

# 3D human pose estimation in video with temporal convolutions and semi-supervised training

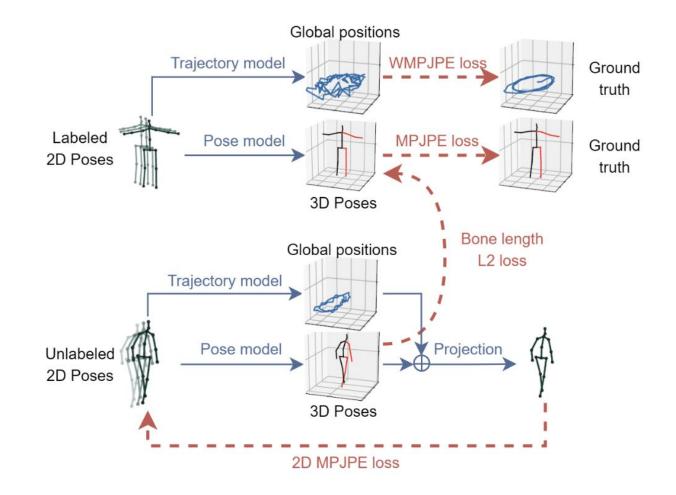
In CVPR 2019

Facebook Al Research

2024. 04 17.

## Contents

- 1. Purpose
- 2. background
  - 2-1. TCN
  - 2-2. casual network
  - 2-3. dilated TCN
- 3. Total Flow
  - 3-1. Network
  - 3-2. semi-supervised
- 4. evaluation

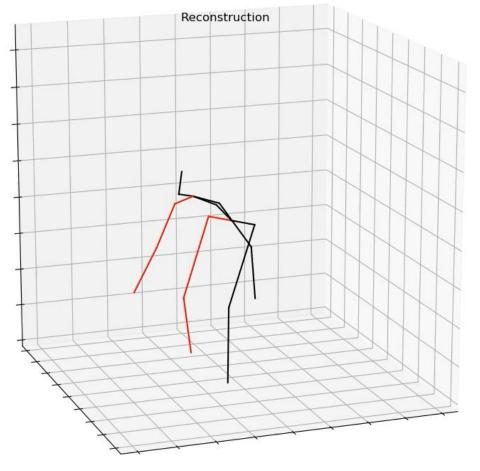


2024. 04. 17.

# Purpose

## 3D human pose estimation in video

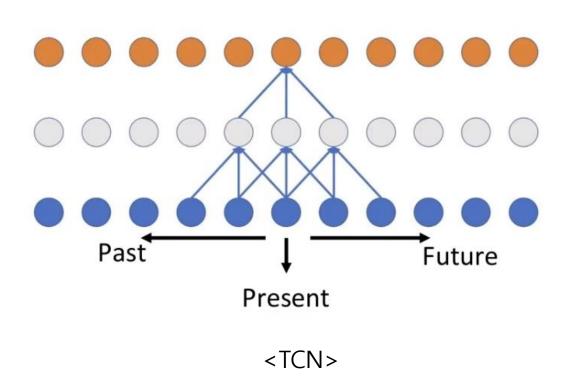


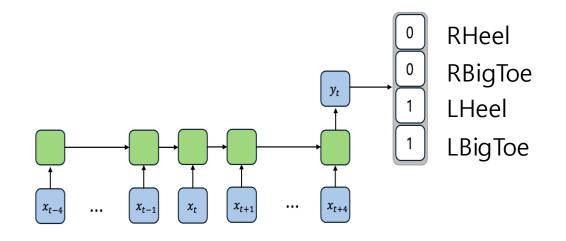


# background – TCN

## Temporal Convolution Network

# Standard Convolution



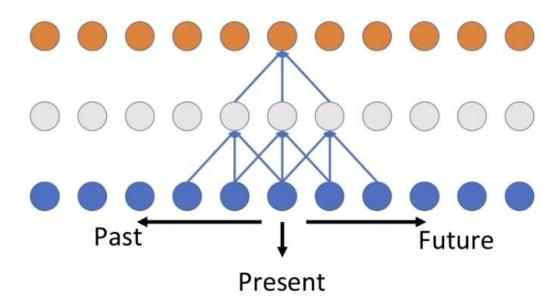


<RNN>

# background – TCN

## Temporal Convolution Network

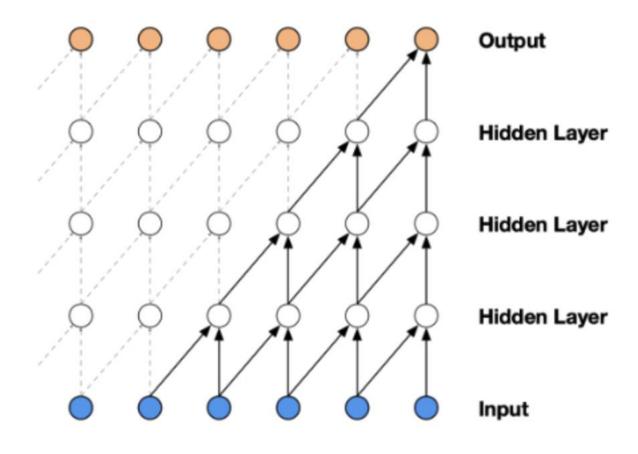
#### Standard Convolution



## <Advantages>

- 1. Parallelism
- 2. Flexible receptive field size
- 3. Stable gradients(exploding / vanishing gradient problem along the time axis)

# background – Causal TCN

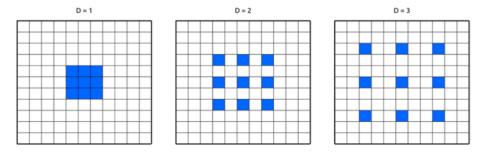


input sequence :  $x_0, ..., x_T$ output sequence :  $y_0, ..., y_T$ 

 $y_t < -input(x_i, ..., x_t)$ 

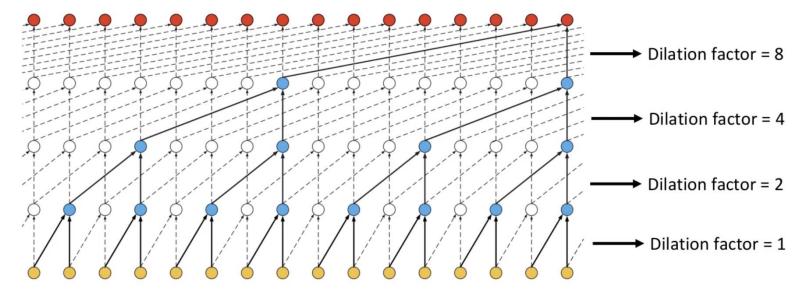
# background – Dilated TCN

#### **Dilated Convolution**



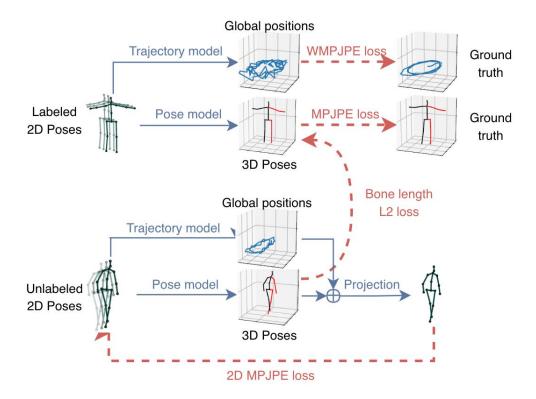
<Advantages>
An effective way to make the Receptive
Field large -> increase in computation x

#### Dilated TCN



## **Total Flow**

### temporal convolutions and semi-supervised training



<data>

1. labeled 2D Poses <-> mocap data

8

2. Un-labeled 2D Poses

Bone length L2 loss 
$$=\sum_{j=1}^{J}\left(\|\mathbf{b}_{j}^{ ext{pred}}-\mathbf{b}_{j}^{ ext{gt}}\|
ight)^{2}$$

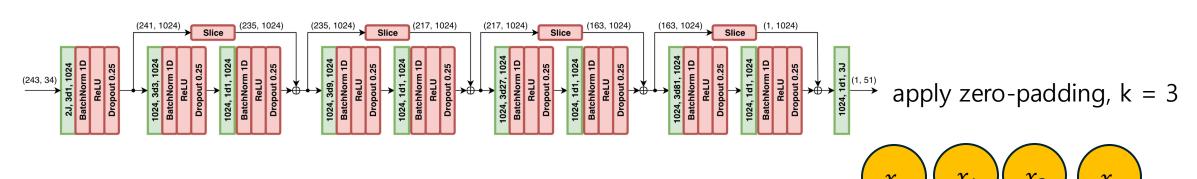
$$ext{MPJPE} = rac{1}{N} \sum_{i=1}^{N} \| \mathbf{p}_i^{ ext{pred}} - \mathbf{p}_i^{ ext{gt}} \|$$

WMPJPE = 
$$\frac{1}{\mathbf{y}_z} \| f(\mathbf{x}) - \mathbf{y} \|$$

 $y_z$ : gt depth in camera space

<each network>
The two networks have the same architecture but do not share any weights

## Network – Temporal dilated convolutional model

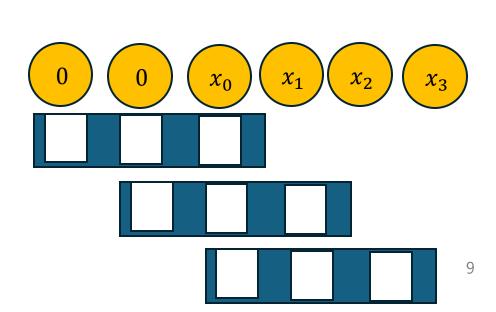


input : (x, y) x joint x Timesteps

output: 3D poses for all frame

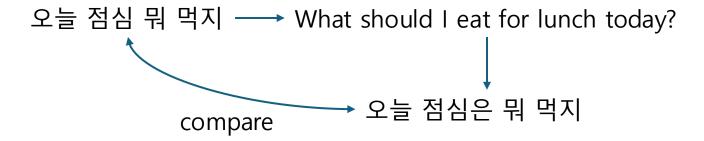
hyper-parameter) kernel size : k

dilation factor: D



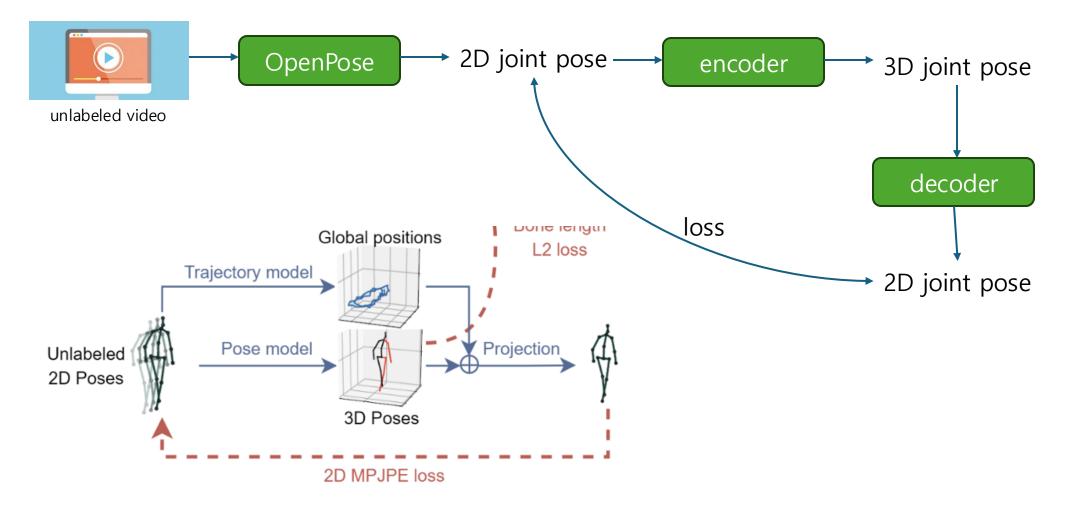
# Network – Semi-supervised approach

main idea : cycle consistency



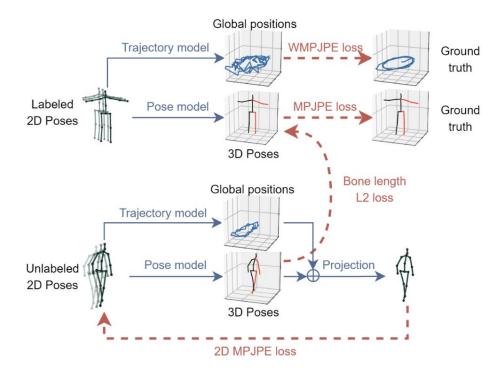
Cycle consistency) maintaining its consistency when returning to its original state through a cyclical process

# Network – Semi-supervised approach



11

# Network – Trajectory



## Trajectory model: For perspective projection

- 2D pose on the screen depends on...
- 1. the global position of the human root joint(trajectory)
- 2. 3D pose

#### <reason>

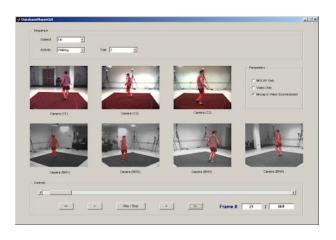
1. without global position, the subject would always be reprojected at the center of the screen with a fixed scale.

## Evaluation – Datasets and Evaluation



#### Human3.6M

- 3.6 million video frames for 11 subjects, of which seven are annotated with 3D poses



#### HumanEva-I

much smaller dataset, with three subjects recorded from three camera views

## Evaluation – Datasets and Evaluation

	Dir.	Disc.	Eat	Greet	Phone	Photo	Pose	Purch.	Sit	SitD.	Smoke	Wait	WalkD.	Walk	WalkT.	Avg
Pavlakos <i>et al.</i> [41] CVPR'17 (*)	67.4	71.9	66.7	69.1	72.0	77.0	65.0	68.3	83.7	96.5	71.7	65.8	74.9	59.1	63.2	71.9
Tekin et al. [52] ICCV'17	54.2	61.4	60.2	61.2	79.4	78.3	63.1	81.6	70.1	107.3	69.3	70.3	74.3	51.8	63.2	69.7
Martinez <i>et al</i> . [34] ICCV'17 (*)	51.8	56.2	58.1	59.0	69.5	78.4	55.2	58.1	74.0	94.6	62.3	59.1	65.1	49.5	52.4	62.9
Sun et al. [50] ICCV'17 (+)	52.8	54.8	54.2	54.3	61.8	67.2	53.1	53.6	71.7	86.7	61.5	53.4	61.6	47.1	53.4	59.1
Fang <i>et al</i> . [10] AAAI'18	50.1	54.3	57.0	57.1	66.6	73.3	53.4	55.7	72.8	88.6	60.3	57.7	62.7	47.5	50.6	60.4
Pavlakos <i>et al</i> . [40] CVPR'18 (+)	48.5	54.4	54.4	52.0	59.4	65.3	49.9	52.9	65.8	71.1	56.6	52.9	60.9	44.7	47.8	56.2
Yang et al. [56] CVPR'18 (+)	51.5	58.9	50.4	57.0	62.1	65.4	49.8	52.7	69.2	85.2	57.4	58.4	43.6	60.1	47.7	58.6
Luvizon <i>et al.</i> [33] CVPR'18 $(*)(+)$	49.2	51.6	47.6	50.5	51.8	60.3	48.5	51.7	61.5	70.9	53.7	48.9	57.9	44.4	48.9	53.2
Hossain & Little [16] ECCV'18 (†)(*)	48.4	50.7	57.2	55.2	63.1	72.6	53.0	51.7	66.1	80.9	59.0	57.3	62.4	46.6	49.6	58.3
Lee et al. [27] ECCV'18 (†)(*)	40.2	49.2	47.8	52.6	50.1	75.0	50.2	43.0	55.8	73.9	54.1	55.6	58.2	43.3	43.3	52.8
Ours, single-frame	47.1	50.6	49.0	51.8	53.6	61.4	49.4	47.4	59.3	67.4	52.4	49.5	55.3	39.5	42.7	51.8
Ours, 243 frames, causal conv. (†)	45.9	48.5	44.3	47.8	51.9	57.8	46.2	45.6	59.9	68.5	50.6	46.4	51.0	34.5	35.4	49.0
Ours, 243 frames, full conv. (†)	45.2	46.7	<u>43.3</u>	45.6	48.1	<b>55.1</b>	<u>44.6</u>	<u>44.3</u>	57.3	65.8	47.1	44.0	49.0	<u>32.8</u>	33.9	46.8
Ours, 243 frames, full conv. $(\dagger)(*)$	45.1	<u>47.4</u>	42.0	<u>46.0</u>	<u>49.1</u>	<u>56.7</u>	44.5	44.4	<u>57.2</u>	<u>66.1</u>	<u>47.5</u>	<u>44.8</u>	<u>49.2</u>	32.6	<u>34.0</u>	47.1

(a) Protocol 1: reconstruction error (MPJPE).

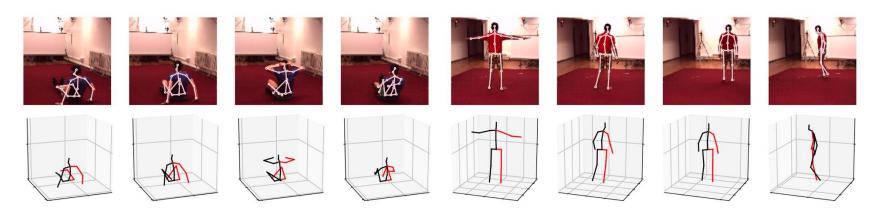


Figure 4: Qualitative results for two videos. **Top:** video frames with 2D pose overlay. **Bottom:** 3D reconstruction.