

# Facial Micro-Expression Analysis in ADOS Videos

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# Face and Facial Expression



Happiness

Sadness

Disgust

Surprise

Contempt

Anger

Fear

# Facial Expression Analysis

- Has attracted great interest over the past years
  - Electronic / digital health
  - Human machine interaction
  - Human behavior analysis
  - Video communication



Digital Health [1]



Human Machine Interaction



Video Communication [3]

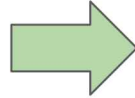


Behavior Analysis [2]

[1] Figure from <https://www.hospitaltimes.co.uk/digital-access-a-cause-and-solution-to-health-inequality/>  
[2] Figure from <https://hertasecurity.com/news/a-new-unprecedented-solution-for-human-behavior-analysis/>  
[3] Figure from <https://www.youtube.com/watch?v=1XyJNPa8Fwc>

# Socio-emotion Interaction in ASD

- **Socio-emotion interaction** difficulties
  - communication disorder
  - emotional dysregulation
  - rigid and repetitive behaviors



- **Problems** related to performance of
  - Expressive language
  - Social
  - Emotional adaptive skills

All individuals diagnosed with ASD, experience either **one or more aforementioned difficulties**, regardless of the severity levels of diagnosis

# Emotions in ASD

- Usually **do not show** the emotions **in a way that normal people** would be able to **recognize and understand**

➤ either they do **not respond** emotionally



➤ or their emotional responses might sometimes seem **over-reaction**



- With the advancement of CV and ML technology, there are much research that have been embarked around **recognizing human emotions**, particularly for autistic children and individuals
- This study focuses on showing **how to analyze emotions felt by the autistic persons**, not the capacity of the person to recognize emotions in others

# Existing Study on Emotion Recognition for ASD

- **Types of stimuli** to invoke emotions
  - **Picture**
    - a series of images that invoke the desired emotions
  - **Video**
    - bi-sensory stimulation that combines audio and visual stimulus
  - **Task**
    - directive real-life situation that elicit the desired emotions
- The **evoked emotions** includes some basic emotions, like
  - fear, anger, happiness, sadness, disgust, surprise, etc.
- Types of **extracted feature** about emotions
  - Optical flow, spatio-temporal, thermal intensity value(TIV) from thermal images, EEG, texture, etc.
- Types of **classifier**
  - K-NN, K-means, SVM, Decision Tree, GMM, CNN, etc.

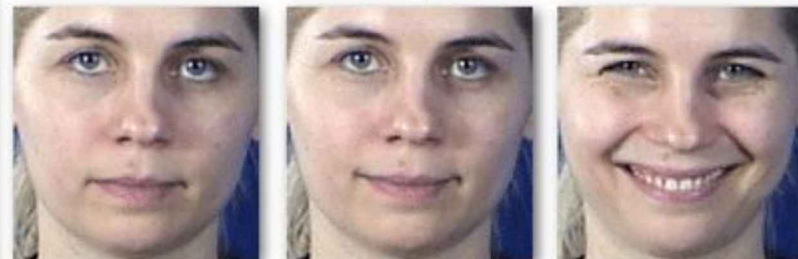
# Micro- vs. Macro- Expression

Traditional Emotion Recognition

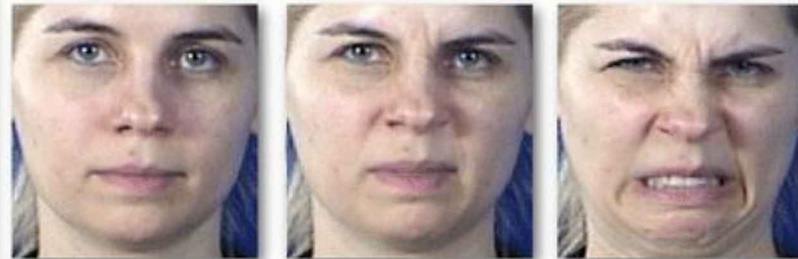
Micro expression

Macro expression

Happiness



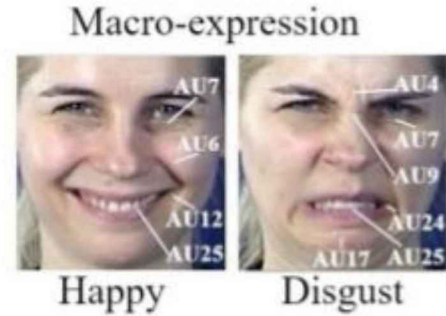
Disgust



Difference of motion intensity between micro and macro expression

## Main differences between macro- and micro- expressions

Difference	Micro-expression	Macro-expression
Noticeability	Easy to ignore	Easily noticed
Time interval	Short duration (0.065-0.5 seconds)	Long duration (0.5-4 seconds)
Motion intensity	Slight variation	Large variation
Subjectivity	Involuntary (uncontrollable)	Voluntary (under control)
Action areas	Fewer	Almost all areas



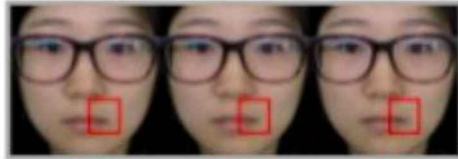
FACS, AU: Action Unit

- **posed** facial expressions (i.e., voluntary facial actions)
- **spontaneous** facial expressions (i.e., involuntary emotion)



# Micro-Expressions

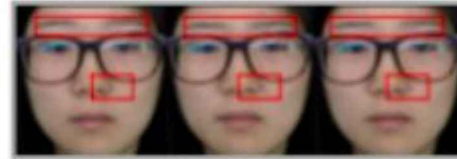
- Are characterized by **short durations** (less than **0.5** seconds), **involuntary** generation and **low intensity**
- Regarded as unique cues showing one's **hidden real emotions**
- Only **a few facial action units** are involved



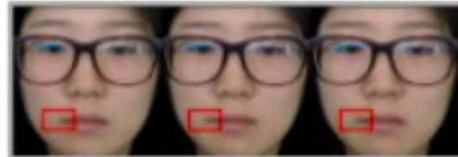
Happiness



Surprise



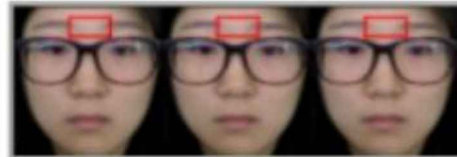
Disgust



Fear



Sadness



Others

# Emotion Classes

- There are many **different basic emotion definitions**

Characteristics	Datasets													
	MEVIEW	SMIC			CASME	CASME II	CAS(ME) <sup>2</sup>	SAMM	MMEW					
		HS	VIS	NIR										
Num of samples*	40	164	71	71	195	247	57	159	300					
Participants*	16	16	8	8	35	35	22	32	36					
Frame rate	25	100	25	25	60	200	30	200	90					
Resolution	1280×720	640×480			640×480 & 1280×720	640×480	640×480	2040×1088	1920×1080					
Facial resolution	N/A	190×230			150×190	280×340	N/A	400×400	400×400					
Emotion classes	7 categories: Happiness (6) Anger (2) Disgust (1) Surprise (9) Contempt (6) Fear (3) Unclear (13)		3 categories: Positive (107) Negative (116) Surprise (83)		8 categories: Amusement (5) Disgust (88) Sadness (6) Contempt (3) Fear (2) Tense (28) Surprise (20) Repression (40)		5 categories: Happiness (33) Repression (27) Surprise (25) Disgust (60) Others (102)		4 categories: Positive (8) Negative (21) Surprise (9) Others (19)		7 categories: Happiness (24) Surprise (13) Anger (20) Disgust (8) Sadness (3) Fear (7) Others (84)		7 categories: Happiness (36) Anger (8) Surprise (89) Disgust (72) Fear (16) Sadness (13) Others (102)	
Available labels	Emotion/FACS	Emotion		Emotion/FACS		Emotion/FACS		Emotion/FACS/ Video type		Emotion/FACS		Emotion/FACS		

- Most popular specific emotions: **7 types**
  - anger, disgust, fear, happiness, sadness, surprise, and contempt

# FACS: Facial Action Units of Different Micro-Expressions

## ● Happiness

- Eye area
  - Cheeks are raised
  - Lower eyelid may show wrinkles or be tense
- Mouth area
  - Corners of the lips are pulled
  - A wrinkle runs from outer nose to outer lip



## ● Sadness

- Eye area
  - Inner corners of the eyebrows are drawn in and then up
  - Skin below the eyebrows is triangulated, with inner corner up
  - Drooping upper eyelids
  - Losing focus in eyes
- Mouth area
  - Corner of the lips are drawn down
  - Jaw comes up



## ● Surprise

- Eye area
  - The eyebrows are raised and curved
  - Skin below the brow is stretched
  - Horizontal wrinkles show across the forehead
  - Eyes are widened; Eyelids are opened, white of the eye showing above and below.
- Mouth area
  - Jaw drops open and teeth are parted but there is no tension or stretching of the mouth.
  - Lip is Puckered



## ● Fear

- Eye area
  - Eyebrows are drawn together, usually in a flat line
  - Wrinkles in the forehead are in the center between the eyebrows, not across
  - Upper eyelid is raised, but the lower lid is tense and drawn up
  - Eyes have the upper white showing, but not the lower white
- Mouth area
  - Mouth is open or lips are slightly tensed or stretched and drawn back
  - Nasolabial area is deepened



- **Anger**

- **Eye area**

- **Eyelid is tightened**

- The eyebrows are lowered and drawn together

- Vertical lines appear between the eyebrows

- **Eyes are in hard stare** or bulging



- **Nose area**

- Nose is wrinkled

- Nostrils may be dilated



- **Mouth area**

- **Lower lip is tensed**

- Lips can be pressed firmly together, with corners down, or in a square shape as if shouting

- The lower jaw juts out



## ● Disgust

- Eye area
  - Eyes are narrowed
  - Eyebrows are lowered
  - Eyelids are tightened
- Nose area
  - Nose is wrinkled
- Mouth area
  - Upper lip is raised
  - Upper teeth may be exposed
- Cheeks
  - Cheeks are raised



## ● Contempt

- Mouth area
  - Lip corner is tightened and raised on only one side of the face



# Applications of Micro-expression (ME) Analysis

- Micro-expressions are **involuntary** and **transient** facial expressions
- Can reflect the **true emotions** that a person try to suppress, hide, mask, or conceal
- Provide **more important information** than macro-expression



**Lie Detection**



**Criminal Investigation**



**Clinical Diagnosis**

# Public Micro-Expression Databases

- Due to the intrinsic characteristics, such as
  - involuntariness
  - short duration
  - slight variation
- Eliciting micro-expressions in a controlled environment is challenging
- So far, a few micro-expression datasets have been developed
- However, most of these still have some deficiencies as regards their
  - elicitation paradigms
  - labelling methods
  - or small data size



### Seven Publicly Released Micro-Expression Datasets

Characteristics	Datasets								
	MEVIEW	SMIC			CASME	CASME II	CAS(ME) <sup>2</sup>	SAMM	MMEW
		HS	VIS	NIR					
Num of samples*	40	164	71	71	195	247	57	159	300
Participants*	16	16	8	8	35	35	22	32	36
Frame rate	25	100	25	25	60	200	30	200	90
Mean age	N/A	N/A			22.03	22.03	22.59	33.24	22.35
Ethnicities	N/A	3			1	1	1	13	1
Resolution	1280×720	640×480			640×480 & 1280×720	640×480	640×480	2040×1088	1920×1080
Facial resolution	N/A	190×230			150×190	280×340	N/A	400×400	400×400
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Available labels	Emotion/FACS	Emotion			Emotion/FACS	Emotion/FACS	Emotion/FACS/ Video type	Emotion/FACS	Emotion/FACS
Download URL	<a href="http://cmp.felk.cvut.cz/cechj/ME/">http://cmp.felk.cvut.cz/cechj/ME/</a>	<a href="http://www.cs.e.oulu.fi/SMIC/Database">http://www.cs.e.oulu.fi/SMIC/Database</a>			<a href="http://fu.psych.ac.cn/CASME/casme-en.php">http://fu.psych.ac.cn/CASME/casme-en.php</a>	<a href="http://fu.psych.ac.cn/CASME/casme2-en.php">http://fu.psych.ac.cn/CASME/casme2-en.php</a>	<a href="http://fu.psych.ac.cn/CASME/cas(me)2-en.php">http://fu.psych.ac.cn/CASME/cas(me)2-en.php</a>	<a href="http://www2.docm.mmu.ac.uk/STAFF/M.Yap/dataset.php">http://www2.docm.mmu.ac.uk/STAFF/M.Yap/dataset.php</a>	<a href="http://www.dpailab.com/database.html">http://www.dpailab.com/database.html</a>

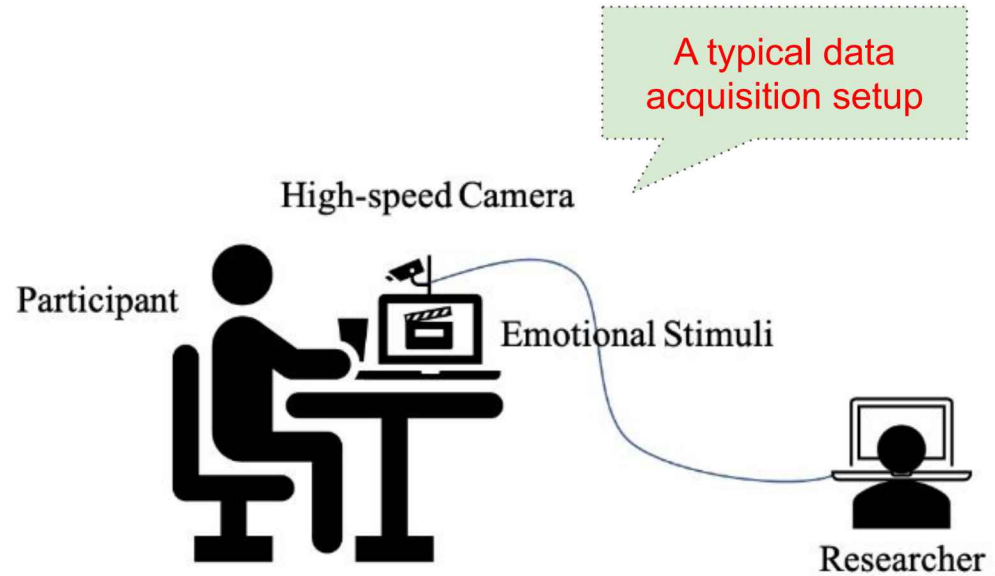
Here, samples\* refers to the samples of micro-expressions. Note that CAS(ME)<sup>2</sup> and MMEW also contain samples of macro-expressions. Not all participants produce micro-expressions; here, Participants\* refers to the number of participants who generate micro-expressions. The parentheses enclose the number of emotional video clips in that category.

# HS: high-speed camera; VIS: normal visual camera; NIR: near-infrared camera

# Paradigm of Elicitation

An effective method

- **Watching an emotional video while maintaining a neutral expression** (i.e. suppressing emotions)

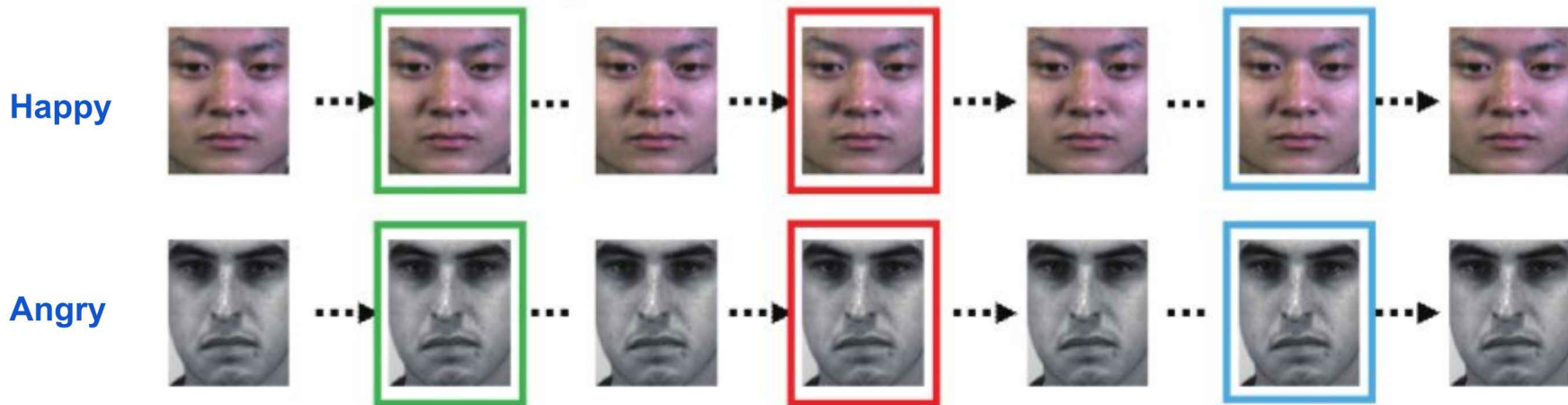


**Figure 13.** For the purposes of data collection, participants watch an emotional video while their faces are imaged by high-speed camera.

Five datasets — SMIC, CASME, CASME II, SAMM and CAS(ME)2 — used this elicitation paradigm.

# Micro-Expression Spotting

- to find the time interval([onset, apex, offset]) at which micro-expression are detected
  - **onset**: the first frame at which a ME starts
    - i.e., changing from the baseline, which is usually the neutral facial expression
  - **apex**: the frame at which the highest intensity of the facial expression is reached
  - **offset**: the last frame at which a ME ends
    - i.e., returning back to the neutral facial expression

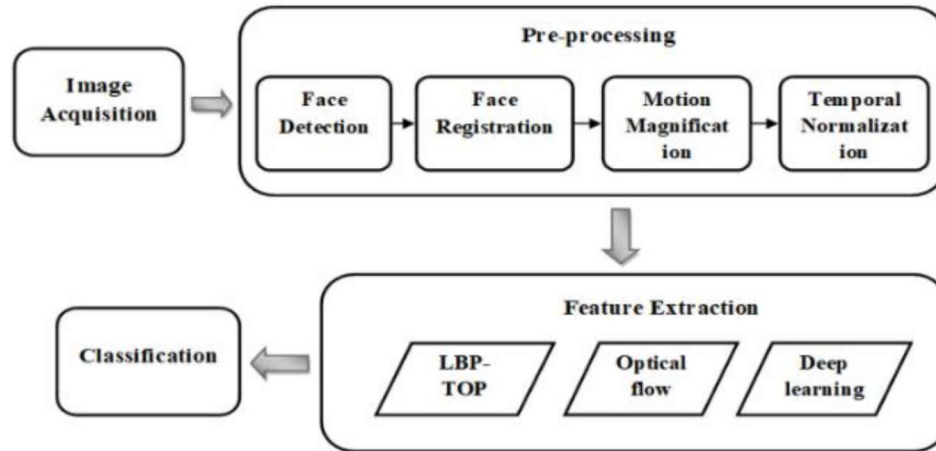


- A prerequisite for advanced facial ME analysis
- Proper spotting can decrease the redundant information for further analysis
- **Three major challenges**
  - (1) relies on setting the optimal thresholds to detect ME for any given feature
  - (2) different people may perform different extra facial actions
    - some people blink habitually, while other people sniff more frequently
  - (3) when recording videos, many comprehensive factors may significantly influence the micro-expression spotting
    - head movement
    - physical activity
    - recording environment
    - Lighting condition

- **A discriminative feature**
  - should be able to capture the **difference in both spatial and temporal domains**
  - should be able to capture the **micro-level differences**
- Existing spotting algorithms can be broadly divided into **two classes**:
  - **Hand-crafted Feature**
    - Optical flow based
    - Feature descriptor based
      - Gradient based Feature: histogram of gradients (HOG), HOG-TOP
      - Local Binary Pattern (LBP) based Feature
      - LBP-TOP variants
  - **Learning based Feature**
    - Single stream learning: 2D CNN, 3D CNN (motion)
    - Shallow learning: limited and imbalanced data
    - Two-step learning: spatio+temporal feature

# Micro-Expression Recognition

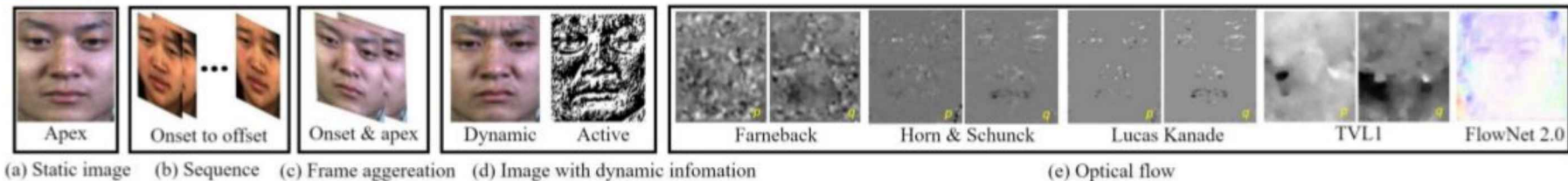
- Attempts to **categorize a video** of micro-expressions into one of the expression classes
  - happiness, sadness, disgust, fear, contempt, anger, surprise, etc.



- Various inputs

Input	Strength	Shortcoming
Apex	Efficient; Take advantage of massive facial images	Require magnification and apex detection
Sequence	Process directly	Not efficient; Redundancy
Frame aggregation	Efficiently leverage key temporal information	Require apex detection
Image with dynamic information	Efficiently embedding spatio-temporal information	Required dynamic information computation
Optical flow	Remove identity in some degree	Optical flow computation is necessary
Combination	Fully exploring spatial and temporal information	High computation

multiple inputs



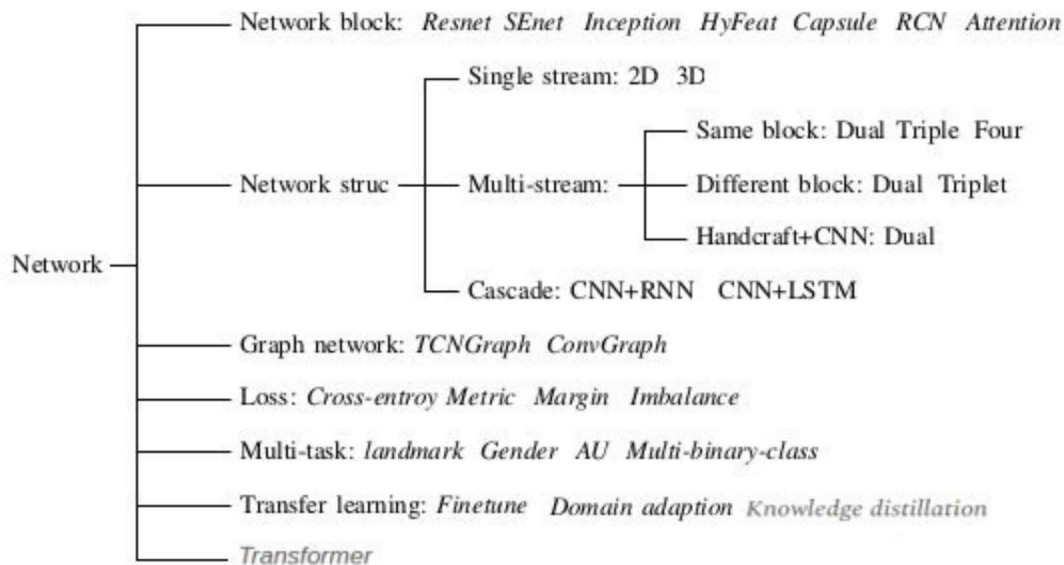
\* **Image with dynamic information**: a standard image that holds the dynamics of an entire video sequence in a single instance.

- Existing methods

- Traditional classifiers

- SVM, KNN, extreme learning machine (ELM), etc.

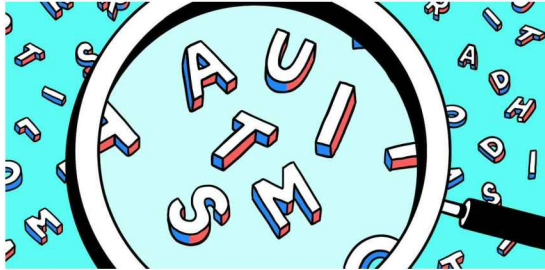
- Deep learning based





# Our Work

- Previous study on facial expression analysis for ASD mainly focus on macro-expression analysis
- However, micro-expression can reflect more true feelings of individuals



In our work:

- Use computer vision and machine learning technology
- to **analyze facial micro-expressions** of participants
- in **hour-long ADOS video sequences**
- for the diagnosis of ASD

# ADOS Videos

- Structured but natural discussion
- Participants with ASD underwent the ADOS interview
- ADOS: Autism Diagnostic Observation Schedule
  - 15 observation activities
  - Response to each activity is inspected



1. Construction Task

2. Telling A Story

3-4. Describing A  
Picture & Talking

5-7. Conversation on  
School, Work, Social  
Difficulties & Emotions

8. Demonstration Task

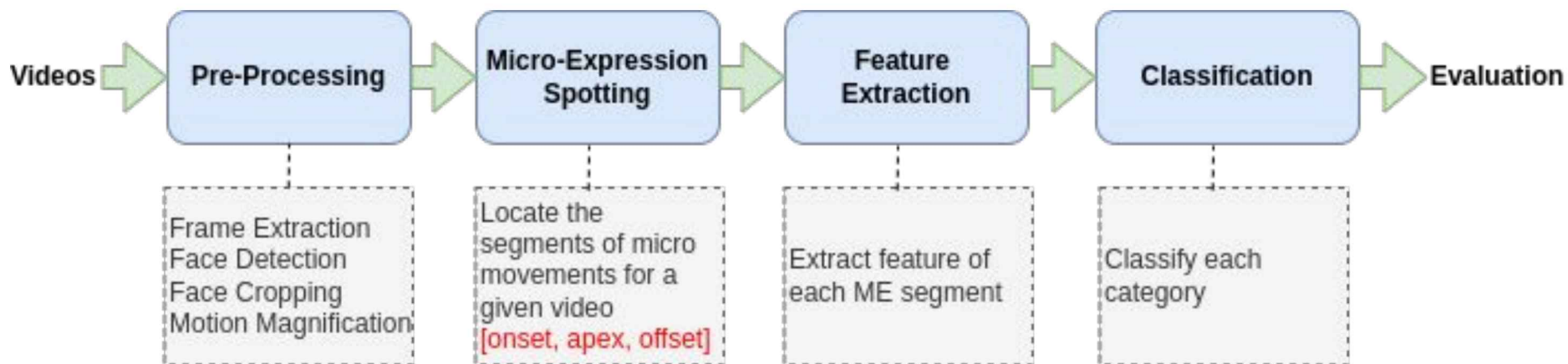
9. Cartoons

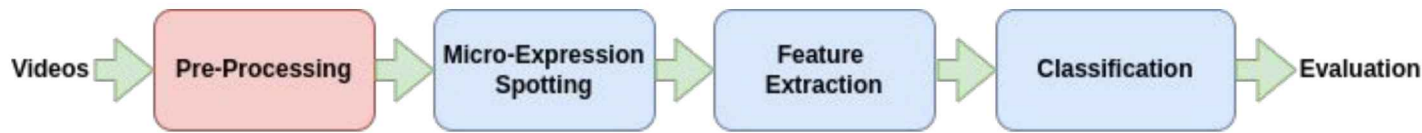
10. Break

11-14. Conversation on  
Daily Living,  
Relationships, Plans

15. Creating A Story

# Pipeline





# Pre-process

- Split the whole video into 15 separate subvideos based on the 15 tasks
- For the chosen tasks (5-7, 11-14)
  - Extract frames
  - Detect the face of the participant
  - Crop square face regions

**Extract Frames**

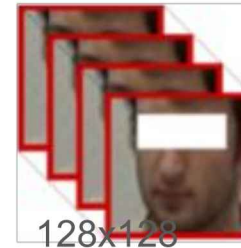


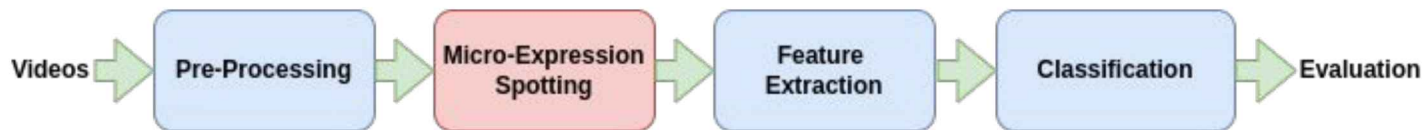
**Video**

**Face Detection & Cropping**



Dlib toolbox

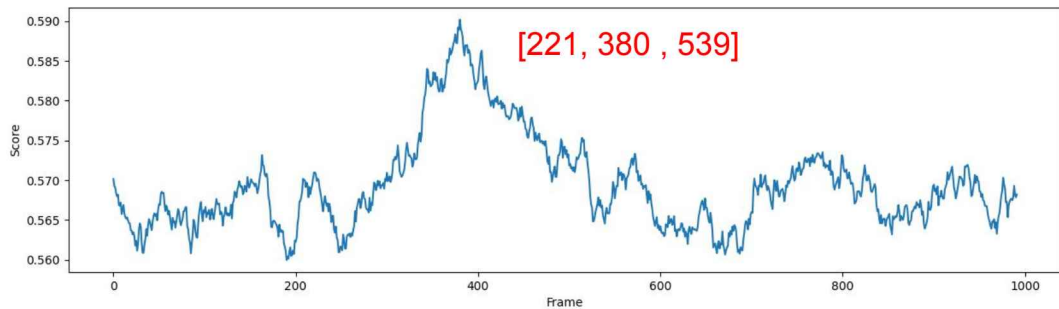




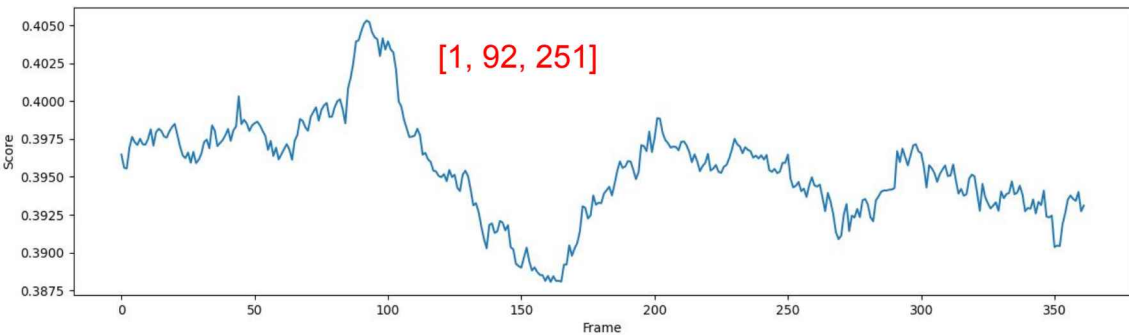
- In our experiment

- Spotting model is trained on **CAS(ME)<sup>2</sup>**
  - **4 categories** (more general terms):
    - Positive: **Happiness**
    - Negative: **Anger, Disgust, Sadness, Fear**
    - Surprise: **Surprise**
    - Others: **Contempt**, sympathy, confusion, helplessness, etc.
  - The emotion labeling criteria is based on the **FACS coding** results
- For each spotted segment in ADOS videos
  - **35 frames** for 30 fps (frame rate) videos
    - Calculated by averaging segment length of CAS(ME)<sup>2</sup>

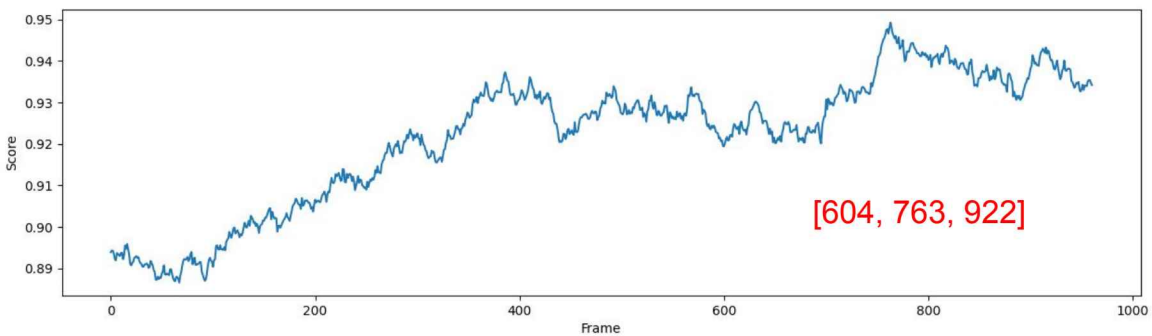
\* 7 scenes



Eyebrows raised and pulled together



Narrowing of the lips; Drooping upper eyelids



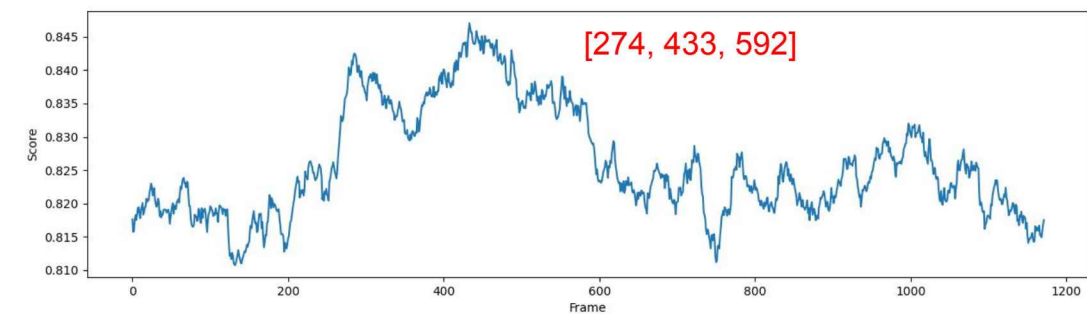
Onset

Apex

Offset



Narrowing of the lips; slight pulling down of lip corners; Drooping upper eyelids



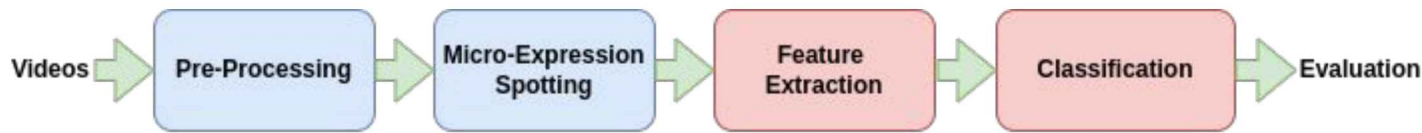
Onset

Apex

Offset



Lose focus in eyes



## Next work

- So far, we have obtained all the spotted intervals of ADOS videos
- Next step:
  - extract discriminative feature of each interval for autism diagnosis
- Basic Idea: **use a transformer and optical flow**
  - Transformer
    - inductive bias – modeling long-range relations between input patches
    - the pattern of ME is **subtle** and appears only in **a part of each frame** in the video
    - **might seem more suitable for micro-expression analysis**
  - optical flow
    - as an effective **motion feature**
    - **can complement the lack of data**

Transformer: Liu, Z.; Ning, J.; Cao, Y.; Wei, Y.; Zhang, Z.; Lin, S.; Hu, H. Video swin transformer. arXiv 2021, arXiv:2106.13230.

OF: Farneback, G. Two-frame motion estimation based on polynomial expansion. In Scandinavian Conference on Image Analysis; Springer: Berlin/Heidelberg, Germany, 2003; pp. 363–370.



# Evaluation Protocol

- **10-fold cross-validation**

- measures the performance of the model using 10% videos as a validation set and the rest as a training set.
- We repeat the training and validation process 10 times
- 10 **Accuracy/F1-score** values are averaged to produce the final result.

# Summary

- Gave a brief **overview** of existing macro and micro expression analysis technology, and applications in ASD diagnosis field
- Reviewed the **pipeline** of the micro-expression analysis in ADOS videos
- Showed some **spotting result**
- Discussed a **feasible plan** for next step