Deep Learning for Visual Computing Motivation, Image Classification

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Topics

Deep Learning

- Motivation
- ► Introduction

Image classification

- Challenges
- Datasets
- Manual approach



Course is called Deep Learning for Visual Computing

► Very generic term (includes computer graphics etc.)

We'll focus on Computer Vision

- Make computers gain high-level understanding of images
- Goal is human-like understanding



Image Classification is a common example

► "What thing is shown in this image?"

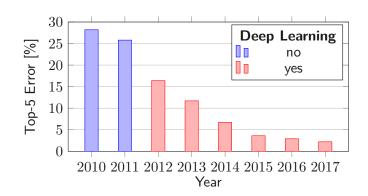






ImageNet benchmark performance over time

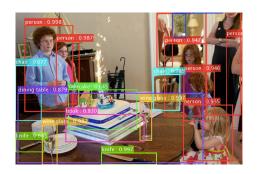
▶ 100k test images of 1000 classes



"Describe the image with keywords"



"Detect notable objects"



"Detect traffic participants"



"Be an artist"





link

"Estimate people's poses"



link

"Generate videos that look real"



link

Other Forms of Image Understanding

"Describe the image with a sentence"



A little girl in a pink shirt is looking at a toy doll.



A woman is riding a bicycle on the pavement.



A girl with a red cap, hair tied up and a gray shirt is fishing in a calm lake.

paper



All these examples are based on Deep Learning

- ▶ Would be impossible otherwise at this quality
- ▶ We will take a closer look throughout the lecture

Deep Learning is state of the art

- In virtually any Computer Vision task
- In other fields as well (e.g. speech recognition)



Deep Learning Introduction

Sounds like magic? Not quite!

- ► Implemented using Neural Networks
- With neurons adapted for image data (convolutional)
- ► And arranged in many layers (hence "deep")

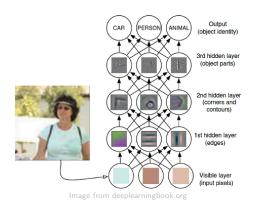
Key ingredients

- Large datasets
- Lots of processing power (GPUs)



Deep Learning Introduction

Networks learn high-level concepts for lower-level ones





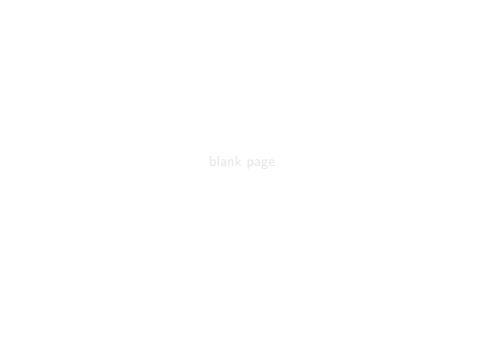
Deep Learning Introduction

The approach is 30+ years old

▶ But data and processing power were limited until recently

Image from Wikipedia





Fundamental Computer Vision task

Definition

- ► Given a set of class labels (e.g. {bird, cat, dog})
- ▶ Which class does the given image belong to?



Image from youtube.com

Image belongs to exactly one class in the set

- Comparatively easy task (but still challenging)
- On some datasets Deep Learning outperforms humans

Simple problem formulation

- Great for learning Deep Learning basics
- ▶ Why we will stick to classification for now



Image Classification Challenges – Pose and Viewpoint



Image adapted from warrenphotographic.co.uk

Image Classification Challenges – Illumination



Image from studioddt.com

Image Classification Challenges – Deformation









Image from cs231n.github.io

Image Classification Challenges – Occlusion







Image from cs231n.github.io

Image Classification Challenges – Background



Image from cs231n.github.io

Challenges – Intraclass Variation



Image from cs231n.github.io



A good classifier must cope with these challenges

- ▶ To verify this we need a representative dataset
- Such datasets are usually large

If we employ Machine Learning we also need training data

- Datasets must be disjoint (so need even more data)
- Deep Learning requires lots of data



Dataset acquisition takes lots of effort

- ► Collect many (thousands or more) of images
- Assign class labels to enable automatic training and testing

Data acquisition and processing is central in Deep Learning

- ▶ Often the most time-consuming task
- Usually main bottleneck for performance

Thankfully many public datasets are available



Image Classification Datasets – CIFAR-10

10 classes, 60k images



Image from cs.toronto.edu



Image Classification Datasets – ImageNet

20k classes, 14m images

▶ Main driver for Deep Learning performance



Image from umich.edu



Image Classification Datasets – COCO

300k images, labels for classification, detection, segmentation, \dots



Image from cocodataset.org

Let's build an image classifier

- ► Should support the classes {dog, cat}
- ▶ Using the CIFAR-10 dataset



Image from cs.toronto.edu

How can we write an algorithm for this purpose?



Image from cs.toronto.edu

We cannot!

- ► No obvious unique and reliable features
- ▶ Not clear how to represent and use them



Image from cs.toronto.edu

We humans are incredible image classifiers

But we cannot describe formally how we do so

► Thus the standard if {} else {} approach fails

This applies to most vision problems

► Reason we need Machine and Deep Learning