Evaluation of Facial Landmark Detection on MOBIO Database

Na Zhang

Problem

- MOBIO [1] database was usually used for…before?
- Seldom used in facial landmark detection
- In our problem,
  - Choose several state-of-art facial landmark detection methods
  - Execute landmark detection on MOBIO database
  - Evaluate the performance of popular detection methods on MOBIO

[*] Chris McCool, Sébastien Marcel, Abdenour Hadid, Matti Pietikäinen, Pavel Matějka, Jan Černocký, Norman Poh, Josef Kittler, Anthony Larcher, Christophe Lévy, Driss Matrouf, Jean-François Bonastre, Phil Tresadern, and Timothy Cootes, "Bi-Modal Person Recognition on a Mobile Phone: using mobile phone data", in IEEE ICME Workshop on Hot Topics in Mobile Mutlimedia, 2012.

MOBIO Database Description

- Mobile Biometrics Database
- Diverse Bi-modal database
- Consists of bi-modal data
  - Audio
  - Video
- Taken from 152 people
- Female-Male ratio: 1:2
  - 100 males
  - 52 females
- Collected from August 2008 until July 2010 in six different sites from five different countries with both native and non-native English speakers
- Source download link: https://www.idiap.ch/dataset/mobio

- 12 sessions were captured for each client
  - 6 sessions for Phase I
    - Consists of 21 questions with the question types ranging from:
    - Short Response Questions, Short Response Free Speech, Set Speech, and Free Speech
  - 6 sessions for Phase II
    - Consists of 11 questions with the question types ranging from:
    - Short Response Questions, Set Speech, and Free Speech
- Recorded using 2 mobile devices
  - A mobile phone: NOKIA N93i
  - A laptop computer: standard 2008 MacBook
- The laptop was only used to capture part of the first session
- The first session consists of data captured on both the laptop and the mobile phone

Detailed Description of Questions

- Short Response Questions
  The short response questions consisted of five pre-defined questions, which were:
  - What is your name? – the user supplies their fake name
  - What is your address? – the user supplies their fake address
  - What is your birthdate? – the user supplies their fake birthdate
  - What is your license number? – the user supplied their fake ID card number (the same for each person)
  - What is your credit card number? – the user supplies their fake Card number

- Short Response Free Speech
  - There were five random questions taken form a list of 30-40 questions.
  - The user had to answer these questions by speaking for approximately 5 seconds of recording (sometimes more and sometimes less).

- Set Speech
  - The users were asked to read pre-defined text out aloud
  - This text was designed to take longer than 10 seconds to utter and the participants were allowed to correct themselves while reading these paragraphs.

- Free Speech
  - Consisted of 10 random questions from a list of approximately 30 questions
  - The answers to each of these questions took approximately 10 seconds (sometimes less and sometimes more)

- In our problem:
  - Extract frames from video data
  - Just collect still face images
  - 20,600 face images with 640*480 size

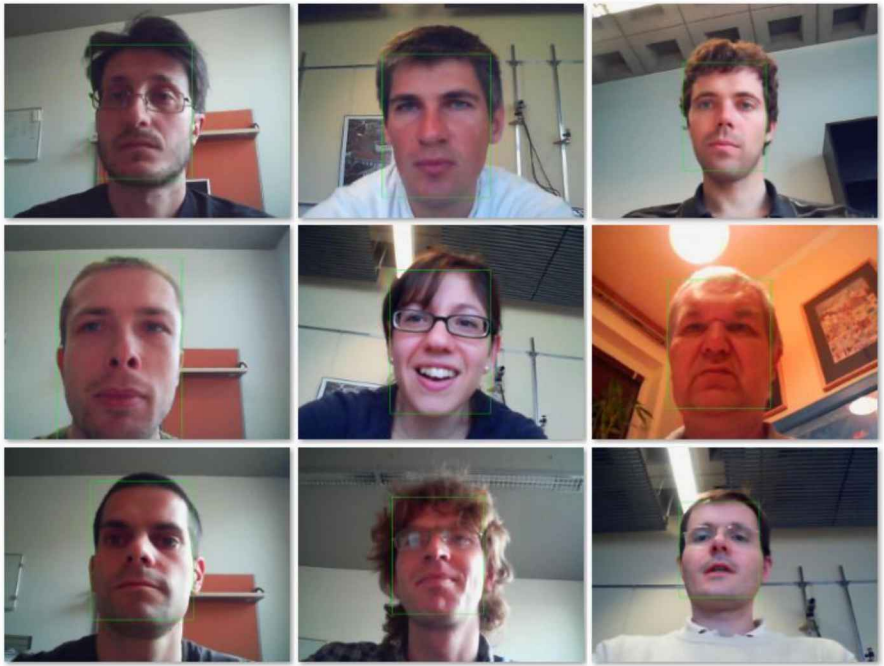
Preprocess Data

- Face Detection
  - MTCNN [*] vs. MatLab Dlib
  - MatLab Dlib:
    - 18, 483 images with one correct face
    - 537 with multiple faces
    - 1,580 with no face

[*] Zhang, Kaipeng, et al. "Joint face detection and alignment using multitask cascaded convolutional networks." IEEE Signal Processing Letters 23.10 (2016): 1499-1503.

- MTCNN:
  - 20,275 images with one right face
  - 211 with multiple faces
  - 114 with no face
  - **Finally, 20,481 images detected**

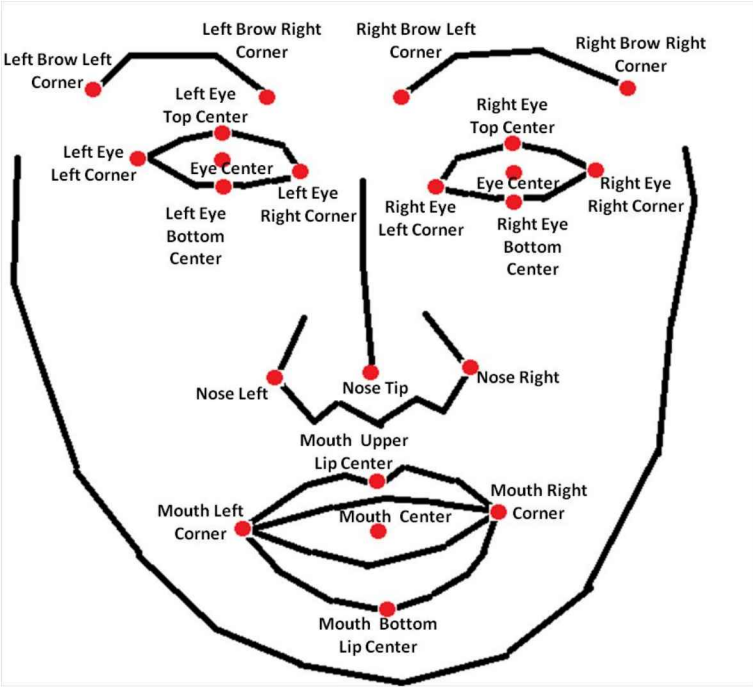
- Choose MTCNN!!!



- Face Cropping
  - Crop into square shape with fixed size by bounding box information
  - 300 * 300
- Face Resize
  - Different facial landmark detection methods have different input size
    - 256 * 256
    - 227 * 227
    - 224*224
    - …

Generate Ground Truth

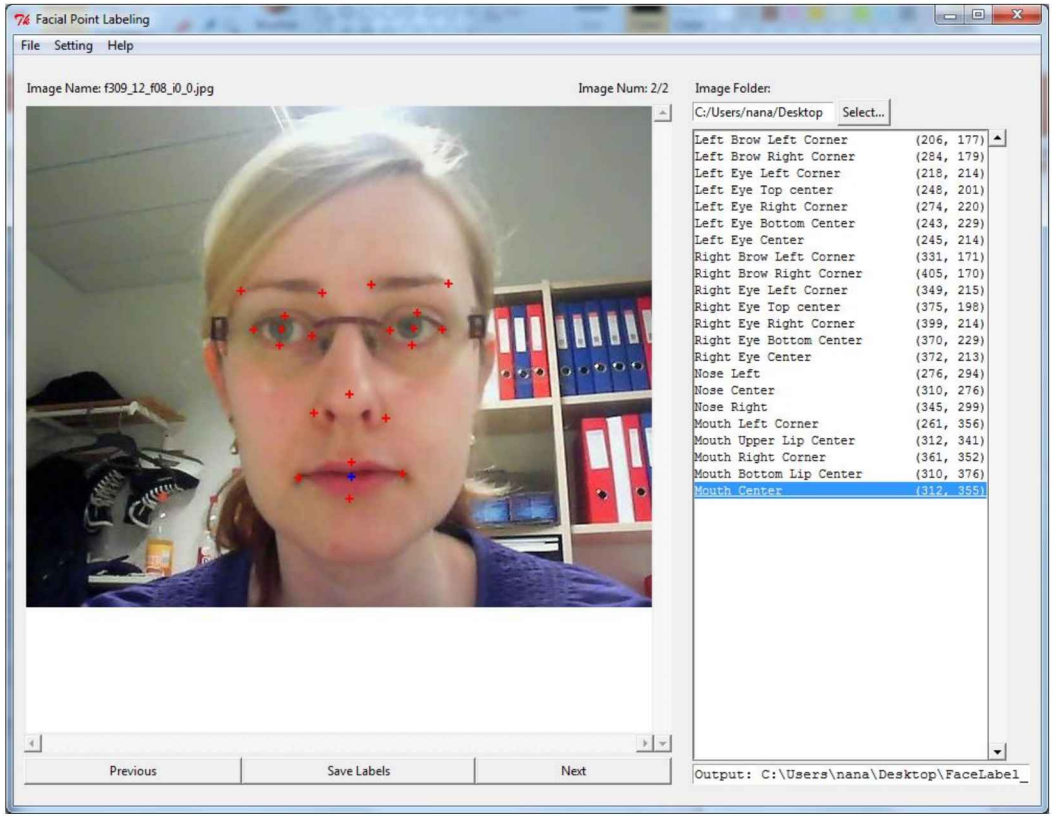
- 1.Left brow left corner
- 2.Left brow right corner
- 3.Right brow left corner
- 4.Right brow right corner
- 5.Left eye left corner
- 6.left eye top center
- 7.Left eye right corner
- 8.left eye bottom center
- 9.left eye center
- 10.Right eye left corner
- 11.right eye top center
- 12.Right eye right corner
- 13.right eye bottom center
- 14.right eye center
- 15.Nose tip
- 16.Nose left
- 17.Nose right
- 18.Mouth left corner
- 19.mouth upper lip center
- 20.Mouth right corner
- 21.Mouth bottom lip center
- 22.Mouth center



- Manually label 22 facial landmarks
- During 2014 to 2017
- Develop a Labeling Tool
  - Named FaceLabel_App
  - The result saved in .txt files

Face Label App

- Run in windows system
- Label images one by one
- In order



Experiment & Evaluation

- Choose several facial landmark detection methods to detect landmarks
- Compare the points with ground truth for evaluation
- **Measure metric**
  - NME: Normalized Mean Error
  - CED: Cumulative Error Distribution Curve
  - AUC: Area Under the error Curve
  - Failure rate

Mean Normalized Error

- The Euclidean Distance ($L_2\ norm$) between estimated points and ground truth are normalized by inter-ocular/ outer eye corner distance

$$e_i = \frac{\left\| X_{(i)}^{e} - X_{(i)}^{g} \right\|_2}{d_{io}}$$

e_i: the i-th error value
$X_{(i)}^{e}$: the i-th estimated points
$X_{(i)}^{g}$: the i-th ground truth
d_{io}: IOD, the inter-ocular distance, i.e. Euclidean distance between two eye centers

- NME can be:
  - Sample-wise
  - Landmark-wise, like above
  - Overall
- Heavy impacted by outliers

- Use the **distance of two outer eye corners from ground truth** to normalize
- Use landmark-wise NME
- For every face image:
  - Calculate Euclidean distance of 2 outer eye centers: d
  - Calculate the sum of Euclidean distances for 15/16/5 facial landmarks: $\sum_{i=1}^{15} D_i$
  - Calculate normalized mean error: $error = \frac{\sum_{i=1}^{15} D_i}{15 * d}$

Notes: [68 points: 15] ; [19 points: 16]; [5 points: 5]

Cumulative Error Distribution

- Cumulative distribution function of normalized errors
- Evaluate the fraction of facial landmarks changes as error threshold changes
- Better way to handle outliers
- In our experiment,
  - We set error value threshold is 0.08
  - Partition the error value range [0, 0.08] into 80 segments with equal step size 0.001
  - For each error value point X, Calculate the fraction of face images whose error value <= X as Y

AUC

- The area under the error curve CED

$$AUC_{\alpha} = \int_{0}^{\alpha} f(e)de$$

e: Normalized error

f(e):cumulative error distribution function

α:upper bound, used to calculate the define integration

Failure Rate

- Count the fraction of faces whose error value is greater than error value threshold, e.g. 0.08

Facial Landmark Detection Methods

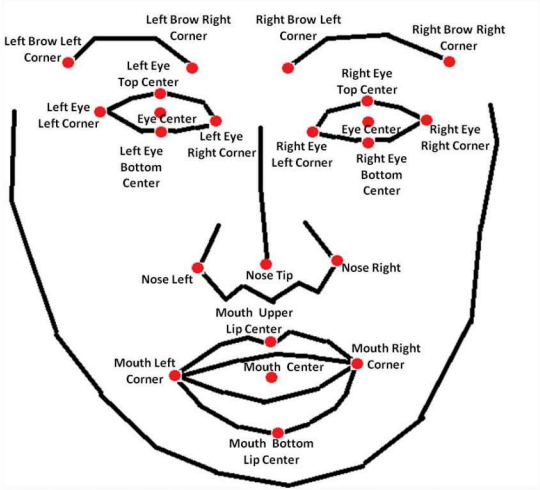
- Tweaked CNN
- WingLoss
- DAC-CSR
- PA-CNN
- OpenPose
- ECT
- TCDCN

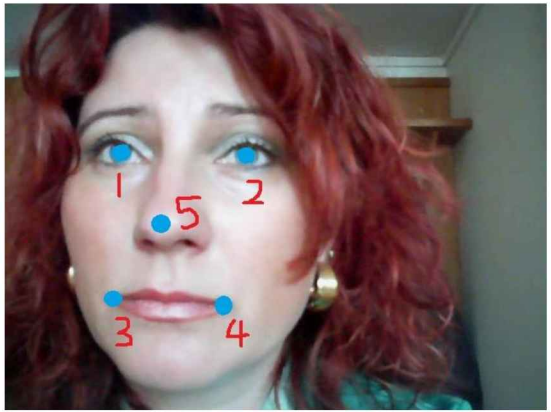
MTCNN

- Python3.0 + mtcnn
- 18,392
- 5 points
- Input original images

- 1. left eye center
- 2. right eye center
- 3. mouth left corner
- 4. mouth right corner
- 5. nose tip

- For 5 detected landmarks:
  - Find the facial points that can be get their corresponding points in those 22 ground truth points for evaluation
- Find 5 points in total
  - 9.left eye center -- 1
  - 14.right eye center -- 2
  - 18.Mouth left corner -- 3
  - 20.Mouth right corner – 4
  - 15.Nose tip -- 5



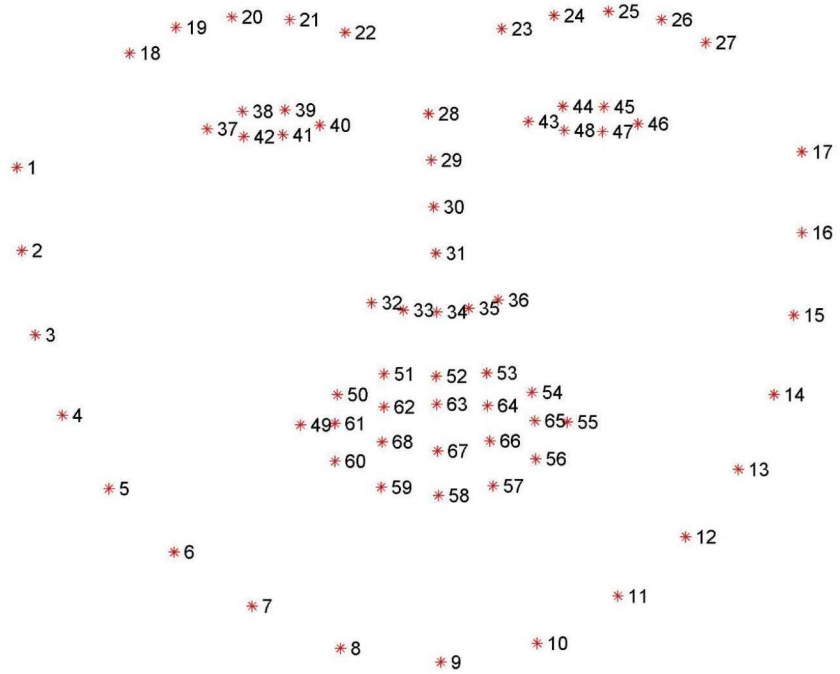


ECT Model

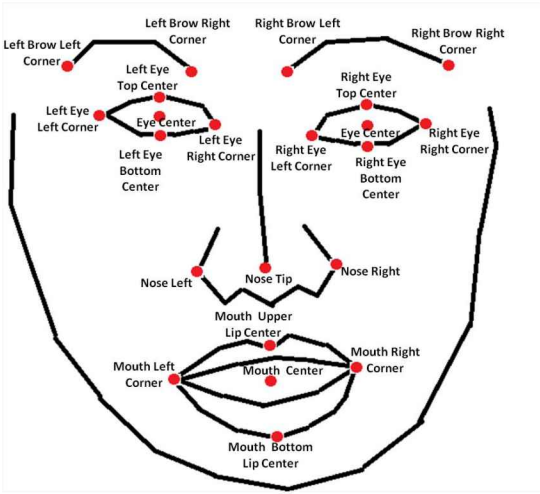
- Estimation Correction Tuning Deep Model
  - Data-driven model: FCN; compute response maps (textural appearance information)
  - Model-driven model: Maximum points fitted with PDM
  - RLMS: fine-tune facial shape iteratively, correct outliers of landmarks
- Pre-trained deep model
- On Caffe + Python
- Input: 256 * 256
- Output: 68 facial points

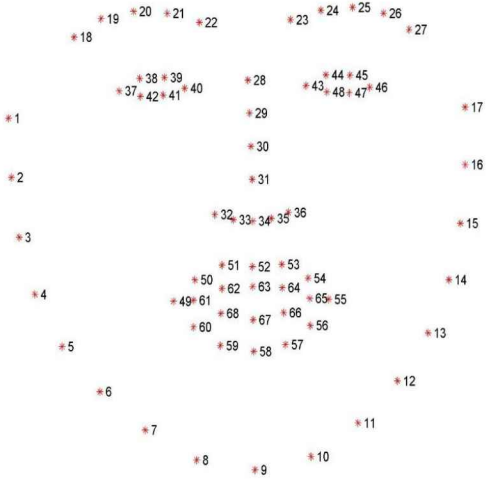
68 facial points

- 51 facial features points
  - 5+5 brow
  - 6+6 eyes
  - 9 nose
  - 20 mouth
- 17 face contour points

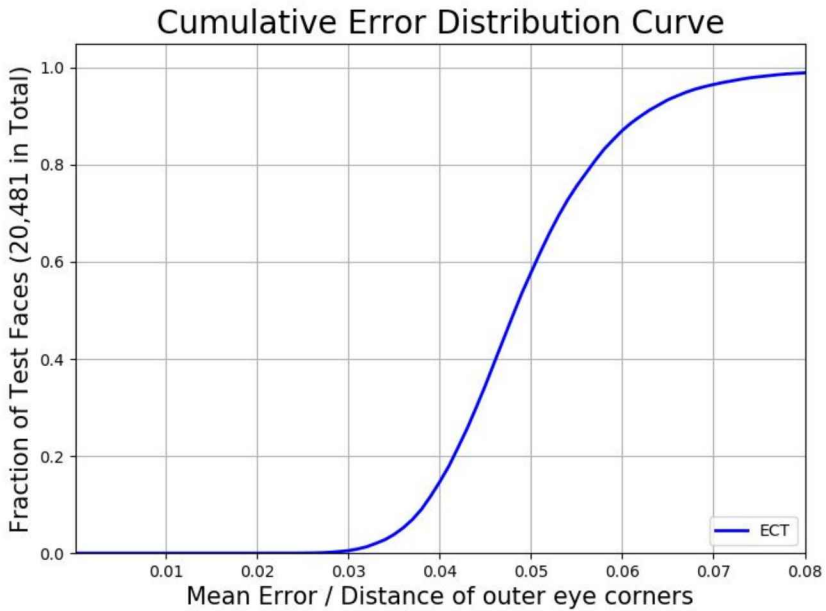


- For 68 detected landmarks:
  - Find the facial points that can be get their corresponding points in those 22 ground truth points for evaluation
- Find 15 points in total
  - 1.Left brow left corner -- 18
  - 2.Left brow right corner -- 22
  - 3.Right brow left corner -- 23
  - 4.Right brow right corner -- 27
  - 5.Left eye left corner -- 37
  - 7.Left eye right corner -- 40
  - 10.Right eye left corner -- 43
  - 12.Right eye right corner -- 46
  - 15.Nose tip -- 31
  - 16.Nose left -- 32
  - 17.Nose right -- 36
  - 18.Mouth left corner -- 49
  - 20.Mouth right corner -- 55
  - 19.Mouth upper lip center -- 52
  - 21.Mouth bottom lip center -- 58





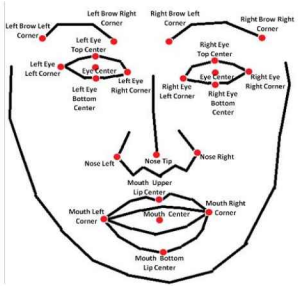
- 20,481 normalized mean errors
- Set error threshold=0.08
- Step size=0.001
- AUC=38.226405
- Failure rate: 1.079049%

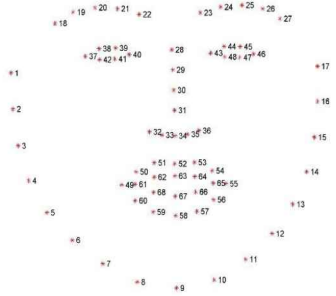


PA-CNN

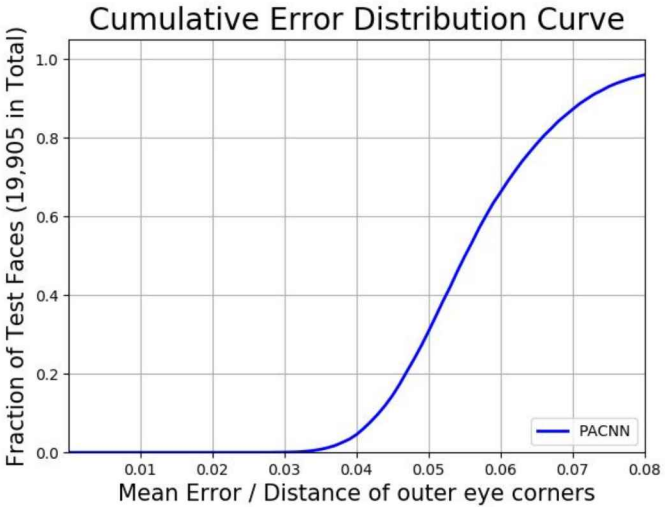
- Part-Aware Deep CNN
- End-to-end regression framework
  - Encode image into feature maps shared by all landmarks
  - The feature are sent into 2 sub-nets to regress 2 types of landmarks
    - Contour landmark: 17
    - Inner landmark: 51
  - **Can directly detect landmarks on original images**
  - **Does not need to detect, crop, and resize face**
- Caffe + Python + Dlib + OpenCV3
- Output: 68 points

- In total
  - 19,505 faces are detected
  - 1,095 faces fail to be detected
- Evaluation is similar with Method 1
- Adopt same 15 landmarks for evaluation





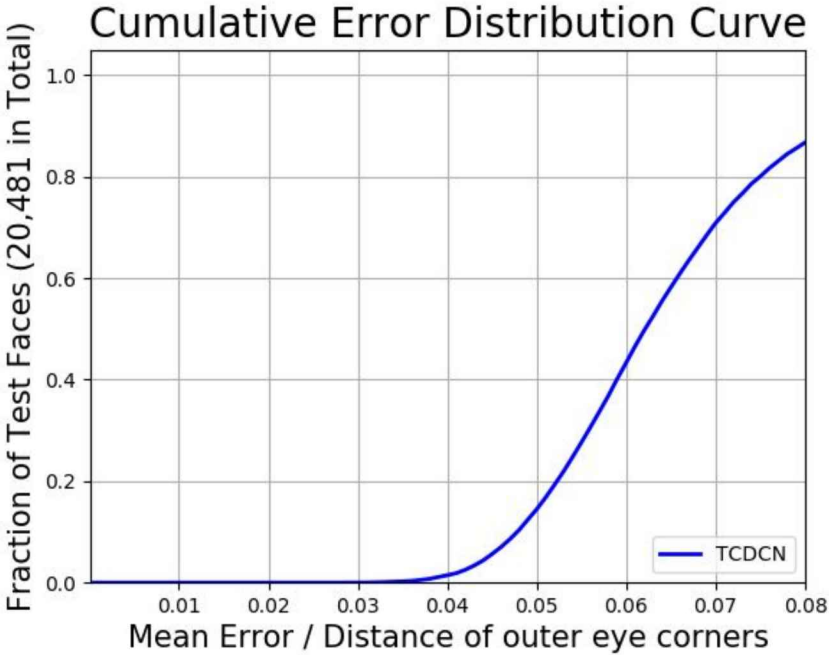
- 19,905 normalized mean errors
- Set error threshold=0.08
- Step size=0.001
- AUC=29.574564
- Failure rate: 4.029736%



TCDCN

- 68 points:
- input: original images,
- bbox: [left, top, width, height]
- output:
  - 20,481 images:
  - 68 facial landmark: (x1,y1,x2,y2....x68,y68).
- Evaluation is similar with Method 1
- Adopt same 15 landmarks for evaluation

- 20,481 normalized mean errors
- Set error threshold=0.08
- Step size=0.001
- AUC (%) = 21.545304
- Failure Rate (%) = 13.290367

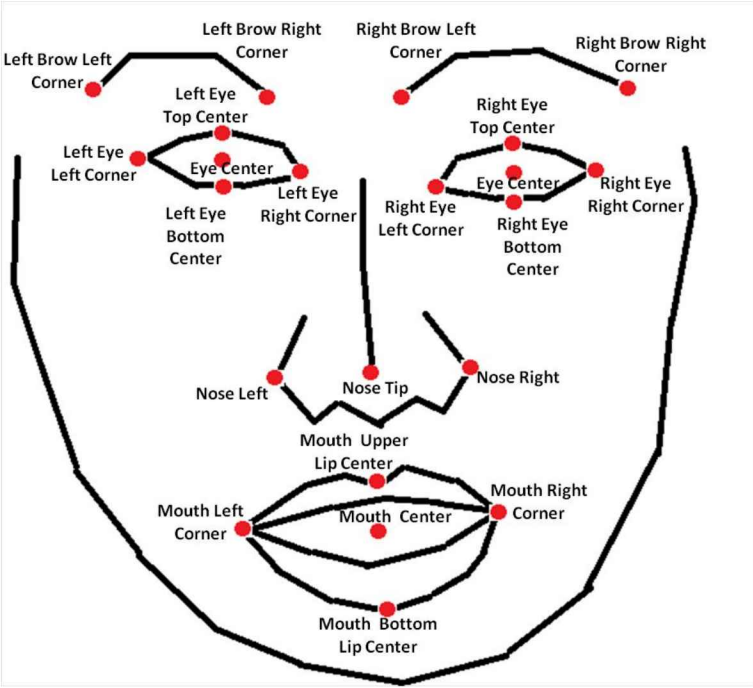


WingLoss

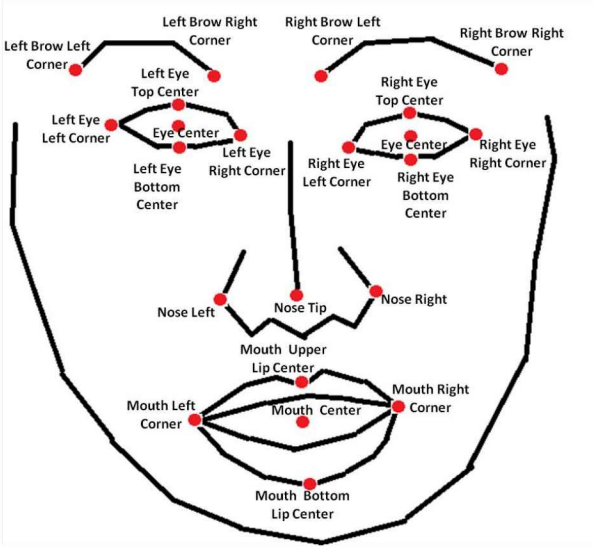
- input: in 256*256 size;
- MTCNN face detection
- output: 19 landmarks.
- (x1,x2,x3,....x19, y1,y2,y3,.....y19)

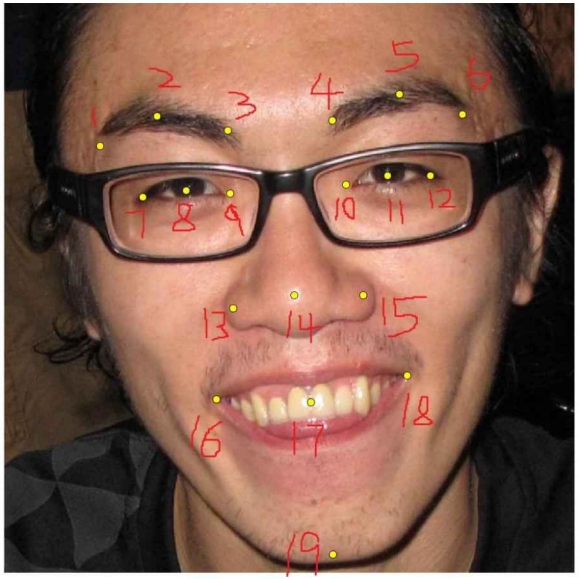
Ground Truth

- 1.Left brow left corner
- 2.Left brow right corner
- 3.Right brow left corner
- 4.Right brow right corner
- 5.Left eye left corner
- 6.left eye top center
- 7.Left eye right corner
- 8.left eye bottom center
- 9.left eye center
- 10.Right eye left corner
- 11.right eye top center
- 12.Right eye right corner
- 13.right eye bottom center
- 14.right eye center
- 15.Nose tip
- 16.Nose left
- 17.Nose right
- 18.Mouth left corner
- 19.mouth upper lip center
- 20.Mouth right corner
- 21.Mouth bottom lip center
- 22.Mouth center



- For 19 detected landmarks:
  - Find the facial points that can be get their corresponding points in those 22 ground truth points for evaluation
- Find 16 points in total
  - 1.Left brow left corner -- 1
  - 2.Left brow right corner -- 3
  - 3.Right brow left corner -- 4
  - 4.Right brow right corner -- 6
  - 5.Left eye left corner -- 7
  - 9.left eye center -- 8
  - 7.Left eye right corner -- 9
  - 10.Right eye left corner -- 10
  - 14.right eye center -- 11
  - 12.Right eye right corner -- 12
  - 16.Nose left -- 13
  - 15.Nose tip -- 14
  - 17.Nose right -- 15
  - 18.Mouth left corner -- 16
  - 22.Mouth center -- 17
  - 20.Mouth right corner -- 18



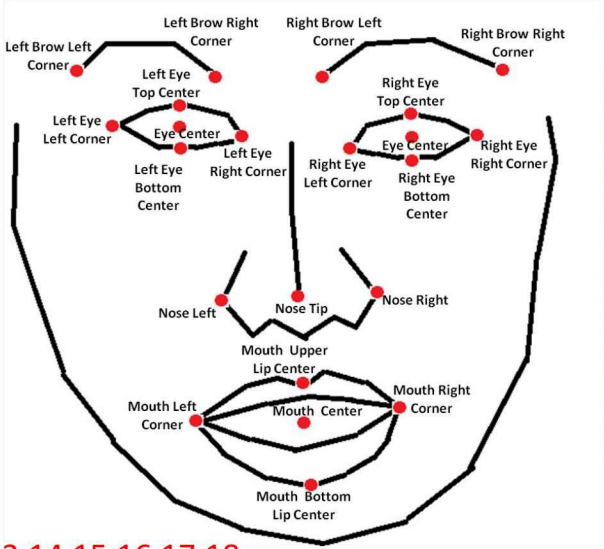


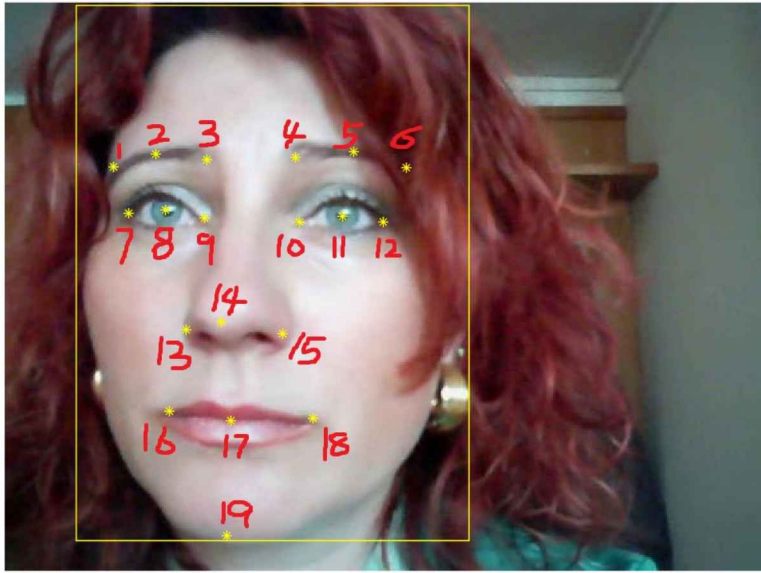
- WingLoss: 1,3,4,6,7,8,9,10,11,12,13,14,15,16,17,18,

DAC-CSR

- img_list.txt:
- record source image path and bbox [x1, y1, width, height].
- input:
- Original images, 20481 images
- MTCNN bbox
- output: 19 landmarks
- [x1, x2,...,x19, y1, y2,.....y19]

- For 19 detected landmarks:
  - Find the facial points that can be get their corresponding points in those 22 ground truth points for evaluation
- Find 16 points in total
  - 1.Left brow left corner -- 1
  - 2.Left brow right corner -- 3
  - 3.Right brow left corner -- 4
  - 4.Right brow right corner -- 6
  - 5.Left eye left corner -- 7
  - 9.left eye center -- 8
  - 7.Left eye right corner -- 9
  - 10.Right eye left corner -- 10
  - 14.right eye center -- 11
  - 12.Right eye right corner -- 12
  - 16.Nose left -- 13
  - 15.Nose tip -- 14
  - 17.Nose right -- 15
  - 18.Mouth left corner -- 16
  - 22.Mouth center -- 17
  - 20.Mouth right corner -- 18





- WingLoss: 1,3,4,6,7,8,9,10,11,12,13,14,15,16,17,18,

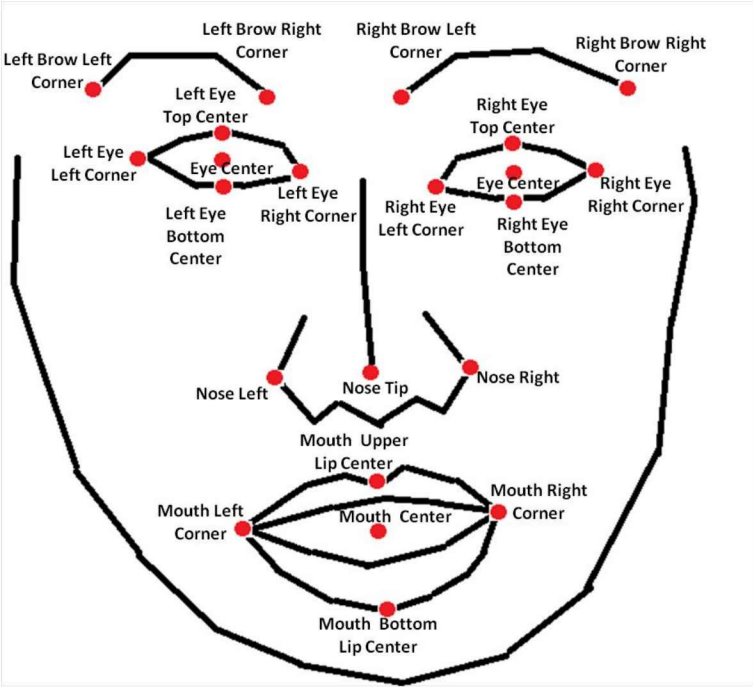
VillianCNN

- 1: left eye center
- 2: right eye center
- 3: nose tip
- 4: left mouth corner
- 5: right mouth corner

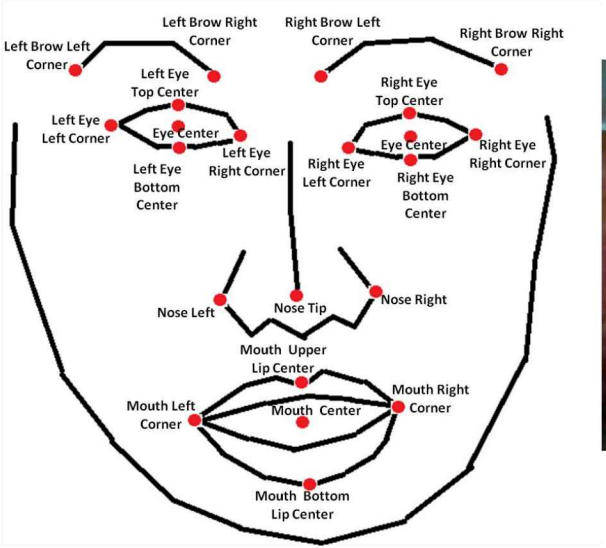
- input any size images,
- -- MOBIO Faces, we input 256*256!
- 20,481 images in total.
- resize to 40*40
- output: 40*40

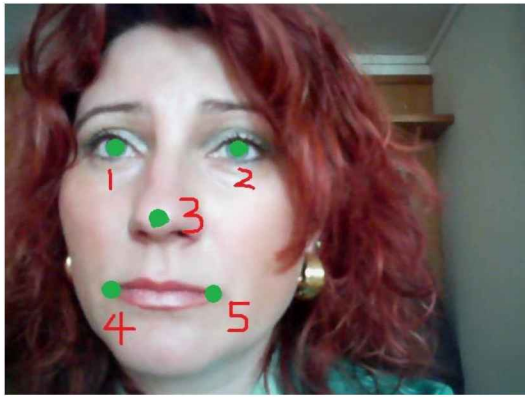
Ground Truth

- 1.Left brow left corner
- 2.Left brow right corner
- 3.Right brow left corner
- 4.Right brow right corner
- 5.Left eye left corner
- 6.left eye top center
- 7.Left eye right corner
- 8.left eye bottom center
- 9.left eye center
- 10.Right eye left corner
- 11.right eye top center
- 12.Right eye right corner
- 13.right eye bottom center
- 14.right eye center
- 15.Nose tip
- 16.Nose left
- 17.Nose right
- 18.Mouth left corner
- 19.mouth upper lip center
- 20.Mouth right corner
- 21.Mouth bottom lip center
- 22.Mouth center



- For 5 detected landmarks:
  - Find the facial points that can be get their corresponding points in those 22 ground truth points for evaluation
- Find 5 points in total
  - 9.left eye center -- 1
  - 14.right eye center -- 2
  - 15.Nose tip -- 3
  - 18.Mouth left corner -- 4
  - 20.Mouth right corner -- 5
- VillianCNN: 1,2,3,4,5

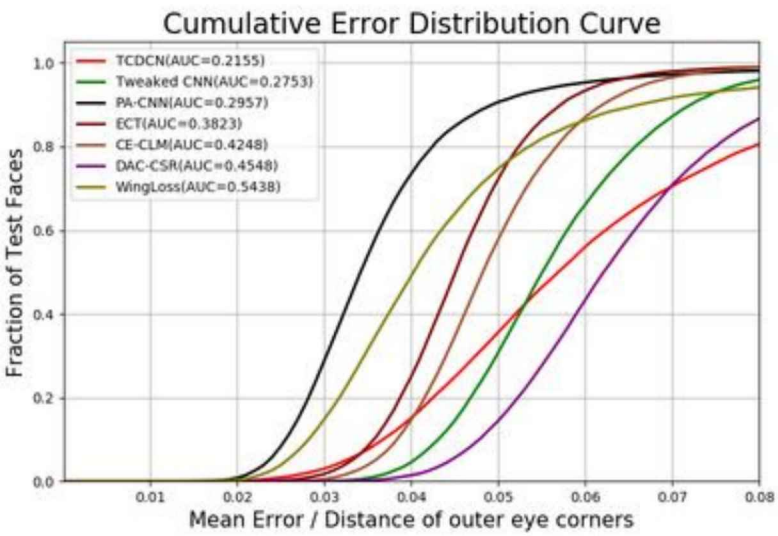




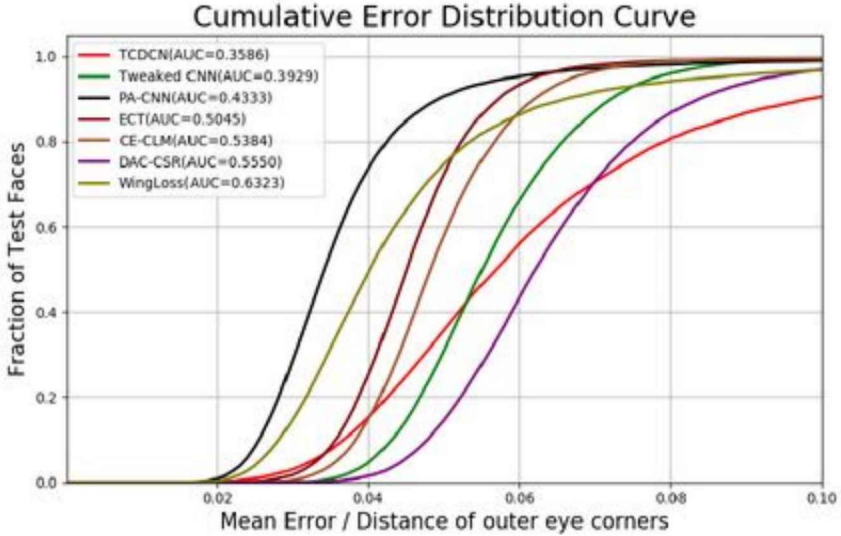
Final Result Comparison

TABLE II
EVALUATION RESULTS OF FACIAL LANDMARK DETECTION ON DEEP MODELS

Method	Normalized Mean Error (10^{-2})	Threshold=0.08		Threshold=0.10	
		AUC (%)	Failure Rate (%)	AUC (%)	Failure Rate (%)
Tweaked CNN [54]	6.4739049	27.533598	19.334993	39.288243	9.462429
WingLoss [40]	**3.8777522**	**54.384399**	1.904204	**63.232557**	1.010693
DAC-CSR [51]	4.6757547	45.475898	5.849324	55.507959	3.251794
PA-CNN [52]	5.7171261	29.574564	4.029736	43.333145	0.630608
CE-CLM [53]	4.7493759	42.482611	**0.990872**	53.840948	0.536926
ECT [55]	5.0704699	38.226405	1.079049	50.450906	**0.502905**
TCDCN [43]	6.5829441	21.545304	13.290367	35.863483	3.071139


Cumulative Error Distribution Curve
TCDCN(AUC=0.2155)
Tweaked CNN(AUC=0.2753)
PA-CNN(AUC=0.2957)
ECT(AUC=0.3823)
CE-CLM(AUC=0.4248)
DAC-CSR(AUC=0.4548)
WingLoss(AUC=0.5438)
Fraction of Test Faces
Mean Error / Distance of outer eye corners
0.0
0.2
0.4
0.6
0.8
1.0
0.01
0.02
0.03
0.04
0.05
0.06
0.07
0.08

(a)


Cumulative Error Distribution Curve
TCDCN(AUC=0.3586)
Tweaked CNN(AUC=0.3929)
PA-CNN(AUC=0.4333)
ECT(AUC=0.5045)
CE-CLM(AUC=0.5384)
DAC-CSR(AUC=0.5550)
WingLoss(AUC=0.6323)
Fraction of Test Faces
Mean Error / Distance of outer eye corners
0.0
0.2
0.4
0.6
0.8
1.0
0.02
0.04
0.06
0.08
0.10

(b)