

Chapter 64

Direct Perception and Action Decision for Unknown Object Grasping

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ABSTRACT

This paper discusses the direct perception of an unknown object and the action decision to grasp an unknown object using depth sensor for social robots. Conventional methods estimate the accurate physical parameters when a robot wants to grasp an unknown object. Therefore, we propose a perceptual system based on an invariant concept in ecological psychology, which perceives the information relevant to the action of the robot. Firstly, we proposed the plane detection based approach for perceiving an unknown object. In this paper, we propose the sensation of grasping which is expressed by using inertia tensor, and applied with fuzzy inference using the relation between principle moment of inertia. The sensation of grasping encourages the decision for the grasping action directly without inferring from physical value such as size, posture and shape. As experimental results, we show that the sensation of grasping expresses the relative position and posture between the robot and the object, and the embodiment of the robot arm by one parameter. And, we verify the validity of the action decision from the sensation of grasping.

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INTRODUCTION

Recently, various types of social robots have been developed for the next generation society of humankind (Beetz, 2011). For example, rehabilitation robots, education robots and therapy robots will be expected for the progress of society (Robins, 2005). Moreover, amusement robots, service robots and partner robots are expected in order to have a comfortable life. The social robots have to work not only in specific environments like factories, but also in general environments such as public facilities and homes (Mitsunaga, 2008). In a general environment, a social robot should take action flexibly in order to fulfill specific tasks, even in an unknown environment. A social robot needs human like perception and action decision in unknown environment. To discuss social robots, there are important consideration such as embodiment, social interaction, perception, interpretation, communication, experience, learning and so on (Fong, 2003). Especially, Wainer et al. insists that robots with physical bodies has a measurable effect on performance and perception of social interaction (Wainer, 2006). A perception of social robots cannot separate with the embodiment of a robot. Therefore, we focus on the discussion of the perception of a robot which has an embodiment to realize a social robot working in an unknown environment.

There are many previous researches that focus on the unknown object recognition method in a real environment (Bay, 2008; Comaniciu, 2001). To perceive an unknown object, F. Jurie et al. had proposed the real-time 3D template matching to detect objects in 3D space (Jurie, 2001). However, this method needs a minimal level of predefined knowledge which is given by the operator, or achieved by oneself. Therefore, a robot cannot perceive the environment and take suitable action immediately in an unexpected situation. All of the 3D template data makes nearly impossible to provide beforehand by human operator in real environment. On the other, there are segmentation methods for detecting an unknown object. Random sample consensus (RANSAC) is a popular segmentation method (Fischler, 1981). RANSAC and progressive RANSAC (PROSAC) (Chum, 2005) are installed in the Point Cloud Library (PCL) (Aldoma, 2012). These methods can detect same features from point cloud data using depth sensor. However, these methodologies require a significant volume of a computational cost to obtain accurate results.

Moreover, decision making of robotic action requires accurate object information such as position, orientation, size and so on. 3D object recognition methods are improved to recognize more accurate information (Dube, 2011; Alenya, 2014; Atanasov, 2014). However, accurate and complete information is difficult to measure by sensors in a real environment because of the noise and specification of hardware. Furthermore, a lot time is needed to make an accurate 3D map by 3D reconstruction method (Overby, 2004), a robot has to move around in order to acquire an unseen side. The robot is required to make a decision for an action in real time from poor and imperfect information.

On the other hand, people can perceive various types of necessary information for action easily, even in occlusion and imperfect information. Therefore, we have discussed the perception and action cycle for a robot (Masuta, 2008). Especially, we focus on the direct perception concept that is discussed in ecological psychology. The ecological approach discusses the perception and action cycle interacting with an environment (Gibson, 1979). Especially, importance is that perception is not inferred from sense organs, but only detection of invariant information (Turvey, 1999). Gibson insisted that the world contained “invariant” information. Invariant information was directly accessible to the perceptual systems of humans. Human just picks up this invariant information. This is called “direct perception” (Gibson, 1979). The invariant information extracted as the perceptual information can be obtained based on the coupling structure of the physical body with the environment. A robot will be able to perceive efficiently

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