facial keypoints

June 4, 2020

1 Facial Keypoints Detection

The data can be downloaded from https://www.kaggle.com/c/facial-keypoints-detection/data

The notebook was originally a kaggle notebook, the link to which is https://www.kaggle.com/vikrantdey/facial-keypoints

```
[1]: import numpy as np
import pandas as pd

import os
for dirname, _, filenames in os.walk('/kaggle/input'):
    for filename in filenames:
        print(os.path.join(dirname, filename))
```

/kaggle/input/facial-keypoints-detection/IdLookupTable.csv /kaggle/input/facial-keypoints-detection/SampleSubmission.csv /kaggle/input/facial-keypoints-detection/training.zip /kaggle/input/facial-keypoints-detection/test.zip

```
[6]: print("Contents of input/facial-keypoints-detection directory: ")
!ls ../input/facial-keypoints-detection/

print("\nExtracting .zip dataset files to working directory ...")
!unzip -u ../input/facial-keypoints-detection/test.zip
!unzip -u ../input/facial-keypoints-detection/training.zip

print("\nCurrent working directory:")
!pwd
print("\nContents of working directory:")
!ls
```

Contents of input/facial-keypoints-detection directory:
IdLookupTable.csv SampleSubmission.csv test.zip training.zip

Extracting .zip dataset files to working directory ...

Archive: ../input/facial-keypoints-detection/test.zip
inflating: test.csv

```
Archive: ../input/facial-keypoints-detection/training.zip
       inflating: training.csv
     Current working directory:
     /kaggle/working
     Contents of working directory:
     __notebook_source__.ipynb test.csv training.csv
 [2]: from sklearn.model_selection import train_test_split
     from matplotlib import pyplot as plt
     %matplotlib inline
 [3]: | IdLookupTable = pd.read_csv('/kaggle/input/facial-keypoints-detection/
      IdLookupTable.info()
     <class 'pandas.core.frame.DataFrame'>
     RangeIndex: 27124 entries, 0 to 27123
     Data columns (total 4 columns):
         Column
                     Non-Null Count Dtype
     --- -----
                     -----
         RowId
                     27124 non-null int64
      0
      1
                      27124 non-null int64
         ImageId
      2 FeatureName 27124 non-null object
         Location
                      0 non-null
                                     float64
     dtypes: float64(1), int64(2), object(1)
     memory usage: 847.8+ KB
 [4]: IdLookupTable.head()
 [4]:
        RowId ImageId
                                   FeatureName Location
            1
                    1
     0
                             left_eye_center_x
                                                    NaN
     1
            2
                     1
                            left_eye_center_y
                                                    NaN
     2
            3
                    1
                            right eye center x
                                                    NaN
     3
            4
                            right_eye_center_y
                                                     NaN
                     1 left_eye_inner_corner_x
                                                    {\tt NaN}
[10]: training = pd.read_csv('/kaggle/working/training.csv')
     training.info()
     <class 'pandas.core.frame.DataFrame'>
     RangeIndex: 7049 entries, 0 to 7048
     Data columns (total 31 columns):
      # Column
                                    Non-Null Count Dtype
     --- -----
      0 left_eye_center_x
                                   7039 non-null
                                                  float64
                                   7039 non-null float64
         left_eye_center_y
```

```
4
          left_eye_inner_corner_x
                                      2271 non-null
                                                       float64
      5
          left_eye_inner_corner_y
                                      2271 non-null
                                                       float64
      6
          left eye outer corner x
                                      2267 non-null
                                                       float64
      7
          left_eye_outer_corner_y
                                      2267 non-null
                                                       float64
      8
          right eye inner corner x
                                      2268 non-null
                                                       float64
          right_eye_inner_corner_y
                                      2268 non-null
                                                       float64
         right_eye_outer_corner_x
                                      2268 non-null
                                                       float64
      11
          right_eye_outer_corner_y
                                      2268 non-null
                                                       float64
          left_eyebrow_inner_end_x
                                      2270 non-null
                                                       float64
      12
          left_eyebrow_inner_end_y
                                      2270 non-null
                                                       float64
      13
          left_eyebrow_outer_end_x
                                                       float64
                                      2225 non-null
          left_eyebrow_outer_end_y
                                      2225 non-null
                                                       float64
          right_eyebrow_inner_end_x
      16
                                      2270 non-null
                                                       float64
          right_eyebrow_inner_end_y
                                      2270 non-null
                                                       float64
      17
      18
          right_eyebrow_outer_end_x
                                      2236 non-null
                                                       float64
      19
          right_eyebrow_outer_end_y
                                      2236 non-null
                                                       float64
      20
          nose_tip_x
                                      7049 non-null
                                                       float64
      21
          nose tip y
                                      7049 non-null
                                                       float64
                                                       float64
      22
          mouth left corner x
                                      2269 non-null
      23
          mouth left corner y
                                      2269 non-null
                                                       float64
          mouth_right_corner_x
                                      2270 non-null
                                                       float64
          mouth right corner y
                                      2270 non-null
                                                       float64
      25
      26
          mouth_center_top_lip_x
                                      2275 non-null
                                                       float64
      27
          mouth_center_top_lip_y
                                      2275 non-null
                                                       float64
      28
          mouth_center_bottom_lip_x
                                      7016 non-null
                                                       float64
      29
          mouth_center_bottom_lip_y
                                      7016 non-null
                                                       float64
                                      7049 non-null
          Image
                                                       object
      30
     dtypes: float64(30), object(1)
     memory usage: 1.7+ MB
[11]: training.head(5)
         left_eye_center_x left_eye_center_y right_eye_center_x \
                                                         30.227008
      0
                 66.033564
                                     39.002274
      1
                 64.332936
                                     34.970077
                                                         29.949277
      2
                 65.057053
                                     34.909642
                                                         30.903789
      3
                 65.225739
                                     37.261774
                                                         32.023096
      4
                 66.725301
                                     39.621261
                                                         32.244810
                             left_eye_inner_corner_x
         right_eye_center_y
                                                       left_eye_inner_corner_y \
      0
                  36.421678
                                            59.582075
                                                                      39.647423
      1
                  33.448715
                                            58.856170
                                                                      35.274349
      2
                  34.909642
                                            59.412000
                                                                      36.320968
      3
                  37.261774
                                            60.003339
                                                                      39.127179
```

7036 non-null

7036 non-null

float64

float64

right_eye_center_x

right_eye_center_y

2 3

[11]:

4

38.042032

58.565890

39.621261

```
left_eye_outer_corner_x left_eye_outer_corner_y right_eye_inner_corner_x
     0
                      73.130346
                                                39.969997
                                                                           36.356571
     1
                      70.722723
                                                36.187166
                                                                           36.034723
     2
                      70.984421
                                                36.320968
                                                                           37.678105
     3
                      72.314713
                                                38.380967
                                                                           37.618643
     4
                      72.515926
                                                39.884466
                                                                           36.982380
        right_eye_inner_corner_y ... nose_tip_y mouth_left_corner_x \
                       37.389402 ...
                                       57.066803
                                                             61.195308
     0
                       34.361532 ...
                                       55.660936
                                                             56.421447
     1
     2
                       36.320968 ... 53.538947
                                                             60.822947
                                                             65.598887
     3
                       38.754115 ... 54.166539
     4
                       39.094852 ...
                                       64.889521
                                                             60.671411
        mouth_left_corner_y mouth_right_corner_x mouth_right_corner_y
     0
                  79.970165
                                         28.614496
                                                                77.388992
                  76.352000
                                         35.122383
                                                                76.047660
     1
     2
                  73.014316
                                         33.726316
                                                               72.732000
     3
                  72,703722
                                         37.245496
                                                                74.195478
                  77.523239
                                         31.191755
                                                                76.997301
        mouth_center_top_lip_x mouth_center_top_lip_y mouth_center_bottom_lip_x \
                     43.312602
                                              72.935459
                                                                          43.130707
     0
     1
                     46.684596
                                              70.266553
                                                                          45.467915
     2
                     47.274947
                                              70.191789
                                                                          47.274947
     3
                     50.303165
                                              70.091687
                                                                          51.561183
                     44.962748
                                             73.707387
                                                                          44.227141
        mouth_center_bottom_lip_y
     0
                        84.485774
     1
                        85.480170
     2
                        78.659368
     3
                        78.268383
                        86.871166
                                                     Image
    0 238 236 237 238 240 240 239 241 241 243 240 23...
     1 219 215 204 196 204 211 212 200 180 168 178 19...
     2 144 142 159 180 188 188 184 180 167 132 84 59 ...
     3 193 192 193 194 194 194 193 192 168 111 50 12 ...
     4 147 148 160 196 215 214 216 217 219 220 206 18...
     [5 rows x 31 columns]
[8]: test = pd.read_csv('/kaggle/working/test.csv')
     test.info()
```

```
<class 'pandas.core.frame.DataFrame'>
     RangeIndex: 1783 entries, 0 to 1782
     Data columns (total 2 columns):
         Column
                  Non-Null Count Dtype
                  -----
         ImageId 1783 non-null
                                 int64
         Image
                 1783 non-null
                                 object
     dtypes: int64(1), object(1)
     memory usage: 28.0+ KB
[9]: test.head(5)
[9]:
        ImageId
                                                           Image
              1 182 183 182 182 180 180 176 169 156 137 124 10...
              2 76 87 81 72 65 59 64 76 69 42 31 38 49 58 58 4...
     1
              3 177 176 174 170 169 169 168 166 166 166 161 14...
     2
     3
              4 176 174 174 175 174 174 176 176 175 171 165 15...
              5 50 47 44 101 144 149 120 58 48 42 35 35 37 39 ...
[12]: training = training.dropna()
     training.shape, type(training)
[12]: ((2140, 31), pandas.core.frame.DataFrame)
\rightarrowdtype=int, sep=' ').reshape((96,96)))
[14]: def get image and dots(df, index):
         image = plt.imshow(df['Image'][index],cmap='gray')
         1 = []
         for i in range(1,31,2):
             l.append(plt.plot(df.loc[index][i-1], df.loc[index][i], 'ro'))
         return image, 1
[16]: fig = plt.figure(figsize=(8, 8))
     fig.subplots_adjust(
         left=0, right=1, bottom=0, top=1, hspace=0.05, wspace=0.05)
     for i in range(16):
         ax = fig.add_subplot(4, 4, i + 1, xticks=[], yticks=[])
         get_image_and_dots(training, i)
     plt.show()
     plt.savefig('fkeypoints.png')
```



<Figure size 432x288 with 0 Axes>

```
[19]: (numpy.ndarray, (2140, 30))
[20]: X_train, X_test, y_train, y_test = train_test_split(X, Y, test_size=0.25, □ → random_state=42)
```

1.0.1 Model

```
[21]: from keras.models import Sequential from keras.layers import Dense, Conv2D, Flatten, AvgPool2D, BatchNormalization, Dropout, Activation, MaxPooling2D from keras.optimizers import Adam from keras import regularizers from keras.layers.advanced_activations import LeakyReLU from keras.models import Sequential, Model from keras.layers import Activation, Convolution2D, MaxPooling2D, BatchNormalization, Flatten, Dense, Dropout, Conv2D, MaxPool2D, ZeroPadding2D
```

Using TensorFlow backend.

```
[25]: model = Sequential()
      model.add(Convolution2D(32, (3,3), padding='same', use_bias=False,__
       \rightarrowinput_shape=(96,96,1)))
      model.add(LeakyReLU(alpha = 0.1))
      model.add(BatchNormalization())
      model.add(Convolution2D(32, (3,3), padding='same', use_bias=False))
      model.add(LeakyReLU(alpha = 0.1))
      model.add(BatchNormalization())
      model.add(MaxPool2D(pool_size=(2, 2)))
      model.add(Convolution2D(64, (3,3), padding='same', use_bias=False))
      model.add(LeakyReLU(alpha = 0.1))
      model.add(BatchNormalization())
      model.add(Convolution2D(64, (3,3), padding='same', use_bias=False))
      model.add(LeakyReLU(alpha = 0.1))
      model.add(BatchNormalization())
      model.add(MaxPool2D(pool_size=(2, 2)))
      model.add(Convolution2D(96, (3,3), padding='same', use_bias=False))
      model.add(LeakyReLU(alpha = 0.1))
      model.add(BatchNormalization())
      model.add(Convolution2D(96, (3,3), padding='same', use_bias=False))
      model.add(LeakyReLU(alpha = 0.1))
```

```
model.add(BatchNormalization())
model.add(MaxPool2D(pool_size=(2, 2)))
model.add(Convolution2D(128, (3,3),padding='same', use_bias=False))
# model.add(BatchNormalization())
model.add(LeakyReLU(alpha = 0.1))
model.add(BatchNormalization())
model.add(Convolution2D(128, (3,3),padding='same', use bias=False))
model.add(LeakyReLU(alpha = 0.1))
model.add(BatchNormalization())
model.add(MaxPool2D(pool_size=(2, 2)))
model.add(Convolution2D(256, (3,3),padding='same',use_bias=False))
model.add(LeakyReLU(alpha = 0.1))
model.add(BatchNormalization())
model.add(Convolution2D(256, (3,3),padding='same',use_bias=False))
model.add(LeakyReLU(alpha = 0.1))
model.add(BatchNormalization())
model.add(MaxPool2D(pool_size=(2, 2)))
model.add(Convolution2D(512, (3,3), padding='same', use_bias=False))
model.add(LeakyReLU(alpha = 0.1))
model.add(BatchNormalization())
model.add(Convolution2D(512, (3,3), padding='same', use_bias=False))
model.add(LeakyReLU(alpha = 0.1))
model.add(BatchNormalization())
model.add(Flatten())
model.add(Dense(512,activation='relu'))
model.add(Dropout(0.1))
model.add(Dense(30))
model.summary()
```

Model: "sequential_3"

Output Shape	Param #
(None, 96, 96, 32)	288
(None, 96, 96, 32)	0
(None, 96, 96, 32)	128
	=======================================

conv2d_26 (Conv2D)	(None,	96,	96,	32)	9216
leaky_re_lu_26 (LeakyReLU)	(None,	96,	96,	32)	0
batch_normalization_26 (Batc	(None,	96,	96,	32)	128
max_pooling2d_11 (MaxPooling	(None,	48,	48,	32)	0
conv2d_27 (Conv2D)	(None,	48,	48,	64)	18432
leaky_re_lu_27 (LeakyReLU)	(None,	48,	48,	64)	0
batch_normalization_27 (Batc	(None,	48,	48,	64)	256
conv2d_28 (Conv2D)	(None,	48,	48,	64)	36864
leaky_re_lu_28 (LeakyReLU)	(None,	48,	48,	64)	0
batch_normalization_28 (Batc	(None,	48,	48,	64)	256
max_pooling2d_12 (MaxPooling	(None,	24,	24,	64)	0
conv2d_29 (Conv2D)	(None,	24,	24,	96)	55296
leaky_re_lu_29 (LeakyReLU)	(None,	24,	24,	96)	0
batch_normalization_29 (Batc	(None,	24,	24,	96)	384
conv2d_30 (Conv2D)	(None,	24,	24,	96)	82944
leaky_re_lu_30 (LeakyReLU)	(None,	24,	24,	96)	0
batch_normalization_30 (Batc	(None,	24,	24,	96)	384
max_pooling2d_13 (MaxPooling	(None,	12,	12,	96)	0
conv2d_31 (Conv2D)	(None,	12,	12,	128)	110592
leaky_re_lu_31 (LeakyReLU)	(None,	12,	12,	128)	0
batch_normalization_31 (Batc	(None,	12,	12,	128)	512
conv2d_32 (Conv2D)	(None,	12,	12,	128)	147456
leaky_re_lu_32 (LeakyReLU)	(None,	12,	12,	128)	0
batch_normalization_32 (Batc	(None,	12,	12,	128)	512

```
max_pooling2d_14 (MaxPooling (None, 6, 6, 128)
    ._____
    conv2d_33 (Conv2D)
                     (None, 6, 6, 256)
                                            294912
    leaky_re_lu_33 (LeakyReLU) (None, 6, 6, 256) 0
    batch normalization 33 (Batc (None, 6, 6, 256)
                         (None, 6, 6, 256)
    conv2d 34 (Conv2D)
                                            589824
    leaky_re_lu_34 (LeakyReLU) (None, 6, 6, 256)
    batch_normalization_34 (Batc (None, 6, 6, 256)
    max_pooling2d_15 (MaxPooling (None, 3, 3, 256)
    conv2d_35 (Conv2D)
                        (None, 3, 3, 512) 1179648
    leaky_re_lu_35 (LeakyReLU) (None, 3, 3, 512) 0
    batch_normalization_35 (Batc (None, 3, 3, 512)
    conv2d_36 (Conv2D)
                        (None, 3, 3, 512)
                                            2359296
    leaky_re_lu_36 (LeakyReLU) (None, 3, 3, 512)
    batch_normalization_36 (Batc (None, 3, 3, 512)
                                            2048
                         (None, 4608)
    flatten_3 (Flatten)
    ._____
    dense_5 (Dense)
                         (None, 512)
                                             2359808
    _____
    dropout_3 (Dropout)
                    (None, 512)
    dense 6 (Dense)
                        (None, 30)
                                             15390
    -----
    Total params: 7,268,670
    Trainable params: 7,264,318
    Non-trainable params: 4,352
[26]: model.compile(optimizer='Adam',
               loss='mse',
               metrics=['mae'])
    model.fit(X_train, y_train, epochs=600)
```

Epoch 1/600

```
8.2838
Epoch 2/600
4.0540
Epoch 3/600
3.6241
Epoch 4/600
3.4118
Epoch 5/600
3.1581
Epoch 6/600
3.0203
Epoch 7/600
2.9459
Epoch 8/600
2.7325
Epoch 9/600
2.7494
Epoch 10/600
2.5696
Epoch 11/600
2.4580
Epoch 12/600
2.6962
Epoch 13/600
2.4514
Epoch 14/600
2.3462
Epoch 15/600
2.3842
Epoch 16/600
2.4269
Epoch 17/600
```

```
2.3308
Epoch 18/600
2.1542
Epoch 19/600
2.1801
Epoch 20/600
2.1331
Epoch 21/600
2.1339
Epoch 22/600
2.2461
Epoch 23/600
2.3583
Epoch 24/600
2.0723
Epoch 25/600
2.0194
Epoch 26/600
2.0914
Epoch 27/600
1.8878
Epoch 28/600
2.0949
Epoch 29/600
2.1124
Epoch 30/600
1.9360
Epoch 31/600
1.8371
Epoch 32/600
1.9249
Epoch 33/600
```

```
1.8696
Epoch 34/600
1.9371
Epoch 35/600
1.9093
Epoch 36/600
2.2844
Epoch 37/600
1.8746
Epoch 38/600
2.0004
Epoch 39/600
1.7710
Epoch 40/600
1.8833
Epoch 41/600
1.8392
Epoch 42/600
1.9392
Epoch 43/600
1.8675
Epoch 44/600
1.8701
Epoch 45/600
1.7298
Epoch 46/600
2.4432
Epoch 47/600
1.7269
Epoch 48/600
1.6432
Epoch 49/600
```

```
1.7887
Epoch 50/600
1.7754
Epoch 51/600
1.8397
Epoch 52/600
1.8257
Epoch 53/600
1.7824
Epoch 54/600
1.7511
Epoch 55/600
1.7265
Epoch 56/600
1.6839
Epoch 57/600
1.6897
Epoch 58/600
2.0732
Epoch 59/600
1.6399
Epoch 60/600
1.5681
Epoch 61/600
1.7199
Epoch 62/600
1.6264
Epoch 63/600
1.6378
Epoch 64/600
1.7927
Epoch 65/600
```

```
1.6939
Epoch 66/600
1.5340
Epoch 67/600
1.6026
Epoch 68/600
1.5160
Epoch 69/600
1.6915
Epoch 70/600
1.5766
Epoch 71/600
1.5926
Epoch 72/600
1.7028
Epoch 73/600
1.6910
Epoch 74/600
1.6454
Epoch 75/600
1.8963
Epoch 76/600
1.8278
Epoch 77/600
1.5216
Epoch 78/600
1.6099
Epoch 79/600
1.6247
Epoch 80/600
1.5229
Epoch 81/600
```

```
1.4921
Epoch 82/600
1.4947
Epoch 83/600
1.5640
Epoch 84/600
1.5200
Epoch 85/600
1.6368
Epoch 86/600
1.4845
Epoch 87/600
1.5809
Epoch 88/600
1.5074
Epoch 89/600
1.4709
Epoch 90/600
1.6025
Epoch 91/600
1.5904
Epoch 92/600
1.4484
Epoch 93/600
1.5005
Epoch 94/600
1.4795
Epoch 95/600
1.5742
Epoch 96/600
1.5155
Epoch 97/600
```

```
1.4782
Epoch 98/600
1.4414
Epoch 99/600
1.4841
Epoch 100/600
1.6241
Epoch 101/600
1.5985
Epoch 102/600
1.7865
Epoch 103/600
1.4981
Epoch 104/600
1.4734
Epoch 105/600
1.6906
Epoch 106/600
1.4671
Epoch 107/600
1.5060
Epoch 108/600
1.5113
Epoch 109/600
1.5458
Epoch 110/600
1.5179
Epoch 111/600
1.4304
Epoch 112/600
1.4524
Epoch 113/600
```

```
1.5536
Epoch 114/600
1.4252
Epoch 115/600
1.4633
Epoch 116/600
1.3688
Epoch 117/600
1.6194
Epoch 118/600
1.4350
Epoch 119/600
1.5636A: 2s - 1
Epoch 120/600
1.4031
Epoch 121/600
1.5492
Epoch 122/600
1.5759
Epoch 123/600
1.5051
Epoch 124/600
1.4196
Epoch 125/600
1.4705
Epoch 126/600
1.3974
Epoch 127/600
1.4894
Epoch 128/600
1.6959
Epoch 129/600
```

```
1.5480
Epoch 130/600
1.4973
Epoch 131/600
1.5033
Epoch 132/600
1.5103
Epoch 133/600
1.6729
Epoch 134/600
1.7289
Epoch 135/600
1.5955
Epoch 136/600
1.4551
Epoch 137/600
1.4512
Epoch 138/600
1.4311
Epoch 139/600
1.3873
Epoch 140/600
1.3788
Epoch 141/600
1.3258
Epoch 142/600
1.4065
Epoch 143/600
1.6206
Epoch 144/600
1.4940
Epoch 145/600
```

```
1.4411
Epoch 146/600
1.4736
Epoch 147/600
1.4627
Epoch 148/600
1.6488
Epoch 149/600
1.4948
Epoch 150/600
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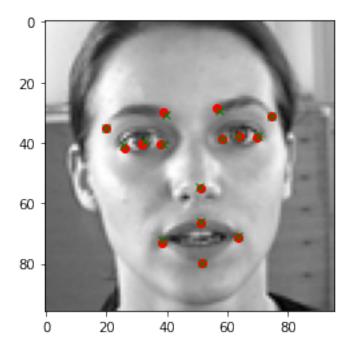
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1.2613
   Epoch 594/600
   1.2338
   Epoch 595/600
   1.2239A: Os - loss: 2.5357
   Epoch 596/600
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   Epoch 597/600
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   Epoch 598/600
   1.1787
   Epoch 599/600
   1.2412
   Epoch 600/600
   1.2052
[26]: <keras.callbacks.callbacks.History at 0x7fd53c1a4e10>
[27]: score = model.evaluate(X_test, y_test, verbose=0)
   print('Test loss:', score[0])
   print('Test accuracy:', score[1])
   Test loss: 1.6637967334729489
   Test accuracy: 0.9066632390022278
[28]: model.save('keypoint_model03.h5')
[29]: test['Image'] = test['Image'].apply(lambda x: np.fromstring(x, dtype=int, sep='u
   \hookrightarrow').reshape((96,96)))
   test.shape, type(test)
[29]: ((1783, 2), pandas.core.frame.DataFrame)
[30]: test_X = np.asarray([test['Image']], dtype=np.uint8).reshape(test.
    \hookrightarrowshape [0], 96, 96, 1)
   test_res = model.predict(test_X)
   train_predicts = model.predict(X_train)
```

```
[31]: n = 12

xv = X_train[n].reshape((96,96))
plt.imshow(xv,cmap='gray')

for i in range(1,31,2):
    plt.plot(train_predicts[n][i-1], train_predicts[n][i], 'ro')
    plt.plot(y_train[n][i-1], y_train[n][i], 'x', color='green')

plt.show()
```



```
[38]: lookid_dir = '../input/facial-keypoints-detection/IdLookupTable.csv'
lookid_data = pd.read_csv(lookid_dir)

[42]: lookid_list = list(lookid_data['FeatureName'])
    imageID = list(lookid_data['ImageId']-1)
    pre_list = list(test_res)
    rowid = lookid_data['RowId']
    rowid=list(rowid)
    feature = []
    for f in list(lookid_data['FeatureName']):
        feature.append(lookid_list.index(f))
    preded = []
    for x,y in zip(imageID,feature):
        preded.append(pre_list[x][y])
```

```
rowid = pd.Series(rowid,name = 'RowId')
loc = pd.Series(preded,name = 'Location')
submission = pd.concat([rowid,loc],axis = 1)
submission.to_csv('face_key_detection_submission.csv',index = False)
```