

000
001
002
003
004
005
006
007
008
009
010054
055
056
057
058
059
060
061
062
063
064

1. The summary of Haar cascade face detector

This algorithm can be broken down into the following three steps:

1.1. Extracting features

The Haar feature is used to extract the features of the original image. In this algorithm, the main reason for using features instead of pixels for recognition is that features can contain prior knowledge [1].

1.2. Selecting features

The features extracted by Haar feature should be reduced because their number is too many. A variant of Adaboost is used to select and train these features [1]. A weak feature is composed of a feature and its corresponding threshold. Multiple weak classifiers can be combined to form a strong classifier.

1.3. Using cascade structure to combine classifiers

In a picture, there are many regions that do not contain face elements. If all of the features are applied to one sub-window, the computing efficiency will be greatly reduced. Therefore, this algorithm uses cascade structure to combine the classifiers. This structure contains many stages, and each stage contains multiple classifiers. Only when all classifiers in all stages determine that the current sub-window is a face, the algorithm confirms that this sub-window is a face. If any classifier in any of the stages determines that the current sub-window is not a face, then the algorithm determines that the current sub-window is not a face, and the algorithm will be terminated at this stage. Several simple classifiers can be deployed on the first several stages of the cascade structure. In this way, only after passing the detection of simple classifiers in the first several stages for each sub-window, will it be detected by a large number of more complex classifiers. Such a cascade structure can exclude sub-windows that do not contain faces at an early stage, thus greatly reducing the number of sub-windows that need to be determined by all features.

2. The process of designing the camouflage

I have designed two types of camouflage altogether. I will introduce these two kinds of camouflage respectively. Two sets of pictures are shown below. The faces in the first group (Figure 1 and Figure 2) were expressionless, while the faces in the second group (Figure 3 and Figure 4) were smiling. Figure 2 and Figure 4 shows the results of this algorithm that identifies the Figure 1 and Figure 2 which don't have any camouflage.



Figure 1

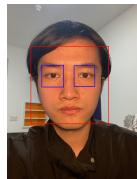


Figure 2



Figure 3

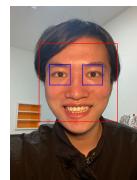


Figure 4

It can be seen that the recognition results of this algorithm on these pictures are correct. I will compare these two pictures with the results of using two types of camouflage that I used below.

2.1. The camouflage of covering the eyes

2.1.1 The description of this camouflage

I covered both of my eyes with a piece of white paper.

2.1.2 The principle of this camouflage

In the official website of OpenCV, I learned that a facial feature that is extracted and is used in this algorithm is that there is a significant difference in intensity between the eyes area and the upper cheeks area, and the eyes area is always darker than the upper cheeks area[1]. So I chose to use the

108 paper which is more white than my upper cheeks to cover
 109 my eye area. Because of this, the classifier that contains
 110 this feature may decide that this image is not a face. Since
 111 this algorithm uses cascade structure, even if the classifiers
 112 containing other features consider this picture to be a face,
 113 the whole algorithm will also determine that this picture is
 114 not a face.
 115

116 2.1.3 The pictures of experimental results

118 Figure 5 shows the results of this algorithm which identifies
 119 the image that I covered the eyes of the face with no expres-
 120 sion, and Figure 6 shows the results of this algorithm which
 121 identifies the image that I covered the eyes of the face with
 122 smile. It can be seen that this camouflage is effective for
 123 both expressionless and smiling faces.
 124



Figure 5



Figure 6

134 2.2. The camouflage of blacking the nos

135 2.2.1 The description of this camouflage

136 I black the nose area. Since I don't have a black makeup
 137 pen, I had to black the nose area of the facial picture. I
 138 think these two methods have the same effect.
 139

141 2.2.2 The principle of this camouflage

143 One feature extracted and used in this algorithm is that the
 144 eyes area is always darker than the bridge of the nose, so
 145 I black the area around the bridge of the nose to make the
 146 bridge of the nose's intensity to be greater than the eyes'.
 147 Because of this, the classifier that contains this feature may
 148 decide that this image is not a face. Since this algorithm
 149 uses cascade structure, even if the classifiers containing
 150 other features consider this picture to be a face, the whole
 151 algorithm will also determine that this picture is not a face.
 152

153 2.2.3 The pictures of experimental results

155 Figure 7 shows the results of this algorithm which identi-
 156 fies the image that I black the the bridge of the nose of the
 157 face with no expression, and Figure 8 shows the results of
 158 this algorithm which identifies the image that I black the
 159 the bridge of the nose of the face with smile. It can be seen
 160 that this camouflage is effective for both expressionless and
 161 smiling faces.

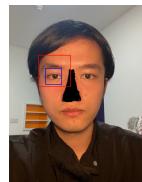


Figure 7

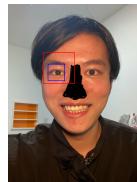


Figure 8

172 2.2.4 The analysis of experimental results

173 The important part of this camouflage is that the bridge of
 174 the nose should be blacked. If only the tip of the nose but
 175 not the bridge of the nose is blacked, the algorithm can still
 176 correctly identify the face. This indicates that the feature
 177 extracted by this algorithm is the intensity comparison be-
 178 tween the bridge of the nose and the eyes, rather than the
 179 intensity comparison between the nose and the eyes. This
 180 conclusion can be confirmed by the following figures. Fig-
 181 ure 9 is this algorithm's recognition result of the image
 182 without blacking the nose bridge, and Figure 10 is this al-
 183 gorithm's recognition result of the image with blacking the
 184 nose bridge.
 185

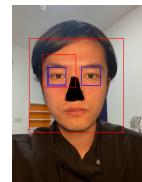


Figure 9

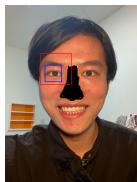


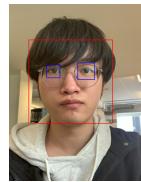
Figure 10

197 3. Further discussion

198 During the experiment, I have been using images with
 199 low pixels (231*308) for the algorithm to recognize. The al-
 200 gorithm always correctly identified the position of my face
 201 and eyes of the images without the camouflage. When I
 202 tried to get the algorithm to identify images with high pixels
 203 (2361*3148), the algorithm could still find the exact loca-
 204 tion of my face, but at the same time, it might identify the
 205 places that do not contain faces as faces, and the places that
 206 do not contain eyes as eyes. So I guess that as the pixel
 207 increases, more and more non-face regions in the image
 208 wrongly meet all the features extracted from the algorithm
 209 by accident. For example, the algorithm may identify the
 210 area where two black dots are close together on clothes as
 211 a face, so I don't think this algorithm can do high precision
 212 face recognition work. Figure 11 is the recognition result
 213 of this algorithm for high pixel images. Figure 12 is the
 214 recognition result of low pixel image by this algorithm.
 215

216
217
218
219
220
221

222 Figure 11



223 Figure 12

224

225 **References**

226

- 227 [1] Paul Viola and Michael Jones. Robust real-time face detection.
- International journal of computer vision*
- , 57(2):137–154,
-
- 228 2004. 1

229

230

231

232

233

234

235

236

237

238

239

240

241

242

243

244

245

246

247

248

249

250

251

252

253

254

255

256

257

258

259

260

261

262

263

264

265

266

267

268

269

270

271

272

273

274

275

276

277

278

279

280

281

282

283

284

285

286

287

288

289

290

291

292

293

294

295

296

297

298

299

300

301

302

303

304

305

306

307

308

309

310

311

312

313

314

315

316

317

318

319

320

321

322

323