

Early Pragmatic Language Development for an Infant Robot

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The Problem: How can a developing creature acquire natural language in a meaningful way? At an early stage of a human infant's life, certain verbal behaviors - both instinctive and learned - enable the infant to begin acquiring the caretakers' system of communication. The problem is to provide an infant robot with a mechanism for exhibiting such verbal behaviors, which would also have the function of acquiring words of a natural language.

Motivation: Motivation comes from the general direction of research on Kismet[1], a robot built to be an embodied developing agent situated in the human world, whose behavior follows its own internal agenda and reacts to the world and human presence in it. Like a human infant, Kismet was designed to elicit nurturing responses from humans, which would scaffold its development and learning. As the robot acquires more of the basics of a natural language, communication with humans becomes easier, and the range of possible concepts and tasks to learn increases. The approach taken here, based on the pragmatics of language acquisition, is motivated by a desire to produce a system that acquires words together with functional meanings.

Previous Work: We draw extensively on research in the fields of developmental psychology, phonetics and linguistics, and more recently, behavioral robotics and ethology. Some researchers [6, 7] approached the problem of acquisition of natural categories and labels by robots from the point of view of perceptual grounding. The robot analyzes the visual scene and the speech stream into segments, the best correlation between which will form a perceptual concept-label pair which is acquired by the robot. The project also utilizes the concepts of proto-linguistic behaviors [4] and referential mapping [5].

Approach: The emphasis of this project is on the pragmatic functionality of acquired concepts. The bases of the natural language that Kismet is developing are grounded not only in its sensory streams but also, primarily, in its "drive" to manipulate its environment in such a way as to achieve the satiation of its "desires". After Halliday [4], we identify a number of requisite behaviors in early linguistic development: request, calling, greeting, self-assertion, question, naming, and categorization. The development starts with a finite set of pre-programmed vocal responses to certain "emotional" states such as frustration. They are activated as responses to combinations of releasers, which signal the presence of visual and audio stimuli as well as internal state such as the currently active behavior. These proto-verbal responsive behaviors are arranged in an architecture inspired by the subsumption architecture [2] with the additions of randomness, state in the form of short-term and long-term memories, and timings (Fig.1), and are implemented using a script language Zac [3] designed for programming robotic control architectures. These responses establish an initial regulatory system of communication between the robot and the caretaker.

The next step up from such a closed formulaic and manipulative system is the actual acquisition of adult-like words and their appropriate use in the scenarios of naming and categorization. This stage of development is achieved through a reinforcement learning mechanism operating on a network of concepts. Nodes can be dynamically created and deleted and this updating of the network constitutes the learning in the system. Nodes are activated by recognizing the presence of specific releasers to which they are sensitive. Releasers here can be thought of as virtual sensors with particular receptive fields, or filters, which may be hierarchically organized.

Many nodes may be more or less active at any one time. What Kismet actually says is determined in a distributed way through the timings of node activations.

The robot learns the language on different planes. New reinforcement signals (e.g., labels such as "good" and "bad") are learned through correlation with pre-programmed ones. New meanings are created from combinations of releasers and internal virtual sensors. Labels for these are learned by a reinforcement algorithm. These meanings

