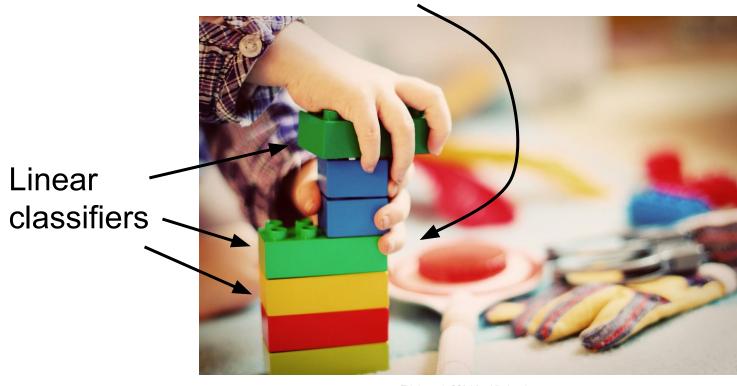
Linear Classification

Neural Network



This image is CC0 1.0 public domain

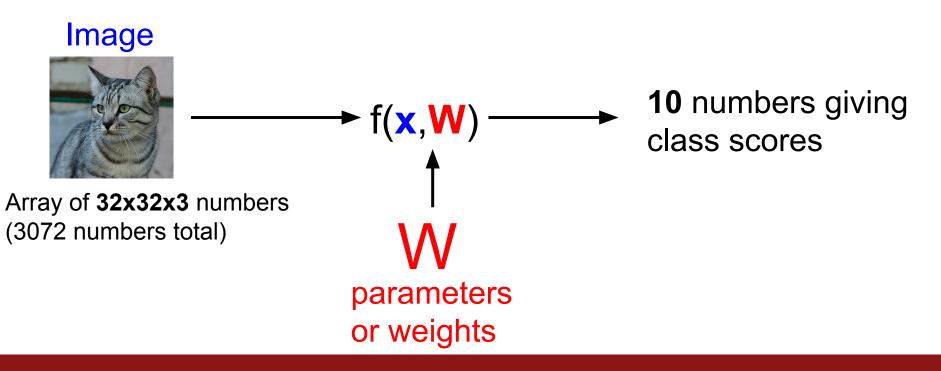
Recall CIFAR10



50,000 training images each image is **32x32x3**

10,000 test images.

Parametric Approach

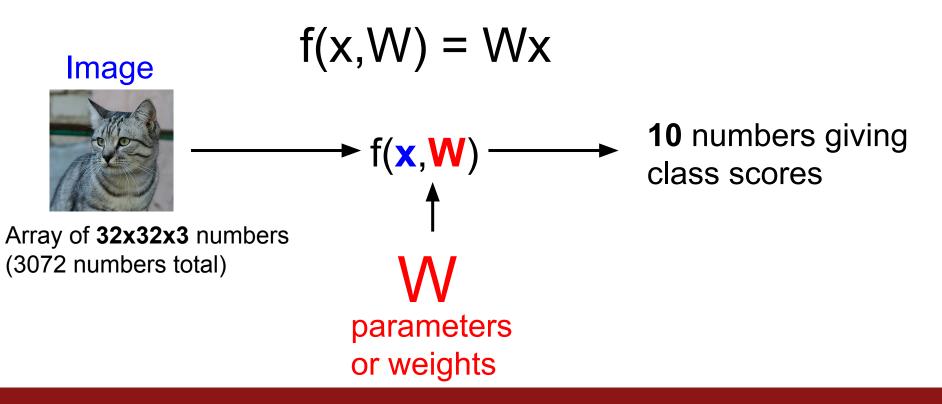


Fei-Fei Li & Justin Johnson & Serena Yeung

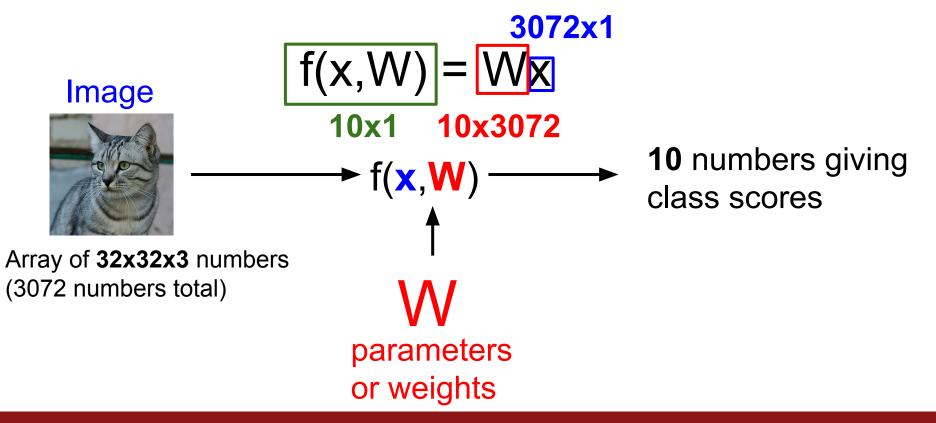
Lecture 2 - 51

April 5, 2018

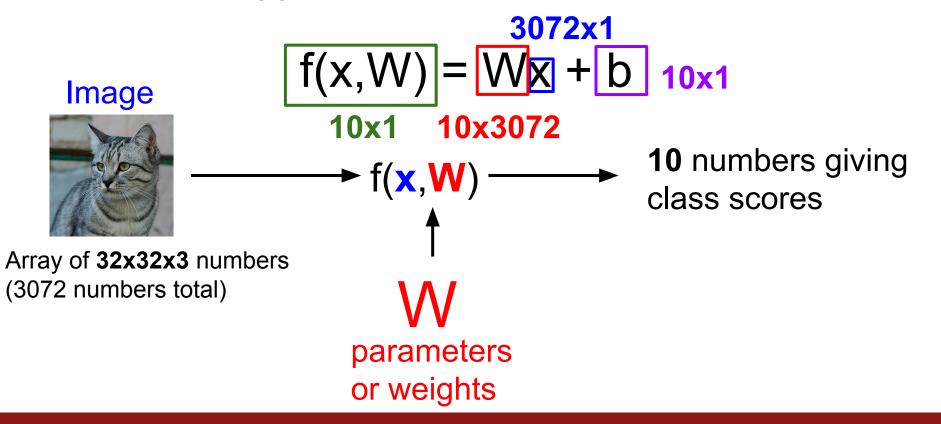
Parametric Approach: Linear Classifier

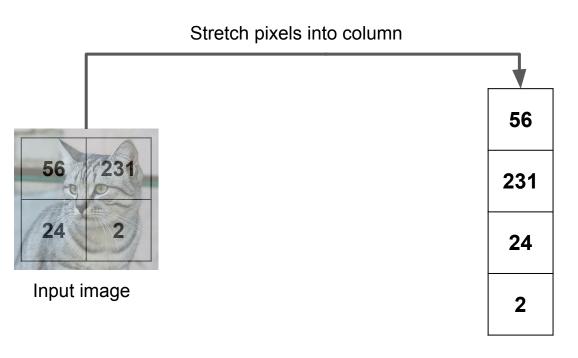


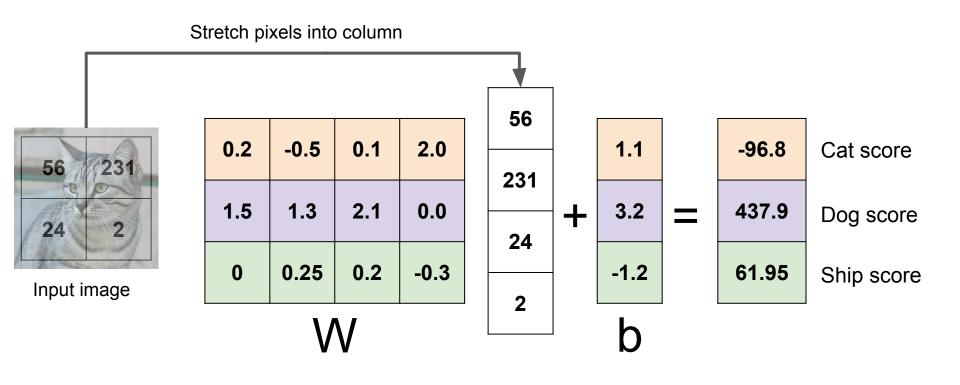
Parametric Approach: Linear Classifier



Parametric Approach: Linear Classifier

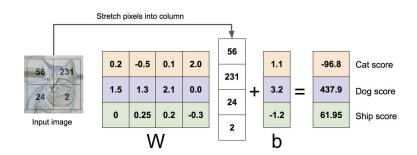


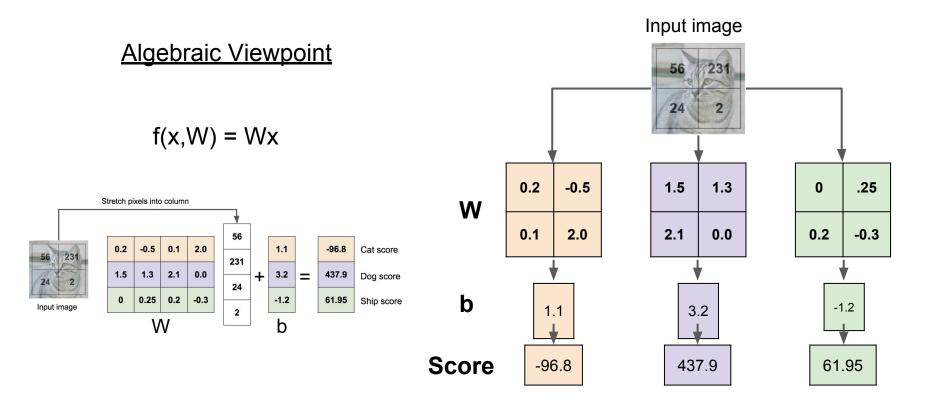




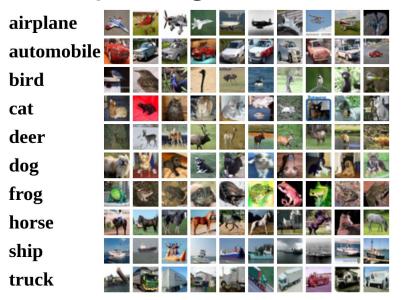
Algebraic Viewpoint

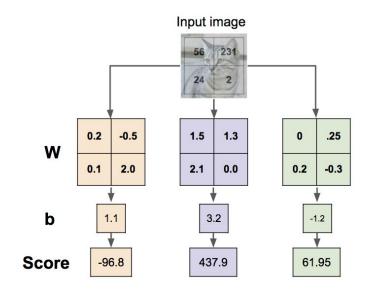
$$f(x,W) = Wx$$



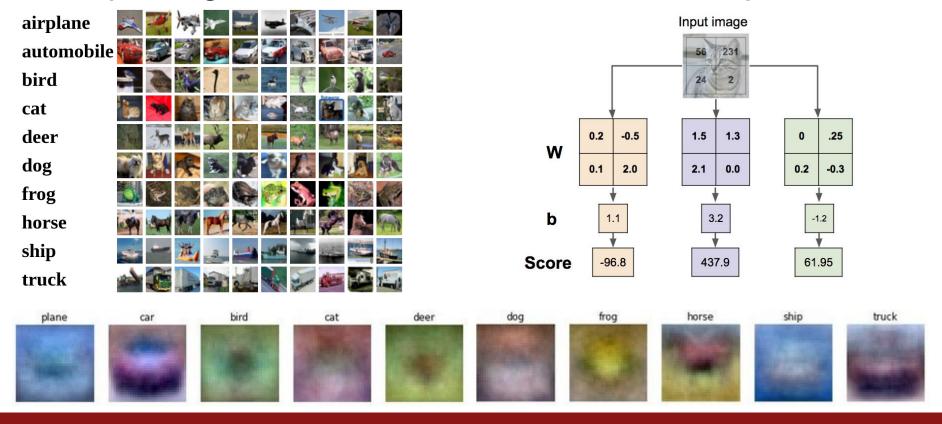


Interpreting a Linear Classifier

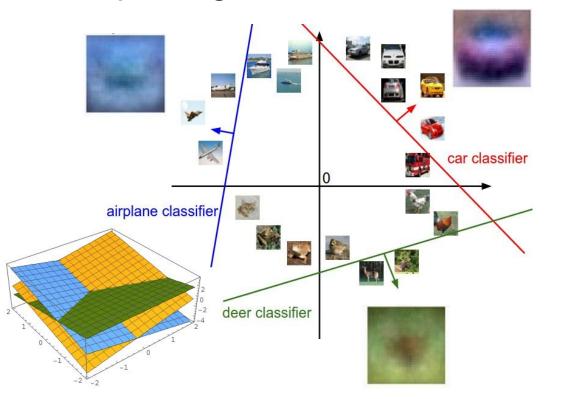




Interpreting a Linear Classifier: Visual Viewpoint



Interpreting a Linear Classifier: Geometric Viewpoint



$$f(x,W) = Wx + b$$



Array of **32x32x3** numbers (3072 numbers total)

Plot created using Wolfram Cloud

Cat image by Nikita is licensed under CC-BY 2.0

Hard cases for a linear classifier



First and third quadrants

Class 2

Second and fourth quadrants

Class 1:

1 <= L2 norm <= 2

Class 2

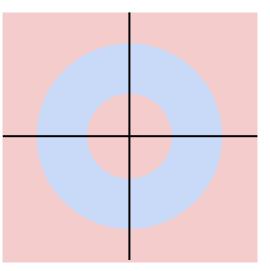
Everything else

Class 1:

Three modes

Class 2

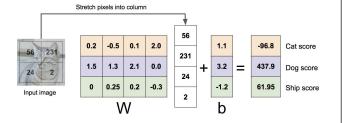
Everything else



Linear Classifier: Three Viewpoints

Algebraic Viewpoint

$$f(x,W) = Wx$$



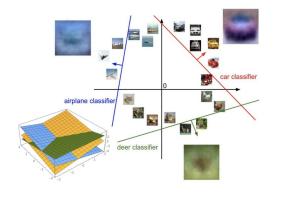
Visual Viewpoint

One template per class



Geometric Viewpoint

Hyperplanes cutting up space



So far: Defined a (linear) score function f(x,W) = Wx + b

Example class scores for 3 images for some W:

How can we tell whether this W is good or bad?





airplane	-3.45	-0.51	3.42
automobile	-8.87	6.04	4.64
bird	0.09	5.31	2.65
cat	2.9	-4.22	5.1
deer	4.48	-4.19	2.64
dog	8.02	3.58	5.55
frog	3.78	4.49	-4.34
horse	1.06	-4.37	-1.5
ship	-0.36	-2.09	-4.79
truck	-0.72	-2.93	6.14