourself it is robust

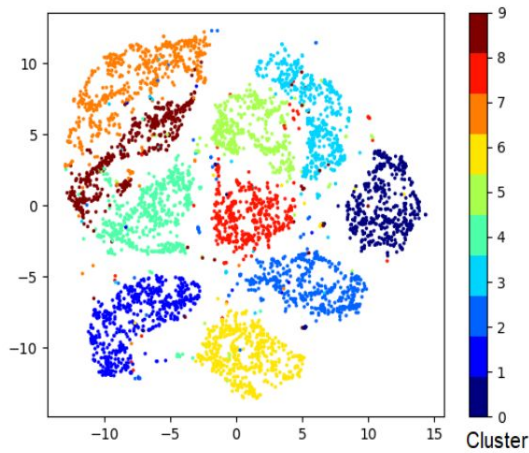


Clustering

Finding spatial/spectral/temporal patterns, outliers, etc

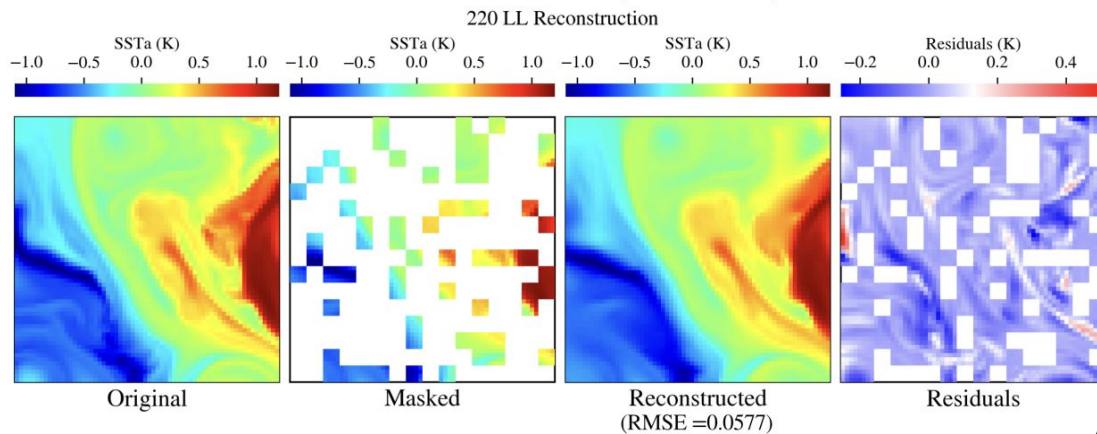
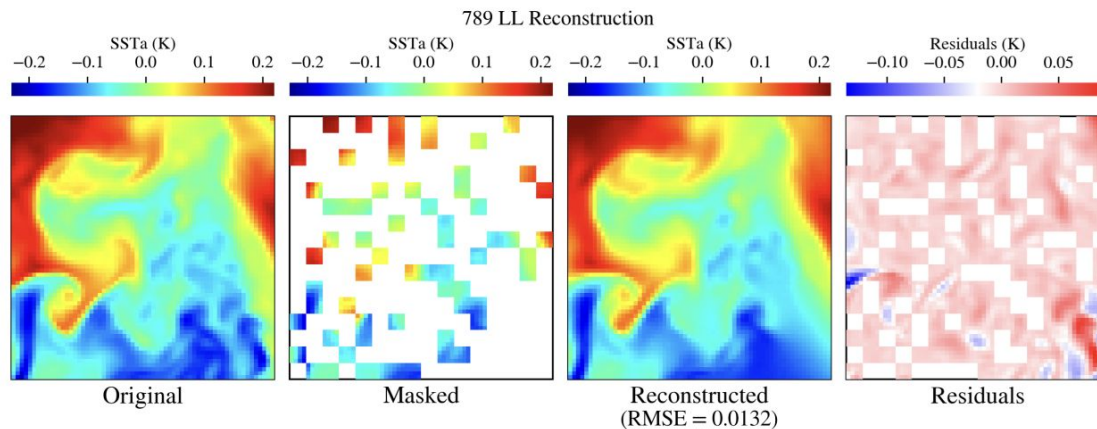


(a) k-Means

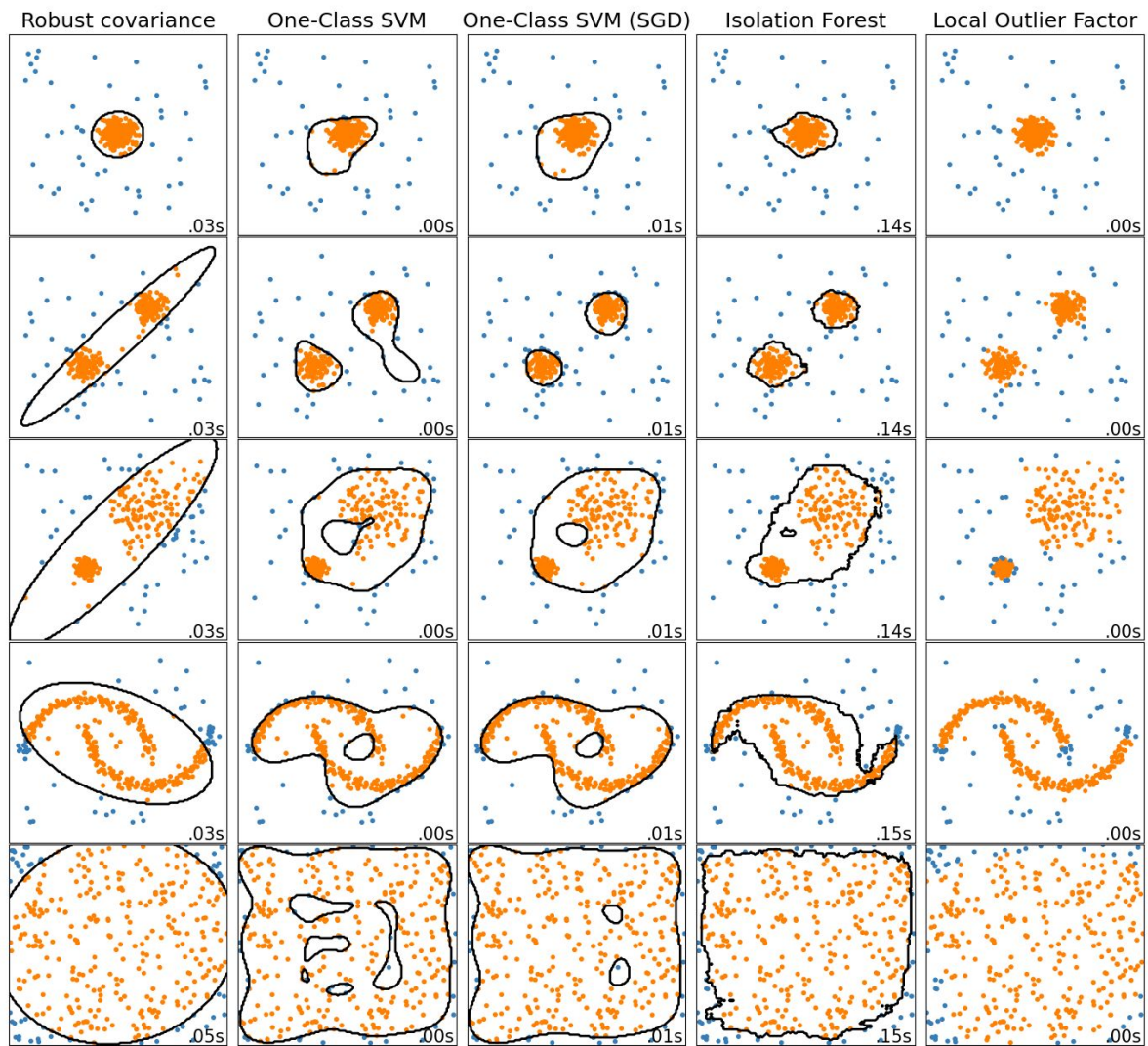


(b) Autoencoder + k-Means

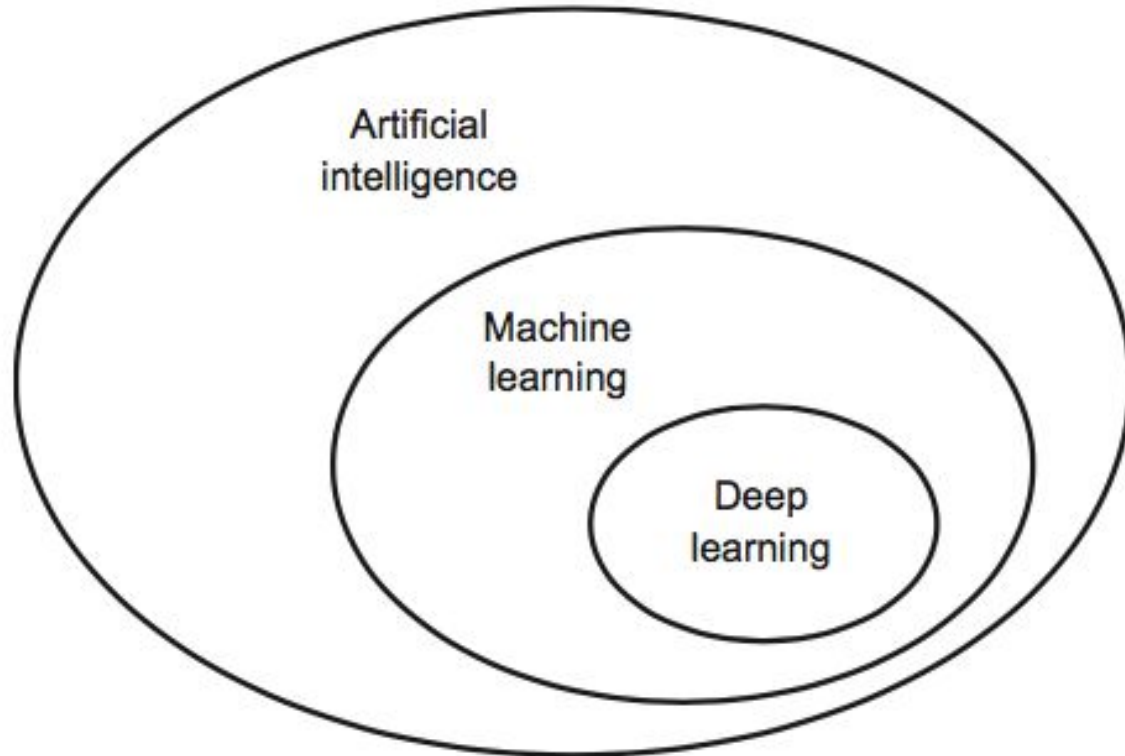
Interpolation



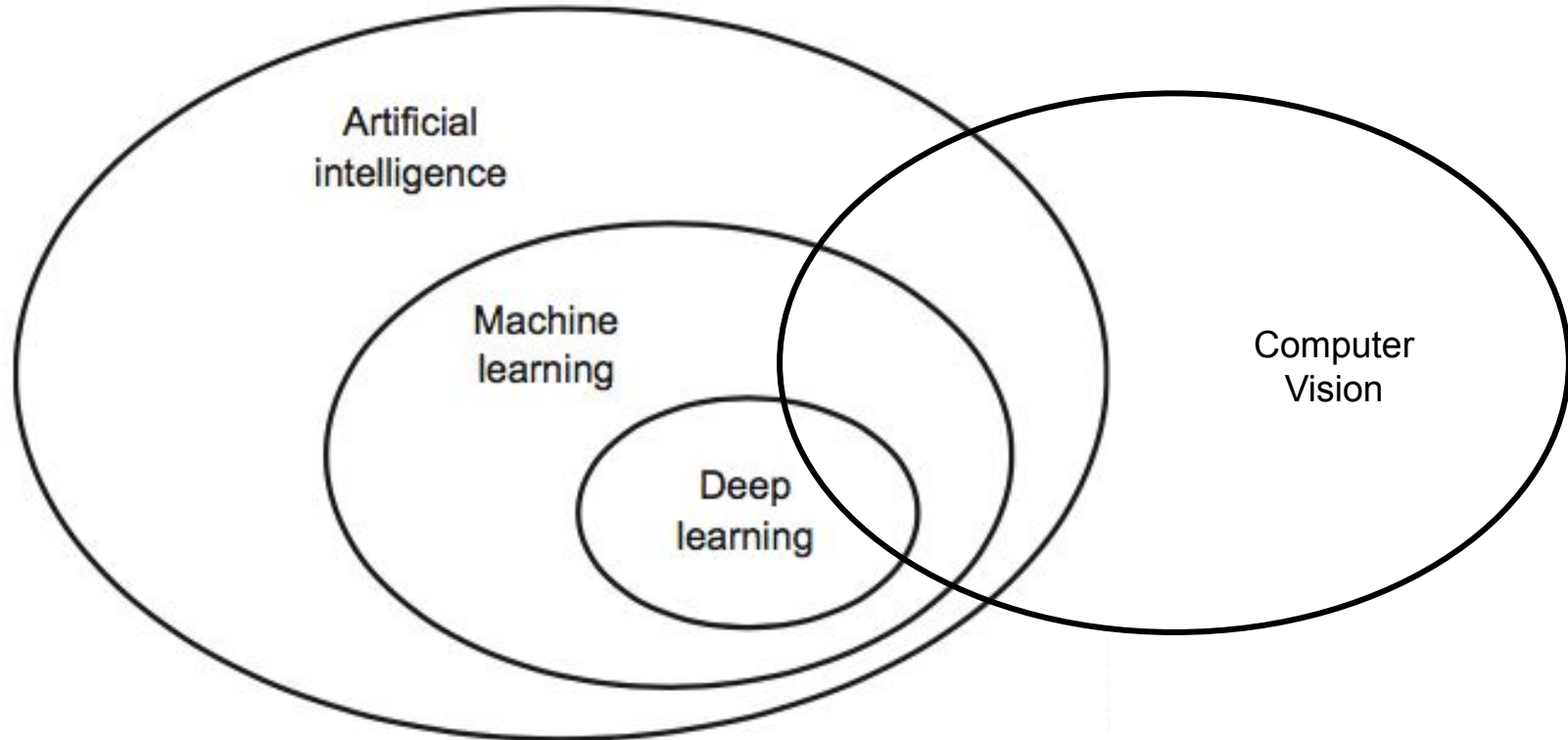
Outlier Detection



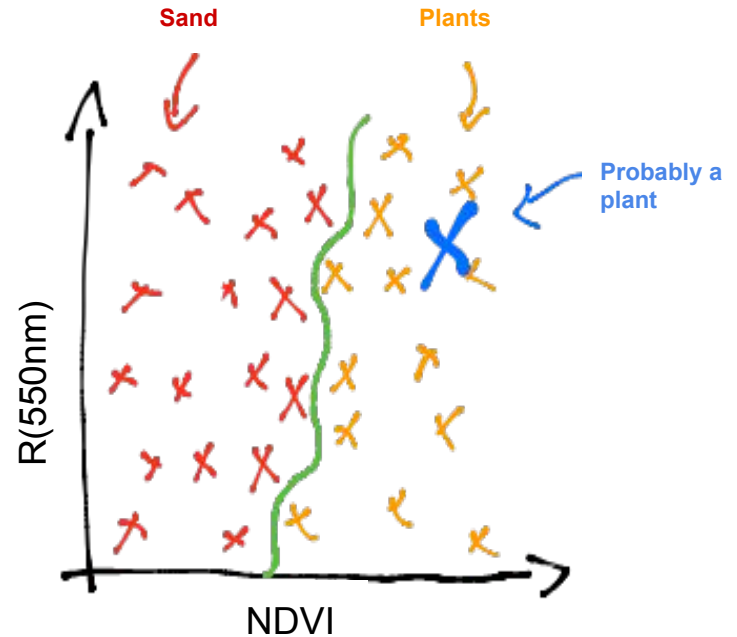
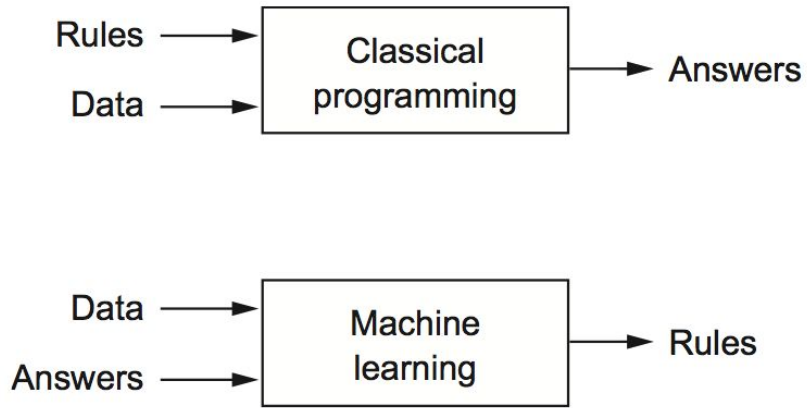
So what is Deep Learning?



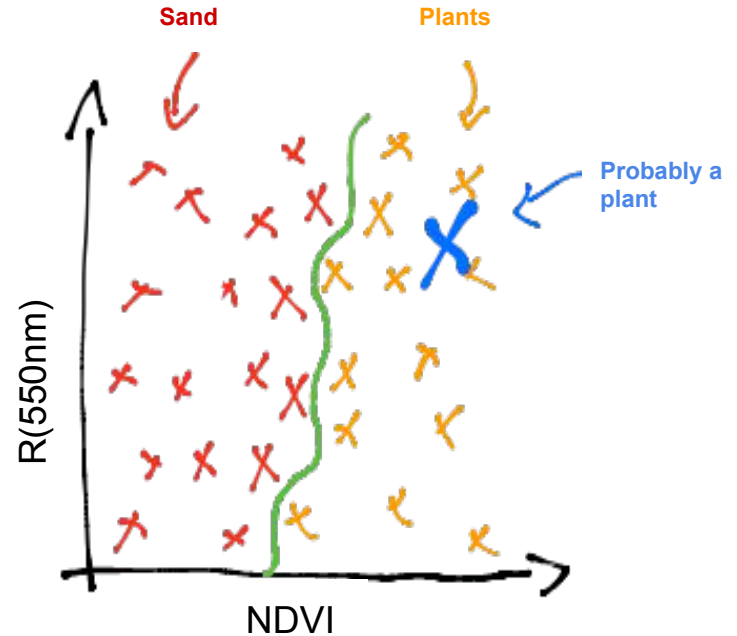
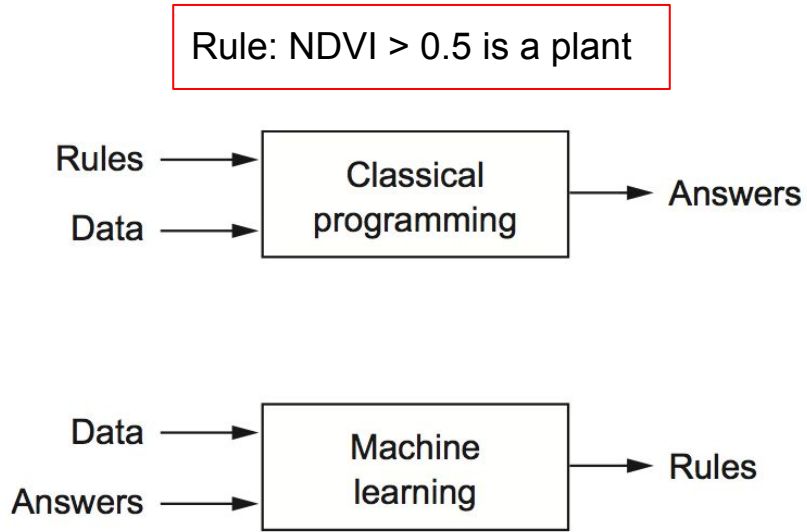
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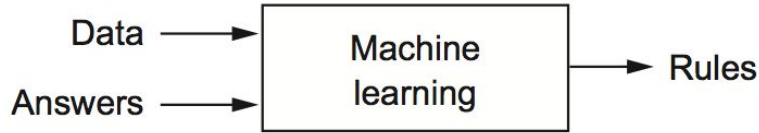
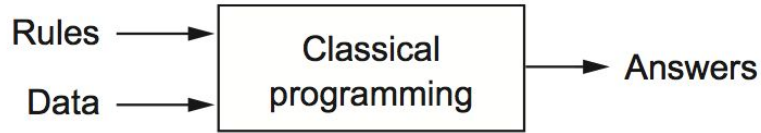


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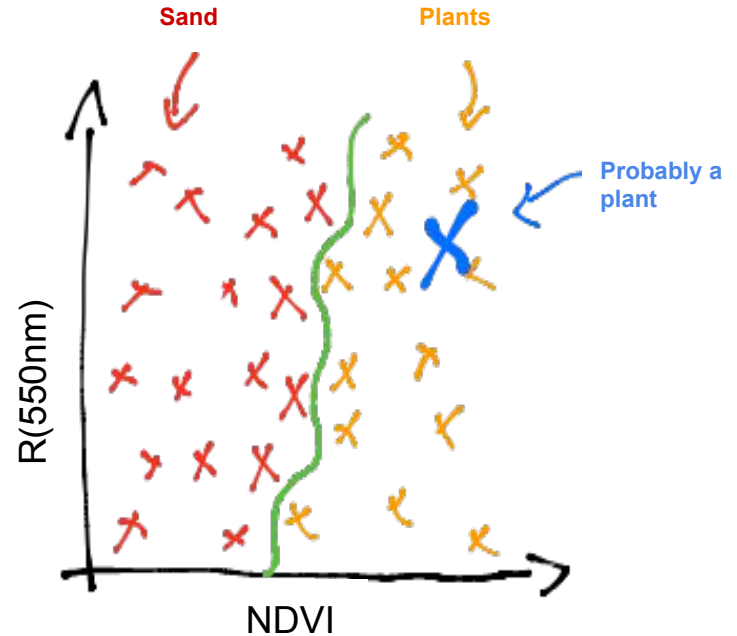


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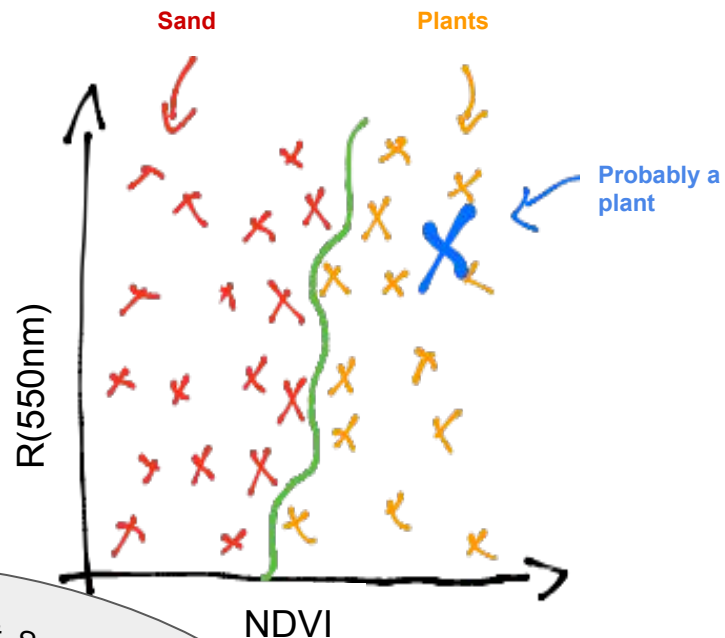
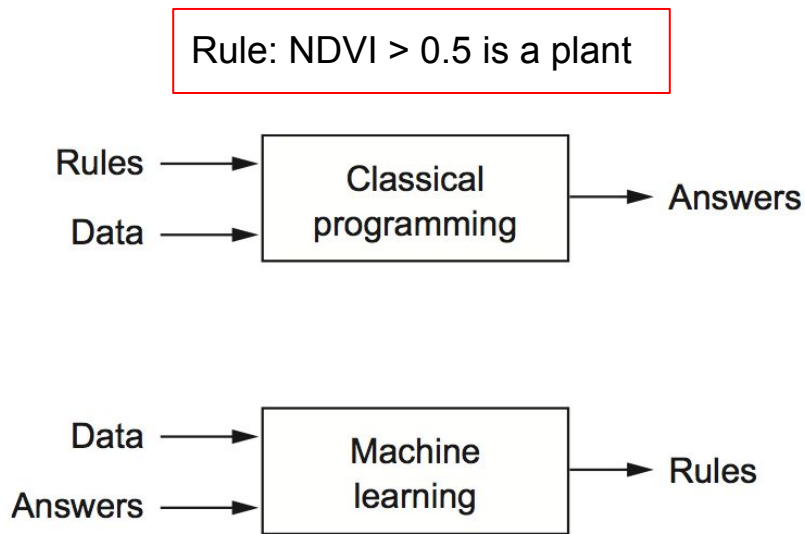
Rule: $\text{NDVI} > 0.5$ is a plant



Instead: give it NDVI and pixels you know are plants



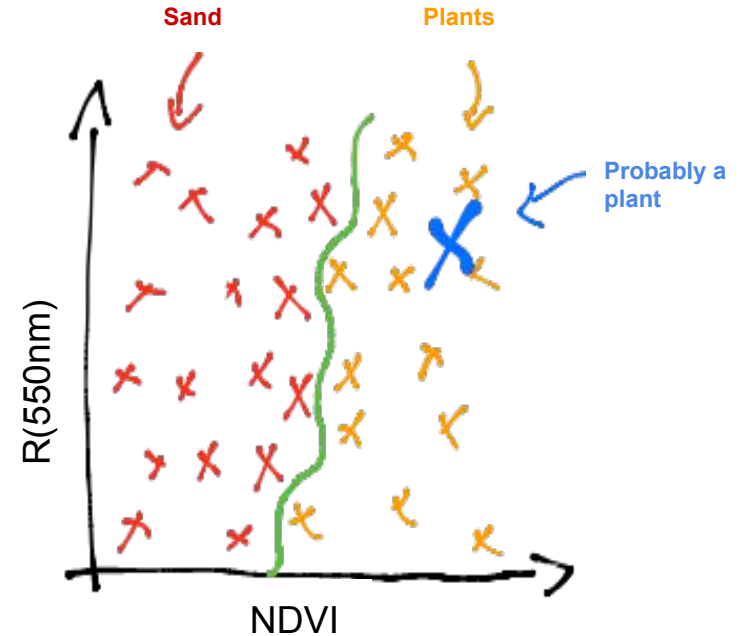
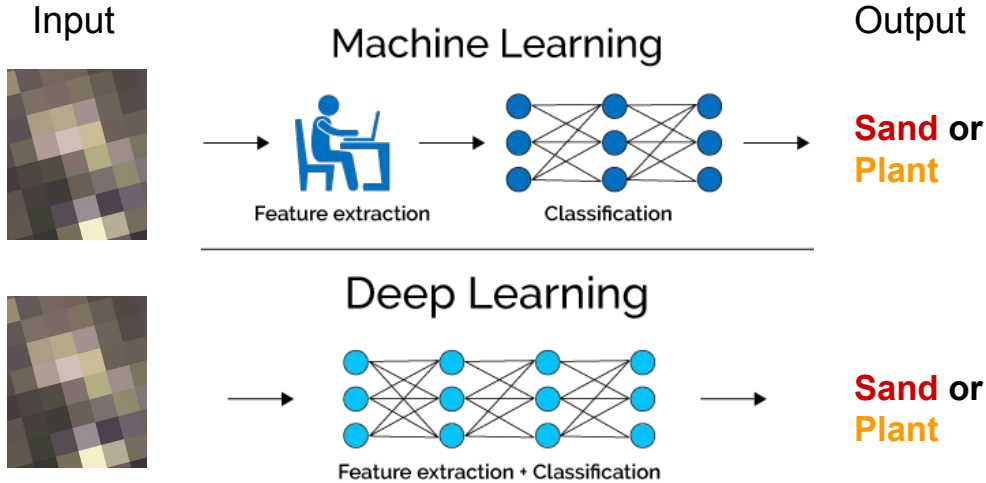
So what is Deep Learning?



Instead: give it NDVI and labeled pixels you know are plants

Random Forest, Support Vector Machine, Naive Bayes, Max Likelihood, K-Nearest Neighbor, AdaBoost, Spectral Angle Mapper, etc

So what is Deep Learning?



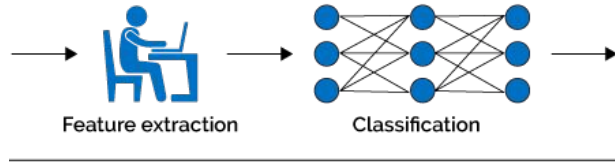
So what is Deep Learning?

Feature: NDVI

Input



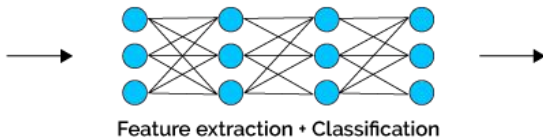
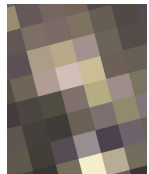
Machine Learning



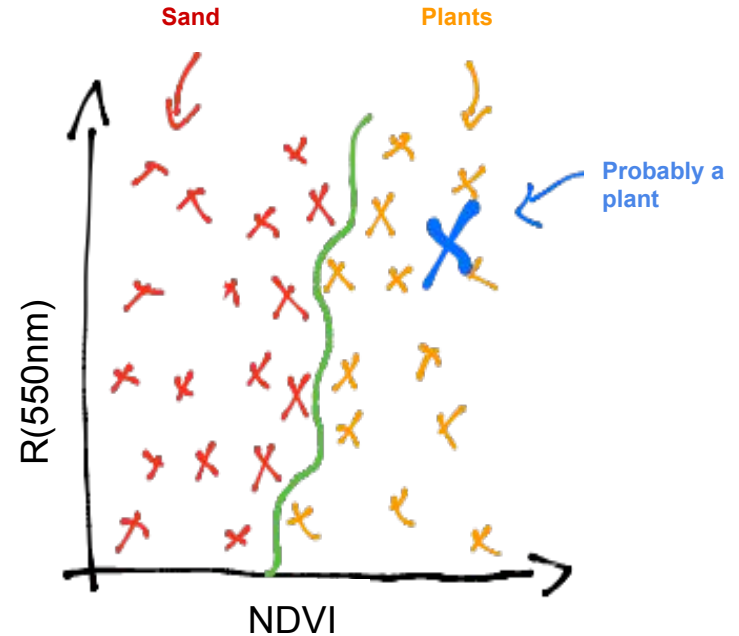
Output

Sand or Plant

Deep Learning



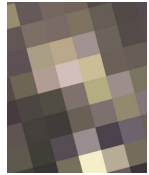
Sand or Plant



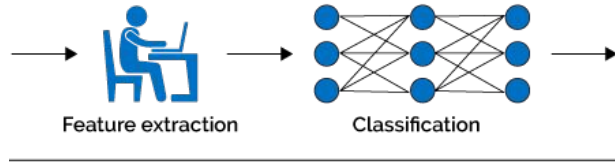
So what is Deep Learning?

Feature: NDVI

Input



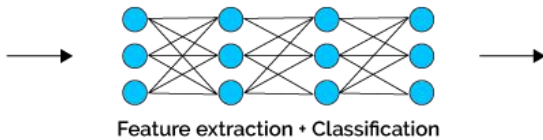
Machine Learning



Output

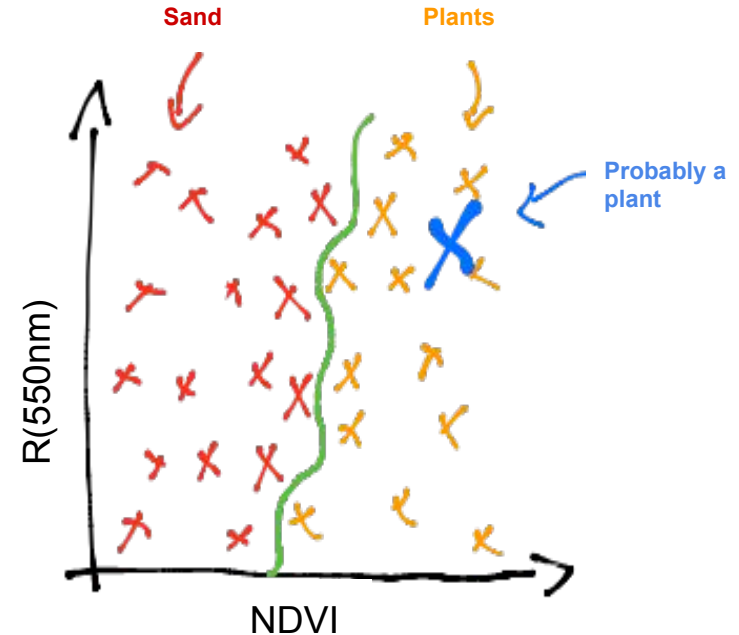
Sand or Plant

Deep Learning



Sand or Plant

Instead: give it the raw spectra and let it discover important combinations itself



So what is Deep Learning?

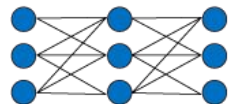
Input



Machine Learning



Feature extraction

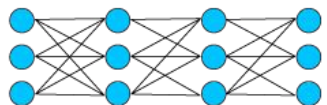


Classification

Output

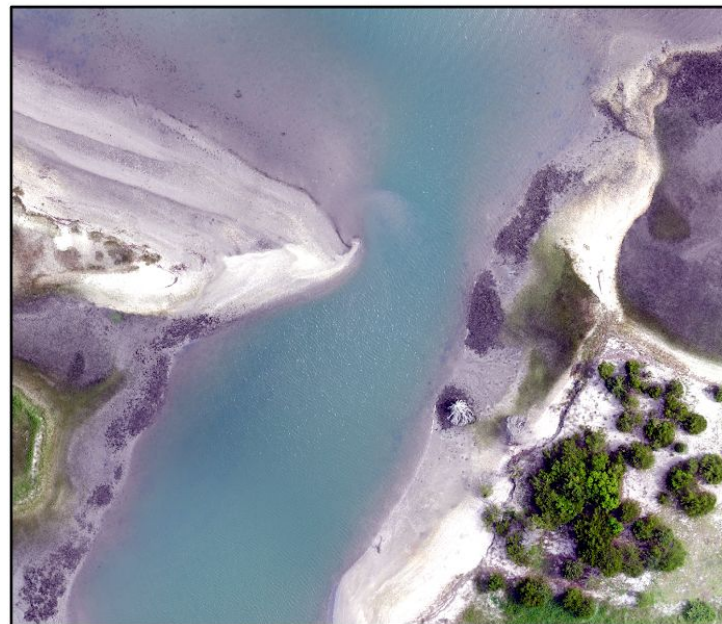
Tree or oyster
or sand or
water or marsh

Deep Learning



Feature extraction + Classification

Tree or oyster
or sand or
water or marsh



So what is Deep Learning?

Feature: ???

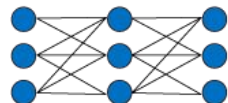
Input



Machine Learning



Feature extraction

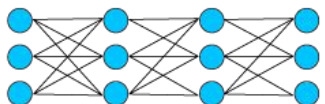


Classification

Output

Tree or oyster
or sand or
water or marsh

Deep Learning



Feature extraction + Classification

Tree or oyster
or sand or
water or marsh



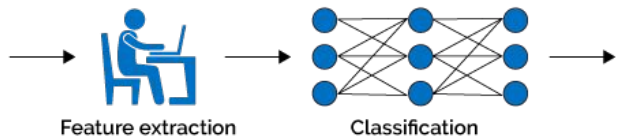
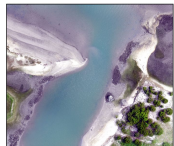
So what is Deep Learning?

Feature: ???

Input

Machine Learning

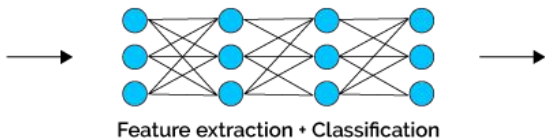
Output



Tree or oyster
or sand or
water or marsh

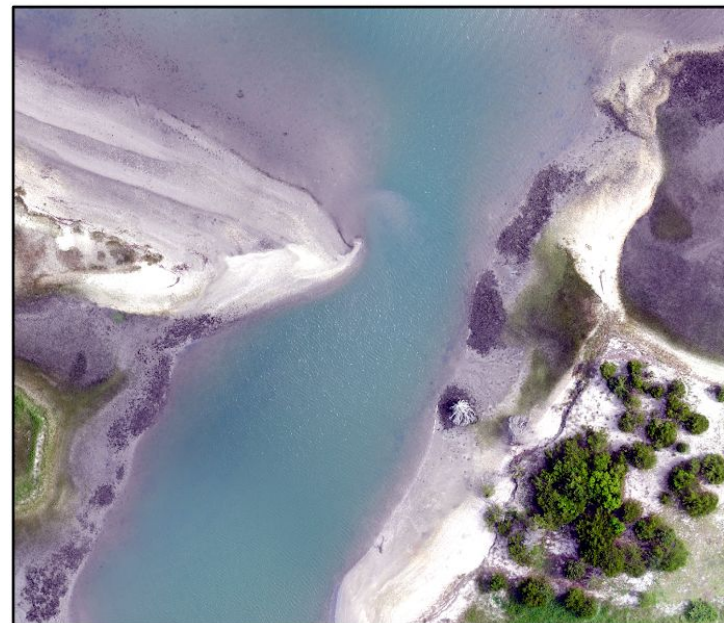


Deep Learning



Tree or oyster
or sand or
water or marsh

Instead: give it the multispectral imagery and let it discover important features itself



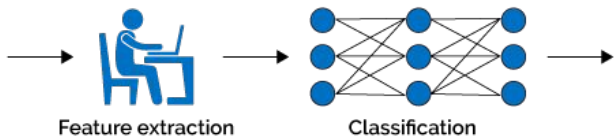
So what is Deep Learning?

Feature: ???

Input

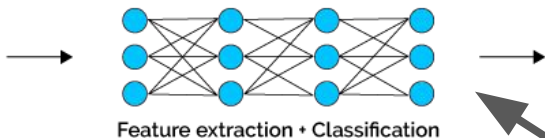
Machine Learning

Output



Tree or oyster
or sand or
water or marsh

Deep Learning



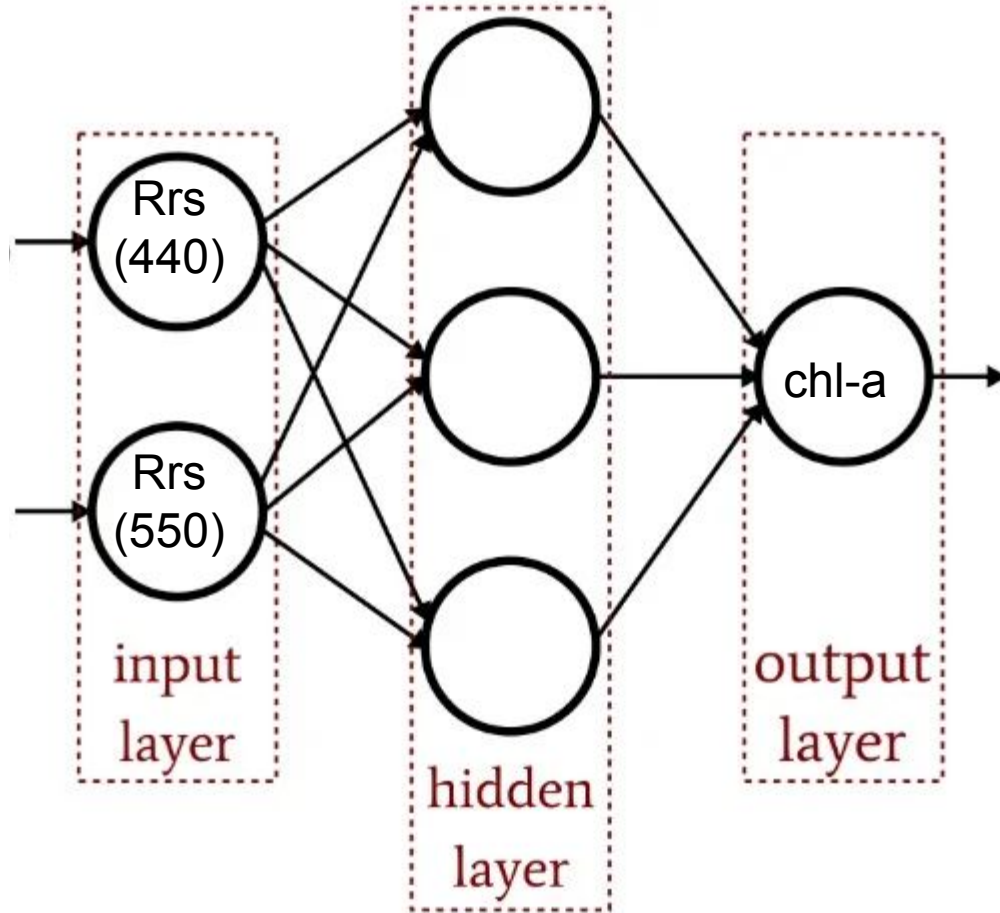
Tree or oyster
or sand or
wat



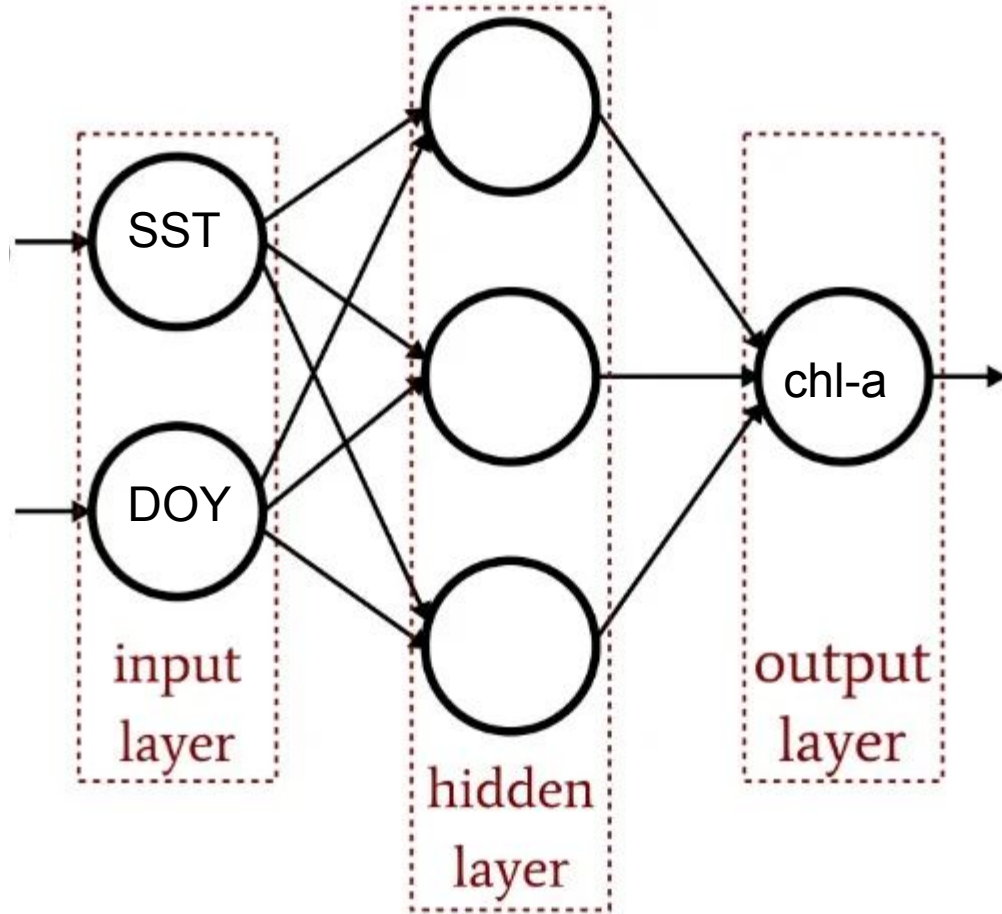
Instead: give it the multispectral imagery
let it discover important features itself

Neural networks

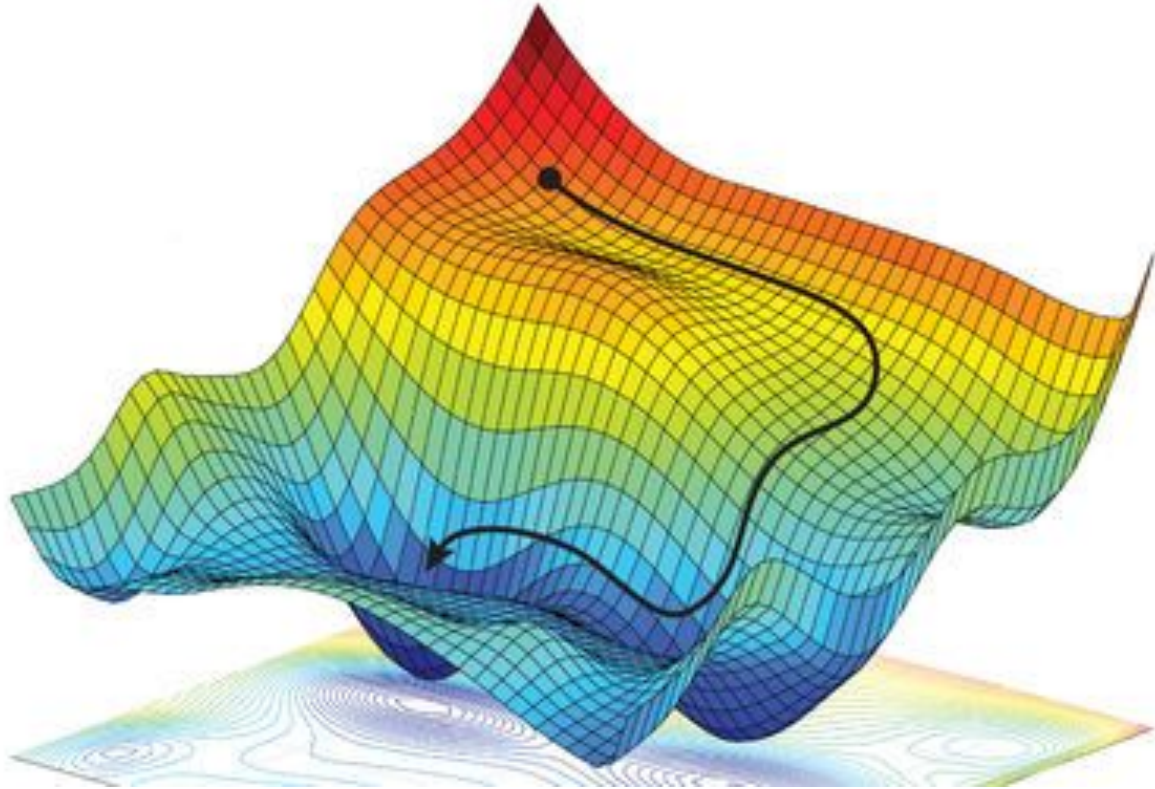
The Simplest Neural Network



The Simplest Neural Network

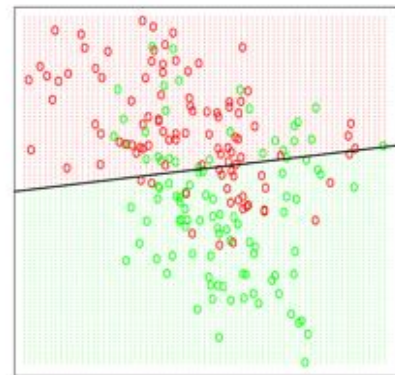
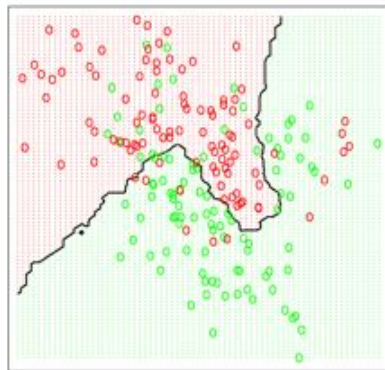


So what is Deep Learning?

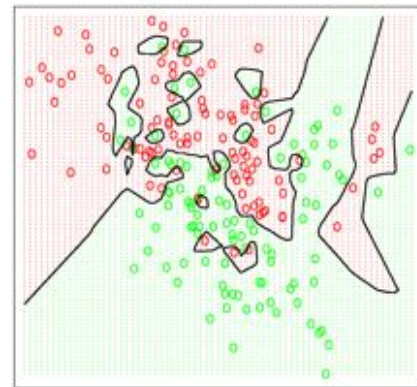


So what is Deep Learning?

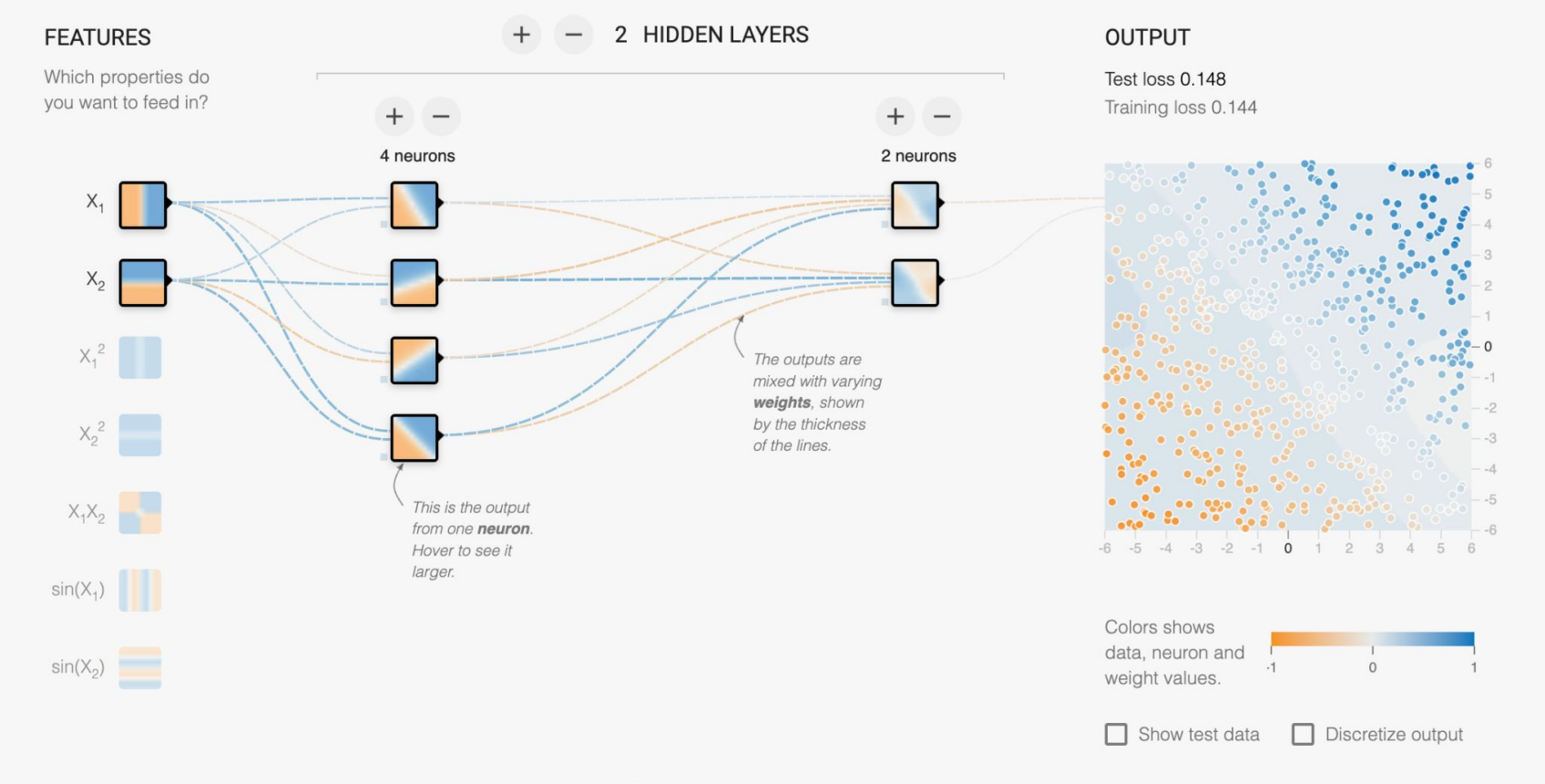
Learning means finding a combination of model parameters that minimizes a loss function for a given set of training data samples and their corresponding targets.



Overfitting?



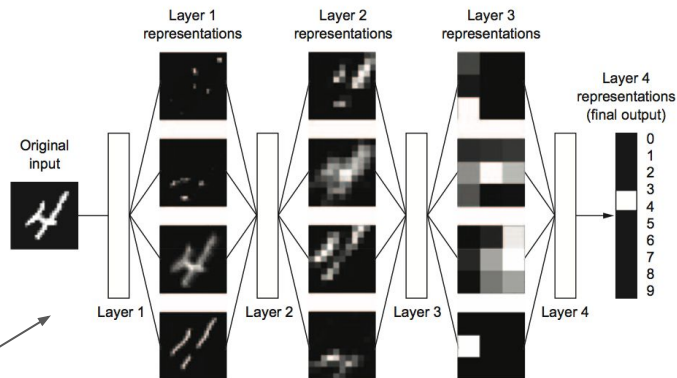
Play with it: <http://playground.tensorflow.org/>



So what is Deep Learning?

Convolutional neural networks...

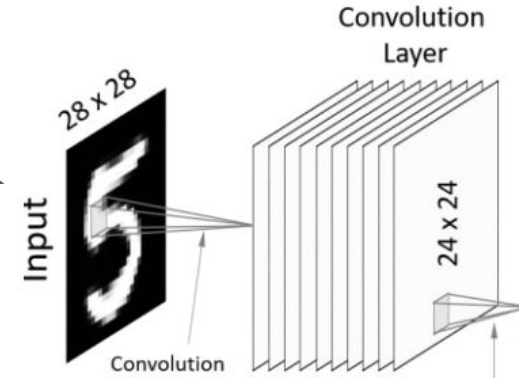
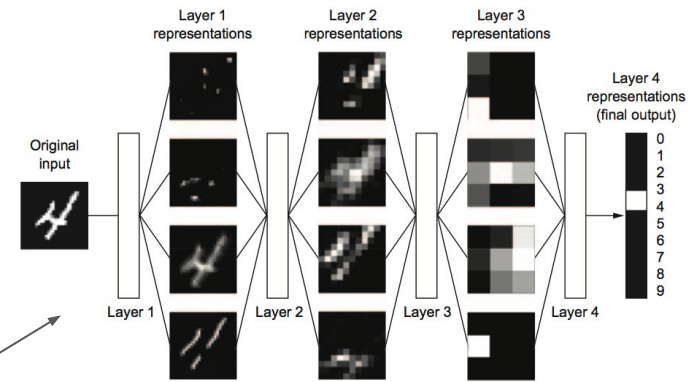
- learn to recognize high-level structure in images by building hierarchical representations
- extract features via spatial convolutions with filters
- learn filters via iterative minimization of a risk function
- **have shown capabilities beyond human performance for image analysis**



So what is Deep Learning?

Convolutional neural networks...

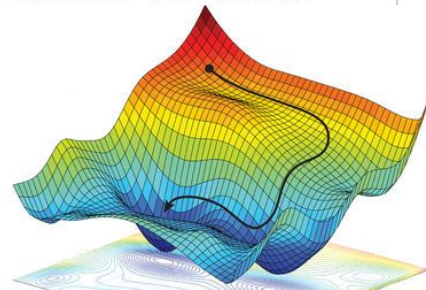
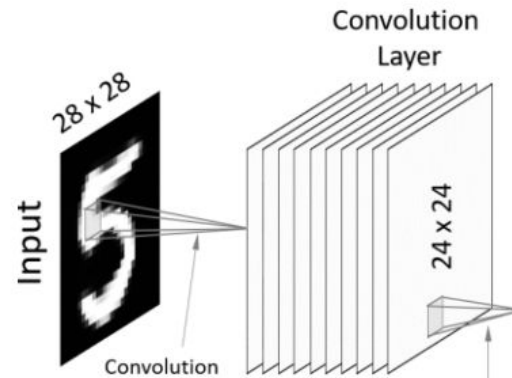
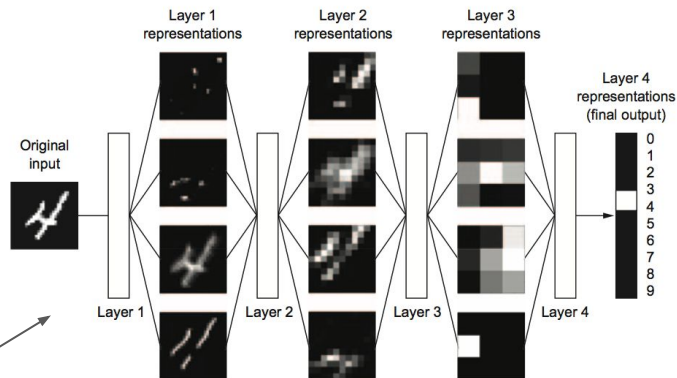
- learn to recognize high-level structure in images by building hierarchical representations
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- learn filters via iterative minimization of a risk function
- **have shown capabilities beyond human performance for image analysis**



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Recurrent neural networks...

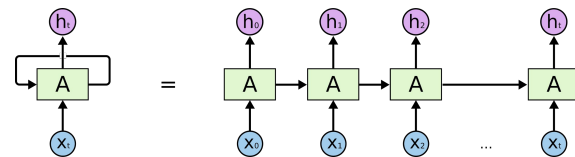
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Recurrent neural networks...

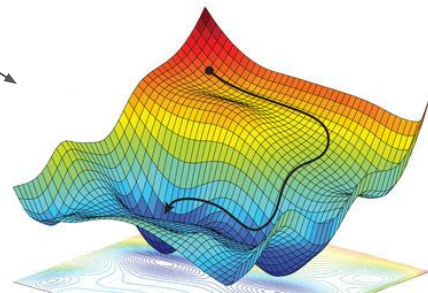
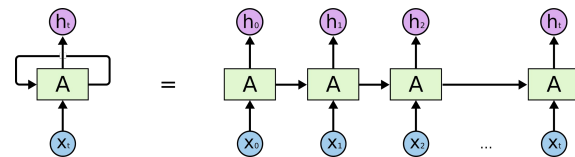
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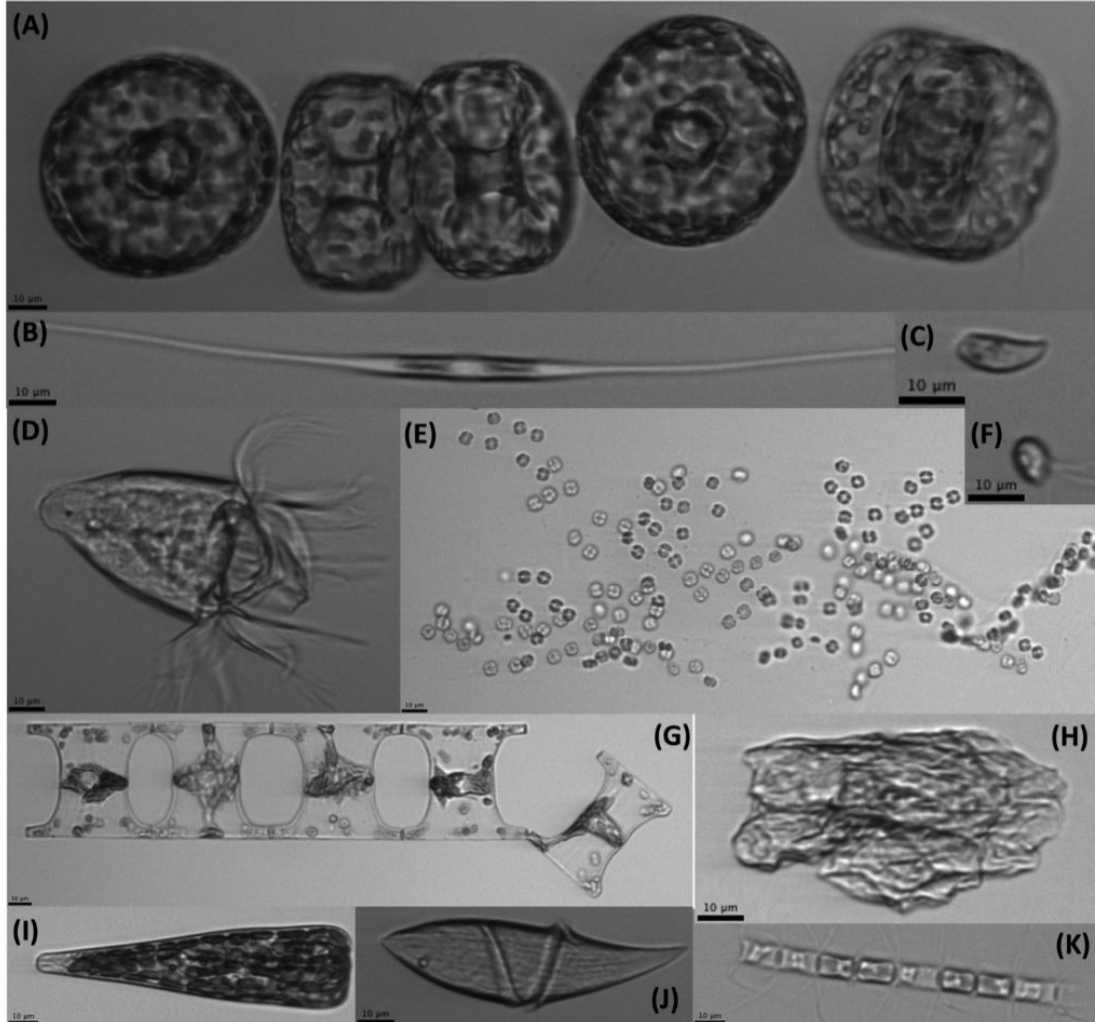
Recap: Why Machine Learning in Remote Sensing?

It is useful for its ability to represent complex, nonlinear relationships across spectral, temporal, and spatial dimensions.

Recap: Why Machine Learning in Remote Sensing?

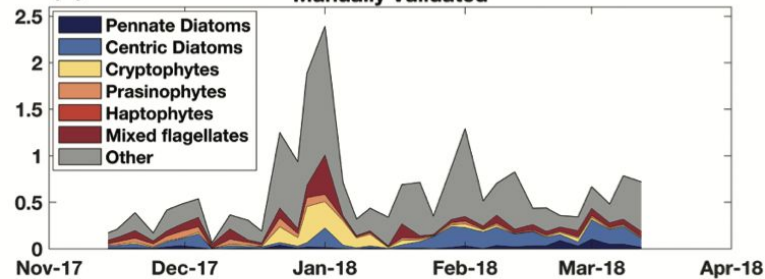
It is useful for its ability to represent complex, nonlinear relationships across spectral, temporal, and spatial dimensions.

- Spectral relationships between bands (and relations between those relations)
- Spatial relationships such as seascape context
- Temporal relationships such as phenology or bloom cycles
- Combining sensing modalities (e.g. Optical + SAR)

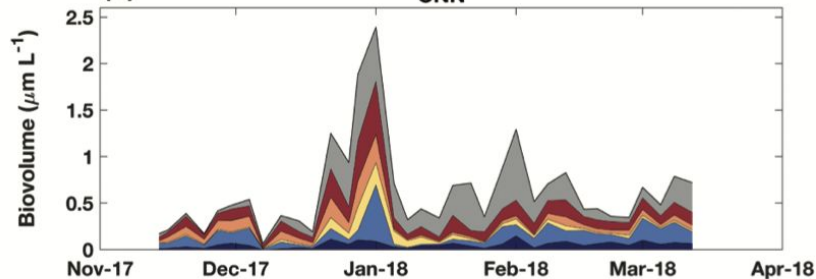


2017-2018

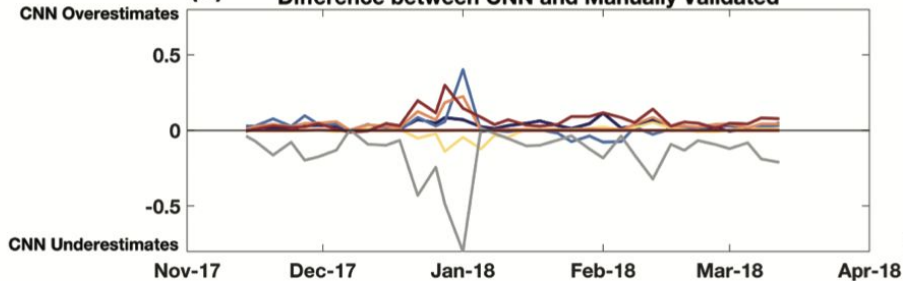
(A) Manually Validated



(B) CNN

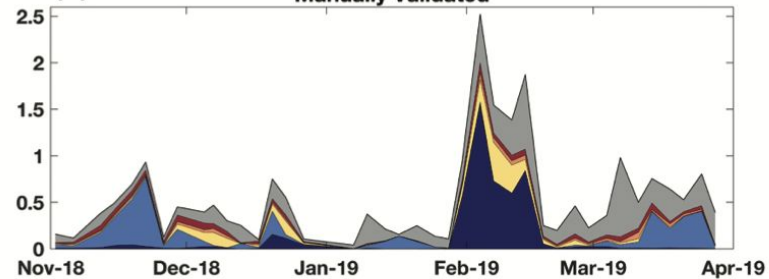


(C) Difference between CNN and Manually Validated

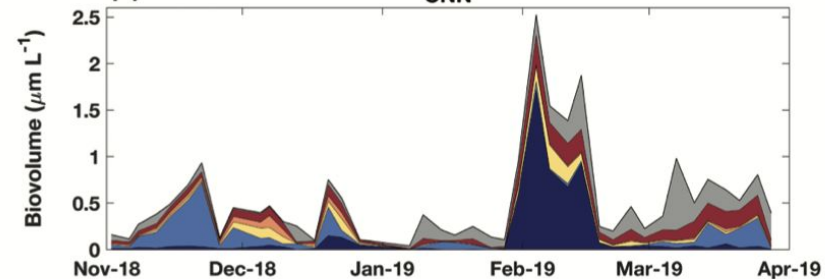


2018-2019

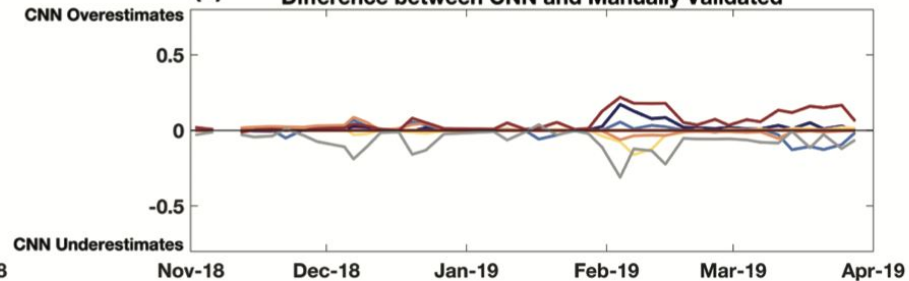
(D) Manually Validated

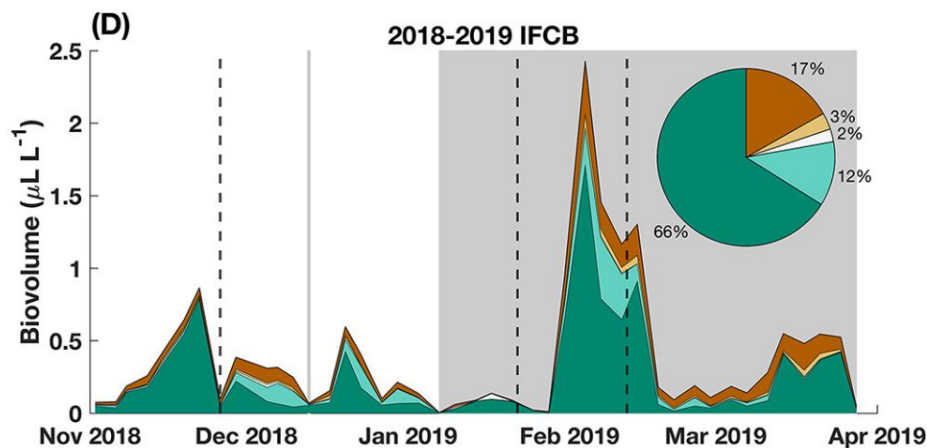
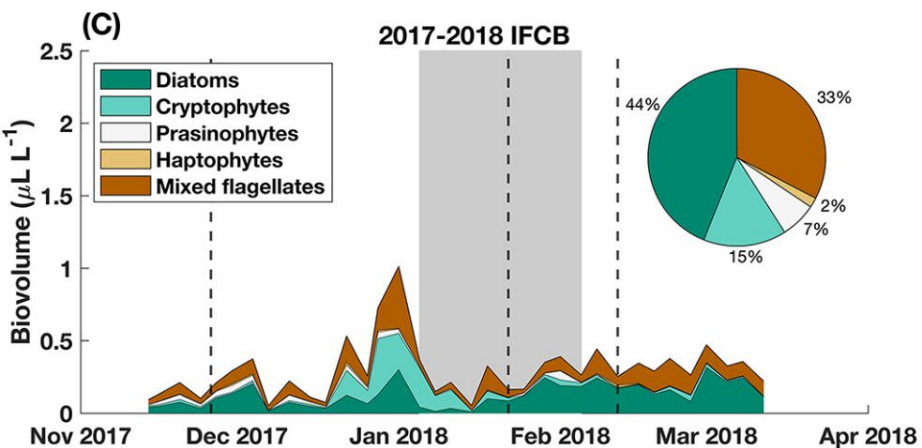
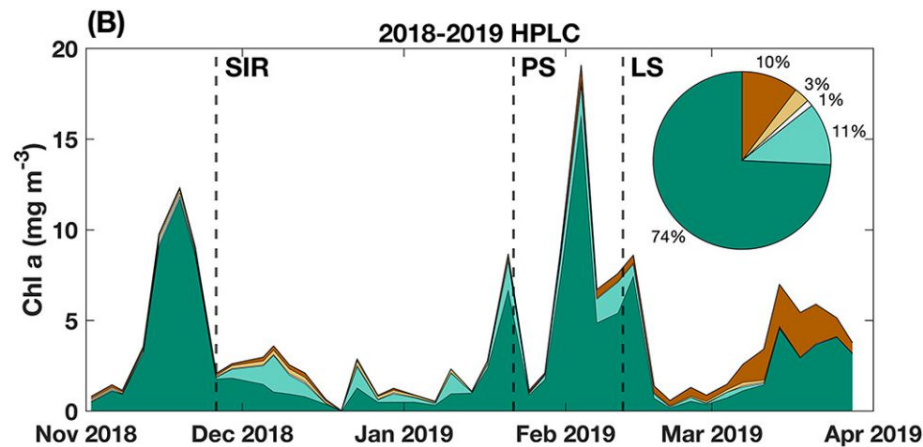
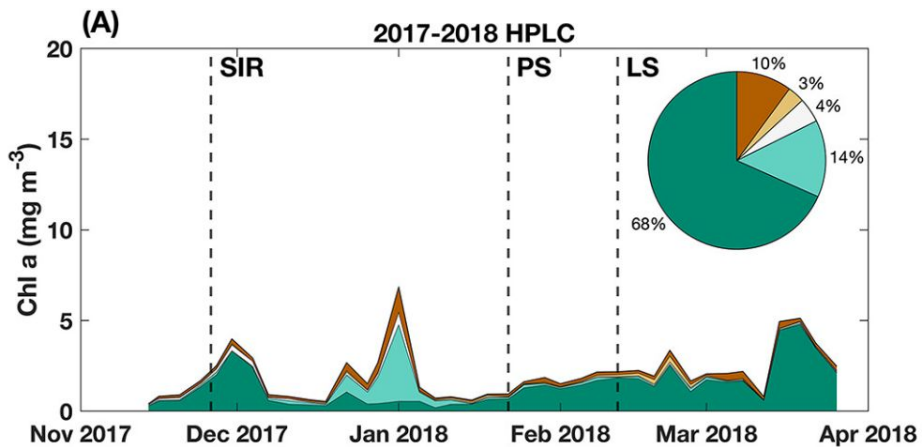


(E) CNN

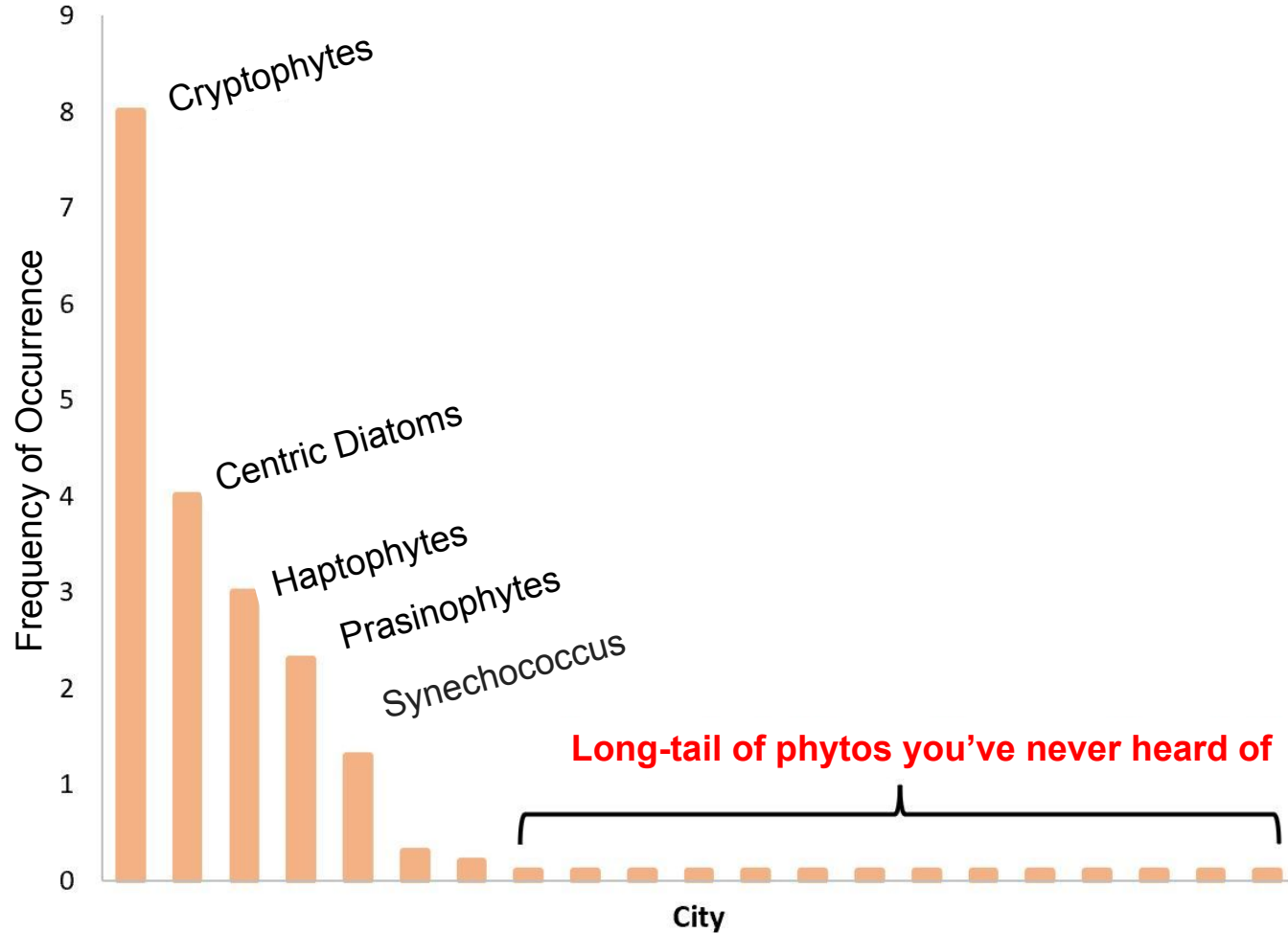


(F) Difference between CNN and Manually Validated

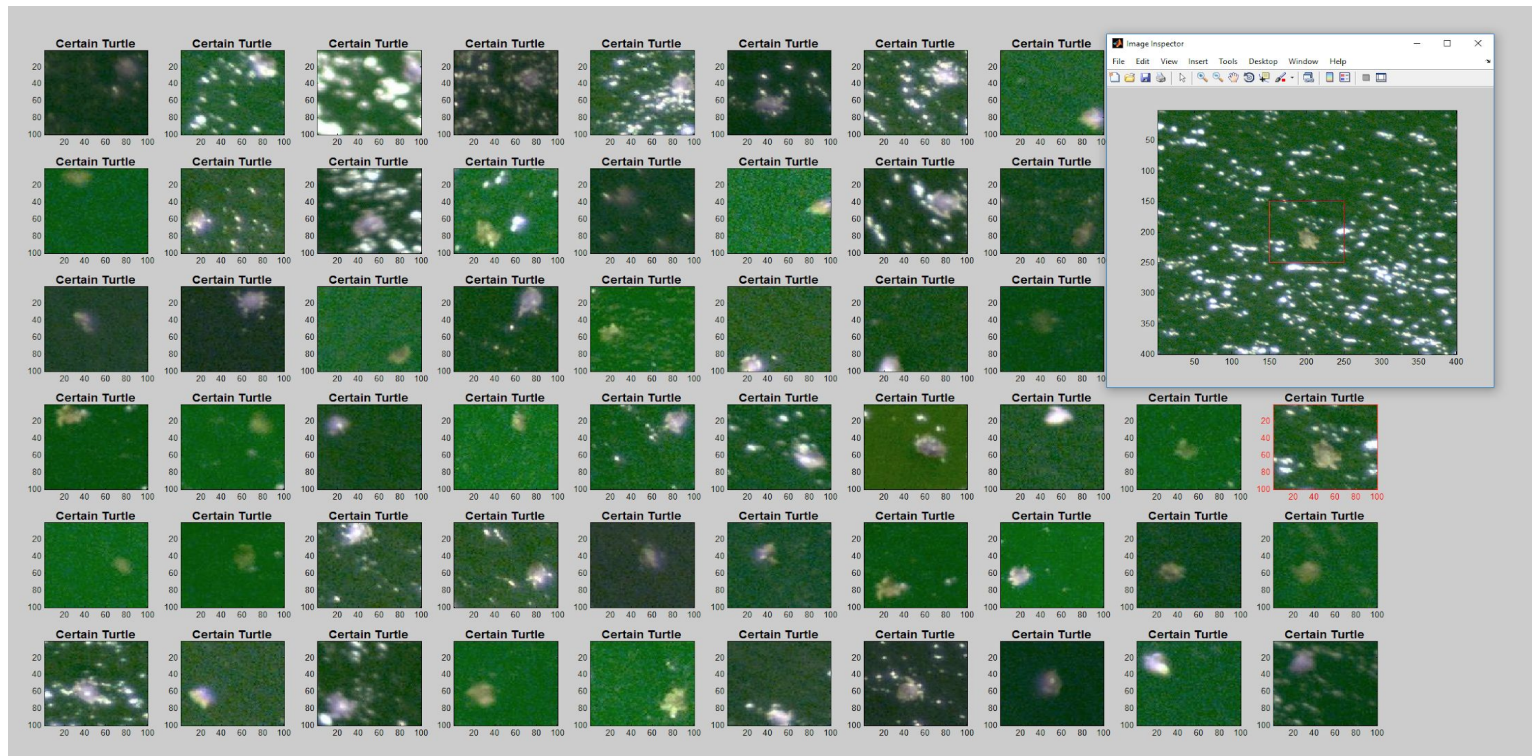




Long Tailed Distribution of Phytoplankton Species



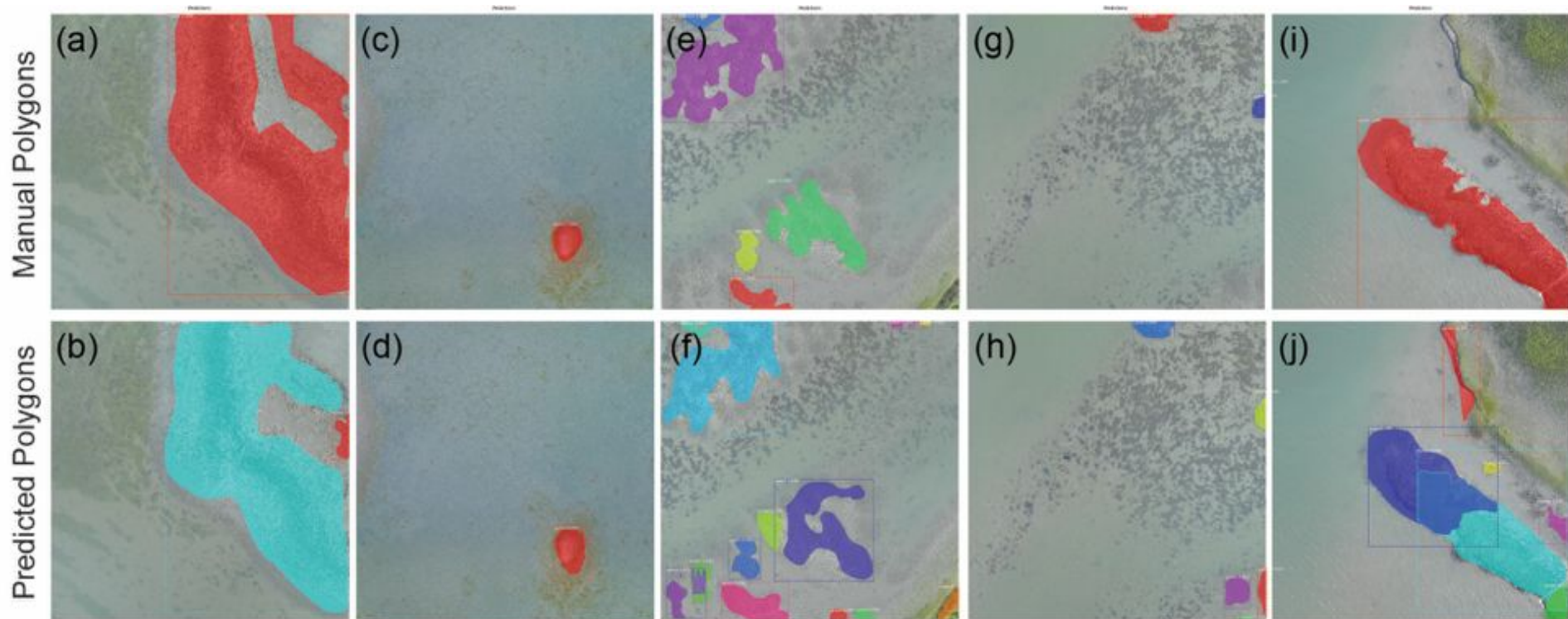
Object Detection: Patch Based



Object Detection: Bounding Box



Object Detection: Semantic Segmentation



Recap: Why Machine Learning in Remote Sensing?

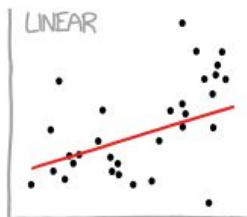
It is useful for its ability to represent **complex**, nonlinear relationships across spectral, temporal, and spatial dimensions.

But simplicity is beautiful if it is feasible.

Specifically for Remote Sensing Inversions

- you can't extrapolate beyond train/test data
- the simpler the better and comparing to a simple baseline is critical
- assessment needs to be spatially and temporally independent
- if we improve 1-2% over a polynomial regression of a band ratio (to predict chl_a) with a complex NN is this beneficial?
- sometimes what we care about in the ocean in the context of climate change is anomalies and things different from expectations
- think about encodings - you might be interested in the encoding of a light attenuation profile into a single K_d - or you might be interested in the encoding of a spatial pattern into some more semantically meaningful vector.

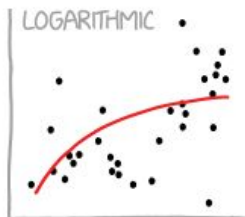
CURVE-FITTING METHODS AND THE MESSAGES THEY SEND



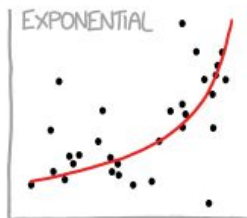
"HEY, I DID A REGRESSION."



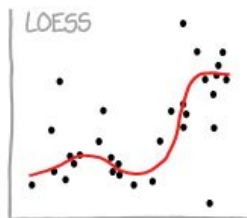
"I WANTED A CURVED LINE, SO I MADE ONE WITH MATH."



"LOOK, IT'S TAPERING OFF!"



"LOOK, IT'S GROWING UNCONTROLLABLY!"



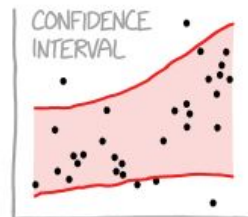
"I'M SOPHISTICATED, NOT LIKE THOSE BUMBLING POLYNOMIAL PEOPLE."



"I'M MAKING A SCATTER PLOT BUT I DON'T WANT TO."



"I NEED TO CONNECT THESE TWO LINES, BUT MY FIRST IDEA DIDN'T HAVE ENOUGH MATH."



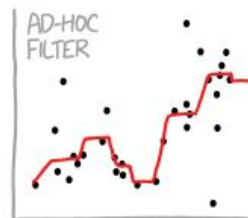
"LISTEN, SCIENCE IS HARD. BUT I'M A SERIOUS PERSON DOING MY BEST."



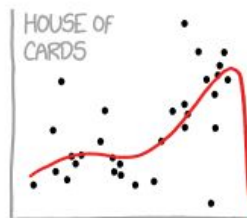
"I HAVE A THEORY, AND THIS IS THE ONLY DATA I COULD FIND."



"I CLICKED 'SMOOTH LINES' IN EXCEL."



"I HAD AN IDEA FOR HOW TO CLEAN UP THE DATA. WHAT DO YOU THINK?"

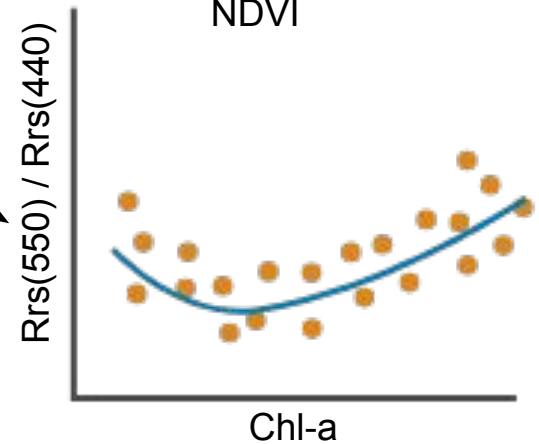
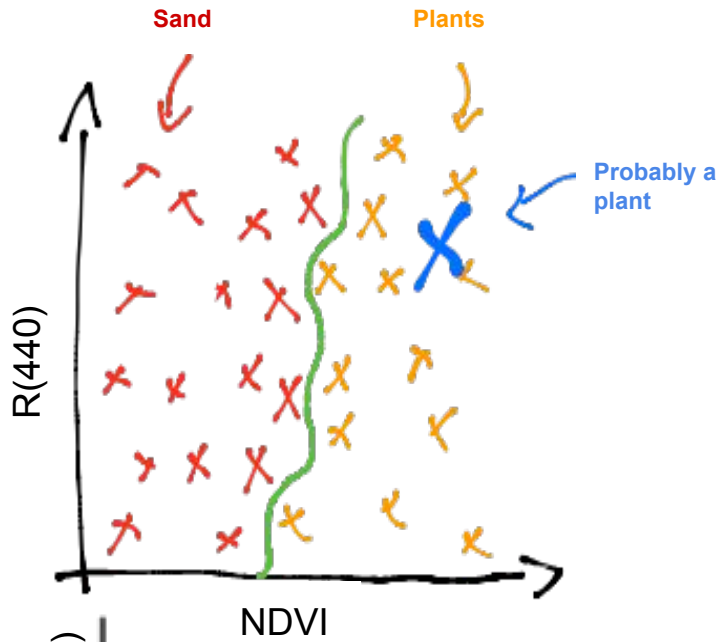
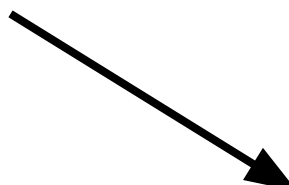
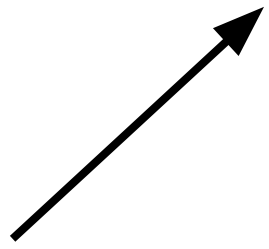
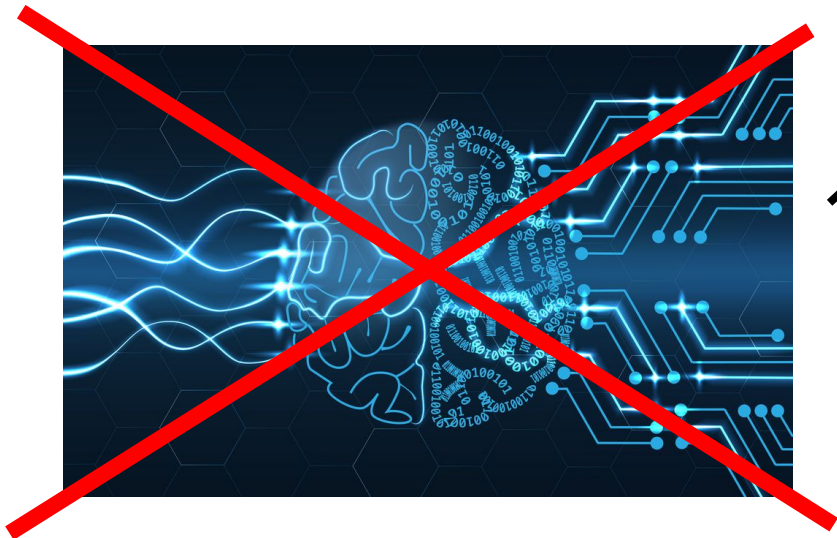


"AS YOU CAN SEE, THIS MODEL SMOOTHLY FITS THE- WAIT NO NO DON'T EXTEND IT AAAAAA!!!"

Python and Machine Learning Resources

- Building a strong but basic python foundation (e-textbook):
 - <https://www.pythonlikeyoumeanit.com/>
- Become python proficient (textbook):
 - <https://www.amazon.com/dp/0134034287/>
- Open Geospatial Tutorials using Python for Geospatial and Remote Sensing Analysis
 - <https://github.com/patrickcgray/open-geo-tutorial/>
- ML Crash Course (online class from Google):
 - <https://developers.google.com/machine-learning/crash-course/>
- Python and Deep Learning with Keras (textbook):
 - <https://www.amazon.com/Deep-Learning-Python-Francois-Chollet/dp/1617294438>
- Neural Networks Video Series (highly recommend)
 - <https://www.youtube.com/watch?v=aircAruvnKk>
- Papers:
 - Ma et al. Deep learning in remote sensing applications: A meta-analysis and review
 - Zhang et al. Deep learning for remote sensing data: A technical tutorial on the state of the art
 - Lecun et al. Deep Learning

Questions?



Quick Exploration

https://github.com/patrickcgray/oceanoptics2021/blob/main/oci_ml_oo23.ipynb

Deep Learning Technical Overview



Surpassing Human Performance

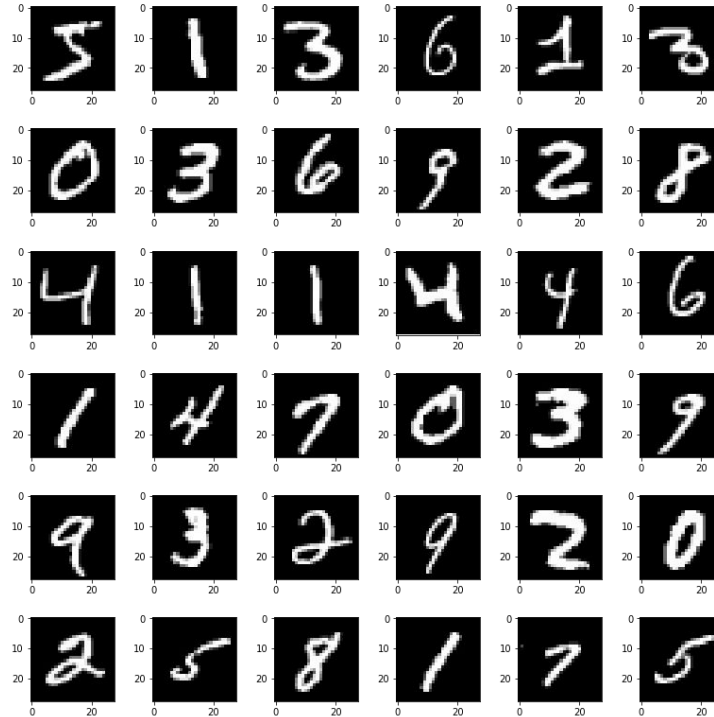
DL is now beating humans at many tasks:

- Chess
- Driving
- Language Translation
- Voice Recognition
- Predicting stock prices
- Object Detection
- Go

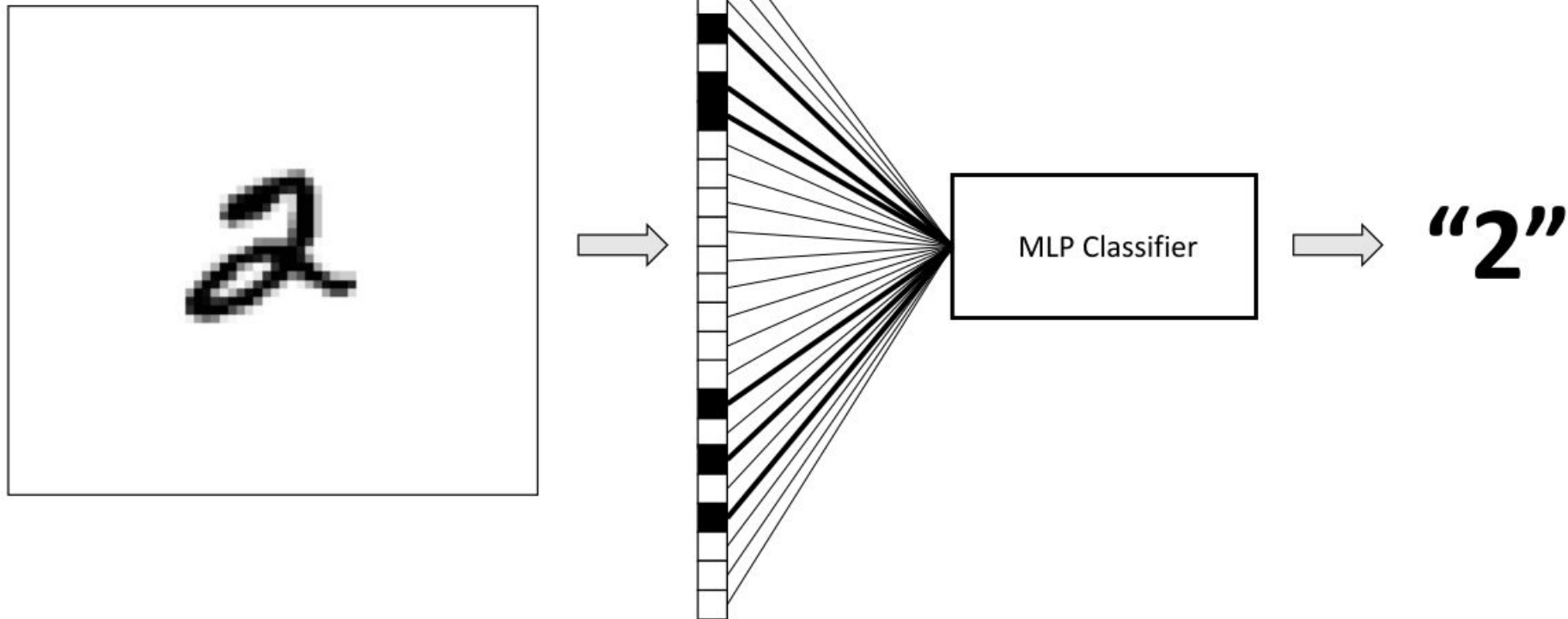


So what is Deep Learning?

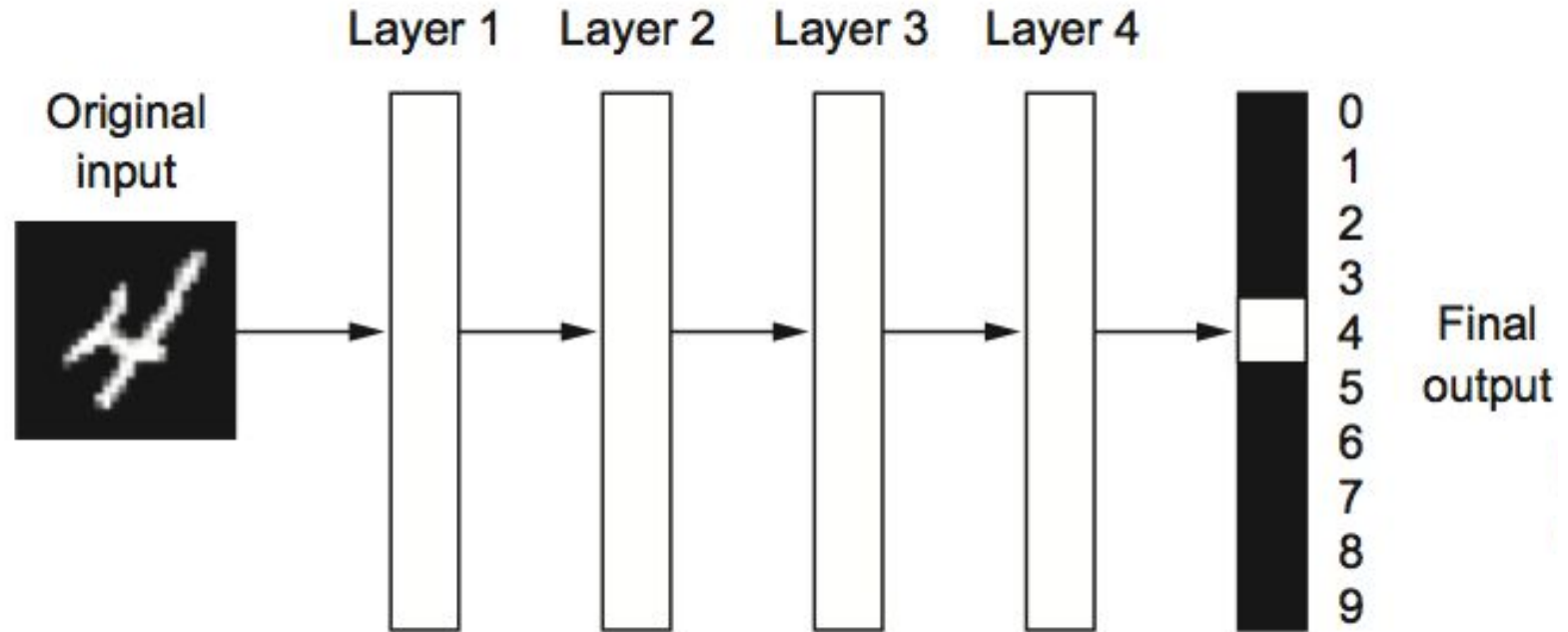
First 36 images in MNIST



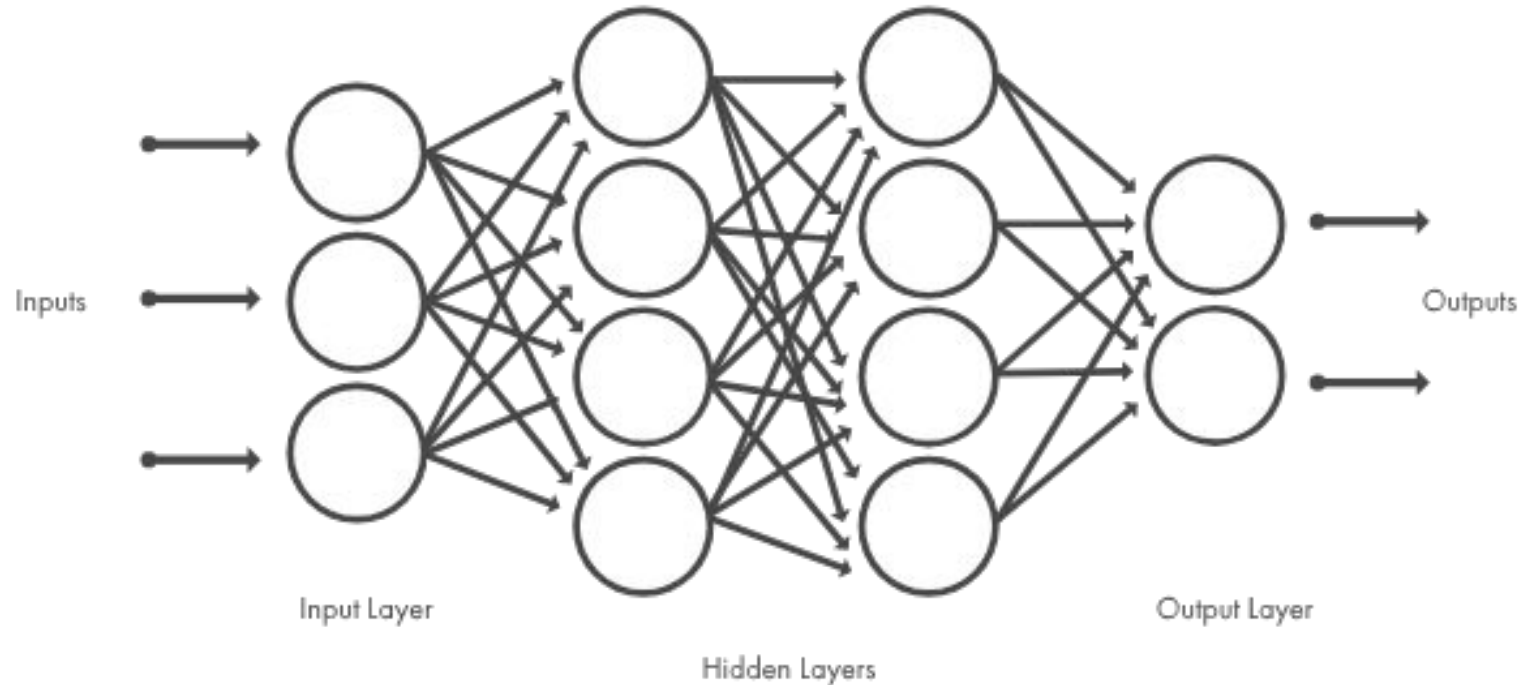
So what is Deep Learning?



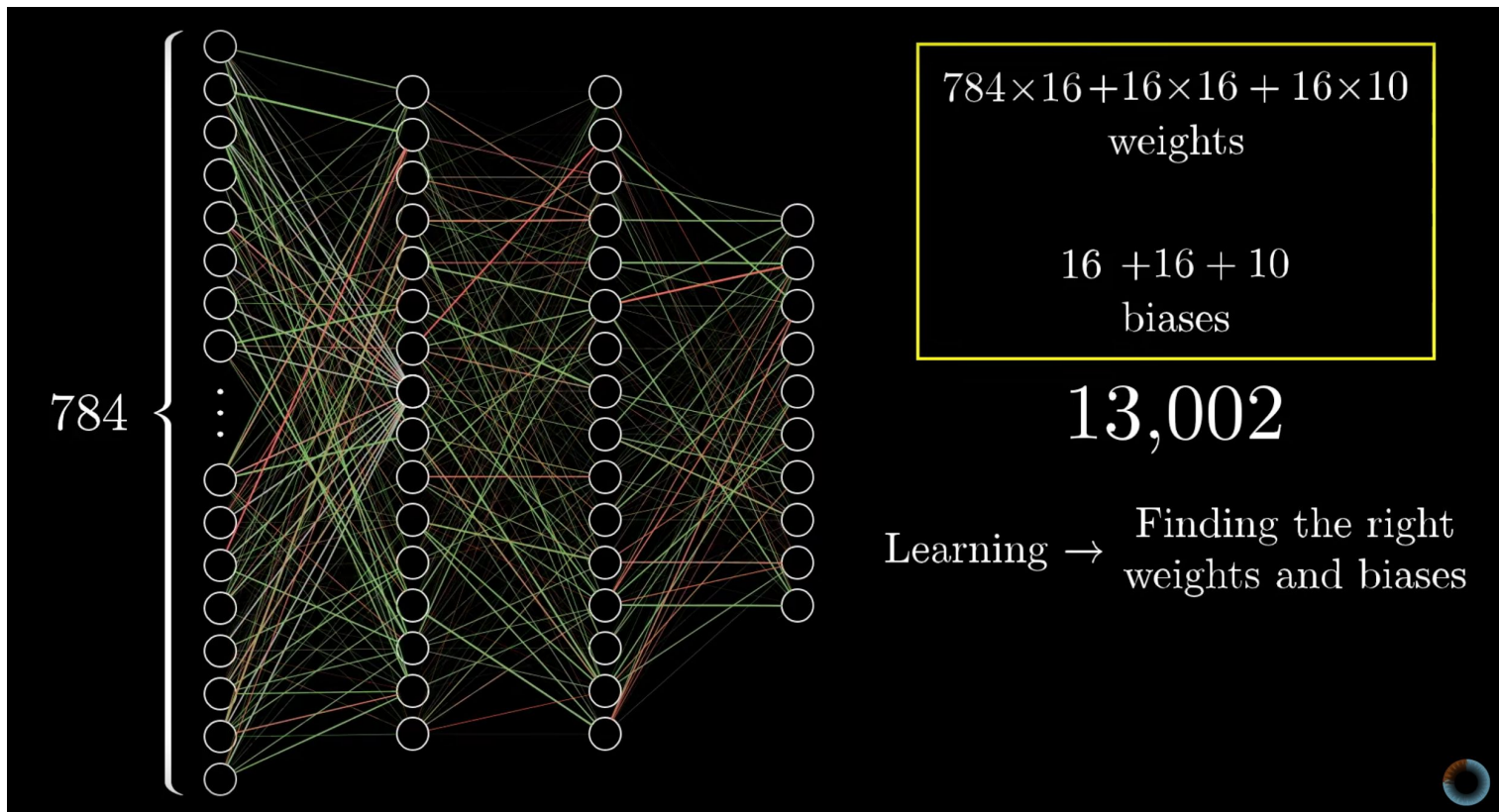
So what is Deep Learning?



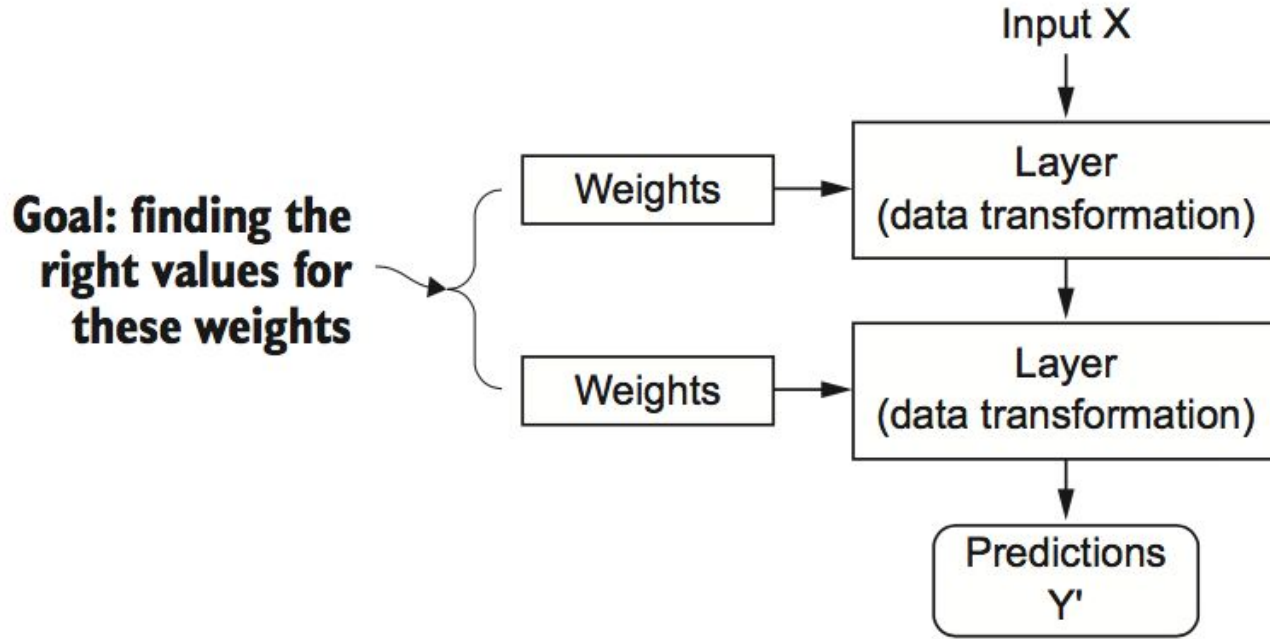
So what is Deep Learning?



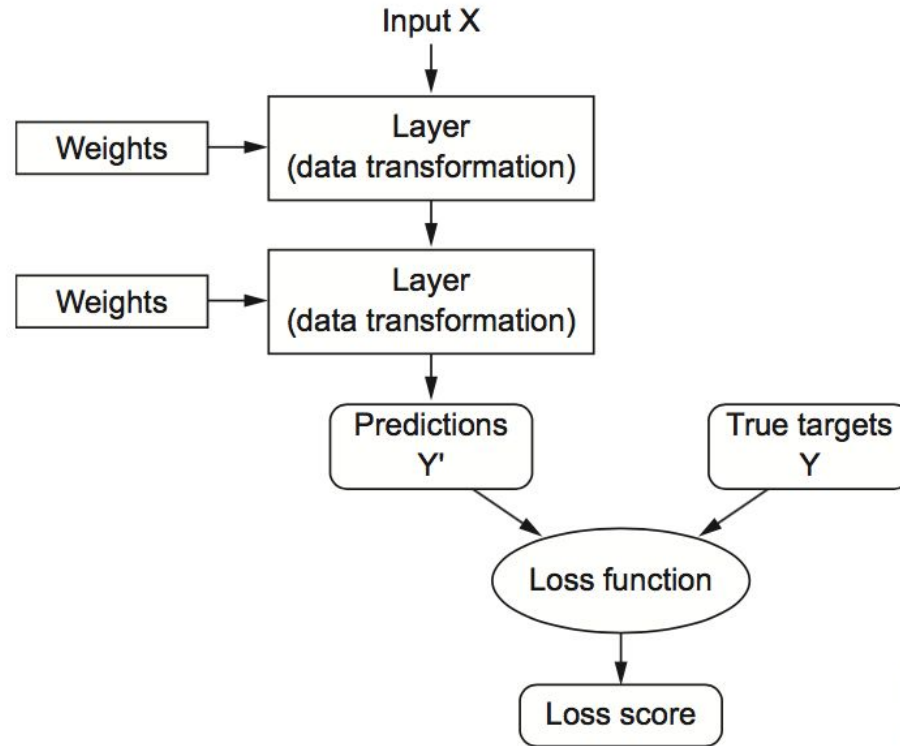
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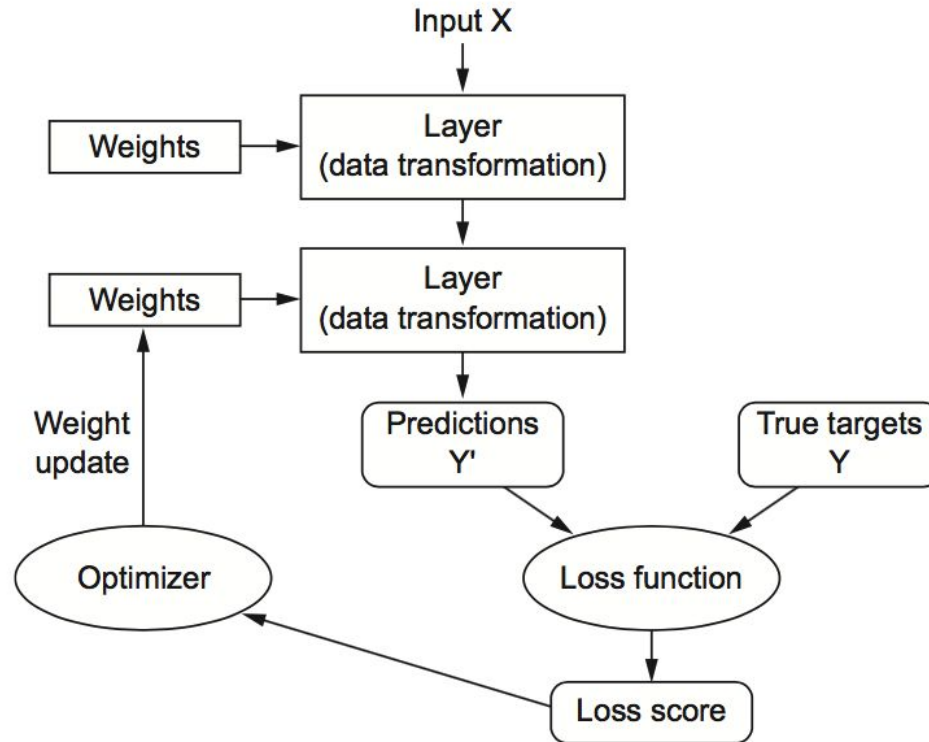
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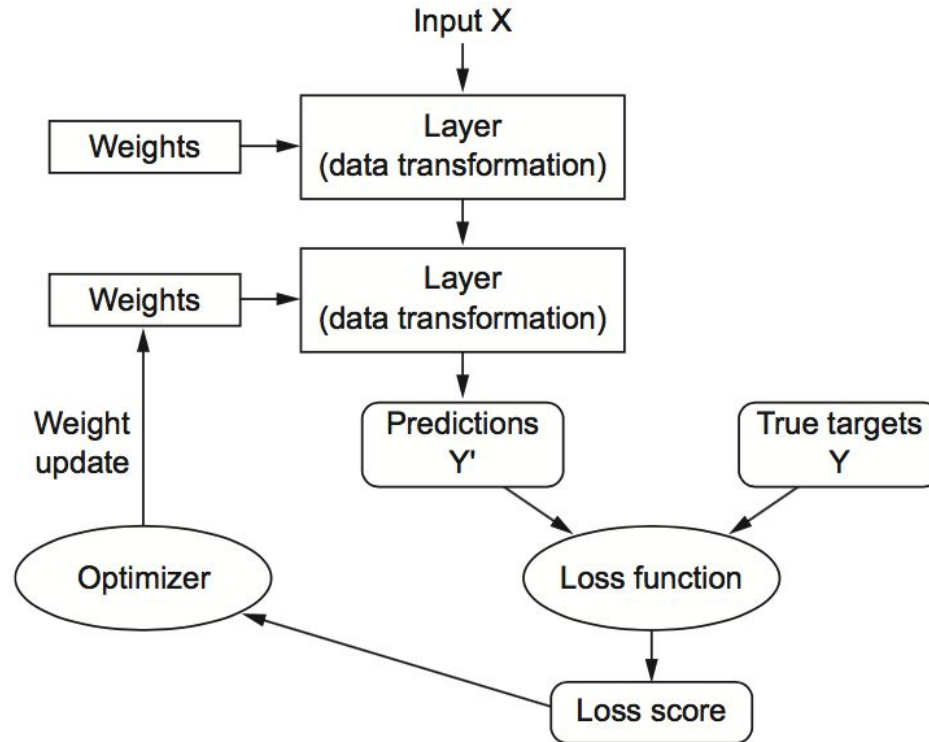
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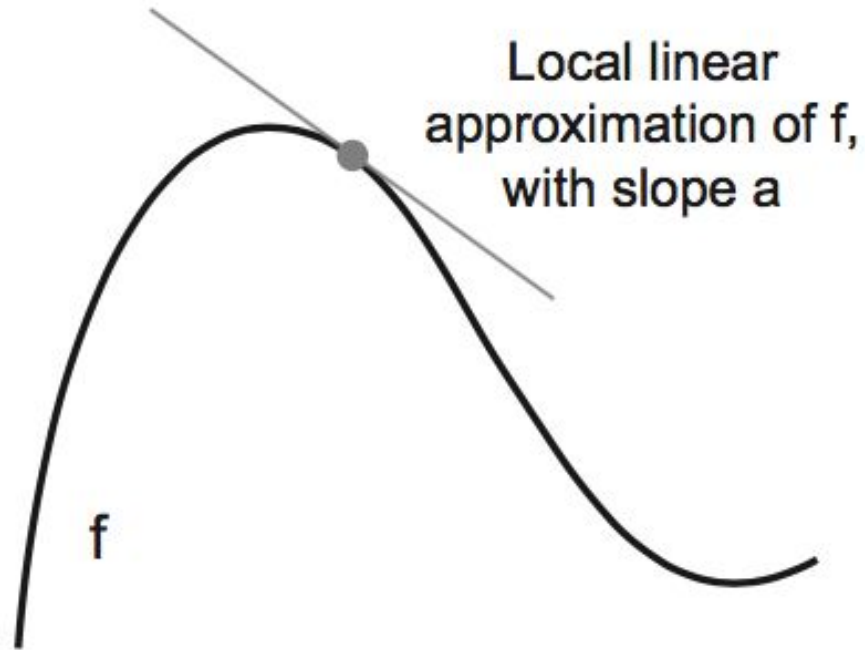
So what is Deep Learning?



So what is Deep Learning? - Questions???



So what is Deep Learning?



So what is Deep Learning?

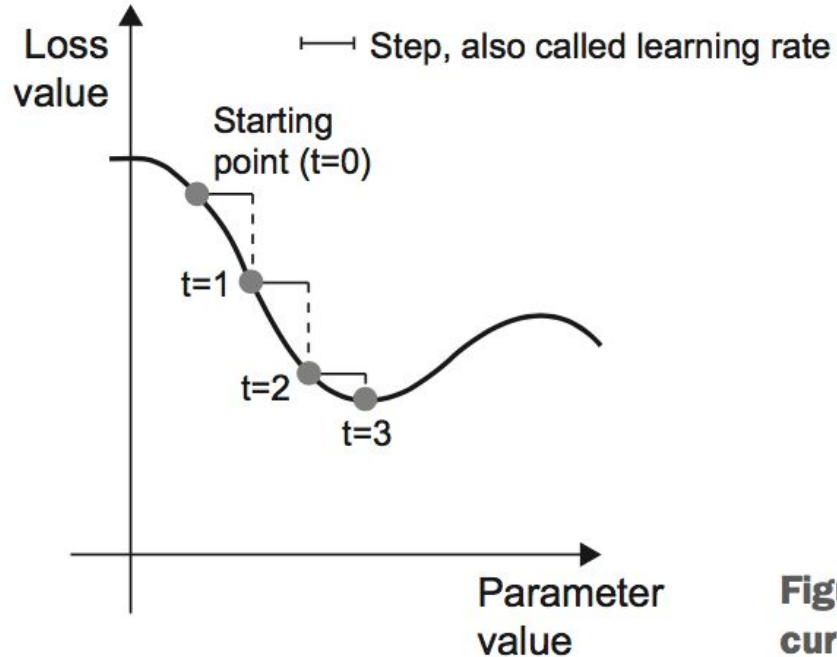


Figure 2.11 SGD down a 1D loss curve (one learnable parameter)

So what is Deep Learning?

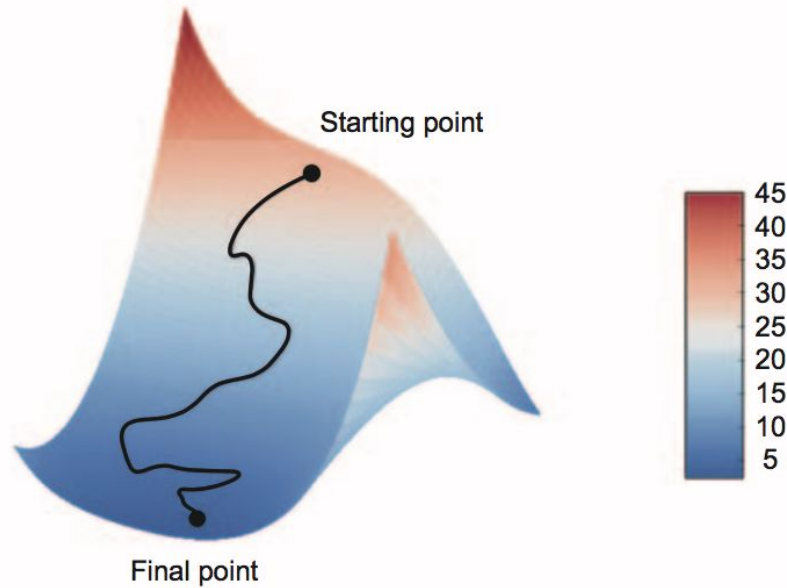


Figure 2.12 Gradient descent down a 2D loss surface (two learnable parameters)

So what is Deep Learning?

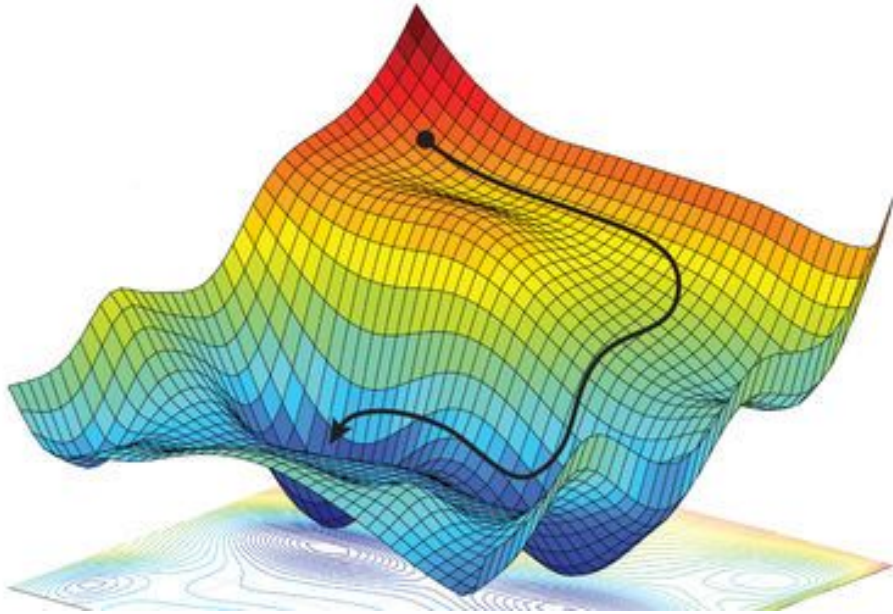
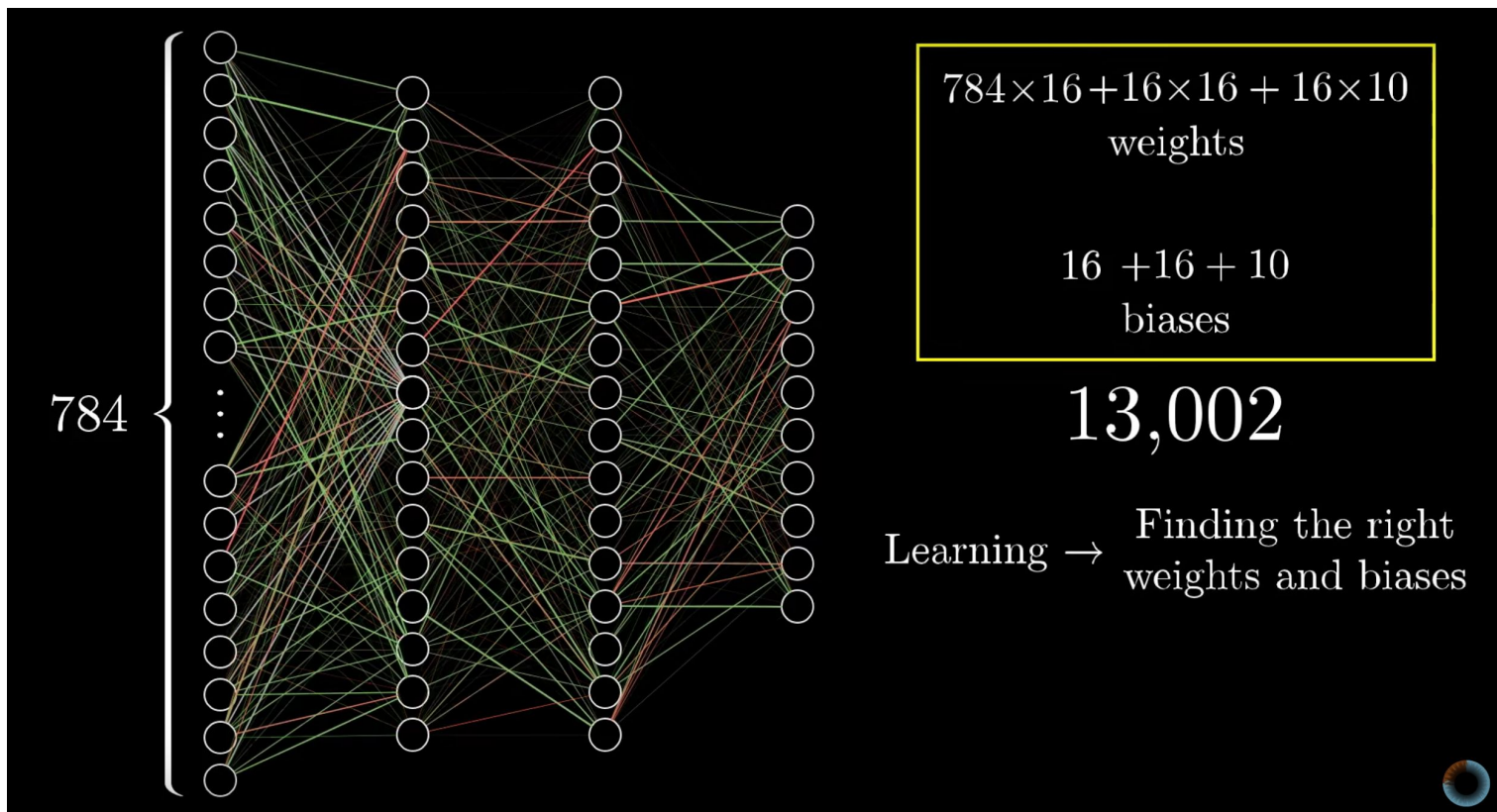


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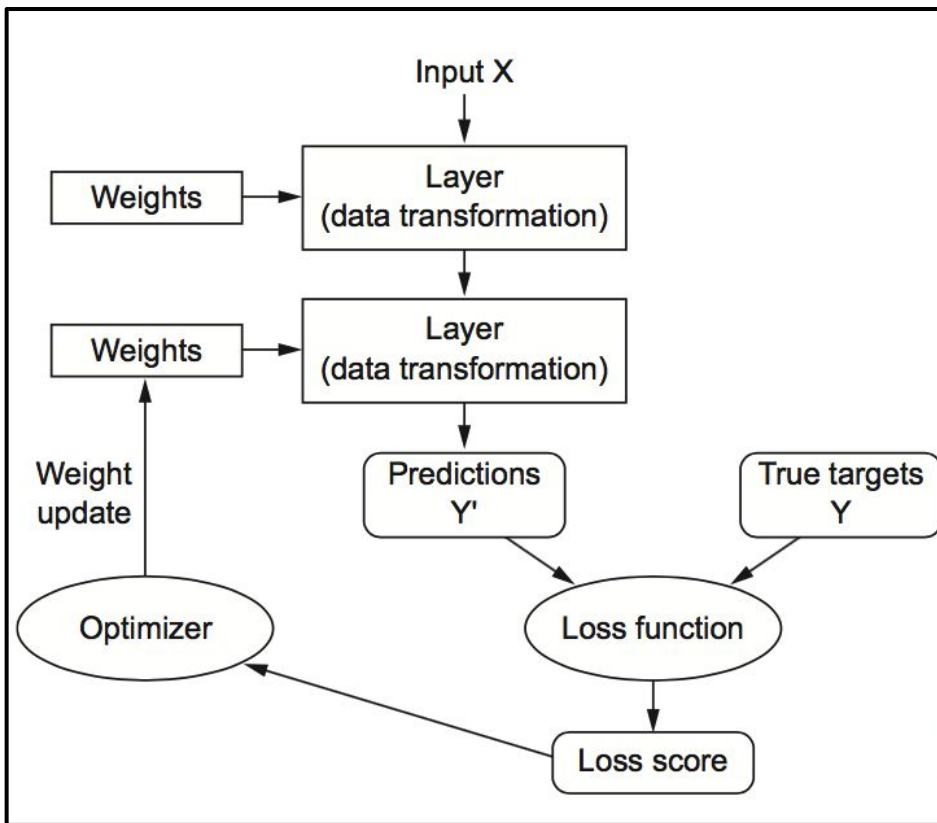


So what is Deep Learning?

Learning means finding a combination of model parameters that minimizes a loss function for a given set of training data samples and their corresponding targets.

Loss is the quantity you'll attempt to minimize during training → so it should represent a measure of success for the task you're trying to solve.

The optimizer specifies the exact way in which the gradient of the loss will be used to update parameters in your model.

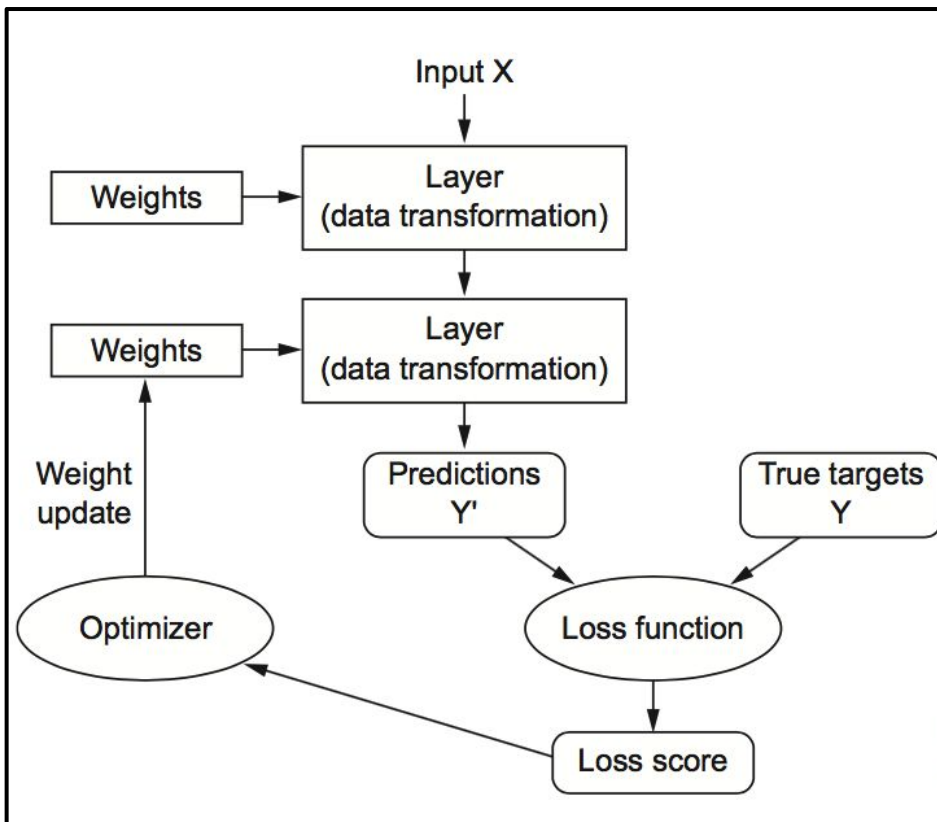


So what is Deep Learning?

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Navigate to:

<http://playground.tensorflow.org/>

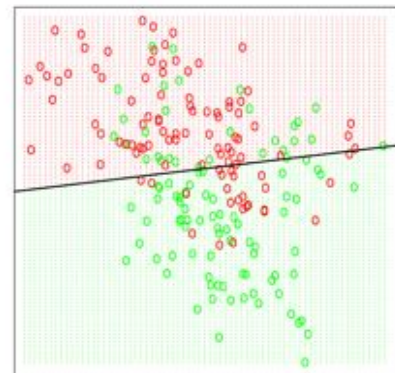
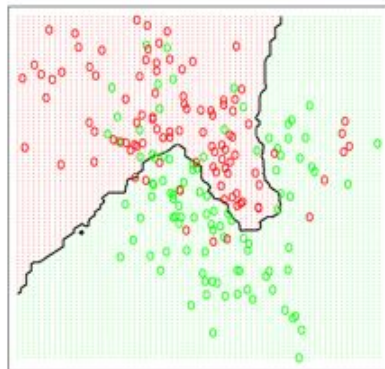


So what is Deep Learning?

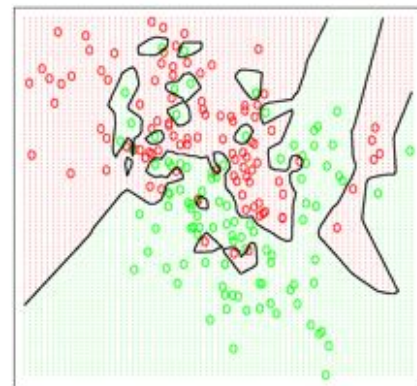
Learning means finding a combination of model parameters that minimizes a loss function for a given set of training data samples and their corresponding targets.

Navigate to:

<http://playground.tensorflow.org/>



Overfitting?



Let's build a super simple Neural Network!

Navigate to:

https://github.com/patrickcgray/deep_learning_ecology



So what is Deep Learning?

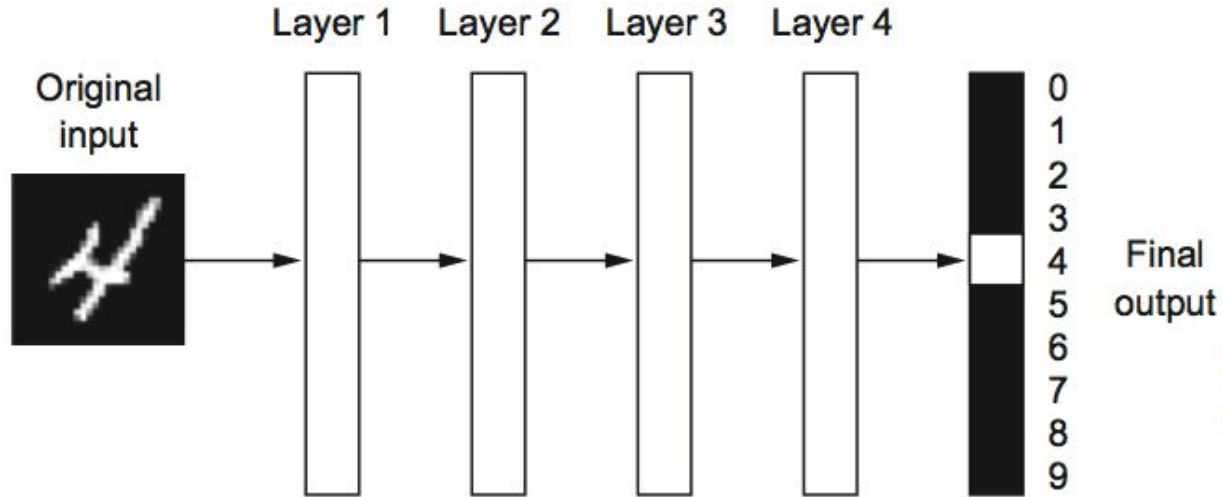
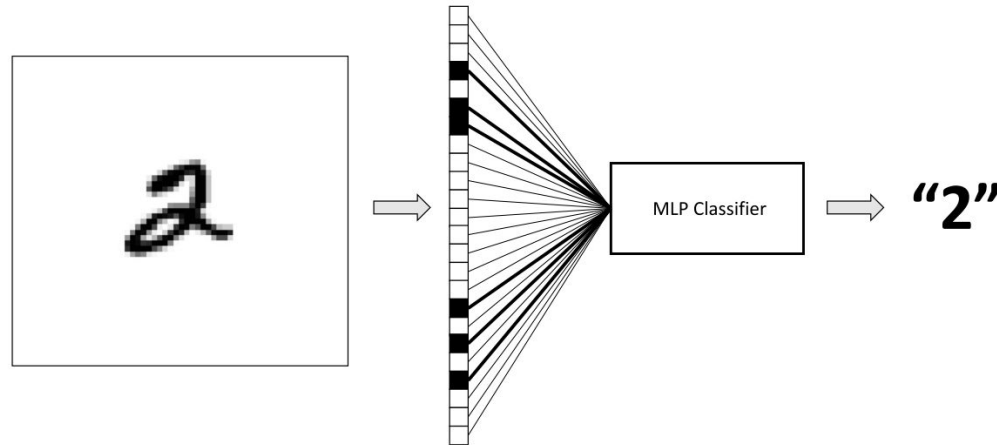


Figure 1.5 A deep neural network for digit classification

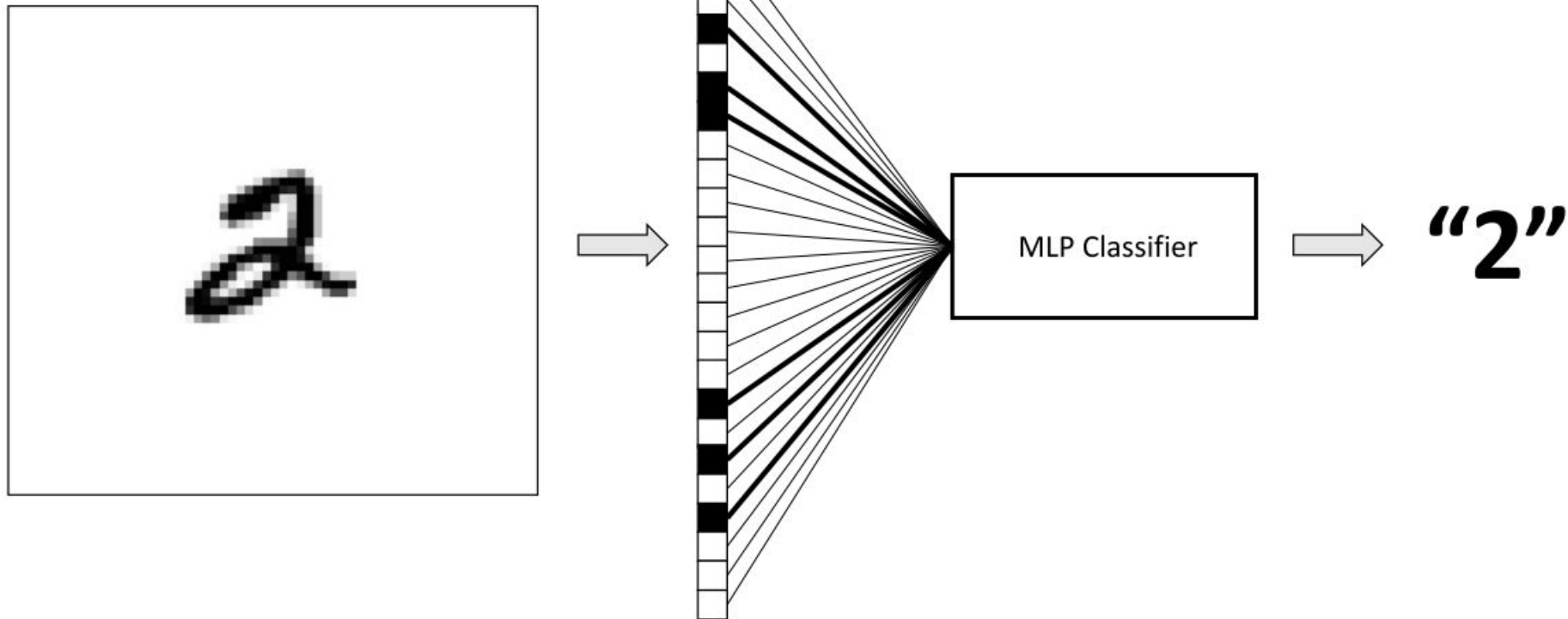
Train your own Multi-layer Perceptron!

Navigate to:

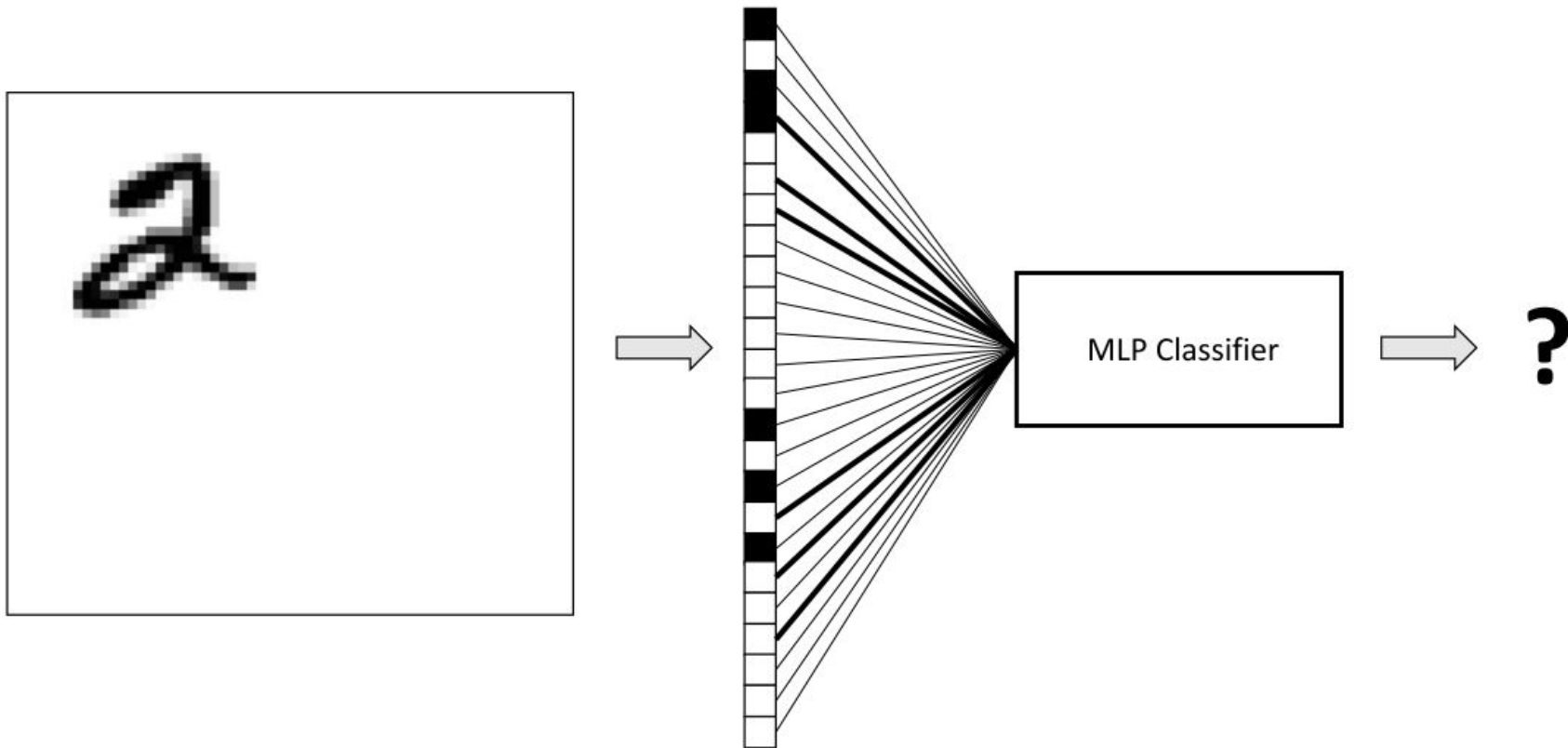
https://github.com/patrickcgray/deep_learning_ecology



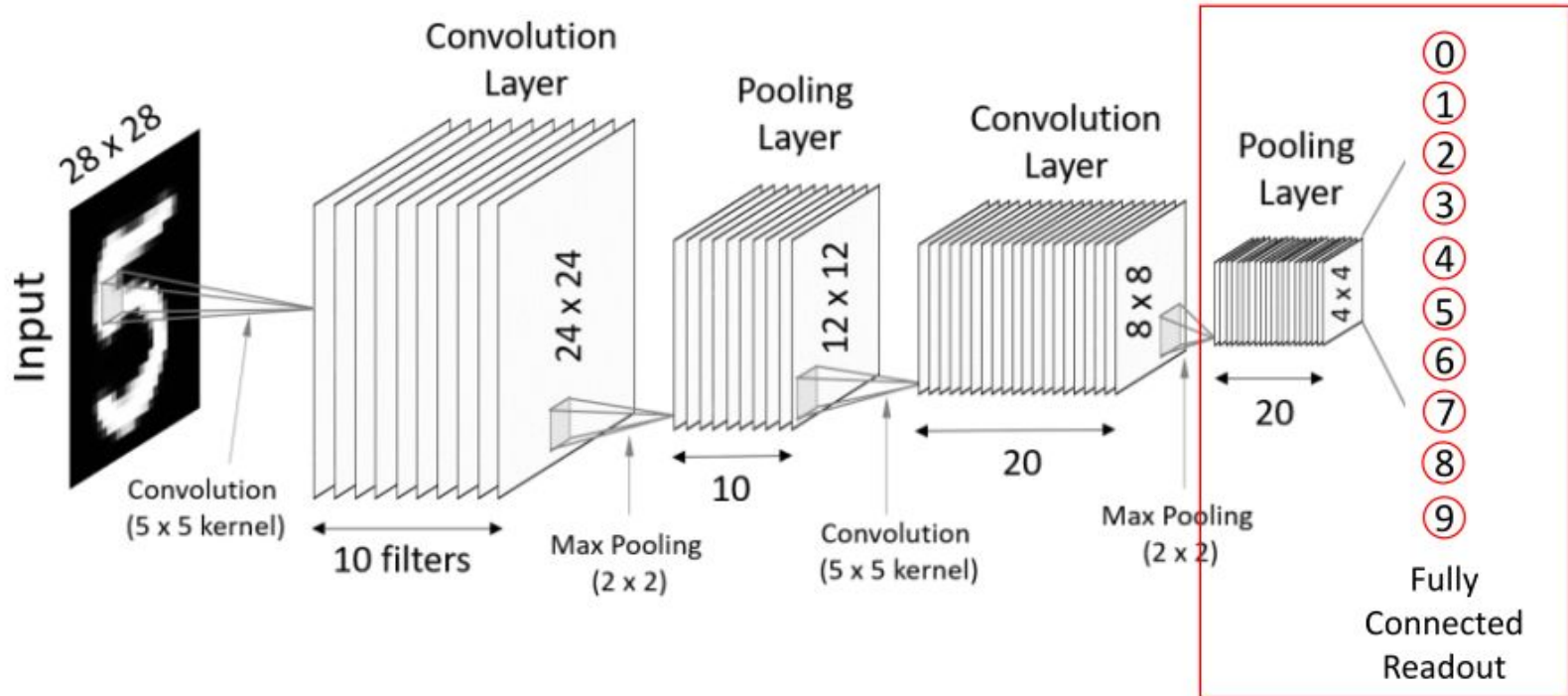
So what is Deep Learning?



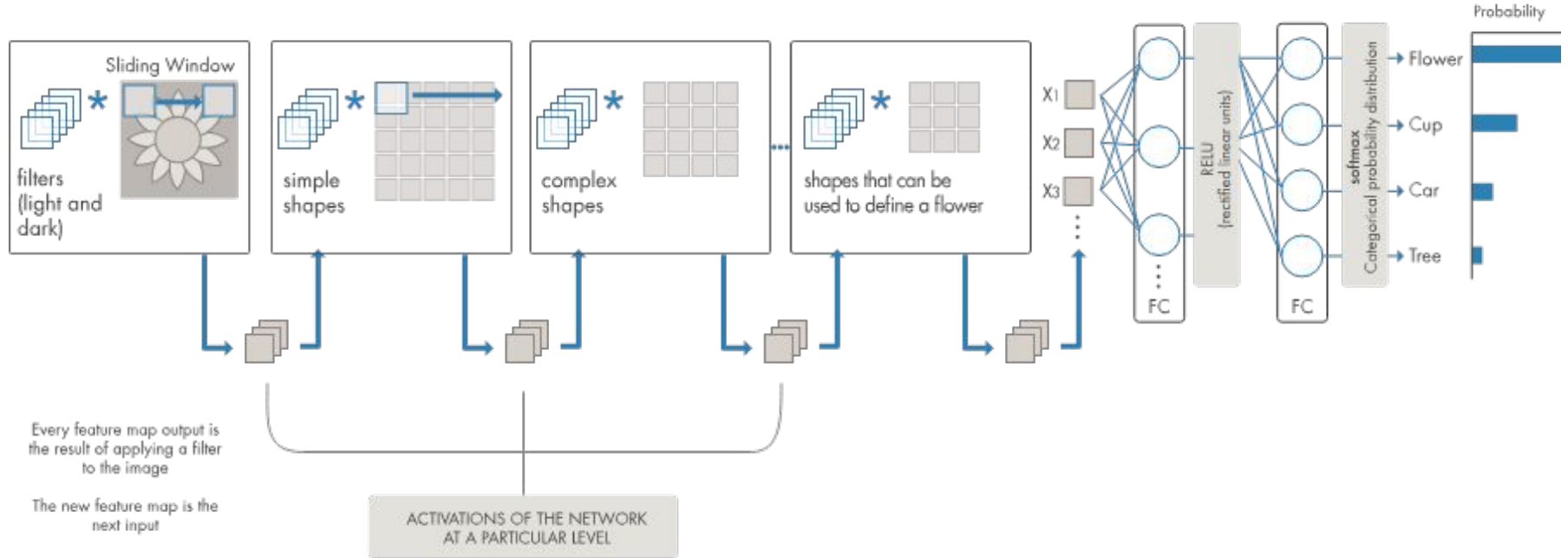
So what is Deep Learning?



So what is Deep Learning?



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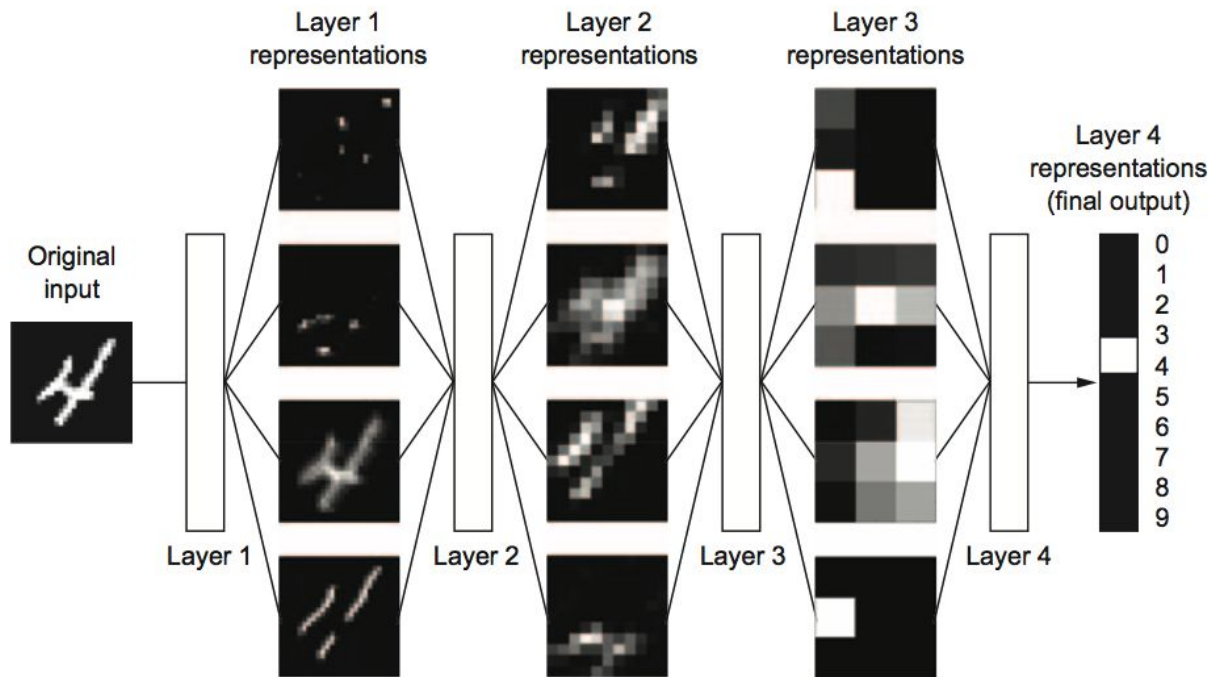
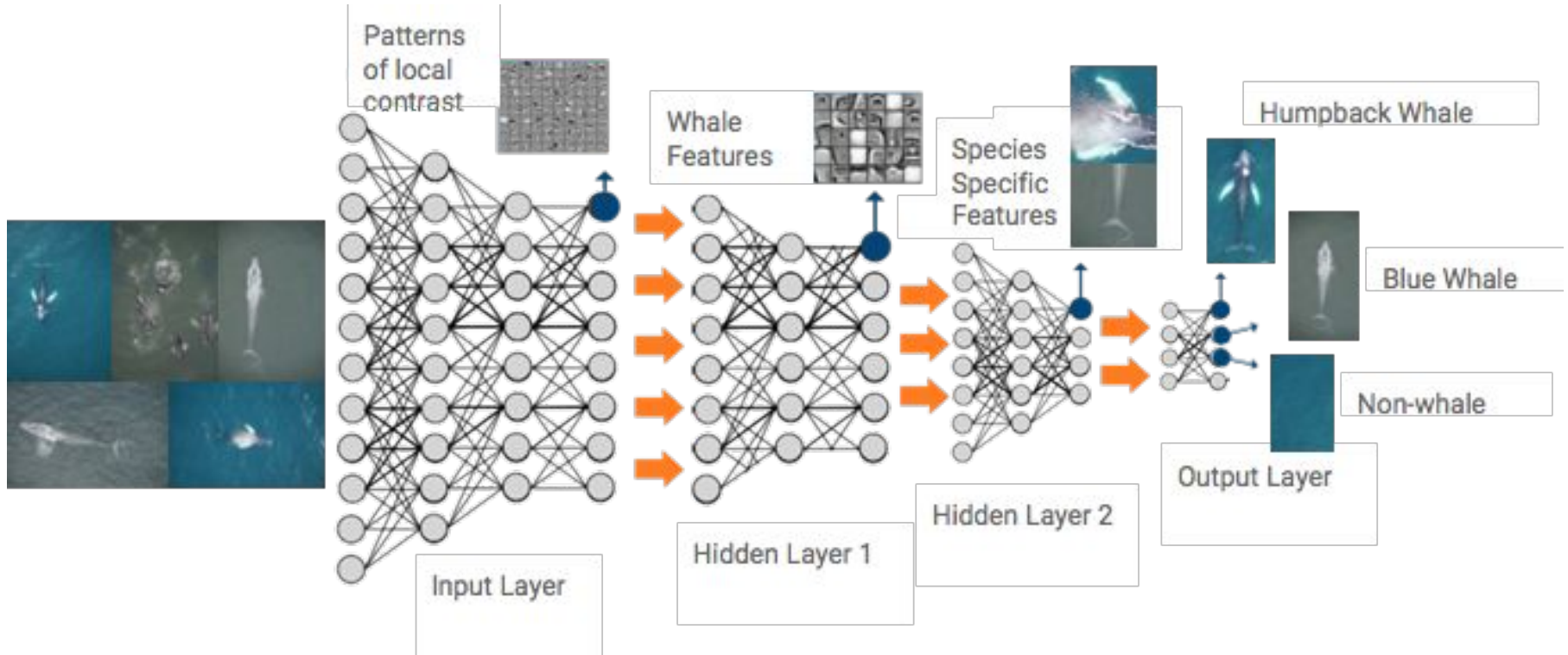


Figure 1.6 Deep representations learned by a digit-classification model

So what is Deep Learning?



Train your own Convolutional Neural Network!

Navigate to:

https://github.com/patrickcgray/deep_learning_ecology

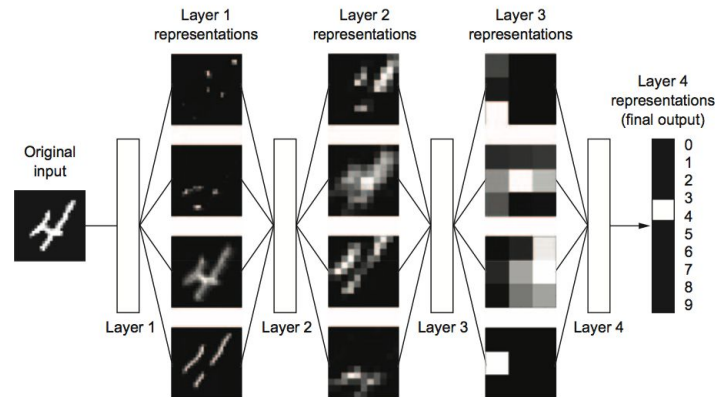


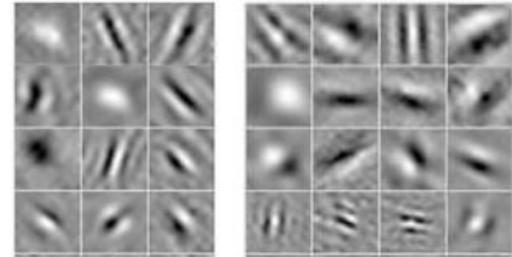
Figure 1.6 Deep representations learned by a digit-classification model

So what is Deep Learning?

Layer 1 Filters,
Convolutional Neural Network



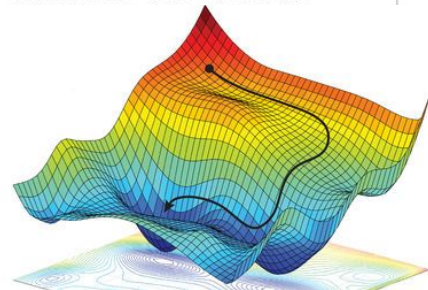
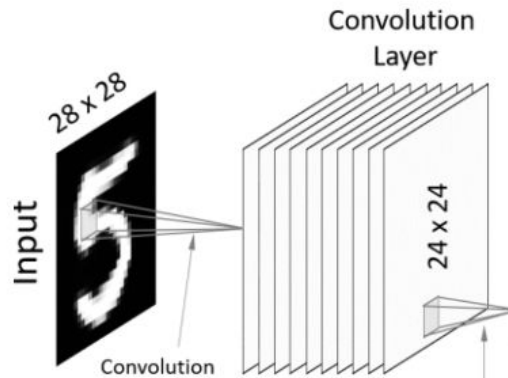
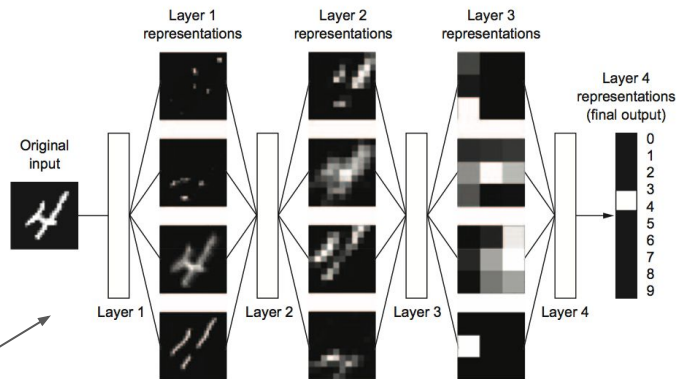
Neuron Receptive Fields,
Macaque Visual Cortex



So what is Deep Learning?

Convolutional neural networks...

- learn to recognize high-level structure in images by building hierarchical representations
- extract features via spatial convolutions with filters
- learn filters via iterative minimization of a risk function
- **have shown capabilities beyond human performance for image analysis**



Train your own CNN for Real small Images!

Navigate to:

https://github.com/patrickcgray/deep_learning_ecology