

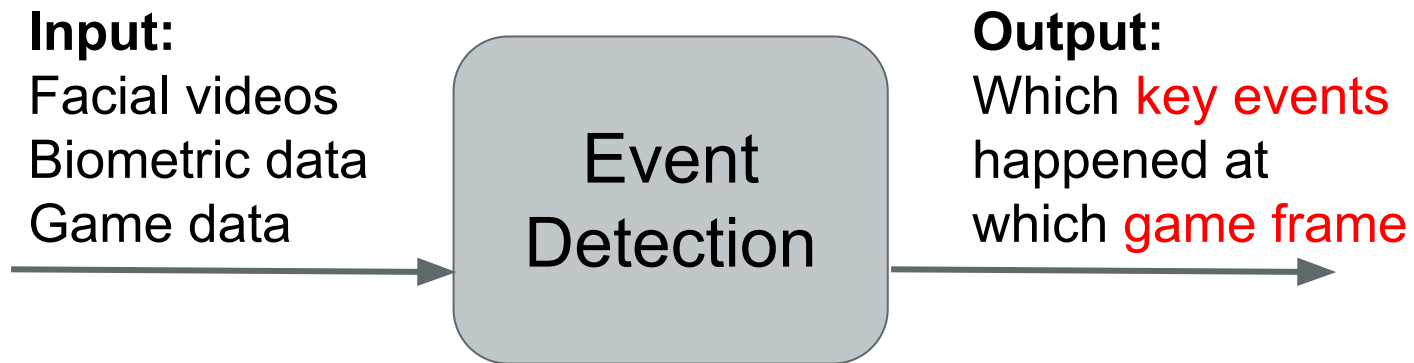
Emotional Mario: A Games Analytics Challenge

Mediaeval 2021

[Link](#)

Introduction

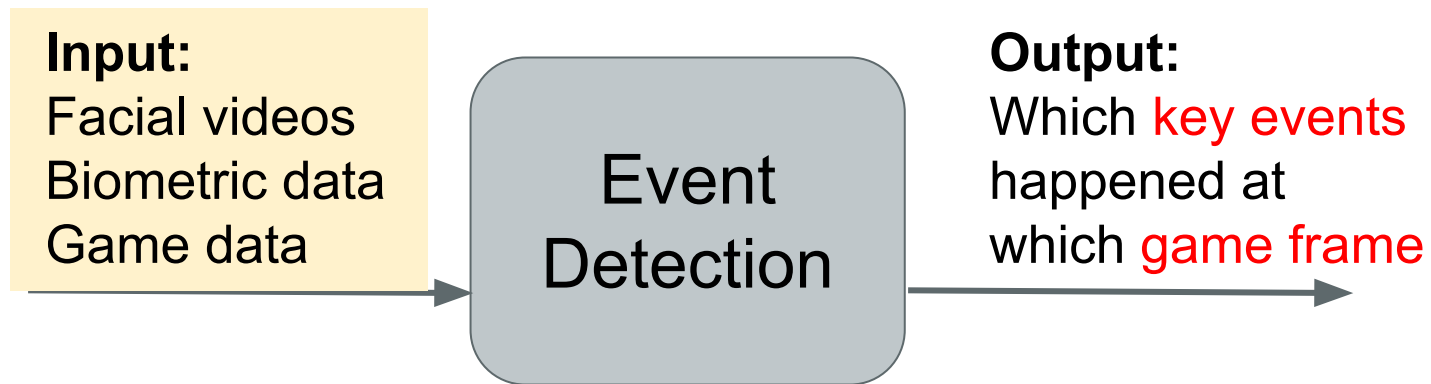
- A grand challenge from [MediaEval](#)
- Motivation: game affects users' emotions → the connection between emotion and games
- **Task 1: Event detection**



- **Task 2: Gameplay summarization**

Introduction

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- **Task 2: Gameplay summarization**

Dataset: Toadstool

- The dataset consists of video, biometric data collected from 10 participants playing Super Mario Bros
- Participants wore wristband to collect the biometric data
- The dataset is separated into two main folders:

- - consent.pdf
 - LICENSE
 - **participants**
 - protocol.pdf
 - questionnaire_answers.csv
 - questionnaire.pdf
 - README.txt
 - **scripts**

- **participants:** facial videos, biometric data, and game data from the ten participants
- **scripts:**
 - Scripts about how they synchronize the game data with other data
 - A script to get mario game frame

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- **scripts:**
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Dataset: Toadstool-Participants




- In the participants folder, we have data for the 10 participants
- E.g., data in participant_0

```
.
├── participant_0_gap_info.json
├── participant_0_sensor
│   ├── ACC.csv
│   ├── BVP.csv
│   ├── EDA.csv
│   ├── HR.csv
│   ├── IBI.csv
│   ├── info.txt
│   └── TEMP.csv
├── participant_0_sensor_raw
│   ├── ACC.csv
│   ├── BVP.csv
│   ├── EDA.csv
│   ├── HR.csv
│   ├── IBI.csv
│   ├── info.txt
│   ├── tags.csv
│   └── TEMP.csv
├── participant_0_session.json
├── participant_0_video.avi
└── participant_0_video_info.json
```

Dataset: Toadstool-Facial video in Participants

- In the participants folder, we have data for the 10 participants
- E.g., data in participant_0

- participant_0_gap_info.json
- **participant_0_sensor**
 - ACC.csv
 - BVP.csv
 - EDA.csv
 - HR.csv
 - IBI.csv
 - info.txt
 - TEMP.csv
- **participant_0_sensor_raw**
 - ACC.csv
 - BVP.csv
 - EDA.csv
 - HR.csv
 - IBI.csv
 - info.txt
 - tags.csv
 - TEMP.csv
- participant_0_session.json
- **participant_0_video.avi**
- participant_0_video_info.json



A facial video

`{"start_time": 1579785027.233421, "stop_time": 1579787135.8056052}`

Dataset: Toadstool-Biometric Data in Participants

- In the participants folder, we have data for the 10 participants
- E.g., data in participant_0

```
├── participant_0_gap_info.json
├── participant_0_sensor
│   ├── ACC.csv
│   ├── BVP.csv
│   ├── EDA.csv
│   ├── HR.csv
│   ├── IBI.csv
│   ├── info.txt
│   └── TEMP.csv
├── participant_0_sensor_raw
│   ├── ACC.csv
│   ├── BVP.csv
│   ├── EDA.csv
│   ├── HR.csv
│   ├── IBI.csv
│   ├── info.txt
│   ├── tags.csv
│   └── TEMP.csv
├── participant_0_session.json
├── participant_0_video.avi
└── participant_0_video_info.json
```

Raw biometric data collected from the wristband:

- **ACC**: the movement of the wearer
- **EDA**: the electrical conductivity of the skin
- **BVP**: blood Volume Pulse
- **IBI**: the time interval between individual heartbeats → the instantaneous heart rate/heart rate variability
- **HR**: the average heart rate values
- **TEMP**: the temperature of the person 8

Dataset: Toadstool-Biometric Data in Participants

- Data for the 10 participants
- E.g., data in participant_0

```
— participant_0_gap_info.json
— participant_0_sensor
  — ACC.csv
  — BVP.csv
  — EDA.csv
  — HR.csv
  — IBI.csv
  — info.txt
  — TEMP.csv
— participant_0_sensor_raw
  — ACC.csv
  — BVP.csv
  — EDA.csv
  — HR.csv
  — IBI.csv
  — info.txt
  — tags.csv
  — TEMP.csv
— participant_0_session.json
— participant_0_video.avi
— participant_0_video_info.json
```

Synchronized biometric data collected from the wristband to the game frame:

- **ACC**: the movement of the wearer
- **EDA**: the electrical conductivity of the skin
- **BVP**: blood Volume Pulse
- **IBI**: the time interval between individual heartbeats → the instantaneous heart rate/heart rate variability
- **HR**: the average heart rate values
- **TEMP**: the temperature of the person

Dataset: Toadstool-Game Data in Participants

- In the participants folder, we have data for the 10 participants
- E.g., data in participant_0

```
.
├── participant_0_gap_info.json
├── participant_0_sensor
│   ├── ACC.csv
│   ├── BVP.csv
│   ├── EDA.csv
│   ├── HR.csv
│   ├── IBI.csv
│   ├── info.txt
│   └── TEMP.csv
├── participant_0_sensor_raw
│   ├── ACC.csv
│   ├── BVP.csv
│   ├── EDA.csv
│   ├── HR.csv
│   ├── IBI.csv
│   ├── info.txt
│   ├── tags.csv
│   └── TEMP.csv
├── participant_0_session.json
├── participant_0_video.avi
└── participant_0_video_info.json
```

Where is the game data?

Dataset: Toadstool-Game Data in Participants

- Use `replay_game_session.py` in *scripts* folder with `participant_[num]_session.json` in *participants* folder to get the game frame

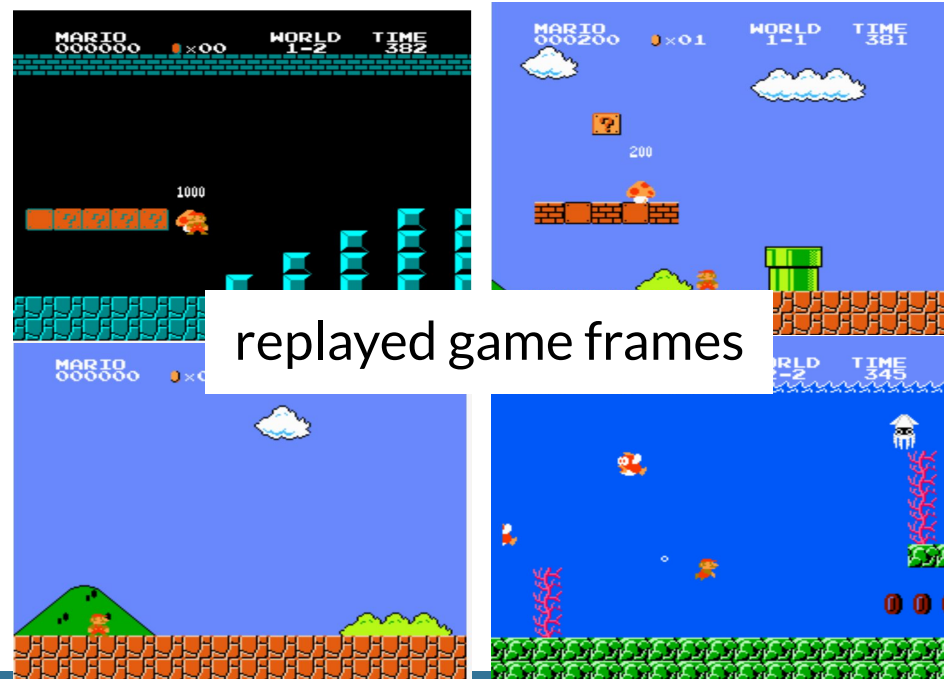
- `participant_0_gap_info.json`
- **`participant_0_sensor`**
 - `ACC.csv`
 - `BVP.csv`
 - `EDA.csv`
 - `HR.csv`
 - `IBI.csv`
 - `info.txt`
 - `TEMP.csv`
- **`participant_0_sensor_raw`**
 - `ACC.csv`
 - `BVP.csv`
 - `EDA.csv`
 - `HR.csv`
 - `IBI.csv`
 - `info.txt`
 - `tags.csv`
 - `TEMP.csv`
- `participant_0_session.json`
- **`participant_0_video.avi`**
- `participant_0_video_info.json`

Data in *scripts* folder:

- `replay_game_session.py`
- `synchronize_game_with_face_data.py`
- `synchronize_game_with_sensor_data.py`
- `synchronize.sh`


Command:

`python3 replay_game_session.py -i [path]/participant_[idx]_session.json -o [folder path to store the generated frame images]`



Dataset: Toadstool

Check out their perprint and github for more information


 OSFHOME ▼

Toadstool - A Dataset for Training Emot... Files Wiki Anal

interesting for a manifold of researchers to explore different exciting questions.


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Has supplemental materials for [Toadstool: A Dataset for Training Emotional Intelligent Machines Playing Super Mario Bros](#) on OSF Preprints

Wiki 

Training Emotional Intelligent Machines Playing Super Mario Bros

[\[preprint\]](#) [\[github\]](#)

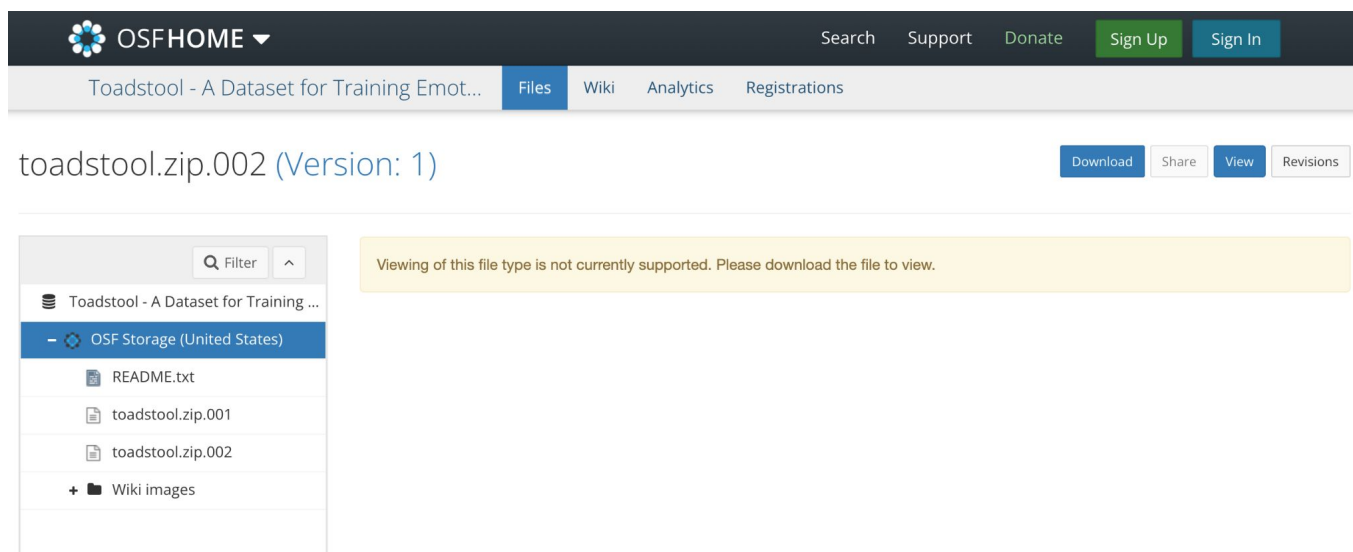


[Read More](#)



How to Download the Dataset?

- Click the [dataset link](#)
- Download toadstool.zip.001 and toadstool.zip.002
- Use 7 zip to unzip the dataset
- Ubuntu command:
 - `sudo apt install p7zip-full`
 - `7z x toadstool.zip.001`

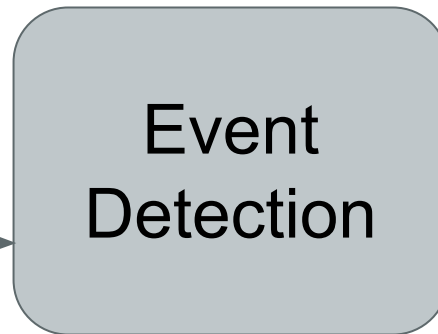


Introduction

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- **Task 1: Event detection**

Input:

Facial videos
Biometric data
Game data



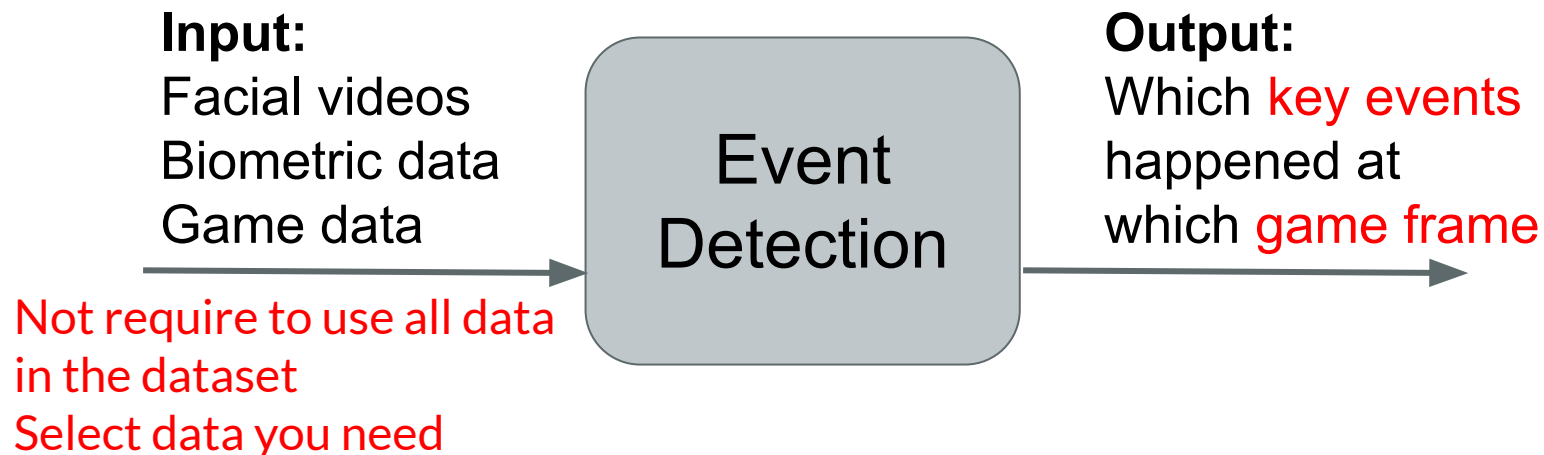
Not require to use all data
in the dataset

Select data you need

- **Task 2: Gameplay summarization**

Introduction

- A grand challenge from [MediaEval](#)
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- **Task 2: Gameplay summarization**

Output

Frame number:

1

Select numbers within “obs_n” in
participant_{idx}_session.json

```
— participant_0_gap_info.json
— participant_0_sensor
  — ACC.csv
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  — EDA.csv
  — HR.csv
  — IBI.csv
  — info.txt
  — TEMP.csv
— participant_0_sensor_raw
  — ACC.csv
  — BVP.csv
  — EDA.csv
  — HR.csv
  — IBI.csv
  — info.txt
  — tags.csv
  — TEMP.csv
— participant_0_session.json
— participant_0_video.avi
— participant_0_video_info.json
```

Key events:

2

- **new_stage**: at the very beginning of a new stage, except the first one, which starts at frame_number 1
- **flag_reached**: when the flag, i.e. the level end is reached
- **status_up**: when a mushroom or flower (power up) is consumed by the player
- **status_down**: when a player encounters a monster and loses a power up
- **life_lost**: when a player loses one of Mario's lives, note that the game is in endless mode

Sample Output

- Output your results into a csv files in format: **frame_num**,**event**
- There must be a “,” between **frame_num** and **event**



	frame_num	event
1	575	stauts_down
2	1675	stauts_down
3	1725	status_up
4	1875	flag_reached
5	2125	life_lost
6	2825	new_stage
7	2975	stauts_down
8	3025	flag_reached
9	3425	life_lost

- Name the output csv file as “**participant_[num]_events.csv**”, replace [num] with the index of participants
- **Do not** put headers in the csv file

Groundtruth

- Download the groundtruth [here](#)
- We only provide the groundtruth of 7 participants
- The rest of the groundtruth (participants 2, 4, 7) are hidden testcases

```
[  
  {"event": "status_up", "frame_number": 812},  
  {"event": "flag_reached", "frame_number": 4229},  
  {"event": "new_stage", "frame_number": 4230},  
  {"event": "status_up", "frame_number": 4719},  
  {"event": "status_down", "frame_number": 5849},  
  ...  
]
```

Evaluation Script

- We will provide an evaluation script
- Usage: **python3 evaluate.py**
 - i [input file folder]**
 - t [groundtruth folder]**
 - n [index of participant]**
 - You can put multiple number behind -n to evaluate multiple results at one time
 - Put the python files and the two file folders under the same folder
- You need to name your input files in the format of
“participant_[num]_events.csv”

```
run
├── participant_2_events.csv
├── participant_4_events.csv
├── participant_7_events.csv
└── README.md

truth
├── participant_0_events.json
├── participant_1_events.json
├── participant_2_events.json
├── participant_3_events.json
├── participant_4_events.json
├── participant_5_events.json
├── participant_6_events.json
├── participant_7_events.json
├── participant_8_events.json
├── participant_9_events.json
└── README.md
```

```
> python3 evaluate.py -i run -t truth -n 2 4 7
[2, 4, 7]
run identifier,precision,recall,f1
participant_2_events.json,0.007680491551459293,0.025380710659898477,0.01179245283018868
participant_4_events.json,0.007877813504823151,0.30434782608695654,0.015358094342579535
participant_7_events.json,0.006752411575562701,0.2937062937062937,0.0132013201320132
```

Evaluation Details and Grading

1. Search the frame number shown in groundtruth in your input file (Frame number +/- 25 is accepted)
2. Compare the event title in ground truth and your results
3. Calculate f1 score

$$F_1 = 2 \cdot \frac{\text{precision} \cdot \text{recall}}{\text{precision} + \text{recall}}$$

Grading:

- After your presentation, you will need to generate the results of participants 2, 4, and 7 and send the csv files to TAs right away
- We will use the avg. **f1 score** of participants 2, 4, and 7 to grade your project. The number will be rounded to four decimal places.
E.g. 0.3456 → 0.346
- **Benchmark** Scores
 - top 1/3: 10 points
 - next 1/3: 7 points
 - last 1/3: 4 points
 - **Note: if your f1 score is lower than baseline (0.02), you will get 0 points**

Baseline

- We will provide the baseline code: baseline1.py
- It can be separated into 2 steps:
 - Find event may happen at which frame
 - Read BVP.csv for the participants as an array
 - Find the local maxima of the BVP array
 - Select the frames with local maxima as the target frames
 - Guess the event of the target frames randomly
 - We have 5 events. We set each event a number (0 ~ 4)
 - Randomly select a number from 0~9 for each target frame
 - If the number is between 0 to 4, we set an event to the frame based on the number
- Average F1 score for the testing set: ~ 0.02
- Run the baseline1.py:

```
python3 baseline1.py -p [path to participants directory in the dataset]  
-i [index of the participants] -o [location to save the output csv file]
```

Other Ideas

- Create meaningful input data from the dataset
 - Use [FER](#) (facial expression recognition) to recognize the emotion from facial videos
 - [Output sample results](#): csv files with emotion analysis for each participant

```
|,box,angry,disgust,fear,happy,sad,surprise,neutral
0,"(273, 269, 139, 139)",0.01,0.0,0.04,0.07,0.06,0.0,0.81
1,"(275, 270, 138, 138)",0.01,0.0,0.03,0.08,0.03,0.0,0.86
2,"(274, 272, 137, 137)",0.01,0.0,0.04,0.05,0.05,0.0,0.85
3,"(276, 273, 134, 134)",0.01,0.0,0.04,0.08,0.03,0.0,0.85
4,"(278, 275, 130, 130)",0.0,0.0,0.04,0.07,0.02,0.0,0.86
```

- Classifiers
 - Use random forest to classify each frame
 - Use RNN to consider contextual data
 - ...

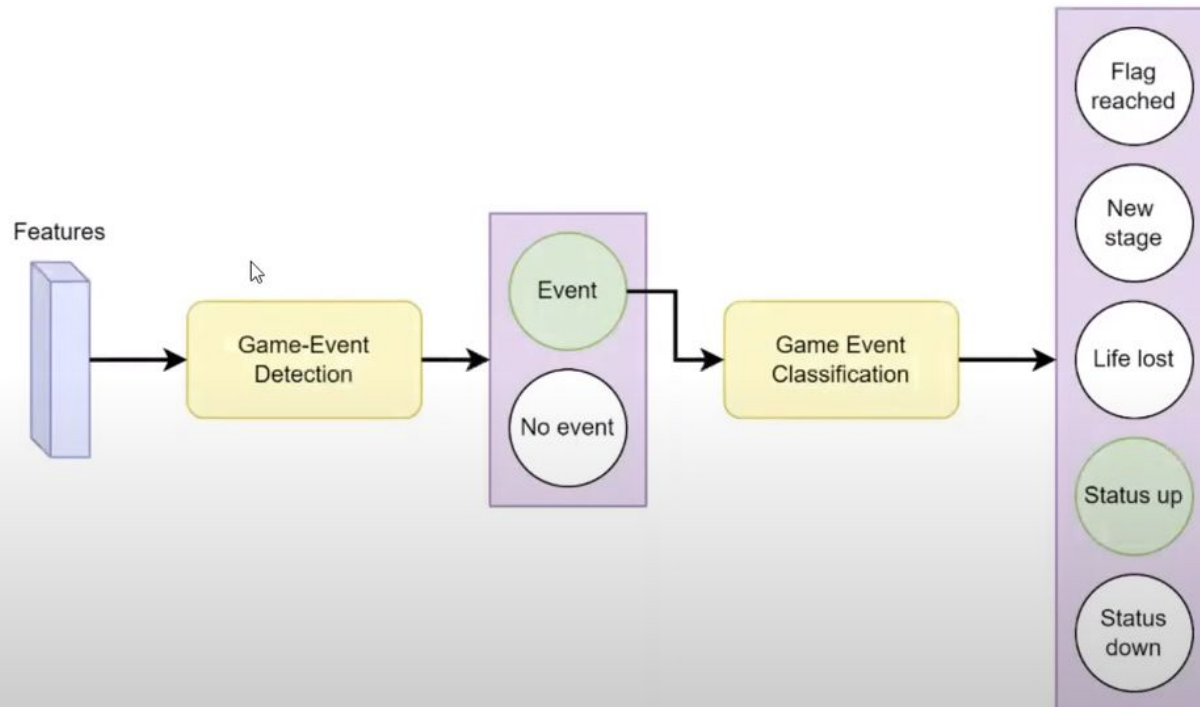
Conclusion



- Get the input data from the dataset: [toadstool](#)
- The groundtruth of the training set: [groundtruth](#)
- We will provide the evaluation script, the baseline scripts, and a sample output: participant_0_events.csv on elearn or the website
- **Demo** to generate the results of the testing set (participants 2, 4, and 7) and send the csv files to TAs **before the report deadline**. We will announce the detailed demo time later.
- The format of the csv files must follow what we just show in the previous slides
- Source code should be uploaded to elearn as well

Possible Solution

- Game-Event Detection:
 - Input: FER, BVP, Electrodermal Activity
- Game Event Classification:
 - Input: game played images, BVP
 - Use ResNet50 to classify the event



Any Questions?

Chia-Ying Hsieh

cyinghsieh@gmail.com

Tzu-Yi Fan

joyfan2@gmail.com
