

# OpenCV deep learning module samples

## Model Zoo

Check [a wiki](#) for a list of tested models.

If OpenCV is built with [Intel's Inference Engine support](#) you can use [Intel's pre-trained](#) models.

There are different preprocessing parameters such mean subtraction or scale factors for different models. You may check the most popular models and their parameters at [models.yml](#) configuration file. It might be also used for aliasing samples parameters. In example,

```
python object_detection.py opencv_fd --model /path/to/caffemodel --config
/path/to/prototxt
```

Check `-h` option to know which values are used by default:

```
python object_detection.py opencv_fd -h
```

## Sample models

You can download sample models using `download_models.py`. For example, the following command will download network weights for OpenCV Face Detector model and store them in FaceDetector folder:

```
python download_models.py --save_dir FaceDetector opencv_fd
```

You can use default configuration files adopted for OpenCV from [here](#).

You also can use the script to download necessary files from your code. Assume you have the following code inside `your_script.py`:

```
from download_models import downloadFile

filepath1 = downloadFile("https://drive.google.com/uc?
export=download&id=0B3gersZ2cHIXRm5PMWRoTkDhdHc", None,
filename="MobileNetSSD_deploy.caffemodel", save_dir="save_dir_1")
filepath2 = downloadFile("https://drive.google.com/uc?
export=download&id=0B3gersZ2cHIXRm5PMWRoTkDhdHc",
"994d30a8afaa9e754d17d2373b2d62a7dfbaaf7a",
filename="MobileNetSSD_deploy.caffemodel")
print(filepath1)
print(filepath2)
# Your code
```

By running the following commands, you will get **MobileNetSSD\_deploy.caffemodel** file:

```
export OPENCV_DOWNLOAD_DATA_PATH=download_folder
python your_script.py
```

**Note** that you can provide a directory using **save\_dir** parameter or via **OPENCV\_SAVE\_DIR** environment variable.

### Face detection

[An origin model](#) with single precision floating point weights has been quantized using [TensorFlow framework](#). To achieve the best accuracy run the model on BGR images resized to `300x300` applying mean subtraction of values `(104, 177, 123)` for each blue, green and red channels correspondingly.

The following are accuracy metrics obtained using [COCO object detection evaluation tool](#) on [Fddb dataset](#) (see [script](#)) applying resize to `300x300` and keeping an origin images' sizes.

AP - Average Precision				FP32/FP16	UINT8	
FP32/FP16   UINT8						
AR - Average Recall				300x300	300x300	any
size	any size					
----- ----- ----- -----						
----- ----- ----- -----						
AP @[ IoU=0.50:0.95   area=	all   maxDets=100 ]		0.408		0.408	0.378
0.328 (-0.050)						
AP @[ IoU=0.50   area=	all   maxDets=100 ]		0.849		0.849	0.797
0.790 (-0.007)						
AP @[ IoU=0.75   area=	all   maxDets=100 ]		0.251		0.251	0.208
0.140 (-0.068)						
AP @[ IoU=0.50:0.95   area=	small   maxDets=100 ]		0.050		0.051 (+0.001)	0.107
0.070 (-0.037)						
AP @[ IoU=0.50:0.95   area=	medium   maxDets=100 ]		0.381		0.379 (-0.002)	0.380
0.368 (-0.012)						
AP @[ IoU=0.50:0.95   area=	large   maxDets=100 ]		0.455		0.455	0.412
0.337 (-0.075)						
AR @[ IoU=0.50:0.95   area=	all   maxDets= 1 ]		0.299		0.299	0.279
0.246 (-0.033)						
AR @[ IoU=0.50:0.95   area=	all   maxDets= 10 ]		0.482		0.482	0.476
0.436 (-0.040)						
AR @[ IoU=0.50:0.95   area=	all   maxDets=100 ]		0.496		0.496	0.491
0.451 (-0.040)						
AR @[ IoU=0.50:0.95   area=	small   maxDets=100 ]		0.189		0.193 (+0.004)	0.284
0.232 (-0.052)						
AR @[ IoU=0.50:0.95   area=	medium   maxDets=100 ]		0.481		0.480 (-0.001)	0.470
0.458 (-0.012)						
AR @[ IoU=0.50:0.95   area=	large   maxDets=100 ]		0.528		0.528	0.520
0.462 (-0.058)						

## References

- [Models downloading script](#)
- [Configuration files adopted for OpenCV](#)
- [How to import models from TensorFlow Object Detection API](#)
- [Names of classes from different datasets](#)