白迎辰_2018201670_深度学习第三次作业

In [18]:

```
import numpy as np import matplotlib.pyplot as plt

import tensorflow as tf from keras import Model from keras.layers import Dense,Flatten,Input,InputLayer from keras.optimizers import SGD,RMSprop,Adam from keras.models import Sequential # 导入models模块中的Sequential容器

from keras.utils import to_categorical
```

In [2]:

```
## 载入mnist数据集
mnist = tf.keras.datasets.mnist
(X_train, y_train), (X_test, y_test) = mnist.load_data()

## 转换为one - hot型向量
Y_train=to_categorical(y_train)
Y_test=to_categorical(y_test)

print(Y_train.shape)
print(Y_train[0])
```

(60000, 10) [0. 0. 0. 0. 0. 1. 0. 0. 0. 0.]

In [3]:

```
实验:构建 Multi-layer Nueral Network 模型
## 第一步 创建模型结构 ##
IMSIZE = 28
                                # MNIST图像为28*28的单层图片
input_layer = Input([IMSIZE,IMSIZE])
x = input_layer
                             # 将28*28*1的Tensor拉直为784维向量
x = Flatten()(input layer)
x = Dense(1000,activation = 'relu')(x) # 全连接到1000个节点,并采用relu激活函数
x = Dense(10,activation = 'softmax')(x) # 全连接到10个节点,并采用softmax激活函数转化为(0,1)取值
output_layer=x
model=Model(input_layer,output_layer) # Model函数将input_layer 和 output_layer中间的部分连接起来
model.summary()
## 第二步 模型编译 ##
model.compile(loss='categorical_crossentropy',optimizer=SGD(lr=0.001),metrics=['accuracy'])
## 第三步 模型拟合 ##
```

```
history1 = model.fit(X_train, Y_train, validation_data=(X_test, Y_test), batch_size=1000, epochs=50)
#第四部 提取loss指标
# model.fit会返回一个history对象,里面记录了训练集和测试集的loss以及acc
# 我们将这些指标取出,绘制折线图
train_loss1 = history1.history["loss"]
Model: "model 1"
Layer (type) Output Shape
                 Param #
______
input_1 (InputLayer) (None, 28, 28)
                  0
flatten_1 (Flatten)
         (None, 784) 0
dense 1 (Dense)
         (None, 1000)
                  785000
dense_2 (Dense) (None, 10)
                  10010
______
Total params: 795,010
Trainable params: 795,010
Non-trainable params: 0
Train on 60000 samples, validate on 10000 samples
Epoch 1/50
s: 2.5986 - val_accuracy: 0.8942
Epoch 2/50
1.8495 - val accuracy: 0.9095
Epoch 3/50
1.5483 - val accuracy: 0.9181
Epoch 4/50
1.3622 - val accuracy: 0.9249
Epoch 5/50
1.2538 - val_accuracy: 0.9293
Epoch 6/50
1.1744 - val accuracy: 0.9314
Epoch 7/50
1.1088 - val_accuracy: 0.9325
Epoch 8/50
1.0637 - val_accuracy: 0.9351
Epoch 9/50
1.0316 - val accuracy: 0.9354
Epoch 10/50
1.0036 - val accuracy: 0.9354
Epoch 11/50
0.9753 - val accuracy: 0.9367
Epoch 12/50
0.9592 - val accuracy: 0.9369
```

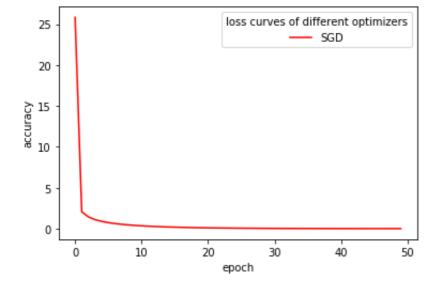
Epoch 13/50

```
60000/60000 [===============] - 1s 9us/step - loss: 0.2656 - accuracy: 0.9736 - val_loss:
0.9378 - val accuracy: 0.9357
Epoch 14/50
0.9219 - val_accuracy: 0.9363
Epoch 15/50
0.9154 - val_accuracy: 0.9355
Epoch 16/50
s: 0.9017 - val_accuracy: 0.9376
Epoch 17/50
s: 0.8939 - val accuracy: 0.9361
Epoch 18/50
0.8825 - val_accuracy: 0.9378
Epoch 19/50
0.8761 - val_accuracy: 0.9371
Epoch 20/50
0.8757 - val_accuracy: 0.9367
Epoch 21/50
0.8672 - val_accuracy: 0.9381
Epoch 22/50
s: 0.8637 - val accuracy: 0.9382
Epoch 23/50
0.8577 - val_accuracy: 0.9379
Epoch 24/50
0.8504 - val accuracy: 0.9385
Epoch 25/50
0.8491 - val accuracy: 0.9380
Epoch 26/50
0.8468 - val accuracy: 0.9380
Epoch 27/50
0.8497 - val_accuracy: 0.9377
Epoch 28/50
0.8417 - val_accuracy: 0.9379
Epoch 29/50
0.8400 - val accuracy: 0.9378
Epoch 30/50
0.8342 - val_accuracy: 0.9383
Epoch 31/50
0.8338 - val_accuracy: 0.9388
Epoch 32/50
0.8349 - val_accuracy: 0.9392
Epoch 33/50
0.8305 - val. accuracy: 0.9392
```

```
Epoch 34/50
0.8307 - val accuracy: 0.9398
Epoch 35/50
0.8273 - val accuracy: 0.9393
Epoch 36/50
0.8264 - val accuracy: 0.9394
Epoch 37/50
0.8260 - val accuracy: 0.9387
Epoch 38/50
0.8222 - val accuracy: 0.9393
Epoch 39/50
0.8223 - val_accuracy: 0.9393
Epoch 40/50
0.8221 - val_accuracy: 0.9394
Epoch 41/50
0.8212 - val accuracy: 0.9393
Epoch 42/50
0.8192 - val_accuracy: 0.9394
Epoch 43/50
0.8177 - val accuracy: 0.9395
Epoch 44/50
0.8190 - val_accuracy: 0.9399
Epoch 45/50
0.8150 - val accuracy: 0.9399
Epoch 46/50
0.8148 - val accuracy: 0.9396
Epoch 47/50
0.8154 - val accuracy: 0.9397
Epoch 48/50
0.8133 - val accuracy: 0.9399
Epoch 49/50
0.8126 - val accuracy: 0.9398
Epoch 50/50
0.8115 - val accuracy: 0.9401
```

In [4]:

```
x = np.arange(50) # 生成0:49的连续整数代表epoch y = train_loss1 # 将history对象的loss属性命名为y plt.plot(x,y,color='red',label='SGD') # 绘制x-y散点图并用红色平滑曲线连接,图例为SGD plt.legend(title='loss curves of different optimizers') # 设置图例标题为不同优化器下的损失曲线 plt.xlabel('epoch') # 设置横轴表示epoch plt.ylabel('accuracy') # 设置纵轴表示accuracy plt.show() # 绘制图像
```



In [5]:

```
optim_list = ['SGD','RMSprop','Adagrad','Adadelta','Adam','Adamax','Nadam'] # 新建一个列表来存储所有优化 ▲
器的名称
color list = ['red','orange','yellow','green','blue','purple','grey'] # 新建一个列表来储存不同优化器对应损失曲线
的颜色
for i in range(7): #生成0:6的连续整数代表7种优化器的下标
 print('Optimizer:') # 输出当前模型使用的优化器名称
 print(optim list[i])
 IMSIZE = 28 # 设置输入图像的长宽为28像素
 model = Sequential([InputLayer([IMSIZE,IMSIZE]),Flatten(),Dense(1000,activation='relu'),Dense(10,activati
on='softmax')]) # 使用sequential容器配置训练模型
 model.summary() # 输出模型信息
 model.compile(loss='categorical crossentropy',optimizer=optim list[i],metrics=['accuracy']) # 模型编译,损
失函数为交叉熵,优化器为优化器列表中的下标为i的优化器,性能度量指标为accuracy准确率
trained model = model.fit(X train, Y train, validation data=(X test, Y test), batch size=1000, epochs=50) #
训练模型,每批1000张图,一共50批
 x = np.arange(50) # 生成0:49 的连续整数代表epoch
 y = trained model.history["loss"] # 将history对象的loss属性命名为y
plt.plot(x,y,color=color_list[i],label=optim_list[i])# 绘制当前优化器下的损失曲线,并设置对应的图例
plt.legend(title='loss curves of different optimizers') # 设定图片标题
plt.ylim((0,1))
plt.xlabel('epoch') # 设置横坐标为epoch
plt.ylabel('accuracy') # 设置纵坐标为accuracy
plt.show() # 绘制图像
```

Optimizer:

SGD

Model: "sequential_1"

Layer (type)	Output Shape	Param #
flatten_2 (Flatten)	(None, 784)	0
dense_3 (Dense)	(None, 1000)	785000
dense_4 (Dense)	(None, 10)	10010

Total params: 795,010 Trainable params: 795,010 Non-trainable params: 0

Train on 60000 samples validate on 10000 samples

```
Epoch 1/50
ss: 0.6862 - val accuracy: 0.8235
Epoch 2/50
0.4978 - val_accuracy: 0.8761
Epoch 3/50
0.3678 - val accuracy: 0.8987
Epoch 4/50
0.2968 - val accuracy: 0.9171
Epoch 5/50
0.3213 - val_accuracy: 0.9117
Epoch 6/50
0.2921 - val_accuracy: 0.9201
Epoch 7/50
0.2454 - val_accuracy: 0.9320
Epoch 8/50
0.2326 - val accuracy: 0.9354
Epoch 9/50
0.2257 - val_accuracy: 0.9387
Epoch 10/50
0.2206 - val accuracy: 0.9397
Epoch 11/50
0.2132 - val accuracy: 0.9416
Epoch 12/50
0.2091 - val accuracy: 0.9436
Epoch 13/50
0.2058 - val accuracy: 0.9480
Epoch 14/50
s: 0.2092 - val accuracy: 0.9439
Epoch 15/50
s: 0.1954 - val accuracy: 0.9485
Epoch 16/50
s: 0.2624 - val accuracy: 0.9296
Epoch 17/50
s: 0.1967 - val_accuracy: 0.9487
Epoch 18/50
s: 0.1899 - val_accuracy: 0.9503
Epoch 19/50
s: 0.1959 - val accuracy: 0.9506
Epoch 20/50
s: 0.2042 - val accuracy: 0.9468
Epoch 21/50
               1 10 10 up/stop | loop: 0 1069 | copurable 0 0675 | vol. loop
e^{0000/e0000}
```

```
s: 0.1910 - val accuracy: 0.9512
Epoch 22/50
s: 0.1865 - val_accuracy: 0.9531
Epoch 23/50
s: 0.1899 - val_accuracy: 0.9524
Epoch 24/50
s: 0.1878 - val_accuracy: 0.9539
Epoch 25/50
s: 0.1862 - val_accuracy: 0.9553
Epoch 26/50
s: 0.1959 - val accuracy: 0.9517
Epoch 27/50
0.1850 - val accuracy: 0.9566
Epoch 28/50
0.1844 - val_accuracy: 0.9555
Epoch 29/50
s: 0.1823 - val accuracy: 0.9563
Epoch 30/50
s: 0.1843 - val_accuracy: 0.9548
Epoch 31/50
s: 0.1911 - val accuracy: 0.9536
Epoch 32/50
s: 0.1834 - val_accuracy: 0.9552
Epoch 33/50
s: 0.1842 - val_accuracy: 0.9558
Epoch 34/50
s: 0.1832 - val accuracy: 0.9547
Epoch 35/50
s: 0.1832 - val_accuracy: 0.9561
Epoch 36/50
s: 0.1838 - val_accuracy: 0.9555
Epoch 37/50
s: 0.1848 - val_accuracy: 0.9562
Epoch 38/50
s: 0.1834 - val accuracy: 0.9566
Epoch 39/50
s: 0.1827 - val_accuracy: 0.9575
Epoch 40/50
s: 0.1825 - val_accuracy: 0.9565
Epoch 41/50
s: 0.1864 - val accuracy: 0.9576
```

```
EDUCII 42/50
s: 0.1840 - val_accuracy: 0.9566
Epoch 43/50
0.1861 - val_accuracy: 0.9586
Epoch 44/50
0.1814 - val_accuracy: 0.9594
Epoch 45/50
0.1830 - val_accuracy: 0.9572
Epoch 46/50
0.1853 - val accuracy: 0.9590
Epoch 47/50
0.1876 - val accuracy: 0.9570
Epoch 48/50
0.1818 - val accuracy: 0.9594
Epoch 49/50
0.1868 - val accuracy: 0.9601
Epoch 50/50
0.1837 - val_accuracy: 0.9594
Optimizer:
RMSprop
Model: "sequential_2"
Layer (type) Output Shape Param #
______
flatten_3 (Flatten) (None, 784) 0
dense_5 (Dense) (None, 1000) 785000
dense_6 (Dense) (None, 10) 10010
_____
Total params: 795,010
Trainable params: 795,010
Non-trainable params: 0
Train on 60000 samples, validate on 10000 samples
Epoch 1/50
s: 5.7740 - val_accuracy: 0.8655
Epoch 2/50
0.8965 - val_accuracy: 0.9566
Epoch 3/50
s: 0.7946 - val_accuracy: 0.9536
Epoch 4/50
```

s: 0.5843 - val accuracy: 0.9603

s: 0.4999 - val accuracy: 0.9698

Epoch 5/50

```
Epoch //50
s: 0.5392 - val accuracy: 0.9689
Epoch 8/50
0.5838 - val_accuracy: 0.9708
Epoch 9/50
0.6996 - val_accuracy: 0.9682
Epoch 10/50
1.1087 - val_accuracy: 0.9538
Epoch 11/50
0.7261 - val_accuracy: 0.9717
Epoch 12/50
s: 0.5916 - val_accuracy: 0.9741
Epoch 13/50
0.6339 - val accuracy: 0.9731
Epoch 14/50
s: 0.6572 - val accuracy: 0.9753
Epoch 15/50
s: 0.6110 - val_accuracy: 0.9764
Epoch 16/50
s: 0.7481 - val_accuracy: 0.9751
Epoch 17/50
0.6473 - val accuracy: 0.9765
Epoch 18/50
0.7424 - val accuracy: 0.9753
Epoch 19/50
s: 0.8345 - val accuracy: 0.9747
Epoch 20/50
s: 0.7422 - val_accuracy: 0.9773
Epoch 21/50
0.8816 - val_accuracy: 0.9726
Epoch 22/50
0.8109 - val accuracy: 0.9760
Epoch 23/50
0.7151 - val accuracy: 0.9788
Epoch 24/50
0.8573 - val_accuracy: 0.9778
Epoch 25/50
0.7429 - val accuracy: 0.9796
Epoch 26/50
s: 0.8046 - val_accuracy: 0.9779
Epoch 27/50
```

```
0.7927 - val_accuracy: 0.9794
Epoch 28/50
0.8404 - val_accuracy: 0.9786
Epoch 29/50
0.7879 - val_accuracy: 0.9803
Epoch 30/50
0.7560 - val accuracy: 0.9810
Epoch 31/50
0.8784 - val accuracy: 0.9796
Epoch 32/50
0.8396 - val_accuracy: 0.9792
Epoch 33/50
s: 0.8777 - val accuracy: 0.9799
Epoch 34/50
s: 0.9000 - val accuracy: 0.9803
Epoch 35/50
0.9287 - val accuracy: 0.9789
Epoch 36/50
0.9804 - val accuracy: 0.9782
Epoch 37/50
0.9415 - val accuracy: 0.9798
Epoch 38/50
0.9183 - val_accuracy: 0.9797
Epoch 39/50
0.9111 - val accuracy: 0.9820
Epoch 40/50
1.0465 - val accuracy: 0.9772
Epoch 41/50
1.3392 - val_accuracy: 0.9725
Epoch 42/50
1.1656 - val accuracy: 0.9770
Epoch 43/50
s: 1.0539 - val_accuracy: 0.9779
Epoch 44/50
s: 1.1126 - val accuracy: 0.9790
Epoch 45/50
s: 0.9085 - val accuracy: 0.9803
Epoch 46/50
s: 1.2239 - val accuracy: 0.9765
Epoch 47/50
s: 1.1020 - val_accuracy: 0.9798
Epoch 48/50
```

```
1.0508 - val accuracy: 0.9814
Epoch 49/50
1.0498 - val accuracy: 0.9800
Epoch 50/50
1.0733 - val_accuracy: 0.9785
Optimizer:
Adagrad
Model: "sequential 3"
Layer (type)
       Output Shape Param #
flatten_4 (Flatten) (None, 784)
dense 7 (Dense)
        (None, 1000) 785000
dense_8 (Dense) (None, 10) 10010
_____
Total params: 795,010
Trainable params: 795,010
Non-trainable params: 0
Train on 60000 samples, validate on 10000 samples
Epoch 1/50
s: 3.7914 - val accuracy: 0.9213
Epoch 2/50
2.2328 - val accuracy: 0.9287
Epoch 3/50
1.5186 - val accuracy: 0.9429
Epoch 4/50
s: 1.2905 - val accuracy: 0.9448
Epoch 5/50
1.1208 - val_accuracy: 0.9475
Epoch 6/50
1.0737 - val accuracy: 0.9486
Epoch 7/50
0.9939 - val accuracy: 0.9508
Epoch 8/50
0.9768 - val accuracy: 0.9500
Epoch 9/50
0.9820 - val accuracy: 0.9491
Epoch 10/50
s: 0.8391 - val accuracy: 0.9556
Epoch 11/50
s: 0.8312 - val_accuracy: 0.9565
Epoch 12/50
0.8104 - val_accuracy: 0.9573
```

Epoch 13/50

```
s: 0.8557 - val_accuracy: 0.9524
Epoch 14/50
0.7833 - val_accuracy: 0.9564
Epoch 15/50
s: 0.7831 - val_accuracy: 0.9556
Epoch 16/50
0.7640 - val_accuracy: 0.9573
Epoch 17/50
0.7615 - val_accuracy: 0.9567
Epoch 18/50
s: 0.7508 - val_accuracy: 0.9573
Epoch 19/50
60000/60000 [==============] - 1s 11us/step - loss: 0.0457 - accuracy: 0.9938 - val_los
s: 0.7374 - val accuracy: 0.9571
Epoch 20/50
s: 0.7366 - val accuracy: 0.9568
Epoch 21/50
s: 0.7216 - val_accuracy: 0.9561
Epoch 22/50
s: 0.7217 - val accuracy: 0.9574
Epoch 23/50
s: 0.7202 - val_accuracy: 0.9576
Epoch 24/50
s: 0.7099 - val accuracy: 0.9579
Epoch 25/50
s: 0.7229 - val accuracy: 0.9570
Epoch 26/50
s: 0.7143 - val_accuracy: 0.9580
Epoch 27/50
s: 0.7277 - val_accuracy: 0.9549
Epoch 28/50
s: 0.7052 - val accuracy: 0.9573
Epoch 29/50
s: 0.7165 - val accuracy: 0.9568
Epoch 30/50
s: 0.7067 - val_accuracy: 0.9566
Epoch 31/50
s: 0.7034 - val accuracy: 0.9583
Epoch 32/50
s: 0.7087 - val_accuracy: 0.9568
Epoch 33/50
0.7027 - val accuracy: 0.9577
```

```
Epoch 34/50
0.7024 - val accuracy: 0.9578
Epoch 35/50
s: 0.7038 - val_accuracy: 0.9573
Epoch 36/50
s: 0.7028 - val accuracy: 0.9580
Epoch 37/50
s: 0.7029 - val accuracy: 0.9577
Epoch 38/50
s: 0.7022 - val accuracy: 0.9574
Epoch 39/50
s: 0.7035 - val_accuracy: 0.9571
Epoch 40/50
s: 0.7055 - val_accuracy: 0.9573
Epoch 41/50
s: 0.7005 - val accuracy: 0.9580
Epoch 42/50
s: 0.7044 - val_accuracy: 0.9569
Epoch 43/50
s: 0.7018 - val_accuracy: 0.9577
Epoch 44/50
loss: 0.7054 - val_accuracy: 0.9573
Epoch 45/50
loss: 0.7048 - val accuracy: 0.9566
Epoch 46/50
loss: 0.7004 - val_accuracy: 0.9576
Epoch 47/50
loss: 0.7007 - val accuracy: 0.9576
Epoch 48/50
oss: 0.7003 - val accuracy: 0.9575
Epoch 49/50
oss: 0.7004 - val_accuracy: 0.9576
Epoch 50/50
loss: 0.7011 - val_accuracy: 0.9578
Optimizer:
Adadelta
Model: "sequential 4"
Layer (type)
       Output Shape
              Param #
flatten_5 (Flatten)
       (None, 784)
```

dense_9 (Dense)

(None, 1000)

dense 10 (Dense) (None 10)

785000

10010

Total params: 795,010 Trainable params: 795,010 Non-trainable params: 0 Train on 60000 samples, validate on 10000 samples Epoch 1/50 s: 8.1005 - val_accuracy: 0.8608 Epoch 2/50 1.5753 - val_accuracy: 0.9492 Epoch 3/50 s: 3.1414 - val accuracy: 0.9017 Epoch 4/50 s: 0.9695 - val_accuracy: 0.9636 Epoch 5/50 0.9718 - val_accuracy: 0.9565 Epoch 6/50 s: 0.8210 - val accuracy: 0.9602 Epoch 7/50 60000/60000 [===============] - 1s 10us/step - loss: 0.5538 - accuracy: 0.9764 - val_los s: 0.6200 - val accuracy: 0.9673 Epoch 8/50 s: 0.5801 - val_accuracy: 0.9711 Epoch 9/50 s: 0.5146 - val accuracy: 0.9722 Epoch 10/50 s: 0.4688 - val_accuracy: 0.9735 Epoch 11/50 s: 0.4883 - val_accuracy: 0.9725 Epoch 12/50 s: 0.5079 - val accuracy: 0.9723 Epoch 13/50 s: 0.4494 - val_accuracy: 0.9744 Epoch 14/50 s: 0.5182 - val_accuracy: 0.9751 Epoch 15/50 s: 0.4195 - val_accuracy: 0.9758 Epoch 16/50 s: 0.4697 - val_accuracy: 0.9736 Epoch 17/50 s: 0.4955 - val accuracy: 0.9735 Epoch 18/50 s: 0.5393 - val_accuracy: 0.9728 Epoch 19/50

__1 1c 12uc/ctop loce: 0.0220 accuracy: 0.0059 val loc

 $=1.0000/60000 I_{--}$

```
--| 00000/00000 |
s: 0.5560 - val_accuracy: 0.9716
Epoch 20/50
s: 0.5161 - val_accuracy: 0.9768
Epoch 21/50
s: 0.5281 - val accuracy: 0.9755
Epoch 22/50
s: 0.5113 - val_accuracy: 0.9760
Epoch 23/50
s: 0.5459 - val_accuracy: 0.9749
Epoch 24/50
s: 0.4790 - val_accuracy: 0.9780
Epoch 25/50
s: 0.5384 - val_accuracy: 0.9770
Epoch 26/50
s: 0.5418 - val_accuracy: 0.9770
Epoch 27/50
s: 0.4966 - val accuracy: 0.9782
Epoch 28/50
s: 0.5215 - val accuracy: 0.9788
Epoch 29/50
0.5199 - val_accuracy: 0.9779
Epoch 30/50
0.6079 - val accuracy: 0.9738
Epoch 31/50
s: 0.5273 - val_accuracy: 0.9767
Epoch 32/50
s: 0.5011 - val accuracy: 0.9799
Epoch 33/50
s: 0.5141 - val_accuracy: 0.9789
Epoch 34/50
s: 0.5388 - val_accuracy: 0.9777
Epoch 35/50
s: 0.6140 - val accuracy: 0.9770
Epoch 36/50
s: 0.4927 - val_accuracy: 0.9796
Epoch 37/50
s: 0.5490 - val_accuracy: 0.9778
Epoch 38/50
s: 0.5913 - val accuracy: 0.9767
Epoch 39/50
s: 0.4937 - val_accuracy: 0.9789
```

```
s: 0.5282 - val_accuracy: 0.9812
Epoch 41/50
s: 0.5557 - val_accuracy: 0.9790
Epoch 42/50
s: 0.5164 - val accuracy: 0.9784
Epoch 43/50
0.5315 - val accuracy: 0.9798
Epoch 44/50
0.5254 - val_accuracy: 0.9794
Epoch 45/50
s: 0.5646 - val_accuracy: 0.9791
Epoch 46/50
s: 0.5115 - val accuracy: 0.9800
Epoch 47/50
s: 0.5442 - val accuracy: 0.9798
Epoch 48/50
s: 0.5115 - val_accuracy: 0.9810
Epoch 49/50
s: 0.5313 - val_accuracy: 0.9797
Epoch 50/50
0.5219 - val accuracy: 0.9809
Optimizer:
Adam
Model: "sequential 5"
Layer (type) Output Shape Param #
_____
flatten 6 (Flatten) (None, 784) 0
dense_11 (Dense)
       (None, 1000) 785000
dense_12 (Dense) (None, 10) 10010
Total params: 795,010
Trainable params: 795,010
Non-trainable params: 0
Train on 60000 samples, validate on 10000 samples
Epoch 1/50
s: 2.9179 - val accuracy: 0.9319
Epoch 2/50
1.4070 - val_accuracy: 0.9466
Epoch 3/50
s: 1.0741 - val_accuracy: 0.9521
Epoch 4/50
s: 0.9146 - val_accuracy: 0.9539
```

Epoch 40/30

```
Epoch 5/50
s: 0.8119 - val accuracy: 0.9577
Epoch 6/50
s: 0.7619 - val accuracy: 0.9607
Epoch 7/50
s: 0.7128 - val accuracy: 0.9646
Epoch 8/50
s: 0.7085 - val_accuracy: 0.9615
Epoch 9/50
s: 0.7737 - val_accuracy: 0.9624
Epoch 10/50
0.6888 - val_accuracy: 0.9649
Epoch 11/50
s: 0.7269 - val_accuracy: 0.9635
Epoch 12/50
s: 0.7246 - val_accuracy: 0.9650
Epoch 13/50
s: 0.6914 - val accuracy: 0.9651
Epoch 14/50
s: 0.7690 - val_accuracy: 0.9643
Epoch 15/50
s: 0.7368 - val_accuracy: 0.9653
Epoch 16/50
s: 0.6810 - val_accuracy: 0.9664
Epoch 17/50
s: 0.6963 - val accuracy: 0.9661
Epoch 18/50
s: 0.7146 - val_accuracy: 0.9675
Epoch 19/50
s: 0.7408 - val_accuracy: 0.9658
Epoch 20/50
s: 0.7480 - val accuracy: 0.9655
Epoch 21/50
s: 0.7109 - val_accuracy: 0.9664
Epoch 22/50
s: 0.7811 - val_accuracy: 0.9666
Epoch 23/50
0.7773 - val accuracy: 0.9666
Epoch 24/50
0.7664 - val_accuracy: 0.9684
Epoch 25/50
```

```
0.7604 - val_accuracy: 0.9672
Epoch 26/50
0.7110 - val accuracy: 0.9707
Epoch 27/50
0.7508 - val_accuracy: 0.9678
Epoch 28/50
s: 0.8248 - val_accuracy: 0.9697
Epoch 29/50
0.7281 - val accuracy: 0.9692
Epoch 30/50
s: 0.6960 - val accuracy: 0.9709
Epoch 31/50
s: 0.7387 - val_accuracy: 0.9721
Epoch 32/50
0.7564 - val_accuracy: 0.9727
Epoch 33/50
0.7740 - val_accuracy: 0.9713
Epoch 34/50
s: 0.7451 - val_accuracy: 0.9713
Epoch 35/50
s: 0.7864 - val accuracy: 0.9719
Epoch 36/50
s: 0.7195 - val_accuracy: 0.9700
Epoch 37/50
s: 0.7436 - val accuracy: 0.9726
Epoch 38/50
s: 0.6843 - val_accuracy: 0.9737
Epoch 39/50
s: 0.7222 - val_accuracy: 0.9720
Epoch 40/50
s: 0.7086 - val_accuracy: 0.9753
Epoch 41/50
0.7089 - val accuracy: 0.9735
Epoch 42/50
0.7707 - val accuracy: 0.9712
Epoch 43/50
s: 0.7301 - val accuracy: 0.9739
Epoch 44/50
s: 0.7107 - val_accuracy: 0.9750
Epoch 45/50
s: 0.7106 - val accuracy: 0.9741
Epoch 46/50
```

```
s: 0.8351 - val accuracy: 0.9705
Epoch 47/50
s: 0.7328 - val_accuracy: 0.9740
Epoch 48/50
s: 0.7592 - val_accuracy: 0.9745
Epoch 49/50
0.6937 - val accuracy: 0.9743
Epoch 50/50
0.8517 - val accuracy: 0.9694
Optimizer:
Adamax
Model: "sequential_6"
Layer (type)
       Output Shape Param #
flatten_7 (Flatten)
        (None, 784)
               0
dense_13 (Dense)
         (None, 1000)
                785000
dense_14 (Dense) (None, 10)
                10010
_____
Total params: 795,010
Trainable params: 795,010
Non-trainable params: 0
Train on 60000 samples, validate on 10000 samples
Epoch 1/50
s: 4.8528 - val_accuracy: 0.9223
Epoch 2/50
2.6672 - val accuracy: 0.9376
Epoch 3/50
1.9301 - val accuracy: 0.9442
Epoch 4/50
1.6077 - val accuracy: 0.9473
Epoch 5/50
1.3715 - val_accuracy: 0.9499
Epoch 6/50
1.1693 - val accuracy: 0.9533
Epoch 7/50
1.0628 - val accuracy: 0.9556
Epoch 8/50
0.9966 - val accuracy: 0.9557
Epoch 9/50
0.9581 - val_accuracy: 0.9560
Epoch 10/50
s: 0.9175 - val accuracy: 0.9566
```

Epoch 11/50

```
0.8782 - val_accuracy: 0.9568
Epoch 12/50
0.8753 - val accuracy: 0.9559
Epoch 13/50
s: 0.8689 - val_accuracy: 0.9560
Epoch 14/50
s: 0.8472 - val accuracy: 0.9574
Epoch 15/50
s: 0.8385 - val accuracy: 0.9583
Epoch 16/50
s: 0.8087 - val accuracy: 0.9582
Epoch 17/50
s: 0.8172 - val_accuracy: 0.9578
Epoch 18/50
s: 0.8084 - val_accuracy: 0.9586
Epoch 19/50
s: 0.8057 - val accuracy: 0.9598
Epoch 20/50
s: 0.8002 - val_accuracy: 0.9600
Epoch 21/50
s: 0.7857 - val accuracy: 0.9601
Epoch 22/50
s: 0.8062 - val_accuracy: 0.9586
Epoch 23/50
s: 0.7907 - val_accuracy: 0.9596
Epoch 24/50
loss: 0.7882 - val_accuracy: 0.9608
Epoch 25/50
loss: 0.7860 - val accuracy: 0.9601
Epoch 26/50
loss: 0.7876 - val_accuracy: 0.9600
Epoch 27/50
loss: 0.7860 - val accuracy: 0.9604
Epoch 28/50
oss: 0.7898 - val_accuracy: 0.9601
Epoch 29/50
oss: 0.7856 - val_accuracy: 0.9605
Epoch 30/50
oss: 0.7869 - val_accuracy: 0.9606
Epoch 31/50
oss: 0.7860 - val accuracy: 0.9604
```

```
Epoch 32/50
oss: 0.7864 - val accuracy: 0.9607
Epoch 33/50
oss: 0.7859 - val_accuracy: 0.9605
Epoch 34/50
oss: 0.7856 - val_accuracy: 0.9607
Epoch 35/50
oss: 0.7857 - val accuracy: 0.9605
Epoch 36/50
oss: 0.7857 - val accuracy: 0.9604
Epoch 37/50
oss: 0.7859 - val_accuracy: 0.9604
Epoch 38/50
oss: 0.7852 - val_accuracy: 0.9604
Epoch 39/50
oss: 0.7854 - val_accuracy: 0.9604
Epoch 40/50
oss: 0.7852 - val accuracy: 0.9606
Epoch 41/50
oss: 0.7852 - val accuracy: 0.9605
Epoch 42/50
oss: 0.7848 - val_accuracy: 0.9605
Epoch 43/50
oss: 0.7847 - val_accuracy: 0.9606
Epoch 44/50
loss: 0.7849 - val_accuracy: 0.9606
Epoch 45/50
loss: 0.7847 - val accuracy: 0.9606
Epoch 46/50
oss: 0.7848 - val accuracy: 0.9607
Epoch 47/50
oss: 0.7848 - val accuracy: 0.9606
Epoch 48/50
oss: 0.7841 - val accuracy: 0.9608
Epoch 49/50
oss: 0.7843 - val accuracy: 0.9606
Epoch 50/50
oss: 0.7842 - val_accuracy: 0.9605
Optimizer:
Nadam
Model: "sequential 7"
```

Laver (type) Output Shape Param #

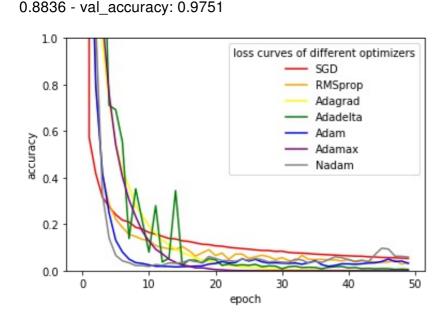
```
______
flatten_8 (Flatten) (None, 784) 0
dense 15 (Dense) (None, 1000) 785000
dense_16 (Dense) (None, 10) 10010
______
Total params: 795,010
Trainable params: 795,010
Non-trainable params: 0
Train on 60000 samples, validate on 10000 samples
Epoch 1/50
s: 2.1557 - val accuracy: 0.9373
Epoch 2/50
1.4074 - val_accuracy: 0.9462
Epoch 3/50
s: 1.1759 - val accuracy: 0.9531
Epoch 4/50
s: 0.6907 - val accuracy: 0.9628
Epoch 5/50
0.5747 - val_accuracy: 0.9693
Epoch 6/50
0.6096 - val accuracy: 0.9686
Epoch 7/50
0.6306 - val_accuracy: 0.9671
Epoch 8/50
0.6069 - val_accuracy: 0.9683
Epoch 9/50
0.6243 - val_accuracy: 0.9691
Epoch 10/50
0.6369 - val accuracy: 0.9711
Epoch 11/50
0.6736 - val accuracy: 0.9664
Epoch 12/50
0.5899 - val_accuracy: 0.9713
Epoch 13/50
0.5925 - val_accuracy: 0.9708
Epoch 14/50
0.6231 - val_accuracy: 0.9687
Epoch 15/50
0.6032 - val_accuracy: 0.9709
Epoch 16/50
0.6412 - val accuracy: 0.9705
Epoch 17/50
```

60000/60000 [-----0.000 0.000

```
0.6221 - val_accuracy: 0.9724
Epoch 18/50
0.6241 - val accuracy: 0.9707
Epoch 19/50
0.6430 - val_accuracy: 0.9709
Epoch 20/50
0.7044 - val accuracy: 0.9716
Epoch 21/50
0.6400 - val_accuracy: 0.9708
Epoch 22/50
0.6576 - val_accuracy: 0.9729
Epoch 23/50
0.5967 - val_accuracy: 0.9742
Epoch 24/50
0.6807 - val accuracy: 0.9733
Epoch 25/50
0.7000 - val accuracy: 0.9735
Epoch 26/50
0.6769 - val_accuracy: 0.9745
Epoch 27/50
0.6583 - val_accuracy: 0.9744
Epoch 28/50
0.6210 - val accuracy: 0.9744
Epoch 29/50
0.7284 - val accuracy: 0.9728
Epoch 30/50
0.6707 - val accuracy: 0.9725
Epoch 31/50
0.6618 - val accuracy: 0.9751
Epoch 32/50
60000/60000 [==============] - 1s 9us/step - loss: 0.0356 - accuracy: 0.9960 - val_loss:
0.6382 - val_accuracy: 0.9766
Epoch 33/50
s: 0.9157 - val accuracy: 0.9676
Epoch 34/50
s: 0.6474 - val_accuracy: 0.9779
Epoch 35/50
s: 0.6614 - val_accuracy: 0.9767
Epoch 36/50
s: 0.6453 - val_accuracy: 0.9753
Epoch 37/50
0.7098 - val_accuracy: 0.9754
```

Enach 20/50

```
LP0011 30/30
0.7396 - val accuracy: 0.9727
Epoch 39/50
0.7679 - val_accuracy: 0.9742
Epoch 40/50
0.6952 - val accuracy: 0.9759
Epoch 41/50
0.7414 - val accuracy: 0.9728
Epoch 42/50
0.6957 - val accuracy: 0.9761
Epoch 43/50
60000/60000 [==============] - 0s 8us/step - loss: 0.0450 - accuracy: 0.9961 - val_loss:
0.7169 - val_accuracy: 0.9761
Epoch 44/50
0.7498 - val accuracy: 0.9759
Epoch 45/50
0.9646 - val accuracy: 0.9728
Epoch 46/50
1.0710 - val accuracy: 0.9653
Epoch 47/50
0.7609 - val accuracy: 0.9760
Epoch 48/50
0.8329 - val accuracy: 0.9740
Epoch 49/50
0.7712 - val accuracy: 0.9778
Epoch 50/50
```



深度学习优化算法大概经历了 SGD -> SGDM -> NAG -> Adagrad -> Adadelta(RMSprop) -> Adam -> Nadam 这样的发展历程。

在keras.optimizers子模块中,它们基本上都有对应的类的实现。

In [17]:

import keras

• SGD, 默认参数为纯SGD, 设置momentum参数不为0实际上变成SGDM, 考虑了一阶动量, 设置 nesterov为 True后变成NAG, 即 Nesterov Accelerated Gradient, 在计算梯度时计算的是向前走一步所在位置的梯度。

In [10]:

keras.optimizers.SGD(lr=0.01, momentum=0.0, decay=0.0, nesterov=False)

Out[10]:

<keras.optimizers.SGD at 0x7ff3b56bebd0>

Adagrad,考虑了二阶动量,对于不同的参数有不同的学习率,即自适应学习率。缺点是学习率单调下降,可能后期学习速率过慢乃至提前停止学习。

In [11]:

keras.optimizers.Adagrad(lr=0.01, epsilon=None, decay=0.0)

Out[11]:

<keras.optimizers.Adagrad at 0x7ff3b56bec10>

• RMSprop, 考虑了二阶动量,对于不同的参数有不同的学习率,即自适应学习率,对Adagrad进行了优化,通过指数平滑只考虑一定窗口内的二阶动量。

In [12]:

keras.optimizers.RMSprop(lr=0.001, rho=0.9, epsilon=**None**, decay=0.0)



Out[12]:

<keras.optimizers.RMSprop at 0x7ff3b558d290>

• Adadelta, 考虑了二阶动量,与RMSprop类似,但是更加复杂一些,自适应性更强。

In [13]:

keras.optimizers.Adadelta(lr=1.0, rho=0.95, epsilon=None, decay=0.0)



Out[13]:

<keras.optimizers.Adadelta at 0x7ff3b558db90>

• Adam,同时考虑了一阶动量和二阶动量,可以看成RMSprop上进一步考虑了一阶动量。

In [14]:

keras.optimizers.Adam(lr=0.001, beta_1=0.9, beta_2=0.999, epsilon=**None**, decay=0.0, amsgrad=**False**)



Out[14]:

<keras.optimizers.Adam at 0x7ff3b558e350>

• Adamax, 它是Adam算法基于无穷范数(infinity norm)的变种。

In [15]:

keras.optimizers.Adamax(lr=0.002, beta_1=0.9, beta_2=0.999, epsilon=**None**, decay=0.0)

Out[15]:

<keras.optimizers.Adamax at 0x7ff3b558eb90>

• Nadam, 在Adam基础上进一步考虑了 Nesterov Acceleration。

In [16]:

keras.optimizers.Nadam(Ir=0.002, beta_1=0.9, beta_2=0.999, epsilon=None, schedule_decay=0.004)



Out[16]:

<keras.optimizers.Nadam at 0x7ff3b559cad0>