Introduction to OpenMMLab:
An Open-source Algorithm Platform for Computer Vision

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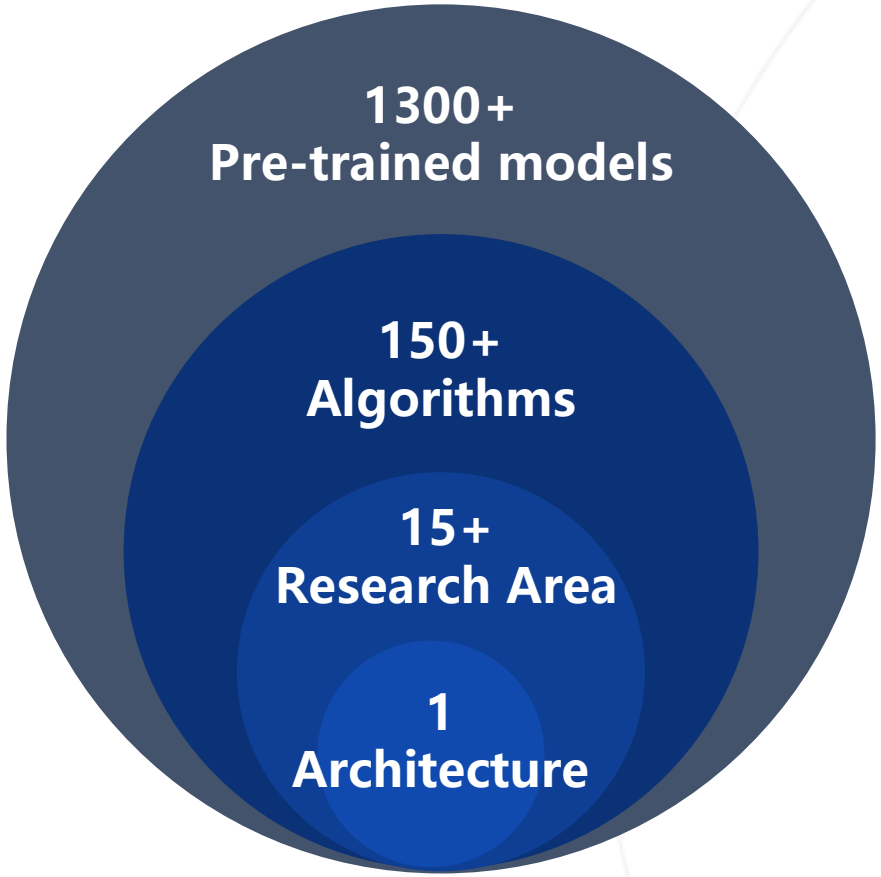
OpenMMLab

Overview



1 Architecture

- A unified architecture for all codebases

15+ Research Area

- Cover various areas for computer vision research

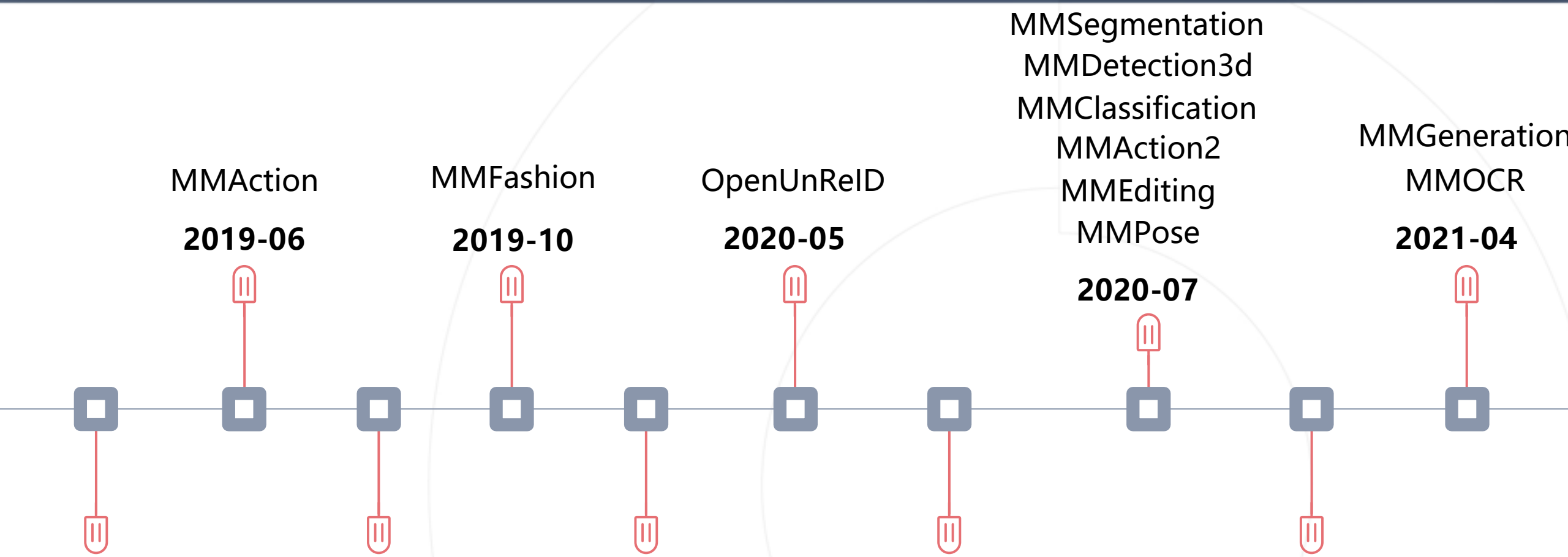
150+ Algorithms

- Implement SOTA algorithms

1300+ Pretrained Models

- Benchmark and out-of-box usage

History


2018-08
MMCV
MMDetection
MMAction
2019-06
2019-08
MMSR
MMSkeleton
MMFashion
2019-10
2020-03
OpenPCDet
OpenUnReID
2020-05
2020-06
OpenSelfSup
MMSegmentation
MMDetection3d
MMClassification
MMAction2
MMEditing
MMPose
2020-07
2021-01
MMTracking
MMGeneration
MMOCR
2021-04

Framework

OpenMMLab

Codebases & Datasets

15+ Codebases

- MMClassification
- MMDetection
- MMSegmentation
- MMGeneration
- MMEditing
- MMTracking
- MMDetection3D
- MMAction2
- MMPose

40+ Datasets

- WIDER
- FineGym
- MovieNet
- Placepedia

MMCV

- Common Utilities
- Training Interface

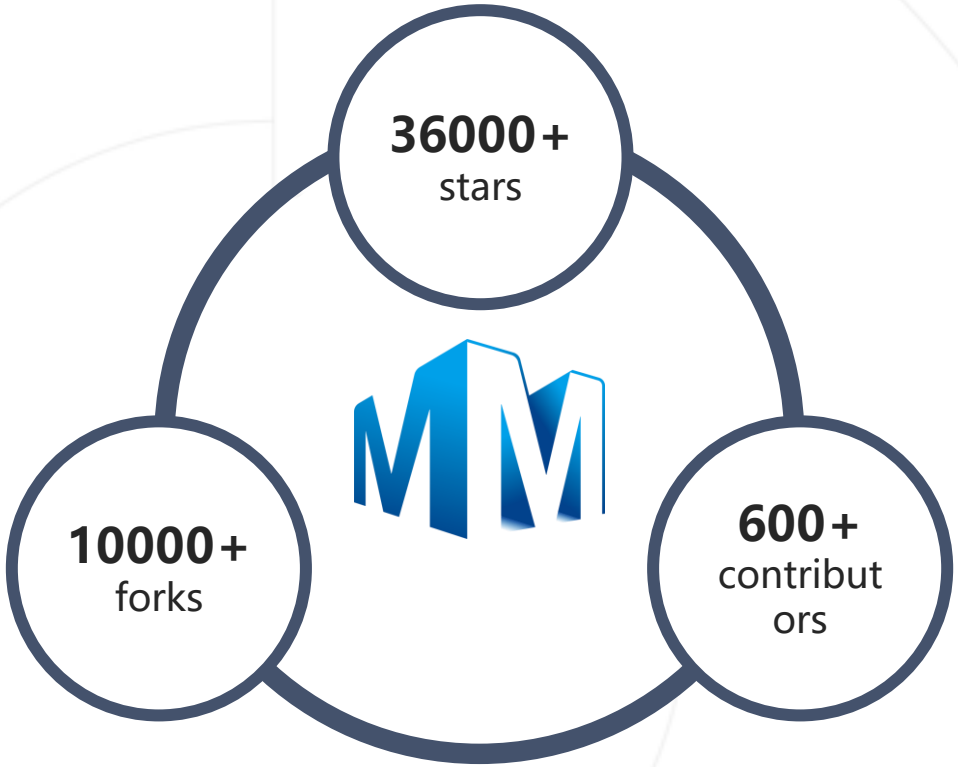
DL Framework

PyTorch

Community Impact

GitHub Statistics

- More than 36k stars
- 600+ contributors from academia and industry
- More than 800k page views of GitHub pages of all projects



Academic Impact

- MMDetection gained more than **450** citations, and selected as PapersWithCode 2020 Top10 Trending Libraries
- Many codebases such as MMDetection, MMDetection3D, MMSegmentation, MMPose are used by winners of various challenges
- More than **35** papers in CVPR 2021 adopt OpenMMLab projects as their implementation or benchmark

Major Features

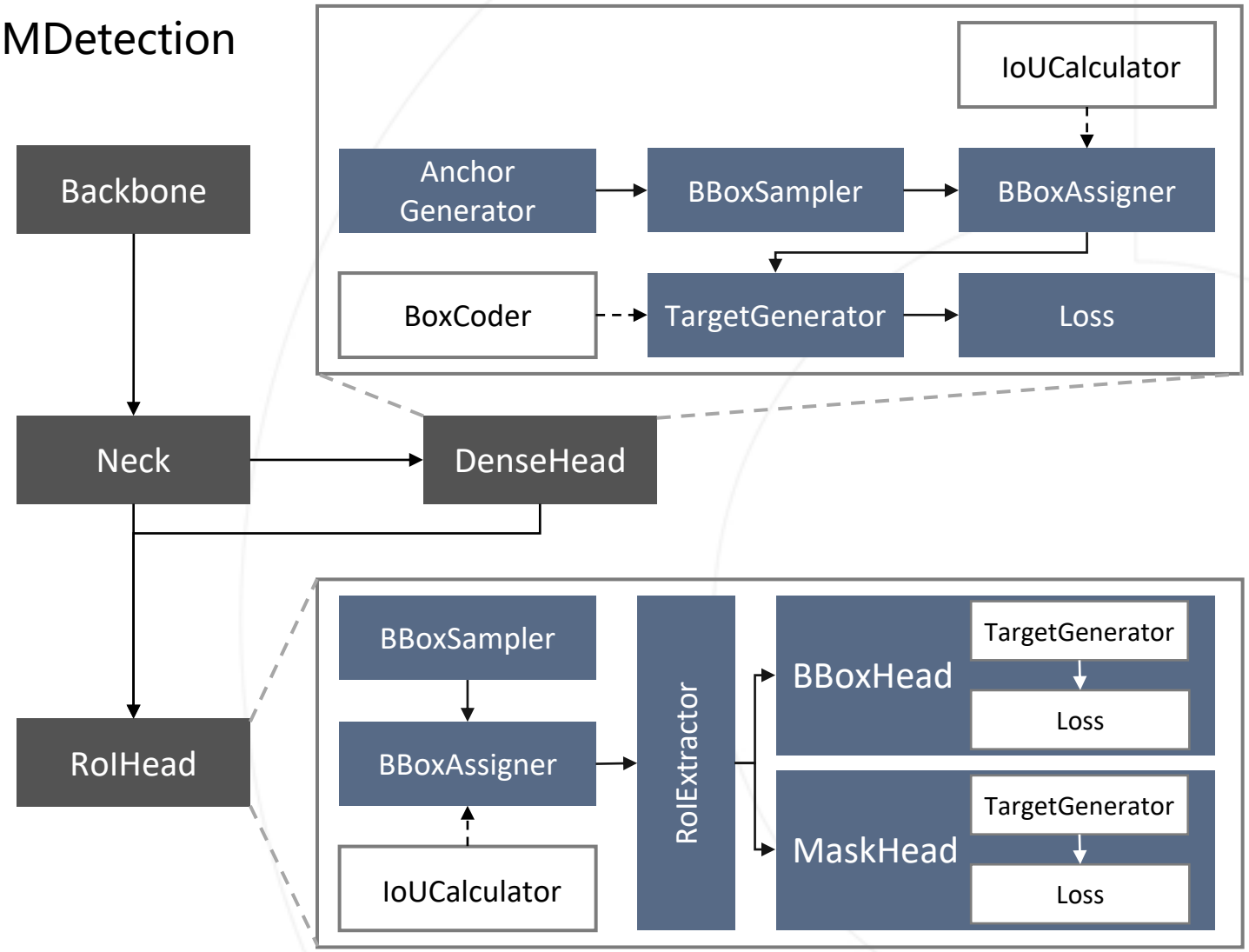
- **Unified benchmark**: Provide a fair research platform and various baselines
- **Modular design**: Fast to develop and try new components
- **High-quality implementation**: Efficiency, accuracy, code style

Unified Benchmark

	Paper A	Paper B	Paper C
1 backbone	• standard ResNet • torchvision pretrain	• ResNet variant • third-party pretrain	• ResNet variant • third-party pretrain
2 augmentation	• 769x769 crop • horizontal flip	• 769x769 crop • horizontal flip • color jittor	• 512x1024 crop • horizontal flip • color jittor
3 method	• method A	• method B	• method C
4 training schedule	• 200 epoch • initial LR 0.1 • poly	• 120k iteration • initial LR 0.01 • step	• 100 epoch • initial LR 0.01 • poly
5 inference	• multi-scale • crop	• single-scale • crop	• single-scale • whole image
	mIoU(A)	> mIoU(B)	> mIoU(C)

Modular Design

Example of MMDetection



Architecture Design

Core concepts for training

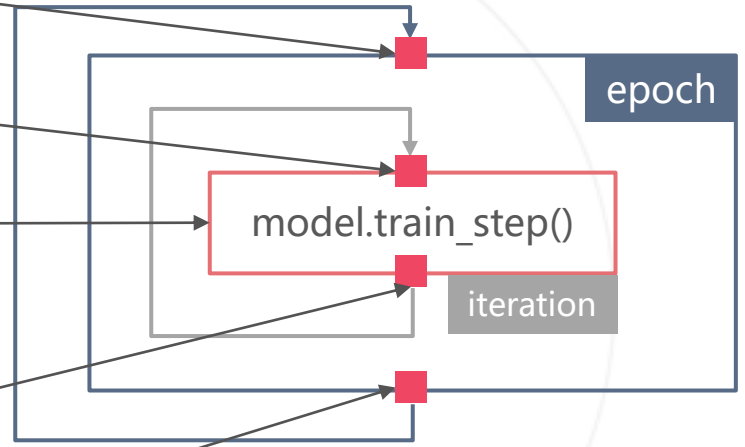
- Runner&Hook
- Registry

Concept: Runner&Hook

ImageNet Example

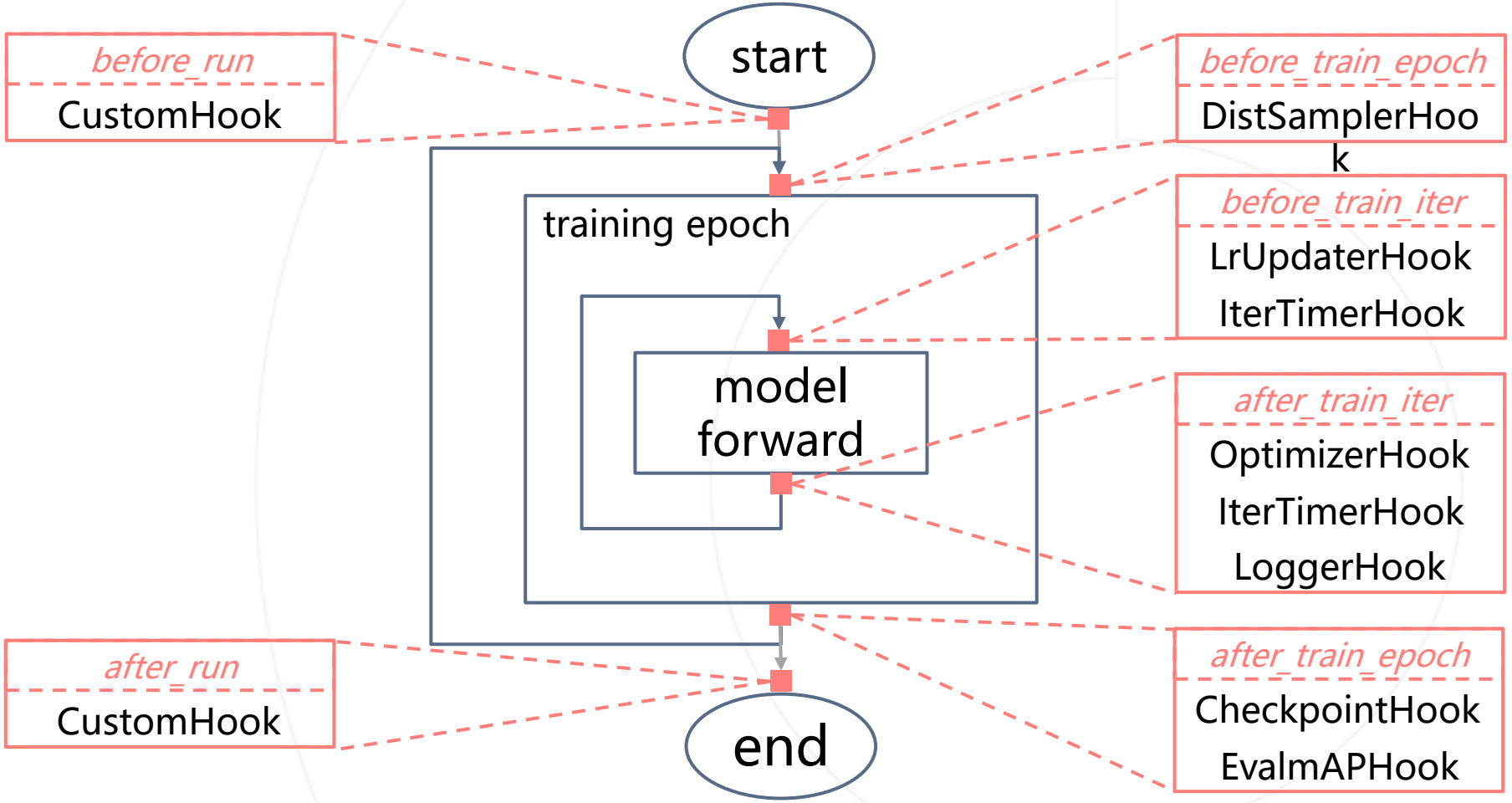
```
model = torch.nn.parallel.DistributedDataParallel(SomeNet(), device_ids=[args.gpu])
optimizer = torch.optim.SGD(...)
train_loader = torch.utils.data.DataLoader(...)

def adjust_learning_rate():
  pass

def record_and_log_loss():
  pass

for epoch in range(args.epochs):
  adjust_learning_rate(optimizer, epoch, args)

  # train for one epoch
  for i, (images, target) in enumerate(train_loader):
    # measure data loading time
    data_time.update(time.time() - end)

    # compute output
    output = model(images)
    loss = criterion(output, target)

    # compute gradient and do SGD step
    optimizer.zero_grad()
    loss.backward()
    optimizer.step()

    # measure accuracy and record loss
    record_and_log_loss(loss)

    # measure elapsed time
    batch_time.update(time.time() - end)
    end = time.time()

    # print progress
    if i % args.print_freq == 0:
      progress.display(i)

  # evaluate on validation set
  save_checkpoint()
```

Runner: execution loop and core logic (fetch data and forward model)
Hook: custom logic and facilities (logging, visualization, lr scheduler, checkpointing, etc)



Concept: Runner&Hook

Another example



Concept: Registry

Build an instance with custom configs

1. Register

```
BACKBONES = Registry('backbones')

@BACKBONES.register_module()
class ResNet(nn.Module):

    pass
```

Registry

BACKBONES

'ResNet' -> <class 'ResNet'>

2. Build

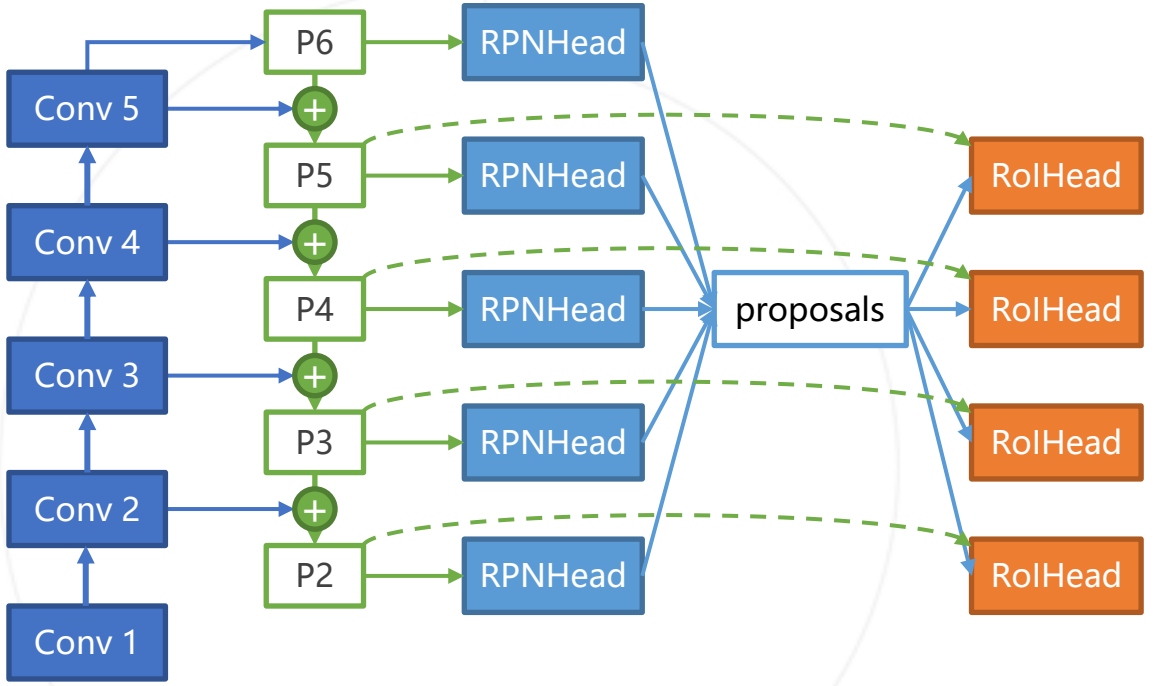
```
config = dict(type='ResNet')
backbone = build_backbone(config, BACKBONES)
```

Concept: Registry

Config

```
model = dict(
    type='FasterRCNN',
    pretrained='torchvision://resnet50',
    backbone=dict(
        type='ResNet',
        depth=50,
        ...),
    neck=dict(
        type='FPN',
        ...),
    rpn_head=dict(
        type='RPNHead',
        ...),
    roi_head=dict(
        type='StandardRoIHead',
        bbox_roi_extractor=dict(
            type='SingleRoIExtractor',
            ...),
        bbox_head=dict(
            type='Shared2FCBBoxHead',
            ...))
)
```

Module



Develop with OpenMMLab Codebases

1. Develop a new algorithm based on some existing codebase
2. Develop a new codebase based on existing codebases
3. Develop a new codebase following the architecture design of OpenMMLab

Develop with OpenMMLab Codebases