

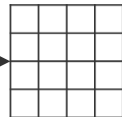


## Faster R-CNN算法改进：多个检测层



64x64输入图像

stride=16



4x4特征



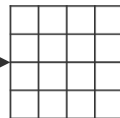


## Faster R-CNN算法改进：多个检测层



64x64输入图像

stride=16

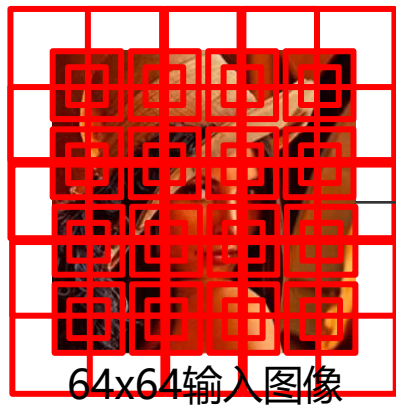


4x4特征

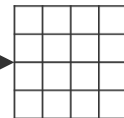




## Faster R-CNN算法改进：多个检测层



stride=16

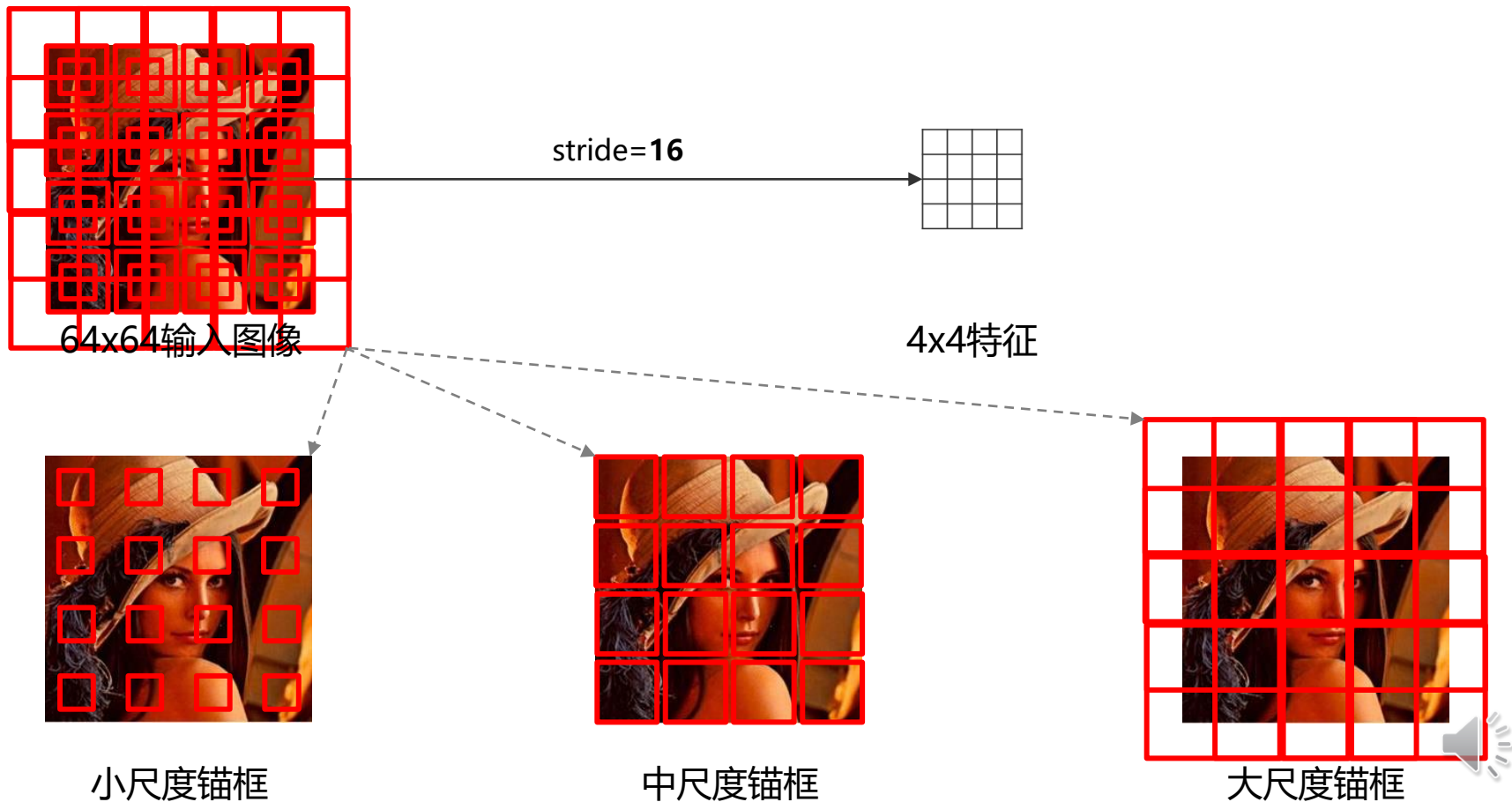


4x4特征



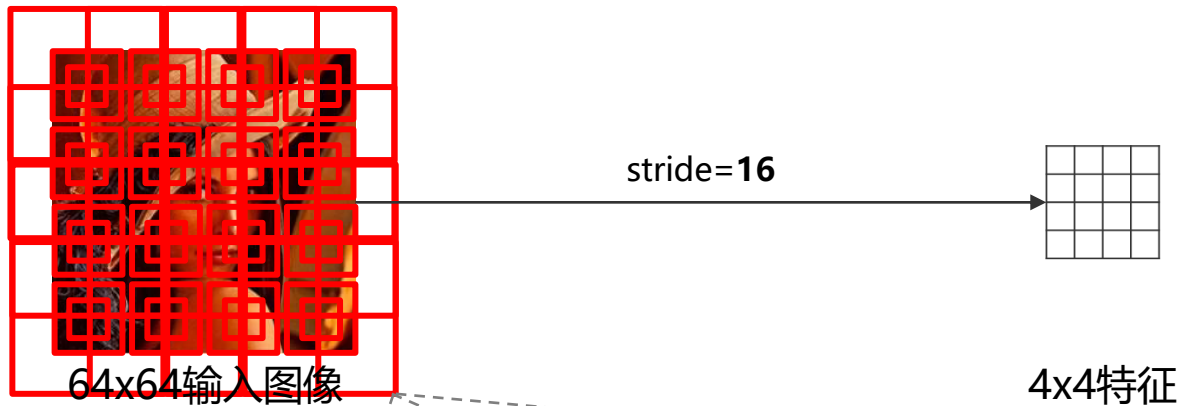


## Faster R-CNN算法改进：多个检测层





## Faster R-CNN算法改进：多个检测层

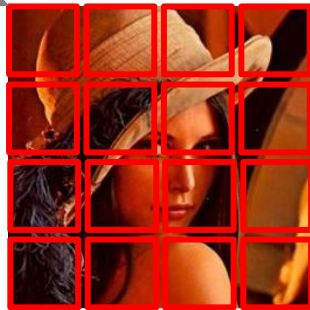


### 单个检测层的问题

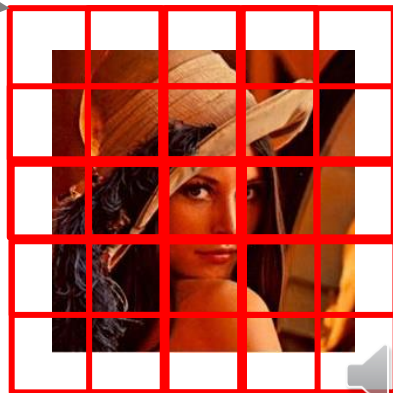
- ① 小尺度锚框铺设太稀疏
- ② 大尺度锚框铺设太稠密



小尺度锚框



中尺度锚框



大尺度锚框

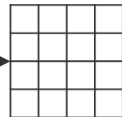


## Faster R-CNN算法改进：多个检测层



64x64输入图像

stride=16



4x4特征



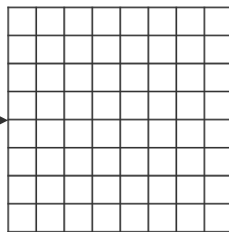


## Faster R-CNN算法改进：多个检测层



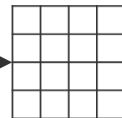
64x64输入图像

stride=8



8x8特征

stride=2



4x4特征

stride=2

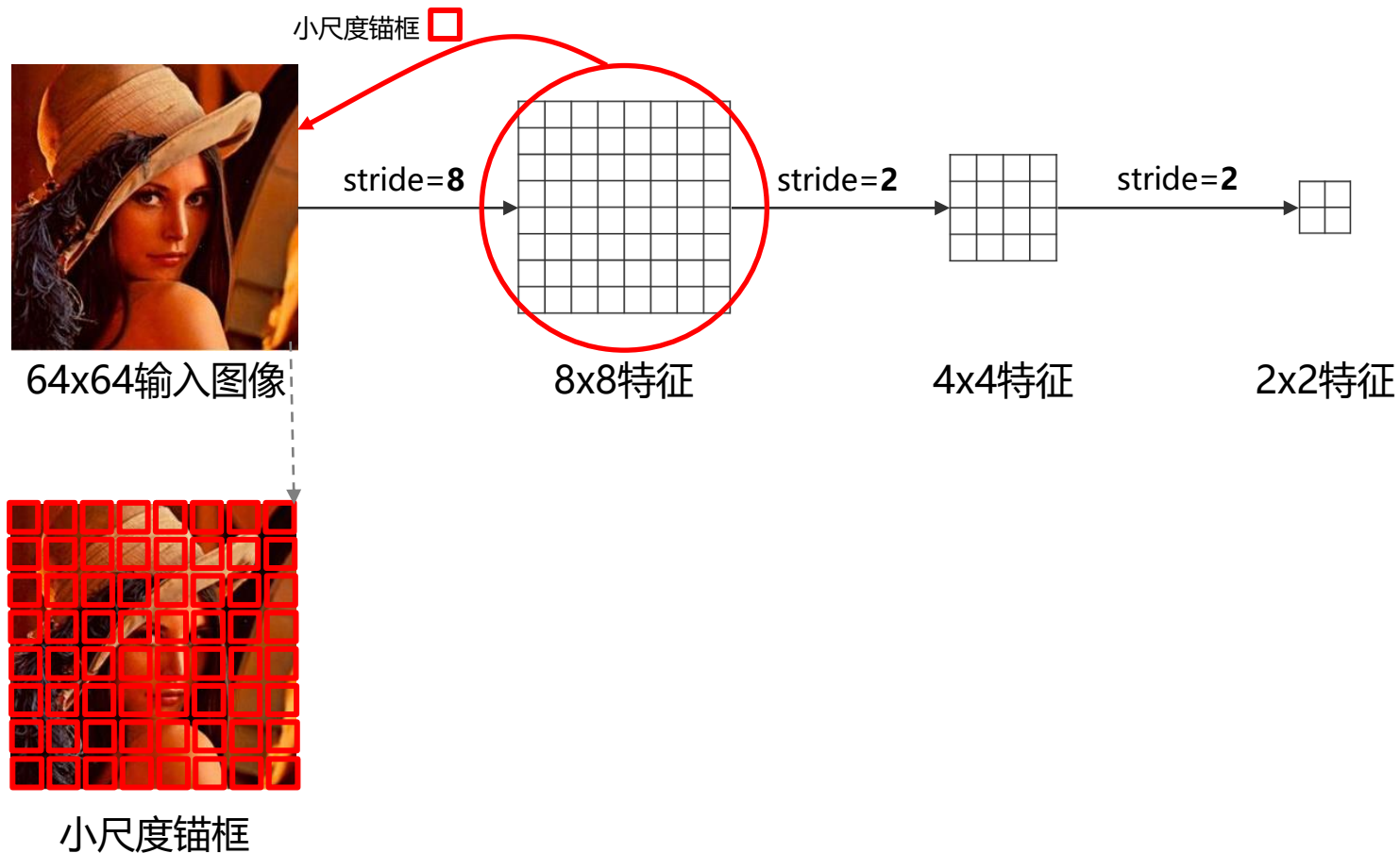


2x2特征





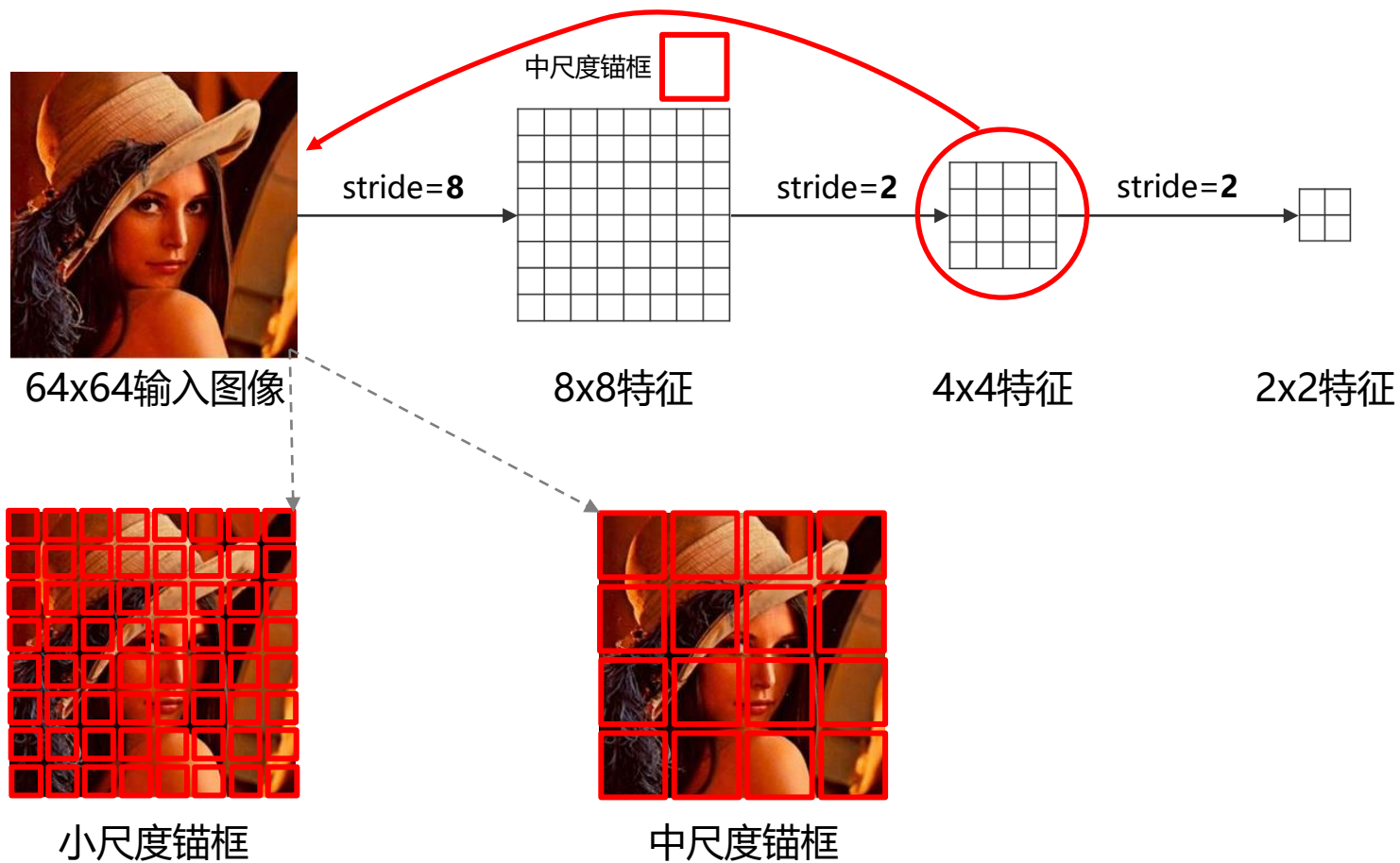
## Faster R-CNN算法改进：多个检测层





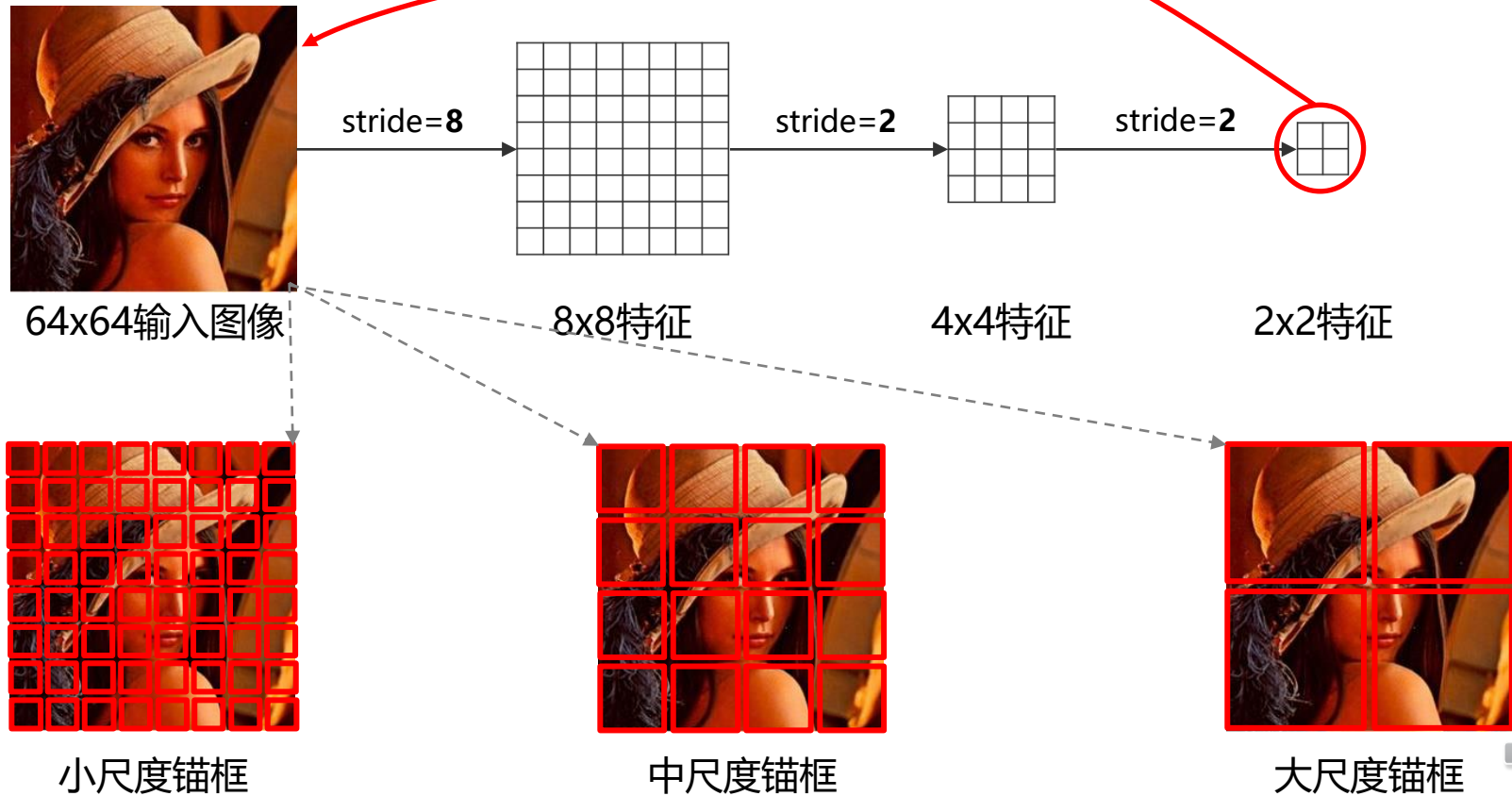


## Faster R-CNN算法改进：多个检测层



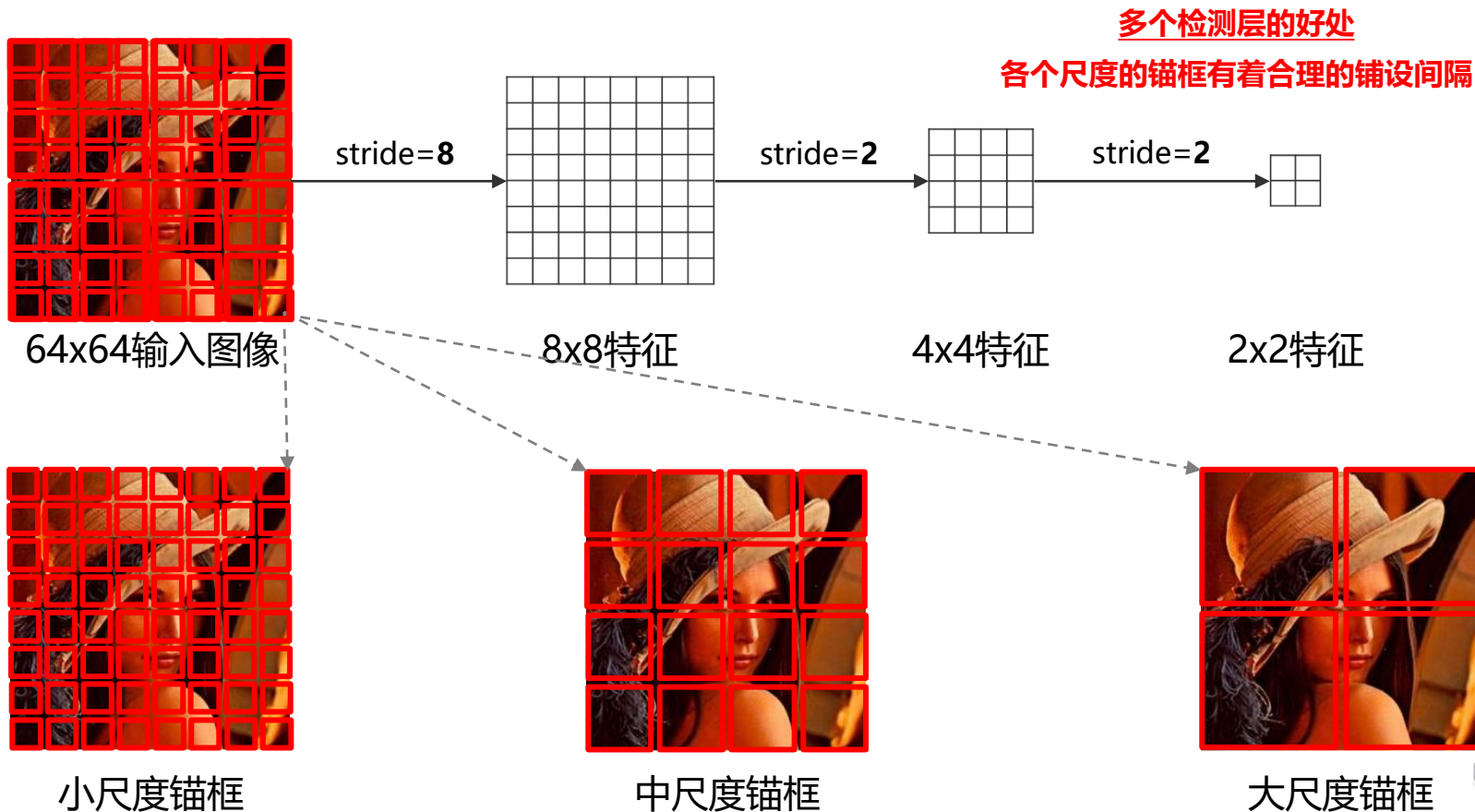


## Faster R-CNN算法改进：多个检测层



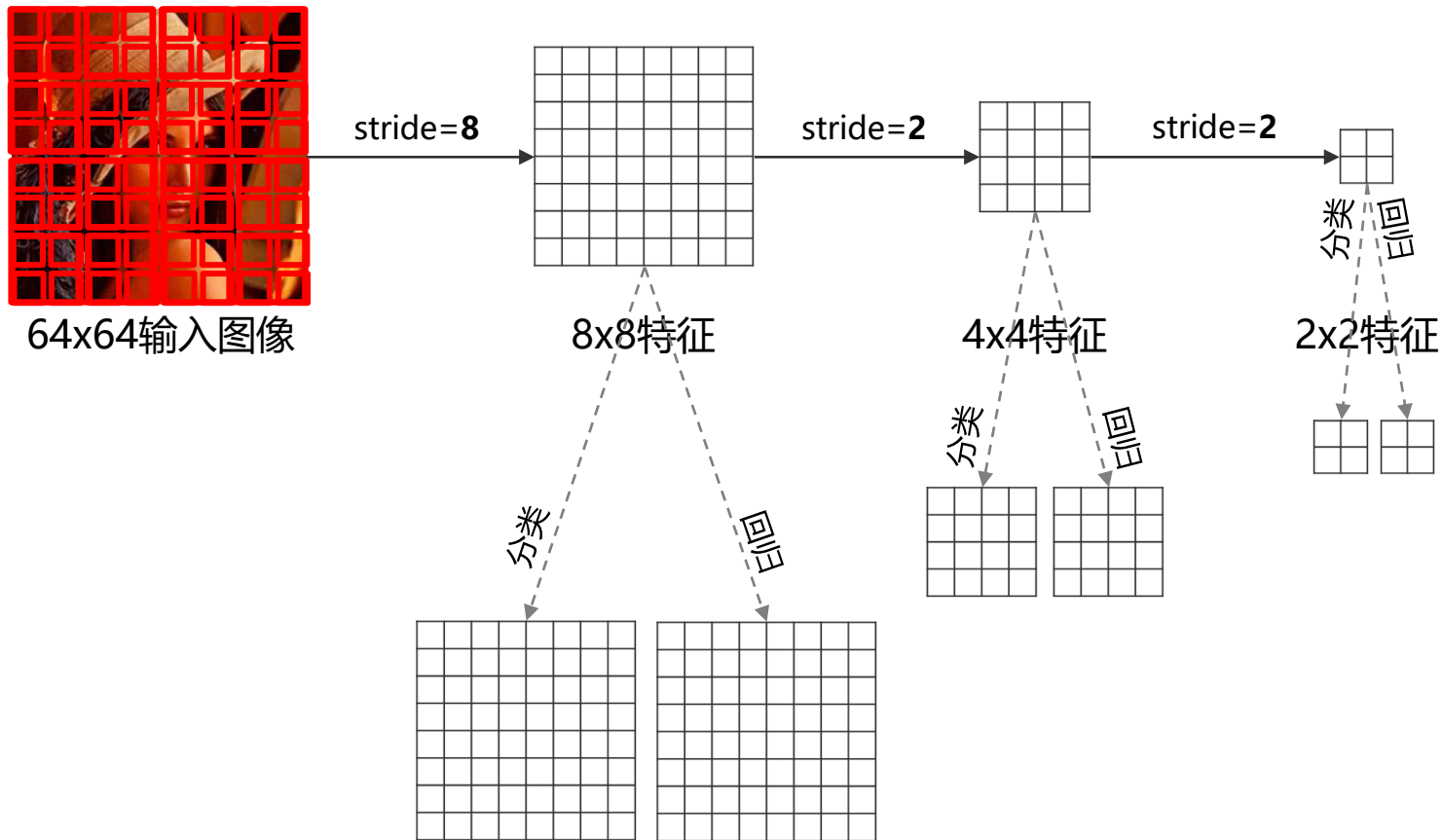


## Faster R-CNN算法改进：多个检测层



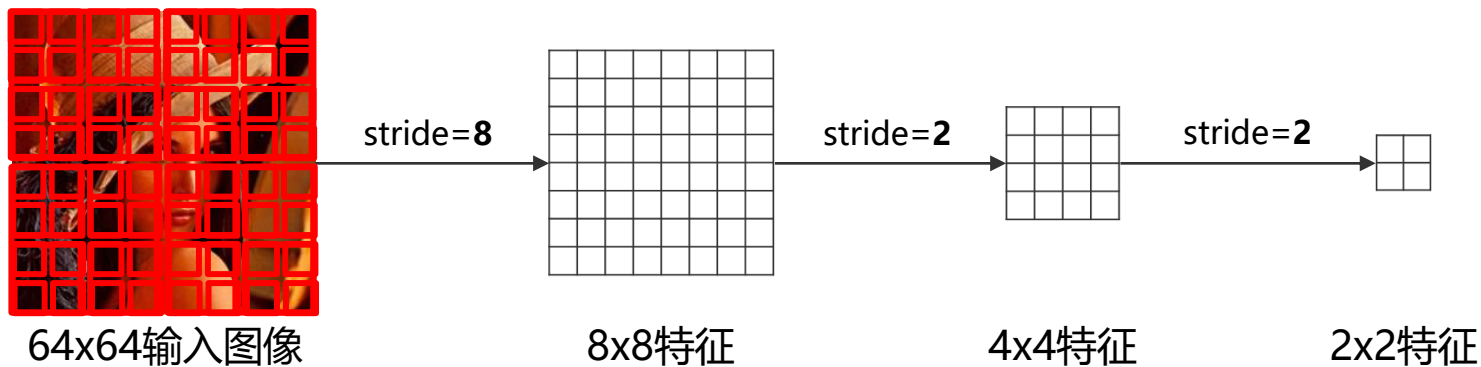


## Faster R-CNN算法改进：多个检测层





## Faster R-CNN算法改进：特征金字塔 (FPN, Feature Pyramid Network)

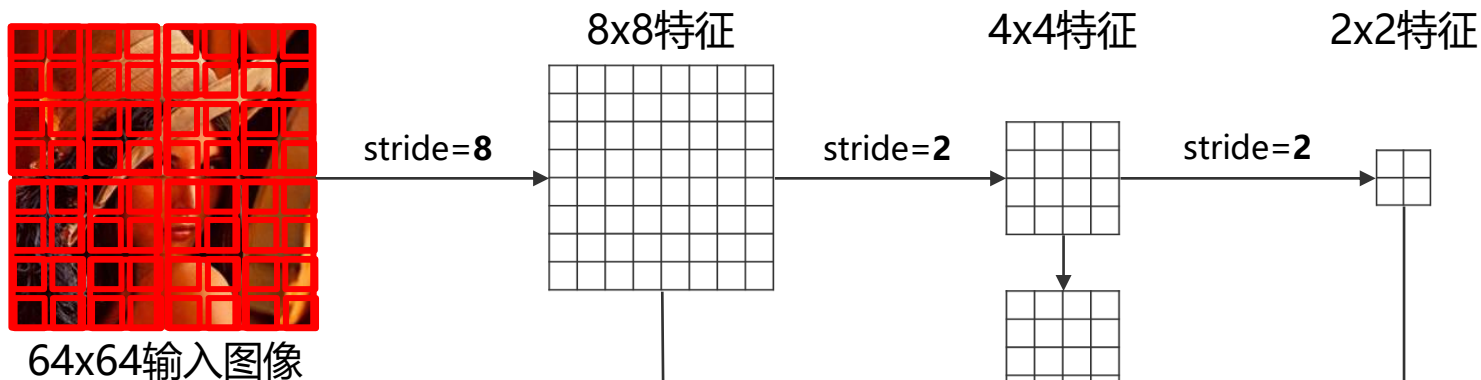


- 物体检测 = 物体分类 + 物体定位
- 物体分类：网络层数较深时，分辨率较小，语义信息比较丰富，利于分类
- 物体回归：网络层数较浅时，分辨率较大，细节信息比较丰富，利于回归
- 存在问题：现有分类网络中，某层的特征不能同时满足这两点，即语义和细节都丰富
- 解决方案：特征金字塔从后到前对特征进行融合，让每层的特征都具备丰富的语义和细节

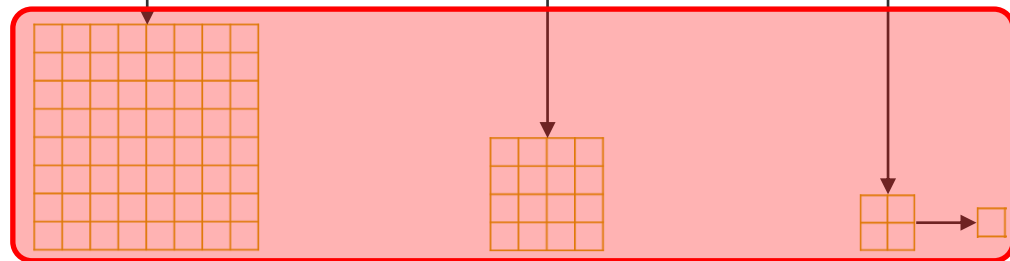




# Faster R-CNN算法改进：特征金字塔 (FPN, Feature Pyramid Network)



- 2x上采样：最近邻或双线性插值
- Conv后没有使用ReLU
- 深层语义信息+浅层细节信息
- 增加很少的额外计算来提升性能
- 后续有很多关于它的改进
- 物体检测算法基本都会使用



特征金字塔  
增强后的特征



## Faster R-CNN代码架构解读：配置文件

```
faster_rcnn_R_50_FPN_1x.yaml x
1  _BASE_: "../Base-RCNN-FPN.yaml"
2  MODEL:
3    WEIGHTS: "detectron2://ImageNetPretrained/MSRA/R-50.pkl"
4    MASK_ON: False
5    RESNETS:
6      DEPTH: 50
```

- faster\_rcnn\_R\_50\_FPN\_1x.yaml





## Faster R-CNN代码架构解读：配置文件

```
faster_rcnn_R_50_FPN_1x.yaml x
1  _BASE_: "../Base-RCNN-FPN.yaml"
2  MODEL:
3    WEIGHTS: "detectron2://ImageNetPretrained/MSRA/R-50.pkl"
4    MASK_ON: False
5    RESNETS:
6      DEPTH: 50
```

- faster\_rcnn\_R\_50\_FPN\_1x.yaml
- 基础网络是ImageNet预训练的ResNet50，加上FPN







## Faster R-CNN代码架构解读：配置文件

```
faster_rcnn_R_50_FPN_1x.yaml x
1  _BASE_: "../Base-RCNN-FPN.yaml"
2  MODEL:
3    WEIGHTS: "detectron2://ImageNetPretrained/MSRA/R-50.pkl"
4    MASK_ON: False
5    RESNETS:
6      DEPTH: 50
```

- faster\_rcnn\_R\_50\_FPN\_1x.yaml
- 基础网络是ImageNet预训练的ResNet50, 加上FPN

```
Base-RCNN-FPN.yaml x
1  MODEL:
2    META_ARCHITECTURE: "GeneralizedRCNN"
3    BACKBONE:
4      NAME: "build_resnet_fpn_backbone"
5    RESNETS:
6      OUT_FEATURES: ["res2", "res3", "res4", "res5"]
7    FPN:
8      IN_FEATURES: ["res2", "res3", "res4", "res5"]
9    ANCHOR_GENERATOR:
10     SIZES: [[32], [64], [128], [256], [512]] # One size for each in feature map
11     ASPECT RATIOS: [[0.5, 1.0, 2.0]] # Three aspect ratios (same for all feature maps)
12    RPN:
13     IN_FEATURES: ["p2", "p3", "p4", "p5", "p6"]
14     PRE_NMS_TOPK_TRAIN: 2000 # Per FPN level
15     PRE_NMS_TOPK_TEST: 1000 # Per FPN level
16     # Detectron1 uses 2000 proposals per-batch,
17     # (See "modeling/rpn/rpn_outputs.py" for details of this legacy issue)
18     # which is approximately 1000 proposals per-image since the default batch size is 2
19     POST_NMS_TOPK_TRAIN: 1000
20     POST_NMS_TOPK_TEST: 1000
21    ROI_HEADS:
22     NAME: "StandardROIHeads"
23     IN_FEATURES: ["p2", "p3", "p4", "p5"]
24    ROI_BOX_HEAD:
25     NAME: "FastRCNNConvFCHead"
26     NUM_FC: 2
27     POOLER_RESOLUTION: 7
28    ROI_MASK_HEAD:
29     NAME: "MaskRCNNConvUpsampleHead"
30     NUM_CONV: 4
31     POOLER_RESOLUTION: 14
32    DATASETS:
33     TRAIN: ("coco_2017_train",)
34     TEST: ("coco_2017_val",)
35    SOLVER:
36     IMS_PER_BATCH: 16
37     BASE_LR: 0.02
38     STEPS: (60000, 80000)
39     MAX_ITER: 90000
40    INPUT:
41     MIN_SIZE_TRAIN: (640, 672, 704, 736, 768, 800)
42    VERSION: 2
```





## Faster R-CNN代码架构解读：配置文件

```
faster_rcnn_R_50_FPN_1x.yaml
1  _BASE_: "../Base-RCNN-FPN.yaml"
2  MODEL:
3    WEIGHTS: "detectron2://ImageNetPretrained/MSRA/R-50.pkl"
4    MASK_ON: False
5    RESNETS:
6      DEPTH: 50
```

- faster\_rcnn\_R\_50\_FPN\_1x.yaml
- 基础网络是ImageNet预训练的ResNet50，加上FPN
- 初始检测层：res2、res3、res4、res5
- 最终检测层：P2、 P3、 P4、 P5、 **P6**

```
Base-RCNN-FPN.yaml
1  MODEL:
2    META_ARCHITECTURE: "GeneralizedRCNN"
3    BACKBONE:
4      NAME: "build_resnet_fpn_backbone"
5    RESNETS:
6      OUT_FEATURES: ["res2", "res3", "res4", "res5"]
7    FPN:
8      IN_FEATURES: ["res2", "res3", "res4", "res5"]
9    ANCHOR_GENERATOR:
10     SIZES: [[32], [64], [128], [256], [512]] # One size for each in feature map
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32    DATASETS:
33     TRAIN: ("coco_2017_train",)
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35    SOLVER:
36     IMS_PER_BATCH: 16
37     BASE_LR: 0.02
38     STEPS: (60000, 80000)
39     MAX_ITER: 90000
40    INPUT:
41     MIN_SIZE_TRAIN: (640, 672, 704, 736, 768, 800)
42    VERSION: 2
```





## Faster R-CNN代码架构解读：配置文件

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faster_rcnn_R_50_FPN_1x.yaml
1  _BASE_: "../Base-RCNN-FPN.yaml"
2  MODEL:
3    WEIGHTS: "detectron2://ImageNetPretrained/MSRA/R-50.pkl"
4    MASK_ON: False
5    RESNETS:
6      DEPTH: 50
```

- faster\_rcnn\_R\_50\_FPN\_1x.yaml
- 基础网络是ImageNet预训练的ResNet50，加上FPN
- 初始检测层：res2、res3、res4、res5
- 最终检测层：P2、P3、P4、P5、P6
- 检测层stride：4、8、16、32、64
- 锚框大小：32、64、128、256、512
- 锚框比例：每层都是[0.5, 1.0, 2.0]

```
Base-RCNN-FPN.yaml
1  MODEL:
2    META_ARCHITECTURE: "GeneralizedRCNN"
3    BACKBONE:
4      NAME: "build_resnet_fpn_backbone"
5    RESNETS:
6      OUT_FEATURES: ["res2", "res3", "res4", "res5"]
7    FPN:
8      IN_FEATURES: ["res2", "res3", "res4", "res5"]
9    ANCHOR_GENERATOR:
10     SIZES: [[32], [64], [128], [256], [512]] # One size for each in feature map
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13      IN_FEATURES: ["p2", "p3", "p4", "p5", "p6"]
14      PRE_NMS_TOPK_TRAIN: 2000 # Per FPN level
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18      # which is approximately 1000 proposals per-image since the default batch size is 2.
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20      POST_NMS_TOPK_TEST: 1000
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31      POOLER_RESOLUTION: 14
32    DATASETS:
33      TRAIN: ("coco_2017_train",)
34      TEST: ("coco_2017_val",)
35    SOLVER:
36      IMS_PER_BATCH: 16
37      BASE_LR: 0.02
38      STEPS: (60000, 80000)
39      MAX_ITER: 90000
40    INPUT:
41      MIN_SIZE_TRAIN: (640, 672, 704, 736, 768, 800)
42    VERSION: 2
```





# Faster R-CNN代码架构解读：配置文件

```
faster_rcnn_R_50_FPN_1x.yaml
1 _BASE_: "../Base-RCNN-FPN.yaml"
2 MODEL:
3   WEIGHTS: "detectron2://ImageNetPretrained/MSRA/R-50.pkl"
4   MASK_ON: False
5   RESNETS:
6     DEPTH: 50
```

- faster\_rcnn\_R\_50\_FPN\_1x.yaml
- 基础网络是ImageNet预训练的ResNet50，加上FPN
- 初始检测层：res2、res3、res4、res5
- 最终检测层：P2、P3、P4、P5、P6
- 检测层stride：4、8、16、32、64
- 锚框大小：32、64、128、256、512
- 锚框比例：每层都是[0.5, 1.0, 2.0]

```
Base-RCNN-FPN.yaml
1 MODEL:
2   META_ARCHITECTURE: "GeneralizedRCNN"
3   BACKBONE:
4     NAME: "build_resnet_fpn_backbone"
5   RESNETS:
6     OUT_FEATURES: ["res2", "res3", "res4", "res5"]
7   FPN:
8     IN_FEATURES: ["res2", "res3", "res4", "res5"]
9   ANCHOR_GENERATOR:
10    SIZES: [[32], [64], [128], [256], [512]] # One size for each in feature map
11    ASPECT RATIOS: [[0.5, 1.0, 2.0]] # Three aspect ratios (same for all feature maps)
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16    # Detectron1 uses 2000 proposals per-batch,
17    # (See "modeling/rpn/rpn_outputs.py" for details of this issue,
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24  ROI_BOX_HEAD:
25    NAME: "FastRCNNConvFCHead"
26    NUM_FC: 2
27    POOLER_RESOLUTION: 7
28  ROI_MASK_HEAD:
29    NAME: "MaskRCNNConvUpsampleHead"
30    NUM_CONV: 4
31    POOLER_RESOLUTION: 14
32  DATASETS:
33    TRAIN: ("coco_2017_train",)
34    TEST: ("coco_2017_val",)
35  SOLVER:
36    IMS_PER_BATCH: 16
37    BASE_LR: 0.02
38    STEPS: (60000, 80000)
39    MAX_ITER: 90000
40  INPUT:
41    MIN_SIZE_TRAIN: (640, 672, 704, 736, 768, 800)
42  VERSION: 2
```

RPN生成  
候选区域  
的参数

Fast RCNN  
的参数

训练和测试  
的数据

训练参数

多尺度训练





## Faster R-CNN代码架构解读：算法流程

### 构建

- 模型构建
- 优化器构建
- 数据构建



### 训练

- 数据处理
- 特征提取
- RPN训练
- RPN生成候选区域
- Fast R-CNN训练



### 测试

- 数据处理
- 特征提取
- RPN生成候选区域
- Fast R-CNN输出结果





## Faster R-CNN代码架构解读：算法流程

### 构建

- 模型构建
- 优化器构建
- 数据构建





## Faster R-CNN代码架构解读：构建

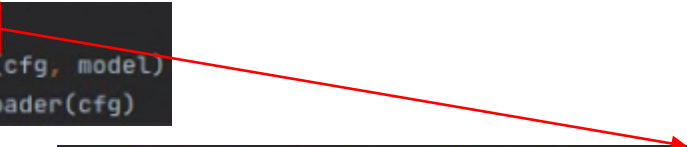
```
model = self.build_model(cfg)
optimizer = self.build_optimizer(cfg, model)
data_loader = self.build_train_loader(cfg)
```





## Faster R-CNN代码架构解读：模型构建

```
model = self.build_model(cfg)
optimizer = self.build_optimizer(cfg, model)
data_loader = self.build_train_loader(cfg)
```



```
self.backbone = build_backbone(cfg)
self.proposal_generator = build_proposal_generator(cfg, self.backbone.output_shape())
self.roi_heads = build_roi_heads(cfg, self.backbone.output_shape())
```







## Faster R-CNN代码架构解读：模型构建

```
model = self.build_model(cfg)
optimizer = self.build_optimizer(cfg, model)
data_loader = self.build_train_loader(cfg)
```

```
self.backbone = build_backbone(cfg)
self.proposal_generator = build_proposal_generator(cfg, self.backbone.output_shape())
self.roi_heads = build_roi_heads(cfg, self.backbone.output_shape())
```

```
def build_resnet_fpn_backbone(cfg, input_shape: ShapeSpec):
    """
    Args:
        cfg: a detectron2 CfgNode

    Returns:
        backbone (Backbone): backbone module, must be a subclass of :class:`Backbone`
    """
    bottom_up = build_resnet_backbone(cfg, input_shape)
    in_features = cfg.MODEL.FPN.IN_FEATURES
    out_channels = cfg.MODEL.FPN.OUT_CHANNELS
    backbone = FPN(
        bottom_up=bottom_up,
        in_features=in_features,
        out_channels=out_channels,
        norm=cfg.MODEL.FPN.NORM,
        top_block=LastLevelMaxPool(),
        fuse_type=cfg.MODEL.FPN.FUSE_TYPE,
    )
    return backbone
```

ResNet构建

FPN构建





## Faster R-CNN代码架构解读：模型构建

```
model = self.build_model(cfg)
optimizer = self.build_optimizer(cfg, model)
data_loader = self.build_train_loader(cfg)
```

```
self.backbone = build_backbone(cfg)
self.proposal_generator = build_proposal_generator(cfg, self.backbone.output_shape())
self.roi_heads = build_roi_heads(cfg, self.backbone.output_shape())
```

```
def build_resnet_fpn_backbone(cfg, input_shape: ShapeSpec):
    """
    Args:
        cfg: a detectron2 CfgNode

    Returns:
        backbone (Backbone): backbone module, must be a subclass of :class:`Backbone`
    """
    bottom_up = build_resnet_backbone(cfg, input_shape)
    in_features = cfg.MODEL.FPN.IN_FEATURES
    out_channels = cfg.MODEL.FPN.OUT_CHANNELS
    backbone = FPN(
        bottom_up=bottom_up,
        in_features=in_features,
        out_channels=out_channels,
        norm=cfg.MODEL.FPN.NORM,
        top_block=LastLevelMaxPool(),
        fuse_type=cfg.MODEL.FPN.FUSE_TYPE,
    )
    return backbone
```

```
self.anchor_generator = build_anchor_generator(
    cfg, [input_shape[f] for f in self.in_features])
self.box2box_transform = Box2BoxTransform(weights=cfg.MODEL.RPN.BBOX_REG_LOSS_TYPE)
self.anchor_matcher = Matcher(
    cfg.MODEL.RPN.IOU_THRESHOLDS, cfg.MODEL.RPN.IOU_LABELS, allow_low_quality_matches=True)
self.rpn_head = build_rpn_head(cfg, [input_shape[f] for f in self.in_features])
```

锚框的构建

锚框变换的构建

锚框匹配的构建

锚框预测的构建





## Faster R-CNN代码架构解读：模型构建

```
model = self.build_model(cfg)
optimizer = self.build_optimizer(cfg, model)
data_loader = self.build_train_loader(cfg)
```

```
self.backbone = build_backbone(cfg)
self.proposal_generator = build_proposal_generator(cfg, self.backbone.output_shape())
self.roi_heads = build_roi_heads(cfg, self.backbone.output_shape())
```

```
def build_resnet_fpn_backbone(cfg, input_shape: ShapeSpec):
    """
    Args:
        cfg: a detectron2 CfgNode

    Returns:
        backbone (Backbone): backbone module, must be a subclass of :class:`Backbone`
    """
    bottom_up = build_resnet_backbone(cfg, input_shape)
    in_features = cfg.MODEL.FPN.IN_FEATURES
    out_channels = cfg.MODEL.FPN.OUT_CHANNELS
    backbone = FPN(
        bottom_up=bottom_up,
        in_features=in_features,
        out_channels=out_channels,
        norm=cfg.MODEL.FPN.NORM,
        top_block=LastLevelMaxPool(),
        fuse_type=cfg.MODEL.FPN.FUSE_TYPE,
    )
    return backbone
```

```
self.anchor_generator = build_anchor_generator(
    cfg, [input_shape[f] for f in self.in_features]
)
self.box2box_transform = Box2BoxTransform(weights=cfg.MODEL.RPN.BBOX_REG_WEIGHTS)
self.anchor_matcher = Matcher(
    cfg.MODEL.RPN.IOU_THRESHOLDS, cfg.MODEL.RPN.IOU_LABELS, allow_low_quality_matches=True
)
self.rpn_head = build_rpn_head(cfg, [input_shape[f] for f in self.in_features])
```

```
self.box_pooler = box_pooler RoI Pooling构建
self.box_head = box_head 特征增强网络构建
self.box_predictor = box_predictor 预测模块构建
```





## Faster R-CNN代码架构解读：优化器构建

```
model = self.build_model(cfg)
optimizer = self.build_optimizer(cfg, model)
data_loader = self.build_train_loader(cfg)
```

```
self.scheduler = self.build_lr_scheduler(cfg, optimizer)
# Assume no other objects need to be checkpointed.
# We can later make it checkpoint the stateful hooks
self.checkpointer = DetectionCheckpointer(
    # Assume you want to save checkpoints together with logs/statistics
    model,
    cfg.OUTPUT_DIR,
    optimizer=optimizer,
    scheduler=self.scheduler,
)

self.start_iter = 0
self.max_iter = cfg.SOLVER.MAX_ITER
self.cfg = cfg
```

学习率策略

训练模型保存

训练迭代次数



## Faster R-CNN代码架构解读：数据构建

```
model = self.build_model(cfg)
optimizer = self.build_optimizer(cfg, model)
data_loader = self.build_train_loader(cfg)
```

```
dataset_dicts = get_detection_dataset_dicts(
    cfg.DATASETS.TRAIN,
    filter_empty=cfg.DATALOADER.FILTER_EMPTY_ANNOTATIONS,
    min_keypoints=cfg.MODEL.ROI_KEYPOINT_HEAD.MIN_KEYPOINTS_PER_IMAGE
    if cfg.MODEL.KEYPOINT_ON
    else 0,
    proposal_files=cfg.DATASETS.PROPOSAL_FILES_TRAIN if cfg.MODEL.LOAD_PROPOSALS else None,
)

dataset = DatasetFromList(dataset_dicts, copy=False)

if mapper is None:
    mapper = DatasetMapper(cfg, True)
dataset = MapDataset(dataset, mapper)

sampler_name = cfg.DATALOADER.SAMPLER_TRAIN
logger = logging.getLogger(__name__)
logger.info("Using training sampler {}".format(sampler_name))
# TODO avoid if-else?
if sampler_name == "TrainingSampler":
    sampler = samplers.TrainingSampler(len(dataset))
elif sampler_name == "RepeatFactorTrainingSampler":
    sampler = samplers.RepeatFactorTrainingSampler(
        dataset_dicts, cfg.DATALOADER.REPEAT_THRESHOLD
    )
else:
    raise ValueError("Unknown training sampler: {}".format(sampler_name))
return build_batch_data_loader(
    dataset,
    sampler,
    cfg.SOLVER.IMS_PER_BATCH,
    aspect_ratio_grouping=cfg.DATALOADER.ASPECT_RATIO_GROUPING,
    num_workers=cfg.DATALOADER.NUM_WORKERS,
```





## Faster R-CNN代码架构解读：算法流程

### 构建

- 模型构建
- 优化器构建
- 数据构建



### 训练

- 数据处理
- 特征提取
- RPN训练
- RPN生成候选区域
- Fast R-CNN训练





## Faster R-CNN代码架构解读：数据处理

读图像

```
image = utils.read_image(dataset_dict["file_name"], format=self.img_format)
```

图像变换  
(缩放、水平翻转)

```
image, transforms = T.apply_transform_gens(self.tfm_gens, image)
```

标注变换

```
utils.transform_instance_annotations(  
    obj, transforms, image_shape, keypoint_hflip_indices=self.keypoint_hflip_indices
```

过滤标注

```
dataset_dict["instances"] = utils.filter_empty_instances(instances)
```

图像归一化

```
images = [(x - self.pixel_mean) / self.pixel_std for x in images]
```

图像组batch

```
images = ImageList.from_tensors(images, self.backbone.size_divisibility)
```







## Faster R-CNN代码架构解读：特征提取

```
bottom_up_features = self.bottom_up(x)
```

→ ResNet特征提取

```
x = [bottom_up_features[t] for t in self.in_features[::-1]]
```

```
results = []
```

```
prev_features = self.lateral_convs[0](x[0])
```

```
results.append(self.output_convs[0](prev_features))
```

```
for features, lateral_conv, output_conv in zip(
```

```
    x[1:], self.lateral_convs[1:], self.output_convs[1:]
```

```
):
```

```
    top_down_features = F.interpolate(prev_features, scale_factor=2, mode="nearest")
```

```
    lateral_features = lateral_conv(features)
```

```
    prev_features = lateral_features + top_down_features
```

```
    if self.fuse_type == "avg":
```

```
        prev_features /= 2
```

```
    results.insert(0, output_conv(prev_features))
```

→ FPN特征提取

```
if self.top_block is not None:
```

```
    top_block_in_feature = bottom_up_features.get(self.top_block.in_feature, None)
```

```
    if top_block_in_feature is None:
```

```
        top_block_in_feature = results[self._out_features.index(self.top_block.in_feature)]
```

```
    results.extend(self.top_block(top_block_in_feature))
```

```
assert len(self._out_features) == len(results)
```

```
return dict(zip(self._out_features, results))
```







## Faster R-CNN代码架构解读：RPN训练和候选区域生成

```
anchors = self.anchor_generator(features)
```

生成锚框

```
pred_objectness_logits, pred_anchor_deltas = self.rpn_head(features)
```

预测锚框的  
类别和偏移

```
gt_labels, gt_boxes = self.label_and_sample_anchors(anchors, gt_instances)
```

计算锚框类别和  
偏移的真值

```
losses = self.losses(  
    anchors, pred_objectness_logits, gt_labels, pred_anchor_deltas, gt_boxes  
)
```

计算分类和  
回归的误差损失

```
proposals = self.predict_proposals(  
    anchors, pred_objectness_logits, pred_anchor_deltas, images.image_sizes  
)
```

RPN生成  
候选区域





## Faster R-CNN代码架构解读：Fast R-CNN训练

```
proposals = add_ground_truth_to_proposals(gt_boxes, proposals)
```

把真实标注框  
加入候选区域

```
proposals = self.label_and_sample_proposals(proposals, targets)
```

候选区域的分类和回  
归的真值计算

```
box_features = self.box_pooler(features, [x.proposal_boxes for x in proposals])
```

RoIPooling

```
box_features = self.box_head(box_features)
```

候选区域  
特征增强

```
predictions = self.box_predictor(box_features)
```

候选区域分类和  
回归的预测

```
F.cross_entropy(self.pred_class_logits, self.gt_classes, reduction="mean")
```

分类损失函数

```
loss_box_reg = smooth_l1_loss(  
    self.pred_proposal_deltas[fg_inds[:, None], gt_class_cols],  
    gt_proposal_deltas[fg_inds],  
    self.smooth_l1_beta,  
    reduction="sum",  
)
```

回归损失函数





# 请观看演示视频

```
--config-file  
./configs/COCO-Detection/faster_rcnn_R_50_FPN_1x.yaml  
--num-gpus  
1  
SOLVER.IMS_PER_BATCH  
1  
INPUT.MIN_SIZE_TRAIN  
(400,)  
DATASETS.TRAIN  
('coco_2017_val',)  
DATALOADER.NUM_WORKERS  
0
```





## 课程作业

### ■ 单步调试Faster R-CNN代码

1. 利用配好的detectron2物体检测平台，使用PyCharm软件，单步调试如下配置

([https://github.com/facebookresearch/detectron2/blob/master/configs/COCO-Detection/faster\\_rcnn\\_R\\_50\\_FPN\\_1x.yaml](https://github.com/facebookresearch/detectron2/blob/master/configs/COCO-Detection/faster_rcnn_R_50_FPN_1x.yaml))的Faster R-CNN代码

2. 把Faster R-CNN中每个细节与代码对应上，真正弄懂Faster R-CNN的整个流程
  3. (可选) 如果硬件条件允许，可以使用8卡GPU训一个模型，看精度是否与官方一致
- \* 大家一定多花些时间仔细调试Faster R-CNN的代码，有不懂的可以在群里问或者查阅相关资料，Faster R-CNN的代码、原理、细节等都搞明白之后，后续的物体检测算法都很容易上手





结语

感谢各位聆听!

Thanks for Listening

