

# Regularizing Long Short Term Memory with 3D Human-Skeleton Sequences



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#### Problem: Large-scale action recognition







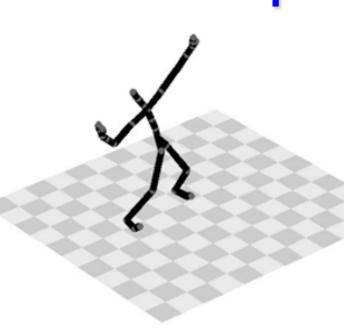






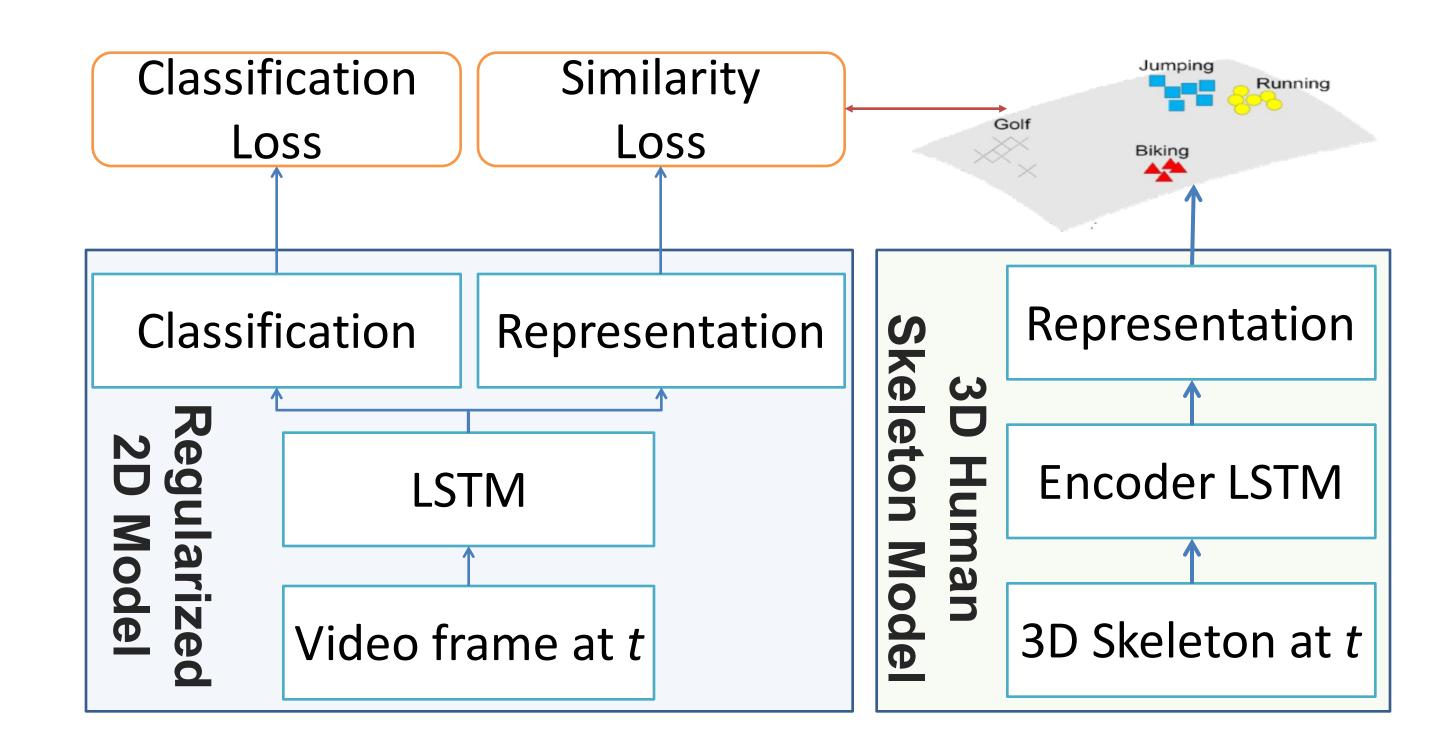
# **Key Idea:** Additionally train with 3D human-skeleton sequences



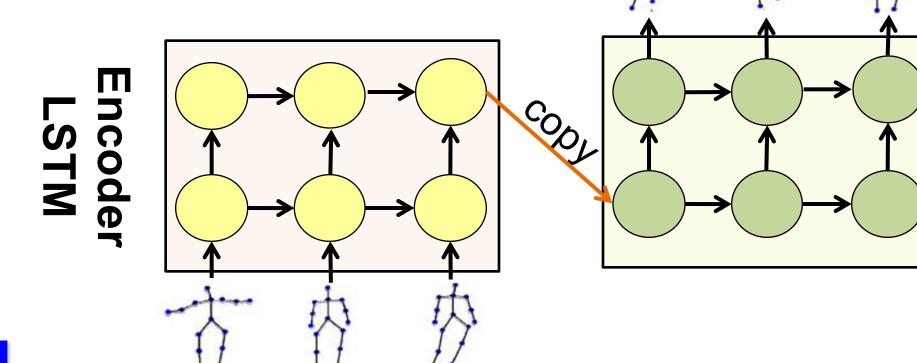


- Instead of typical strategies:
- Making representations more sophisticated (e.g. deeper)
- Increasing supervision (e.g. more videos in training)
- > 3D data is not available at test time
- > Poor coverage of actions in 3D sequences

### Regularized LSTM (R-LSTM)



## **Encoder LSTM**



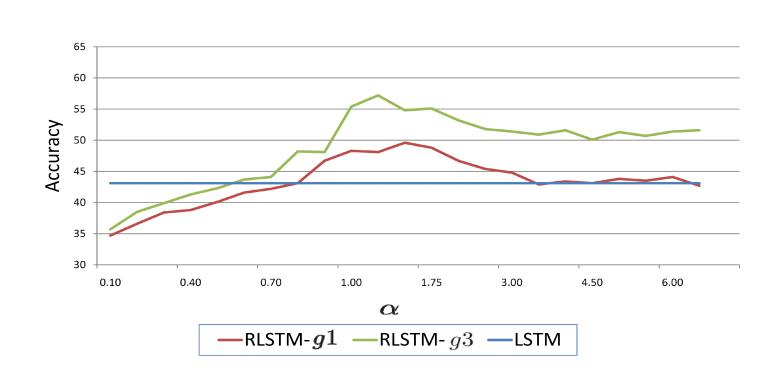
## **Training**

- Constrains
- Class aware:  $\frac{1}{n_1} \sum_{r_s} |r_s r_v| \frac{1}{n_2} \sum_{r_{s'}} |r_{s'} r_v| \le 0$
- Class independent:  $\frac{1}{n}\sum_{r_s}|r_s-r_v| \leq \alpha$
- > Hybrid backpropagation through time

#### Accuracy in [%] on Sports-1M

Method	Hit@1	Hit@5
CNN	59.3	77.7
LSTM	71.3	89.9
Karpathy et al. '14	60.9	80.2
Ng et al. '15	72.1	90.6
Tran et al. '15	61.1	85.2
R-LSTM	75.9	91.7

Actions	Acc Change	
Running	3.7	
Badminton	1.5	
Track cycling	2.2	
Road cycling	1.3	
Down hill biking	0.7	
BMX	0.4	
Wind Surfing	-1.2	
Fishing	-1.0	
Land Surfing	-0.9	



# Accuracy in [%] on UCF101 & HMDB-51

Method	UCF- 101	HMDB- 51
Karpathy et al. '14	65.4	-
Srivastava et al. '15	75.8	44.1
Donahue et al. '14	71.12	-
Simonyan et al. '14	72.8	40.5
Zha et al. '15	79.34	-
Tran et al. '15	85.2	-
R-LSTM	86.9	55.3

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