

# ACTIVE: Activity Concept Transitions in Video Event Classification

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## Introduction

### ➤ Goal

- Classify high level events from unconstrained web videos

### ➤ Challenges

- Complex human-object interactions
- Diverse video quality (e.g. YouTube)
- Large scale dataset

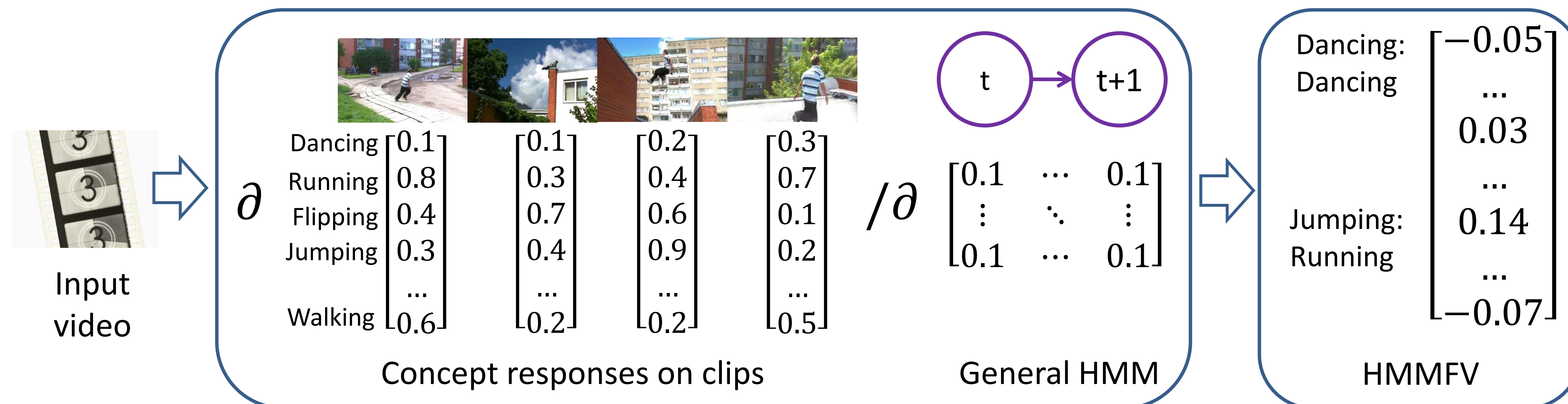
### ➤ Motivation

- Encoding activity concept transitions



kiss	0.08	-0.02	0.06
walk	-0.07	-0.03	0.01
dance	-0.02	0.05	0.14
	kiss	walk	dance

## Hidden Markov Model Fisher Vector (HMMFV)



### ➤ Video Representation: A sequence of activity concept responses

### ➤ Fisher Kernel: Partial derivatives of log-likelihood function

### ➤ HMMFV: Partial derivatives about HMM transition parameters

$$\frac{\partial}{\partial \tau_{i|j}} \log P(X|\theta, \tau) = \sum_t \left[ \frac{\xi_t(i, j)}{\tau_{i|j}} - \gamma_{t-1}(j) \right]$$

$$FV(i, j) \sim \sum_t \alpha_{t-1}(j) [\theta_{x_t|i} \beta_t(i) - \beta_{t-1}(j)]$$

$$\xi_t(i, j) = P(s_t = i, s_{t-1} = j | X, \theta, \tau)$$

$$\gamma_{t-1}(j) = P(s_{t-1} = j | X, \theta, \tau)$$

$$\alpha_t(i) = P(x_1, \dots, x_t, s_t = i | \theta, \tau)$$

$$\beta_t(i) = P(x_{t+1}, \dots, x_n | s_t = i, \theta, \tau)$$

## Experimental Setup

### ➤ Dataset

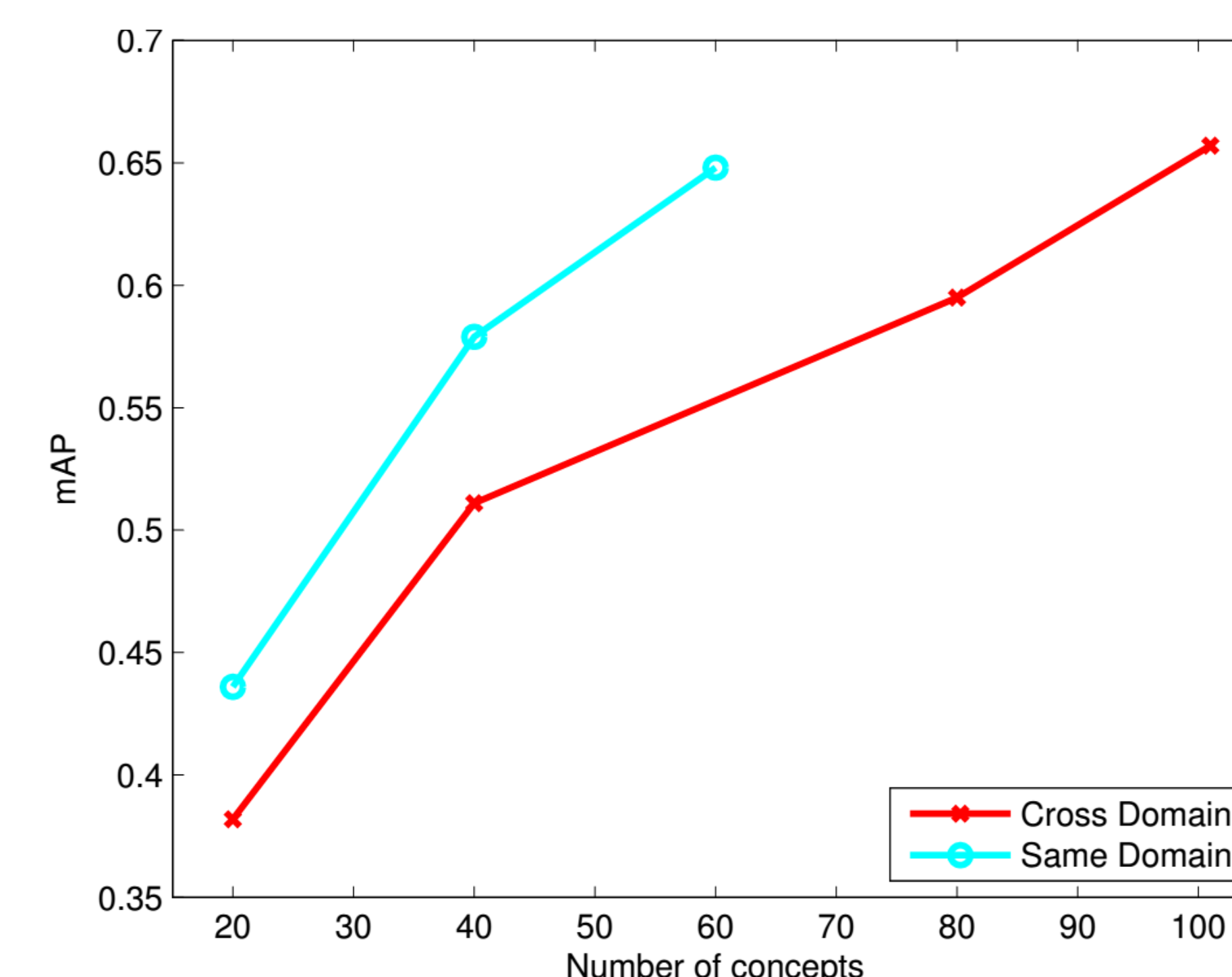
- TRECVID MED 2011 Event Kit
- 70% for training, 30% for testing

### ➤ Setup

- Gaussian kernel SVM, 5-fold cross validation

### ➤ Activity Concepts

- Same domain (Event Kit annotations [1])
- Cross domain (UCF 101)

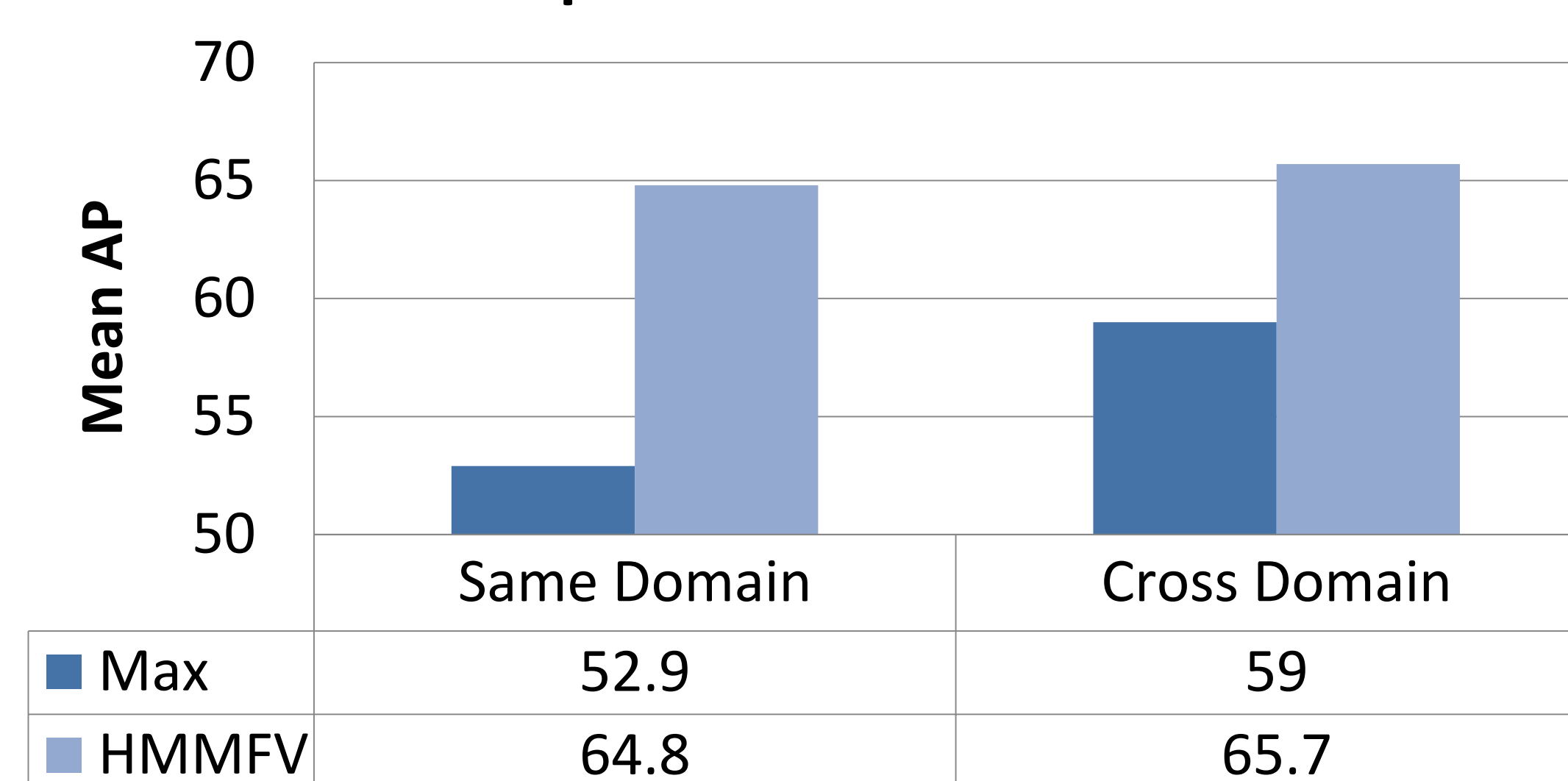


## Top Activity Concept Transitions Visualization

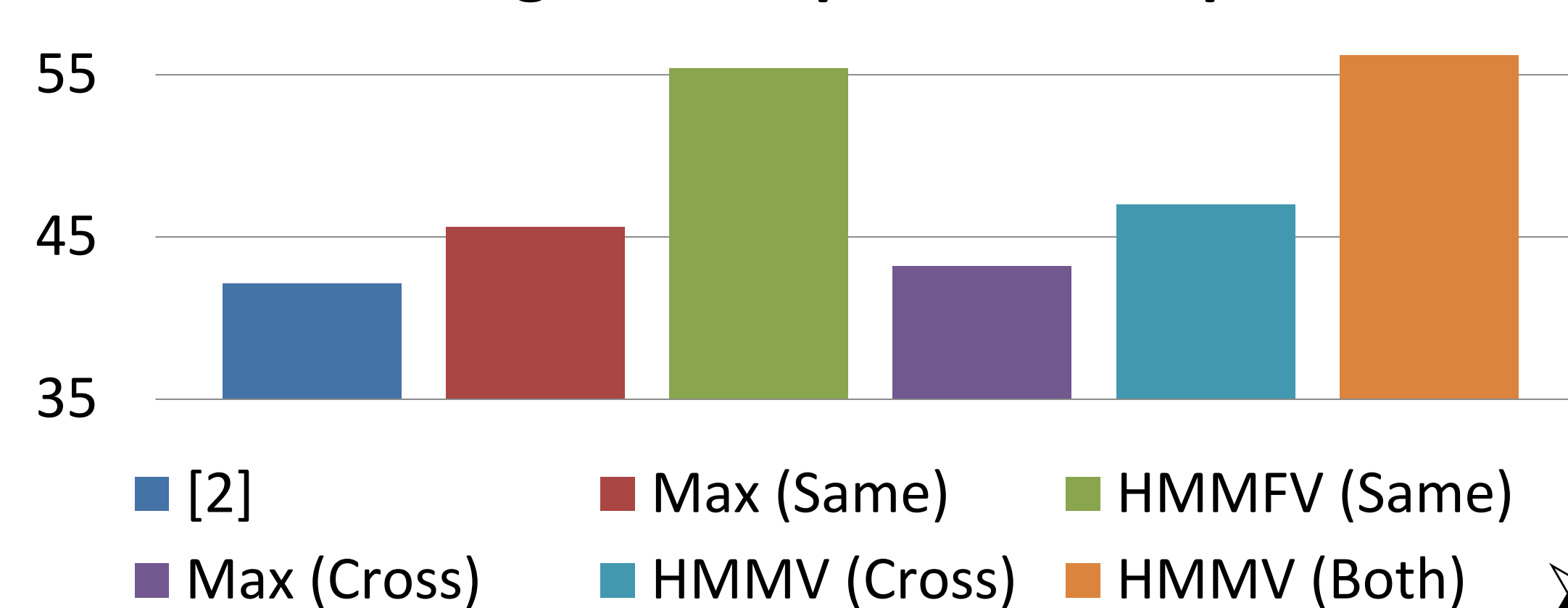


## Quantitative results

### Comparison with baseline



### Training with 10 positive samples



### Comparison with state-of-the-art

Event	Joint [1]	LL [2]	HMMFV	HMMFV+[2]
Board trick	75.7	81.3	<b>88.5</b>	88.2
Feed animal	<b>56.5</b>	36.3	46.8	46.1
Wedding	72.2	74.3	<b>79.6</b>	78.9
Land fish	67.5	<b>82.9</b>	77.1	81.1
Woodworking	<b>65.3</b>	49.6	60.4	62.3
Birthday party	78.2	73.6	81.1	<b>81.4</b>
Change tire	47.7	<b>54.1</b>	48.2	51.8
Flash mob	<b>91.9</b>	86.8	87.3	87.7
Vehicle unstuck	69.1	76.9	75.6	<b>77.2</b>
Groom animal	51.0	57.9	61.7	<b>63.4</b>
Make sandwich	41.9	51.5	47.6	<b>52.4</b>
Parade	72.4	72.0	77.0	<b>77.0</b>
Parkour	66.4	79.2	88.6	<b>89.0</b>
Repair appliance	<b>78.2</b>	66.1	61.9	63.4
Sewing	57.5	60.8	57.5	<b>62.1</b>
Mean AP	66.1	66.9	69.3	<b>70.8</b>

### ➤ Conclusion

- Coding temporal transitions of activities by HMMFV improves performance
- Activity concepts coded by HMMFV is desirable with limited training samples
- May be useful for video event recounting (description)

[1] H. Izadinia and M. Shah. Recognizing complex events using large margin joint low-level event model. In ECCV, 2012.  
[2] C. Sun and R. Nevatia. Large-scale Web Video Event Classification by use of Fisher Vectors. In WACV, 2013