

CNN-based Face Recognition

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Problem Description



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MTCNN



My Work

```
cnn = CNN(img_width=img_width, input_size=img_width ** 2, output_size=num)

x_image = cnn.format_input()

conv_output1 = cnn.conv(x_image, 1, kernel_size1, feature_map1_num, 'conv1')
conv_output2 = cnn.conv(conv_output1, feature_map1_num, kernel_size2, feature_map2_num, 'conv2')
conv_output3 = cnn.conv(conv_output2, feature_map2_num, kernel_size3, feature_map3_num, 'conv3')

conv_output3_flat, flat_size = cnn.flat(conv_output3, feature_map3_num)

fc_output1 = cnn.fc(conv_output3_flat, flat_size, fully_connect_size, 'fc1')
y_conv = cnn.output_layer(fc_output1, fully_connect_size, num, 'fc2')

cnn.train()
```

Results

	人脸朝向	人脸表情	人脸名字	是否戴太阳镜	FaceScrub
CNN	98.2%	18.7%	98.3%	94.5%	76%~82%
ANN	92%	17%	97.5%	87%	

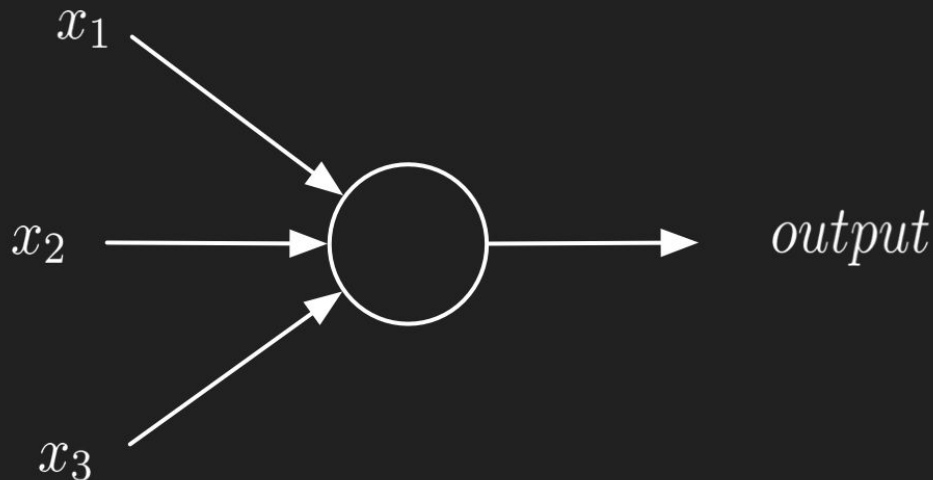
Procedure

1. choosing model
2. data preprocessing
3. training network
4. improve the profermance



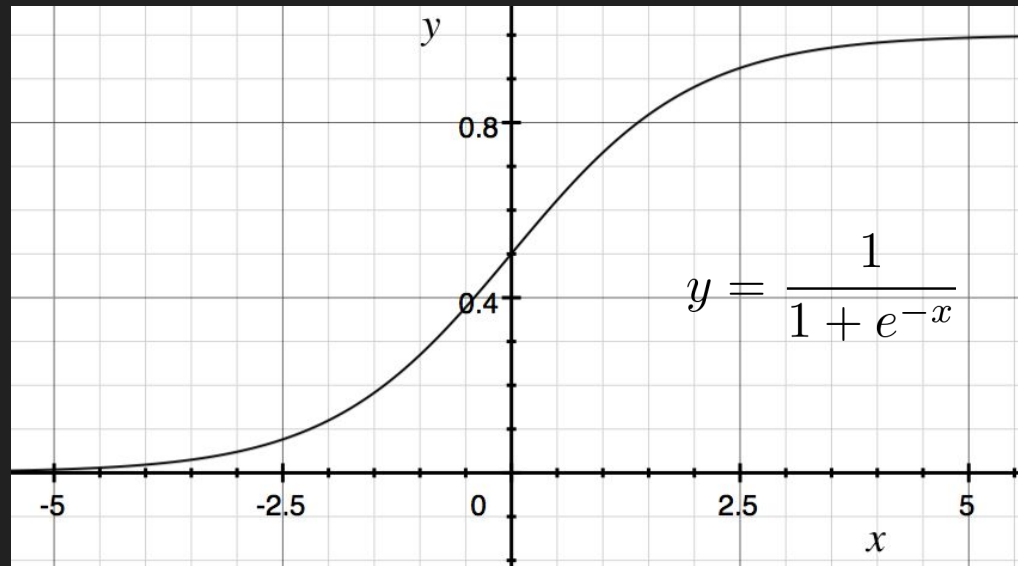
Neural Network

Perceptrons

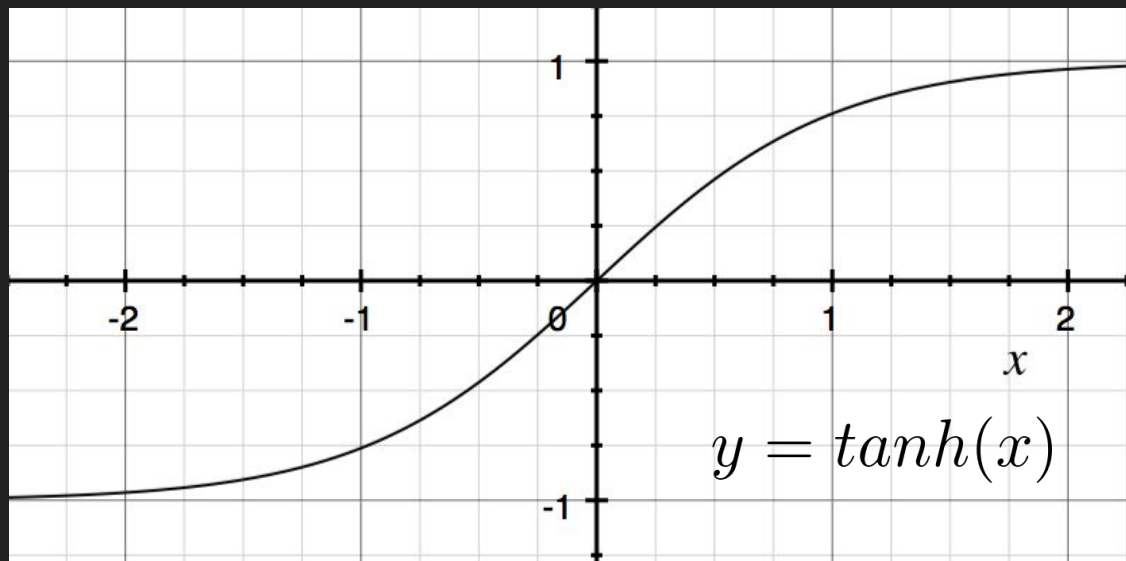


$$output = f(x_1, x_2, x_3) = f\left(\sum_{i=1}^n w_i x_i - \theta\right)$$

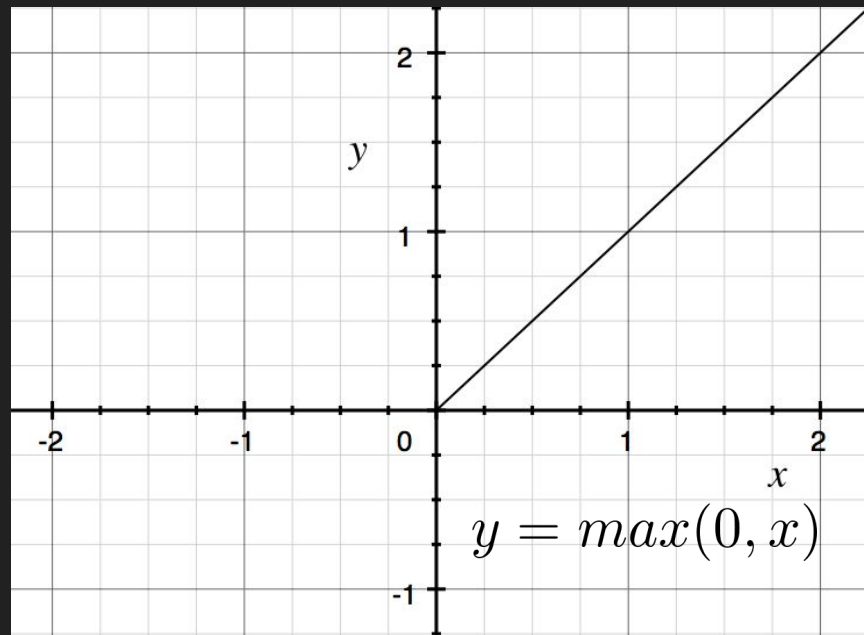
Activation Function



Activation Function

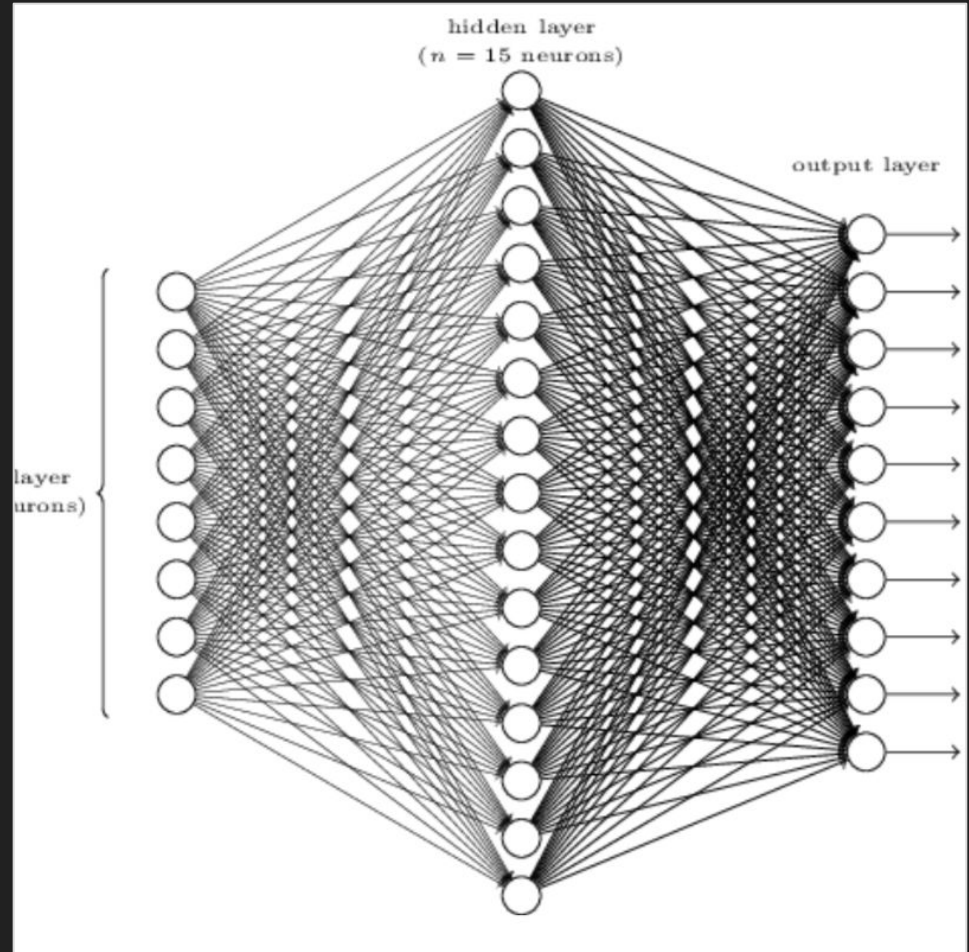


Activation Function



ANN

1. Imitating biological neuron structure
2. Fully connected



Loss Function

$$C = \frac{1}{2n} \sum_x ||y - a||^2$$

$$C = -\frac{1}{n} \sum_x [y \ln a + (1 - y) \ln(1 - a)]$$

y is the label of data, and a is the prediction of network

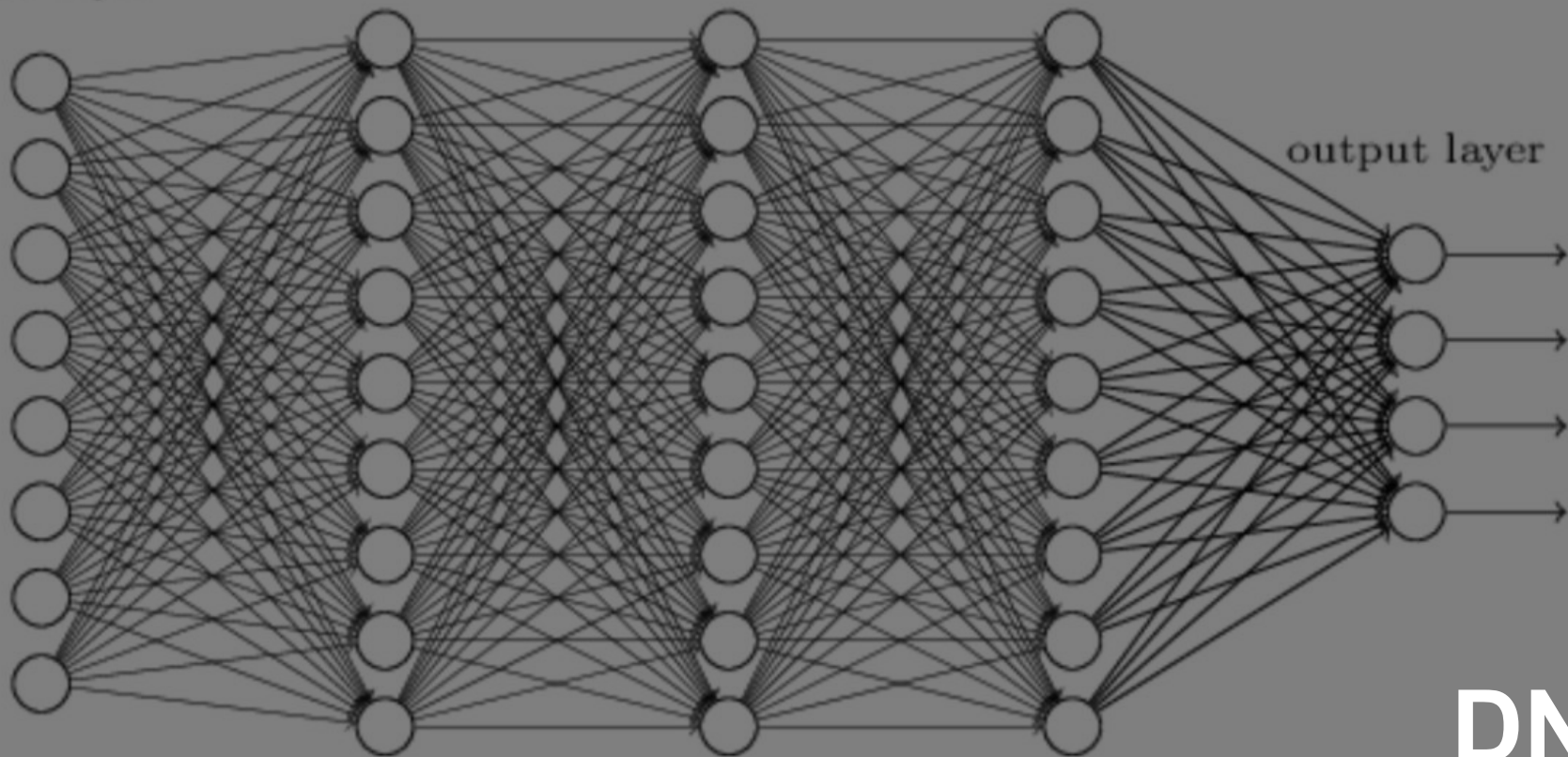
input layer

hidden layer 1

hidden layer 2

hidden layer 3

output layer

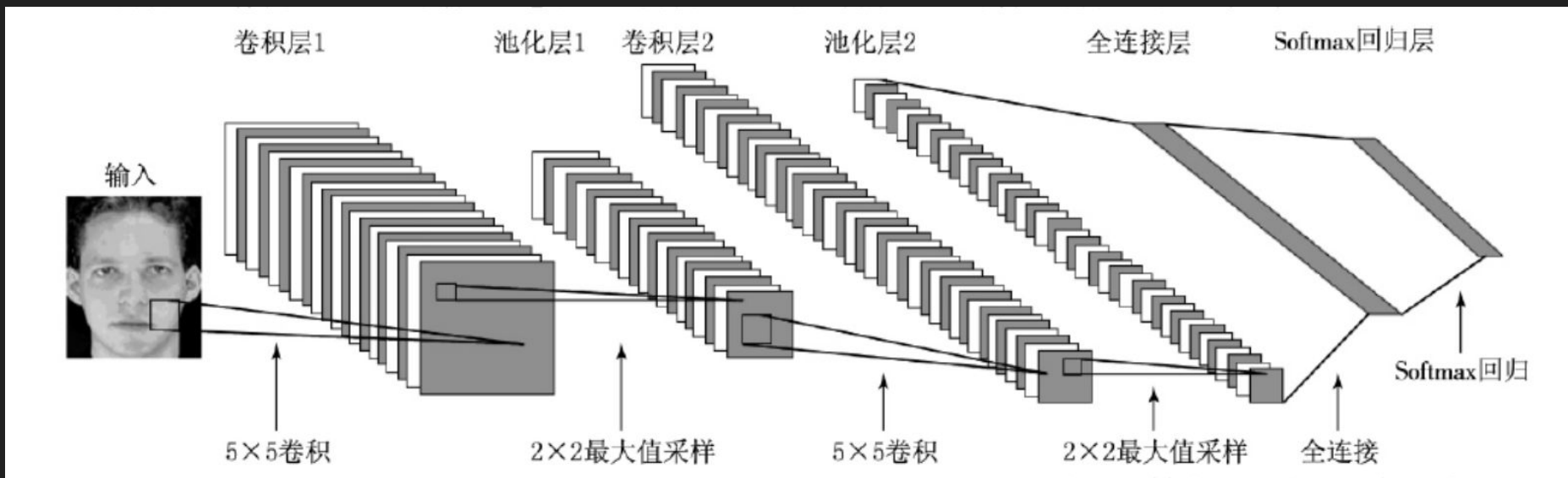


DN

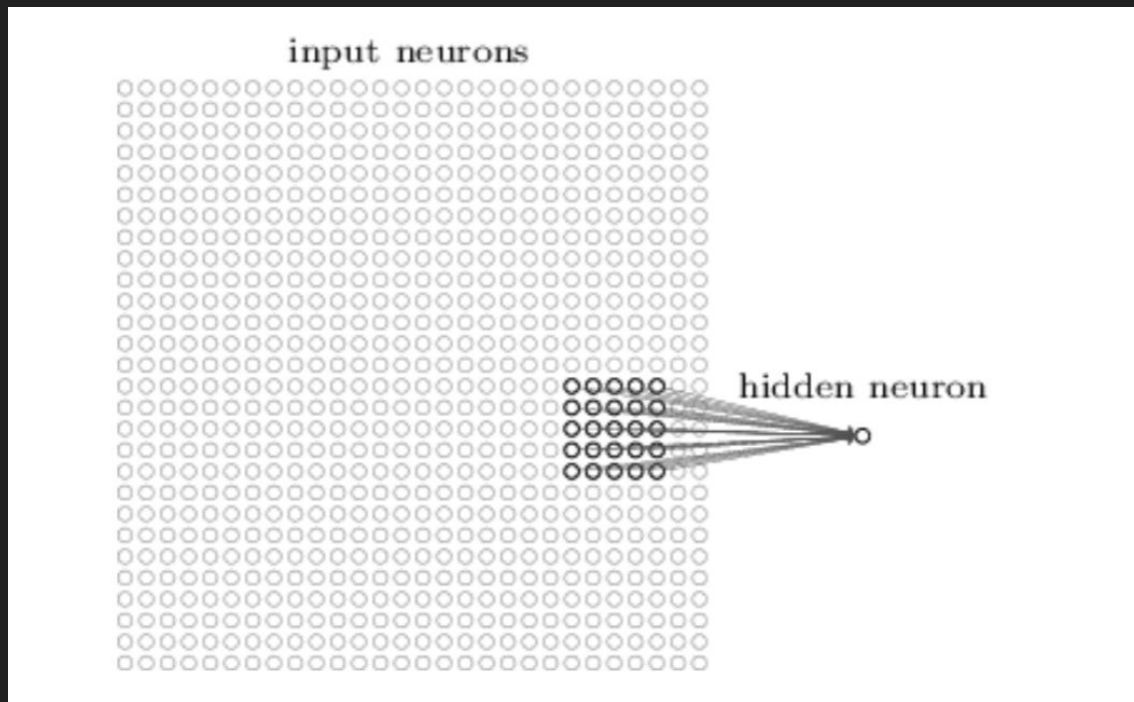
The background features three overlapping circles of varying shades of gray, creating a layered effect. The circles are positioned on the right side of the frame, with the darkest circle on the left and the lightest on the right.

Convolutional Neural Network

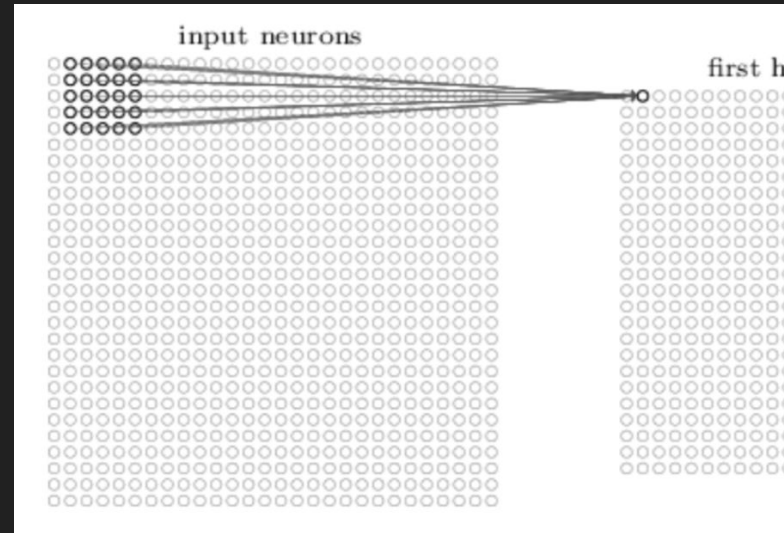
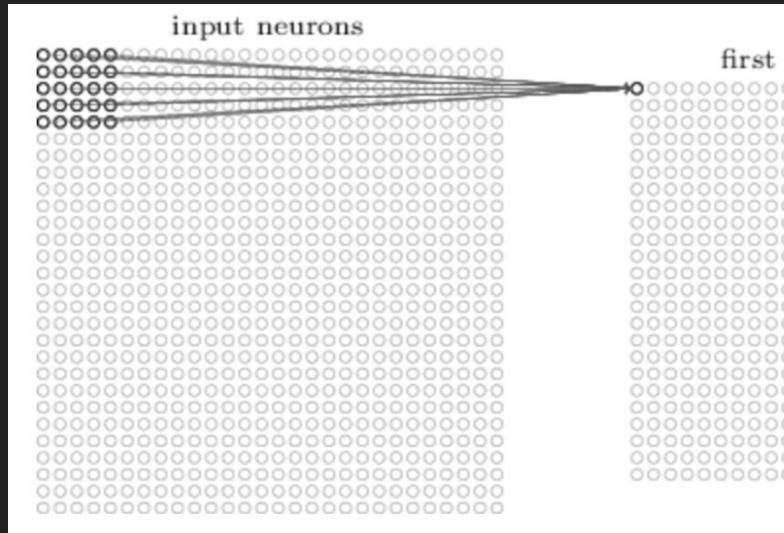
Convolutional Neural Network



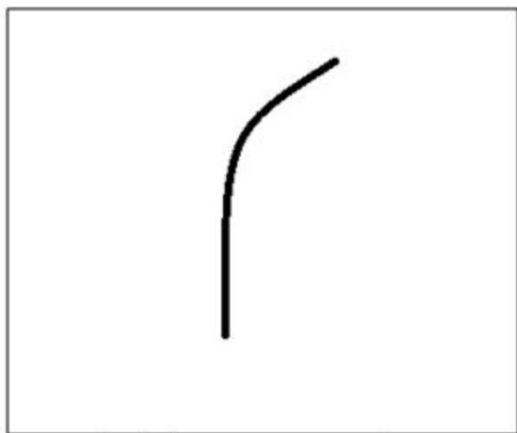
local receptive fields



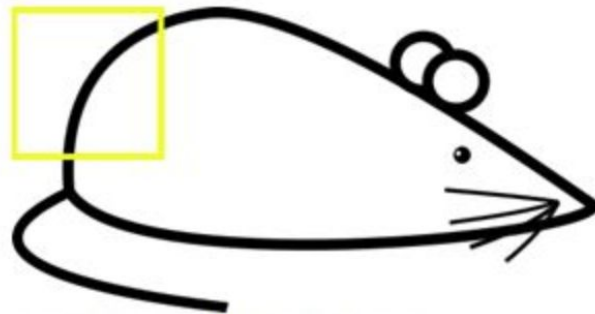
kernel



convolution



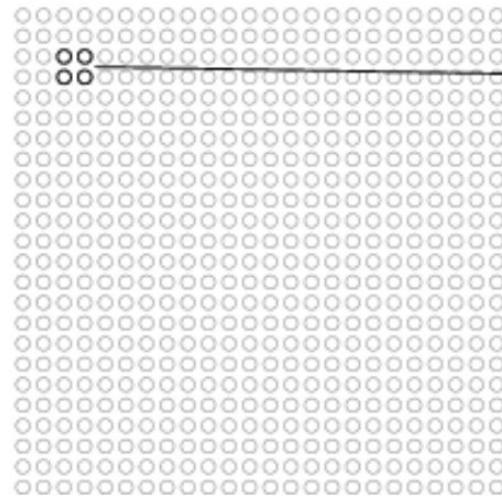
Visualization of a curve detector filter



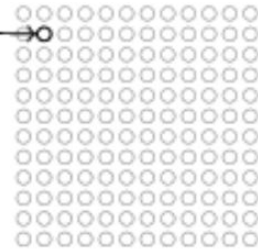
Visualization of the filter on the image

pooling

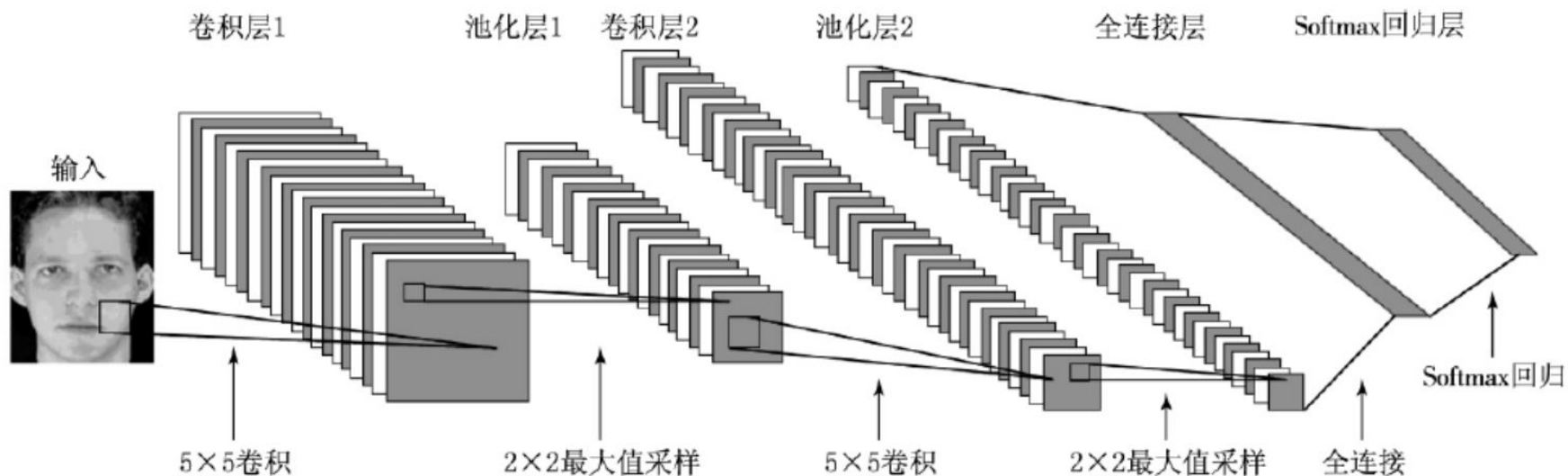
hidden neurons (output from feature map)



max-pooling units



architecture





Computing Graph Model

TensorFlow



Data Preprocessing

1.



Data Preprocessing

1. Extract faces from input pictures
2. Change face images to grayscale images
3. Mapping the pixel values of the images to the $[-1, 1]$ interval
4. Whitening(unfinished)
5. Compress the image to one dimension



Training Network

1.



TensorBoard

loss, accuracy





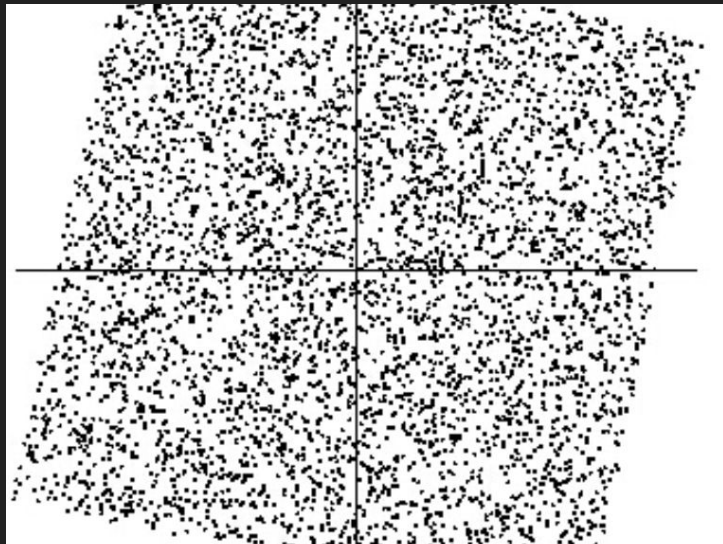
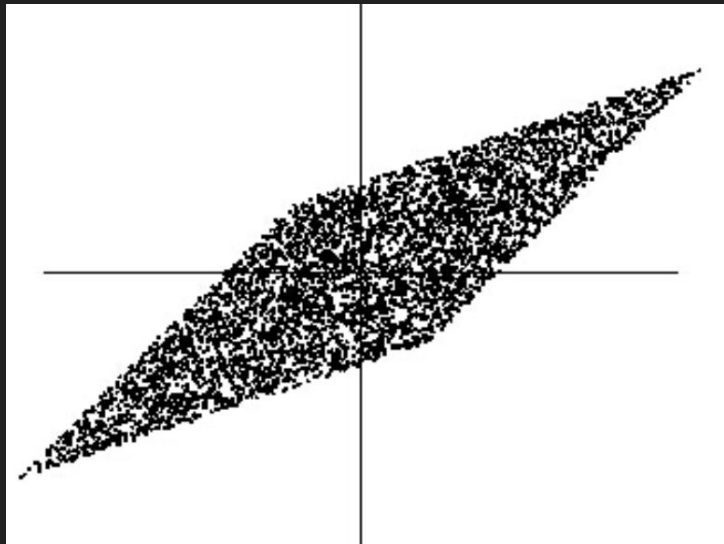
Improve the Profermance



Learning Rate



AdamOptimizer



Whitening

Batch Normalization

Whitening every layer approximately in the network



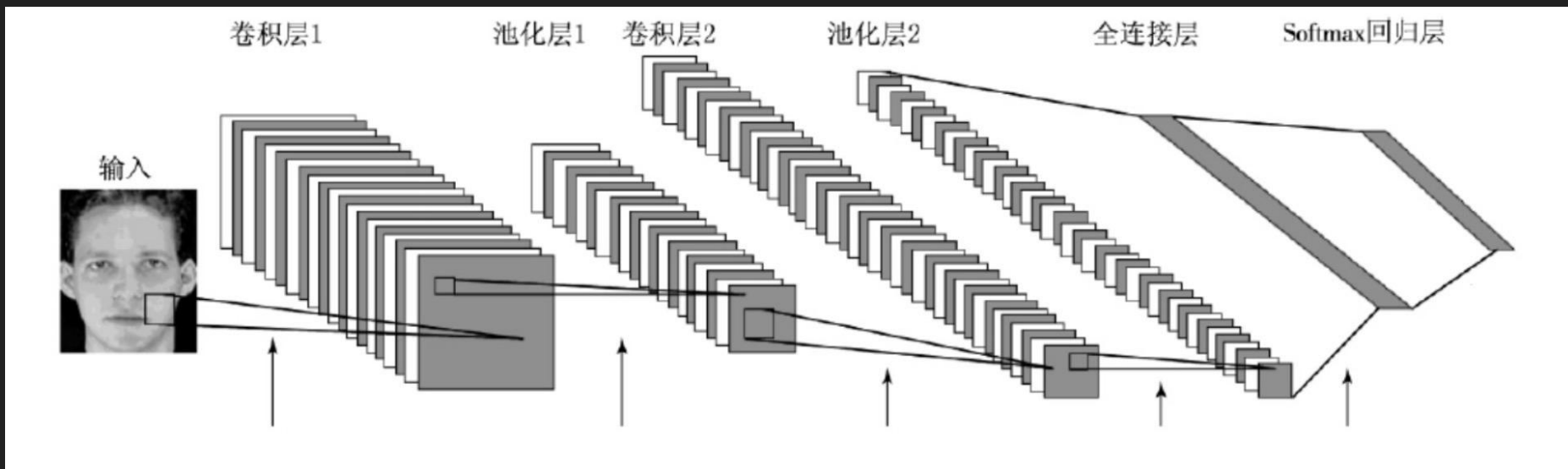
Batch Normalization

$$\mu_B \leftarrow \frac{1}{m} \sum_{i=1}^m x_i$$

$$\sigma_B^2 \leftarrow \frac{1}{m} \sum_{i=1}^m (x_i - \mu_B)^2$$

$$\hat{x}_i \leftarrow \frac{x_i - \mu_B}{\sqrt{\sigma_B^2 + \epsilon}}$$

$$y_i \leftarrow \gamma \hat{x}_i + \beta \equiv BN_{\gamma, \beta}(x_i)$$



Batch Normalization

Distribution



WHY

神经网络喜欢独立同分布的数据，更喜欢白化后的数据

BN可以加快训练速度，可以防止过拟合，可以更快地收敛

Reference

Michael A. Nielsen, "Neural Networks and Deep Learning", Determination Press, 2015

Sergey Ioffe, Christian Szegedy, "Batch Normalization: Accelerating Deep Network Training by Reducing Internal Covariate Shift", 2015

Thank you