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HandiViz project: clinical validation of a driving assistance for electrical wheelchair

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Abstract—Autonomy and independence in daily life, whatever the impairment of mobility, constitute fundamental needs that participate to the self-esteem and the well-being of disabled people. In this context, assistive technologies are a relevant answer. To address the driving assistance issue, we propose in this paper a unified shared control framework able to smoothly correct the trajectory of the electrical wheelchair. The system integrates the manual control with sensor-based constraints by means of a dedicated optimization strategy. The resulting low-complex and low-cost embedded system is easily plugged onto on-the-shelf wheelchairs. The robotic solution has been then validated through clinical trials that have been conducted within the Rehabilitation Center of Pôle Saint Hélér (France) with 25 volunteering patients presenting different disabling neuro-pathologies. This assistive tool is shown to be intuitive with 25 volunteering patients presenting different disabling neuro-pathologies. This assistive tool is shown to be intuitive and robust as it respects the user intention, it does not alter perception while reducing the number of collisions in case of hazardous maneuvers or in crowded environment.

I. INTRODUCTION

Global population ageing as well as disability compensation constitute major challenging societal and economic issues \cite{1}. In particular, achieving autonomy remains a fundamental need that contributes to the wellness and the well-being. In this context, innovative and smart technologies are designed to achieve independence while matching user individual wants \cite{2}.

Assistive technologies are then considered as a great opportunity to improve the quality of life. Sensorial, physical and mental limitations can be exceeded by enhancing existing abilities. Assistive robots can then be seen as a solution for realizing daily living tasks.

For people suffering of motor disabilities (due to coordination limitations, dexterity incapacities, injury, accidents...), electrical wheelchair remains one of the most used assistive technology equipment as it is synonym of freedom of navigation and travel \cite{3}. However, operating a wheelchair in a secure way requires cognitive skills (typically to anticipate obstacles and to plan a safe trajectory) as well as performant visual-perceptual abilities \cite{4}. As a consequence, because of inadequate and dangerous reactions encountered while navigating, some disabled people are not allowed to drive electrical wheelchair, thus dramatically reducing their autonomy \cite{5}. Therefore, designing a robotic assistive solution related to wheelchair navigation remains of major importance.

Recent advanced studies then deal with Smart Wheelchairs in order to improve user quality of experience. As for example, the NavChair \cite{6}, the European FP7 Radhar project \cite{7} and the recent SYSIASS \cite{8} and COALAS projects \cite{9} were able to design systems that take partial or full control from the user for safe and effective navigation assistance. However, the main difficulty related to Smart Wheelchair systems is to associate low-cost embedded solutions and efficient and robust framework \cite{10}.

The French HandiViz project aims then at realizing an intuitive and low-cost driving assistance. The idea is to progressively correct the trajectory in a smooth manner so that to avoid static or dynamic obstacles. The related robotic system should be easily plugged onto any off-the-shelf wheelchairs thus targeting a widespread usage as a commercialization of the assistance system is envisaged.

To this aim, different control frameworks based on a shared control strategy have been previously developed. However, in the literature, such robotic assistance systems typically require expensive multi-sensor system (e.g. laser-range finder \cite{11}, \cite{12}), specific instrumentation or adapted interfaces \cite{13}, \cite{14}. In this paper, we design a generic shared control system that is independant of the used sensor types, as soon as these sensors are able to provide a distance information. The proposed solution fuses then the user input with a dedicated control law without any a priori knowledge of the environment, in order to design a reactive local approach. In this domain, state-of-the-art techniques typically use visual servoing frameworks \cite{15}, \cite{16} or Potential Field Methods \cite{17}, \cite{12} that act as repulsion forces. In our case, contrary to \cite{12} that abruptly switches to obstacle avoidance/full control states, we define a unified framework able to progressively and adaptively modulate the applied trajectory correction with respect to the distance with the detected obstacles. It results in a smooth correction thus improving the quality of navigation.

Besides, assessing a robotic system remains a critical step in the development of an assistive technology. As disability is a complex notion, it is difficult to provide a universal bench-
Hélier in Rennes (France) with the help of 23 patients.

A. Modelling

As shown on Fig.1,

- let \( \mathbf{v}_{c_i} = (u_{c_i}, \omega_{c_i}) \) be the velocity of the sensor \( c_i \),
- let \( x_i \) be the distance from the sensor \( c_i \) to the obstacle,
- let \( x_i^* \) be a minimum distance from the sensor \( c_i \) to the obstacle,
- let \( e_{c_i} = x_i - x_i^* \) be the error between \( x_i \) and \( x_i^* \).

We can define a Jacobian \( \mathbf{J} \) such as

\[
\mathbf{v}_{c_i} = \mathbf{J} u
\]

We obtain

\[
\mathbf{v}_{c_i} = \mathbf{J} u + \mathbf{J}_\omega \omega
\]

with \( \mathbf{J}_u \) of size 6 x 1 and \( \mathbf{J}_\omega \) of size 6 x 1

- Let \( \mathbf{L}_{x_i} \) be the interaction matrix for the sensor \( c_i \).

We have

\[
\dot{x}_i = \mathbf{L}_{x_i} \mathbf{v}_{c_i}
\]

with \( \mathbf{L}_{x_i} \), of size 1 x 6

By combining equations (2) and (3), we obtain

\[
\dot{x}_i = \mathbf{L}_{x_i} \mathbf{J}_u u + \mathbf{L}_{x_i} \mathbf{J}_\omega \omega
\]

To avoid collision with the obstacle, we constrain \( \dot{x}_i \) by a minimum value \(-\lambda e_{c_i}\) corresponding to a proportional corrector. Consequently, we get

\[
\dot{x}_i \geq -\lambda e_{c_i}
\]

By combining equation (4) and inequation (5), we get

\[
\mathbf{L}_{x_i} \mathbf{J}_u u + \mathbf{L}_{x_i} \mathbf{J}_\omega \omega \geq -\lambda e_{c_i}
\]

As \( \mathbf{J}_u \) is of size 6 x 1 and \( \mathbf{L}_{x_i} \) is of size 1 x 6, we can define a scalar \( a_{c_i} \) such as

\[
a_{c_i} = \mathbf{L}_{x_i} \mathbf{J}_u
\]

Similarly, we can define a scalar \( b_{c_i} \) such as

\[
b_{c_i} = \mathbf{L}_{x_i} \mathbf{J}_\omega
\]

Then, we get

\[
a_{c_i} u + b_{c_i} \omega \geq -\lambda e_{c_i}
\]

which defines a half-plane in the \( u - \omega \) plane as shown on Fig.2.

![Fig. 1. Definition of the robot frame](image1)

![Fig. 2. Definition of half plan in the \( u - \omega \) plane](image2)

Inequation (9) can be rewritten as

\[
\mathbf{A} \mathbf{u} \geq \mathbf{B}
\]

with \( \mathbf{A} = \begin{bmatrix} a_{c_1} & b_{c_1} \\ a_{c_2} & b_{c_2} \\ \vdots & \vdots \\ a_{c_N} & b_{c_N} \end{bmatrix} \) and \( \mathbf{B} = \begin{bmatrix} -\lambda e_{c_0} \\ -\lambda e_{c_1} \\ \vdots \\ -\lambda e_{c_N} \end{bmatrix} \).

When considering \( N \) sensors, we can rewrite inequation (10) using

\[
\mathbf{A} = \begin{bmatrix} a_{c_0} & b_{c_0} \\ a_{c_1} & b_{c_1} \\ \vdots & \vdots \\ a_{c_{N-1}} & b_{c_{N-1}} \end{bmatrix} \quad \text{and} \quad \mathbf{B} = \begin{bmatrix} -\lambda e_{c_0} \\ -\lambda e_{c_1} \\ \vdots \\ -\lambda e_{c_{N-1}} \end{bmatrix}.
\]
**B. Computing control values**

- Let $\mathbf{u}_{\text{op}} = (u_{\text{op}}, \omega_{\text{op}})$ be the velocity input from the user.
- Let $\mathbf{u}_{\text{cmd}} = (u_{\text{cmd}}, \omega_{\text{cmd}})$ be the velocity sent to the robot.

$\mathbf{u}_{\text{cmd}}$ is computed from $\mathbf{u}_{\text{op}}$ under constraints (10) by minimizing a cost function $f$. This can written as an optimization problem using

$$
\begin{align*}
\mathbf{u}_{\text{cmd}} &= \min_{\mathbf{u}} f(\mathbf{u}_{\text{op}}, \mathbf{u}) \\
\mathbf{A}\mathbf{u} &\geq \mathbf{B}
\end{align*}
$$

(11)

We define the function $f$ such as

$$
f(\mathbf{u}_{\text{op}}, \mathbf{u}) = \alpha(u - u_{\text{op}})^2 + \beta(\omega - \omega_{\text{op}})^2
$$

(12)

with $\alpha \geq 0$ and $\beta \geq 0$.

Then the problem can be written as a classical quadratic optimization problem using

$$
\begin{align*}
\mathbf{u}_{\text{cmd}} &= \min_{\mathbf{u}} \mathbf{g}^T \mathbf{u} + \frac{1}{2} \mathbf{u}^T \mathbf{H} \mathbf{u} \\
\mathbf{A}\mathbf{u} &\geq \mathbf{B}
\end{align*}
$$

(13)

with $\mathbf{g} = \mathbf{u}_{\text{op}}^T \begin{bmatrix} -\alpha & 0 \\ 0 & -\beta \end{bmatrix}$ and $\mathbf{H} = \begin{bmatrix} \alpha & 0 \\ 0 & \beta \end{bmatrix}$.

As shown on Fig. 3, we can define two sets $\mathbf{P}_0$ and $\mathbf{P}_1$ such as

$$
\begin{align*}
\mathbf{P}_0 &= \{ \mathbf{u} | \mathbf{A}\mathbf{u} \geq \mathbf{B} \} \\
\mathbf{P}_1 &= \{ \mathbf{u} | \mathbf{A}\mathbf{u} > \mathbf{B} \}
\end{align*}
$$

(14)

Hence the problem (13) can be resolved only if $\mathbf{P}_0 \neq \emptyset$. If $\mathbf{P}_0 = \emptyset$, $x_i^*$ can be dynamically decreased to expand the size of $\mathbf{P}_0$.

In fact, the quadratic optimization problem (13) can be solved by observing two different cases:

- Case 1: if $\mathbf{u}_{\text{op}} \in \mathbf{P}_0$, then $\mathbf{u}_{\text{cmd}} = \mathbf{u}_{\text{op}}$,
- Case 2: otherwise, as the cost function is convex and its minimum is reached when $\mathbf{u} = \mathbf{u}_{\text{op}}$, $\mathbf{u}_{\text{cmd}} \in \mathbf{P}_0 - \mathbf{P}_1$ meaning that it exists at least one $i$ such as $a_i u + b_i \omega = -\lambda e_i$.

A search along $\mathbf{P}_0 - \mathbf{P}_1$ gives the result of the optimisation problem 13.

To sum up, the proposed shared control solution allows the user to control the wheelchair while observing safety constraints. The formulation of the problem leads to solve a simple quadratic system under constraints. The resulting algorithm provides then a progressive trajectory correction as no singularities can be observed.

Next section is devoted to the clinical trials and the evaluation of this assistive tool.

**III. CLINICAL TRIALS**

Clinical trials are mandatory when considering robotic assistance device for wheelchair navigation purposes. This study constitutes the first step in clinical validation. Before envisaging a widespread usage of the system, we have to first demonstrate the robustness and the acceptability of the solution, along with its ability to reduce collision.

This evaluation study has received a favorable opinion from the ethics committee of Pontchaillou Hospital (Rennes, France).

**A. Experimental setup**

The wheelchair used in the clinical trials is based on an off-the-shelf YouQ Luca wheelchair. This wheelchair has 5 wheels, where 2 of which are actuated and the 3 others are caster wheels. The user joystick and the R-Net wheelchair electronics come from Penny & Giles.

In Section II we described a generic shared control process which has been designed independently of the sensors used. To cope with the low-cost requirement that should guarantee in the end a widespread usage, the wheelchair has been equipped with 15 ultrasound sensors installed all around it, as shown on Fig. 4.

When considering the hardware architecture, an ARM CPU board is embedded onto the wheelchair to run the actual algorithm. With this setup, a latency of less than 10 ms is to be expected due to the nature of the communications encountered on the R-Net CAN bus. This latency has been shown to be undetectable by wheelchair users.

A user interface is installed behind the wheelchair: it allows the clinicians to start/stop the experiments and enable or disable the assistance. The volunteers are then not aware of the activation or not of the assistance module.

**B. Evaluation methodology**

The objective of this study is to validate the robustness and the efficiency of the system, as well as this acceptability. Hence we have recruited 23 disabled people who are considered as experts in wheelchair driving. These patients present different pathologies that induce different degrees in motor impairments: tetraplegia, cerebral palsy, amputee, brain injured, stroke. In order to avoid a bias in the evaluation process, we consider only one configuration of the system: users perform then in the same conditions. Then the parametrisation of the wheelchair is realized only once, which can slightly differs from the parametrisation of the wheelchair of each participant. In addition, the maximum speed is set to 0.8 m/s for all the volunteers.
Tests have been performed at Rehabilitation Center Pôle Saint Hélier in Rennes, France. To this aim, an ecological circuit has been defined: Fig. 5 shows the complete path to be achieved by the users. The navigation is realized in real conditions, meaning that the corridors are crowded, and that dynamic or static obstacles are disposed along the circuit. In addition, users were supposed to realize difficult maneuvers in narrow spaces (turning around, rolling back at the end in a small corridor).

At the beginning of the experiments, volunteers were asked to reach the therapeutic apartment, then going back the same way until reaching a narrow corridor where they have to realize a 90° rotation before entering backwards. We dispose no landmark on the floor or on the walls. Volunteers receive no further indications so that they could choose the trajectory they want.

Random double-blind trials were then conducted: for each volunteer, experiments were performed twice with and without assistance without the patient knowing which one is which. Each participant took about four minutes to complete each trial. A short QUEST-like questionnaire was fulfilled at the end of each trial [19].

C. Results

Table I sums up the results from the clinical trials as described in the previous section. For each participant, we first determine whether the first experiment is realized with or without assistance. Then in each case, the number of collisions are recorded by an external observer and the QUEST questionnaire is fulfilled, leading to a score between 0 and 40.

In addition, when the assistance module is activated, the activation ratio, corresponding to the percentage of navigating time during which the correction is applied, is measured.

We can observed that on average, the number of collisions when assistance is engaged is reduced by half when compared to navigation without assistance. Moreover, as the answers to the questionnaire are in favour of the assisted trial, it exhibits the acceptability of the assistance solution.

In particular, it emphasizes the fact that the assistance does not alter the driving experience even if the activation ratio can be up to 35.51%.

Collisions typically occur during difficult maneuvers as illustrated in Table II, for example while reversing in a narrow corridor or crossing a moving obstacle or operating a sharp turn around small static obstacles that could be difficult to perceive for people suffering of visual impairments.

IV. Conclusion

This paper has presented a unified shared control framework able to provide a smooth trajectory correction while driving a wheelchair. Based on an efficient fusion of the user input and the obstacle avoidance task, a progressive modulation of the velocity of the wheelchair is obtained.

Clinical trials involving 23 volunteering disabled people have shown that for individuals who are used to drive their electrical wheelchair daily, the driving assistance is intuitive and does not modify their perception. In addition, while navigating along an ecological circuit, collisions during difficult maneuvers (turning around in a narrow space, rolling backward...) are reduced when the assistance is engaged.

Thus this paper demonstrates the clinical validation of the proposed assistive system. The next step in the clinical validation process will consist in realizing new trials with disabled who experience great driving difficulties or are even not allowed to drive an electrical wheelchair because of their poor navigation skills.

Acknowledgement

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<table>
<thead>
<tr>
<th>Participant</th>
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<th>With Assistance</th>
<th>Activation rate (%)</th>
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**Average** 0.869 32.91 0.435 33.30 8.11

**TABLE I**
Collision results

**TABLE II**
Images from the experiments

![Image a) Collision while reversing](image1)

![Image b) Moving obstacle](image2)

![Image c) Rotation around a dust bin with assistance - No Collision](image3)

![Image d) Rotation around a dust bin without assistance - 1 Collision](image4)
and Luc Le Pape from Ergovie, Eric Bazin from INSA Rennes, Amélie Colin from Healthcare Network Breizh Paralysie Cérébrale and all the volunteers who contributed to this study.

REFERENCES


