



17级种子班

数字图像处理课设

Week5-CNN总结



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CONTENT

●01

整体网络结构

●02

数据预处理及增强

●03

实验结果

整体网络结构

使用框架：TensorFlow2.1

整体网络层数：

算上batch_normalization层及池化层共30层

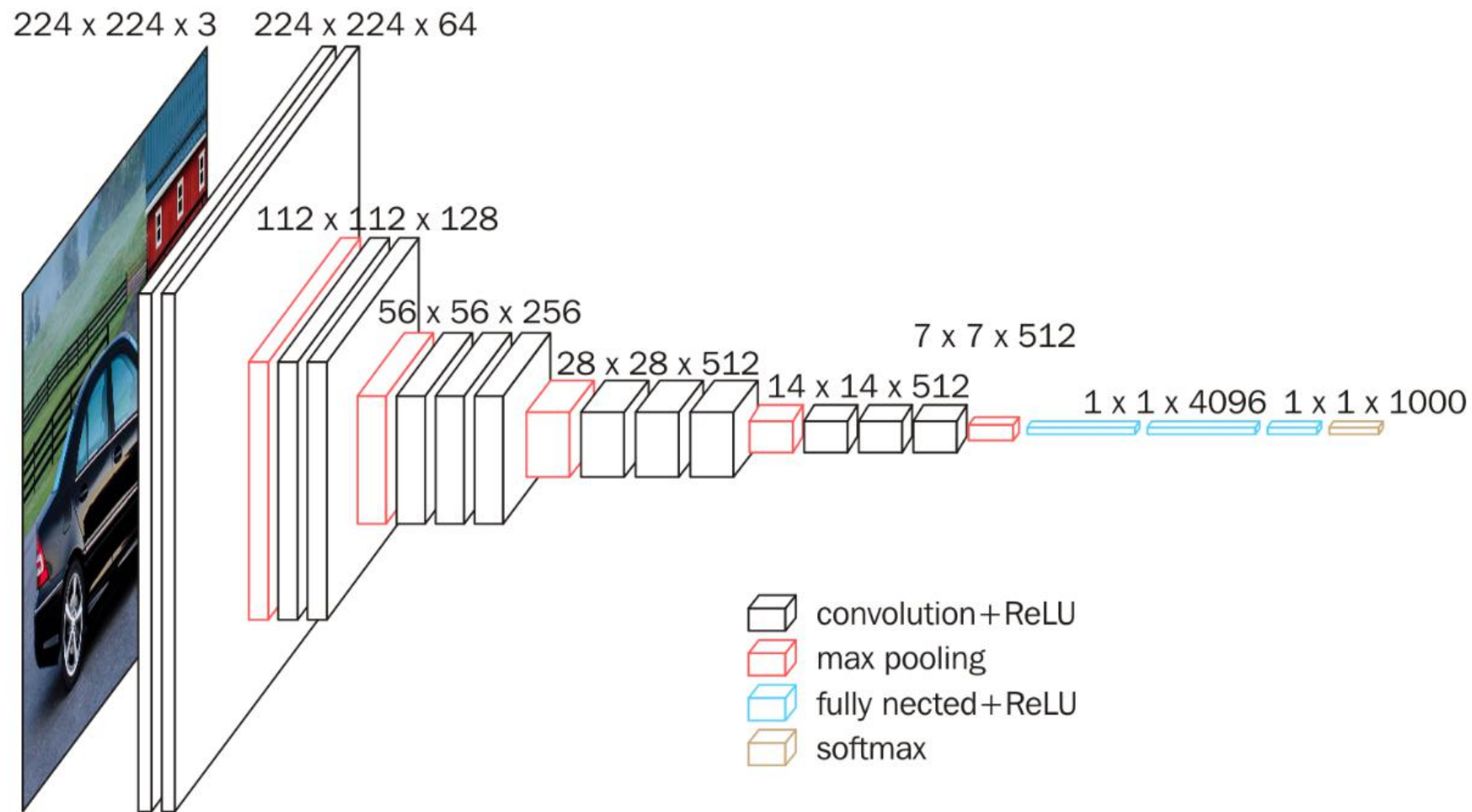
不算batch_normalization层及池化层共13层

初始化：Xavier
损失函数：crossentropy
激活函数：Relu
优化器：Adam

Model: "sequential"		
Layer (type)	Output Shape	Param #
batch_normalization (BatchNo	(None, 32, 32, 3)	12
conv2d (Conv2D)	(None, 32, 32, 64)	1792
batch_normalization_1 (Batch	(None, 32, 32, 64)	256
conv2d_1 (Conv2D)	(None, 32, 32, 64)	36928
batch_normalization_2 (Batch	(None, 32, 32, 64)	256
conv2d_2 (Conv2D)	(None, 32, 32, 64)	36928
max_pooling2d (MaxPooling2D)	(None, 16, 16, 64)	0
batch_normalization_3 (Batch	(None, 16, 16, 64)	256
conv2d_3 (Conv2D)	(None, 16, 16, 128)	73856
batch_normalization_4 (Batch	(None, 16, 16, 128)	512
conv2d_4 (Conv2D)	(None, 16, 16, 128)	147584
max_pooling2d_1 (MaxPooling2	(None, 8, 8, 128)	0
batch_normalization_5 (Batch	(None, 8, 8, 128)	512
conv2d_5 (Conv2D)	(None, 8, 8, 256)	295168
batch_normalization_6 (Batch	(None, 8, 8, 256)	1024
conv2d_6 (Conv2D)	(None, 8, 8, 256)	590080
batch_normalization_7 (Batch	(None, 8, 8, 256)	1024
conv2d_7 (Conv2D)	(None, 8, 8, 256)	590080
max_pooling2d_2 (MaxPooling2	(None, 4, 4, 256)	0
batch_normalization_8 (Batch	(None, 4, 4, 256)	1024
conv2d_8 (Conv2D)	(None, 4, 4, 512)	1180160
batch_normalization_9 (Batch	(None, 4, 4, 512)	2048
conv2d_9 (Conv2D)	(None, 4, 4, 512)	2359808
max_pooling2d_3 (MaxPooling2	(None, 2, 2, 512)	0
batch_normalization_10 (Batc	(None, 2, 2, 512)	2048
flatten (Flatten)	(None, 2048)	0
batch_normalization_11 (Batc	(None, 2048)	8192
dense (Dense)	(None, 512)	1049088
batch_normalization_12 (Batc	(None, 512)	2048
dense_1 (Dense)	(None, 10)	5130
Total params: 6,385,814		
Trainable params: 6,376,208		
Non-trainable params: 9,606		

整体网络结构

参考网络结构：VGG16



整体网络结构

参考网络结构：VGG16



通道数翻倍:

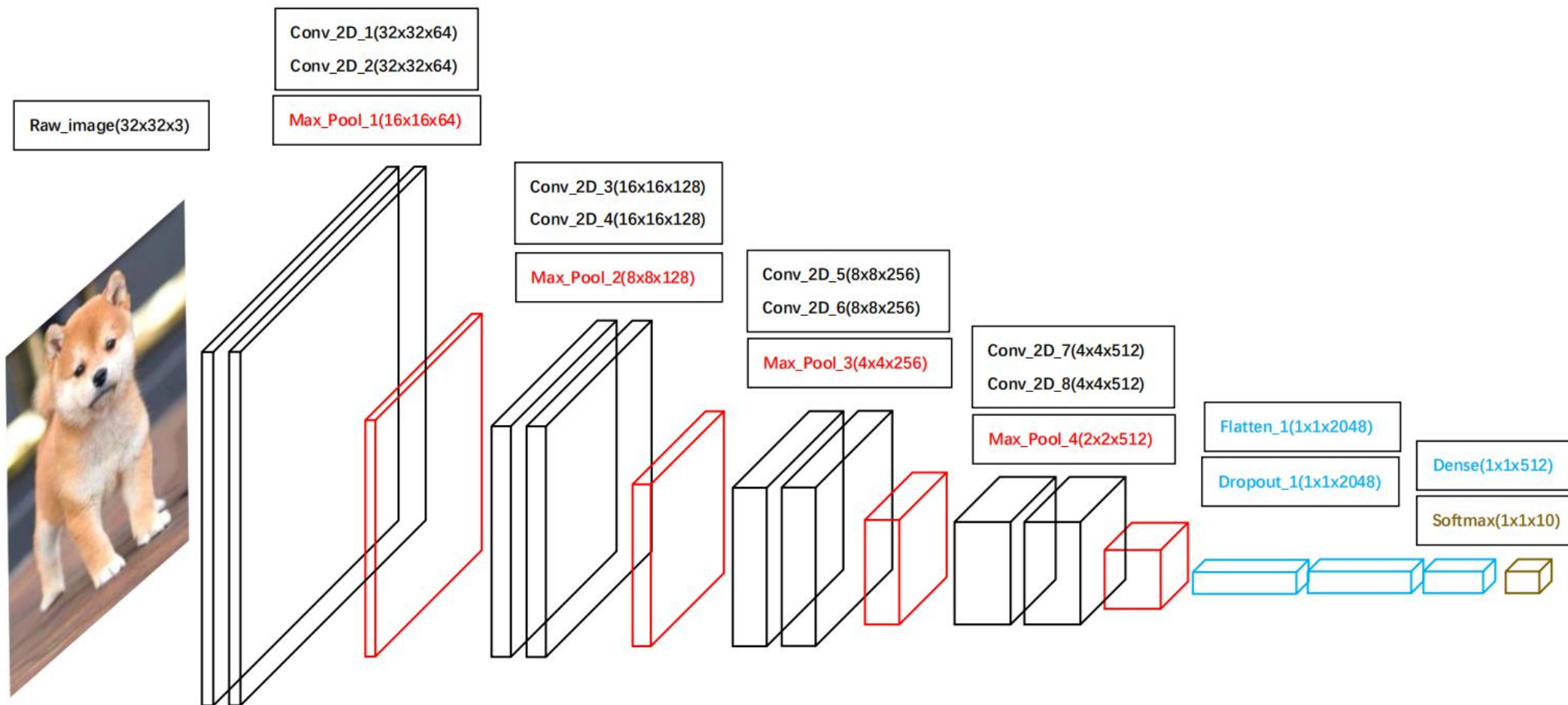
64 \rightarrow 128 \rightarrow 256 \rightarrow 512 \rightarrow 512 \rightarrow 512

高和宽变减半(池化层):

224 \rightarrow 112 \rightarrow 56 \rightarrow 28 \rightarrow 14 \rightarrow 7

整体网络结构

我们的网络结构



整体网络结构

加入Dropout层(0.25, 0.5)

Model: "sequential"

Layer (type)	Output Shape	Param #
conv2d (Conv2D)	(None, 32, 32, 64)	1792
conv2d_1 (Conv2D)	(None, 32, 32, 64)	36928
max_pooling2d (MaxPooling2D)	(None, 16, 16, 64)	0
dropout (Dropout)	(None, 16, 16, 64)	0
conv2d_2 (Conv2D)	(None, 16, 16, 128)	73856
conv2d_3 (Conv2D)	(None, 16, 16, 128)	147584
max_pooling2d_1 (MaxPooling2D)	(None, 8, 8, 128)	0
dropout_1 (Dropout)	(None, 8, 8, 128)	0
conv2d_4 (Conv2D)	(None, 8, 8, 256)	295168
conv2d_5 (Conv2D)	(None, 8, 8, 256)	590080
max_pooling2d_2 (MaxPooling2D)	(None, 4, 4, 256)	0
dropout_2 (Dropout)	(None, 4, 4, 256)	0
conv2d_6 (Conv2D)	(None, 4, 4, 512)	1180160
conv2d_7 (Conv2D)	(None, 4, 4, 512)	2359808
max_pooling2d_3 (MaxPooling2D)	(None, 2, 2, 512)	0
dropout_3 (Dropout)	(None, 2, 2, 512)	0
flatten (Flatten)	(None, 2048)	0
dropout_4 (Dropout)	(None, 2048)	0
dense (Dense)	(None, 512)	1049088
dense_1 (Dense)	(None, 10)	5130

Total params: 5,739,594
Trainable params: 5,739,594
Non-trainable params: 0

0.25

0.5

整体网络结构

Dropout层修改(0.1-0.5)

Model: "sequential"

Layer (type)	Output Shape	Param #
conv2d (Conv2D)	(None, 32, 32, 64)	1792
conv2d_1 (Conv2D)	(None, 32, 32, 64)	36928
max_pooling2d (MaxPooling2D)	(None, 16, 16, 64)	0
dropout (Dropout)	(None, 16, 16, 64)	0
conv2d_2 (Conv2D)	(None, 16, 16, 128)	73856
conv2d_3 (Conv2D)	(None, 16, 16, 128)	147584
max_pooling2d_1 (MaxPooling2D)	(None, 8, 8, 128)	0
dropout_1 (Dropout)	(None, 8, 8, 128)	0
conv2d_4 (Conv2D)	(None, 8, 8, 256)	295168
conv2d_5 (Conv2D)	(None, 8, 8, 256)	590080
max_pooling2d_2 (MaxPooling2D)	(None, 4, 4, 256)	0
dropout_2 (Dropout)	(None, 4, 4, 256)	0
conv2d_6 (Conv2D)	(None, 4, 4, 512)	1180160
conv2d_7 (Conv2D)	(None, 4, 4, 512)	2359808
max_pooling2d_3 (MaxPooling2D)	(None, 2, 2, 512)	0
dropout_3 (Dropout)	(None, 2, 2, 512)	0
flatten (Flatten)	(None, 2048)	0
dropout_4 (Dropout)	(None, 2048)	0
dense (Dense)	(None, 512)	1049088
dense_1 (Dense)	(None, 10)	5130
Total params: 5,739,594		
Trainable params: 5,739,594		
Non-trainable params: 0		

0.1

0.2

0.3

0.4

0.5

整体网络结构

batch_normalization层取代Dropout层

Model: "sequential"

Layer (type)	Output Shape	Param #
batch_normalization (Batch Normalization)	(None, 32, 32, 3)	12
conv2d (Conv2D)	(None, 32, 32, 64)	1792
batch_normalization_1 (Batch Normalization)	(None, 32, 32, 64)	256
conv2d_1 (Conv2D)	(None, 32, 32, 64)	36928
max_pooling2d (MaxPooling2D)	(None, 16, 16, 64)	0
batch_normalization_2 (Batch Normalization)	(None, 16, 16, 64)	256
conv2d_2 (Conv2D)	(None, 16, 16, 128)	73856
batch_normalization_3 (Batch Normalization)	(None, 16, 16, 128)	512
conv2d_3 (Conv2D)	(None, 16, 16, 128)	147584
max_pooling2d_1 (MaxPooling2D)	(None, 8, 8, 128)	0
batch_normalization_4 (Batch Normalization)	(None, 8, 8, 128)	512
conv2d_4 (Conv2D)	(None, 8, 8, 256)	295168
batch_normalization_5 (Batch Normalization)	(None, 8, 8, 256)	1024
conv2d_5 (Conv2D)	(None, 8, 8, 256)	590080
max_pooling2d_2 (MaxPooling2D)	(None, 4, 4, 256)	0
batch_normalization_6 (Batch Normalization)	(None, 4, 4, 256)	1024
conv2d_6 (Conv2D)	(None, 4, 4, 512)	1180160
batch_normalization_7 (Batch Normalization)	(None, 4, 4, 512)	2048
conv2d_7 (Conv2D)	(None, 4, 4, 512)	2359808
max_pooling2d_3 (MaxPooling2D)	(None, 2, 2, 512)	0
batch_normalization_8 (Batch Normalization)	(None, 2, 2, 512)	2048
flatten (Flatten)	(None, 2048)	0
batch_normalization_9 (Batch Normalization)	(None, 2048)	8192
dense (Dense)	(None, 512)	1049088
batch_normalization_10 (Batch Normalization)	(None, 512)	2048
dense_1 (Dense)	(None, 10)	5130

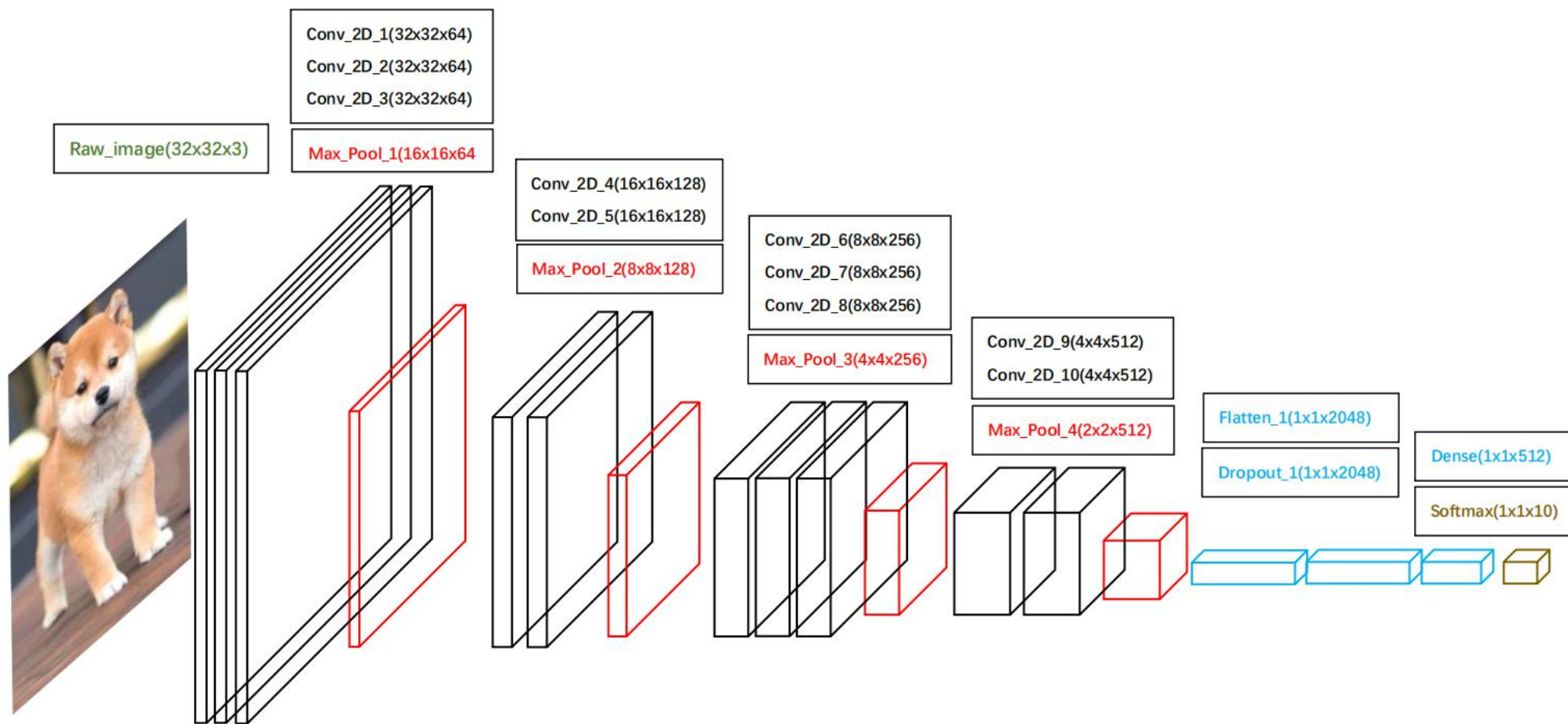
Total params: 5,757,526

Trainable params: 5,748,560

Non-trainable params: 8,966

整体网络结构

网络深度增加



数据预处理及增强

使用ImageDataGenerator

```
train_datagen = keras.preprocessing.image.ImageDataGenerator(  
    featurewise_center=True,           #使输入数据集去中心化(均值为0)  
    featurewise_std_normalization=True, #将输入除以数据集的标准差以完成标准化  
    rotation_range=30,                 #图片随机转动的角度  
    width_shift_range=0.12,            #图片随机水平偏移的幅度  
    height_shift_range=0.12,           #图片随机竖直偏移的幅度  
    zoom_range=0.12,                   #随机缩放的幅度  
    rescale=1./255,                    #重缩放因子(在其他变换前)  
    horizontal_flip=True)              #随机水平翻转  
  
test_datagen = keras.preprocessing.image.ImageDataGenerator(rescale=1./255)
```

实验结果

超参数搜索：

epochs = 30

batch_size_list = [32, 64, 128, 256, 512]

lr_list = [0.0001, 0.0002, 0.0005, 0.001, 0.002, 0.005]

搜索结果：

batch_size 越大，每代训练的训练时间越短，但缩短到一定程度就不再下降

batch_size 越大，模型收敛的越慢，直至不能收敛

batch_size 越大，过拟合会越晚体现出来

lr对结果影响不大，较小的lr表现更好

最终采用：

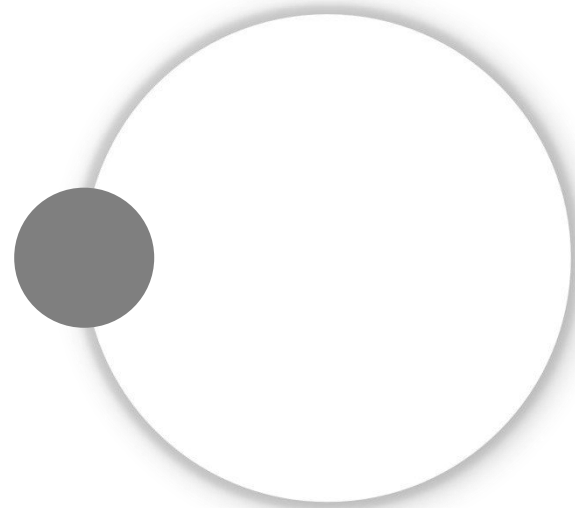
batch_size = 32 lr = 0.0001 epochs = 30/100

acc = 91.2%



实验总结

- CNN在图像识别方面的效果不错
- 在实验中深刻感受到了DeepLearning的玄学及不可解释性
- 对深度学习研究的困惑



Question & Answer

