# Building Real Applications

With Deep Learning

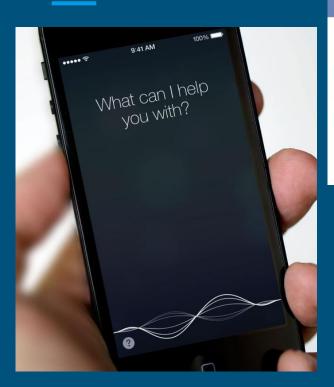
### Plan

What is Deep Learning?

Who is using it?

How do we use it?

# Deep Learning is in Everything!



#### **Photos: Suggest Tags**

This helps your friends label and share their photos, and makes it easier to find out when photos of you are posted.



#### Suggest photos of me to friends

When photos look like me, suggest tagging me

This feature uses a comparison of photos you're tagged in to suggest that friends tag you in new photo





# **Building Intuition**

Given a Problem

How to build a deep learning model to solve the problem?

### Problem

What does it mean when someone gives you a problem?

### Problem

Data (Features + Label)

Comedy?	Length (m)	Female Lead?	Super Hero?	Like?
Yes	90	No	No	No
No	120	Yes	Yes	Yes

Goal (Given features, predict label)

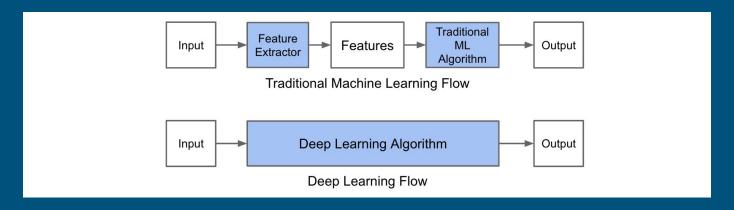
### Deep Learning vs. Other Machine Learning

**Deep Learning** 

Learn a representation for the data

Decision Trees, KNN, etc.

Implement an algorithm, given features



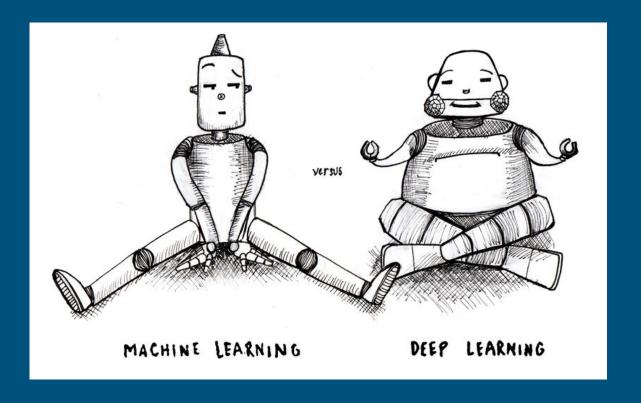
### Why Deep Learning

Powerful

Feature Learning

Software availability

Sounds good



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**Architecture (Layer Types)** 

**Loss Functions** 

**Hyper Parameters** 

Challenges

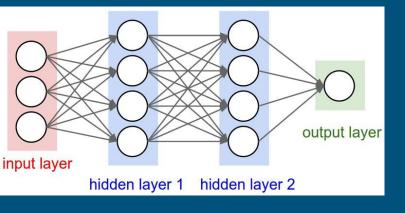
Tricks

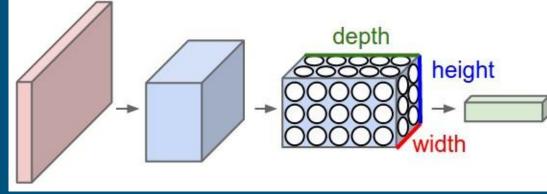
# Convolutional Neural Networks (CNN/ConvNet)

Architecture

Number of layers

Number of neurons in each layer





# Layer Types

Convolutional

**Fully Connected** 

**Pooling** 

Dropout

Activation

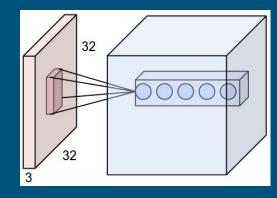
# Layer Types: Convolutional

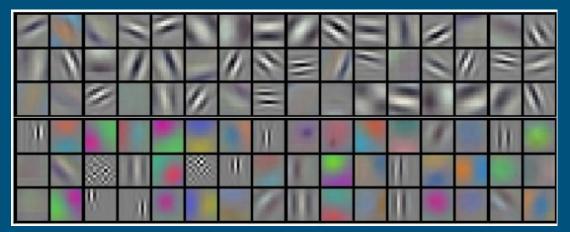
# Filters

Size Filters

Stride

Padding





# Layer Types: Pooling

In practice, most use Max Pooling.

$$f(x,y) = max(x,y)$$

$$\delta f/\delta x = 1 (x >= y)$$

$$\delta f/\delta y = 1 (y >= x)$$

# Layer Types: Pooling

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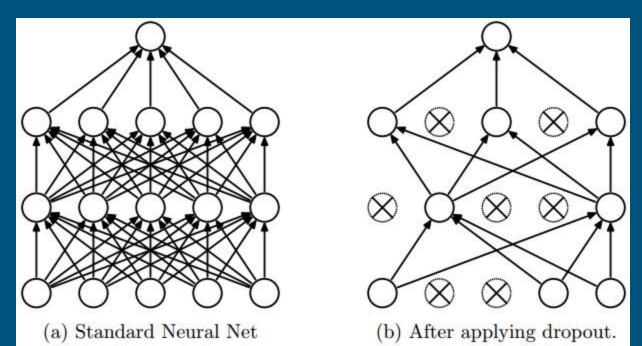
Intuition: x=4, y=2

f(x,y) = 4, and the function is not sensitive to the setting of y.

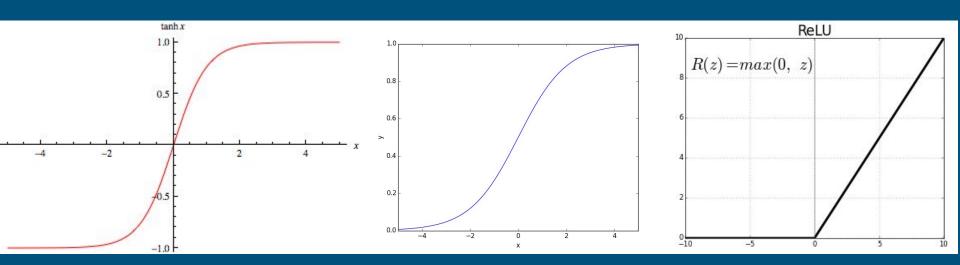
Increase y by small amount h, f(x,y+h) = 4, so there is no effect

### Layer Types: Dropout

Randomly turn off activations in a layer



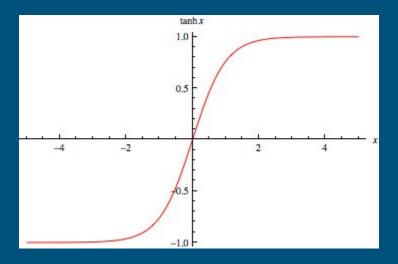
# Layer Types: Activation



### Vanishing/Exploding Gradient

Activation Function: Large input regions mapped to extremely small range.

Large change in input produces small change in output → small gradient



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### Loss Function

**Common Choices:** 

Softmax

Sigmoid Cross Entropy

Euclidean

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### Hyper Parameters

#### **Batch Size**

Updating network parameters with each batch

#### **Learning Rate**

#### Step Size

When to reduce learning rate

#### **Momentum**

Add fraction of previous weight update to current update

#### **Weight Decay**

Reduce the value of weights

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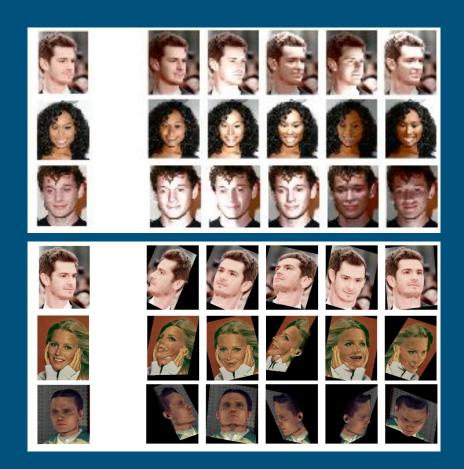
**Challenges** 

Tricks

# Overfitting

**HUGE Problem** 

What can we do to avoid it?
Early Stopping
Reduce # Parameters
Dropout
Data Augmentation



### Training Time/Memory

GPUs have limited memory (6GB)

Training can take hours, days, or weeks!

Biggest factors in deciding your architecture, and hyper parameters

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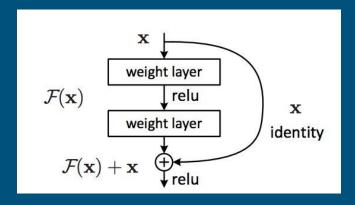
Tricks

### Common Architectures

Alexnet (8 layers) 60M parameters

VGG-Net (19 layers) 138M parameters

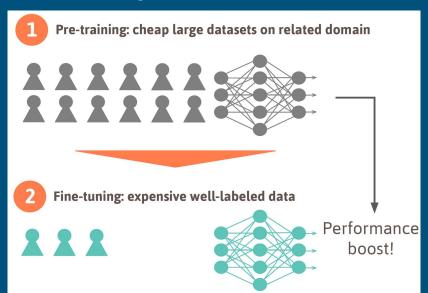
Res-Net (50+ layers) 25M parameters



### Fine-Tuning

Typically what is done in practice

Most problems do not have enough data to train these networks from scratch



### Deeper = Better?



# Debugging (What is going wrong?)

Start with a small amount of data and make sure you can overfit it

Looking at activations

Visualizers



Mouth Open



High Cheekbones



5 o'clock Shadow



Arched Eyebrows

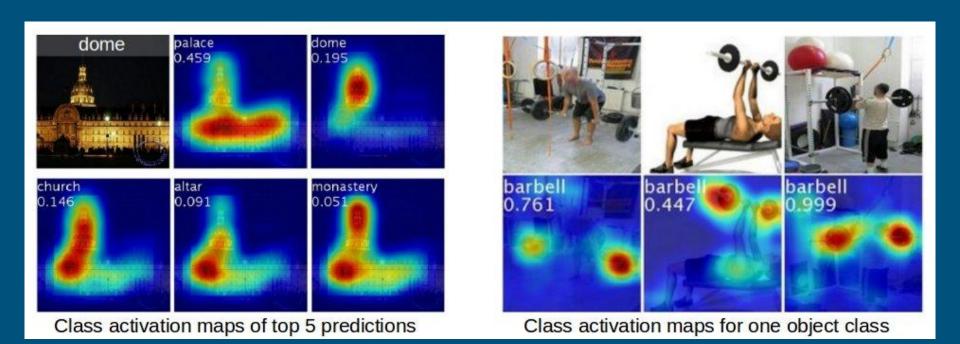


Bushy Eyebrows



Earrings

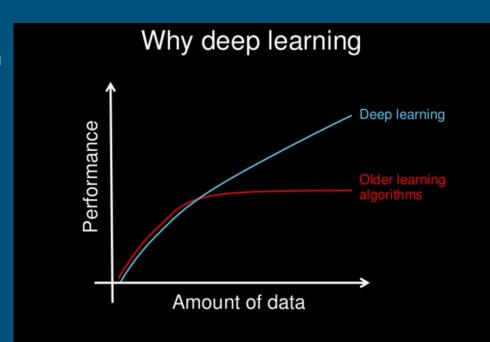
### Activations



# Deep Learning Fails

Happens A LOT

Training/Validation/Testing Distribution



How do data science techniques scale with amount of data?

# **Automatic Caption Generation**



"man in black shirt is playing guitar."



"construction worker in orange safety vest is working on road."



"two young girls are playing with lego toy."



"girl in pink dress is jumping in air."



"black and white dog jumps over bar."



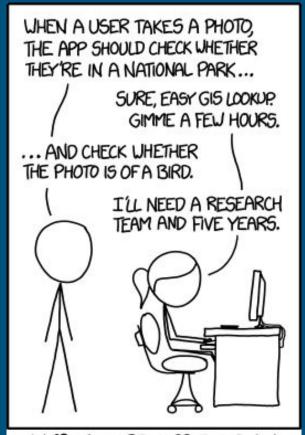
"young girl in pink shirt is swinging on swing."

# Deep Learning Fails



# Deep Learning Fails





IN CS, IT CAN BE HARD TO EXPLAIN THE DIFFERENCE BETWEEN THE EASY AND THE VIRTUALLY IMPOSSIBLE.

### Personal Experience

Getting familiar with software (Caffe, Torch, TensorFlow, MatConvNet)

#### Most Time:

Creating/Choosing Loss Function Augmenting Data

#### Less Time:

Picking Architecture
Choosing Hyperparameters

### Review

Deep Learning is a tool

Be aware of challenges

Combine different ML methods