

# OPEN DATA SCIENCE CONFERENCE

Boston | May 1 - 4 2018



@ODSC



# POWER UP YOUR COMPUTER VISION SKILLS WITH TENSORFLOW-KERAS

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ODSC East - Boston May 2018

<http://bit.ly/odsckerascv>

# **Introductions & Networking**

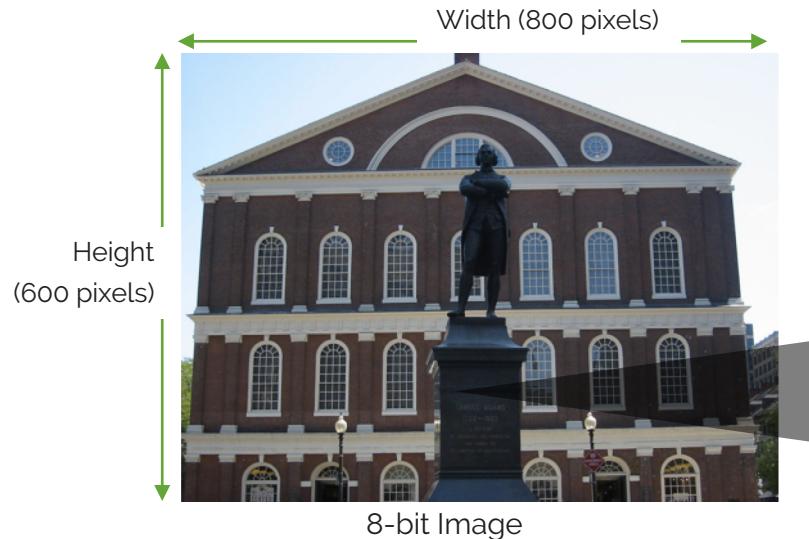
# Agenda

- Computer Vision Overview
- Overview of Keras & TensorFlow
- Getting started with Keras
- Anatomy of Keras Model
- Computer Vision Models using Keras
  - Image Classification
  - Transfer Learning
  - Object Detection
- Working with Computer Vision Datasets
- Deploying Keras Model
- Computer Vision next...

# Computer Vision Overview

# Computer Vision 101

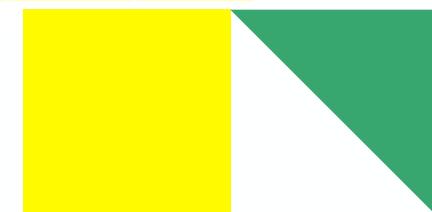
- Image Properties
  - Pixels
  - Dimension
  - Channels
  - Color Depth



Color Depth	Channel Range
1-Bit	0-1
8-Bit	0-255
16-Bit	0-65,535
24-Bit	0-16,777,215

One Pixel

Channel	Red	Green	Blue
Value	0-255	0-255	0-255



# Computer Vision 101

Classification:  
Label Item(s)  
in Image

Description:  
Relationship between objects

Segmentation: Pixel-wise segments of  
objects in any image

Detection/Localization:  
Bounding Boxes around  
objects

Goal: Build models that can perform visual tasks



# Classification & Description



<https://vqa.cloudcv.org/>

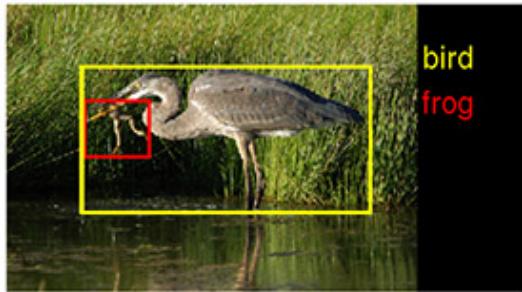
What is the man doing?

Submit

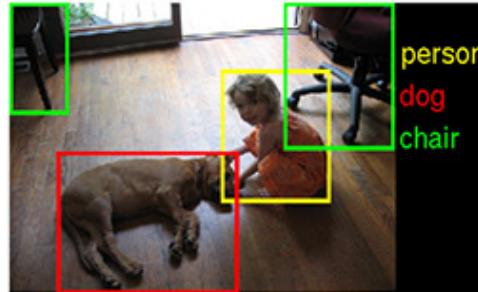
Predicted top-5 answers with confidence:



# Localization



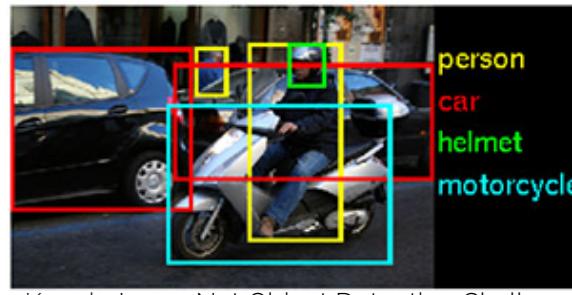
bird  
frog



person  
dog  
chair



person  
hammer  
flower pot  
power drill

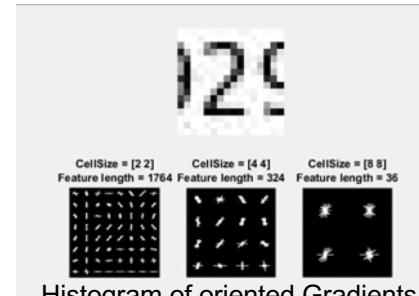


person  
car  
helmet  
motorcycle

Source: Kaggle ImageNet Object Detection Challenge

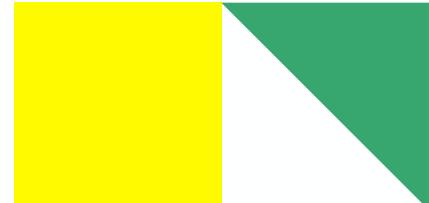
# Computer Vision prior to Deep Learning

- Objectives
  - Extract generalized features (patterns) from images
  - Determine generalized descriptors for feature
  - Use generalized features to perform visual tasks based on descriptors
- What are features?
  - Edges
  - Shapes
  - Change in Colors/Intensity



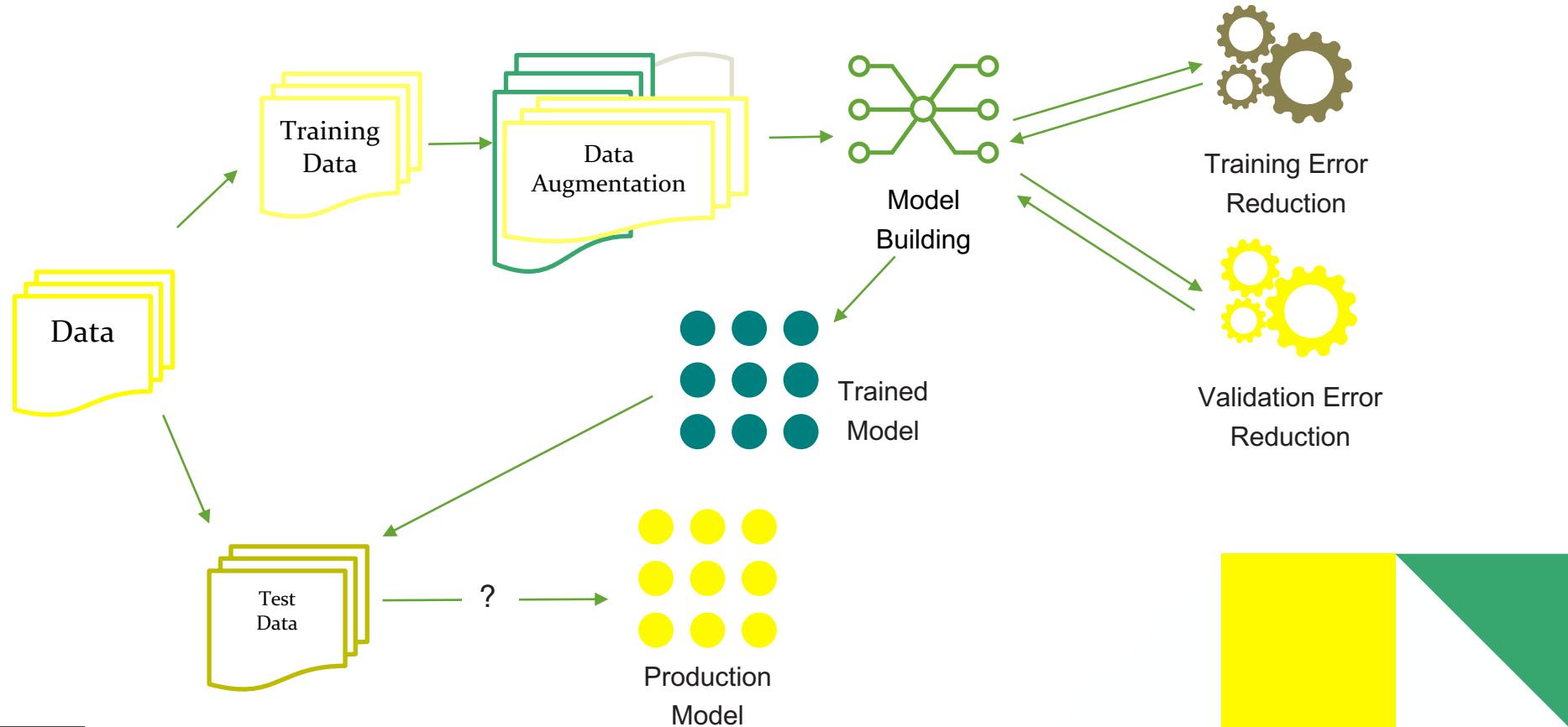
# Traditional Machine Learning → Deep Learning

- Featuring engineering challenges
  - Human scaling: Time consuming to create rules for each object
  - Computational scaling: Translating human curated features to algorithms that machines can perform at scale is tedious
- Feature Learning
  - Design computational models to learn features
- Feature Learning Challenges
  - Need large datasets with annotations
  - Model architectures become complex both to design and run

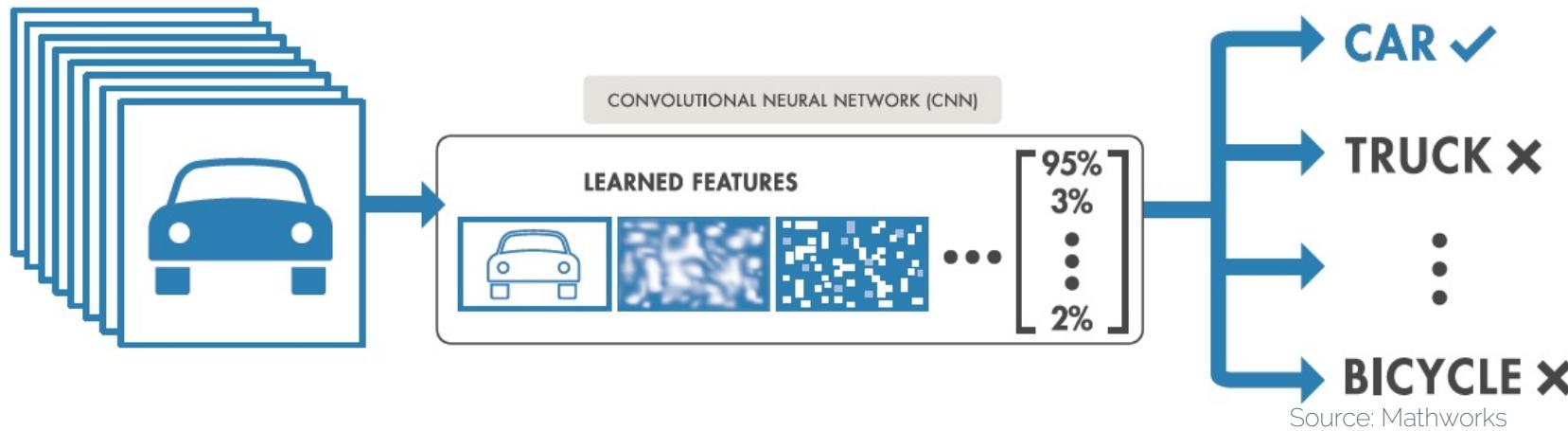


# Convolutional Neural Networks

# Deep (Supervised Machine) Learning Basics

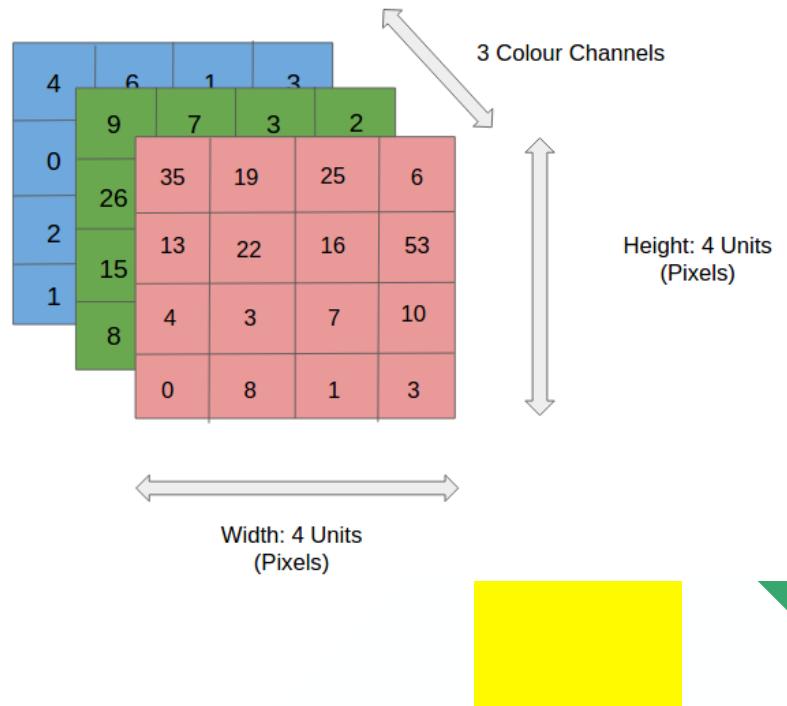


# Convolutional Neural Networks

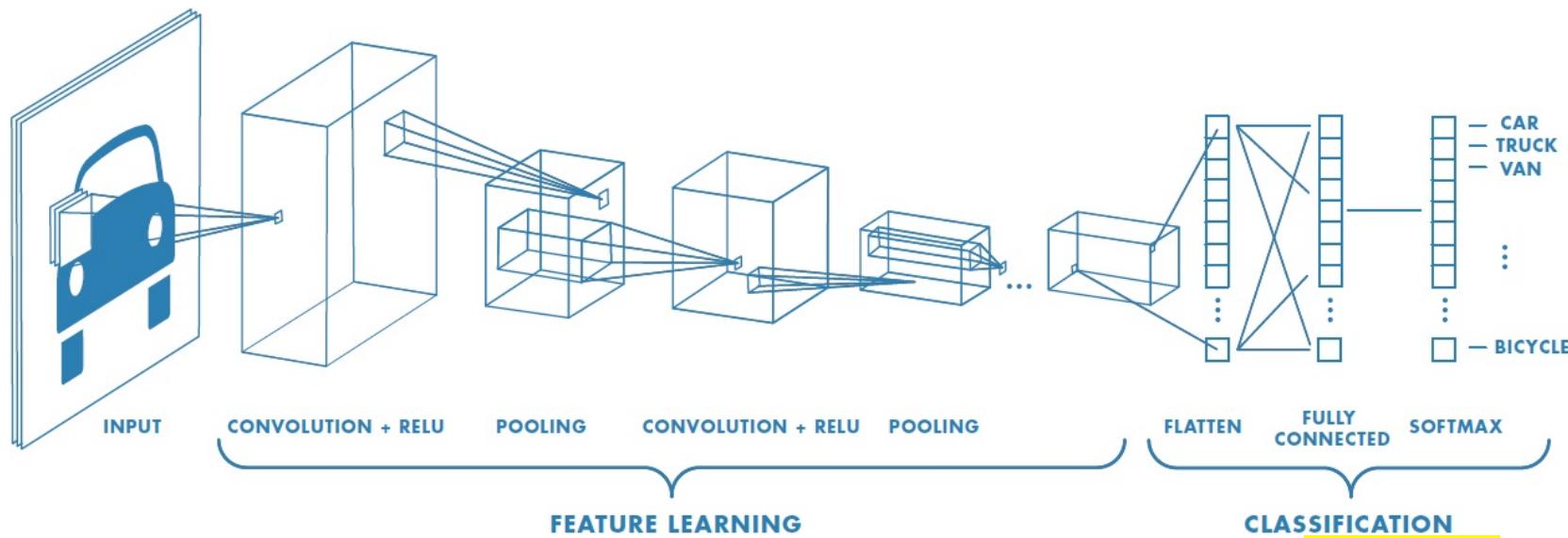


# Convolutional Neural Network Properties

- Learning Low Level Features from High Dimensional Data
- Must Have Data Properties:
  - Shift Invariant
  - Correlated
- Simply put: Changing column order should not change meaning of Data
- Naturally Suitable Data Types:
  - Images, Audio, Video
  - Time Series

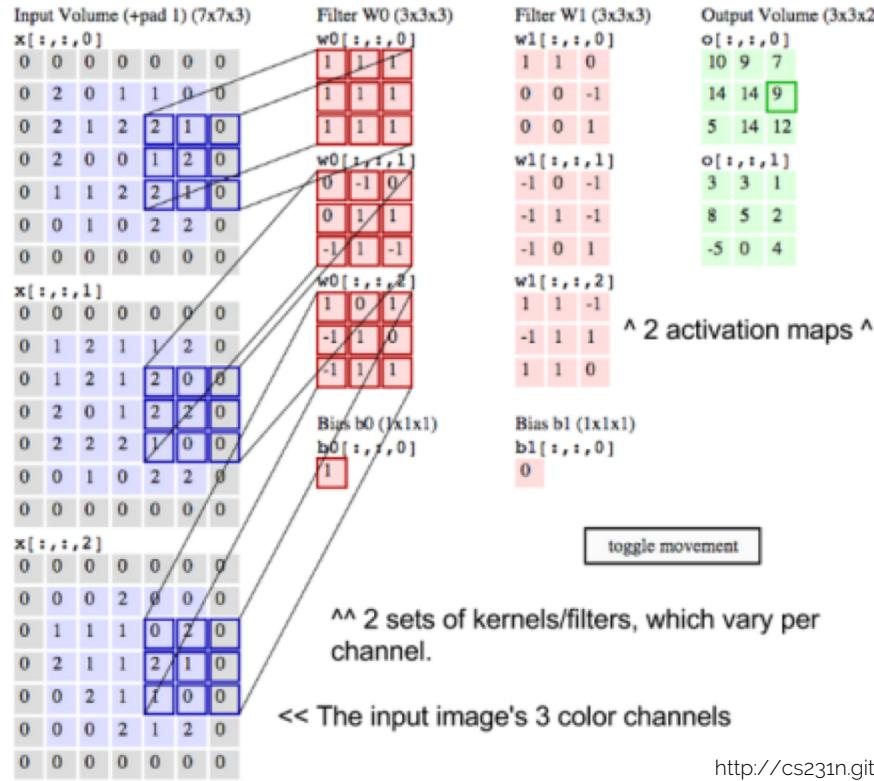


# Convolutional Neural Networks



Source: Mathworks

# Convolution



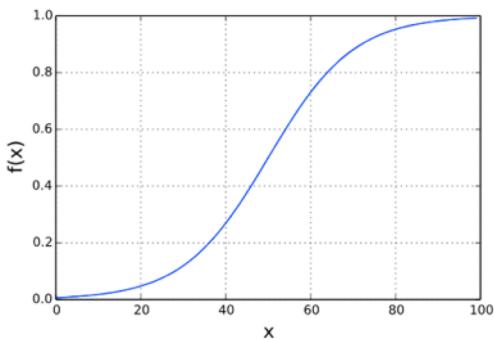
3 <sub>0</sub>	3 <sub>1</sub>	2 <sub>2</sub>	1	0
0 <sub>2</sub>	0 <sub>2</sub>	1 <sub>0</sub>	3	1
3 <sub>0</sub>	1 <sub>1</sub>	2 <sub>2</sub>	2	3
2	0	0	2	2
2	0	0	0	1

12.0	12.0	17.0
10.0	17.0	19.0
9.0	6.0	14.0

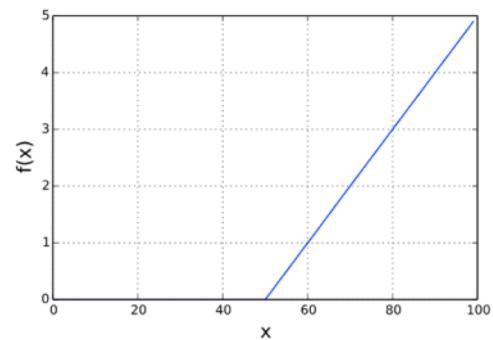
# Activation

- Mechanism that decides to activate or not activate a neuron in a neural network

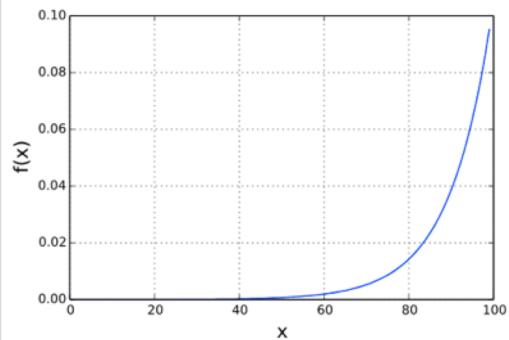
Sigmoid



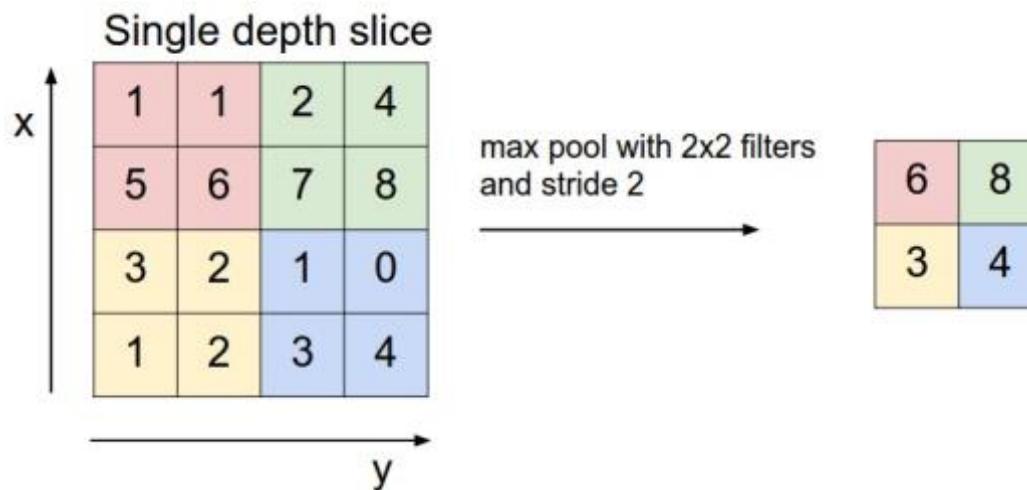
ReLU



Softmax



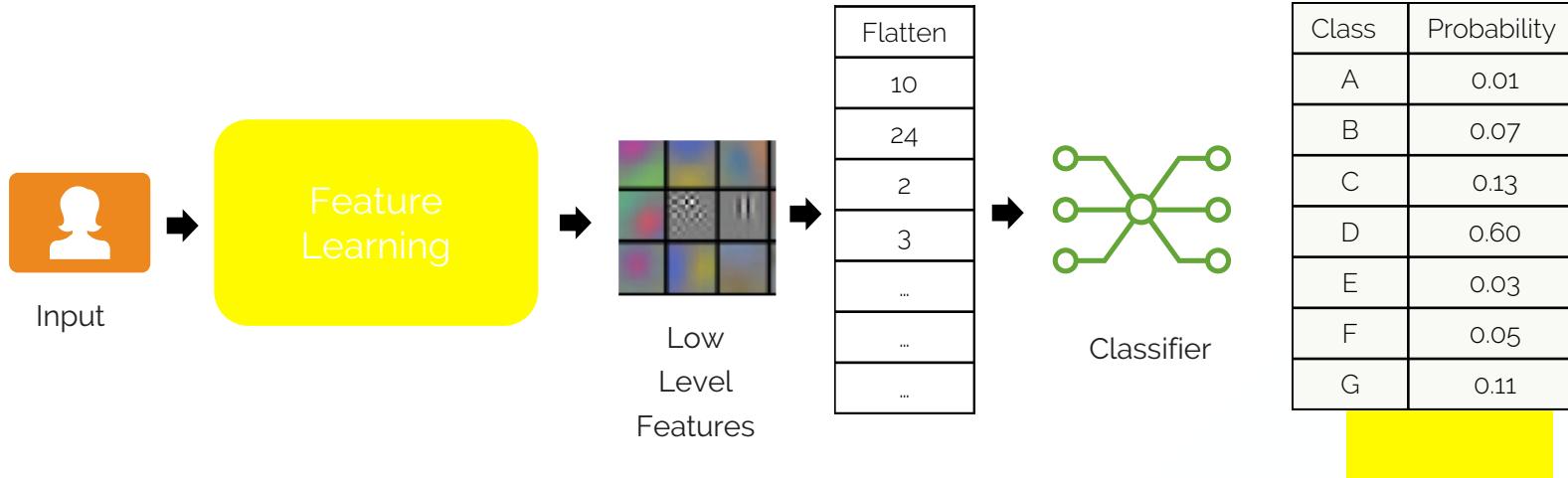
# Pooling



Source: Stanford University CS 231 Course:  
<http://cs231n.github.io/convolutional-networks/#pool>

# Fully Connected

- Traditional Machine Learning: Feature Engineering
- Deep Learning: Feature Learning
- Learned Features are fed to traditional classifier such as Multivariate Logistic Regression (Softmax)



# Loss Reduction

Class	Probability	Truth	Loss/ Error
A	0.01	0	0
B	0.07	0	0
C	0.13	0	0
D	0.60	1	0.22
E	0.03	0	0
F	0.05	0	0
G	0.11	0	0

Cross Entropy Loss

One-Hot Encoded Vector of Class Label

## Back Propagation

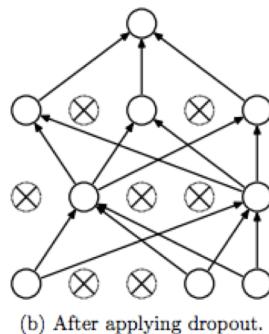
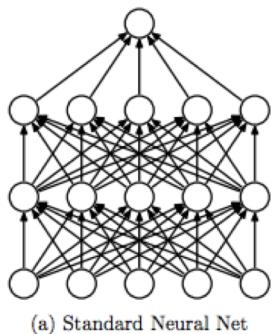
- Account for loss across the weights and biases in the network
- Use cost function provided by Back Propagation to optimize the weights and biases to reduce loss
- Variety of Stochastic Gradient Descent techniques used for optimization

Overview of Gradient Descent Techniques:  
<http://ruder.io/optimizing-gradient-descent/>

# Generalization

## Dropout

Randomly remove learned nodes from network



Dropout: A Simple Way to Prevent Neural Networks from Overfitting  
<http://www.jmlr.org/papers/volume15/srivastava14a/srivastava14a.pdf>

## Batch Normalization

- During training input data shifts as weights and parameter adjust values
- To prevent internal covariate shift, normalize each batch by mean and variance

Batch Normalization: Accelerating Deep Network Training by Reducing Internal Covariate Shift  
<https://arxiv.org/abs/1502.03167>