



# NVIDIA TRANSFER LEARNING TOOLKIT

NVIDIA 开发者社区 何琨

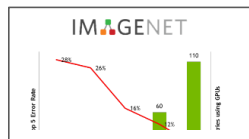
# POWERING THE DEEP LEARNING ECOSYSTEM

## COMPUTER VISION

OBJECT DETECTION



IMAGE CLASSIFICATION



## SPEECH & AUDIO

VOICE RECOGNITION

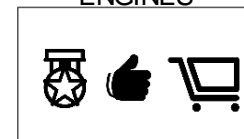


LANGUAGE TRANSLATION



## NATURAL LANGUAGE PROCESSING

RECOMMENDATION ENGINES



SENTIMENT ANALYSIS



## DEEP LEARNING FRAMEWORKS

Caffe



DL4J  
Deeplearning4j

Mocha.jl



K  
KERAS

MatConvNet

Microsoft  
CNTK

MINERVA

mxnet

OpenDeep



Pylearn2

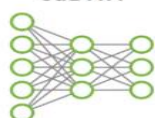


theano



## NVIDIA DEEP LEARNING SDK

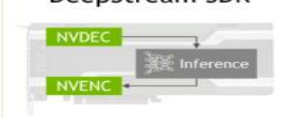
cuDNN



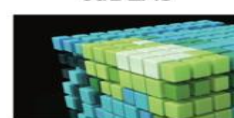
TensorRT



DeepStream SDK



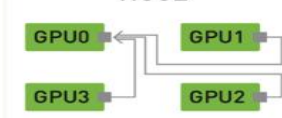
cuBLAS



cuSPARSE



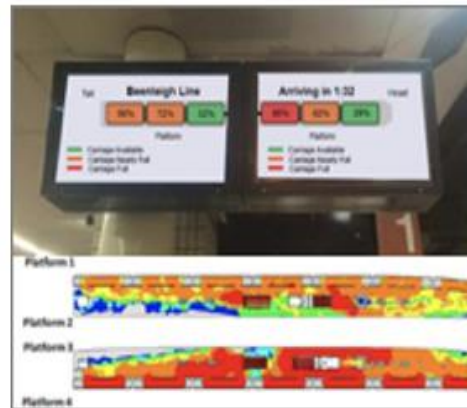
NCCL



# CV IN DEEPLARNING



Access Control



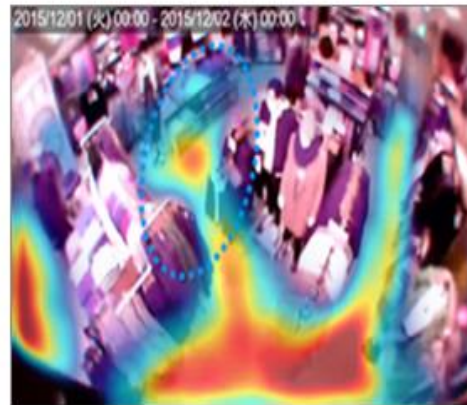
Public Transit



Parking Management



Traffic Engineering



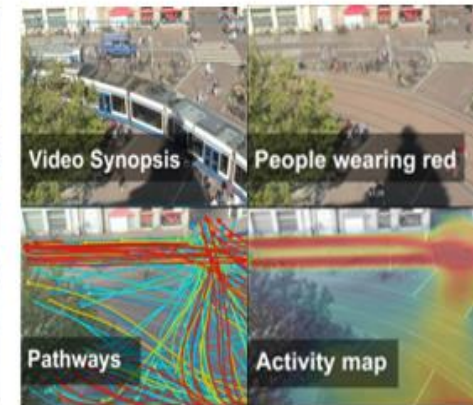
Retail Analytics



Securing Critical Infrastructure



Managing Logistics



Forensic Analysis

# Deep Learning Workflow Management

## Deep Learning Challenges

### Third Party Pre-Trained Models

- Lack accuracy
- Use case limitations
- Model size limitations
- Unoptimized for GPUs

### Deep Learning Training

- Compute Resources
- Time spent training from scratch
- Learning DL frameworks

### Deep Learning Inference

- Unclear workflows for production ready models
- Complex application pipeline

## NVIDIA Deep Learning Solution

### Transfer Learning Toolkit

- GPU accelerated pre trained models
- Incremental Training
- Pruning
- Easy to use
- Abstraction from learning DL frameworks

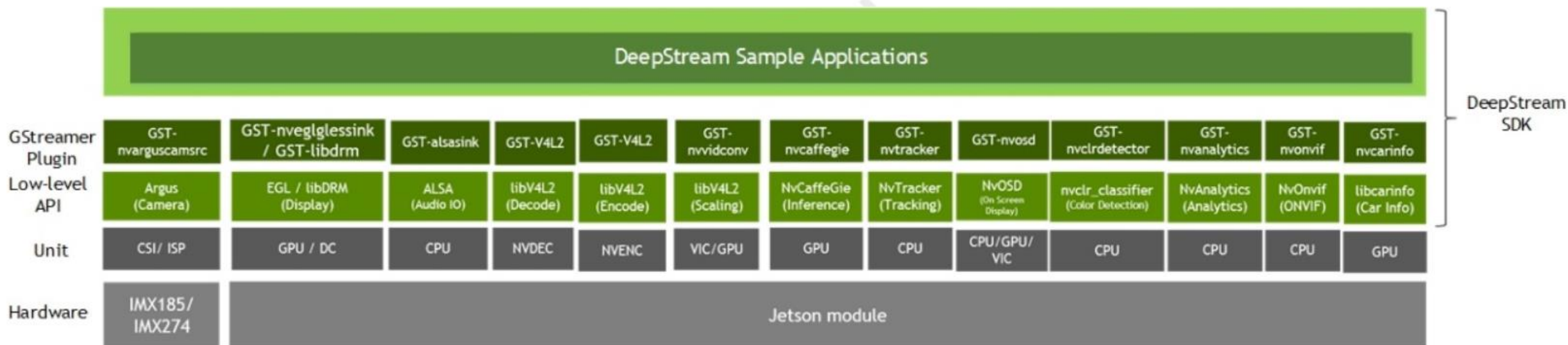
### DeepStream SDK

- Faster intelligent insights
- Track inference
- End to end- easy AI deployment

# DeepStream

NVIDIA DeepStream SDK是为大规模创建和部署基于AI的视频分析应用程序解决方案而设计的，它提供完整的框架和所有基本构建模块。

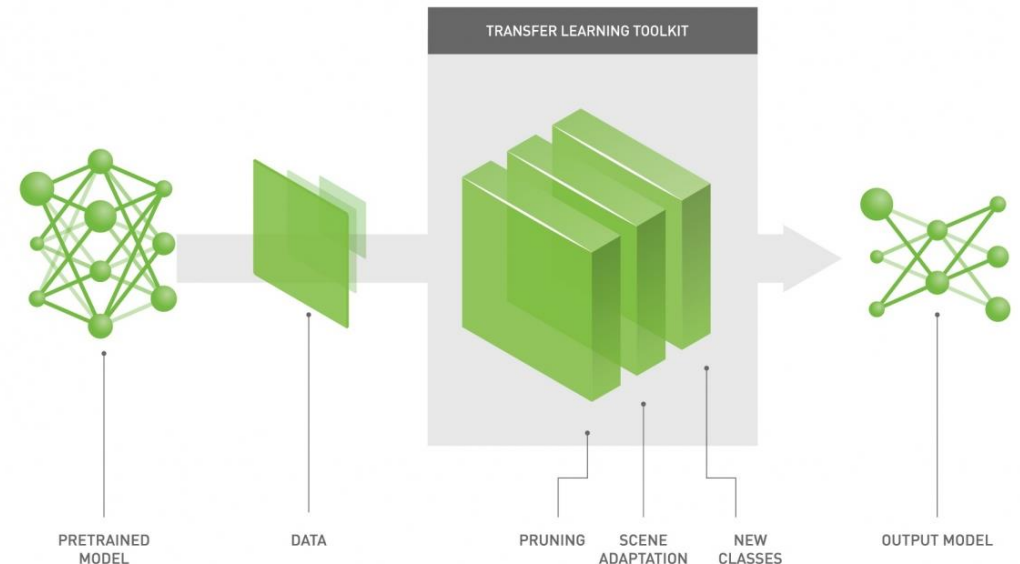
- Gstreamer
- TensorRT



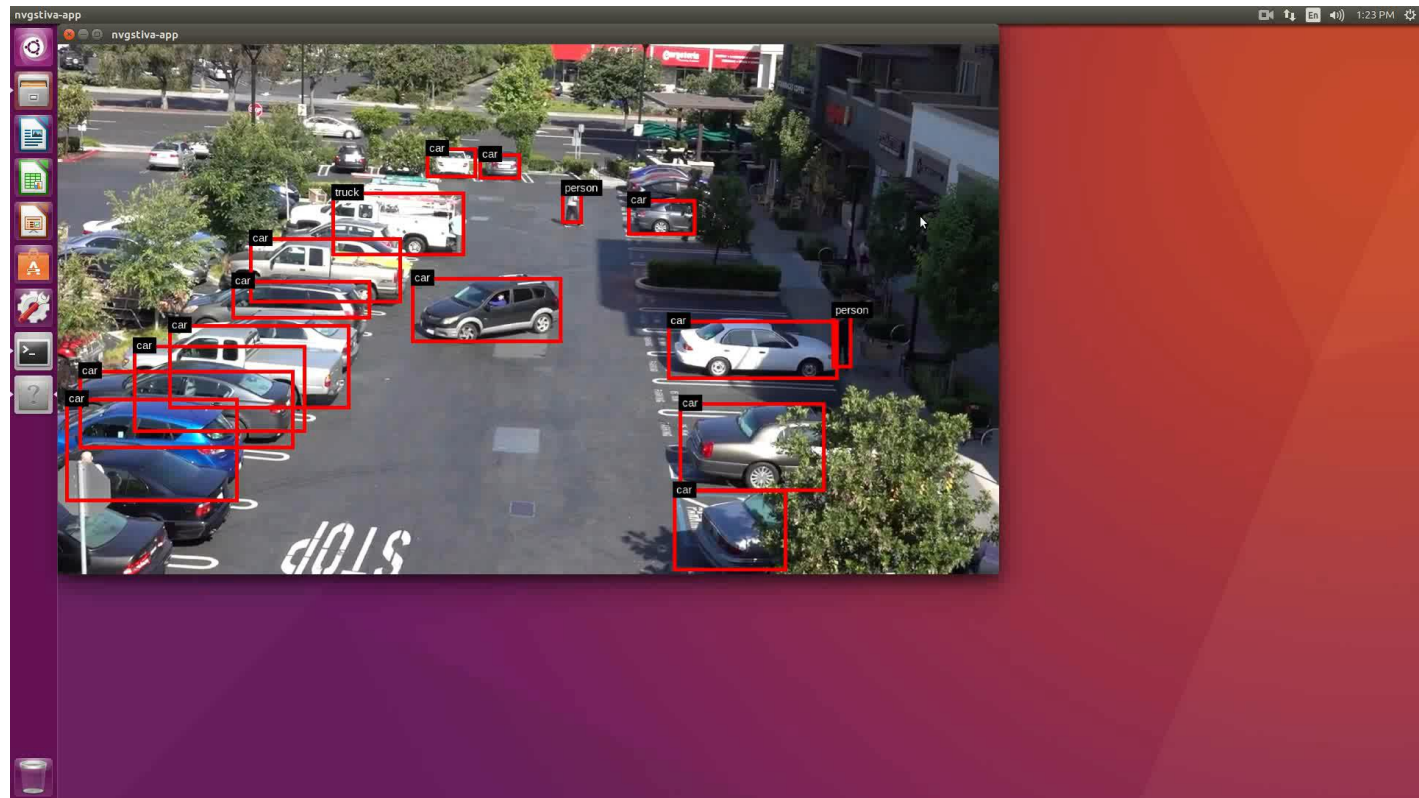
# NVIDIA DEEPSTREAM

- A unified release across all our platforms will allow application portability. This includes support for Jetson Nano, our newest AI platform.
- 50% reduction in memory footprint resulting in exceptional stream processing density.
- Turnkey integration with Azure Edge IoT to build applications and services, leveraging the power of Azure cloud.
- Containerized deployment for Jetson platforms. This enhances the ability to deploy applications to Docker, dramatically enhancing the delivery and maintenance of applications at large scale.
- Plugin sources for inference, message schema converter, and message broker plugins.
- A new reference tracker for robust object tracking.
- Added support for heterogeneous cameras, segmentation networks, monochrome images and hardware accelerated JPEG decode and encode.

## TRANSFER LEARNING TOOLKIT



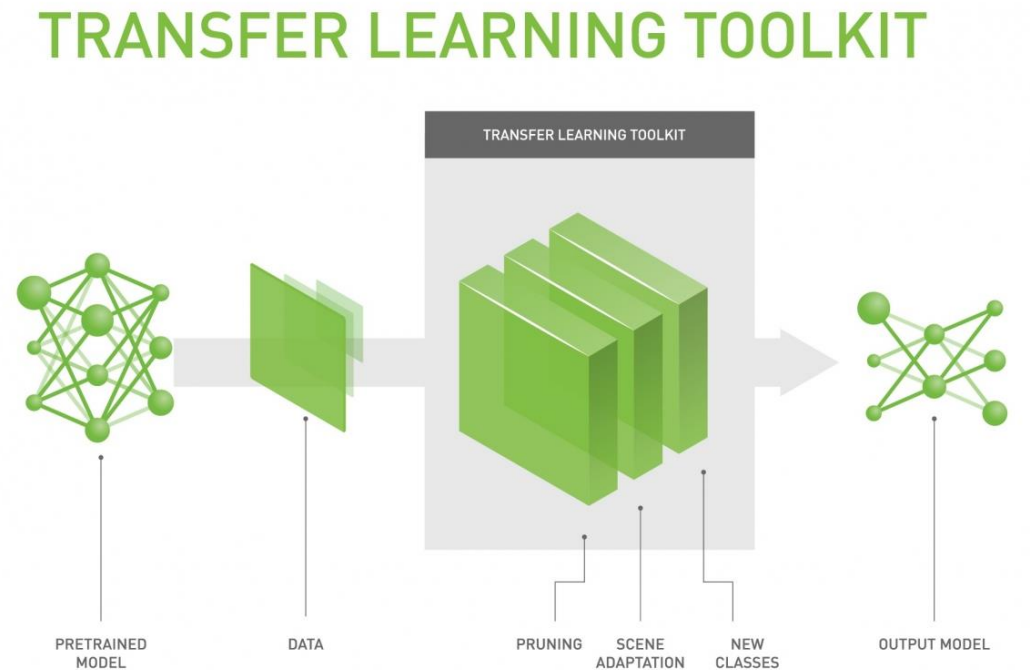
```
Terminal
nvidia@cegra-ubuntu: ~/TensorRTyolo
OPS: 65 res 62 19 x 19 x1024 -> 19 x 19 x1024
66 conv 512 1 x 1 / 1 19 x 19 x1024 -> 19 x 19 x 512 0.379 BFL
OPS: 67 conv 1024 3 x 3 / 1 19 x 19 x 512 -> 19 x 19 x1024 3.407 BFL
OPS: 68 res 65 19 x 19 x1024 -> 19 x 19 x1024
69 conv 512 1 x 1 / 1 19 x 19 x1024 -> 19 x 19 x 512 0.379 BFL
OPS: 70 conv 1024 3 x 3 / 1 19 x 19 x 512 -> 19 x 19 x1024 3.407 BFL
OPS: 71 res 68 19 x 19 x1024 -> 19 x 19 x1024
72 conv 512 1 x 1 / 1 19 x 19 x1024 -> 19 x 19 x 512 0.379 BFL
OPS: 73 conv 1024 3 x 3 / 1 19 x 19 x 512 -> 19 x 19 x1024 3.407 BFL
OPS: 74 res 71 19 x 19 x1024 -> 19 x 19 x1024
75 conv 512 1 x 1 / 1 19 x 19 x1024 -> 19 x 19 x 512 0.379 BFL
OPS: 76 conv 1024 3 x 3 / 1 19 x 19 x 512 -> 19 x 19 x1024 3.407 BFL
OPS: 77 conv 512 1 x 1 / 1 19 x 19 x1024 -> 19 x 19 x 512 0.379 BFL
OPS: 78
```



# NVIDIA TRANSFER LEARNING TOOLKIT

Transfer Learning Toolkit是一个基于python的工具包，它使开发人员能够利用NVIDIA预先训练的模型，并为开发人员提供一系列的工具，使流行的网络架构适应他们自己的数据，并且能够训练、调整、修剪和导出模型以进行部署。它还拥有简单的接口和抽象API，提高了深度学习训练工作流的效率。

- GPU优化的预训练砒码，可用于计算机视觉任务
- 轻松修改配置文件以添加新类并使用自定义数据重新训练模型
- 在异构的多GPU环境中执行模型调整和重新训练
- 使用修剪功能缩小模型尺寸
- 模型导出API，可在具有NVIDIA Tesla和Jetson产品的NVIDIA DeepStream SDK上部署



# NVIDIA TRANSFER LEARNING TOOLKIT

## Efficient Pre-trained Models

GPU-accelerated high performance models trained on large scale datasets.

## Faster Inference with Model Pruning

Model pruning reduces size of the model resulting in faster inference

## Training with Multiple GPUs

Re-training models, adding custom data for multi GPU training using an easy to use tool

## Abstraction

Abstraction from having deep knowledge of frameworks, simple intuitive interface to the features

## Containerization

Packaged in a container easily accessible from NVIDIA GPU Cloud website. All code dependencies are managed automatically

## Integration

Models exported using TLT are easily consumable for inference with Deep Stream SDK

# NVIDIA TRANSFER LEARNING TOOLKIT

在指定的公共数据集上训练的图像分类和目标检测模型，可与Transfer Learning Toolkit一起使用。

## Image Classification

- ResNet10/18/50
- VGG16/19
- MobileNet V1/V2
- AlexNet
- SqueezeNet
- GoogLeNet

Faster RCNN supporting backbones:

- ResNet10/18/50
- VGG16/19
- GoogLeNet
- MobileNet V1/V2

## Object Detection

DetectNet\_v2 supporting backbones:

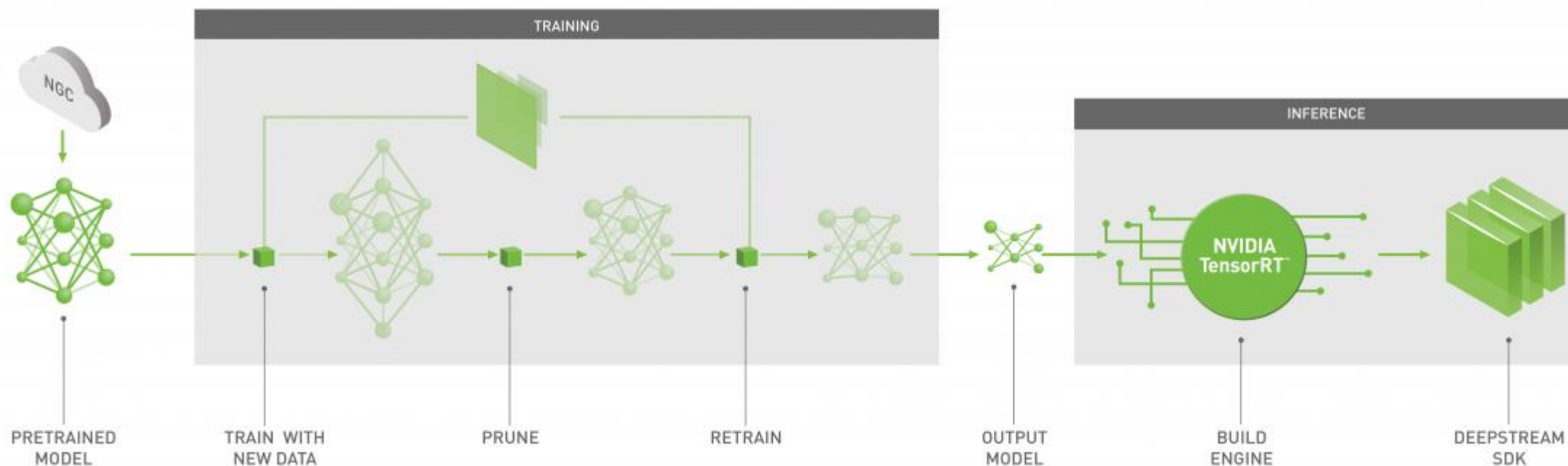
- ResNet10/18/50
- VGG 16/19
- GoogLeNet
- MobileNet V1/V2

SSD:

- ResNet10/18

# NVIDIA TRANSFER LEARNING TOOLKIT

为应用在计算机视觉领域的深度学习工作流程，提供了全方位的便利工具



# 在服务器上部署TRANSFER LEARNING TOOLKIT

## Hardware Requirements

### Minimum

- 4 GB system RAM
- 4 GB of GPU RAM
- Single core CPU
- 1 GPU
- 50 GB of HDD space

### Recommended

- 32 GB system RAM
- 32 GB of GPU RAM
- 8 core CPU
- 4 GPUs
- 100 GB of SSD space

## Software Requirements

- Ubuntu 18.04 LTS/ **16.04 LTS**
  - NVIDIA GPU Cloud account and API key - <https://ngc.nvidia.com/>
  - docker-ce installed, <https://docs.docker.com/install/linux/docker-ce/ubuntu/>
  - nvidia-docker2 installed, instructions: [https://github.com/nvidia/nvidia-docker/wiki/Installation-\(version-2.0\)](https://github.com/nvidia/nvidia-docker/wiki/Installation-(version-2.0))
  - NVIDIA GPU driver v410.xx or above**
- Note:** DeepStream 4.0 - NVIDIA SDK inference <https://developer.nvidia.com/deepstream-sdk> is recommended.

# 在服务器上部署TRANSFER LEARNING TOOLKIT

## Installation Prerequisites

- Install Docker. See: <https://www.docker.com/>.
- NVIDIA GPU driver v410.xx or above. Download from <https://www.nvidia.com/Download/index.aspx?lang=en-us>.
- Install NVIDIA Docker 2 from: <https://github.com/NVIDIA/nvidia-docker>.

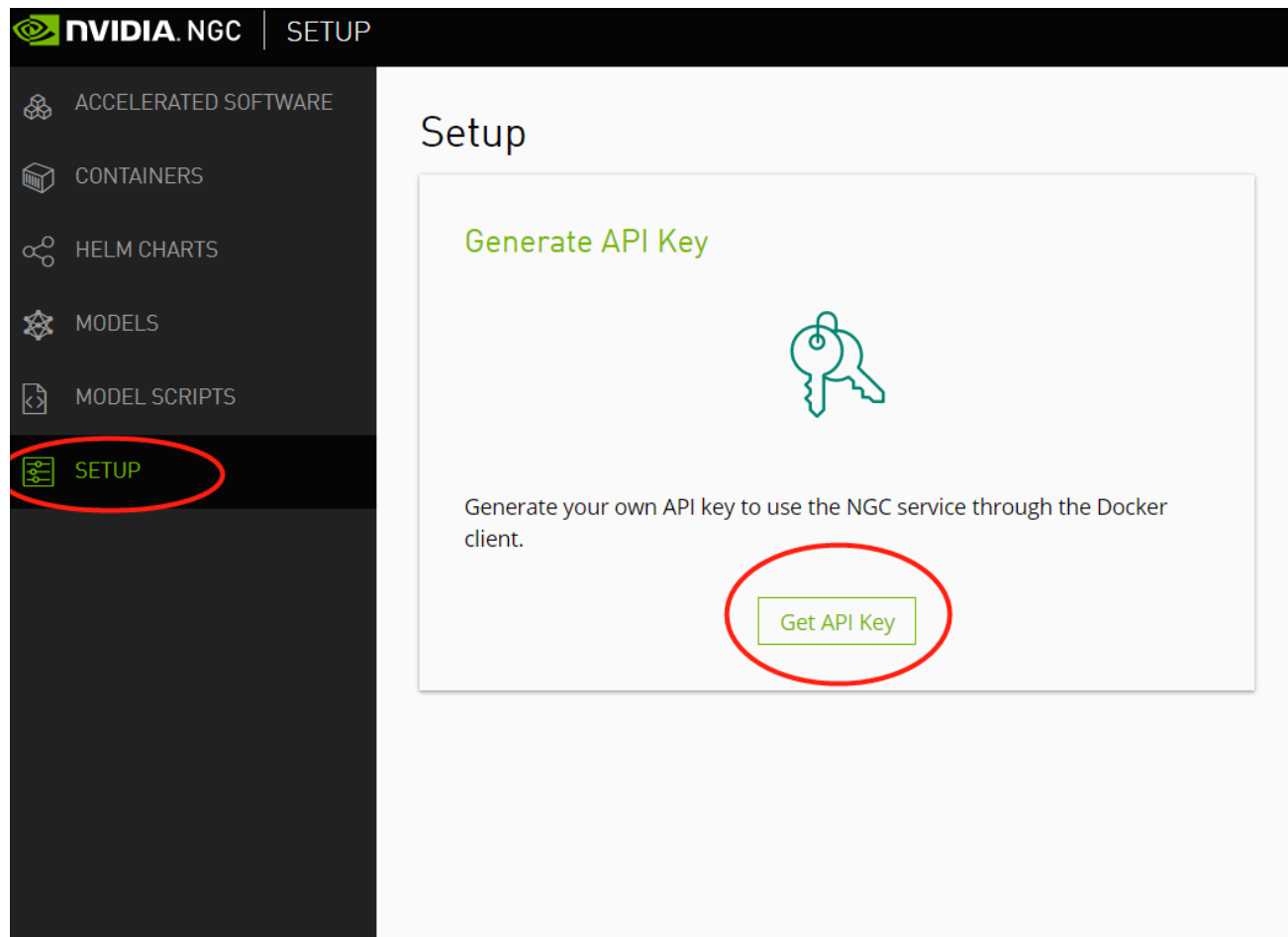
### Get an NGC API key

- NVIDIA GPU Cloud account and API key - <https://ngc.nvidia.com/>
  1. Go to NGC and click the **Transfer Learning Toolkit** container in the **Catalog** tab. This message is displayed, **Sign in to access the PULL feature of this repository.**
  2. Enter your email address and click **Next** or click **Create an Account.**
  3. Choose your **organization** when prompted for Organization/Team.
  4. Click **Sign In.**
  5. Select the **Containers** tab on the left navigation pane and click the **Transfer Learning Toolkit** tile.

### Download the docker container

- Execute docker login nvcr.io from the command line and enter your username and password.
  - Username: \$oauthtoken
  - Password: API\_KEY
- Execute docker pull nvcr.io/nvidia/tlt-streamanalytics:<version>

# 在服务器上部署TRANSFER LEARNING TOOLKIT




**NVIDIA. NGC** | SETUP

- ACCELERATED SOFTWARE
- CONTAINERS
- HELM CHARTS
- MODELS
- MODEL SCRIPTS
- SETUP**

## Setup

### Generate API Key



Generate your own API key to use the NGC service through the Docker client.

**Get API Key**

# 在服务器上部署TRANSFER LEARNING TOOLKIT

Setup > API Key

Generate API Key

## API

### API Information

Generate your own API key to use the NGC service through the Docker client. Anyone with this API Key has access to all services, actions, and resources on your behalf.

Click Generate API Key to create your own API Key. If you have forgotten or lost your API Key, you can come back to this page to create a new one at any time.

### Usage

Use your API key to log in to the NGC registry by entering the following command and following the prompts:

### NGC CLI

```
$ ngc config set
```

### Docker™

For the username, enter '\$oauthtoken' exactly as shown. It is a special authentication token for all users.

```
$ docker login nvcr.io
```

```
Username: $oauthtoken
```

```
Password: <Your Key>
```

# 在服务器上部署TRANSFER LEARNING TOOLKIT

在服务器上运行TLT的镜像

1.Run the image using this command.

```
docker run --runtime=nvidia -it nvcr.io/nvidia/tlt-streamanalytics:v1.0.1_py2 /bin/bash
```

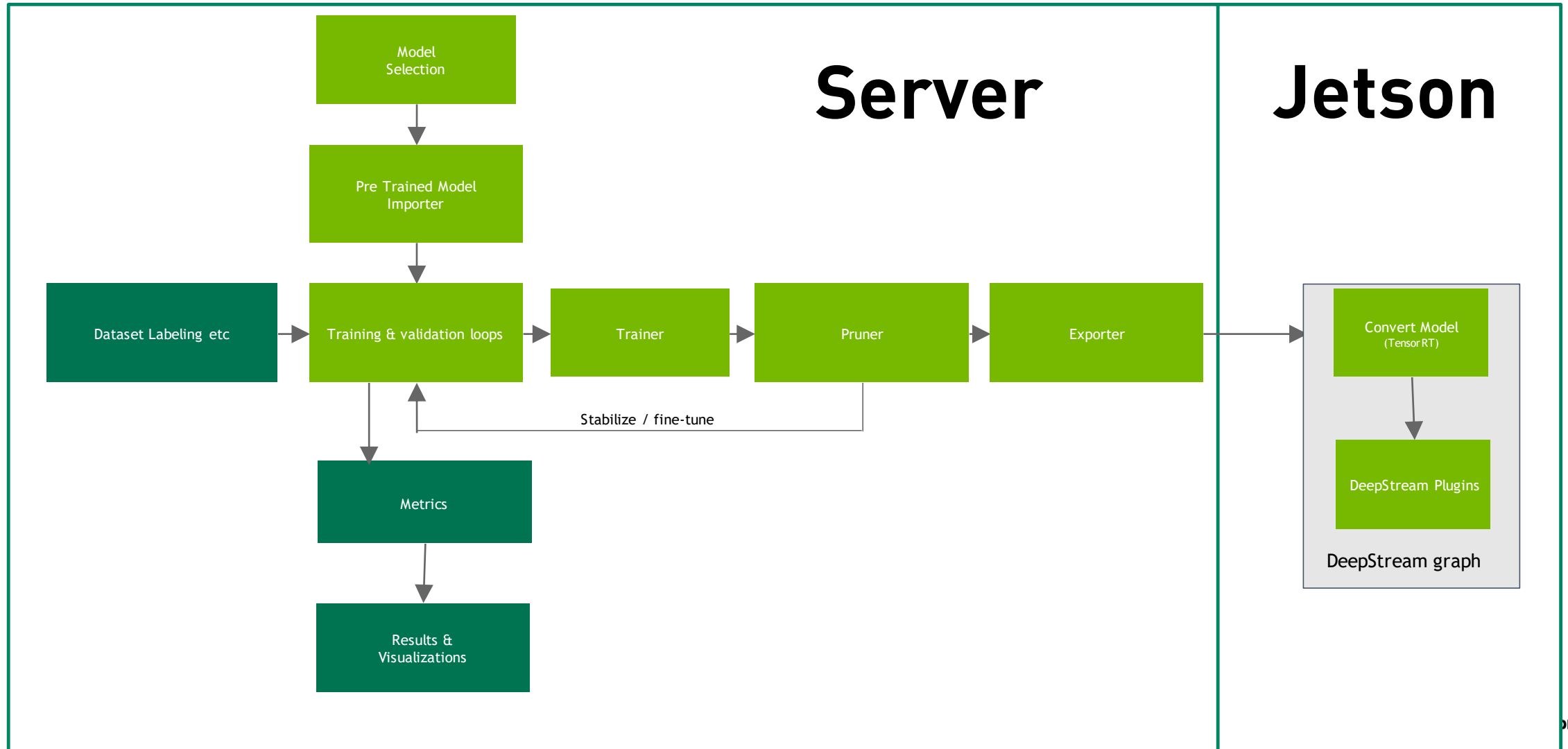
2.Mount local directories using -v and expose the docker ports to the host using -p

```
docker run --runtime=nvidia -it \  
-v "/path/to/dir/on/host":"/path/to/dir/in/docker" \ -p 8888:8888 \  
nvcr.io/nvidia/tlt-streamanalytics:v1.0_py2 /bin/bash
```

3.Invoke the jupyter notebook using the following command

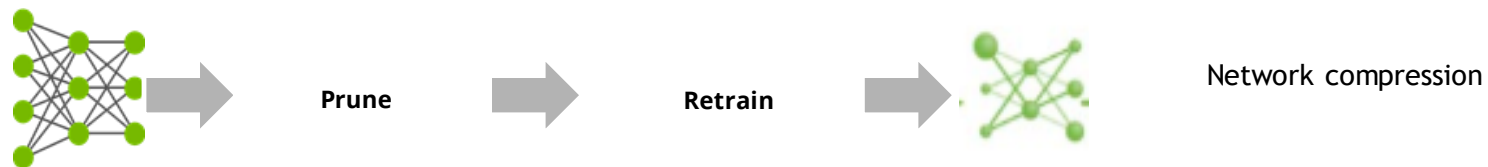
```
jupyter notebook --ip 0.0.0.0 --port 8888 --allow-root
```

# TRANSFER LEARNING TOOLKIT的工作流程

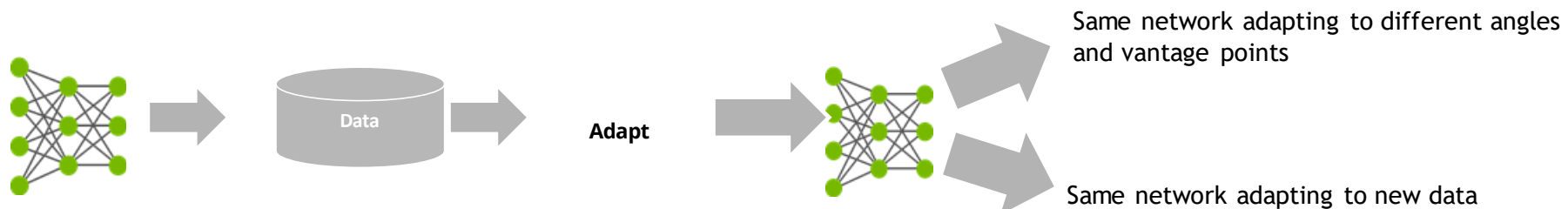


# TRANSFER LEARNING TOOLKIT的工作流程

## Pruning

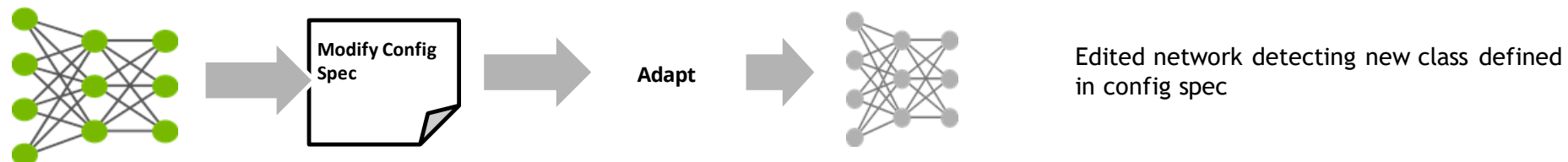


## Scene Adaptation

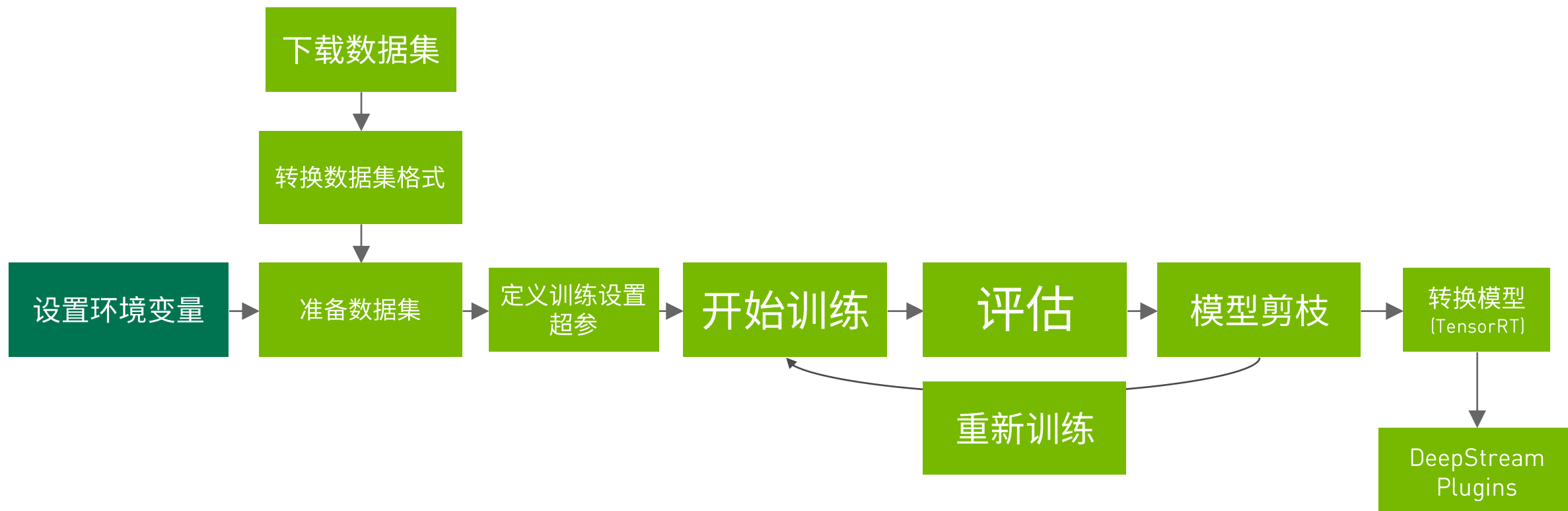


Train with new data from another vantage point, camera location, or added attribute

## New Classes



# TRANSFER LEARNING TOOLKIT的工作流程



# 总结

- NVIDIA Transfer Learning Toolkit为深度学习训练部署流程提供了完整的工具链
- Transfer Learning Toolkit 的安装部署需要使用NGC
- 把训练和剪裁好的模型部署在边缘设备(Jetson 平台)上时，需要在边缘设备上转换成TRT的格式

链接：<https://pan.baidu.com/s/1TWT5PMYn-VkYoxbVCWKoWA>  
提取码：cdaq

# 更多资源：

# <https://developer.nvidia-china.com>



何琨-Ken

北京 密云



扫一扫上面的二维码图案，加我微信



THANK YOU

