

Graduate School of Creative Science and Engineering
Waseda University

博士論文概要

Doctoral Thesis Synopsis

論文題目

Thesis Theme

Acquisition and Extension of the Robot Body Model
Using Motor Babbling through Deep Learning

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Today, intelligent robots are expected to play an important role in supporting humans in their daily activities. Conventional robots such as industrial robots are primarily designed to perform designated tasks with high speed and precision and would require extra modifications to perform other tasks. In addition, these robots are typically applicable only in highly controlled environments that have no or negligible uncertainty and are free of human movement. Recently, there has been a growing interest in the development of robots working closely with humans. In real environments, where humans exist, it is essential that robots would be required to adapt to dynamic and uncertain environments and cope with new tasks and situations in a short duration. Robot with tool use skills would be particularly useful in human society, as this would enable the robot to expand its capabilities in performing tasks. To perform various tasks, robots should have complex structure such as large numbers of applicable sensors and degrees of freedom (DOFs). Designing robot models manually has become difficult because of the above reasons. To address this challenge, I developed a model in which the robot body and environment is based on experiences the robot acquires through physical (embodied) interactions as opposed to models in which the robot body and environment is pre-designed; that is, the robot learns sensorimotor relationships from experience using machine learning, particularly deep learning. The focus of this approach is embodiment. In this thesis, I focus specifically on the body model acquired from interaction with the environment through embodiment.

Enabling robots to act intelligently is usually referred to as artificial intelligence. Symbolic AI is based on the manipulation of symbols, an assumption defined as the physical symbol systems hypothesis. These approaches are only valid if all possible situations are described. However, it is difficult to describe all possible situations in the real world. In contrast, Brooks proposed a subsumption architecture in which the system and environment are connected using a connection between the sensor and motor without a description of the symbols. This approach is referred to as embodied AI. It is hypothesized that for complex intelligence, in particular for human-like intelligence, a human-like body is necessary for the robot to interact with the environment. To consider the complex relationship between the sensor and motor, machine-learning, particularly by using deep learning, has been proven effective. Deep learning has desirable characteristics, such as the ability to obtain individual features from data. The term deep learning refers to the use of deep neural networks (DNNs) and recurrent neural networks (RNNs). DNNs are feed-forward neural networks, whereas RNNs are recurrent connections.

When considering a robot body, body schema is an important concept used to explain the body model. Body schema involves mapping between the motor and the sensors located on the body and includes information about the size and shape of the body. In addition, body schema is not fixed after acquisition but is, instead, a plastic model. In particular, when humans are proficient in using a tool, they behave as if their bodies extend to the edge of the tool that they use. The phenomenon is referred to as tool-body assimilation. Owing to body schema, humans have the capability of recognizing the tool tip posture, position, and motion because their body schema is plastic with respect to their body and the tool. Hence, body schema can connect the body and tool easily; that is, it is possible to describe the tool using the model to extend the body schema. When a body schema is considered from an engineering perspective, it is a multi-modal association of motion and sensors. In my research, I added the concept of the forward/inverse model (In this study, the forward model predicts the next

state of the arm-tip position from the current state of the joint angles, and the inverse model predicts the next state of the joint angles from the current state of the arm-tip position). I refer to it as the body model, and in particular, I refer to it when considering the dynamics as body dynamics.

The objective of my thesis is to propose a machine learning framework that the robot acquires its body model from the experiences it gains through embodiment without pre-designed robot-environment model to adapt to dynamic and uncertain environments and cope with new tasks and situations in a short duration. Specifically, I focus on the acquisition and extension of the robot body model using motor babbling through deep learning to realize dynamic motion learning for flexible-joint robots and tool use with a tool-body assimilation model.

This thesis is organized into four chapters. Chapter 1 provides the background, research objective, related work, and overview of the proposed approach as an introduction of the current study about acquisition and extension of the robot body model using motor babbling through deep learning.

In chapter 2, I propose a learning strategy for robots with flexible joints having multi-DOFs in order to achieve dynamic motion tasks. In spite of there being several potential benefits of flexible-joint robots such as exploitation of intrinsic dynamics and passive adaptation to environmental changes with mechanical compliance, controlling such robots is challenging because of their increasingly complex dynamics. To achieve dynamic movements exploiting such benefits of flexible-joint dynamics, I introduce a two-phase learning framework of the body dynamics of the robot using a RNN motivated by a recent deep learning strategy consisting of pre-training and fine-tuning. This two-phase learning methodology comprises a pre-training phase with motor babbling and a fine-tuning phase with additional learning of dynamic motion tasks. In the pre-training phase with motor babbling, I consider active and passive exploratory motions in order to efficiently learn body dynamics. In the fine-tuning phase, the learned body dynamics are adjusted for specific tasks. I demonstrate the effectiveness of the proposed methodology in achieving dynamic tasks involving constrained movement requiring interactions with the environment on a simulated flexible joint robot model as well as in hardware experiments using a PR2 robot both of which have a seven DOF redundant arm. The results illustrate a reduction in the required number of training iterations for task learning as well as generalization capabilities for untrained situations with the learned body dynamics through motor babbling. In addition, I discuss the issues regarding the trade-off between task training iterations and the success rate of task execution. Furthermore, I discuss the small number of exploratory motor-babbling motions for body dynamics.

In chapter 3, I propose a tool-body assimilation model that considers grasping during motor babbling for using tools. A robot with tool-use skills could be useful in a human--robot symbiosis situation because this allows the robot to expand its task performing abilities. In the existing tool-use model, the body and tool models are separated. Therefore, the motions during tool use are expressed by each tool. To address these issues, I adopted tool-body assimilation. In a tool-body assimilation model, the tool model is expressed as a body with a tool. Therefore, the motions during tool use are expressed by the body model, and the robot effectively learns the tool functions. Almost all existing studies for robots to use tools require predetermined motions and tool features; the motion patterns are limited and the robots cannot use novel tools. Some of the other past studies fully search for all available parameters for novel tools, but this leads to massive amounts of calculations. Other past studies

approaches were mainly focused on obtaining the functions of the tools, and showed the robot starting its motions with a tool pre-attached to the robot. This implies that the robot would not be able to decide whether and where to grasp the tool. In real life environments, robots would need to consider the possibilities of tool-grasping positions, then grasp the tool. To address these issues, the robot performs motor babbling by grasping and nongrasping the tools to learn both tool functions and the robot's body model. In addition, the robot grasps various parts of the tools to learn the different tool functions from the different grasping positions. These rich motion experiences are learned using deep learning. Tool features were self-organized in parametric bias, modulating the body model according to the tool in use. Finally, I designed a neural network for the robot to generate motion only from the target image. To evaluate the model, I have the robot manipulate an object task without any tools or with several tools of different shapes. I have the robot generate motions after showing the initial and target states by deciding whether and where to grasp the tool. Therefore, the robot is capable of generating the correct motion and grasping decision when the initial and target states are provided to the robot. In Chapter 4, the achievements of a series of both numerical and robot experiments of acquisition and extension of the robot body model. Finally, reviews on the remaining research topics and future directions conclude this thesis.

早稲田大学 博士（工学） 学位申請 研究業績書

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種 類 別 (By Type)	題名、 発表・発行掲載誌名、 発表・発行年月、 連名者 (申請者含む) (theme, journal name, date & year of publication, name of authors inc. yourself)
論文 ○	<p>Tool-body Assimilation Model Considering Grasping Motion through Deep Learning, Robotics and Autonomous Systems, Volume. 91, pp. 115-127, 2017, Kuniyuki Takahashi, Kitae Kim, Tetsuya Ogata, Shigeki Sugano</p> <p>Body Model Transition by Tool Grasping During Motor Babbling using Deep Learning and RNN, Lecture Notes in Computer Science (The 25th International Conference on Artificial Neural Networks (ICANN 2016)), pp. 166-174, September, 2016, Kuniyuki Takahashi, Hadi Tjandra, Tetsuya Ogata, Shigeki Sugano</p> <p>Efficient Motor Babbling Using Variance Predictions from a Recurrent Neural Network, Lecture Notes in Computer Science (22nd International Conference on Neural Information Processing (ICONIP2015)), pp.26-33, November, 2015, Kuniyuki Takahashi, Kanata Suzuki, Tetsuya Ogata, Hadi Tjandra, Shigeki Sugano</p> <p>Effective Motion Learning for a Flexible-Joint Robot using Motor Babbling, 2015 IEEE/RSJ International Conference on Intelligent Robots and Systems(IROS2015), pp. 2723-2728, September, 2015, Kuniyuki Takahashi, Tetsuya Ogata, Hiroki Yamada, Hadi Tjandra, Shigeki Sugano</p> <p>Neural Network based Model for Visual-motor Integration Learning of Robot's Drawing Behavior: Association of a Drawing Motion from a Drawn Image, 2015 IEEE/RSJ International Conference on Intelligent Robots and System (IROS2015), pp. 2736-2741, September, 2015, Kazuma Sasaki, Hadi Tjandra, Kuniaki Noda, Kuniyuki Takahashi, Tetsuya Ogata</p> <p>Tool-body Assimilation Model Based on Body Babbling and Neuro-dynamical System, Mathematical Problems in Engineering, Article ID 837540, vol. 2015, 15 pages, 2015, Kuniyuki Takahashi, Tetsuya Ogata, Hadi Tjandra, Yuki Yamaguchi, Shigeki Sugano</p> <p>Tool-body Assimilation Model based on Body Babbling and a Neuro-dynamical System for Motion Generation, Lecture Notes in Computer Science (The 24th International Conference on Artificial Neural Networks (ICANN 2014)), pp. 363-370, September, 2014, Kuniyuki Takahashi, Tetsuya Ogata, Hadi Tjandra, Shingo Murata, Hiroaki Arie, Shigeki Sugano</p> <p>Tool-body Assimilation Model using a Neuro-dynamical System for Acquiring Representation of Tool Function and Motion, IEEE/ASME International Conference on Advanced Intelligent Mechatronics (AIM2014), ThAT3.6, pp. 1255 - 1260, July, 2014, Kuniyuki Takahashi, Tetsuya Ogata, Hadi Tjandra, Yuki Yamaguchi, Yuki Suga, Shigeki Sugano</p> <p>Handling and Grasp Control with Additional Grasping Point for Dexterous Manipulation of Cylindrical Tool, 2011 IEEE International Conference on Robotics and Biomimetics (ROBIO2011), pp. 733 - 738, December, 2011, Taisuke Sugaiwa, Kuniyuki Takahashi, Hiroyuki Kano, Hiroyasu Iwata, and Shigeki Sugano</p>

早稲田大学 博士（工学） 学位申請 研究業績書

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種 類 別 By Type	題名、 発表・発行掲載誌名、 発表・発行年月、 連名者（申請者含む）(theme, journal name, date & year of publication, name of authors inc. yourself)
講演 ○	<p>多自由度柔軟関節ロボットのためのモーターバブリングを用いた効率的な動的動作の学習, 第 17 回計測自動制御学会システム・インテグレーション部門講演会 (SI 2016), 1C4-1, 札幌, 2016 年 12 月, 高橋城志, 尾形哲也, 中西淳, Gordon Cheng, 菅野重樹</p> <p>Deep Learning による把持動作を考慮した道具身体化モデルの学習と動作生成, 第 17 回計測自動制御学会システム・インテグレーション部門講演会 (SI 2016), 札幌, 2016 年 12 月, 金紀泰, 高橋城志, 尾形哲也, 菅野重樹</p> <p>Deep Learning によるロボット道具使用のための把持を考慮した道具身体化モデル, 第 34 回日本ロボット学会 学術講演会 (RSJ2016), 3Y3-02, 山形, 2016 年 9 月, 金紀泰, 高橋城志, 尾形哲也, 菅野重樹</p> <p>Wizard of Oz と深層学習によるロボットの柔軟物折り畳み作業, 第 34 回日本ロボット学会 学術講演会 (RSJ2016), 2G2-05, 山形, 2016 年 9 月, 陽品駒, 佐々木一磨, 鈴木彼方, 加瀬敬唯, 高橋城志, 菅野重樹, 尾形哲也</p> <p>ディープニューラルネットワークとリカレントニューラルネットワークによる把持位置を考慮した道具使用モデル, 日本機械学会ロボティクス・メカトロニクス講演会 (ROBOMECH2016), 2P1-12b1, 横浜, 2016 年 6 月, 金紀泰, 高橋城志, Hadi Tjandra, 尾形哲也, 菅野重樹</p> <p>道具機能表現を分類するための道具身体化モデルの転移学習, 日本機械学会ロボティクス・メカトロニクス講演会 (ROBOMECH2016), 2P1-12b4, 横浜, 2016 年 6 月, 山田浩貴, 下栗逸爾, 高橋城志, Hadi Tjandra, 菅野重樹, 尾形哲也</p> <p>CNN による二次元物体画像から実ロボットでの把持動作生成, 日本機械学会ロボティクス・メカトロニクス講演会 (ROBOMECH2016), 2P1-12b7, 横浜, 2016 年 6 月, 鈴木彼方, 新古真純, 陽品駒, 高橋城志, 菅野重樹, 尾形哲也</p> <p>モーターバブリングと神経回路モデルを用いた道具身体化～複数道具把持動作による身体遷移過程の学習と生成～, 情報処理学会 第 78 回全国大会 (IPSI2016), 7L-02, 東京, 2016 年 3 月, 金紀泰, 高橋城志, Hadi Tjandra, 尾形哲也, 菅野重樹</p> <p>深層学習を用いた多自由度ロボットによる柔軟物の折り畳み動作生成, 情報処理学会 第 78 回全国大会 (IPSI2016), 1P-04, 東京, 2016 年 3 月, 鈴木彼方, 高橋城志, Gordon Cheng, 尾形哲也</p> <p>力学系学習木による効率的な学習のための階層性を利用した入力ベクトル決定法, 情報処理学会 第 78 回全国大会 (IPSI2016), 4P-02, 東京, 2016 年 3 月, 濱翔平, 平井諒, 高橋城志, 山田浩貴, 尾形哲也, 菅野重樹, 金天海</p> <p>ロボットを用いた身体モデル変調による道具身体化モデル～道具把持動作による身体遷移過程の学習と動作生成～, 第 16 回計測自動制御学会システム・インテグレーション部門講演会 (SI 2015), 2C3-1, 名古屋, 2015 年 12 月, Hadi Tjandra, 高橋城志, 尾形哲也, 菅野重樹</p> <p>Dynamic Motion Learning for a Flexible-Joint Robot using Active-Passive Motor Babbling, The 33rd Annual Conference of the Robotics Society of Japan (第 33 回日本ロボット学会 学術講演会 (RSJ2015)), 2G1-07, Tokyo, September, 2015, Kuniyuki Takahashi, Tetsuya Ogata, Shigeki Sugano, Gordon Cheng</p> <p>受動的な動作を含むモータバブリングによる柔軟関節を有するロボットの動作学習, 第 33 回日本ロボット学会 学術講演会 (RSJ2015), 1B3-07, 東京, 2015 年 9 月, 山田浩貴, 高橋城志, 尾形哲也, Hadi Tjandra, 菅野重樹</p> <p>RTC による深層学習モデルと柔軟関節ロボットの統合 ～道具身体化モデルの学習データ収集と動作実現～, 第 33 回日本ロボット学会 学術講演会 (RSJ2015), 3F3-06, 東京, 2015 年 9 月, 陽品駒, 高橋城志, 尾形哲也, 菅野重樹</p> <p>再帰神経回路モデルによる分散予測を用いた柔軟関節ロボットの身体ダイナミクスの探索, 日本機械学会ロボティクス・メカトロニクス講演会 (ROBOMECH2015), 2P1-S06, 京都, 2015 年 5 月, 鈴木彼方, 高橋城志, Hadi Tjandra, 村田真悟, 菅野重樹, 尾形哲也</p> <p>身体バブリングによる事前学習を用いた柔軟関節ロボットの効果的な動作学習, 情報処理学会 第 77 回全国大会 (IPSI2015), 5T-08, 京都, 2015 年 3 月, 山田浩貴, 高橋城志, 尾形哲也, Hadi Tjandra, 菅野重樹</p>

早稲田大学 博士（工学） 学位申請 研究業績書

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