

The background of the slide features a complex network of glowing blue lines and dots, resembling a neural network or a data connection map, set against a dark blue gradient.

Practical Deep Learning: A quick glance

Image classification and object detection

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Image classification

Convolution layers are the eyes of a CNN



Input

Image classification

Architecture of CNNs:

- Feature extraction
- Dimension reduction

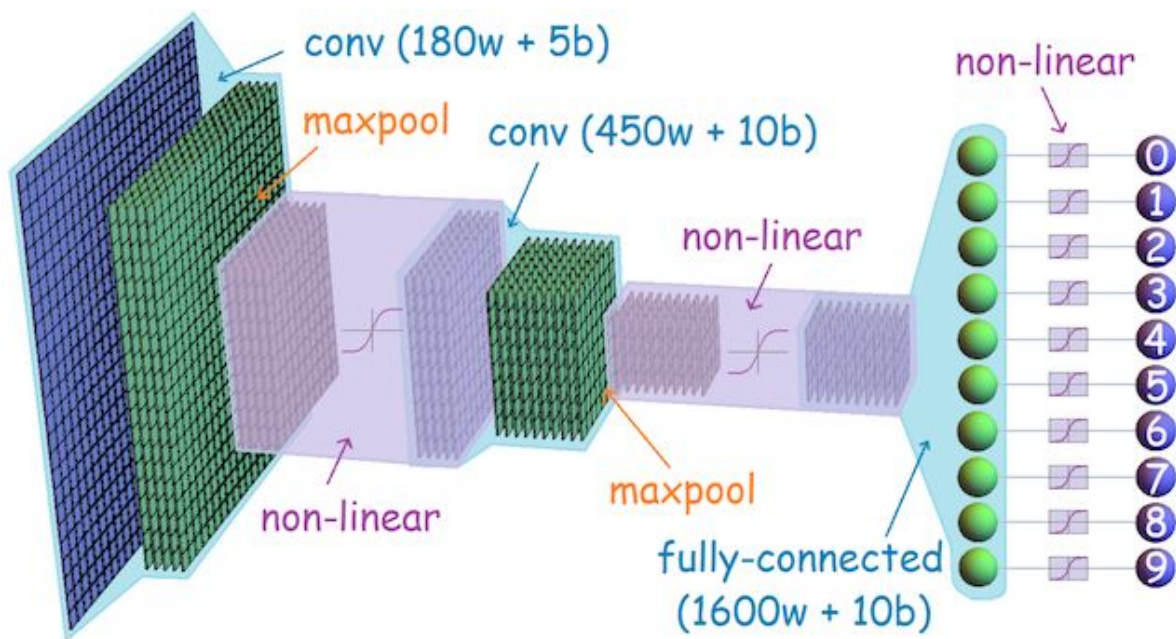


Image classification

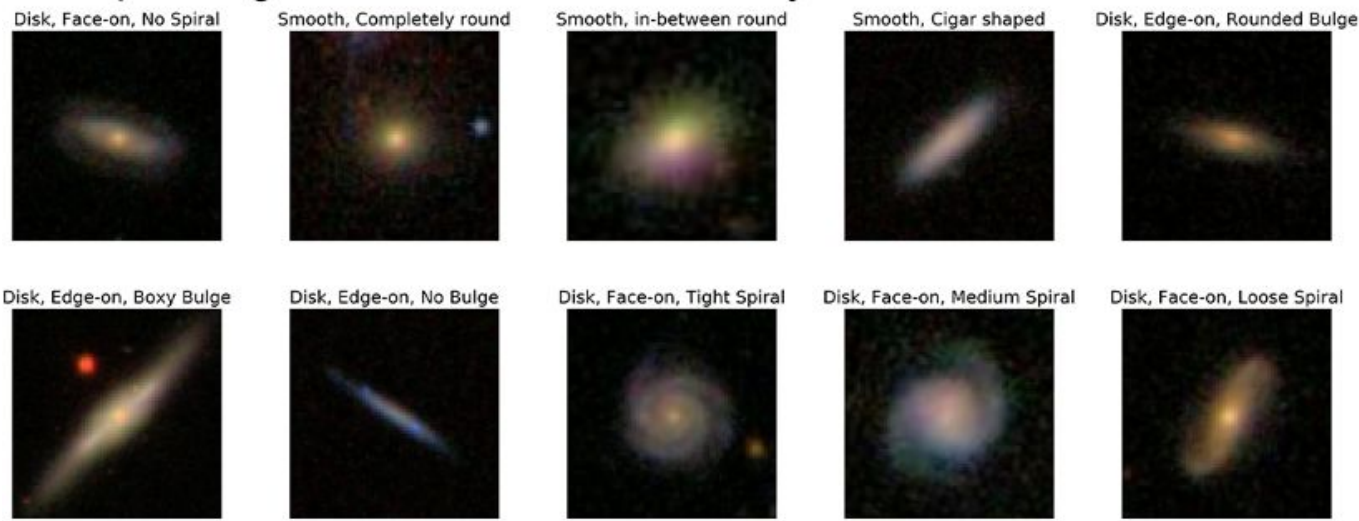
Galaxy10 toy dataset

([astroNN web](#))

Galaxy10 dataset (21785 images)

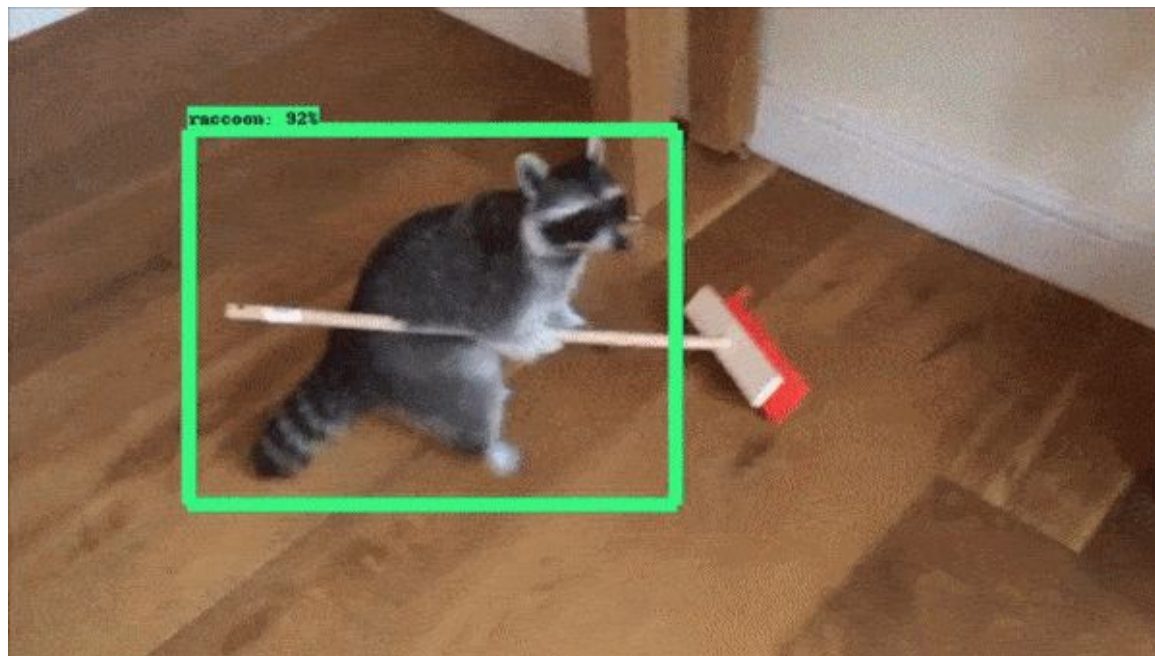
- Class 0 (3461 images): Disk, Face-on, No Spiral
- Class 1 (6997 images): Smooth, Completely round
- Class 2 (6292 images): Smooth, in-between round
- Class 3 (394 images): Smooth, Cigar shaped
- Class 4 (1534 images): Disk, Edge-on, Rounded Bulge
- Class 5 (17 images): Disk, Edge-on, Boxy Bulge
- Class 6 (589 images): Disk, Edge-on, No Bulge
- Class 7 (1121 images): Disk, Face-on, Tight Spiral
- Class 8 (906 images): Disk, Face-on, Medium Spiral
- Class 9 (519 images): Disk, Face-on, Loose Spiral

Example images of each class from Galaxy10 dataset



Object detection

Provide the object (class) and location (bounding box/region)



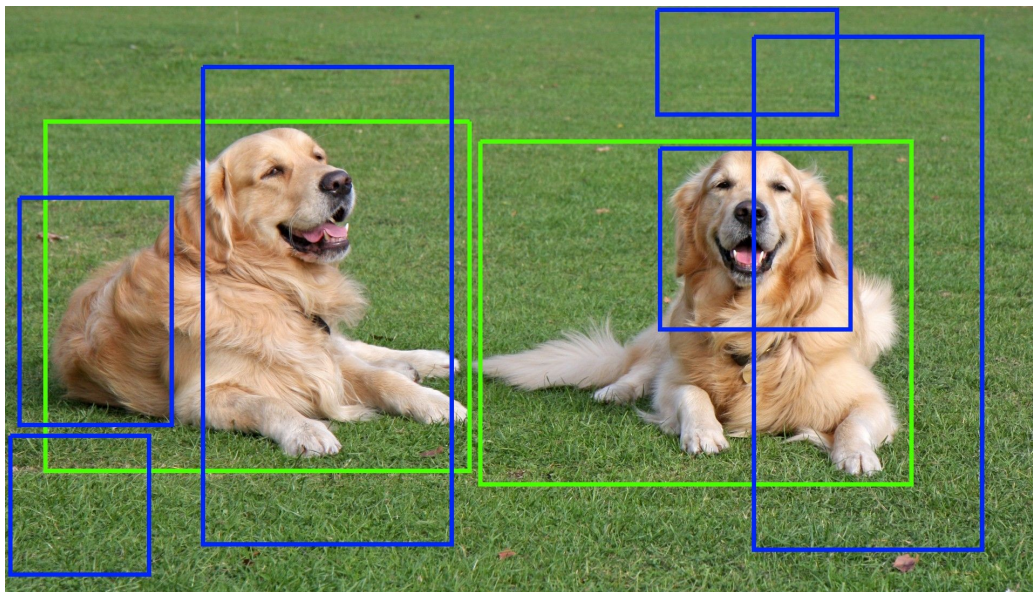
Object detection

Object detection models:

Simultaneously learns

- Region proposal
- Object classification

Output a set of candidates
regions-class and confidence

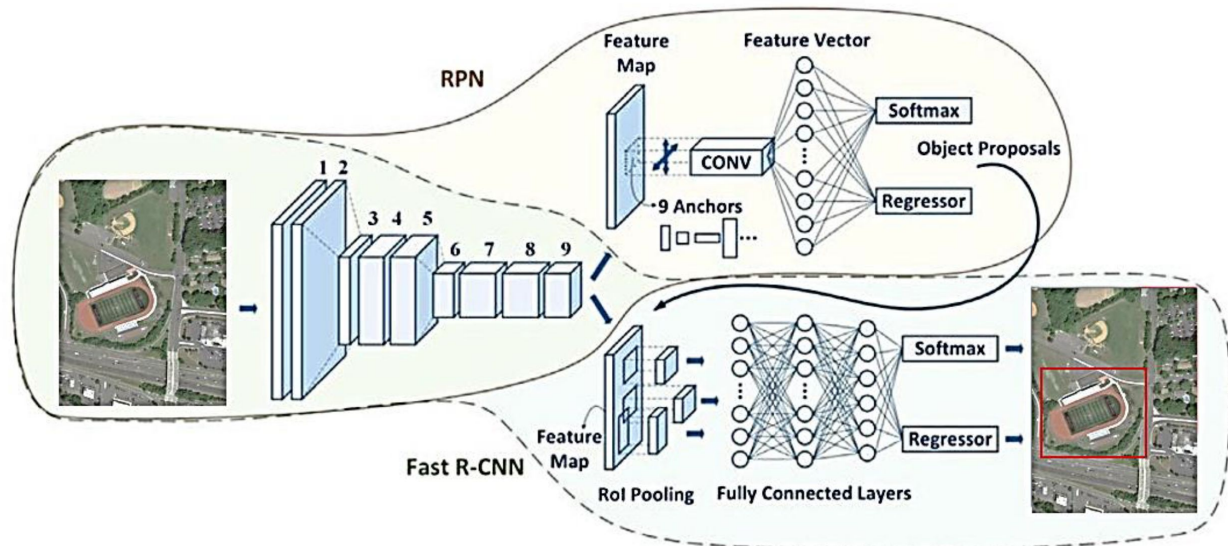


Object detection

Faster R-CNN architecture:

Region proposal based on
a Region Proposal Network

- Slower
- More accurate
- Small objects



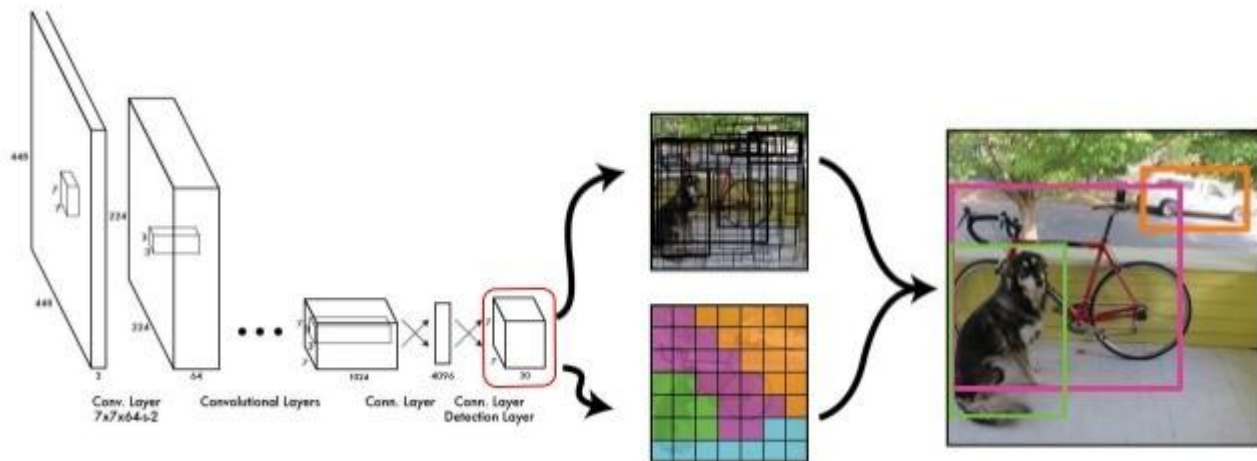
Object detection

SSD or YOLO

Region proposal based
on grid

- Faster
- Less accurate
- Common objects

YOLO: You Only Look Once



Object detection

Detect fiery looping rain on the Sun

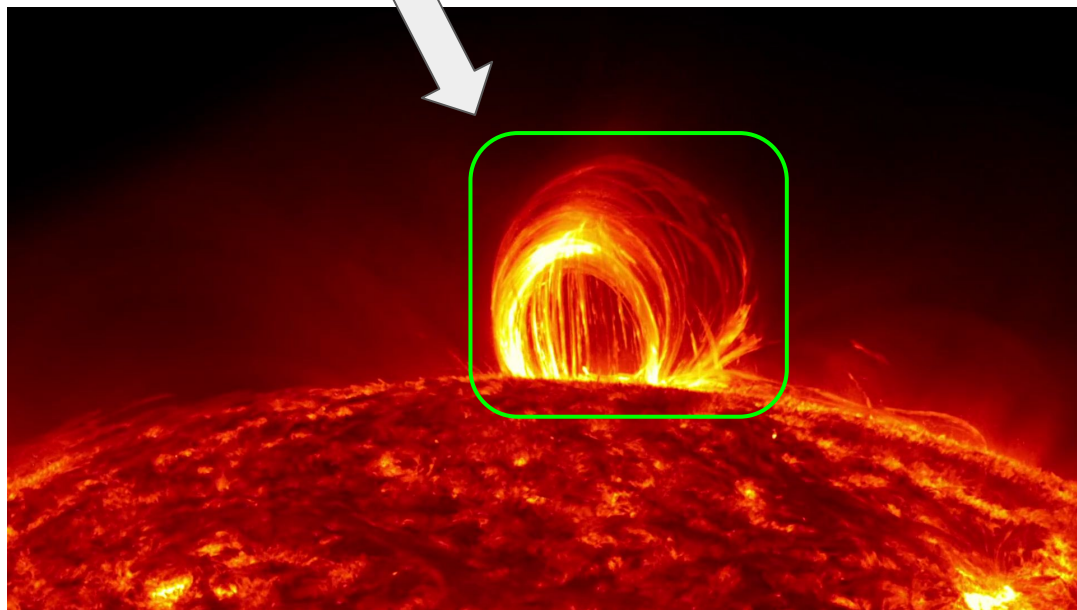
[Video](#)

A recurrent pattern in a structured signal can be learned

- RGB images
- Texture, color, shape

(X, Y, W, H)

{
Location
Color properties
Dimension
Movement
...



Object detection

Detection annotation

- Image information
- Objects
 - Class
 - Region

```
1 <annotation>
2   <folder>train</folder>
3   <filename>FieryLoopingRainSun-187.jpg</filename>
4   <path>/home/alcala/Escritorio/Astro/Solar-FieryLoop/train/FieryLoopingRainSun-187.jpg</path>
5   <source>
6     <database>Unknown</database>
7   </source>
8   <size>
9     <width>1280</width>
10    <height>720</height>
11    <depth>3</depth>
12  </size>
13  <segmented>0</segmented>
14  <object>
15    <name>fiery_loopw</name>
16    <pose>Unspecified</pose>
17    <truncated>0</truncated>
18    <difficult>0</difficult>
19    <bndbox>
20      <xmin>699</xmin>
21      <ymin>376</ymin>
22      <xmax>820</xmax>
23      <ymax>471</ymax>
24    </bndbox>
25  </object>
26 </annotation>
27
```

Object detection

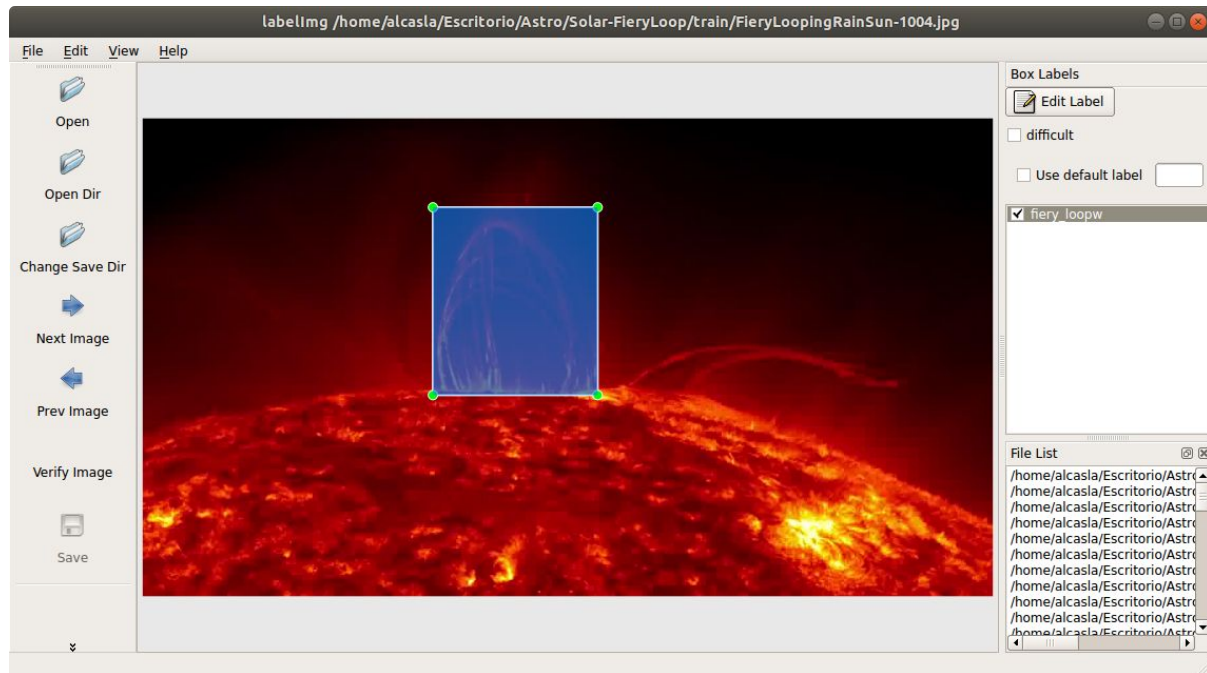
LabelImg

Repository

github.com/tzutalin/labelImg

Installer file

tzutalin.github.io/labelImg/



Object detection

Object detection framework

github.com/tensorflow/models/tree/master/research/object_detection

Software configuration

github.com/spsrc/somachine2020

Build conda environment using .yaml file

[API configuration](#)

Model Zoo

We provide a large collection of models that are trained on several datasets in the [Model Zoo](#).

Guides

- [Configuring an object detection pipeline](#)
- [Preparing inputs](#)
- [Defining your own model architecture](#)
- [Bringing in your own dataset](#)
- [Supported object detection evaluation protocols](#)
- [TPU compatible detection pipelines](#)
- [Training and evaluation guide \(CPU, GPU, or TPU\)](#)

Extras:

- [Exporting a trained model for inference](#)
- [Exporting a trained model for TPU inference](#)
- [Inference and evaluation on the Open Images dataset](#)
- [Run an instance segmentation model](#)
- [Run the evaluation for the Open Images Challenge 2018/2019](#)
- [Running object detection on mobile devices with TensorFlow Lite](#)
- [Context R-CNN documentation for data preparation, training, and export](#)