

Design of a mechatronic device for learning a gesture and strengthening specific muscles.

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Abstract

The objective of this project is to create a mechatronic device that can assist in motor learning. The device will facilitate interaction between a human and a robot, taking into account the kinematics and dynamics of the movement that the human is trying to learn. The potential applications of this development range from learning high-level sports movements to rehabilitating motor skills after trauma or surgery.

Context

This thesis is part of the three themes of the RoBioSS team: 'Dynamics of motion', 'Physical and systemic interactions', and 'Mechatronics and cobotics'. The acquisition of motor skills, ranging from everyday movements to technical gestures and movements leading to high-level sporting performance, is based on two main processes: trajectory learning and dynamic adaptation [16].

We will use the term 'global movement' to refer to the movement of the entire body when performing a gesture (see Figure 1a). We will also use the term 'isolated movement' to refer to the movement of a single joint, which is identical (from a kinematic perspective) to the joint movement performed during the global movement, but without involving the other joints of the body (see Figure 1b).

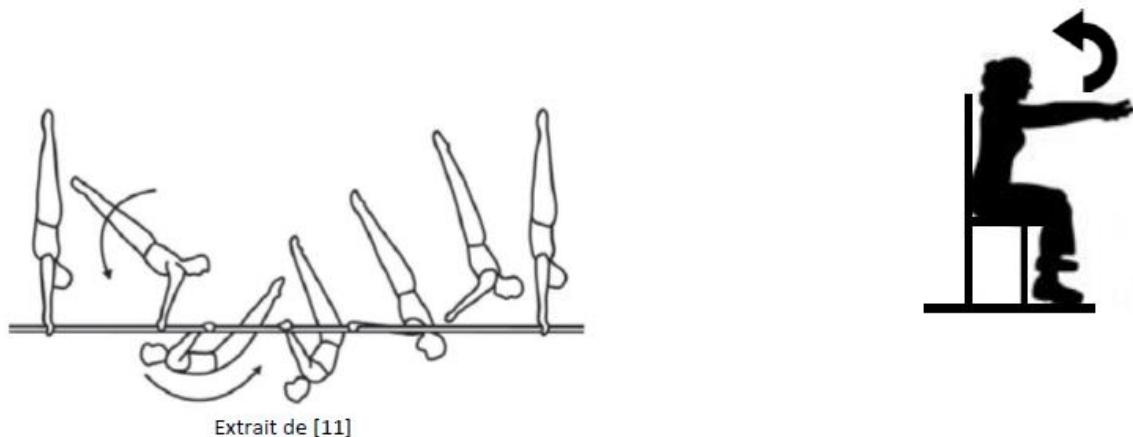


Figure 1: Difference between an antepulsion performed during a "global movement" (a) and during an "isolated movement" (b).

Learning joint kinematics can be initiated by mimicking in a single joint the movements to be performed. However, this approach has limitations when considering the dynamic aspects of the movement. Performing adduction of the right arm in a standing position does not generate the same joint torque in the shoulder as performing it in a backward salto, even though the joint kinematics are identical [13]. Similarly, reproducing flexion-extension movements at knee with feet in

the air when sitting on a rehabilitation table generates less joint effort compared to supporting your body weight while walking on the ground.

The use of robotic systems to aid trajectory learning (guidance) has shown no real direct benefits [8,14]. On the other hand, robotic guidance can be useful when the movement involves several joints and/or several tasks, as it can free up attention from some aspects of the movement to focus on others [9].

With regard to dynamic adaptation, recent research [5,17] has shown that the process of dynamic adaptation process was based on the errors observed during the previous movement. By applying a force, via a robotic device, proportional to the observed deviation from the desired trajectory, it is then possible to correct the trajectory ("Error-based learning"). A mathematical model of this principle was proposed by [4] and tested with subjects walking on a treadmill. However, the question of the trade-off between of learning and the amount of assistance to be provided.

Coaches and technicians have long applied this principle, using their imagination to devise simple devices. For example, hang a bucket from the end of a rope and asking a gymnast learning to do pommel circles to place his or her feet in the bucket (Figure 2), allows the gymnast to create the external effort needed to keep his or her feet above the ground.

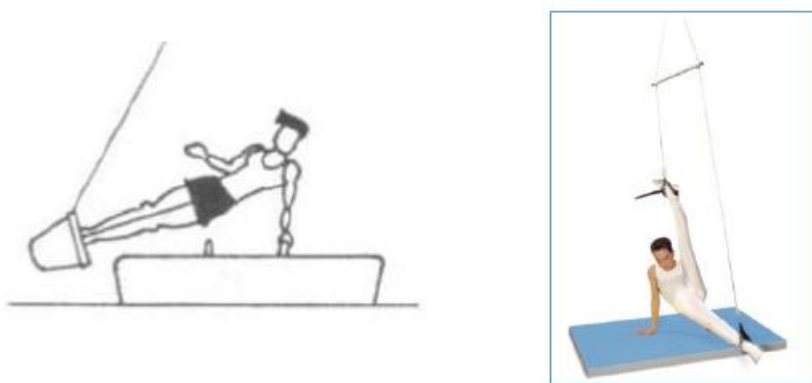


Figure 2: Learning device for pommel ring kinematics.

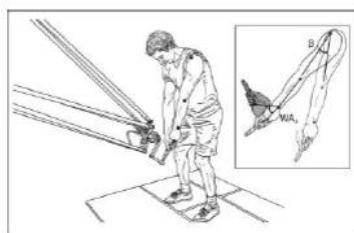
While this device proves effective at the beginning of the learning phase, facilitating the gestures, its prolonged use quickly becomes counter-productive, as it encourages the development of joint efforts that are very different from those that the gymnast will have to provide. What happens when the bucket is removed, thus removing the possibility of creating an external effort on the feet? The use of this type of device therefore calls into question the ability to transfer of learning.

The same analysis can be made for all kinds of assistance provided by a trainer, robotic device (such as an exoskeleton [10]) or even a physiotherapist during the performing a gesture. In reality, these are more assistance than learning devices, in the sense that they compensate for a deficit in effort.

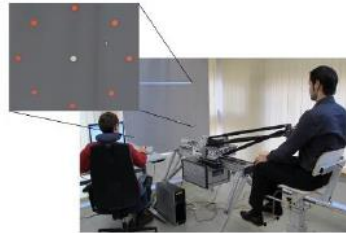
Another robotic assistance strategy, called the "error-augmentation strategy", was then studied by [15]. It consists in having a robotic device apply a force opposite to that which should be applied to correct any deviations in trajectory. If this strategy initially leads to greater trajectory deviations, when the applied force disappears an "after-effect" is observed, leading the subject to follow the desired trajectory perfectly with appropriate muscle activation.

Most of the existing mechanical/mechatronic devices related to movement have been developed for studies in the field of motor control (Figure 3a and 3b), movement analysis and simulation (Figure 3c

and 3d), or for rehabilitation programs (Figure 3e). Few are designed to teach specific movements or strengthen muscles. The design of today's ergometers requires the user/athlete to master gestures so that the efforts felt are perceived as realistic. There are no ergometers equipped with a biomechanical movement simulation model that model, taking into account the subject's anthropometric characteristics, which enable to learn how to develop the necessary articular efforts, with feedback on the quality of the kinematics achieved. The general scientific problem proposed by this research program research program is to develop this type of ergometer, whose use can be extend from the field of functional rehabilitation to that of high-level sports performance.



a) Contrôle du poignet lors du putt au golf.



b) BioMotionBot pour l'analyse de tâches sensori-motrices



c) Ergomètre Kayak instrumenté pour la simulation de gestes sportifs (RoBioSS)



d) Ergocycle instrumenté pour l'analyse du mouvement de pédalage (RoBioSS)



e) Dispositif iso-cinétique destiné à la rééducation
Extrait de <https://www.medimex.fr/humac-norm-by-csmi.html>

Figure 3: Examples of mechanical and mechatronic devices related to human movement (a from [12]; b from [1]).

Objectives

The objectives of this project are to develop a mechatronic device that will allow :

- Learn to achieve a target joint kinematics during an isolated movement WHILE
- Learn to produce, during an isolated movement, the articular efforts developed for this kinematics during the global movement. These efforts will be obtained by optimizing the movement according to the subject's anthropometric characteristics and physical capabilities.
- Define the ideal opposing force to be applied so that it is acceptable to the user.
- Take into account the level of learning and expertise in robotic assistance.
- Define different forms of feedback to optimize learning without making learners dependent.

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