



OpenCV(Python) Deeplearning4j(Java)

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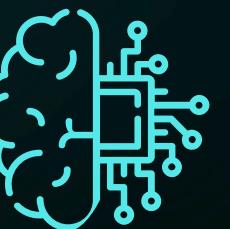
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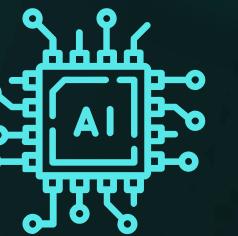
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OpenCV Introduction



1. was started at Intel in 1999 and the first release came out in 2000
2. now supports a multitude of algorithms related to Computer Vision and Machine Learning and is expanding day by day
3. supports a wide variety of programming languages such as C++, Python, Java, etc.
4. is available on different platforms including Windows, Linux, OS X, Android, and iOS.
5. Interfaces for high-speed GPU operations based on CUDA and OpenCL are also under active development
6. OpenCV-Python is the Python API for OpenCV, combining the best qualities of the OpenCV C++ API and the Python language

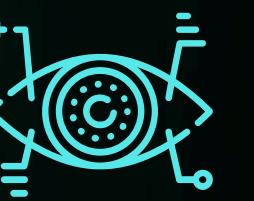
OpenCV-Python



1. is a library of Python bindings designed to solve computer vision problems
2. allows you to perform various tasks related to image and video processing, such as loading and saving images, applying various filters and transformations, detecting and recognizing objects, image segmentation, and much more.
3. It provides a wide range of functions and algorithms to handle images and videos efficiently
4. All the OpenCV array structures are converted to from Numpy arrays

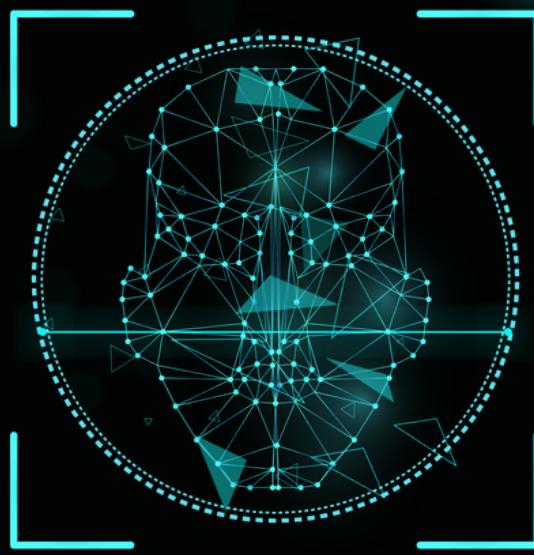


OpenCV-Python

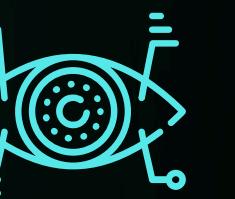


- Image and Video I/O allows you to read and write images and videos in various file formats. You can use functions like `cv2.imread()` to load an image, `cv2.imwrite()` to save an image, `cv2.VideoCapture()` to capture video from a camera or a file, and `cv2.VideoWriter()` to save video output.

- Image Manipulation provides functions for resizing, cropping, rotating, and flipping images. You can change the image properties, such as brightness, contrast, and saturation, using functions like `cv2.resize()`, `cv2.flip()`, `cv2.rotate()`, and others



OpenCV-Python

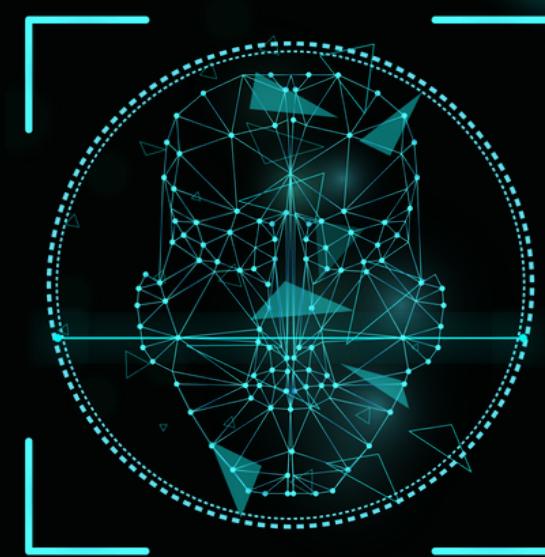


- Image Filtering

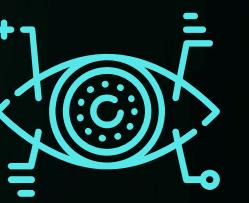
You can apply various filters and enhancements to images using functions like `cv2.filter2D()`, `cv2.GaussianBlur()`, `cv2.medianBlur()`, and `cv2.equalizeHist()`. These functions allow you to perform operations like smoothing, sharpening, noise reduction, and histogram equalization

- Image Segmentation

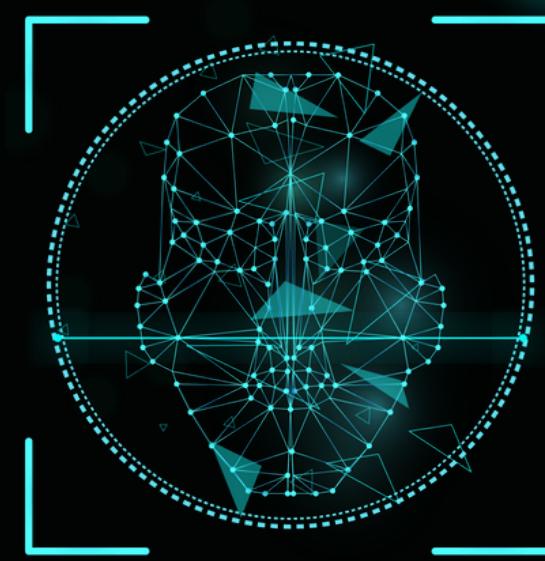
provides functions for image segmentation, which involves dividing an image into meaningful regions or objects. Functions like `cv2.findContours()`, `cv2.drawContours()`, and `cv2.watershed()` can be used for contour detection, region filling, and watershed segmentation



OpenCV-Python



- Object Detection includes pre-trained models and functions for object detection and recognition. The most commonly used method is the Haar cascades classifier, which can be used to detect objects like faces, eyes, and smiles. OpenCV-Python also supports deep learning-based object detection frameworks like TensorFlow and PyTorch.
- Feature Detection offers various algorithms for detecting and describing image features, such as corners, edges, and keypoints. Common functions include `cv2.goodFeaturesToTrack()`, `cv2.Canny()`, `cv2.cornerHarris()`, `cv2.ORB_create()`



Canny Edge Detection in OpenCV



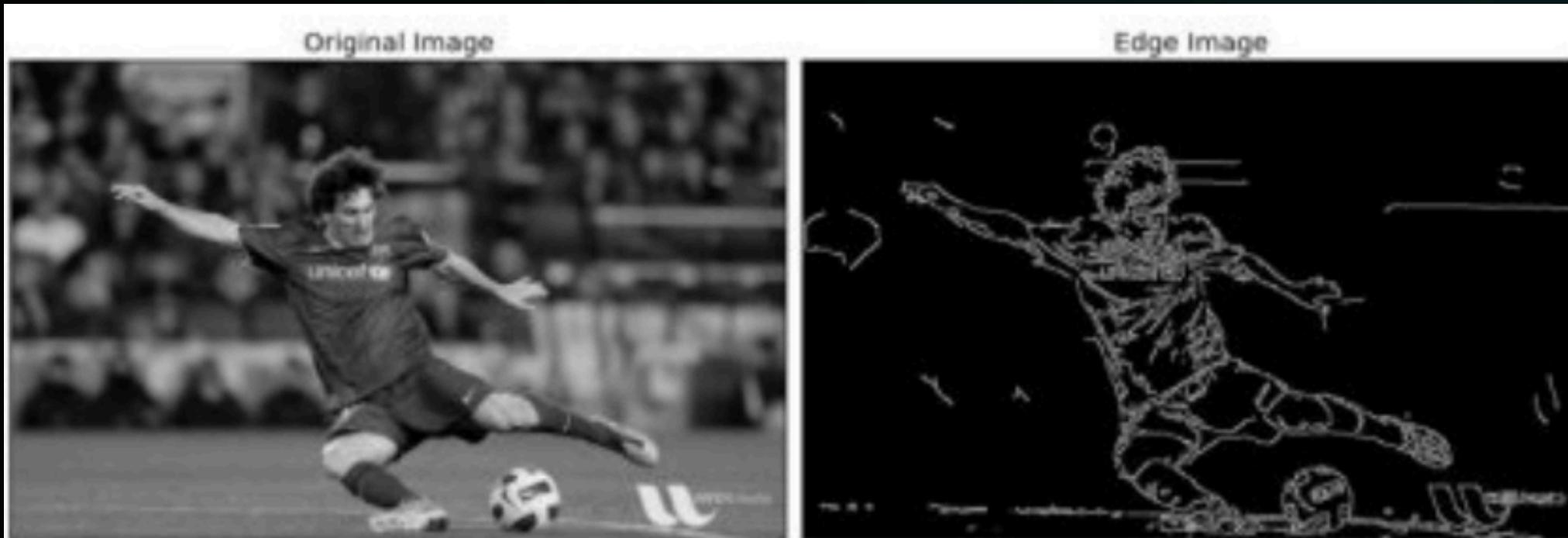
```
import numpy as np  
import cv2 as cv  
from matplotlib import pyplot as plt
```

```
img = cv.imread('messi5.jpg',  
cv.IMREAD_GRAYSCALE)  
assert img is not None
```

```
edges = cv.Canny(img,100,200)
```

```
plt.subplot(121),plt.imshow(img,cmap = 'gray')  
plt.title('Original Image'), plt.xticks([ ]), plt.yticks([ ])  
plt.subplot(122),plt.imshow(edges,cmap = 'gray')  
plt.title('Edge Image'), plt.xticks([ ]), plt.yticks([ ])
```

```
plt.show()
```



Deeplearning4j Introduction



1. Eclipse Deeplearning4j is a suite of tools for running deep learning on the JVM
2. is open-source released under Apache License
3. It is powered by its own open-source numerical computing library, and works with both CPUs and GPUs
4. includes implementations of the restricted Boltzmann machine(RBM), deep belief net, deep autoencoder, recursive neural tensor network, word2vec, doc2vec. These algorithms all include distributed parallel versions that integrate with Apache Hadoop and Spark



Deeplearning4j Introduction



1. includes a vector space modeling and topic modeling toolkit, implemented in Java and integrating with parallel GPUs for performance. It is designed to handle large text sets
2. includes implementations of term frequency-inverse document frequency (tf-idf), deep learning, Mikolov's word2vec algorithm, doc2vec, reimplemented and optimized in Java
3. Tensorflow, Keras and Deeplearning4j work together.
Deeplearning4j can import models from Tensorflow and other Python frameworks if they have been created with Keras



Installation requirements

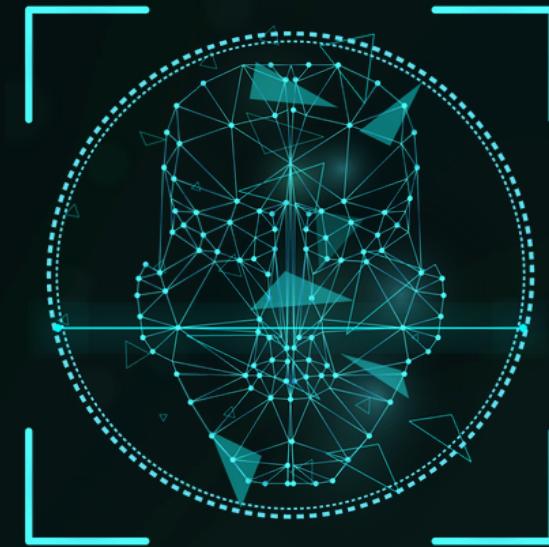


Java (developer version) 11 or later

Apache Maven 3.3.x (automated build and dependency manager)

IntelliJ IDEA or Eclipse

Neural network configuration in Deeplearning4j



```
MultiLayerConfiguration config = new NeuralNetConfiguration.Builder()  
    .optimizationAlgo(OptimizationAlgorithm.STOCHASTIC_GRADIENT_DESCENT)  
    .iterations(1000)  
    .learningRate(0.1)  
    .list()  
    .layer(0, new DenseLayer.Builder().nIn(2).nOut(4).activation(Activation.RELU).build())  
    .layer(1, new OutputLayer.Builder(LossFunctions.LossFunction.MSE)  
        .nIn(4).nOut(1).activation(Activation.SIGMOID).build())  
    .pretrain(false).backprop(true)  
    .build();
```



Thanks for
your
attention! 😊

