

ICPR 2020 - Online

**3D attention mechanism for
fine-grained classification
of table tennis strokes
using a Twin
Spatio-Temporal
Convolutional Neural
Networks**

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MATHEMATIQUES IMAGE APPLICATIONS
UNIVERSITÉ DE LA ROCHELLE

La Rochelle
Université



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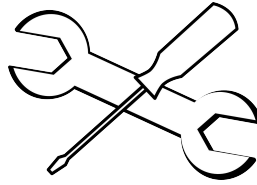
Renaud Péteri

Julien Morlier

Goal



Improve athletes performances

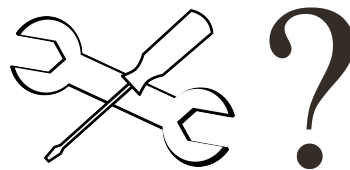


through tools



for trainers and athletes

Goal



Input

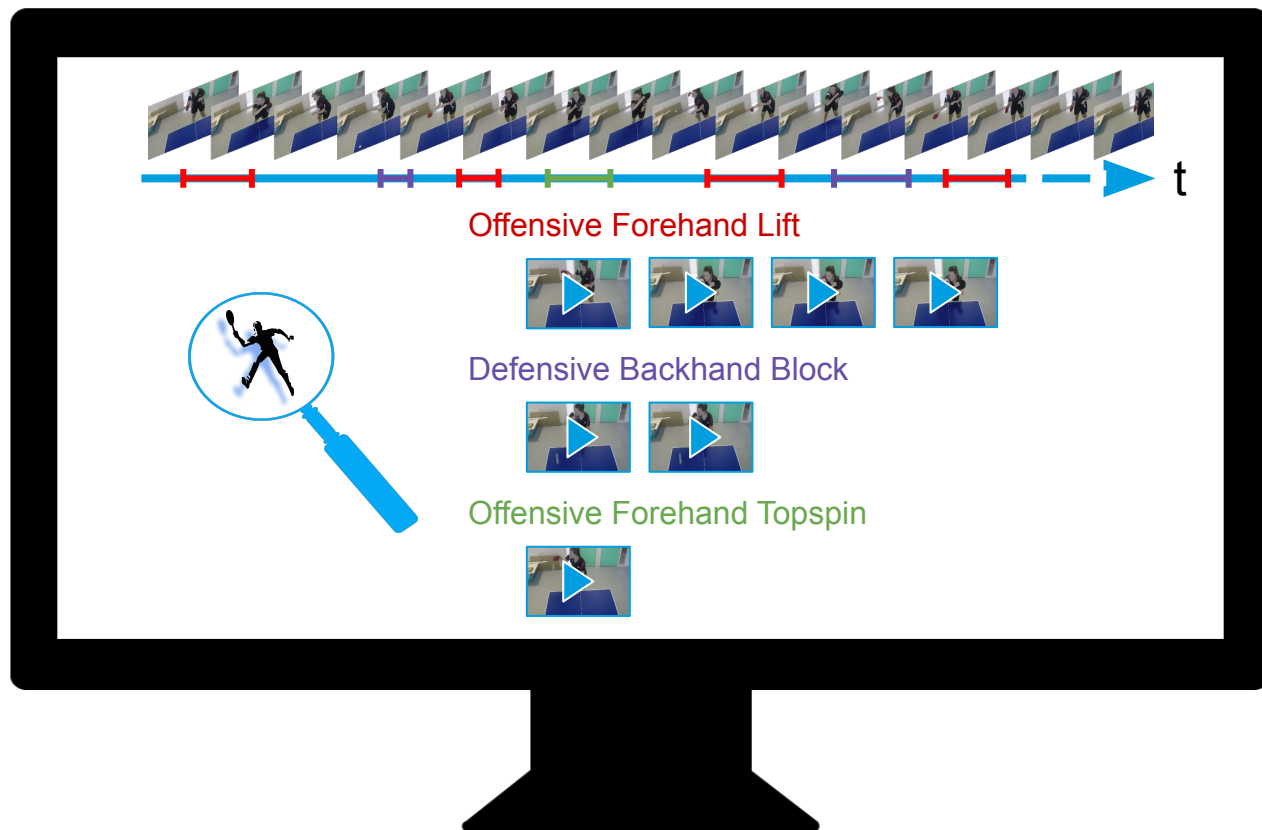


Output



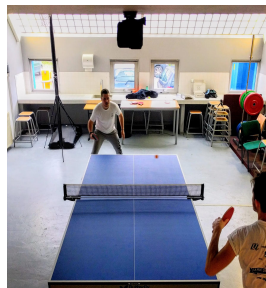
Offensive Backhand Flip

Goal

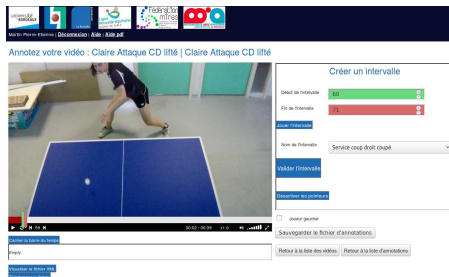


TTStroke-21 dataset [1]

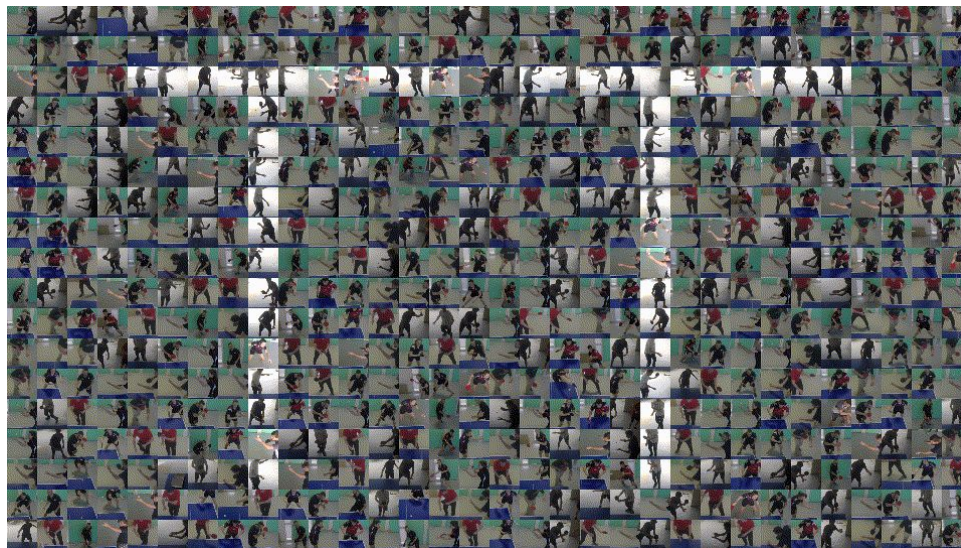
- 241 videos at 25, 30 and 120 fps
- 4 209 annotations
- 3 428 strokes over 20 classes



Acquisition

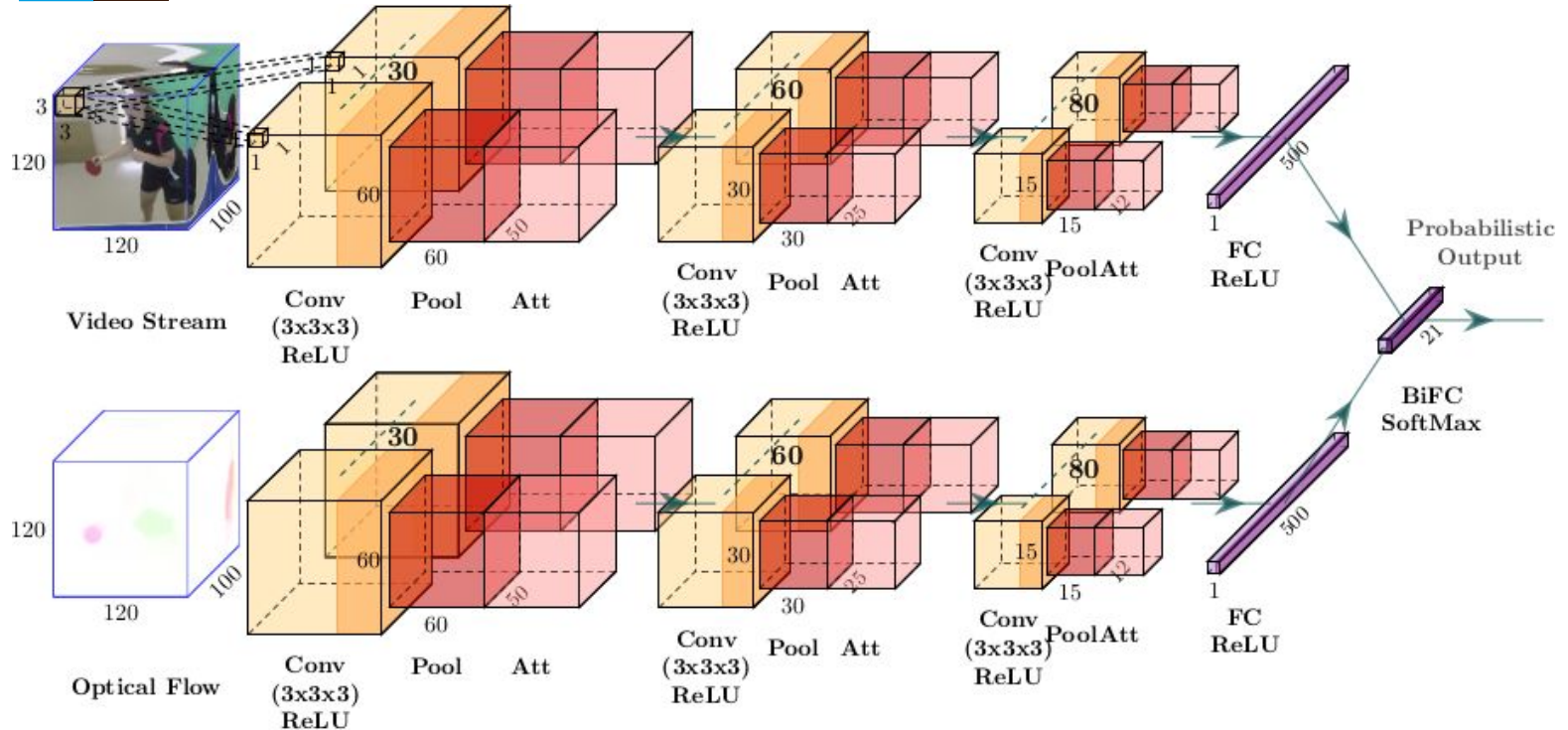


Annotation platform



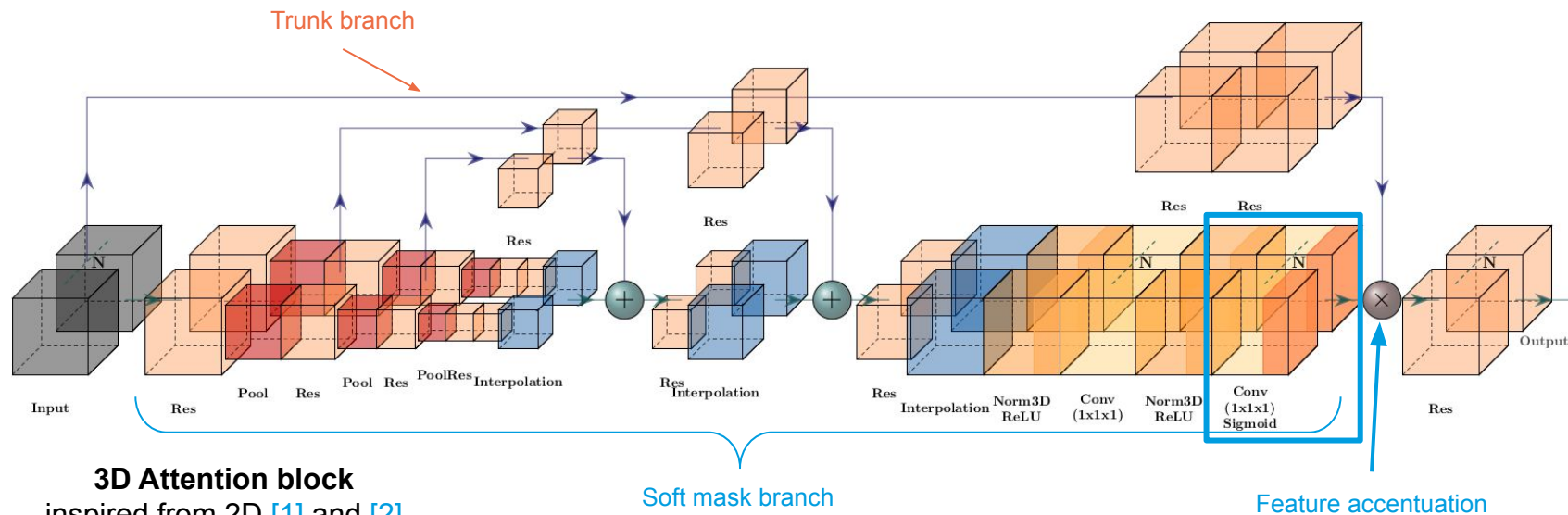
Samples TTStroke-21

Improve with Attention Mechanism



T-STCNN with attention blocks

The 3D Attention Block



3D Attention block
inspired from 2D [1] and [2]

- [1] F. Wang, M. Jiang, C. Qian, S. Yang, C. Li, H. Zhang, X. Wang, and X. Tang, "Residual attention network for image classification", CVPR 2017, pp. 6450–6458.
- [2] K. He, X. Zhang, S. Ren, and J. Sun, "Deep residual learning for image recognition", CoRR, vol. abs/1512.03385, 2015.

Results - Features accentuated



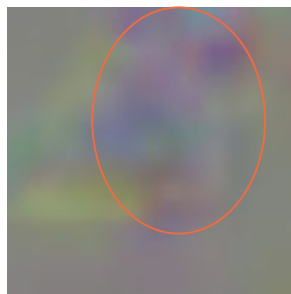
Soft mask branch
output of:

table
border



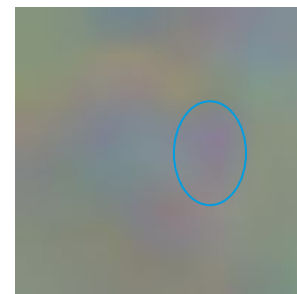
Attention block 1

player's
body
parts



Attention block 2

ball &
racket



Attention block 3

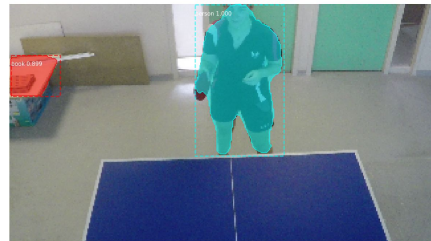
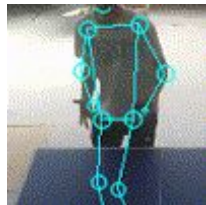
Results - Classification accuracy (in %)

Models	Train	Validation	Test
RGB-I3D [1]	98	72.6	69.8
RGB-STCNN [2]	98.6	87	76.7
RGB-STCNN with Attention	96.9	88.3	85.6 93.2
Flow-I3D [1]	98.9	73.5	73.3
Flow-STCNN [2]	88.5	73.5	74.1 85.6
Flow-STCNN with Attention	96.4	83.5	79.7
Two Stream-I3D [1]	99.2	76.2	75.9
Twin-STCNN [2]	99	86.1	81.9
Twin-STCNN with Attention	97.3	87.8	87.3 95.8

Future work and perspectives

→ Ongoing work:

- › Attention mechanism observation during the training phase
- › Pose [1] for classification
- › UNet for segmentation



→ Later in 2021:

- › MediaEval 2021
- › Pose + Depth [2] for stroke movement modeling with the aim to give a qualitative feedback to the players



[1] Newell Alejandro, Yang Kaiyu, Deng Jia: Stacked Hourglass Networks for Human Pose Estimation. ECCV 2016.

[2] Ramamonjisoa Michaël, Lepetit Vincent: SharpNet: Fast and Accurate Recovery of Occluding Contours in Monocular Depth Estimation. ICCV Workshops 2019.

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