A Robust Open-source Tendon-driven Robot Arm for Learning Control of Dynamic Motions

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1 1 Introduction

Human athletic movements, such as those seen in table tennis 2 or soccer, require a dynamic interplay of power, agility, and 3 precision. They also involve quick reactions to unpredictable 4 stimuli, while effectively managing impacts and maintaining 5 safety during rapid interactions. Such complexities pose chal-6 lenges for robots aiming to emulate or collaborate with humans 7 in sports settings. Most commercial robots are either precise 8 yet fragile or safe yet underpowered. Tendon-driven robots 9 provide a middle ground, lessening impact risks due to low 10 inertia. Still, they often face precision issues due to unpre-11 dictable friction. 12

Our paper presents a newly designed 4-DoF tendon-driven 13 robot arm, powered by pneumatic artificial muscles (PAMs), 14 designed to minimize friction. This design achieves high force, 15 low inertia, backdrivability, and superior precision compared 16 to counterparts. While PAMs add complexities in control due 17 to their nonlinear nature, reinforcement learning (RL) proves 18 effective in handling them [1]. Our robot's design also miti-19 gates RL's collision risks during explorative training. 20

To foster further innovations, both the robot's hardware and software are made open-source. The hardware predominantly employs readily available parts. Our software offers an adaptable API in Python and C++ based on the o80 framework [2], interfacing with the robot's PLC over UDP. We will also opensource a huge proprioceptive data-set (25 days) including motion data at various speeds and forms.

28 2 PAMY2: Design & Implications



For optimized control and durability, we utilize Polytetrafluoroethylene (PTFE) for our Bowden tubes, reducing friction and heat, while enhancing the tendon's lifespan. This design, complemented by an inner tube and custom 3D-printed outer supports, ensures a consistent tendon length during



(a) Collision force measurements



(b) Table tennis smashing motion

Figure 1: Experiments performed to validate PAMY2. (a) Experimental setup for the collision force measurements. A Pilz PRMS is mounted to a table onto which the end effector of the robot is colliding. (b) Table tennis smashing experiment.



Figure 2: Table tennis smashing experiment from [1] with identical setting (incl. hyperparameters) but with PAMY2 instead of PAMY1 [3, 4]. (a) PAMY2 reaches $4 \times$ of the final performance of PAMY1. (b) The velocity of the returned ball is appr. twice as high. (c) Although harder to control precisely, the agent misses the ball less often and (d) returns it more likely to the desired ball landing region.

40 movements, thereby improving precision. We've replaced previous gliding bearings with ball bear-41 ings to further decrease friction. Our updated pneumatics offers better airflow through enhanced 42 tube routing, a buffer reservoir for stable air pressure, and a ring circuit for consistent pressure 43 distribution.

44 **3** Experiments & Evaluations

We assessed our new robot arm's capabilities in terms of impact safety, robustness, ease of control,
 and precision during fast movements.

Impact Safety: Our tendon-driven arm promises enhanced safety
over traditional motor-driven systems due to its lightweight design
achieved by placing the heavy actuators at the base. Using the Pilz
Robot Measurement System (PRMS) to gauge collision forces (Figure 1a), our arm showed superior safety, achieving similar force levels as conventional systems but at nearly quadruple the speed (see
Figure 3).

Robustness: For reinforcement learning to be effective, the robot's longevity is paramount. We tackled friction, a known wear-and-tear contributor in tendon-driven robots. Our study revealed that the redesigned Bowden tubes produce less heat, signifying re-duced friction compared to the system in [3]. We also released a dataset of a 25-day continuous operation, confirming our robot's durability and consistency during dynamic tasks.

Ease of Control: Improving linearity of a system makes control 61 easier. Comparing our new system, PAMY2, with its predecessor, 62 PAMY1, the former exhibited better amplitude responses and sig-63 nificantly reduced nonlinearity during dynamic actions. Further-64 more, replicating a table tennis smashing experiment from [1] (Fig-65 ure 1b), using Proximal Policy Optimization (PPO) [6] for learn-66 ing, we achieve significantly higher ball speeds while improving on 67 precision, demonstrating the benefits of the new design for highly 68 dynamic and precise motions (Figure 2). 69



Figure 3: Force map for PAMY2, showing peak impact forces across impact velocities and contact situations [5].

70 **References**

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