Build a CNN model of facial recognition

We use upper face images as training/validation dataset. The objective is to recognize masked faces face_recognition_CNN_n_squre_02.ipynb for processing non-square images 2021.01.30

Shengping Jiang

```
In [1]:
         1 # face recognition CNN n squre 02.ipynb for processing non-square in
          2 # 2021.01.30
         3 # Shengping Jiang
            import numpy as np
            import pandas as pd
            import os
         7
         8 import re
         9 import matplotlib.pyplot as plt
         10 import cv2
         11 import random
         12 from keras.utils import to_categorical
         13 from keras.layers import Dense, Conv2D, Flatten, MaxPool2D, Dropout
         14 from keras.models import Sequential
         15 from keras.models import load model
         16 from keras.optimizers import SGD
         17 from sklearn.model selection import train test split
         18 from keras import backend as K
         19
         20 np.random.seed(1)
```

Using TensorFlow backend.

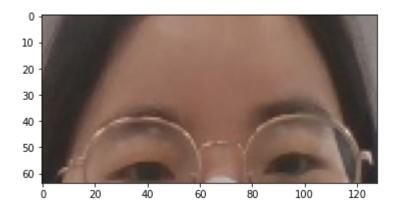
```
In [2]:
          1 # Process images before loading
            extension = ['jpg','png','bmp','jpeg']
          2
          3
            def load data(path name):
          4
                images = []
          5
                labels = []
          6
                class names = []
          7
                 for (root,dirs,files) in os.walk(path_name):
          8
                     #print('root:',root)
          9
                     #We only process images under ./data/train2
         10
                     pattern = '^\w+/train2/'
         11
                     if re.match(pattern, root):
         12
                         print('Read images from:',root)
         13
                         #print('files:',files)
                         #label = root.split('/')[-1]
         14
                         for img in files:
         15
                             img = img.lower()
         16
         17
                             if img.split('.')[1] in extension:
         18
                                 full path = os.path.join(root, img)
         19
                                 #print('full_path:',full_path)
         20
                                 image = cv2.imread(full_path)
                                 image = cv2.cvtColor(image, cv2.COLOR BGR2RGB)
         21
         22
                                 #print('image.shape, img_rows, img_cols:', image
         23
                                 label = full path.split('/')[-2]
         24
                                 # Add image and label to images and lanels
         25
                                 images.append(image)
         26
                                 labels.append(label)
         27
                             #no train folder
                     else:
         28
                         print("Not include folder:",root)
         29
                         #break
         30
                # Converting images to numpy
         31
                images = np.array(images)
         32
                #Change image data type as float and nomarize values to [0~1] fl
         33
                images = images.astype('float32')
         34
                images /= 255
         35
         36
                # This will get how many names in the labels (nb classes)
         37
                class names = pd.get dummies(labels).columns
         38
                nb classes = len(pd.get dummies(labels).columns)
         39
         40
                # Create a name dictionary for prediction
         41
                output = {i:class_names[i] for i in range(len(class_names))}
         42
         43
                # Convert class name to binary value
         44
                labels = pd.get dummies(labels).values
         45
                # change binary values to float from integer
         46
                labels = labels.astype('float32')
         47
                print('image number:', len(images))
         48
                 return images, labels, nb classes, output
         49
         50
```

```
In [3]:
         1 # Example of pd.get dummies()
            names =['Caojun','Shengping','Shengping','Benyuan','Benyuan','Chensh
         3 bivalue = pd.get dummies(names).columns
            print(pd.get dummies(names))
            print(bivalue)
           Benyuan
                    Caojun
                            Chenshu
                                     Shengping
        0
                 0
                         1
                                  0
        1
                 0
                         0
                                  0
                                              1
        2
                 0
                         0
                                  0
                                              1
        3
                 1
                         0
                                  0
                                              0
        4
                 1
                         0
                                  0
                                              0
        5
                 0
                         0
                                  1
                                              0
        6
                 0
                         0
                                  1
                                              0
        Index(['Benyuan', 'Caojun', 'Chenshu', 'Shengping'], dtype='object')
In [4]:
         1 #准备训练与验证数据
          2
            train path = 'data'
         3 \mid \#IMAGE\_SIZE = 64
           images, labels, nb classes, output = load data(train path)
         5
            # Separate images and labels to training group and validation group
           train images, valid images, train labels, valid labels \
          7
                         = train test split(images, labels, test size = 0.2, rar
        Not include folder: data
        Not include folder: data/train2
        Read images from: data/train2/蔡江宸
        Read images from: data/train2/胡潘
        Read images from: data/train2/吴玟君
        Read images from: data/train2/刘璐
        Read images from: data/train2/汤振
        Read images from: data/train2/罗卫
        Not include folder: data/train
        Not include folder: data/train/黄振
        Not include folder: data/train/胡潘
        Not include folder: data/train/刘璐
        Not include folder: data/train/曾子寅
        image number: 124
In [5]:
         1 print('train images.shape:',train images.shape)
          2
            print(type(train images))
         3
            print("train images[0].shape:",train images[0].shape)
            #print("train_labels:",train_labels)
            print("nb classes:",nb classes)
            print("output:",output)
        train images.shape: (99, 64, 128, 3)
        <class 'numpy.ndarray'>
        train images[0].shape: (64, 128, 3)
        nb classes: 6
        output: {0: '刘璐', 1: '吴玟君', 2: '汤振', 3: '罗卫', 4: '胡潘', 5: '蔡江
        宸'}
```

```
In [6]: 1 # Visualizing Training data
2 print(train_labels[1])
3 plt.imshow(train_images[1])
```

[1. 0. 0. 0. 0. 0.]

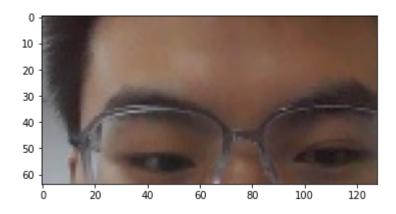
Out[6]: <matplotlib.image.AxesImage at 0x7f4e6ac688d0>



```
In [7]: 1 # Visualizing Training data
2 print(valid_labels[15])
3 plt.imshow(valid_images[15])
```

[0. 0. 1. 0. 0. 0.]

Out[7]: <matplotlib.image.AxesImage at 0x7f4e6ab58748>



```
def model build(inputshape, nb classes):
In [8]:
         1
         2
                model = Sequential()
         3
         4
                #以下代码将顺序添加CNN网络需要的各层,一个add就是一个网络层
         5
                # 1 卷积层1
         6
                model.add(Conv2D(filters=32, kernel size=(3, 3), activation='rel
         7
                # 2 卷积层2
                model.add(Conv2D(filters=32, kernel size=(3, 3), activation='re')
         8
         9
                # 3 池化层1
        10
                model.add(MaxPool2D(pool size=(2, 2)))
                # 4 Dropout层1
        11
                model.add(Dropout(0.25))
        12
        13
                # 5 卷积层3
        14
                model.add(Conv2D(filters=64, kernel size=(3, 3), activation='rel
        15
                # 6 卷积层4
        16
                model.add(Conv2D(filters=64, kernel size=(3, 3), activation='rel
        17
                # 7 池化层2
        18
                model.add(MaxPool2D(pool size=(2, 2)))
        19
                # 8 Dropout层2
        20
                model.add(Dropout(0.25))
                # 9 平化层1
        21
        22
                model.add(Flatten())
        23
                #10 全连接层1
        24
                model.add(Dense(512, activation='relu'))
        25
                # 11 Dropout层3
        26
                model.add(Dropout(0.5))
                # 12 全连接层2. 分类层,输出最终结果
        27
        28
                model.add(Dense(nb classes, activation = 'softmax'))
        29
                return model
```

```
In [9]:
          1
            #Define additional functions
          2
          3
            def save model(model, model path):
          4
                model.save(model path)
          5
          6
            def loadmodel(model path):
                model = load model(model path)
          7
          8
                return model
          9
            def evalu model(model, valid images, valid labels):
         10
                score = model.evaluate(valid images, valid labels, verbose = 1)
         11
         12
                print("%s: %.2f%%" % (model.metrics names[1], score[1] * 100))
         13
```

```
1 # Train the model
In [10]:
            # Images have been resized to 64 x 128 (h x w)
          2
          3
            inputshape = (64, 128, 3)
            # 训练样本的组数(图像样本的人数)
            #nb classes is calculated by load data()
            model = model build(inputshape, nb classes)
            model.summary()
             sqd = SGD(lr = 0.001, decay = 1e-6, momentum = 0.9, nesteroy = True)
            model.compile(loss='categorical crossentropy', optimizer=sgd, metric
            # 加载训练及验证样本,启动训练
            history = model.fit(train images, train labels, batch size = 12, epo
         11
         12
                                       validation_data = (valid_images, valid_land)
         13 # 保存训练模型
         14
            model path = './shengping face model2.h5'
             save model(model, model path)
         15
         16
```

WARNING:tensorflow:From /home/simon/.virtualenvs/codexe/lib/python3. 6/site-packages/keras/backend/tensorflow_backend.py:47: The name tf.g et_default_graph is deprecated. Please use tf.compat.v1.get_default_g raph instead.

WARNING:tensorflow:From /home/simon/.virtualenvs/codexe/lib/python3. 6/site-packages/keras/backend/tensorflow_backend.py:349: The name tf. placeholder is deprecated. Please use tf.compat.v1.placeholder instea d.

WARNING:tensorflow:From /home/simon/.virtualenvs/codexe/lib/python3. 6/site-packages/keras/backend/tensorflow_backend.py:3147: The name t f.random uniform is deprecated. Please use tf.random.uniform instead.

WARNING:tensorflow:From /home/simon/.virtualenvs/codexe/lib/python3. 6/site-packages/keras/backend/tensorflow_backend.py:3014: The name t f.nn.max pool is deprecated. Please use tf.nn.max pool2d instead.

WARNING:tensorflow:From /home/simon/.virtualenvs/codexe/lib/python3. 6/site-packages/keras/backend/tensorflow_backend.py:2683: calling dro pout (from tensorflow.python.ops.nn_ops) with keep_prob is deprecated and will be removed in a future version.

Instructions for updating:

Please use `rate` instead of `keep_prob`. Rate should be set to `rate = 1 - keep prob`.

WARNING:tensorflow:From /home/simon/.virtualenvs/codexe/lib/python3. 6/site-packages/keras/backend/tensorflow_backend.py:1062: calling red uce_prod_v1 (from tensorflow.python.ops.math_ops) with keep_dims is d eprecated and will be removed in a future version.

Instructions for updating:

keep dims is deprecated, use keepdims instead

Layer (type)	Output Shape	Param #
conv2d_1 (Conv2D)	(None, 64, 128, 32)	896
conv2d_2 (Conv2D)	(None, 62, 126, 32)	9248

<pre>max_pooling2d_1 (MaxPooling2</pre>	(None, 31, 63, 32)	0
dropout_1 (Dropout)	(None, 31, 63, 32)	0
conv2d_3 (Conv2D)	(None, 31, 63, 64)	18496
conv2d_4 (Conv2D)	(None, 29, 61, 64)	36928
max_pooling2d_2 (MaxPooling2	(None, 14, 30, 64)	0
dropout_2 (Dropout)	(None, 14, 30, 64)	0
flatten_1 (Flatten)	(None, 26880)	0
dense_1 (Dense)	(None, 512)	13763072
dropout_3 (Dropout)	(None, 512)	0
dense_2 (Dense)	(None, 6)	3078

Total params: 13,831,718.0 Trainable params: 13,831,718.0 Non-trainable params: 0.0

WARNING:tensorflow:From /home/simon/.virtualenvs/codexe/lib/python3. 6/site-packages/keras/optimizers.py:675: The name tf.train.Optimizer is deprecated. Please use tf.compat.v1.train.Optimizer instead.

WARNING:tensorflow:From /home/simon/.virtualenvs/codexe/lib/python3. 6/site-packages/keras/backend/tensorflow_backend.py:2550: calling red uce_sum_v1 (from tensorflow.python.ops.math_ops) with keep_dims is de precated and will be removed in a future version.

Instructions for updating:

keep dims is deprecated, use keepdims instead

WARNING:tensorflow:From /home/simon/.virtualenvs/codexe/lib/python3. 6/site-packages/keras/backend/tensorflow_backend.py:2554: The name t f.log is deprecated. Please use tf.math.log instead.

WARNING:tensorflow:From /home/simon/.virtualenvs/codexe/lib/python3. 6/site-packages/tensorflow_core/python/ops/math_grad.py:1424: where (from tensorflow.python.ops.array_ops) is deprecated and will be remo ved in a future version.

Instructions for updating:

Use tf.where in 2.0, which has the same broadcast rule as np.where WARNING:tensorflow:Variable *= will be deprecated. Use `var.assign(var * other)` if you want assignment to the variable value or `x = x * y` if you want a new python Tensor object.

WARNING:tensorflow:From /home/simon/.virtualenvs/codexe/lib/python3. 6/site-packages/keras/backend/tensorflow_backend.py:766: The name tf. assign add is deprecated. Please use tf.compat.vl.assign add instead.

WARNING:tensorflow:From /home/simon/.virtualenvs/codexe/lib/python3. 6/site-packages/keras/backend/tensorflow_backend.py:519: calling Constant.__init__ (from tensorflow.python.ops.init_ops) with dtype is deprecated and will be removed in a future version.

Instructions for updating:

Call initializer instance with the dtype argument instead of passing

it to the constructor WARNING:tensorflow:From /home/simon/.virtualenvs/codexe/lib/python3. 6/site-packages/keras/backend/tensorflow_backend.py:762: The name tf. assign is deprecated. Please use tf.compat.v1.assign instead.

Train on 99 samples, validate on 25 samples Epoch 1/30

WARNING:tensorflow:From /home/simon/.virtualenvs/codexe/lib/python3. 6/site-packages/keras/backend/tensorflow_backend.py:140: The name tf. get_default_session is deprecated. Please use tf.compat.v1.get_default_session instead.

WARNING:tensorflow:From /home/simon/.virtualenvs/codexe/lib/python3. 6/site-packages/keras/backend/tensorflow_backend.py:145: The name tf. ConfigProto is deprecated. Please use tf.compat.v1.ConfigProto instea d.

WARNING:tensorflow:From /home/simon/.virtualenvs/codexe/lib/python3. 6/site-packages/keras/backend/tensorflow_backend.py:150: The name tf. Session is deprecated. Please use tf.compat.v1.Session instead.

WARNING:tensorflow:From /home/simon/.virtualenvs/codexe/lib/python3. 6/site-packages/keras/backend/tensorflow_backend.py:298: The name tf. global_variables is deprecated. Please use tf.compat.v1.global_variables instead.

WARNING:tensorflow:From /home/simon/.virtualenvs/codexe/lib/python3. 6/site-packages/keras/backend/tensorflow_backend.py:306: The name tf. variables_initializer is deprecated. Please use tf.compat.v1.variable s_initializer instead.

```
99/99 [============= ] - 1s - loss: 1.7843 - acc: 0.1
414 - val_loss: 1.7776 - val_acc: 0.3200
Epoch 2/30
717 - val loss: 1.7694 - val acc: 0.3200
Epoch 3/30
99/99 [============ ] - 1s - loss: 1.7447 - acc: 0.3
030 - val loss: 1.7534 - val acc: 0.3200
Epoch 4/30
99/99 [=========== ] - 1s - loss: 1.7062 - acc: 0.3
636 - val loss: 1.7202 - val acc: 0.4000
Epoch 5/30
343 - val loss: 1.6627 - val acc: 0.5200
Epoch 6/30
747 - val loss: 1.5376 - val acc: 0.5200
Epoch 7/30
99/99 [============ ] - 1s - loss: 1.3865 - acc: 0.5
152 - val loss: 1.3028 - val acc: 0.5600
Epoch 8/30
99/99 [==============] - 1s - loss: 1.1260 - acc: 0.5
253 - val loss: 1.0171 - val acc: 0.8400
Epoch 9/30
99/99 [============== ] - 1s - loss: 0.9219 - acc: 0.6
869 - val loss: 0.9379 - val acc: 0.5200
```

```
Epoch 10/30
566 - val loss: 0.6766 - val acc: 0.8400
Epoch 11/30
778 - val loss: 0.4910 - val acc: 0.9600
Epoch 12/30
384 - val_loss: 0.4176 - val_acc: 0.9200
Epoch 13/30
99/99 [============= ] - 1s - loss: 0.3914 - acc: 0.8
889 - val loss: 0.3280 - val acc: 0.9200
Epoch 14/30
788 - val loss: 0.3136 - val acc: 0.9600
Epoch 15/\overline{30}
889 - val loss: 0.2292 - val acc: 0.9200
Epoch 16/30
99/99 [============= ] - 1s - loss: 0.2423 - acc: 0.8
990 - val loss: 0.3307 - val acc: 0.9600
Epoch 17/30
586 - val loss: 0.2062 - val acc: 0.9600
Epoch 18/30
192 - val loss: 0.1734 - val_acc: 0.9600
Epoch 19/30
192 - val loss: 0.1689 - val acc: 0.9200
Epoch 20/30
99/99 [=========== ] - 1s - loss: 0.1586 - acc: 0.9
697 - val loss: 0.1307 - val acc: 0.9600
Epoch 21/30
99/99 [========== ] - 1s - loss: 0.1515 - acc: 0.9
596 - val loss: 0.1764 - val acc: 0.9600
Epoch 22/30
99/99 [==============] - 1s - loss: 0.1211 - acc: 0.9
596 - val loss: 0.1323 - val acc: 0.9200
Epoch 23/30
99/99 [==============] - 1s - loss: 0.1258 - acc: 0.9
697 - val loss: 0.1826 - val_acc: 0.9600
Epoch 24/30
899 - val loss: 0.1253 - val acc: 0.9200
Epoch 25/\overline{30}
99/99 [========== ] - 1s - loss: 0.1325 - acc: 0.9
394 - val loss: 0.1251 - val acc: 0.9200
Epoch 26/30
99/99 [==============] - 1s - loss: 0.2402 - acc: 0.9
192 - val loss: 0.1469 - val_acc: 0.9600
Epoch 27/30
99/99 [========== ] - 1s - loss: 0.1223 - acc: 0.9
697 - val_loss: 0.1269 - val_acc: 0.9600
Epoch 28/30
```

Model parameters analysis

```
conv2d: (input_channels 3 x kernel_size 3 x 3+bians 1)x filters 32=896 parameters conv2d_1: (Pre_layer filters 32 x kernel_size 3 x 3+bians 1) x filters 32=9248 parameters conv2d_2: (Pre_layer filters 32 x kernel_size 3 x 3+bians 1) x filters 64=18496 parameters
```

conv2d_3: (Pre_layer filters 64 x kernel_size 3 x 3+bians 1) x filters 64=36928 parameters max_pooling2d_1: output 64 images of 14x30. The flatten transfers the output to 14x30x64=26880 neurons

dense (Dense): (Pre_layer(flatten) outputs unit 26880 +bians 1) x units 512 = 13763072 parameters

dense 2: (Pre layer(dense)outputs unit 512 +bians 1) x units 6 =3078 parameters

Image shape analysis

```
conv2d: input_shape: 64x128, 3 channels. Output: 64x128, 32 channels --Because padding='same', input_shape(64x128)=Output(64x128)
```

```
conv2d_1: input_shape: 64x128, 32 channels. Output: 62x62, 32 channels --Because padding is not defined, default is valid, output_shape = ceil((input_shape - (kernel_size - 1)) / stride) = ((64x128 - ((3,3) - (1,1))) / 1) = 64x128 - 2x2 = 62x126
```

max_pooling2d: input_shape: 62x126, 32 channels. Output: 31x63, 32 channels --Because pool_size=(2,2), no strides, strides=pool_size=(2,2). Not define padding, padding is valid. output_shape = (input_shape-pool_size + 1)/stride = (62x126-2x2+(1,1))/2=61x125/2 (Rounded up)=31x63

conv2d_2: input_shape: 31x63, 32 channels. Output: 31x63, 64 channels --Because padding='same', input_shape(31x63)=Output(31x63)

conv2d_3: input_shape: 31x63, 64 channels. Output: 29x61, 64 channels --Because padding is not given, it is valid. Output_shape = ((31x63 - ((3,3) - (1,1))) / 1) = 31x63 - 2x2 = 29x61

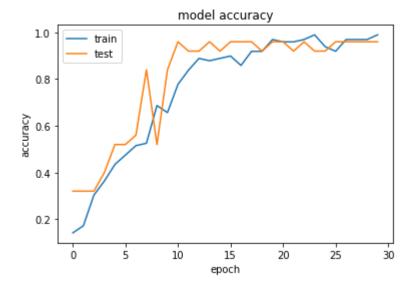
max_pooling2d_1: input_shape: 29x61, 64 channels. Output: 14x30, 64 channels --Because pool_size=(2,2), no strides, strides=pool_size=(2,2). No padding is given, padding is valid. output_shape = (input_shape-pool_size + 1)/stride = (29x61-2x2+ (1,1))/2=28x60/2 (rounded up)=14x30

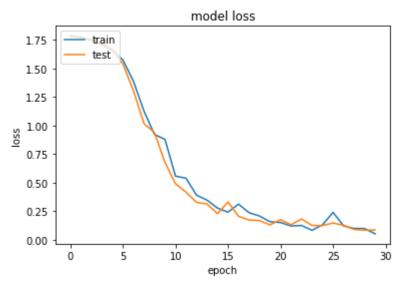
flatten: input_shape: 14x30, 64 channels. Output: 14x30x64=26880, 1 dimension vector --Flatten transfers all units to one dimension vector: pixel (14x30) x channel (64) =26880

dense: input_shape: 26880, one dimension. Output: 512, one dimension --dense connects 26880 units to each output unit. output units are 512

dense_1: input_shape: 512, one dimension. Output: 6, one dimension --dense connects 512 units to each output unit. output units are 6

```
In [11]:
            # 读出保存的模型并进行验证
            model_path = './shengping_face_model2.h5'
          3
            model2 = loadmodel(model path)
            evalu model(model2, valid images, valid labels)
          5
            eva result = model2.evaluate(valid images, valid labels)
            print('eva result:',eva result)
        25/25 [======== ] - 0s
        acc: 96.00%
        25/25 [======== ] - 0s
        eva result: [0.08653917908668518, 0.9599999785423279]
In [12]:
            # summarize history for accuracy
          2
            plt.plot(history.history['acc'])
          3
            plt.plot(history.history['val acc'])
            plt.title('model accuracy')
          5
            plt.ylabel('accuracy')
            plt.xlabel('epoch')
            plt.legend(['train', 'test'], loc='upper left')
            plt.show()
```





```
In [15]:
             # Testing predictions and compare to actual label
           2
           3
             def index_predict(n):
           4
                  checkImage = valid images[n:n+1]
                  checklabel = valid_labels[n:n+1]
           5
                  #print('checkImage:',checkImage)
           6
           7
                  predicts = model.predict(checkImage)
           8
                  print("predicts:",predicts)
           9
                  print("Actual binary, name: ",checklabel[0], output[int(np.argmaxet)]
          10
                  for i in range(len(predicts)):
          11
                      print("Predicted :- ",output[int(np.argmax(predicts[i]))])
          12
          13
                      #print('predict:',predict[i])
```

```
In [16]:
          1 for i in range(len(valid labels)):
                index predict(i)
          2
        predicts: [[9.8787097e-04 8.9721994e-05 3.7342799e-03 3.3260537e-03
        9.8862737e-01
          3.2347268e-0311
        Actual binary, name: [0.0.0.0.1.0.] 胡潘
        Predicted:- 胡潘
        predicts: [[0.01379944 0.0062185 0.0835894 0.172029
                                                              0.6970752 0.
        0272884211
        Actual binary, name: [0.0.0.0.1.0.] 胡潘
        Predicted :- 胡潘
        predicts: [[9.9999774e-01 4.3099405e-07 1.5926870e-09 1.2036265e-14
        3.2242593e-08
           1.7944269e-0611
        Actual binary, name: [1. 0. 0. 0. 0. 0.] 刘璐
        Predicted:- 刘璐
        predicts: [[5.0224871e-03 9.9459672e-01 3.5654526e-07 1.6302594e-05
        9.7832555e-05
          2.6628820e-04]]
        Actual binary, name: [0.1.0.0.0.0.] 吴玟君
        Predicted :- 吴玟君
        predicts: [[1.3158612e-07 9.9999785e-01 3.9044394e-14 1.8744620e-07
        4.1303676e-09
          1.7519769e-0611
        Actual binary, name: [0.1.0.0.0.0.] 吴玟君
        Predicted:- 吴玟君
        predicts: [[7.4822474e-06 1.5423456e-07 7.6317250e-05 6.9135058e-09
        7.3153578e-06
          9.9990869e-0111
        Actual binary, name: [0.0.0.0.0.1.] 蔡江宸
        Predicted :- 蔡江宸
        predicts: [[7.5593976e-05 1.5592390e-04 3.1925845e-01 6.1089249e-04
        6.7039531e-01
          9.5038889e-0311
        Actual binary, name: [0.0.0.0.1.0.] 胡潘
        Predicted :- 胡潘
        predicts: [[5.5492474e-05 4.4090531e-05 2.8060514e-01 1.9161115e-04
         7.1754968e-01
          1.5539433e-0311
        Actual binary, name: [0.0.1.0.0.0.] 汤振
        Predicted :- 胡潘
        predicts: [[5.2539972e-06 1.0214385e-04 5.5464305e-05 9.8813802e-01
         1.1452772e-02
          2.4628398e-0411
        Actual binary, name: [0.0.0.1.0.0.] 罗卫
        Predicted :- 罗卫
        predicts: [[2.0502725e-06 2.9544822e-08 9.9965024e-01 1.6773693e-07
        2.7637204e-04
          7.1231421e-0511
        Actual binary, name: [0.0.1.0.0.0.] 汤振
        Predicted :- 汤振
        predicts: [[9.9999499e-01 4.2410156e-06 9.7286606e-12 2.9399673e-15
        3.6305801e-09
           7.2935194e-0711
        Actual binary, name: [1. 0. 0. 0. 0. 0.] 刘璐
```

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Predicted:- 刘璐
predicts: [[1.3795243e-05 9.9997723e-01 1.2531072e-11 3.0065672e-07
1.2772068e-06
  7.2973521e-0611
Actual binary, name: [0.1.0.0.0.0.] 吴玟君
Predicted :- 吴玟君
predicts: [[7.2040139e-06 1.7859188e-07 9.9615508e-01 2.9848431e-06
3.5697147e-03
  2.6483263e-0411
Actual binary, name: [0.0.1.0.0.0.] 汤振
Predicted :- 汤振
predicts: [[5.0115813e-07 4.8307154e-08 9.9935335e-01 1.0921337e-06
6.0370937e-04
  4.1210744e-0511
Actual binary, name: [0.0.1.0.0.0.] 汤振
Predicted :- 汤振
predicts: [[2.0491282e-06 1.6856793e-06 9.8777175e-01 2.0218498e-05
1.1815689e-02
  3.8861079e-04]]
Actual binary, name: [0.0.1.0.0.0.] 汤振
Predicted :- 汤振
predicts: [[1.8987097e-05 1.2039476e-06 9.8543108e-01 2.5053219e-05
1.4206828e-02
  3.1696452e-0411
Actual binary, name: [0.0.1.0.0.0.] 汤振
Predicted :- 汤振
predicts: [[2.6513455e-06 1.6776983e-04 3.4698495e-04 9.7680640e-01
2.2110252e-02
  5.6606013e-0411
Actual binary, name: [0.0.0.1.0.0.] 罗卫
Predicted:- 罗卫
predicts: [[2.2742568e-06 7.5841999e-05 6.7673514e-05 9.8732007e-01
1.2372936e-02
  1.6110601e-0411
Actual binary, name: [0.0.0.1.0.0.] 罗卫
Predicted:- 罗卫
predicts: [[2.3127021e-05 8.3085788e-06 4.0442496e-06 1.1634216e-06
8.7443390e-04
  9.9908888e-0111
Actual binary, name: [0.0.0.0.0.1.] 蔡江宸
Predicted :- 蔡江宸
predicts: [[5.3416265e-06 2.2987253e-07 2.7305944e-04 2.2093610e-08
3.9871054e-05
  9.9968147e-0111
Actual binary, name: [0.0.0.0.0.1.] 蔡江宸
Predicted :- 蔡江宸
predicts: [[3.8484282e-06 1.1713930e-04 6.6535542e-04 5.0491234e-03
9.9239606e-01
  1.7684379e-0311
Actual binary, name: [0.0.0.0.1.0.] 胡潘
Predicted :- 胡潘
predicts: [[4.3506166e-06 6.5332047e-06 1.9676934e-05 3.0994611e-06
2.8017085e-04
  9.9968624e-0111
Actual binary, name: [0.0.0.0.0.1.] 蔡江宸
Predicted :- 蔡江宸
predicts: [[2.6747308e-05 1.8773864e-06 9.8036915e-01 3.2552882e-05
```

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1.9175127e-02
           3.9447338e-0411
         Actual binary, name:
                              [0.0.1.0.0.0.] 汤振
         Predicted :- 汤振
         predicts: [[9.9999201e-01 6.9059211e-06 2.2567129e-11 7.5456626e-15
         5.7915059e-09
           1.1313166e-0611
         Actual binary, name: [1. 0. 0. 0. 0. 0.] 刘璐
         Predicted:- 刘璐
         predicts: [[7.01824320e-05 1.04334084e-07 3.85002140e-03 1.08676206e-
         98
           2.00488768e-03 9.94074881e-01]]
         Actual binary, name:
                              [0.0.0.0.0.1.] 蔡江宸
         Predicted:- 蔡江宸
             for i in range(len(valid_labels)):
In [17]:
                 print('i, valid labels[i]:',i,valid_labels[i] )
          2
          3
             #plt.imshow(valid images[20])
         i, valid labels[i]: 0 [0. 0. 0. 0. 1. 0.]
         i, valid_labels[i]: 1 [0. 0. 0. 0. 1. 0.]
         i, valid labels[i]: 2 [1. 0. 0. 0. 0. 0.]
         i, valid labels[i]: 3 [0. 1. 0. 0. 0. 0.]
         i, valid labels[i]: 4 [0. 1. 0. 0. 0. 0.]
         i, valid labels[i]: 5 [0. 0. 0. 0. 0. 1.]
         i, valid_labels[i]: 6 [0. 0. 0. 0. 1. 0.]
         i, valid labels[i]: 7 [0. 0. 1. 0. 0. 0.]
         i, valid labels[i]: 8 [0. 0. 0. 1. 0. 0.]
         i, valid labels[i]: 9 [0. 0. 1. 0. 0. 0.]
         i, valid labels[i]: 10 [1. 0. 0. 0. 0. 0.]
         i, valid labels[i]: 11 [0. 1. 0. 0. 0. 0.]
         i, valid labels[i]: 12 [0. 0. 1. 0. 0. 0.]
         i, valid labels[i]: 13 [0. 0. 1. 0. 0. 0.]
         i, valid labels[i]: 14 [0. 0. 1. 0. 0. 0.]
         i, valid labels[i]: 15 [0. 0. 1. 0. 0. 0.]
         i, valid labels[i]: 16 [0. 0. 0. 1. 0. 0.]
         i, valid labels[i]: 17 [0. 0. 0. 1. 0. 0.]
         i, valid_labels[i]: 18 [0. 0. 0. 0. 0. 1.]
         i, valid labels[i]: 19 [0. 0. 0. 0. 0. 1.]
         i, valid labels[i]: 20 [0. 0. 0. 0. 1. 0.]
         i, valid labels[i]: 21 [0. 0. 0. 0. 0. 1.]
         i, valid labels[i]: 22 [0. 0. 1. 0. 0. 0.]
         i, valid_labels[i]: 23 [1. 0. 0. 0. 0. 0.]
         i, valid labels[i]: 24 [0. 0. 0. 0. 0. 1.]
In [ ]:
          1
```