# PS3 (2)

May 16, 2018

## 1 Part 1: Image classification

- 1. Set your random seed to 1234
- 2. Load the MNIST dataset

```
In [3]: import random
       from keras.datasets import mnist
       from keras import models
       from keras import layers
       from keras import regularizers
       from keras.utils import to_categorical
       from sklearn.model_selection import train_test_split
       import matplotlib.pyplot as plt
       %matplotlib inline
       random.seed(1234)
       (train_images, train_labels), (test_images, test_labels) = mnist.load_data()
       train_images = train_images.reshape((60000, 28 * 28)).astype('float32') / 255
       test_images = test_images.reshape((10000, 28 * 28)).astype('float32') / 255
       train_labels = to_categorical(train_labels)
       test_labels = to_categorical(test_labels)
       train_images, valid_images, train_labels, valid_labels = train_test_split(train_images
Using TensorFlow backend.
Downloading data from https://s3.amazonaws.com/img-datasets/mnist.npz
3.Implement a series of neural network models
i.Initial test
In [6]: network_org = models.Sequential()
       network_org.add(layers.Dense(512, activation='relu', input_shape= (28 * 28,)))
       network_org.add(layers.Dense(512, activation='relu'))
       network_org.add(layers.Dense(512, activation='relu'))
```

```
network_org.add(layers.Dense(10, activation='softmax'))
    network_org.compile(optimizer='rmsprop', loss='categorical_crossentropy', metrics=['ac
    result_org = network_org.fit(train_images, train_labels,validation_data=(valid_images,
Train on 50000 samples, validate on 10000 samples
Epoch 1/200
Epoch 2/200
Epoch 3/200
Epoch 4/200
50000/50000 [============== ] - 5s 95us/step - loss: 0.0661 - acc: 0.9799 - val
Epoch 5/200
50000/50000 [============== ] - 5s 95us/step - loss: 0.0463 - acc: 0.9859 - val
Epoch 6/200
50000/50000 [============== ] - 5s 96us/step - loss: 0.0356 - acc: 0.9894 - val
Epoch 7/200
Epoch 8/200
Epoch 9/200
Epoch 10/200
Epoch 11/200
Epoch 12/200
50000/50000 [============== ] - 5s 106us/step - loss: 0.0163 - acc: 0.9955 - va
Epoch 13/200
50000/50000 [============== ] - 5s 102us/step - loss: 0.0119 - acc: 0.9966 - va
Epoch 14/200
Epoch 15/200
50000/50000 [============== ] - 5s 104us/step - loss: 0.0123 - acc: 0.9971 - va
Epoch 16/200
50000/50000 [============= ] - 6s 113us/step - loss: 0.0119 - acc: 0.9971 - va
Epoch 17/200
50000/50000 [============= ] - 5s 105us/step - loss: 0.0088 - acc: 0.9975 - va
Epoch 18/200
50000/50000 [============== ] - 5s 101us/step - loss: 0.0107 - acc: 0.9973 - va
Epoch 19/200
50000/50000 [============== ] - 5s 100us/step - loss: 0.0094 - acc: 0.9975 - va
Epoch 20/200
Epoch 21/200
```

network\_org.add(layers.Dense(512, activation='relu'))

```
Epoch 22/200
Epoch 23/200
Epoch 24/200
Epoch 25/200
Epoch 26/200
Epoch 27/200
50000/50000 [=============== ] - 5s 98us/step - loss: 0.0085 - acc: 0.9978 - val
Epoch 28/200
Epoch 29/200
Epoch 30/200
50000/50000 [============== ] - 5s 98us/step - loss: 0.0067 - acc: 0.9983 - val
Epoch 31/200
Epoch 32/200
Epoch 33/200
Epoch 34/200
50000/50000 [=============== ] - 4s 88us/step - loss: 0.0065 - acc: 0.9984 - val
Epoch 35/200
Epoch 36/200
Epoch 37/200
Epoch 38/200
Epoch 39/200
Epoch 40/200
Epoch 41/200
50000/50000 [=============== ] - 5s 90us/step - loss: 0.0070 - acc: 0.9988 - val
Epoch 42/200
50000/50000 [============== ] - 5s 104us/step - loss: 0.0077 - acc: 0.9988 - va
Epoch 43/200
50000/50000 [=============== ] - 5s 103us/step - loss: 0.0083 - acc: 0.9985 - va
Epoch 44/200
50000/50000 [=============== ] - 5s 101us/step - loss: 0.0084 - acc: 0.9986 - va
Epoch 45/200
50000/50000 [============== ] - 5s 101us/step - loss: 0.0059 - acc: 0.9989 - va
```

```
Epoch 46/200
Epoch 47/200
Epoch 48/200
Epoch 49/200
Epoch 50/200
Epoch 51/200
50000/50000 [============== ] - 5s 91us/step - loss: 0.0070 - acc: 0.9990 - val
Epoch 52/200
50000/50000 [================ ] - 5s 90us/step - loss: 0.0099 - acc: 0.9987 - val
Epoch 53/200
Epoch 54/200
50000/50000 [============== ] - 5s 108us/step - loss: 0.0065 - acc: 0.9988 - va
Epoch 55/200
50000/50000 [============== ] - 5s 103us/step - loss: 0.0055 - acc: 0.9992 - va
Epoch 56/200
Epoch 57/200
Epoch 58/200
50000/50000 [=============== ] - 4s 86us/step - loss: 0.0095 - acc: 0.9986 - val
Epoch 59/200
Epoch 60/200
50000/50000 [============== ] - 5s 100us/step - loss: 0.0052 - acc: 0.9990 - va
Epoch 61/200
Epoch 62/200
Epoch 63/200
Epoch 64/200
Epoch 65/200
Epoch 66/200
Epoch 67/200
Epoch 68/200
50000/50000 [============== ] - 4s 86us/step - loss: 0.0057 - acc: 0.9992 - val
Epoch 69/200
50000/50000 [============== ] - 4s 84us/step - loss: 0.0072 - acc: 0.9989 - val
```

```
Epoch 70/200
Epoch 71/200
Epoch 72/200
50000/50000 [=============== ] - 4s 86us/step - loss: 0.0039 - acc: 0.9994 - val
Epoch 73/200
Epoch 74/200
Epoch 75/200
Epoch 76/200
Epoch 77/200
Epoch 78/200
Epoch 79/200
Epoch 80/200
Epoch 81/200
Epoch 82/200
50000/50000 [============== ] - 5s 98us/step - loss: 0.0074 - acc: 0.9987 - val
Epoch 83/200
Epoch 84/200
Epoch 85/200
Epoch 86/200
Epoch 87/200
Epoch 88/200
Epoch 89/200
Epoch 90/200
Epoch 91/200
Epoch 92/200
Epoch 93/200
50000/50000 [============== ] - 4s 87us/step - loss: 0.0052 - acc: 0.9994 - val
```

```
Epoch 94/200
Epoch 95/200
Epoch 96/200
Epoch 97/200
Epoch 98/200
50000/50000 [=============== ] - 5s 99us/step - loss: 0.0076 - acc: 0.9989 - val
Epoch 99/200
50000/50000 [=============== ] - 5s 98us/step - loss: 0.0080 - acc: 0.9991 - val
Epoch 100/200
Epoch 101/200
Epoch 102/200
50000/50000 [=============== ] - 5s 91us/step - loss: 0.0072 - acc: 0.9990 - val
Epoch 103/200
Epoch 104/200
Epoch 105/200
Epoch 106/200
50000/50000 [============== ] - 5s 102us/step - loss: 0.0079 - acc: 0.9991 - va
Epoch 107/200
Epoch 108/200
Epoch 109/200
Epoch 110/200
Epoch 111/200
Epoch 112/200
Epoch 113/200
50000/50000 [=============== ] - 5s 106us/step - loss: 0.0091 - acc: 0.9990 - va
Epoch 114/200
Epoch 115/200
Epoch 116/200
50000/50000 [============== ] - 5s 109us/step - loss: 0.0066 - acc: 0.9991 - va
Epoch 117/200
50000/50000 [============== ] - 5s 108us/step - loss: 0.0080 - acc: 0.9990 - va
```

```
Epoch 118/200
Epoch 119/200
Epoch 120/200
Epoch 121/200
Epoch 122/200
Epoch 123/200
Epoch 124/200
50000/50000 [=============== ] - 5s 100us/step - loss: 0.0044 - acc: 0.9993 - va
Epoch 125/200
Epoch 126/200
50000/50000 [=============== ] - 4s 87us/step - loss: 0.0072 - acc: 0.9991 - val
Epoch 127/200
Epoch 128/200
Epoch 129/200
50000/50000 [============== ] - 4s 87us/step - loss: 0.0102 - acc: 0.9988 - val
Epoch 130/200
50000/50000 [=============== ] - 4s 86us/step - loss: 0.0063 - acc: 0.9993 - val
Epoch 131/200
Epoch 132/200
Epoch 133/200
Epoch 134/200
Epoch 135/200
Epoch 136/200
Epoch 137/200
Epoch 138/200
Epoch 139/200
Epoch 140/200
50000/50000 [============== ] - 5s 108us/step - loss: 0.0025 - acc: 0.9996 - va
Epoch 141/200
50000/50000 [============== ] - 5s 109us/step - loss: 0.0041 - acc: 0.9994 - va
```

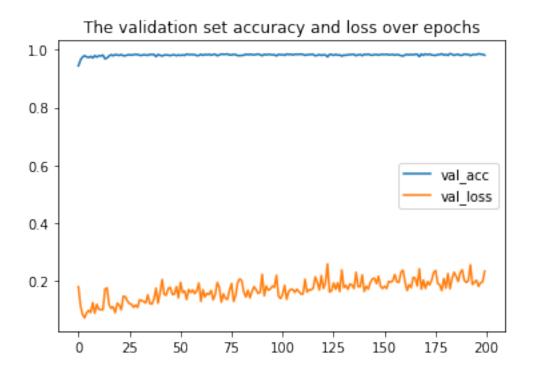
```
Epoch 142/200
Epoch 143/200
Epoch 144/200
Epoch 145/200
Epoch 146/200
Epoch 147/200
Epoch 148/200
50000/50000 [=============== ] - 4s 87us/step - loss: 0.0059 - acc: 0.9993 - val
Epoch 149/200
Epoch 150/200
50000/50000 [============== ] - 4s 87us/step - loss: 0.0072 - acc: 0.9992 - val
Epoch 151/200
Epoch 152/200
Epoch 153/200
50000/50000 [=============== ] - 5s 91us/step - loss: 0.0094 - acc: 0.9989 - val
Epoch 154/200
Epoch 155/200
Epoch 156/200
Epoch 157/200
Epoch 158/200
Epoch 159/200
Epoch 160/200
Epoch 161/200
Epoch 162/200
Epoch 163/200
50000/50000 [=============== ] - 4s 87us/step - loss: 0.0028 - acc: 0.9995 - val
Epoch 164/200
Epoch 165/200
50000/50000 [============== ] - 4s 87us/step - loss: 0.0033 - acc: 0.9995 - val
```

```
Epoch 166/200
Epoch 167/200
Epoch 168/200
50000/50000 [=============== ] - 4s 90us/step - loss: 0.0086 - acc: 0.9990 - val
Epoch 169/200
Epoch 170/200
Epoch 171/200
Epoch 172/200
Epoch 173/200
Epoch 174/200
Epoch 175/200
Epoch 176/200
Epoch 177/200
Epoch 178/200
50000/50000 [============== ] - 4s 87us/step - loss: 0.0072 - acc: 0.9992 - val
Epoch 179/200
Epoch 180/200
Epoch 181/200
Epoch 182/200
Epoch 183/200
Epoch 184/200
Epoch 185/200
50000/50000 [=============== ] - 4s 88us/step - loss: 0.0054 - acc: 0.9993 - val
Epoch 186/200
Epoch 187/200
Epoch 188/200
Epoch 189/200
50000/50000 [============== ] - 4s 87us/step - loss: 0.0068 - acc: 0.9994 - val
```

```
Epoch 190/200
Epoch 191/200
Epoch 192/200
Epoch 193/200
Epoch 194/200
Epoch 195/200
50000/50000 [============== ] - 5s 90us/step - loss: 0.0036 - acc: 0.9996 - val
Epoch 196/200
Epoch 197/200
Epoch 198/200
50000/50000 [=============== ] - 5s 94us/step - loss: 0.0026 - acc: 0.9997 - val
Epoch 199/200
50000/50000 [============== ] - 5s 94us/step - loss: 0.0027 - acc: 0.9997 - val
Epoch 200/200
In [7]: import matplotlib.pyplot as plt
   %matplotlib inline
   val_acc = result_org.history['val_acc']
   val_loss = result_org.history['val_loss']
   plt.plot(val_acc)
   plt.plot(val_loss)
   plt.legend(['val_acc', 'val_loss'])
```

plt.title('The validation set accuracy and loss over epochs')

plt.show()



From around epoch 42, the model starts to perform worse and worse based on the validation dataset.

#### ii. Implement dropout

Epoch 4/200

```
In [9]: network_dropout = models.Sequential()
       network_dropout.add(layers.Dense(512, activation='relu', input_shape=(28 * 28,)))
       network_dropout.add(layers.Dropout(0.5))
       network_dropout.add(layers.Dense(512, activation='relu'))
       network_dropout.add(layers.Dropout(0.5))
       network_dropout.add(layers.Dense(512, activation='relu'))
       network_dropout.add(layers.Dropout(0.5))
       network_dropout.add(layers.Dense(512, activation='relu'))
       network_dropout.add(layers.Dropout(0.5))
       network_dropout.add(layers.Dense(10, activation='softmax'))
       network_dropout.compile(optimizer='rmsprop', loss='categorical_crossentropy', metrics=
       result_dropout = network_dropout.fit(train_images, train_labels,validation_data=(valid
Train on 50000 samples, validate on 10000 samples
Epoch 1/200
Epoch 2/200
50000/50000 [==
                            =======] - 6s 130us/step - loss: 0.2653 - acc: 0.9216 - va
Epoch 3/200
50000/50000 [=====
                       =========] - 6s 128us/step - loss: 0.1967 - acc: 0.9450 - va
```

```
Epoch 5/200
50000/50000 [============== ] - 6s 122us/step - loss: 0.1427 - acc: 0.9596 - va
Epoch 6/200
50000/50000 [============= ] - 6s 130us/step - loss: 0.1289 - acc: 0.9638 - va
Epoch 7/200
Epoch 8/200
50000/50000 [============== ] - 6s 117us/step - loss: 0.1106 - acc: 0.9696 - va
Epoch 9/200
50000/50000 [============== ] - 6s 127us/step - loss: 0.0976 - acc: 0.9725 - va
Epoch 10/200
Epoch 11/200
Epoch 12/200
Epoch 13/200
Epoch 14/200
Epoch 15/200
Epoch 16/200
Epoch 17/200
Epoch 18/200
Epoch 19/200
Epoch 20/200
Epoch 21/200
50000/50000 [============= ] - 7s 130us/step - loss: 0.0637 - acc: 0.9828 - va
Epoch 22/200
50000/50000 [============== ] - 6s 129us/step - loss: 0.0619 - acc: 0.9837 - va
Epoch 23/200
Epoch 24/200
Epoch 25/200
Epoch 26/200
Epoch 27/200
Epoch 28/200
```

```
Epoch 29/200
50000/50000 [============== ] - 7s 133us/step - loss: 0.0568 - acc: 0.9855 - va
Epoch 30/200
Epoch 31/200
50000/50000 [============== ] - 7s 133us/step - loss: 0.0531 - acc: 0.9864 - va
Epoch 32/200
50000/50000 [============== ] - 7s 133us/step - loss: 0.0562 - acc: 0.9865 - va
Epoch 33/200
Epoch 34/200
Epoch 35/200
50000/50000 [============== ] - 7s 140us/step - loss: 0.0567 - acc: 0.9866 - va
Epoch 36/200
Epoch 37/200
Epoch 38/200
Epoch 39/200
Epoch 40/200
Epoch 41/200
Epoch 42/200
Epoch 43/200
Epoch 44/200
50000/50000 [============== ] - 7s 130us/step - loss: 0.0542 - acc: 0.9879 - va
Epoch 45/200
50000/50000 [============= ] - 6s 122us/step - loss: 0.0504 - acc: 0.9888 - va
Epoch 46/200
Epoch 47/200
Epoch 48/200
Epoch 49/200
Epoch 50/200
Epoch 51/200
Epoch 52/200
```

```
Epoch 53/200
50000/50000 [============== ] - 6s 114us/step - loss: 0.0486 - acc: 0.9890 - va
Epoch 54/200
Epoch 55/200
50000/50000 [============== ] - 8s 157us/step - loss: 0.0536 - acc: 0.9884 - va
Epoch 56/200
50000/50000 [============== ] - 7s 148us/step - loss: 0.0526 - acc: 0.9884 - va
Epoch 57/200
Epoch 58/200
Epoch 59/200
50000/50000 [============== ] - 7s 135us/step - loss: 0.0511 - acc: 0.9891 - va
Epoch 60/200
Epoch 61/200
Epoch 62/200
Epoch 63/200
Epoch 64/200
Epoch 65/200
Epoch 66/200
Epoch 67/200
Epoch 68/200
50000/50000 [============== ] - 7s 137us/step - loss: 0.0510 - acc: 0.9907 - va
Epoch 69/200
50000/50000 [============= ] - 7s 135us/step - loss: 0.0560 - acc: 0.9900 - va
Epoch 70/200
50000/50000 [============== ] - 7s 137us/step - loss: 0.0541 - acc: 0.9899 - va
Epoch 71/200
Epoch 72/200
Epoch 73/200
Epoch 74/200
Epoch 75/200
Epoch 76/200
```

```
Epoch 77/200
50000/50000 [============== ] - 8s 151us/step - loss: 0.0523 - acc: 0.9908 - va
Epoch 78/200
Epoch 79/200
50000/50000 [============== ] - 7s 148us/step - loss: 0.0520 - acc: 0.9903 - va
Epoch 80/200
Epoch 81/200
50000/50000 [============== ] - 7s 132us/step - loss: 0.0542 - acc: 0.9902 - va
Epoch 82/200
Epoch 83/200
50000/50000 [============== ] - 7s 134us/step - loss: 0.0575 - acc: 0.9902 - va
Epoch 84/200
Epoch 85/200
Epoch 86/200
Epoch 87/200
Epoch 88/200
Epoch 89/200
Epoch 90/200
Epoch 91/200
50000/50000 [=============== ] - 8s 158us/step - loss: 0.0491 - acc: 0.9914 - va
Epoch 92/200
50000/50000 [============== ] - 8s 151us/step - loss: 0.0605 - acc: 0.9901 - va
Epoch 93/200
50000/50000 [============= ] - 7s 148us/step - loss: 0.0507 - acc: 0.9914 - va
Epoch 94/200
50000/50000 [============== ] - 7s 142us/step - loss: 0.0535 - acc: 0.9911 - va
Epoch 95/200
Epoch 96/200
Epoch 97/200
Epoch 98/200
Epoch 99/200
Epoch 100/200
```

```
Epoch 101/200
Epoch 102/200
50000/50000 [============== ] - 7s 132us/step - loss: 0.0574 - acc: 0.9915 - va
Epoch 103/200
50000/50000 [============== ] - 7s 132us/step - loss: 0.0550 - acc: 0.9915 - va
Epoch 104/200
50000/50000 [============== ] - 7s 132us/step - loss: 0.0559 - acc: 0.9907 - va
Epoch 105/200
50000/50000 [============== ] - 7s 131us/step - loss: 0.0586 - acc: 0.9912 - va
Epoch 106/200
Epoch 107/200
50000/50000 [============== ] - 7s 130us/step - loss: 0.0533 - acc: 0.9917 - va
Epoch 108/200
Epoch 109/200
Epoch 110/200
Epoch 111/200
Epoch 112/200
Epoch 113/200
Epoch 114/200
Epoch 115/200
Epoch 116/200
50000/50000 [============== ] - 7s 146us/step - loss: 0.0649 - acc: 0.9905 - va
Epoch 117/200
50000/50000 [============= ] - 7s 136us/step - loss: 0.0668 - acc: 0.9905 - va
Epoch 118/200
Epoch 119/200
Epoch 120/200
Epoch 121/200
Epoch 122/200
Epoch 123/200
Epoch 124/200
```

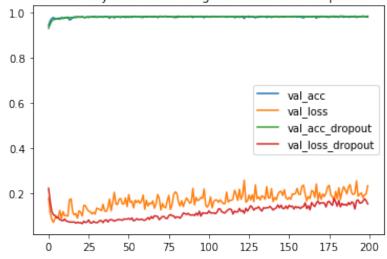
```
Epoch 125/200
50000/50000 [============== ] - 7s 132us/step - loss: 0.0631 - acc: 0.9912 - va
Epoch 126/200
Epoch 127/200
50000/50000 [============== ] - 7s 133us/step - loss: 0.0607 - acc: 0.9919 - va
Epoch 128/200
50000/50000 [============== ] - 7s 134us/step - loss: 0.0673 - acc: 0.9909 - va
Epoch 129/200
50000/50000 [============== ] - 8s 155us/step - loss: 0.0646 - acc: 0.9913 - va
Epoch 130/200
Epoch 131/200
50000/50000 [============== ] - 7s 140us/step - loss: 0.0643 - acc: 0.9909 - va
Epoch 132/200
Epoch 133/200
Epoch 134/200
Epoch 135/200
Epoch 136/200
Epoch 137/200
Epoch 138/200
Epoch 139/200
Epoch 140/200
Epoch 141/200
50000/50000 [============= ] - 7s 134us/step - loss: 0.0699 - acc: 0.9909 - va
Epoch 142/200
50000/50000 [============== ] - 7s 140us/step - loss: 0.0708 - acc: 0.9909 - va
Epoch 143/200
Epoch 144/200
Epoch 145/200
Epoch 146/200
Epoch 147/200
Epoch 148/200
```

```
Epoch 149/200
50000/50000 [============== ] - 7s 132us/step - loss: 0.0681 - acc: 0.9914 - va
Epoch 150/200
50000/50000 [============== ] - 7s 139us/step - loss: 0.0698 - acc: 0.9913 - va
Epoch 151/200
Epoch 152/200
50000/50000 [============== ] - 7s 133us/step - loss: 0.0790 - acc: 0.9904 - va
Epoch 153/200
50000/50000 [============== ] - 7s 131us/step - loss: 0.0681 - acc: 0.9912 - va
Epoch 154/200
Epoch 155/200
Epoch 156/200
Epoch 157/200
Epoch 158/200
Epoch 159/200
Epoch 160/200
Epoch 161/200
Epoch 162/200
Epoch 163/200
Epoch 164/200
50000/50000 [============== ] - 7s 135us/step - loss: 0.0830 - acc: 0.9905 - va
Epoch 165/200
50000/50000 [============= ] - 6s 118us/step - loss: 0.0672 - acc: 0.9922 - va
Epoch 166/200
50000/50000 [============== ] - 6s 121us/step - loss: 0.0831 - acc: 0.9909 - va
Epoch 167/200
Epoch 168/200
Epoch 169/200
Epoch 170/200
Epoch 171/200
Epoch 172/200
```

```
Epoch 173/200
50000/50000 [============== ] - 7s 135us/step - loss: 0.0909 - acc: 0.9907 - va
Epoch 174/200
50000/50000 [============== ] - 7s 140us/step - loss: 0.0860 - acc: 0.9911 - va
Epoch 175/200
Epoch 176/200
50000/50000 [============== ] - 6s 129us/step - loss: 0.0905 - acc: 0.9907 - va
Epoch 177/200
50000/50000 [============== ] - 7s 143us/step - loss: 0.0803 - acc: 0.9914 - va
Epoch 178/200
Epoch 179/200
50000/50000 [============== ] - 6s 112us/step - loss: 0.0799 - acc: 0.9912 - va
Epoch 180/200
Epoch 181/200
Epoch 182/200
Epoch 183/200
Epoch 184/200
Epoch 185/200
Epoch 186/200
Epoch 187/200
Epoch 188/200
50000/50000 [============== ] - 6s 112us/step - loss: 0.0828 - acc: 0.9908 - va
Epoch 189/200
50000/50000 [============= ] - 6s 113us/step - loss: 0.0804 - acc: 0.9914 - va
Epoch 190/200
50000/50000 [============== ] - 6s 111us/step - loss: 0.0804 - acc: 0.9914 - va
Epoch 191/200
Epoch 192/200
Epoch 193/200
Epoch 194/200
Epoch 195/200
Epoch 196/200
```

```
Epoch 197/200
50000/50000 [============== ] - 7s 143us/step - loss: 0.0853 - acc: 0.9908 - va
Epoch 198/200
50000/50000 [====
                          =======] - 7s 144us/step - loss: 0.0869 - acc: 0.9912 - va
Epoch 199/200
50000/50000 [==
                            ========] - 7s 147us/step - loss: 0.0793 - acc: 0.9919 - va
Epoch 200/200
50000/50000 [===
                                ======] - 7s 142us/step - loss: 0.0827 - acc: 0.9917 - va
In [10]: val_acc = result_org.history['val_acc']
       val_loss = result_org.history['val_loss']
       val_acc_dropout = result_dropout.history['val_acc']
       val_loss_dropout = result_dropout.history['val_loss']
       plt.plot(val_acc)
       plt.plot(val_loss)
       plt.plot(val_acc_dropout)
       plt.plot(val_loss_dropout)
       plt.legend(['val_acc', 'val_loss', 'val_acc_dropout', 'val_loss_dropout'])
       plt.title('The validation set accuracy and loss of original model and dropout model or
       plt.show()
```

The validation set accuracy and loss of original model and dropout model over epochs



This model perform slightly better compared with the old model in terms of val\_loss, but the val\_acc are really similar.

```
network_l1.add(layers.Dense(10, activation='softmax'))
            network_l1.compile(optimizer='rmsprop', loss='categorical_crossentropy', metrics=['ac
            result_l1 = network_l1.fit(train_images, train_labels, validation_data=(valid_images, valid_images, validation_data=(valid_images, validation_data=(valid_i
Train on 50000 samples, validate on 10000 samples
Epoch 1/200
Epoch 2/200
50000/50000 [============== ] - 6s 113us/step - loss: 3.0588 - acc: 0.7929 - va
Epoch 3/200
50000/50000 [============== ] - 6s 118us/step - loss: 2.1836 - acc: 0.8357 - va
Epoch 4/200
Epoch 5/200
50000/50000 [============== ] - 6s 115us/step - loss: 1.6821 - acc: 0.8743 - va
Epoch 6/200
50000/50000 [============== ] - 6s 118us/step - loss: 1.5538 - acc: 0.8866 - va
Epoch 7/200
Epoch 8/200
Epoch 9/200
Epoch 10/200
Epoch 11/200
50000/50000 [============== ] - 5s 109us/step - loss: 1.2727 - acc: 0.9221 - va
Epoch 12/200
50000/50000 [============== ] - 6s 116us/step - loss: 1.2387 - acc: 0.9264 - va
Epoch 13/200
Epoch 14/200
50000/50000 [============== ] - 6s 112us/step - loss: 1.1832 - acc: 0.9321 - va
Epoch 15/200
50000/50000 [============= ] - 7s 137us/step - loss: 1.1615 - acc: 0.9348 - va
Epoch 16/200
50000/50000 [============= ] - 6s 118us/step - loss: 1.1450 - acc: 0.9354 - va
Epoch 17/200
Epoch 18/200
50000/50000 [============== ] - 6s 114us/step - loss: 1.1182 - acc: 0.9381 - va
Epoch 19/200
Epoch 20/200
```

network\_l1.add(layers.Dense(512, activation='relu', kernel\_regularizer=regularizers.l network\_l1.add(layers.Dense(512, activation='relu', kernel\_regularizer=regularizers.l network\_l1.add(layers.Dense(512, activation='relu', kernel\_regularizer=regularizers.l

```
Epoch 21/200
Epoch 22/200
Epoch 23/200
Epoch 24/200
Epoch 25/200
Epoch 26/200
Epoch 27/200
Epoch 28/200
Epoch 29/200
Epoch 30/200
Epoch 31/200
Epoch 32/200
50000/50000 [============= ] - 6s 117us/step - loss: 1.0130 - acc: 0.9500 - va
Epoch 33/200
50000/50000 [============== ] - 6s 118us/step - loss: 1.0096 - acc: 0.9493 - va
Epoch 34/200
Epoch 35/200
50000/50000 [============== ] - 6s 118us/step - loss: 1.0030 - acc: 0.9494 - va
Epoch 36/200
Epoch 37/200
Epoch 38/200
Epoch 39/200
Epoch 40/200
Epoch 41/200
Epoch 42/200
Epoch 43/200
50000/50000 [============== ] - 6s 117us/step - loss: 0.9805 - acc: 0.9510 - va
Epoch 44/200
50000/50000 [============== ] - 6s 115us/step - loss: 0.9772 - acc: 0.9524 - va
```

```
Epoch 45/200
Epoch 46/200
Epoch 47/200
Epoch 48/200
Epoch 49/200
Epoch 50/200
50000/50000 [============== ] - 6s 114us/step - loss: 0.9664 - acc: 0.9524 - va
Epoch 51/200
Epoch 52/200
Epoch 53/200
Epoch 54/200
50000/50000 [============== ] - 6s 115us/step - loss: 0.9587 - acc: 0.9539 - va
Epoch 55/200
Epoch 56/200
Epoch 57/200
50000/50000 [============== ] - 5s 101us/step - loss: 0.9530 - acc: 0.9547 - va
Epoch 58/200
Epoch 59/200
50000/50000 [============== ] - 5s 101us/step - loss: 0.9502 - acc: 0.9546 - va
Epoch 60/200
Epoch 61/200
Epoch 62/200
Epoch 63/200
Epoch 64/200
Epoch 65/200
Epoch 66/200
Epoch 67/200
50000/50000 [============== ] - 5s 101us/step - loss: 0.9392 - acc: 0.9560 - va
Epoch 68/200
50000/50000 [============== ] - 5s 101us/step - loss: 0.9401 - acc: 0.9568 - va
```

```
Epoch 69/200
Epoch 70/200
Epoch 71/200
Epoch 72/200
Epoch 73/200
50000/50000 [=============== ] - 5s 101us/step - loss: 0.9342 - acc: 0.9562 - va
Epoch 74/200
Epoch 75/200
Epoch 76/200
Epoch 77/200
50000/50000 [============== ] - 6s 117us/step - loss: 0.9297 - acc: 0.9567 - va
Epoch 78/200
50000/50000 [============== ] - 5s 108us/step - loss: 0.9296 - acc: 0.9571 - va
Epoch 79/200
50000/50000 [============== ] - 5s 101us/step - loss: 0.9299 - acc: 0.9568 - va
Epoch 80/200
50000/50000 [============= ] - 5s 102us/step - loss: 0.9308 - acc: 0.9560 - va
Epoch 81/200
50000/50000 [============== ] - 5s 101us/step - loss: 0.9262 - acc: 0.9575 - va
Epoch 82/200
Epoch 83/200
Epoch 84/200
Epoch 85/200
Epoch 86/200
Epoch 87/200
Epoch 88/200
Epoch 89/200
Epoch 90/200
Epoch 91/200
50000/50000 [============== ] - 5s 101us/step - loss: 0.9185 - acc: 0.9586 - va
Epoch 92/200
50000/50000 [============== ] - 5s 101us/step - loss: 0.9189 - acc: 0.9583 - va
```

```
Epoch 93/200
Epoch 94/200
Epoch 95/200
Epoch 96/200
Epoch 97/200
Epoch 98/200
50000/50000 [============== ] - 5s 101us/step - loss: 0.9168 - acc: 0.9569 - va
Epoch 99/200
50000/50000 [=============== ] - 5s 101us/step - loss: 0.9144 - acc: 0.9584 - va
Epoch 100/200
Epoch 101/200
Epoch 102/200
50000/50000 [============== ] - 5s 102us/step - loss: 0.9121 - acc: 0.9593 - va
Epoch 103/200
50000/50000 [============== ] - 5s 101us/step - loss: 0.9105 - acc: 0.9594 - va
Epoch 104/200
50000/50000 [============= ] - 5s 100us/step - loss: 0.9092 - acc: 0.9598 - va
Epoch 105/200
50000/50000 [============== ] - 5s 107us/step - loss: 0.9088 - acc: 0.9596 - va
Epoch 106/200
50000/50000 [============== ] - 5s 108us/step - loss: 0.9106 - acc: 0.9585 - va
Epoch 107/200
50000/50000 [============== ] - 6s 122us/step - loss: 0.9090 - acc: 0.9583 - va
Epoch 108/200
50000/50000 [=============== ] - 5s 107us/step - loss: 0.9066 - acc: 0.9588 - va
Epoch 109/200
Epoch 110/200
Epoch 111/200
Epoch 112/200
Epoch 113/200
Epoch 114/200
Epoch 115/200
50000/50000 [============== ] - 5s 106us/step - loss: 0.9010 - acc: 0.9604 - va
Epoch 116/200
50000/50000 [============== ] - 5s 101us/step - loss: 0.9015 - acc: 0.9601 - va
```

```
Epoch 117/200
Epoch 118/200
Epoch 119/200
Epoch 120/200
Epoch 121/200
Epoch 122/200
Epoch 123/200
Epoch 124/200
Epoch 125/200
50000/50000 [============== ] - 6s 119us/step - loss: 0.8975 - acc: 0.9588 - va
Epoch 126/200
50000/50000 [============== ] - 6s 121us/step - loss: 0.8946 - acc: 0.9596 - va
Epoch 127/200
50000/50000 [============== ] - 6s 118us/step - loss: 0.8944 - acc: 0.9594 - va
Epoch 128/200
Epoch 129/200
50000/50000 [============== ] - 6s 123us/step - loss: 0.8920 - acc: 0.9600 - va
Epoch 130/200
Epoch 131/200
50000/50000 [============== ] - 6s 118us/step - loss: 0.8949 - acc: 0.9586 - va
Epoch 132/200
Epoch 133/200
Epoch 134/200
Epoch 135/200
Epoch 136/200
Epoch 137/200
Epoch 138/200
Epoch 139/200
50000/50000 [============== ] - 6s 118us/step - loss: 0.8888 - acc: 0.9612 - va
Epoch 140/200
50000/50000 [============== ] - 7s 130us/step - loss: 0.8887 - acc: 0.9618 - va
```

```
Epoch 141/200
Epoch 142/200
Epoch 143/200
Epoch 144/200
Epoch 145/200
Epoch 146/200
50000/50000 [=============== ] - 5s 102us/step - loss: 0.8839 - acc: 0.9622 - va
Epoch 147/200
Epoch 148/200
Epoch 149/200
50000/50000 [============== ] - 6s 123us/step - loss: 0.8847 - acc: 0.9613 - va
Epoch 150/200
Epoch 151/200
50000/50000 [============== ] - 6s 115us/step - loss: 0.8856 - acc: 0.9604 - va
Epoch 152/200
50000/50000 [============= ] - 6s 112us/step - loss: 0.8829 - acc: 0.9616 - va
Epoch 153/200
50000/50000 [============== ] - 6s 114us/step - loss: 0.8829 - acc: 0.9614 - va
Epoch 154/200
Epoch 155/200
50000/50000 [============== ] - 6s 115us/step - loss: 0.8782 - acc: 0.9631 - va
Epoch 156/200
Epoch 157/200
Epoch 158/200
Epoch 159/200
Epoch 160/200
Epoch 161/200
Epoch 162/200
Epoch 163/200
50000/50000 [============== ] - 7s 136us/step - loss: 0.8783 - acc: 0.9617 - va
Epoch 164/200
50000/50000 [============== ] - 6s 121us/step - loss: 0.8789 - acc: 0.9614 - va
```

```
Epoch 165/200
Epoch 166/200
Epoch 167/200
Epoch 168/200
Epoch 169/200
Epoch 170/200
50000/50000 [=============== ] - 5s 104us/step - loss: 0.8804 - acc: 0.9608 - va
Epoch 171/200
Epoch 172/200
Epoch 173/200
Epoch 174/200
50000/50000 [============== ] - 6s 111us/step - loss: 0.8746 - acc: 0.9618 - va
Epoch 175/200
50000/50000 [============== ] - 5s 106us/step - loss: 0.8764 - acc: 0.9613 - va
Epoch 176/200
Epoch 177/200
50000/50000 [============== ] - 5s 101us/step - loss: 0.8733 - acc: 0.9622 - va
Epoch 178/200
Epoch 179/200
50000/50000 [============== ] - 5s 103us/step - loss: 0.8748 - acc: 0.9622 - va
Epoch 180/200
Epoch 181/200
Epoch 182/200
Epoch 183/200
Epoch 184/200
Epoch 185/200
Epoch 186/200
Epoch 187/200
50000/50000 [============== ] - 6s 120us/step - loss: 0.8749 - acc: 0.9616 - va
Epoch 188/200
```

```
Epoch 191/200
Epoch 192/200
Epoch 193/200
Epoch 194/200
50000/50000 [============== ] - 5s 108us/step - loss: 0.8716 - acc: 0.9621 - va
Epoch 195/200
Epoch 196/200
50000/50000 [=============== ] - 5s 102us/step - loss: 0.8701 - acc: 0.9622 - va
Epoch 197/200
50000/50000 [============== ] - 5s 102us/step - loss: 0.8727 - acc: 0.9619 - va
Epoch 198/200
50000/50000 [============== ] - 5s 104us/step - loss: 0.8700 - acc: 0.9624 - va
Epoch 199/200
Epoch 200/200
In [12]: network_12 = models.Sequential()
           network_12.add(layers.Dense(512, activation='relu', input_shape=(28 * 28,), kernel_re
           network_12.add(layers.Dense(512, activation='relu', kernel_regularizer=regularizers.1
           network_12.add(layers.Dense(512, activation='relu', kernel_regularizer=regularizers.1:
           network 12.add(layers.Dense(512, activation='relu', kernel_regularizer=regularizers.1:
           network_12.add(layers.Dense(10, activation='softmax'))
           network_12.compile(optimizer='rmsprop', loss='categorical_crossentropy', metrics=['ac
           result_12 = network_12.fit(train_images, train_labels, validation_data=(valid_images, valid_images, validation_data=(valid_images, validation_data=(valid_i
Train on 50000 samples, validate on 10000 samples
Epoch 1/200
Epoch 2/200
Epoch 3/200
Epoch 4/200
Epoch 5/200
50000/50000 [============== ] - 6s 124us/step - loss: 0.3308 - acc: 0.9648 - va
Epoch 6/200
```

Epoch 189/200

Epoch 190/200

```
Epoch 7/200
Epoch 8/200
Epoch 9/200
Epoch 10/200
Epoch 11/200
Epoch 12/200
50000/50000 [============== ] - 6s 124us/step - loss: 0.1850 - acc: 0.9797 - va
Epoch 13/200
Epoch 14/200
Epoch 15/200
Epoch 16/200
Epoch 17/200
Epoch 18/200
Epoch 19/200
50000/50000 [============== ] - 6s 122us/step - loss: 0.1464 - acc: 0.9841 - va
Epoch 20/200
Epoch 21/200
Epoch 22/200
Epoch 23/200
Epoch 24/200
Epoch 25/200
Epoch 26/200
Epoch 27/200
Epoch 28/200
Epoch 29/200
50000/50000 [============== ] - 6s 119us/step - loss: 0.1237 - acc: 0.9868 - va
Epoch 30/200
50000/50000 [============== ] - 6s 120us/step - loss: 0.1181 - acc: 0.9885 - va
```

```
Epoch 31/200
Epoch 32/200
Epoch 33/200
Epoch 34/200
Epoch 35/200
Epoch 36/200
Epoch 37/200
Epoch 38/200
Epoch 39/200
50000/50000 [============== ] - 6s 121us/step - loss: 0.1075 - acc: 0.9894 - va
Epoch 40/200
50000/50000 [============== ] - 6s 118us/step - loss: 0.1098 - acc: 0.9885 - va
Epoch 41/200
Epoch 42/200
Epoch 43/200
50000/50000 [============== ] - 6s 119us/step - loss: 0.1086 - acc: 0.9887 - va
Epoch 44/200
Epoch 45/200
Epoch 46/200
Epoch 47/200
Epoch 48/200
Epoch 49/200
Epoch 50/200
Epoch 51/200
Epoch 52/200
Epoch 53/200
50000/50000 [============== ] - 6s 117us/step - loss: 0.1025 - acc: 0.9890 - va
Epoch 54/200
50000/50000 [============== ] - 6s 116us/step - loss: 0.0998 - acc: 0.9896 - va
```

```
Epoch 55/200
Epoch 56/200
Epoch 57/200
Epoch 58/200
Epoch 59/200
Epoch 60/200
50000/50000 [============== ] - 6s 119us/step - loss: 0.0980 - acc: 0.9900 - va
Epoch 61/200
Epoch 62/200
Epoch 63/200
Epoch 64/200
Epoch 65/200
50000/50000 [============== ] - 6s 119us/step - loss: 0.0979 - acc: 0.9900 - va
Epoch 66/200
Epoch 67/200
50000/50000 [============== ] - 6s 119us/step - loss: 0.0974 - acc: 0.9890 - va
Epoch 68/200
Epoch 69/200
50000/50000 [============== ] - 6s 121us/step - loss: 0.0927 - acc: 0.9908 - va
Epoch 70/200
Epoch 71/200
Epoch 72/200
Epoch 73/200
Epoch 74/200
Epoch 75/200
Epoch 76/200
Epoch 77/200
50000/50000 [============== ] - 6s 122us/step - loss: 0.0932 - acc: 0.9906 - va
Epoch 78/200
50000/50000 [============== ] - 6s 126us/step - loss: 0.0904 - acc: 0.9912 - va
```

```
Epoch 79/200
Epoch 80/200
Epoch 81/200
Epoch 82/200
Epoch 83/200
Epoch 84/200
50000/50000 [============== ] - 5s 99us/step - loss: 0.0912 - acc: 0.9904 - val
Epoch 85/200
Epoch 86/200
50000/50000 [=============== ] - 5s 101us/step - loss: 0.0917 - acc: 0.9910 - va
Epoch 87/200
50000/50000 [============== ] - 5s 100us/step - loss: 0.0886 - acc: 0.9910 - va
Epoch 88/200
50000/50000 [============== ] - 5s 101us/step - loss: 0.0892 - acc: 0.9909 - va
Epoch 89/200
Epoch 90/200
Epoch 91/200
50000/50000 [============== ] - 5s 100us/step - loss: 0.0862 - acc: 0.9921 - va
Epoch 92/200
Epoch 93/200
50000/50000 [============== ] - 5s 101us/step - loss: 0.0880 - acc: 0.9913 - va
Epoch 94/200
Epoch 95/200
Epoch 96/200
Epoch 97/200
Epoch 98/200
Epoch 99/200
Epoch 100/200
Epoch 101/200
50000/50000 [============== ] - 5s 100us/step - loss: 0.0870 - acc: 0.9913 - va
Epoch 102/200
50000/50000 [============== ] - 5s 100us/step - loss: 0.0841 - acc: 0.9923 - va
```

```
Epoch 103/200
Epoch 104/200
Epoch 105/200
Epoch 106/200
Epoch 107/200
Epoch 108/200
Epoch 109/200
Epoch 110/200
Epoch 111/200
50000/50000 [============== ] - 6s 119us/step - loss: 0.0845 - acc: 0.9914 - va
Epoch 112/200
50000/50000 [============== ] - 6s 116us/step - loss: 0.0837 - acc: 0.9920 - va
Epoch 113/200
50000/50000 [============== ] - 6s 116us/step - loss: 0.0848 - acc: 0.9914 - va
Epoch 114/200
50000/50000 [============= ] - 6s 120us/step - loss: 0.0874 - acc: 0.9911 - va
Epoch 115/200
50000/50000 [============== ] - 6s 120us/step - loss: 0.0862 - acc: 0.9910 - va
Epoch 116/200
Epoch 117/200
50000/50000 [============== ] - 7s 132us/step - loss: 0.0802 - acc: 0.9927 - va
Epoch 118/200
Epoch 119/200
Epoch 120/200
Epoch 121/200
Epoch 122/200
Epoch 123/200
Epoch 124/200
Epoch 125/200
50000/50000 [============== ] - 8s 170us/step - loss: 0.0795 - acc: 0.9927 - va
Epoch 126/200
```

50000/50000 [============== ] - 8s 169us/step - loss: 0.0830 - acc: 0.9911 - va

```
Epoch 127/200
Epoch 128/200
Epoch 129/200
Epoch 130/200
Epoch 131/200
Epoch 132/200
Epoch 133/200
Epoch 134/200
Epoch 135/200
50000/50000 [============== ] - 6s 123us/step - loss: 0.0794 - acc: 0.9922 - va
Epoch 136/200
Epoch 137/200
Epoch 138/200
Epoch 139/200
Epoch 140/200
50000/50000 [============== ] - 6s 122us/step - loss: 0.0790 - acc: 0.9925 - va
Epoch 141/200
50000/50000 [============== ] - 6s 122us/step - loss: 0.0806 - acc: 0.9917 - va
Epoch 142/200
Epoch 143/200
Epoch 144/200
Epoch 145/200
Epoch 146/200
Epoch 147/200
Epoch 148/200
Epoch 149/200
50000/50000 [============== ] - 6s 121us/step - loss: 0.0796 - acc: 0.9922 - va
Epoch 150/200
```

```
Epoch 151/200
Epoch 152/200
Epoch 153/200
Epoch 154/200
Epoch 155/200
Epoch 156/200
Epoch 157/200
Epoch 158/200
Epoch 159/200
50000/50000 [============== ] - 6s 125us/step - loss: 0.0771 - acc: 0.9926 - va
Epoch 160/200
50000/50000 [============== ] - 6s 126us/step - loss: 0.0796 - acc: 0.9923 - va
Epoch 161/200
50000/50000 [============== ] - 8s 164us/step - loss: 0.0777 - acc: 0.9926 - va
Epoch 162/200
Epoch 163/200
50000/50000 [============== ] - 9s 180us/step - loss: 0.0767 - acc: 0.9925 - va
Epoch 164/200
Epoch 165/200
Epoch 166/200
Epoch 167/200
Epoch 168/200
Epoch 169/200
Epoch 170/200
Epoch 171/200
Epoch 172/200
Epoch 173/200
50000/50000 [============== ] - 9s 172us/step - loss: 0.0769 - acc: 0.9923 - va
Epoch 174/200
```

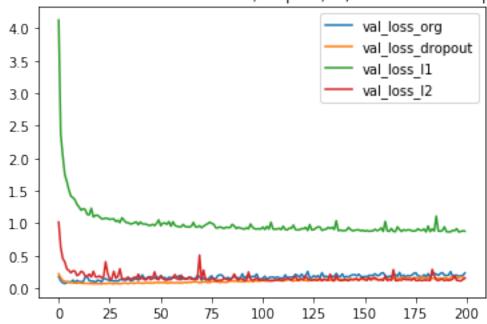
50000/50000 [============== ] - 9s 172us/step - loss: 0.0758 - acc: 0.9928 - va

```
Epoch 175/200
Epoch 176/200
Epoch 177/200
Epoch 178/200
Epoch 179/200
Epoch 180/200
Epoch 181/200
Epoch 182/200
Epoch 183/200
50000/50000 [============== ] - 8s 153us/step - loss: 0.0834 - acc: 0.9905 - va
Epoch 184/200
50000/50000 [============== ] - 7s 132us/step - loss: 0.0728 - acc: 0.9935 - va
Epoch 185/200
50000/50000 [============== ] - 7s 132us/step - loss: 0.0750 - acc: 0.9932 - va
Epoch 186/200
Epoch 187/200
50000/50000 [============== ] - 9s 174us/step - loss: 0.0767 - acc: 0.9923 - va
Epoch 188/200
Epoch 189/200
Epoch 190/200
Epoch 191/200
Epoch 192/200
Epoch 193/200
Epoch 194/200
Epoch 195/200
Epoch 196/200
Epoch 197/200
Epoch 198/200
```

50000/50000 [============== ] - 7s 144us/step - loss: 0.0758 - acc: 0.9928 - va

```
Epoch 199/200
Epoch 200/200
                                  ====] - 7s 144us/step - loss: 0.0746 - acc: 0.9928 - va
50000/50000 [=====
In [13]: val_loss_org = result_org.history['val_loss']
       val_loss_dropout = result_dropout.history['val_loss']
       val_loss_l1 = result_l1.history['val_loss']
       val_loss_12 = result_12.history['val_loss']
       plt.plot(val_loss_org)
       plt.plot(val_loss_dropout)
       plt.plot(val_loss_l1)
       plt.plot(val_loss_12)
       plt.legend(['val_loss_org', 'val_loss_dropout', 'val_loss_11', 'val_loss_12'])
       plt.title('The validation loss for the initial/dropout/L1/L2 model over epochs')
       plt.show()
```

### The validation loss for the initial/dropout/L1/L2 model over epochs



The dropout-adding model appears to perform the best when the epoch number is smaller than 100, and the l2-loss model appears to perform the best when the epoch number is greater than 100.

```
network_best.add(layers.Dropout(0.5))
  network_best.add(layers.Dense(512, activation='relu'))
  network_best.add(layers.Dropout(0.5))
  network_best.add(layers.Dense(512, activation='relu'))
  network best.add(layers.Dropout(0.5))
  network_best.add(layers.Dense(512, activation='relu'))
  network best.add(layers.Dropout(0.5))
  network_best.add(layers.Dense(10, activation='softmax'))
  network_best.compile(optimizer='rmsprop', loss='categorical_crossentropy', metrics=['s
  result_best = network_best.fit(train_images, train_labels, epochs=22, batch_size=512)
Epoch 1/22
Epoch 2/22
Epoch 3/22
Epoch 4/22
Epoch 5/22
Epoch 6/22
Epoch 7/22
Epoch 8/22
Epoch 9/22
Epoch 10/22
Epoch 11/22
Epoch 12/22
Epoch 13/22
Epoch 14/22
Epoch 15/22
Epoch 16/22
Epoch 17/22
Epoch 18/22
```

The test set loss and accuracy of the 0.07913 and 0.9838. Compared with the baseline model from chapter 2.1 in the book, whose test accuracy is 0.9785, my model is better.

### 2 Part 2: Scalar regression

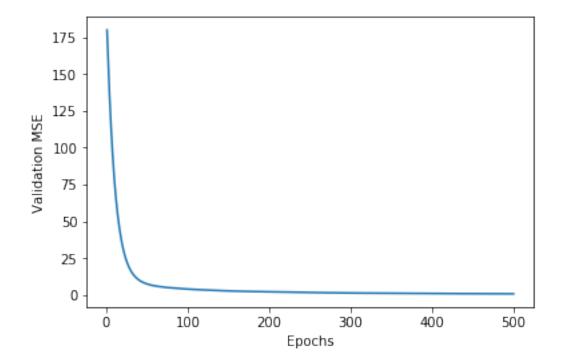
1. Initial model

```
In [1]: from keras.datasets import boston_housing
        (train_data, train_targets), (test_data, test_targets) = boston_housing.load_data()
       mean = train_data.mean(axis=0)
       train_data -= mean
       std = train_data.std(axis=0)
       train_data /= std
       test_data -= mean
       test_data /= std
c:\users\liaoa\appdata\local\programs\python\python36\lib\site-packages\h5py\__init__.py:36: F
  from ._conv import register_converters as _register_converters
Using TensorFlow backend.
Downloading data from https://s3.amazonaws.com/keras-datasets/boston_housing.npz
57344/57026 [============ ] - Os 3us/step
In [2]: from keras import models
       from keras import layers
       def build_model():
           model = models.Sequential()
           model.add(layers.Dense(64, activation='relu', input_shape=(train_data.shape[1],)))
```

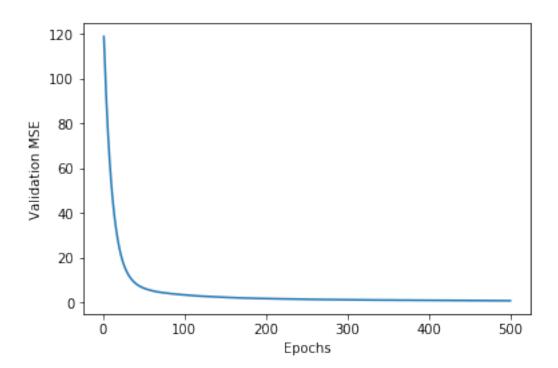
```
model.add(layers.Dense(64, activation='relu'))
           model.add(layers.Dense(1))
            model.compile(optimizer='rmsprop', loss='mse', metrics=['mse'])
            return model
In [15]: import numpy as np
         k = 10
         num_val_samples = len(train_data) // k
         num_epochs = 500
In [7]: def train_model():
           all_mse_histories = []
            for i in range(k):
                print('processing fold #', i)
                val_data = train_data[i * num_val_samples: (i + 1) * num_val_samples]
                val_targets = train_targets[i * num_val_samples: (i + 1) * num_val_samples]
                partial_train_targets = np.concatenate([train_targets[:i * num_val_samples], train_targets[:i * num_val_samples], train_targets[:i * num_val_samples]
                model = build_model()
                history = model.fit(partial_train_data, partial_train_targets, validation_data
                mse_history = history.history['mean_squared_error']
                all_mse_histories.append(mse_history)
            return all_mse_histories
In [13]: import matplotlib.pyplot as plt
         %matplotlib inline
         def smooth_curve(points, factor=0.9):
             smoothed_points = []
             for point in points:
                 if smoothed_points:
                     previous = smoothed_points[-1]
                     smoothed_points.append(previous * factor + point * (1 - factor))
                 else:
                     smoothed_points.append(point)
             return smoothed_points
         def plot_curve(average_mse_history):
             smooth_mse_history = smooth_curve(average_mse_history)
             plt.plot(range(1, len(smooth_mse_history) + 1), smooth_mse_history)
             plt.xlabel('Epochs')
             plt.ylabel('Validation MSE')
             plt.show()
In [17]: all_mse_histories_org = train_model()
         average_mse_history_org = [np.mean([x[n] for x in all_mse_histories_org]) for n in rate
processing fold # 0
processing fold # 1
```

```
processing fold # 2
processing fold # 3
processing fold # 4
processing fold # 5
processing fold # 6
processing fold # 7
processing fold # 8
processing fold # 9
```

In [18]: plot\_curve(average\_mse\_history\_org)

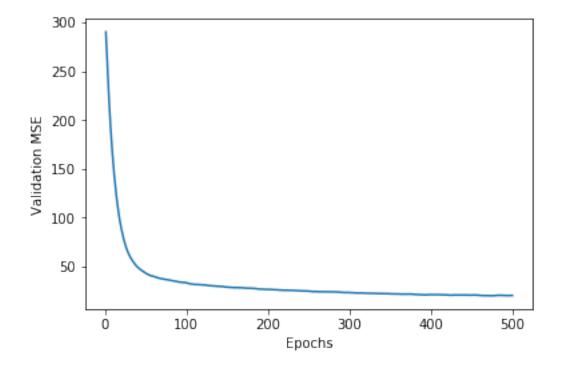


```
102/102 [======== ] - Os Ous/step
test_mse_score 13.935257780785655
   Try on different models - Adding a dense layer
In [25]: def build_model():
            model = models.Sequential()
            model.add(layers.Dense(64, activation='relu', input_shape=(train_data.shape[1],))
             model.add(layers.Dense(64, activation='relu'))
            model.add(layers.Dense(64, activation='relu'))
            model.add(layers.Dense(1))
             model.compile(optimizer='rmsprop', loss='mse', metrics=['mse'])
             return model
In [26]: all_mse_histories_more = train_model()
         average_mse_history_more = [np.mean([x[i] for x in all_mse_histories_more]) for i in :
processing fold # 0
processing fold # 1
processing fold # 2
processing fold # 3
processing fold # 4
processing fold # 5
processing fold # 6
processing fold # 7
processing fold # 8
processing fold # 9
In [27]: plot_curve(average_mse_history_more)
```



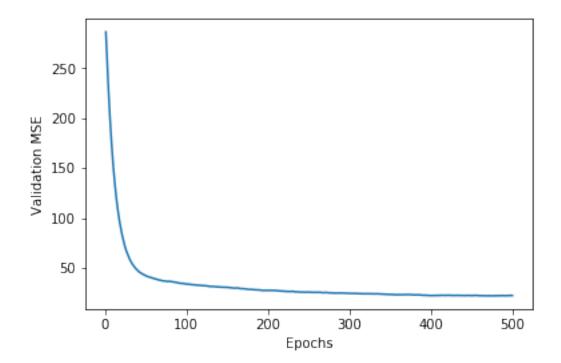
```
model.add(layers.Dropout(0.5))
             model.add(layers.Dense(64, activation='relu'))
             model.add(layers.Dropout(0.5))
             model.add(layers.Dense(64, activation='relu'))
             model.add(layers.Dropout(0.5))
             model.add(layers.Dense(1))
             model.compile(optimizer='rmsprop', loss='mse', metrics=['mse'])
             return model
In [36]: all_mse_histories_dropout = train_model()
         average_mse_history_dropout = [np.mean([x[i] for x in all_mse_histories_dropout]) for
processing fold # 0
processing fold # 1
processing fold # 2
processing fold # 3
processing fold # 4
processing fold # 5
processing fold # 6
processing fold # 7
processing fold # 8
processing fold # 9
```

In [37]: plot\_curve(average\_mse\_history\_dropout)



```
In [38]: model_dropout = build_model()
In [42]: model_dropout.fit(train_data, train_targets, epochs=500, batch_size=16, verbose=0)
        test_mse_score, test_mse_score = model_dropout.evaluate(test_data, test_targets)
        print("test_mse_score", test_mse_score)
102/102 [========== ] - 0s 49us/step
test_mse_score 19.377425848268995
In [40]: model_dropout.fit(train_data, train_targets, epochs=500, batch_size=64, verbose=0)
        test_mse_score, test_mse_score = model_dropout.evaluate(test_data, test_targets)
        print("test_mse_score", test_mse_score)
102/102 [========= ] - Os 67us/step
test_mse_score 17.436082578172872
  Try on different models - Adding penalty
In [46]: from keras import regularizers
        def build_model():
            model = models.Sequential()
            model.add(layers.Dense(64, activation='relu', input_shape=(train_data.shape[1],),
            model.add(layers.Dropout(0.5))
            model.add(layers.Dense(64, activation='relu', kernel_regularizer=regularizers.12(
            model.add(layers.Dropout(0.5))
            model.add(layers.Dense(64, activation='relu', kernel_regularizer=regularizers.12(
            model.add(layers.Dropout(0.5))
            model.add(layers.Dense(1))
            model.compile(optimizer='rmsprop', loss='mse', metrics=['mse'])
            return model
In [47]: all_mse_histories_penalty = train_model()
        average_mse_history_penalty = [np.mean([x[i] for x in all_mse_histories_penalty]) for
processing fold # 0
processing fold # 1
processing fold # 2
processing fold # 3
processing fold # 4
processing fold # 5
processing fold # 6
processing fold # 7
processing fold # 8
processing fold # 9
```

In [48]: plot\_curve(average\_mse\_history\_penalty)



#### The best model is here!