Intro to Deep Learning and Transfer Learning

How to quickly employ deep learning in my Hackathon project

Presentor: Zhizhen Qin

Co-founder and ex-president of Chinese Engineering Society (CES) at UC San Diego Master Student in Computer Science at UC San Diego Research Areas: Deep Learning, Reinforcement Learning, Safe Autonomous Driving

Outline

- What is Deep Learning
- What's "wrong" with training a model from scratch
- What is Transfer Learning (with Demo)
- What else can Deep Learning do

Outline

- What is Deep Learning
- What's "wrong" with training a model from scratch
- What is Transfer Learning (with Demo)
- What else can Deep Learning do

Question: Given some input, what is the output?

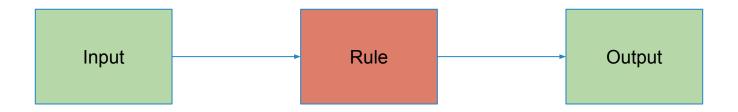


Question: Given some input, what is the output?



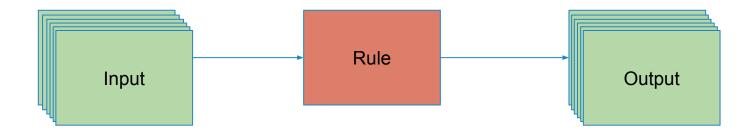
Traditionally: Rule is Known

Question: Given some input, what is the output?



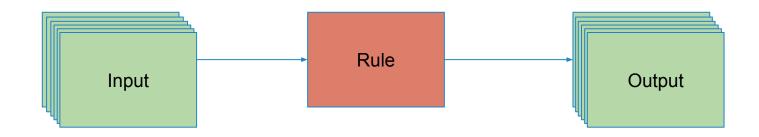
Deep Learning: Learn the rule

Question: Given some input, what is the output?



Deep Learning: Learn the rule

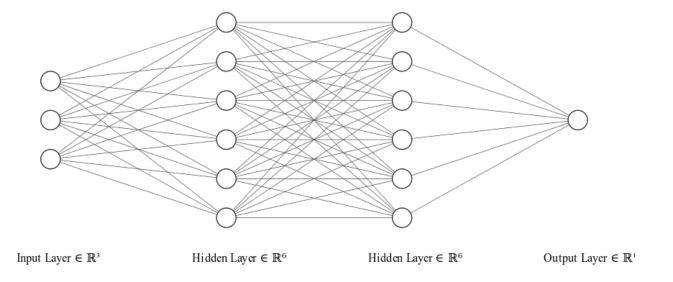
Question: Given some input, what is the output?

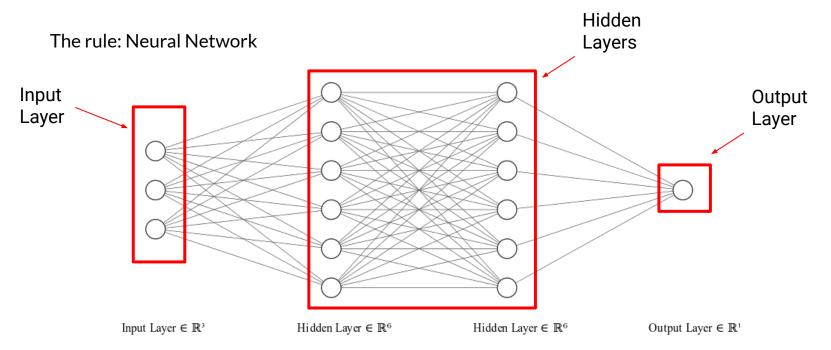


Deep Learning: Learn the rule

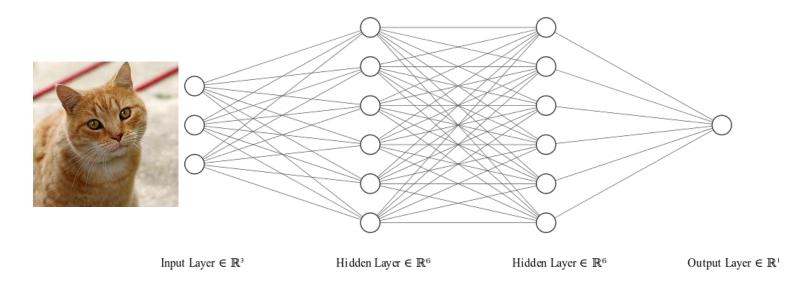
How is this "Rule" represented?

The rule: Neural Network

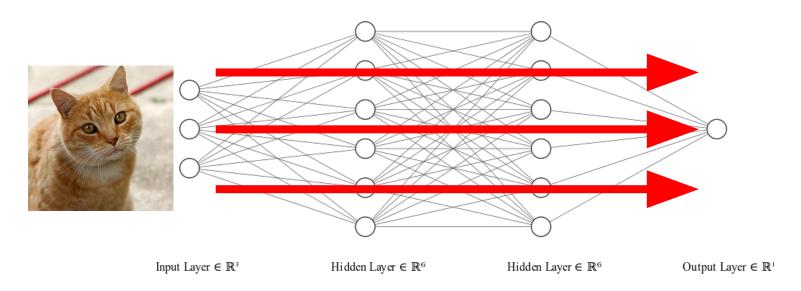




Forward Propagation

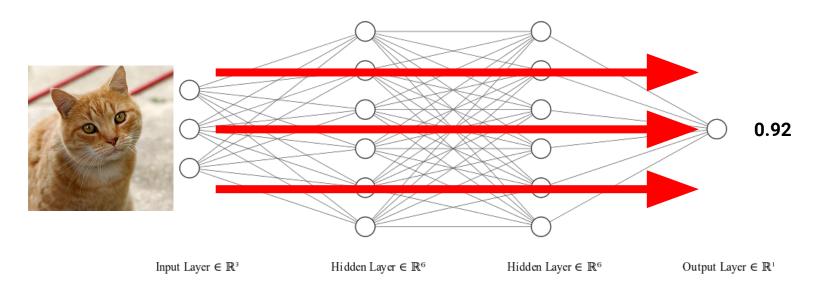


Forward Propagation

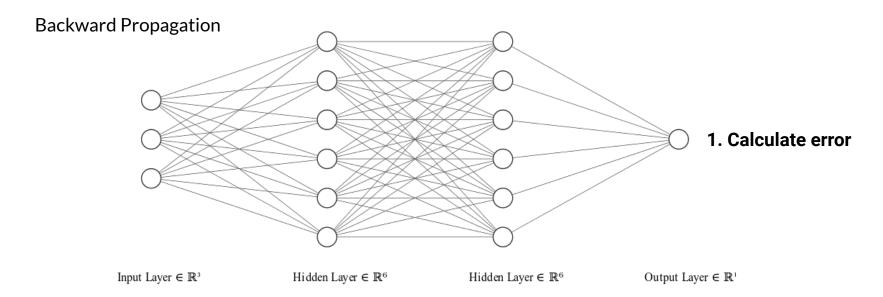


¹²

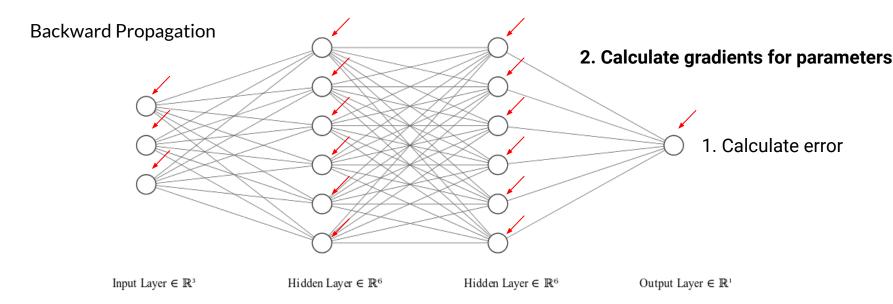
Forward Propagation



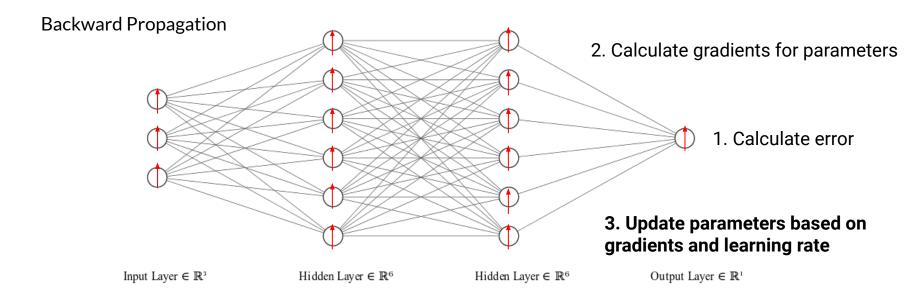
¹³



^{*} Image taken from: http://alexlenail.me/NN-SVG/



^{*} Image taken from: http://alexlenail.me/NN-SVG/



^{*} Image taken from: http://alexlenail.me/NN-SVG/

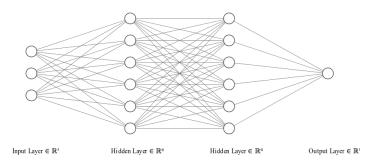
After many forward/backward propagations (oftentimes tens of thousands), the network (rule) is able to accurately perform the classification.

Outline

- What is Deep Learning
- What's "wrong" with training a model from scratch
- What is Transfer Learning (with Demo)
- What else can Deep Learning do

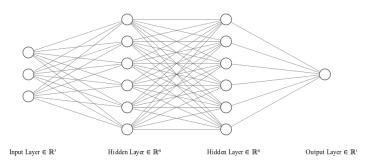
Too many parameters to train

• Too many parameters to train





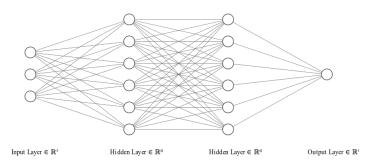
Too many parameters to train





- 256*256 image -> 65536 pixels
- 10 categories -> 10 output units
- 4 layers, with 1024 hidden units each
- Total number of parameters: 65536 * 1024 + 1024 * 1024 + 1024 * 1024 + 1024 * 1024 + 1024 * 1024 + 1024 *

Too many parameters to train

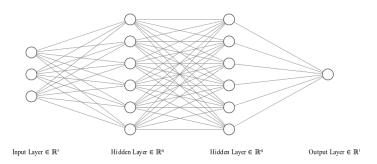




Each iteration would involve a forward path and a backward path of these parameters, and we need at least tens of thousands of such iterations

- 256*256 image -> 65536 pixels
- 10 categories -> 10 output units
- 4 layers, with 1024 hidden units each
- Total number of parameters: 65536 * 1024 + 1024 * 1024 + 1024 * 1024 + 1024 * 1024 + 1024 * 1024 + 1024 * 10 = 70,264,832

Too many parameters to train



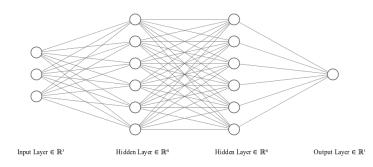


Each iteration would involve a forward path and a backward path of these parameters, and we need at least tens of thousands of such iterations

- 256*256 image -> 65536 pixels
- 10 categories -> 10 output units
- 4 layers, with 1024 hidden units each
- Total number of parameters: 65536 * 1024 + 1024 * 1024 + 1024 * 1024 + 1024 * 1024 + 1024 * 1024 + 1024 * 10 = 70,264,832

Too many parameters to train

This leaves us with a long time to properly train the model...



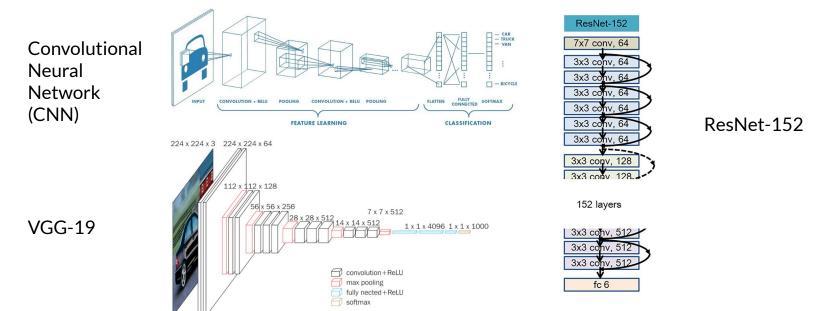


This is REALLY minimal

Each iteration would involve a forward path and a backward path of these parameters, and we need at least tens of thousands of such iterations

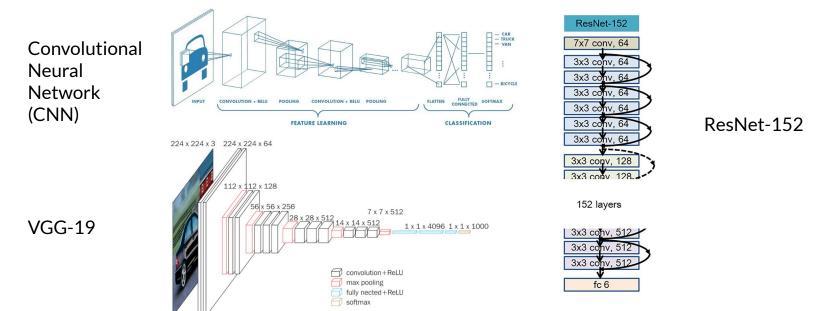
- 256*256 image -> 65536 pixels
- 10 categories -> 10 output units
- 4 layers, with 1024 hidden units each
- Total number of parameters: 65536 * 1024 + 1024 * 1024 + 1024 * 1024 + 1024 * 1024 + 1024 * 1024 + 1024 * 10 = 70,264,832

- Too many parameters to train
- Variations of models



Images: https://towardsdatascience.com/a-comprehensive-guide-to-convolutional-neural-networks-the-eli5-way-3bd2b1164a53 (CNN), https://neurohive.io/en/popular-networks/vgg16/ (VGG-19), https://www.researchgate.net/figure/The-representation-of-model-architecture-image-for-ResNet-152-VGG-19-and-two-layered_fig2_322621180 (ResNet-152)

- Too many parameters to train
- Variations of models



Images: https://towardsdatascience.com/a-comprehensive-guide-to-convolutional-neural-networks-the-eli5-way-3bd2b1164a53 (CNN), <a href="https://towardsdatascience.com/a-comprehensive-guide-to-convolutional

- Too many parameters to train
- Variations of models
- Not enough data

- Too many parameters to train
- Variations of models
- Not enough data

Typically, to train a neural network accurately classifies images, we would typically need at least thousands of samples



Outline

- What is Deep Learning
- What's "wrong" with training a model from scratch
- What is Transfer Learning (with Demo)
- What else can Deep Learning do

Key idea: Take someone else's pretrained model and use it for my application

Key idea: Take someone else's pretrained model and use it for my application

The pretrained model: trained on very large dataset (for example, imagenet, which contains **14,197,122** images and **1000** classes.

Key idea: Take someone else's pretrained model and use it for my application

The pretrained model: trained on very large dataset (for example, imagenet, which contains **14,197,122** images and **1000** classes.

Advantage: the pretrained model has a solid knowledge on the features of objects.

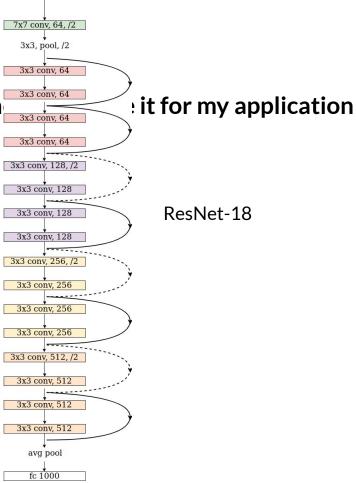
Key idea: Take someone else's pretrained model and use it for my application

How to use "someone else's model"?

Key idea: Take someone else's pretrained m

How to use "someone else's model"?

This layer was for "categorizing 1000 classes"



Image

Key idea: Take someone else's pretrained m

How to use "someone else's model"?

This layer was for "categorizing 1000 classes"

For "Doheny vs Geisel", we want two

Image 7x7 conv, 64, /2 3x3, pool, /2 3x3 conv, 64 3x3 conv. 64 : it for my application 3x3 conv, 64 3x3 conv, 64 3x3 conv, 128, /2 3x3 conv, 128 ResNet-18 3x3 conv. 128 3x3 conv, 128 3x3 conv, 256, /2 3x3 conv. 256 3x3 conv, 256 3x3 conv, 256 3x3 conv. 512. /2 3x3 conv, 512 3x3 conv, 512 3x3 conv, 512 avg pool fc 1000

Key idea: Take someone else's pretrained m

How to use "someone else's model"?

This layer was for "categorizing 1000 classes"

For "Doheny vs Geisel", we want two

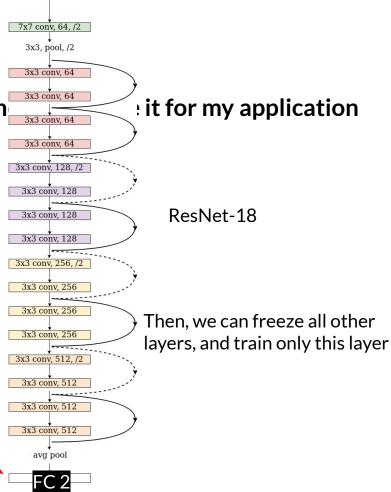
Image 7x7 conv, 64, /2 3x3, pool, /2 3x3 conv, 64 3x3 conv. 64 : it for my application 3x3 conv, 64 3x3 conv, 64 3x3 conv, 128, /2 3x3 conv, 128 ResNet-18 3x3 conv. 128 3x3 conv, 128 3x3 conv, 256, /2 3x3 conv. 256 3x3 conv, 256 3x3 conv, 256 3x3 conv. 512. /2 3x3 conv, 512 3x3 conv, 512 3x3 conv, 512 avg pool

Key idea: Take someone else's pretrained m

How to use "someone else's model"?

This layer was for "categorizing 1000 classes"

For "Doheny vs Geisel", we want two



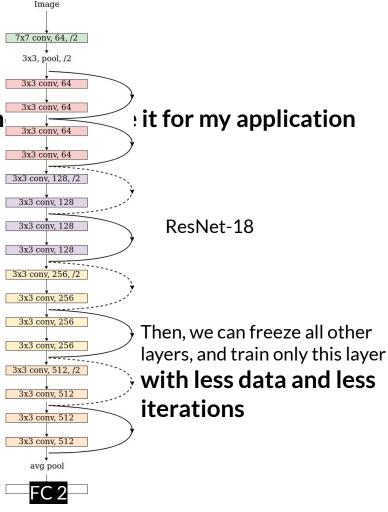
Image

Key idea: Take someone else's pretrained m

How to use "someone else's model"?

This layer was for "categorizing 1000 classes"

For "Doheny vs Geisel", we want two



Key idea: Take someone else's pretrained model and use it for my application

Where to find those "pretrained models"?

Key idea: Take someone else's pretrained model and use it for my application

Where to find those "pretrained models"?

Just search on Google and there are a lot of them

Key idea: Take someone else's pretrained model and use it for my application

Where to find those "pretrained models"?

Just search on Google and there are a lot of them

Even easier why: inside PyTorch: https://pytorch.org/docs/stable/torchvision/models.html

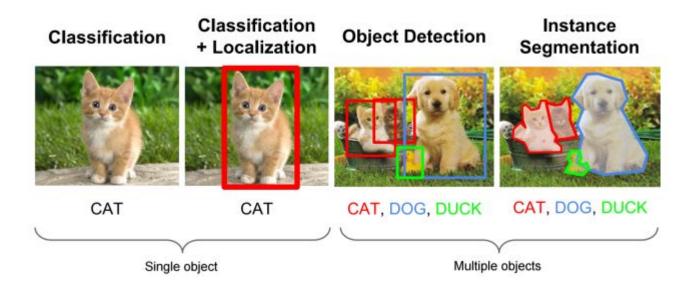
DEMO

Outline

- What is Deep Learning
- What's "wrong" with training a model from scratch
- What is Transfer Learning (with Demo)
- What else can Deep Learning do

What else can Deep Learning do

Compute Vision

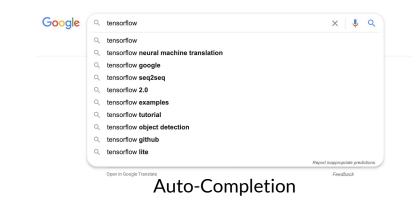


What else can Deep Learning do

Natural Language Processing



Question and Answering



Detect language

Enter text

Translation

Open in Google Translate

English

Feedback

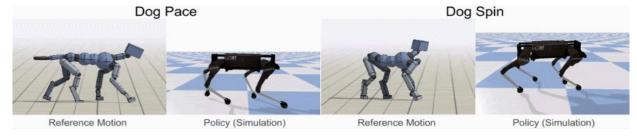
Machine Translation

What else can Deep Learning do

Reinforcement Learning

(e.g. play video games, autonomous driving, robotics, etc.)





Thanks! Questions?

Zhizhen Qin

WeChat: qzz_ckpt

Email: zhizhenqin@ucsd.edu