

Adaptive Prompting for Intelligent Wheelchairs

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6 **Abstract**

7 Cognitive impairments prevent older adults from using powered
8 wheelchairs due to safety concerns, thus reducing their mobility and
9 independence. An intelligent powered wheelchair is proposed to help
10 restore mobility, while ensuring safe navigation. Machine vision and
11 learning techniques are used to help prevent collisions with obstacles, as
12 well as provide navigation assistance through adaptive prompts.

13 14 **1 Introduction**

15 Several residents in long-term-care facilities lack the cognitive and/or visual abilities to
16 safely maneuver powered wheelchairs, and are thus not permitted to use them. This leads to
17 reduced mobility, and, in turn, depression and an increased dependence on caregivers. We
18 propose intelligent wheelchairs to enhance mobility and help improve the quality of life of
19 older adults with cognitive and/or visual impairment, while simultaneously reducing the
20 burden on caregivers.

21 Several intelligent wheelchairs capable of collision avoidance and path planning have been
22 developed recently [1], however these wheelchairs navigate autonomously, thus taking
23 control away from the user. On the other hand, wheelchairs that leave planning and
24 navigation to the user and only provide collision avoidance support are not appropriate for
25 users with cognitive impairment since they often lack planning abilities. We therefore
26 suggest a mixed-initiative control strategy that combines artificial intelligence of the
27 wheelchair with driver abilities to provide supportive, passive navigation assistance that
28 increases independence and ensures safety.

29 Other assistive systems designed for older adults include the Nursebot Project [2], the
30 Assisted Cognition Project [3], and an indoor wayfinding system [4]. Although these
31 systems use AI techniques for planning and/or reminding, they do not incorporate user
32 attitude modeling. A system for assisting persons with dementia during the handwashing task
33 is described in [5]. This system estimates the user's cognitive state using a Partially
34 Observable Markov Decision Process (POMDP) and issues audio prompts. Similar
35 techniques are used in the proposed wheelchair system to determine an appropriate
36 prompting strategy.

37 38 **2 Methods**

39 We have implemented vision-based anti-collision technology to prevent collisions and
40 encourage the user to drive around obstacles [6]. We are using existing mapping and path
41 planning techniques to determine the location of the wheelchair and compute optimal routes
42 to goal locations. We have also completed preliminary work in identifying place types

43 (kitchen, bathroom, etc.) based on visual cues [7]. Finally, we will provide audio prompts to
44 assist the user in navigating to the destination while accounting for his/her needs and
45 preferences, as well as obstacles in his/her path.

46 In this abstract, we present a prototype prompting system, which uses a POMDP to provide
47 adaptive navigation assistance. For our experiments, we use a sample floor plan of an indoor
48 environment. We then pick random start and goal locations and use a path planner, which
49 finds the optimal path as described in [8]. We use a simple interface to simulate wheelchair
50 motion using a keyboard. Noisy observations of wheelchair locations are read at regular time
51 intervals and used to determine the user's most recent behavior (following path, deviating,
52 turning, etc.). If the user is found to be deviating from the optimal route significantly, a new
53 route to the goal is computed and used in subsequent prompting. We use the POMDP to
54 compute the optimal system action (e.g. do nothing, issue prompt, call caregiver) at each
55 time step, based on the user's estimated current status (on-route, off-route, etc.) and his/her
56 cognitive state (ability to navigate independently to his/her desired location, and level of
57 responsiveness to prompts). Transition and reward functions of the model are specified
58 manually using domain knowledge.

59

60 **3 Discussion**

61 We tested the model qualitatively by analyzing the system action for different types of
62 simulated user states. We also analyzed system response to user errors that might occur due
63 to temporary distractions and errors due to noisy sensors. We found that in most cases, the
64 model correctly determines various cognitive states and generates an appropriate action.
65 Further data needs to be collected to incorporate information about the user's driving
66 patterns. In addition, the model must be extended to include various levels of prompting. For
67 example, users with mild dementia might require simple reminders, while more detailed
68 prompts might be necessary to assist users with severe dementia. Future work involves using
69 real wheelchair motion data and maps generated by the stereovision camera, as well as
70 conducting user studies to determine the effectiveness of the system with the target user
71 population. These studies will allow us to assess the level of accuracy needed for robotics
72 and machine learning algorithms to be successfully applied to the field of assistive
73 technology.

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