

Image Classification with PyTorch

PRE-PROCESSING IMAGES TO USE IN MACHINE
LEARNING MODELS



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Overview

Image classification using machine learning

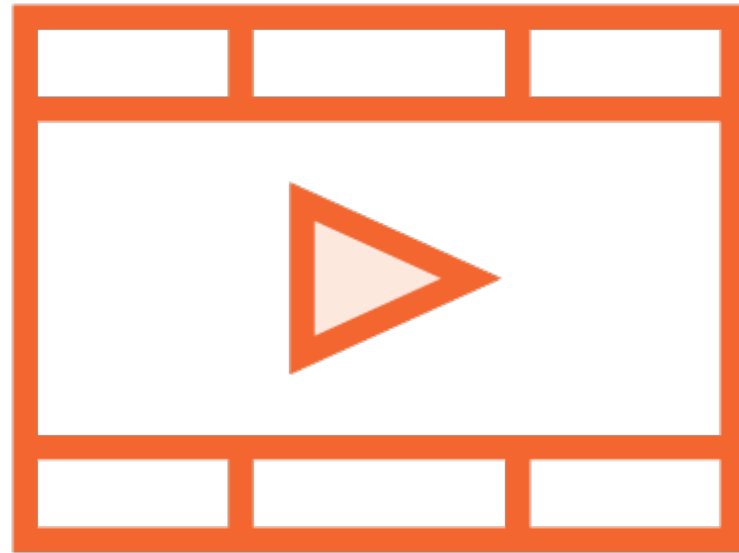
Representing images as tensors

Need for image pre-processing

Common image pre-processing techniques

Prerequisites and Course Outline

Prerequisites

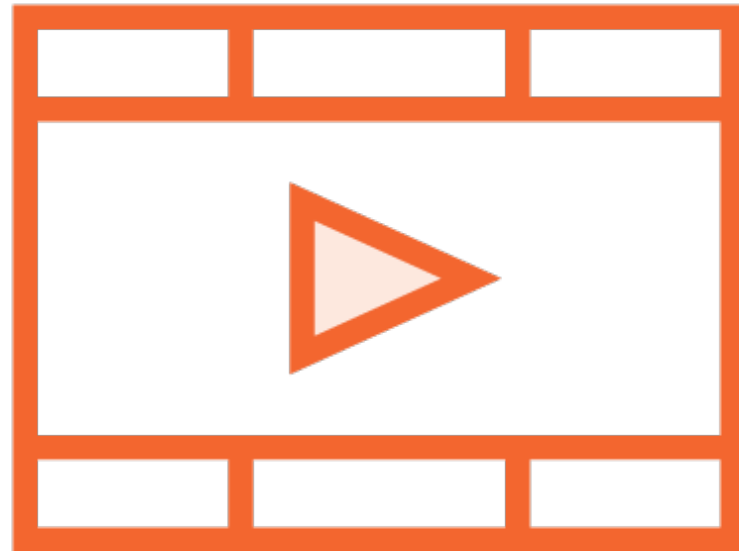


Basic Python programming

Build and training machine learning models

Worked with PyTorch to build simple neural networks

Prerequisite Courses



Foundations of PyTorch

Building your first PyTorch solution

Course Outline



Images as features and pre-processing techniques

Drawbacks of Deep Neural Networks (DNNs) for image classification

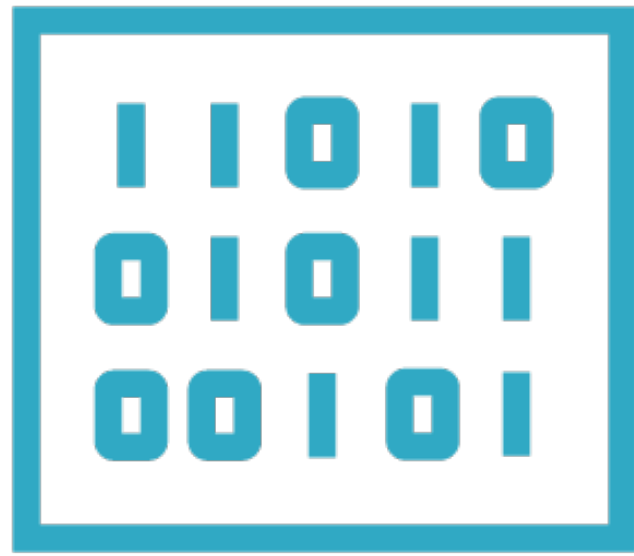
Introducing Convolutional Neural Networks (CNNs)

Hyperparameter tuning

Pre-trained models

Image Recognition

Image Recognition



Images represented
as pixels



Identify edges,
colors, shapes

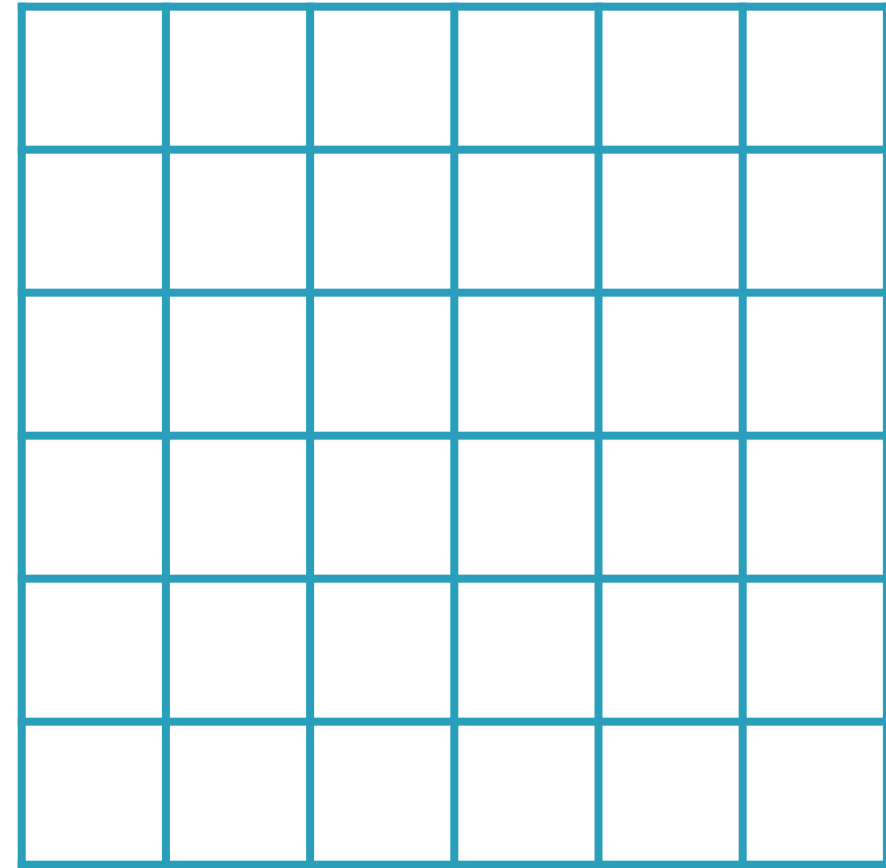


A photo of a
horse

Images as Tensors



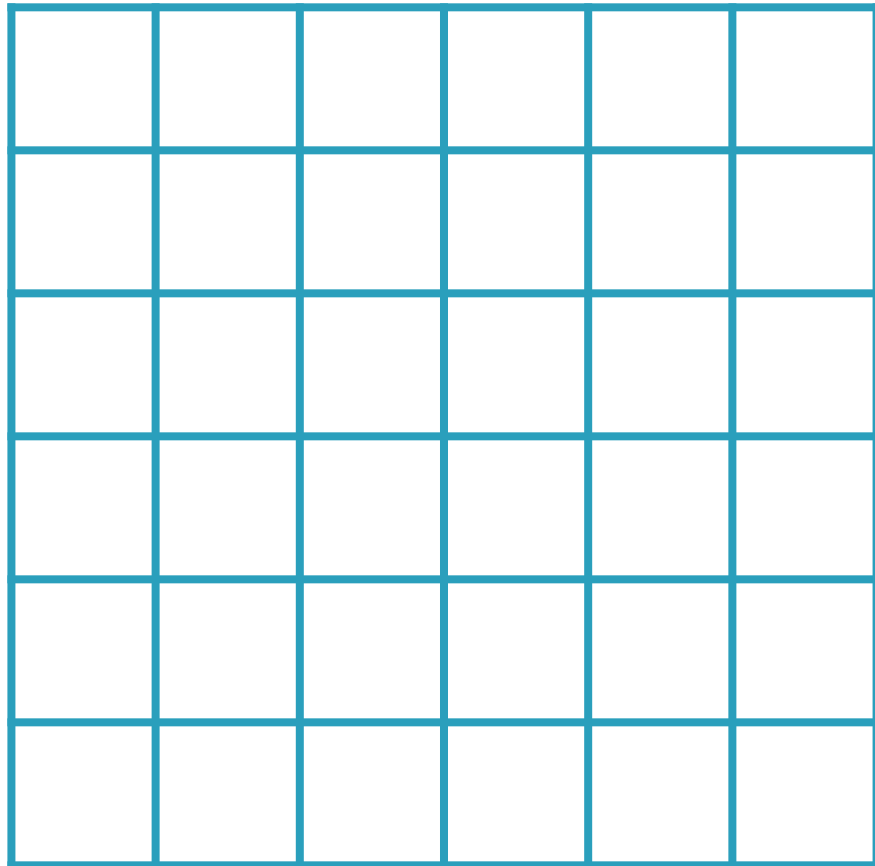
Images as Tensors



Each pixel holds a value based on the type of image



RGB Images

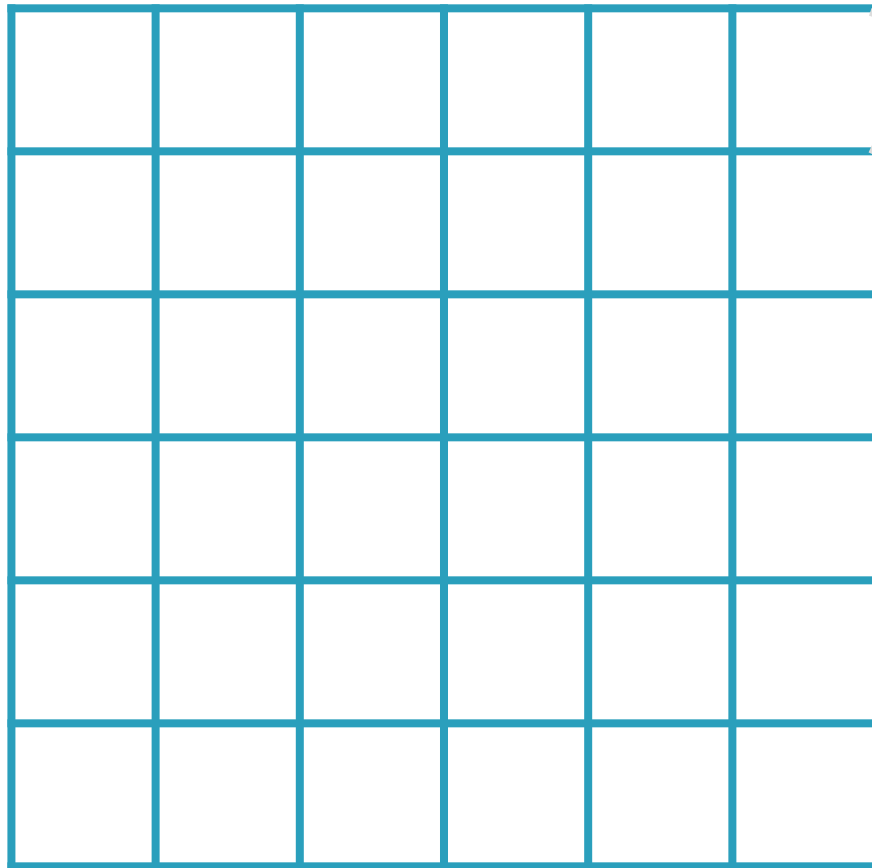


**RGB values are for
color images**

R, G, B: 0-255



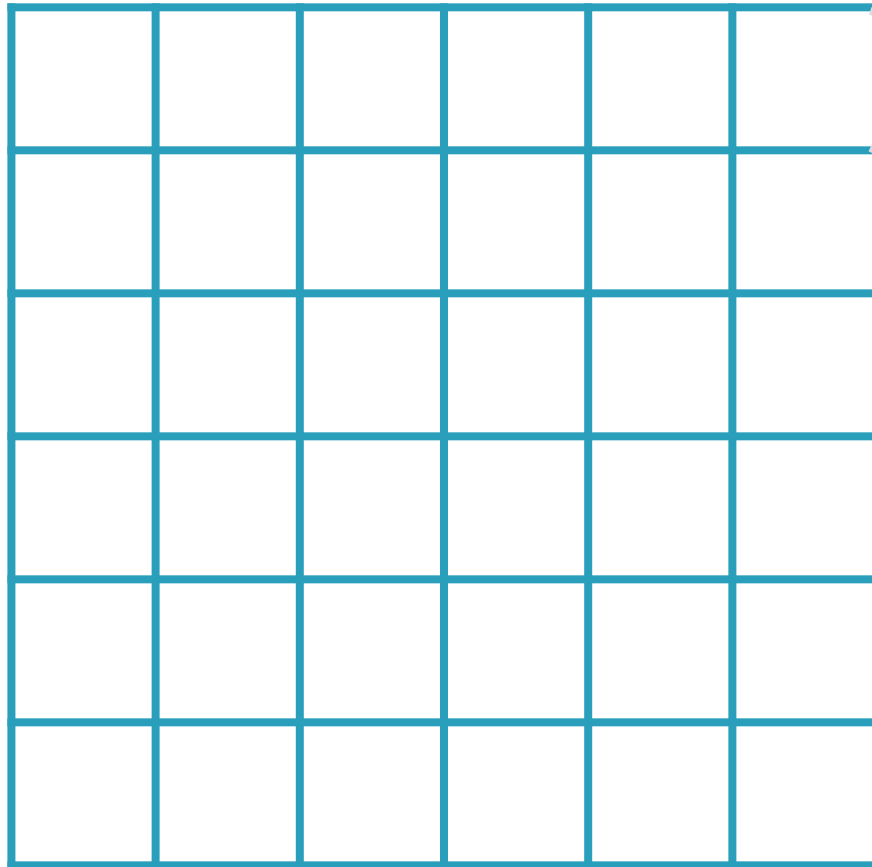
RGB Images



255, 0, 0



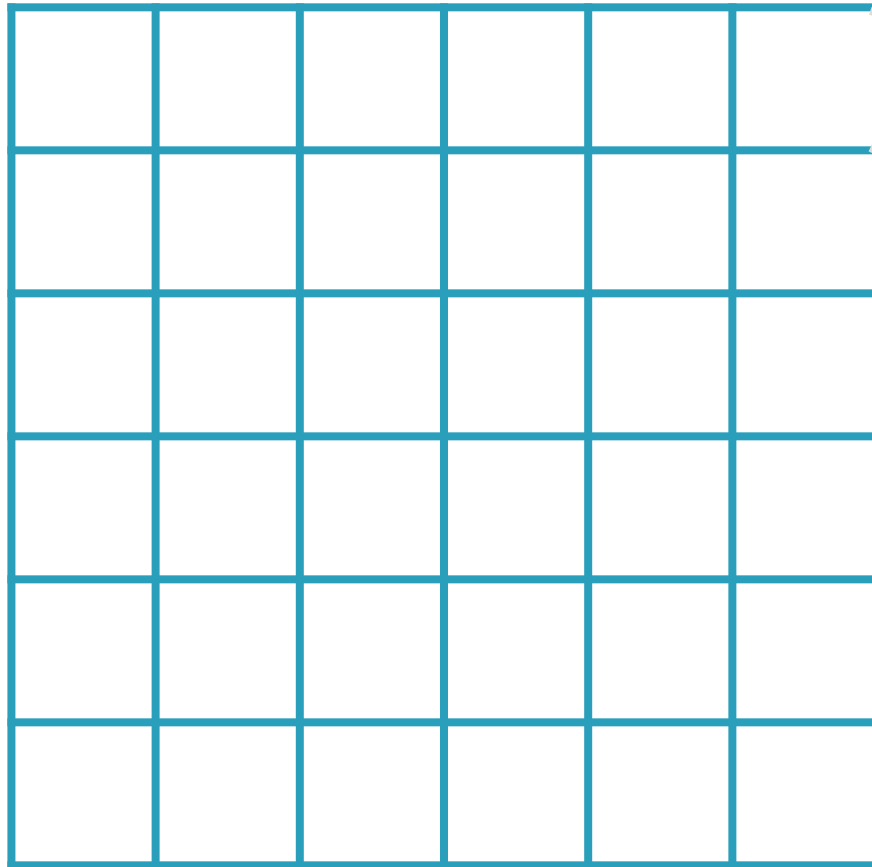
RGB Images



0, 255, 0



RGB Images

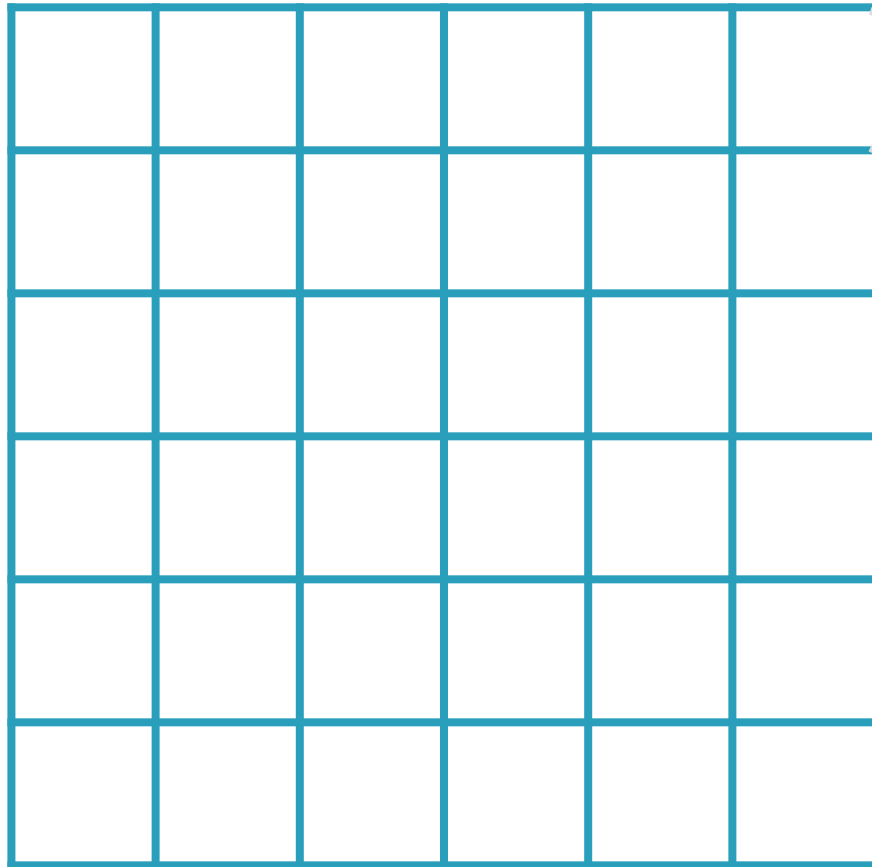


0, 0, 255

3 values to represent
color, **3** channels



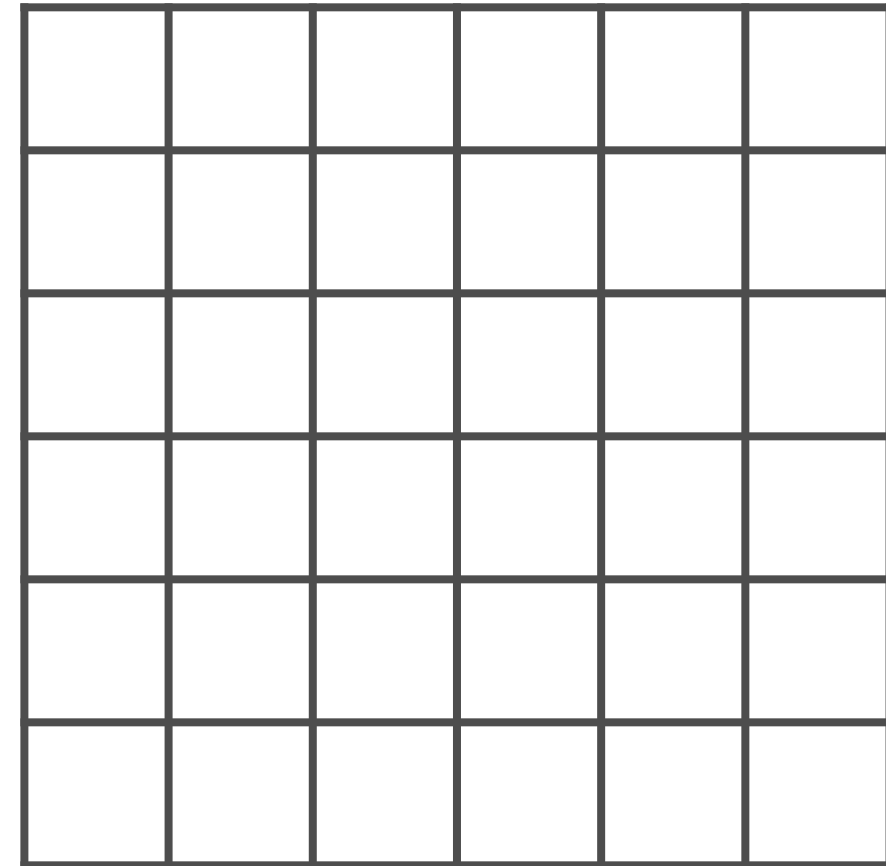
RGB Images



0, 0, 255

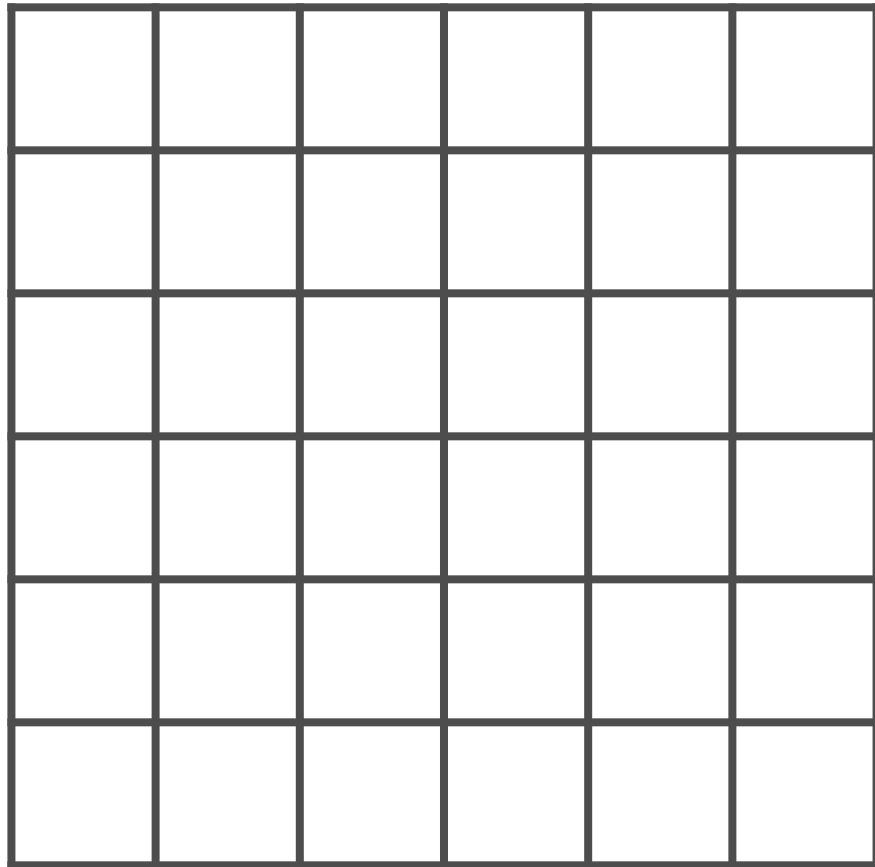
These are often scaled to be in the 0-1 range as neural networks work better with smaller numbers

Grayscale Images





Grayscale Images

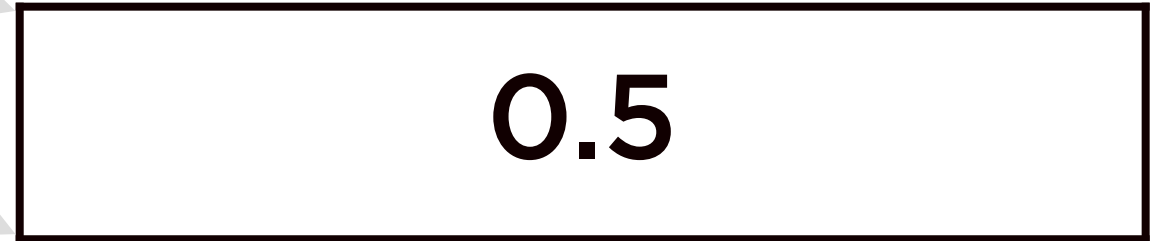
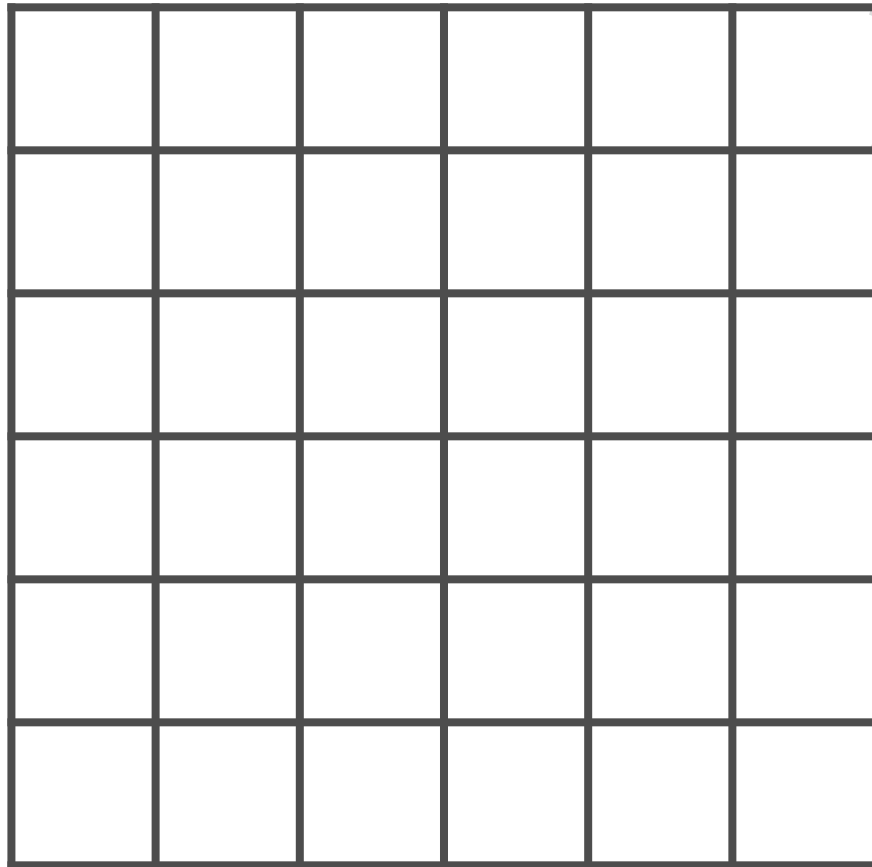


**Each pixel represents
only intensity information**

0.0 - 1.0

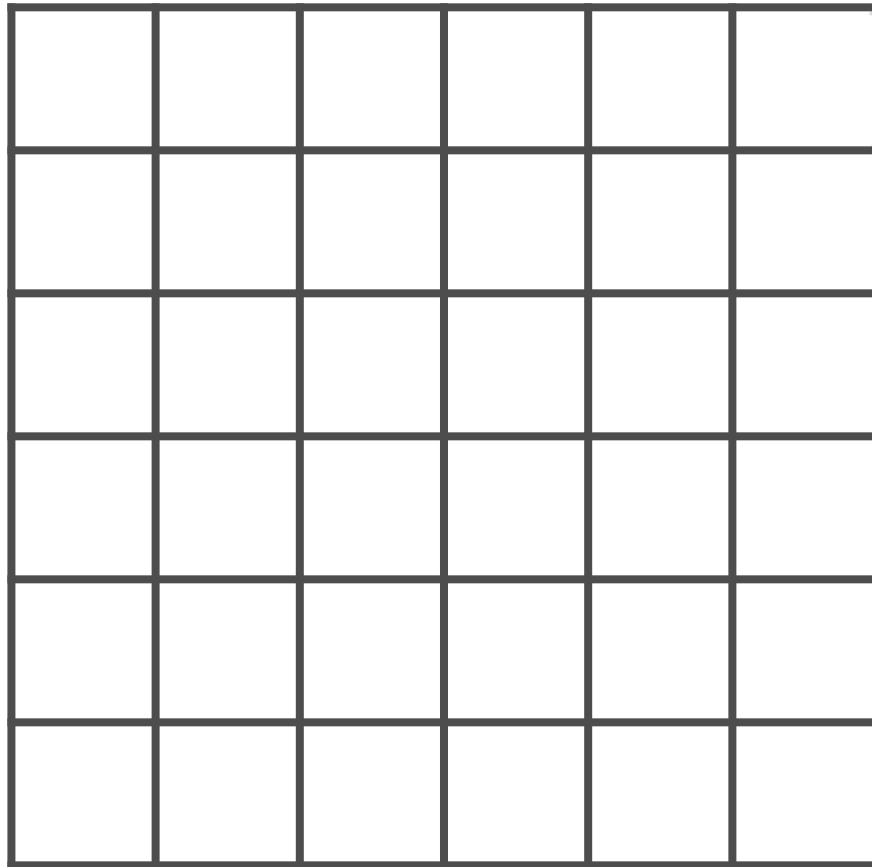


Grayscale Images





Grayscale Images



0.5

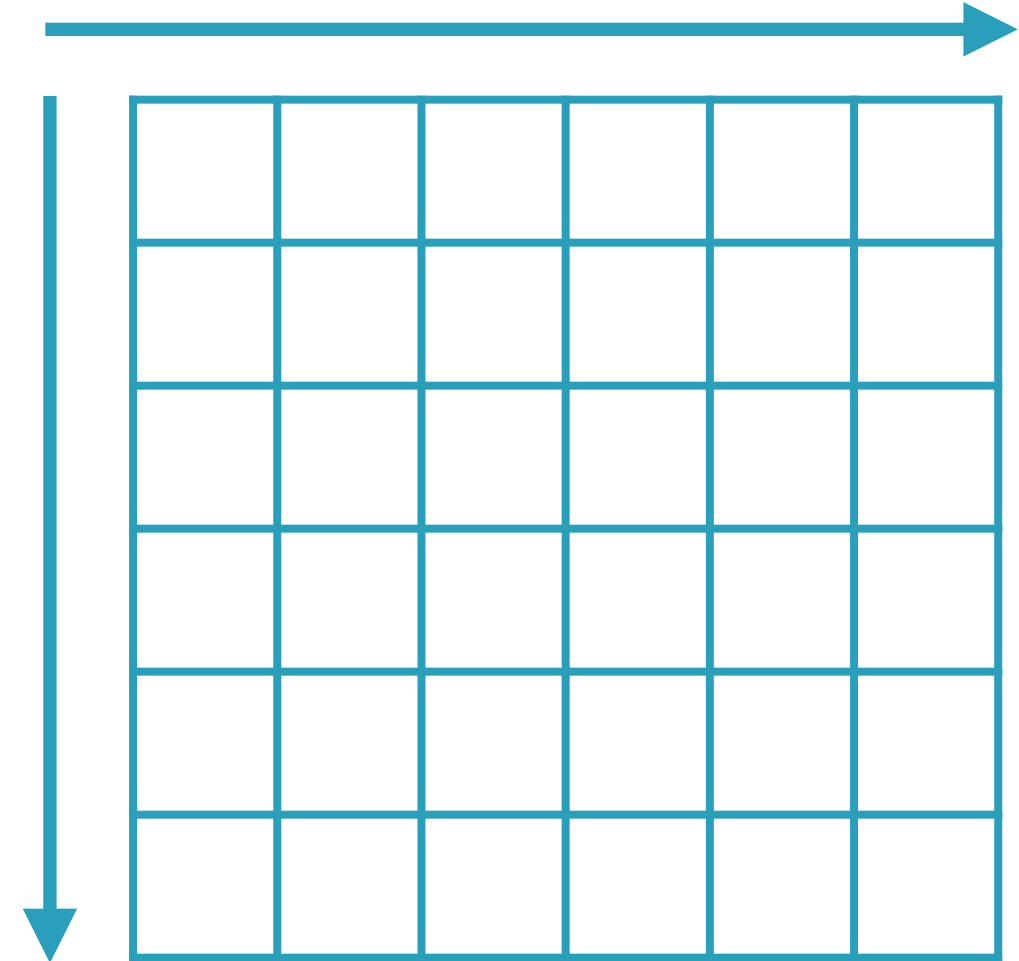
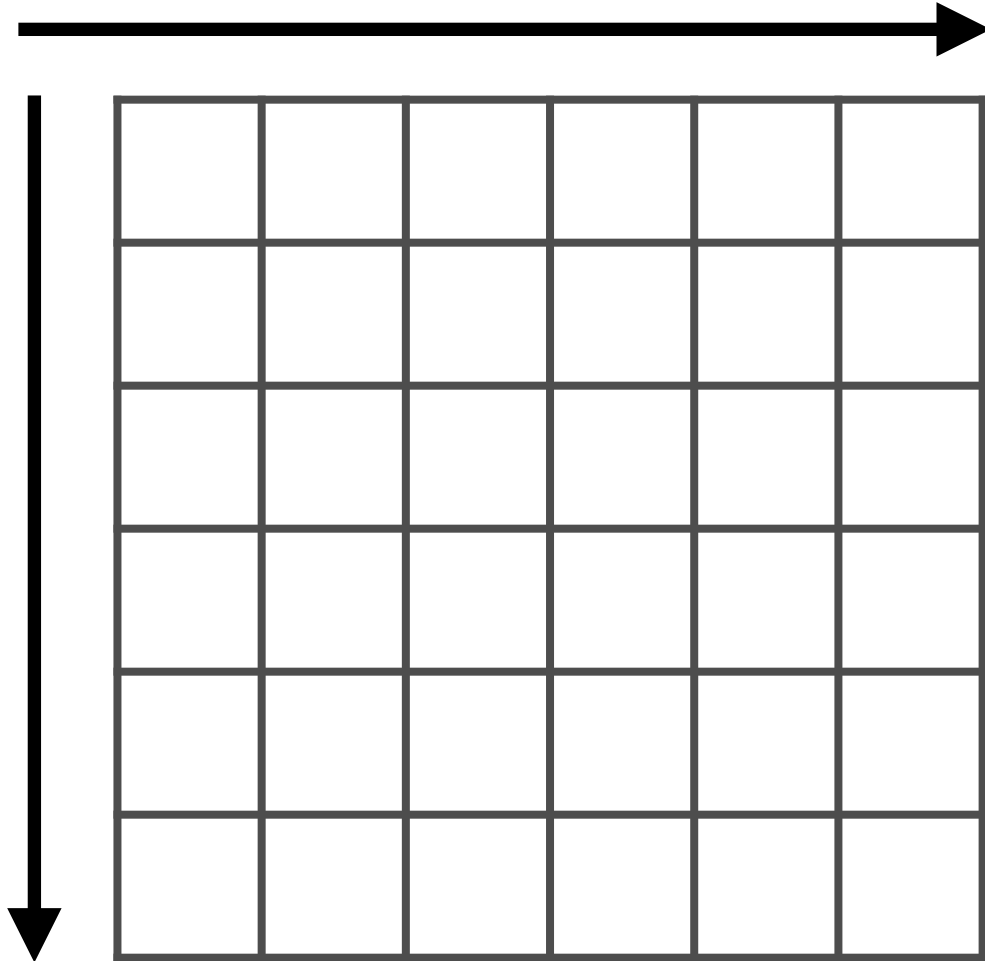
1 value to represent
intensity, **1** channel

Images as Tensors



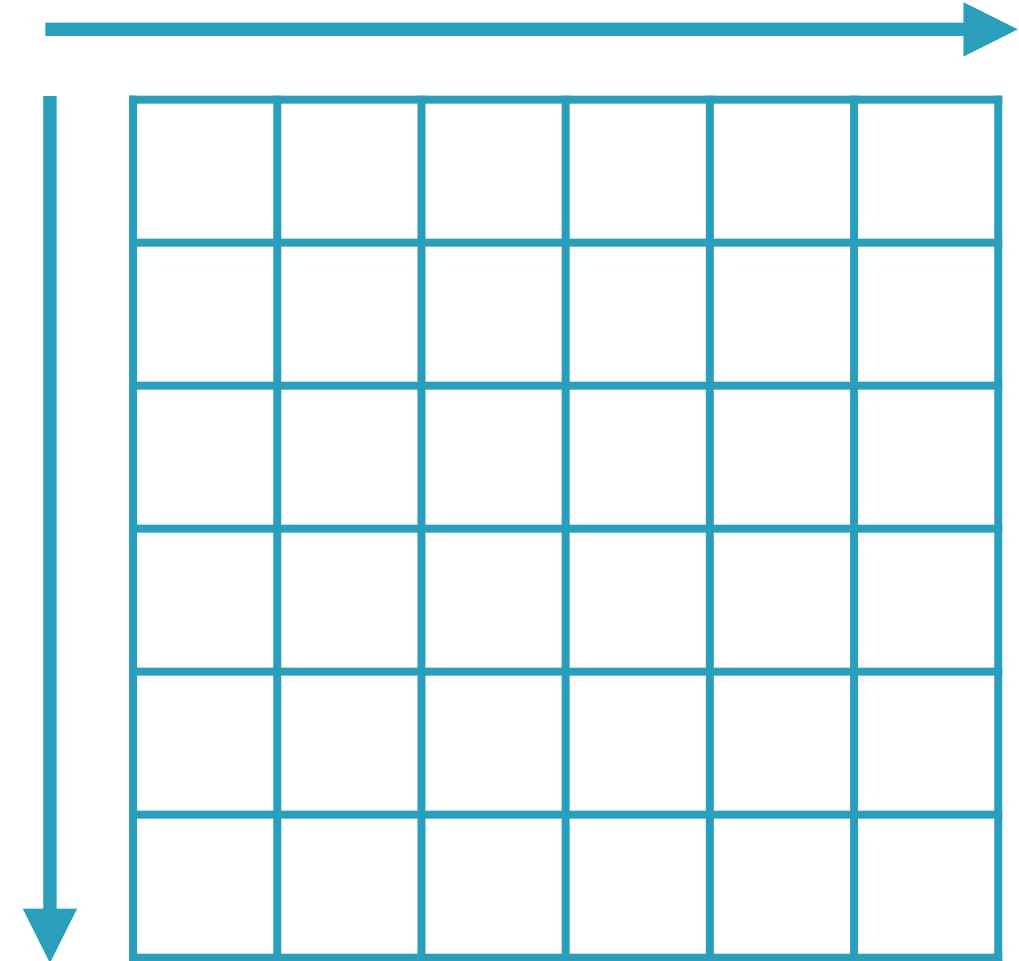
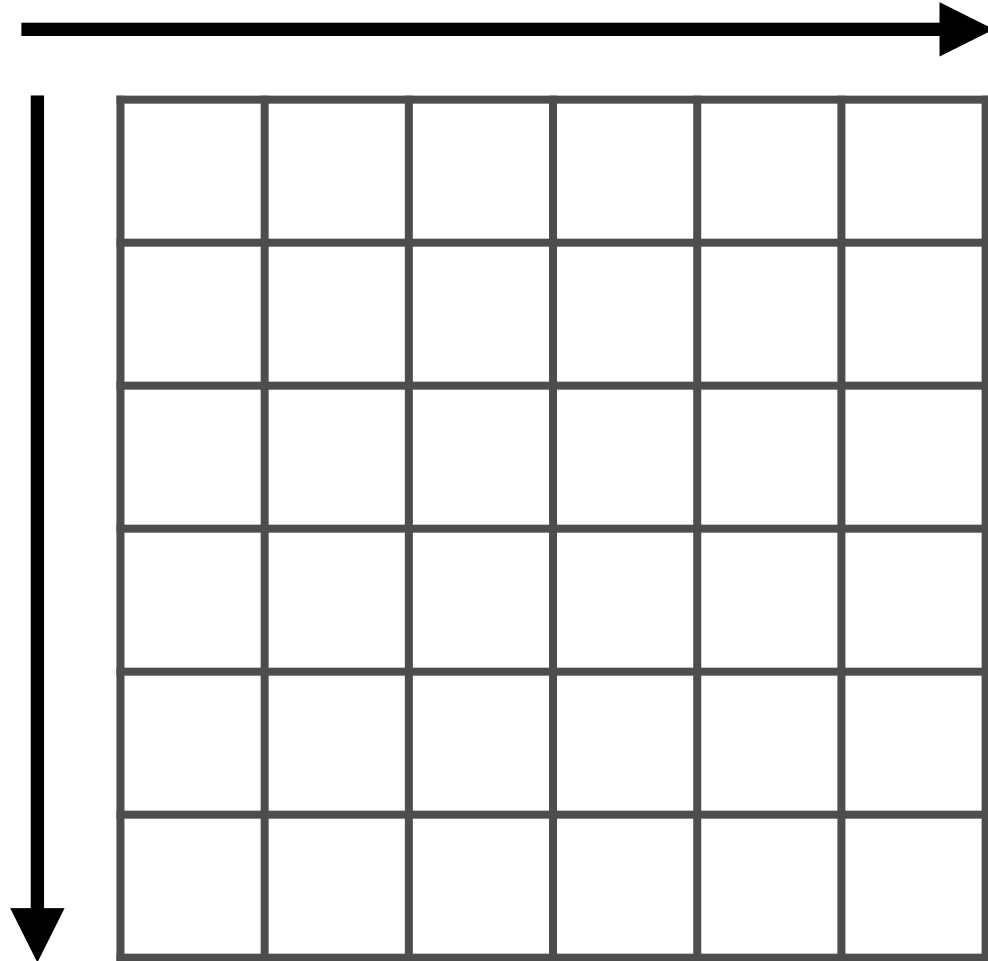
Single channel and multi-channel images

Images as Tensors



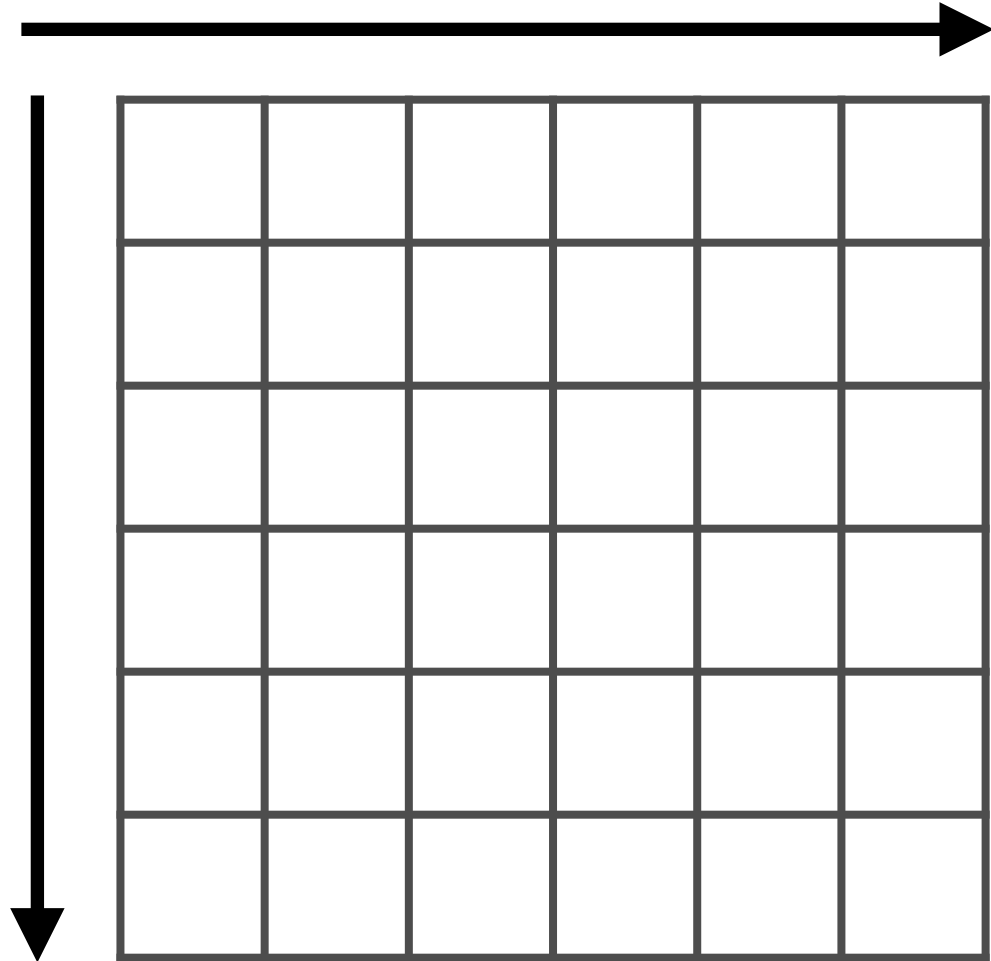
Images can be represented by a 3-D matrix

Images as Tensors

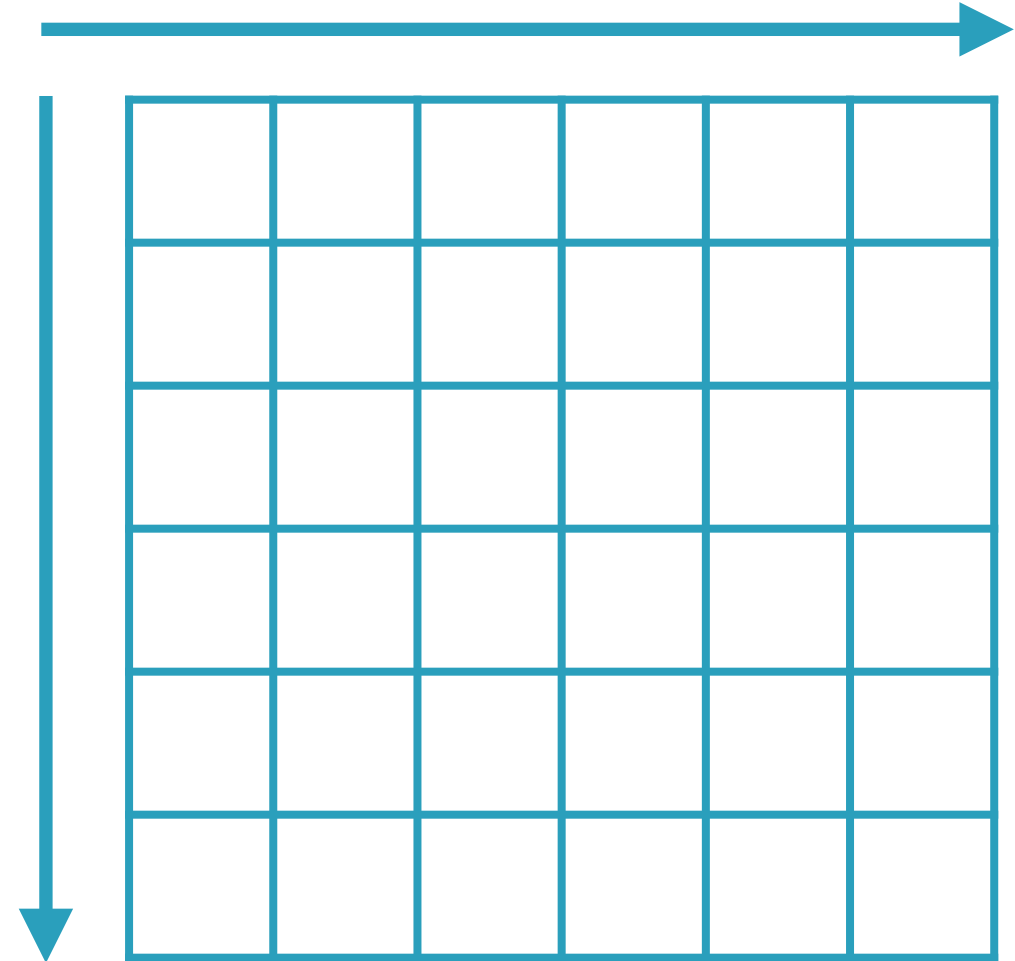


The **number of channels** specifies the **number of elements** in the 3rd dimension

Images as Tensors



(6, 6, 1)



(6, 6, 3)

List of Images



Deep learning frameworks usually deal with a **list of images in one 4-D tensor**

List of Images



The images should all be the same size



List of Images

(10, 6, 6, 3)

The number of channels



List of Images

(10, 6, 6, 3)

The height and width of each image in the list



List of Images

(10, 6, 6, 3)

The number of images

Need for Image Pre-processing

Image Pre-processing Methods

Uniform Aspect
Ratio

Uniform Image Size

Mean and Perturbed
Images

Normalized Image
Inputs

Dimensionality
Reduction

Data Augmentation

Common techniques to improve CNN performance

Image Pre-processing Methods

**Uniform Aspect
Ratio**

Uniform Image Size

Mean and Perturbed
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Dimensionality
Reduction

Data Augmentation

Common techniques to improve CNN performance

Uniform Aspect Ratio

Most models assume square shape

Crop images to be square

Usually, center of image most important

Makes aspect ratio constant

Image Pre-processing Methods

Uniform Aspect
Ratio

Uniform Image Size

Mean and Perturbed
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Common techniques to improve CNN performance

Uniform Image Size

Fit image size to CNN feature maps

250 x 250 image to 100 x 100 image

Down-scaling factor of 0.4

Up-scaling and down-scaling

Image Pre-processing Methods

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Ratio

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Common techniques to improve CNN performance

Mean and Perturbed Images

Mean image: average pixel across entire training dataset

Insights often emerge

E.g. faces usually in center of image

Mean and Perturbed Images

Perturbed image: intentionally distort pixels by varying them from mean image

E.g. to prevent CNN from only focusing on center

Image Pre-processing Methods

Uniform Aspect
Ratio

Uniform Image Size

Mean and Perturbed
Images

**Normalized Image
Inputs**

Dimensionality
Reduction

Data Augmentation

Common techniques to improve CNN performance

Normalized Image Inputs

“Normalize” each pixel

Subtract mean

Divide by standard deviation

**Ensures each pixel has similar data
distribution**

Normalized Image Inputs

Converts pixels to $N(0,1)$ distribution

Then scale to be in $[0,1]$ or $[0,255]$

Helps neural networks converge faster

Image Pre-processing Methods

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Ratio

Uniform Image Size

Mean and Perturbed
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Normalized Image
Inputs

Dimensionality
Reduction

Data Augmentation

Common techniques to improve CNN performance

Dimensionality Reduction

RGB data has 3 channels

Can reduce to grayscale (just 1 channel)

Reduces dimensionality of all image tensors

Reduce the size of the problem so training completes faster

Image Pre-processing Methods

Uniform Aspect
Ratio

Uniform Image Size

Mean and Perturbed
Images

Normalized Image
Inputs

Dimensionality
Reduction

Data Augmentation

Common techniques to improve CNN performance

Data Augmentation

Perturbed images are a form of data augmentation

Scaling, rotation, affine transforms

Makes CNN training more robust

Reduces risk of overfitting

Demo

Set up a deep learning VM on a cloud platform

Demo

Explore common image pre-processing techniques

Demo

**Implement image pre-processing using
PyTorch**

Summary

Image classification using machine learning

Representing images as tensors

Need for image pre-processing

Common image pre-processing techniques