

LEHIGH UNIVERSITY



## Introduction

Automatically recognizing human activities in videos is one of the core tasks in the field of computer vision. Compared to the single-person activity recognition task, group activity recognition requires a more robust scheme that can capture correlated individual actions in group activities.

#### **Common existing approaches:**

Step 1. Identify individual person in video frames. Step 2. Track and recognize individual actions. Step 3. Infer group activities.

#### **Biggest weakness:**

High computation time.

#### **Our Contributions:**

- 1. We propose a novel solution, namely SBGAR, for group activity recognition.
- 2. The proposed scheme is semantics-based. It can generate a semantic representation for each video frame.
- 3. Our solution yields **better performance** than stateof-the-art approaches.

### Intuition

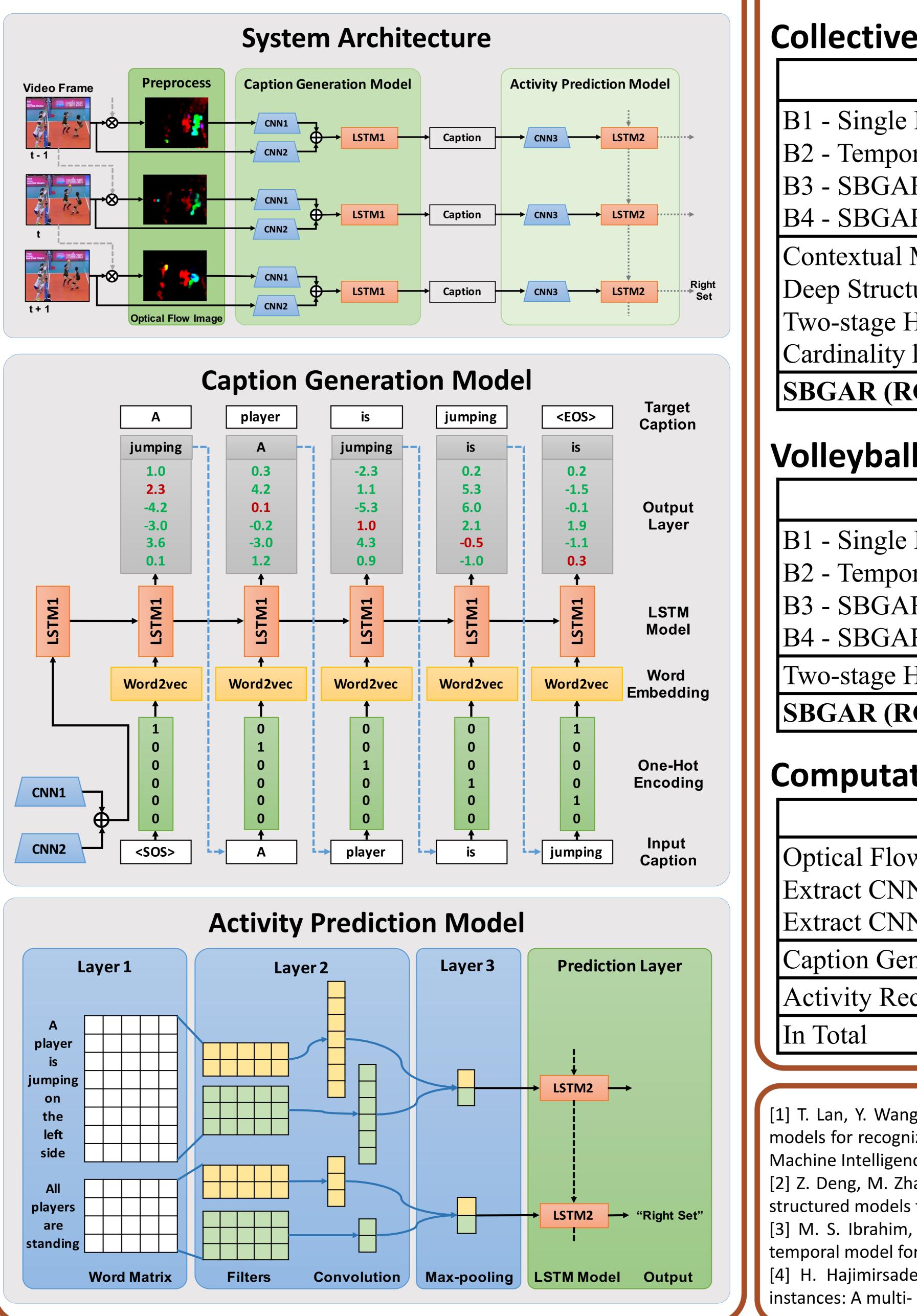
- In a Volleyball Game, given the following descriptions: Frame t-1: There is a player jumping on the right
- side, while others are **standing**.
- Frame t: There is one player spiking on the right side and three players **blocking** on the **left** side, while others are **standing**.
- Frame t+1: All players are standing.
- One can easily infer: **Right team is Spiking**.

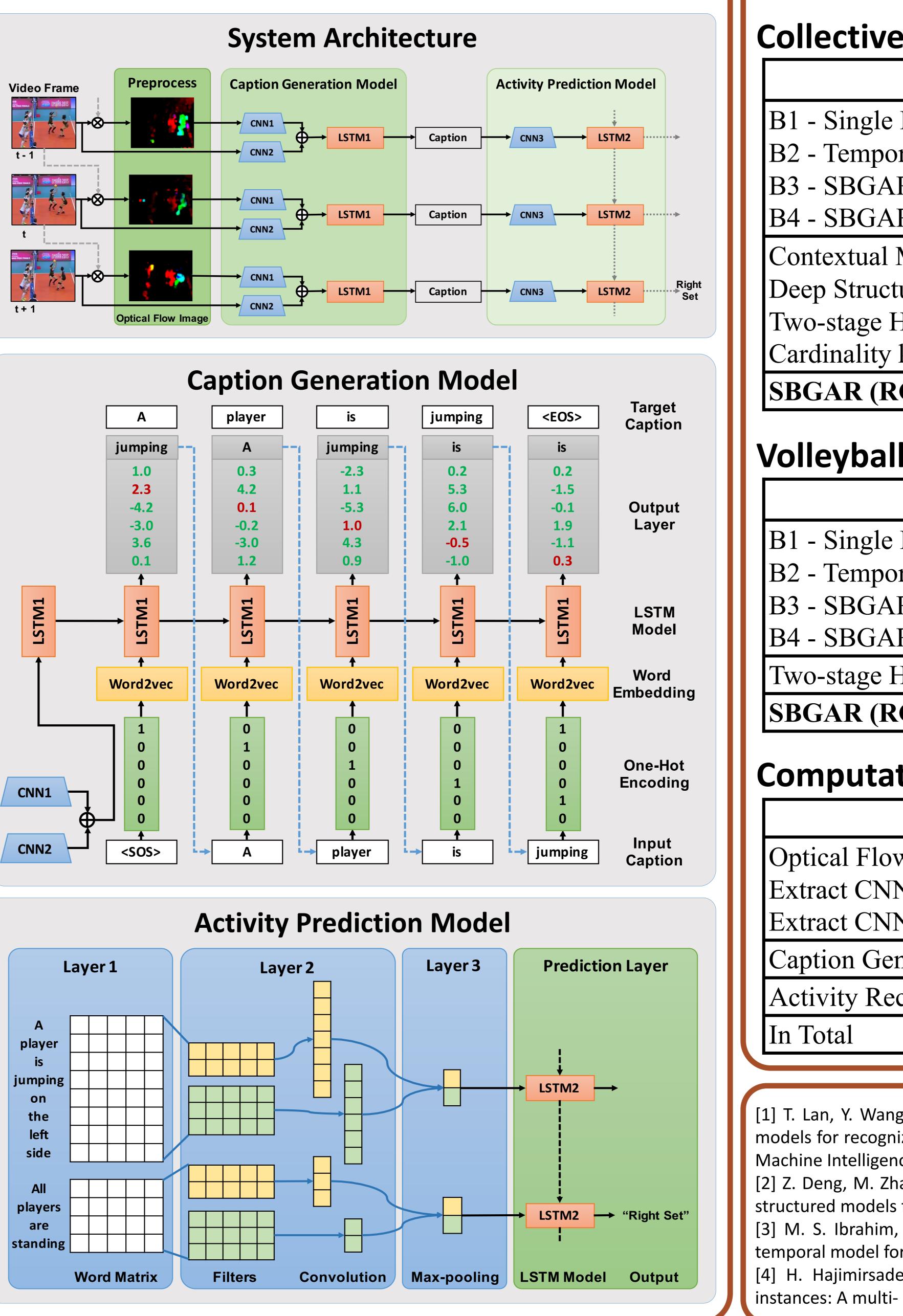
This work is partially supported by a NSF CSR grant 1217379 and a GPU donated by NVIDIA.

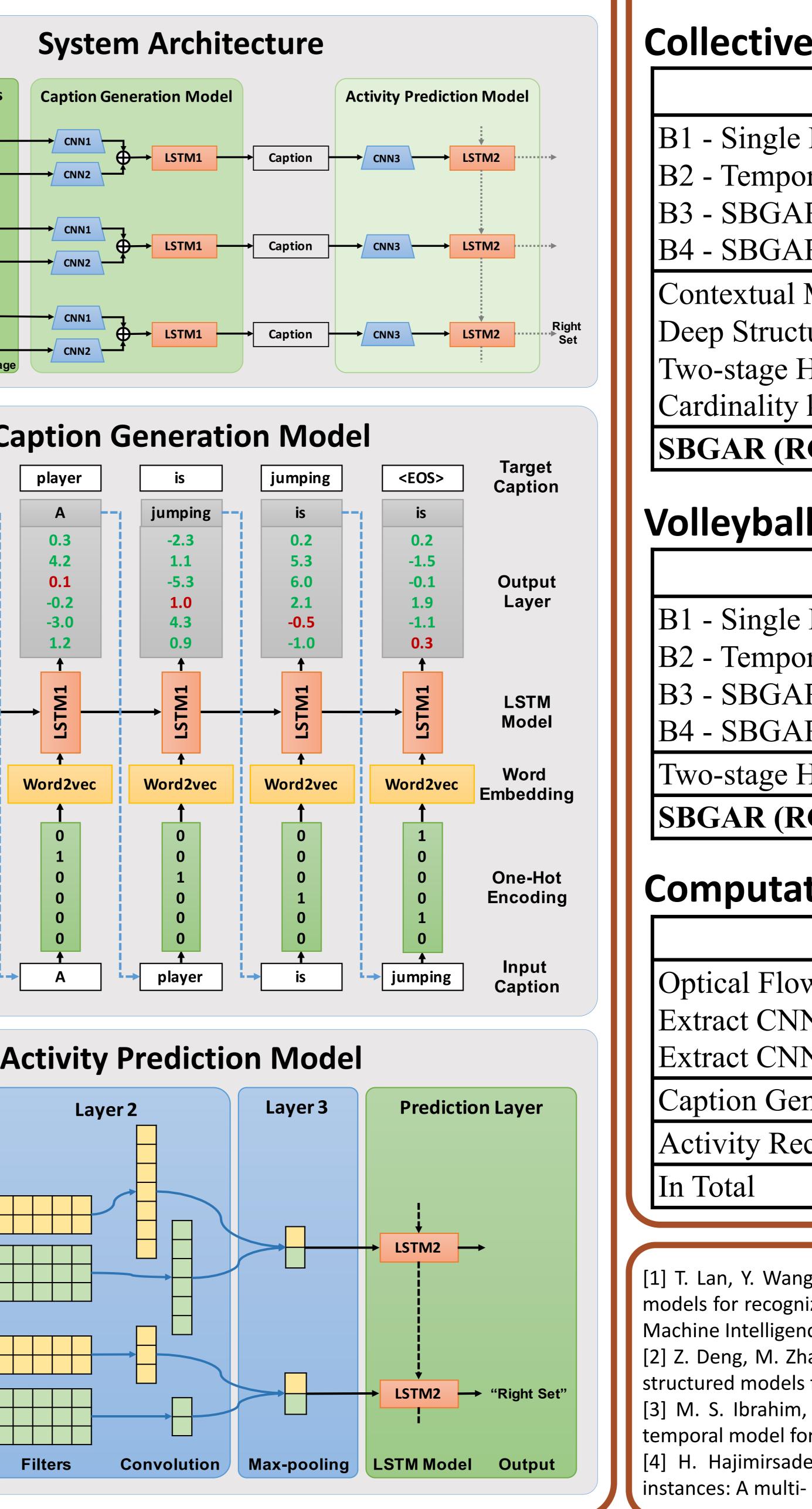
## **SBGAR: Semantics Based Group Activity Recognition** Xin Li, Mooi Choo Chuah **Department of Computer Science and Engineering**

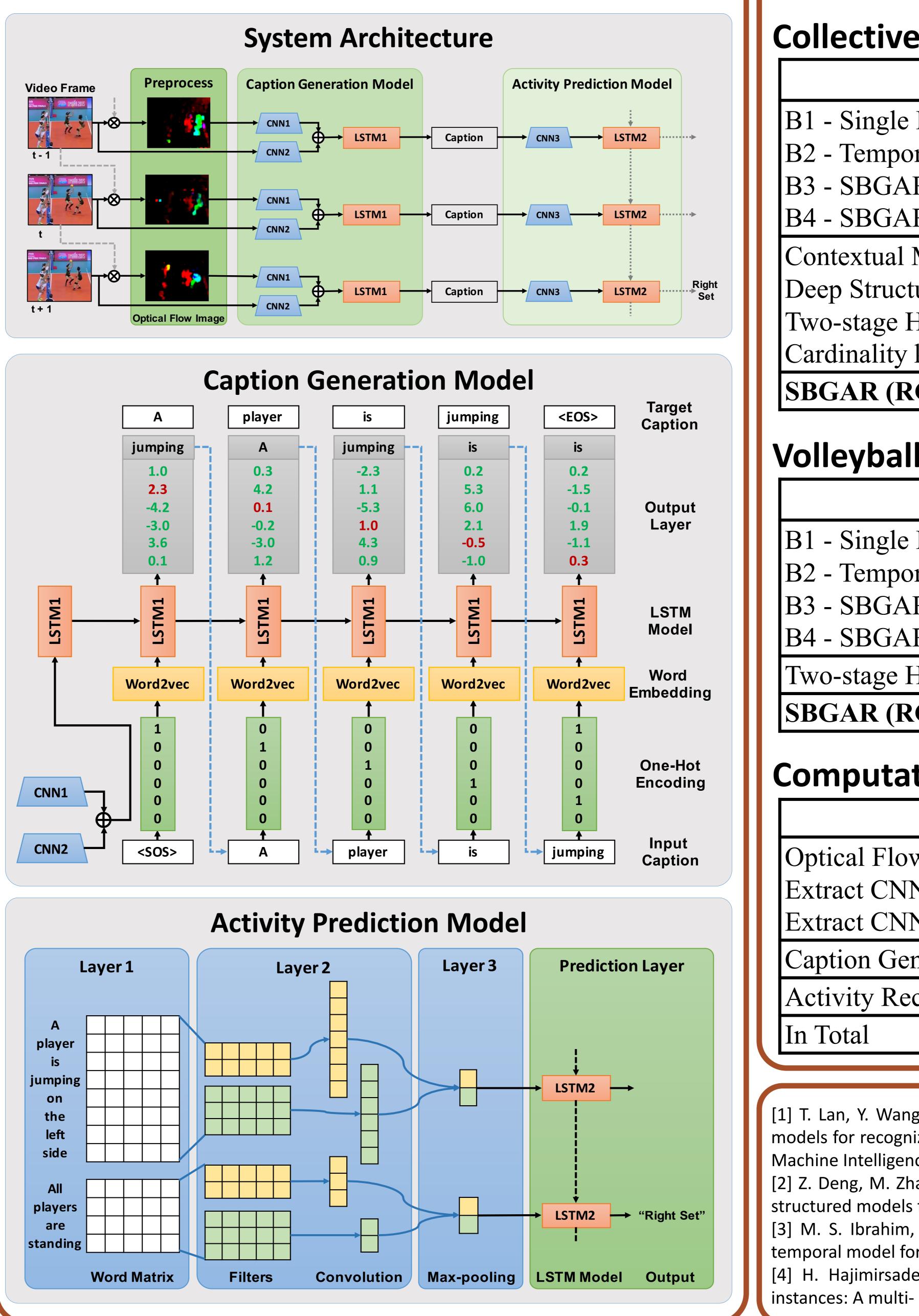
## **Proposed Solution**











## **Experimental Results**

#### **Collective Activity Dataset:** Metho

B1 - Single Frame Classif B2 - Temporal Model with B3 - SBGAR (RGB Fram B4 - SBGAR (Optical Flo

Contextual Model [1] Deep Structured Model [2 Two-stage Hierarchical M Cardinality kernel [4]

**SBGAR (RGB & Optica** 

## **Volleyball Dataset:**

#### Metho

- B1 Single Frame Classif
- B2 Temporal Model with
- B3 SBGAR (RGB Frame
- B4 SBGAR (Optical Flo

Two-stage Hierarchical M

**SBGAR (RGB & Optical** 

# **Computation Time:**

Process Optical Flow Image Extract CNN1 Feature (In Extract CNN2 Feature (In

Caption Generation

Activity Recognition (10

[1] T. Lan, Y. Wang, W. Yang, S. N. Robinovitch, and G. Mori, "discriminativeminative latent models for recognizing contextual group activities," IEEE Transactions on Pattern Analysis and Machine Intelligence, 2012.

[2] Z. Deng, M. Zhai, L. Chen, Y. Liu, S. Muralidha- ran, M. J. Roshtkhari, and G. Mori, "Deep structured models for group activity recognition," arXiv preprint arXiv:1506.04191, 2015. [3] M. S. Ibrahim, S. Muralidharan, Z. Deng, A. Vahdat, and G. Mori, "A hierarchical deep temporal model for group activity recognition." CVPR, 2016. [4] H. Hajimirsadeghi, W. Yan, A. Vahdat, and G. Mori, "Visual recognition by counting instances: A multi- instance cardinality potential kernel," CVPR, 2015.



ds	Accuracy (%)
fication	67.2
th Image Features	68.5
ne Only)	83.7
ow Image Only)	70.1
	79.1
2]	80.6
/Iodel [3]	81.5
	83.4
al Flow)	86.1

ds	Accuracy (%)
fication	41.9
th Image Features	44.3
ne Only)	38.7
ow Image Only)	54.3
/lodel [3]	51.1
al Flow)	66.9

<b>Computation Time (ms)</b>
22.19
27.78
27.78
28.63
2.15
108.53