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## An experimental study of real-time identification of construction workers' unsafe behaviors

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#### ABSTRACT

Construction workers' unsafe behavior is one of the main reasons leading to construction accidents. However, the existing management approach to unsafe behaviors, e.g. Behavior-Based Safety (BBS), relies primarily on manual observation and recording, which not only consumes lots of time and cost but impossibly cover a whole construction site or all workers. To solve this problem and improve safety performance, an image-skeleton-based parameterized method has been proposed in a previous research to real-time identifying construction workers' unsafe behaviors. A theoretical framework has been developed with a preliminary test, but still lacking a comprehensive experiment to verify its validity, particularly in the recognition of the types of unsafe behaviors. This will have a serious impact on the extensive application of the method in real construction sites. Based on the method, this research designs and carries out a series of experiments involving three types of unsafe behaviors to examine its feasibility and accuracy, and determines the value ranges of relevant key parameters. The results of the experiment demonstrate the feasibility and efficiency of the method, being able to identify and distinguish unsafe behaviors in real time, as well as its limitations. This research not only benefits the extensive application of the method in construction safety management, but improves the effectiveness and efficiency of the method by identifying relevant future research focuses. Therefore this paper contributes to the practice as well as the body of knowledge of construction safety management.

### 1. Introduction

Nearly 80% of construction accidents are caused by workers' unsafe behaviors [25]. It is necessary and important for project managers or safety managers to monitor workers' unsafe behaviors in construction sites. Behavior-Based Safety (BBS) is regarded as a promising approach to managing unsafe behaviors on site [25,26]. BBS needs the observation and identification of on-site unsafe behaviors and then makes a feedback to jobsite workers [2,4,6,19,41]. However, this mainly depends on manual observation and recording, which not only consumes lots of time and cost but also impossibly cover a whole construction site or all workers [10], therefore limiting its extensive application in the construction industry.

Automation technologies have been proposed to monitor construction workers' behaviors in construction sites so as to improve the efficiency and effectiveness of unsafe behavior management [9,34,35]. For example, wearable sensors and motion capture technologies have been adopted in existing research [24,26,36]. The former is sensitive and in

time, but affecting workers' normal working or operations [3,23,37,38]. The latter is usually to collect workers' behavior images with cameras and then identify unsafe behaviors by comparing the captured images with the images in unsafe behaviors databases [5,18,28,29,32,39]. The motion capture method requires no wearable devices, but not timely enough, as comparing the similarity of images is of great calculation amount [20,21,29,31,38,42].

Motion capture technologies have arisen the interest of the construction industry, being mainly used in the identification and biomechanical analysis of unsafe behaviors [15,33]. These methods usually contain four steps: 1) collecting sample data (joint sensor data, RGB-D image and stereo camera image), 2) reducing dimension, 3) extracting the features of motions from the sample data, and 4) identifying test motions by comparing their features with the features in Step 3 [7,11,12,14,22,27,32]. Most of these methods are post-analysis, thus not being applied to the real-time identification of unsafe behaviors. The main reason lies in the dimension reduction method. In order to reduce the redundant dimensions of images, previous studies used to

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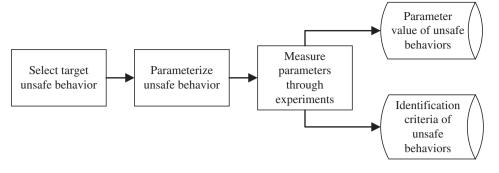
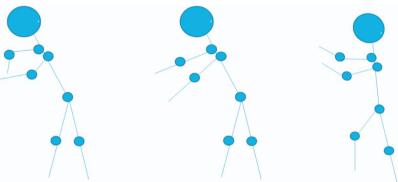


Fig. 1. Experiment process.



**Fig. 2.** The skeleton models of the leading postures of three unsafe behaviors.

(a) leaning on handrails (b) dumping from height (c) climbing ladders

Table 1
Key joint parameters in human skeleton mathematical model.

No.	Abbr.	Description	Skeleton model
θ1	LA	Angle between left arm and torso	† z
θ2	RA	Angle between right arm and torso	<b>6</b>
θ3	LE	Angle of left elbow	Y
θ4	RE	Angle of right elbow	· θ <sub>1</sub> <b>Ω</b>
θ5	LK	Angle of left knee	$\theta_3$
θ6	RK	Angle of right knee	
θ7	UB	Inclination of upper body	$\theta_4$
		(Angle between torso and axis z)	A h
θ8	LB	Inclination of lower body	• 1
		(Angle between line a-o and axis z.	<b>%</b> °
		Point a is the middle point of the connecting line between two feet;	
		Point o is located at the hip joint.)	$\theta_8$
h1	WH	The height difference between two wrists	$\Theta_{5}$
h2	AH	The height difference between two ankles	\ K
			$\setminus \mathbf{Q}/\theta_{\alpha}$
			y
			\\[ \lambda_1 \\ \lambda_1 \
			ha a X
			1"2



Fig. 3. Kinect for Windows 2.0.

adopt machine learning methods, such as Kernal PCA and DTW methods [13,14]. However, this leads to long latency time since it took a lot of time to calculate the value of these features, especially in the case of dynamic behavior rather than static posture [8,11,13,14,23]. To solve the above problems, a simplified image-based unsafe behavior identification method has been developed and tested with one behavior (climbing) of one worker by the authors. The preliminary study shows that it is feasible to real-time identify workers' unsafe behaviors through the simplification of behavior data and the development of a concise identification algorithm. However, various kinds of unsafe behaviors (e.g. dumping, leaning, etc.) are involved in a construction site with different characteristics, which mean different key parameters to describe these behaviors. Even for a same behavior with the same key parameters, different workers may present different value range for

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