

1 Automated objective robot-assisted assessment of
2 wrist passive ranges of motion

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9 Word count: 1692

10 **Abstract**

11 The measurement of wrist passive ranges of motion (ROMs) can provide insight into improvements and allow for
12 effective monitoring during a rehabilitation program. Compared with conventional methods, this study proposed
13 a new robotic assessment technique for measuring passive ROMs of the wrist. The robotic system has a
14 reconfigurable handle structure that allows for multi-dimensional applications of wrist motions. The assessment
15 reliability of this robotic system was analysed on 11 subjects for measuring wrist extension/flexion and
16 radial/ulnar deviation. Preliminary data demonstrated its potential with intraclass correlation coefficient ($ICC_{2,1}$)
17 all greater than 0.857 and standard error of measurement (SEM) less than 3.38° . Future work will focus on the
18 standardization of the assessment protocol of this robotic system for assessment purposes, paving the way for its
19 clinical application.

20 **Keywords:** Robotic, assessment technique, wrist, passive range of motion

21 **1. Introduction**

22 Although the human wrist is a biomechanical marvel when it is intact and functioning,
23 orthopedic or neurological impairments inevitably cause dysfunction to its motion (Skirven et
24 al., 2011). Over the past few decades, a variety of robot-assisted rehabilitation techniques have
25 been developed to restore wrist/hand motor function. Krebs et al. (2007) integrated a wrist
26 robotic device into the shoulder-and-elbow MIT-MANUS. Initial clinical results demonstrated
27 the efficacy of this robot in providing continuous passive motion, strength, sensory, and
28 sensorimotor training for the wrist. Squeri et al. (2014) developed a haptic robot to quantify
29 motor impairment and assist wrist articular movements. The proposed adaptive control strategy
30 showed great potential in maximizing the recovery of the wrist ranges of motion (ROMs). An
31 important feature of robot-assisted therapy is that exercises should be tailored to specific
32 impairments (Sanguineti et al., 2009). Passive ROM is of high clinical importance in therapy
33 and assessment of musculoskeletal disorders. Typically in robot rehabilitation an initial active
34 ROM target is set so that subjects can easily achieve the goal, and that ROM is progressively

35 modified over the course of the entire protocol. However, these existing rehabilitation systems
36 have not been implemented with assessment of passive wrist ROMs, nor has any method been
37 validated to reliability assess passive ROMs.

38 This study proposes a new robot-assisted assessment technique for measuring passive ROMs
39 of the human wrist. The reliability of this method for measuring wrist extension/flexion (E/F)
40 and radial/ulnar deviation (RD/UD) is analysed. To the best of the authors' knowledge, this is
41 the first example of robot-assisted automation of assessing passive wrist ROMs based on
42 predefined joint torque values. The reconfigurable robotic design is also a novel device which
43 allows continuous posture adjustment of the handle.

44 **2. Methods**

45 **2.1 Participants**

46 Eleven healthy subjects (*six males: Age 26.17 ± 4.22 years, Height 173.83 ± 7.28 cm, Weight*
47 *79.83 ± 10.03 kg, and five females: Age 26.40 ± 5.98 years, Height 164.20 ± 3.49 cm, Weight*
48 *55.30 ± 5.72 kg) volunteered to participate in this study. The study was approved by the
49 University of Auckland, Human Participants Ethics Committee (019707) and consents were
50 obtained from all participants.*

51 **2.2 Instrumentation**

52 The proposed assessment technique is implemented using a reconfigurable wrist rehabilitation
53 robot and an adaptive passive assessment strategy, as in Figure 1. The robot mechanically
54 consists of a handle, a handle holder, the base, and the forearm holder with straps. The handle
55 can be rotated along the handle holder for arbitrary posture adjustment. To follow convention

56 (Horger, 1990), the motion measurement of wrist E/F is achieved by setting 0° of the handle
57 (horizontal), and that of RD/UD by adjusting the handle to 90° (vertical).

58 Its electronic control system consists of a flat brushless motor (EC 90, Maxon), a magnetic rotary
59 sensor (AS5048A, AMS), and a static torque sensor (JNNT-1, Zhongwan), which allows the
60 implementation of adaptive passive assessment technique. The robot reverses when real-time
61 human-robot interaction torque triggers a predefined value of wrist passive torque. Data of the
62 angular position of the human wrist are collected from the built-in magnetic rotary sensor.

Figure 1. A reconfigurable wrist rehabilitation robot. (IT: Interaction torque; WPT: Wrist passive
torque; E/F: Extension/flexion; RD/UD: Radial/ulnar deviation)

63 **2.3 Procedures**

64 Subjects sat on a height-adjustable chair with the forearm strapped and hand grasping the
65 handle. The wrist joint was visually adjusted to approximate the rotational axis of the wrist
66 robot, with fingers holding the handle for assessment exercises. The device was set with two
67 configurations: 0° for wrist E/F and 90° for wrist RD/UD. The predefined wrist passive torque
68 was set with two grades: 2.5/3.5 Nm for males and 2/2.5 Nm for females. These values were
69 set using experience from pilot trials, and in clinical practice can be predefined by therapist
70 based on a specific subject and his/her pathology. Thus there are a total of four measures (APA-
71 0-L, APA-0-H, APA-90-L, and APA-90-H), where APA means adaptive passive assessment,
72 0 or 90 is the position of the handle, L or H refers to the level of torque limit (Low level: 2 Nm
73 for females and 2.5 Nm for males, High level: 2.5 Nm for females and 3.5 Nm for males). For
74 each measure, each participant was required to repeat 12 cycles of the passive assessment

75 movement. This device was driven to work in a constant low-velocity environment (5°/s) to
76 mimic clinical wrist rehabilitation exercises.

77 **2.4 Statistics**

78 Intraclass correlation coefficient (ICC) was used to examine the test-retest reliability of the
79 four measures. Absolute reliability was determined by calculating the standard error of
80 measurement (SEM) and smallest real difference (SRD) with 95% confidence interval, as in
81 Eqs (1) and (2) (Weir, 2005), where standard deviation (SD) is the mean SDs of all
82 measurements. SEM % and SRD % were also calculated to facilitate the comparability with
83 other studies. In this study, ICC_{2,1} was selected for data analysis with two-way random,
84 absolute agreement, 95% confidence interval.

$$SEM = SD\sqrt{1-ICC} \quad (1)$$

$$SRD = SEM * 1.96 * \sqrt{2} \quad (2)$$

85 **3. Results**

86 Each participant conducted 12 cycles of passive assessment movement for each measure. The
87 maximum and minimum joint angle values can be extracted for each cycle. To minimize the
88 effects from random factors, such as active engagement with the robot from human users, each
89 data set finally includes 10 elements by removing the maximum and minimum values. To
90 facilitate data analysis in SPSS, **Table 1** presents the mean and SD of measurements over all
91 participants for each measure and each cycle, where wrist ROM values do not vary in a
92 predictable way with the measurement cycle. It is assumed that exercises with 12 cycles of
93 measurements are not enough for an increase of wrist ROM, and the influence of removing
94 maximum and minimum values is negligible.

Table 1. Means and SDs of measurements over all participants.

95 **Table 2** summarizes the statistical results of the test-retest reliability of the selected four
96 measures, where all ICC_{2,1} values are greater than 0.857 and SEM values are less than
97 3.38°. Munro's correlation description (Munro, 2004) is used to interpret ICC values as
98 high reliability of APA-0-H-F and APA-90-H-UD. The reliability of all others is excellent
99 with ICC_{2,1} no less than 0.9.

Table 2. Statistical results of the test-retest reliability of the selected four measures.

100 **4. Discussion and conclusion**

101 To achieve adaptive robotic protocols, assessment is normally required to tune controllers
102 based on training performance or joint capacity. Squeri et al. (2014) explored an adaptation
103 law by increasing task difficulty as a subject succeeds in completing a task, and assessed wrist
104 ROMs by the use of the Fugl-Meyer assessment technique (Gladstone et al., 2002). While this
105 robotic system showed promise in increasing wrist ROMs and decreasing its spasticity, it did
106 not allow for automatic detection of the joint ROMs. In contrast, the proposed robot-assisted
107 technique in this study can conduct automatic assessment of human wrist ROMs. Even though
108 its reliability is satisfactory as in **Table 2**, the clinical efficacy and applications have not been
109 clear yet when comparing with standard assessment tools, such as Fugl-Meyer Assessment
110 (Gladstone et al., 2002) and Modified Ashworth Scale (Gregson et al., 1999).

111 With respect to conventional assessment methods of wrist passive ROMs, the proposed robotic
112 technique has the following features. First, it does not depend on manual operation by raters.
113 This means that the influences of subjective factors from raters can be greatly eliminated.
114 Second, this proposed technique depends on multiple measurements to minimize the influences

115 of objective factors from participants. For passive ROM assessment of the wrist, participants
116 are theoretically required to be fully relaxed and avoid active contribution that may
117 significantly affect the accuracy. Third, the proposed technique has comparable assessment
118 accuracy to manual methods. An example is a manual device developed by Zhang et al. (2015)
119 for measuring ROMs of ankle dorsiflexion and plantarflexion with all ICC_{2,1} values no less
120 than 0.846. The intrarater reliability of goniometric measurements of passive wrist motions
121 was evaluated on 48 subjects (Horger, 1990), with ICC_{2,1} values no less than 0.908 and SEM
122 values no greater than 3.537°. These are in line with the proposed robot-assisted assessment
123 results. Last but most unique, the proposed robotic assessment technique can be reconfigured
124 for continuous adjustments. The most basic two configurations are 0° of the handle for wrist
125 E/F, and 90° for wrist RD/UD.

126 The relevance of the proposed robot-assisted assessment method to therapy is manifested in
127 two aspects. One is to allow robot-assisted assessment of the wrist passive ROM during the
128 training, facilitating intelligent control of the rehabilitation protocol. The participant does not
129 have to get off the device for progress evaluation during robot-assisted therapy. The proposed
130 method, in the other hand, provides a more objective assessment method compared to
131 goniometric measures. Traditional goniometric measures are subject to different raters (Horger,
132 1990), while this technique requires only the determination of an appropriate torque limit.

133 While the proposed assessment technique has many advantages, three limitations exist and
134 should be further investigated. One is the determination of the predefined wrist passive torque.
135 Potential values can be set by physiotherapists based on their clinical experiences or a
136 preliminary examination. Second is the determination of the number of assessment cycles. Both
137 statistical and clinical requirements should be considered to derive the optimal cycle number
138 to trade off assessment accuracy with time. Thirdly, measures should be taken to minimize

139 active contribution from participants. The next prototype will be also implemented with a
140 harmonic reducer to minimize gear clearance for improved assessment accuracy, and with a
141 load-dependent forearm holder to allow for consistent anatomical motion for comfort and
142 safety.

143 This is the first attempt, in the field of robot-assisted rehabilitation to automate the assessment
144 of passive joint ROMs, wherein the wrist robot can be reconfigured to achieve continuous
145 postures of the handle. Preliminary data from 11 healthy subjects demonstrated its potential
146 with ICC_{2,1} all greater than 0.857 and SEM less than 3.38°. Future work will mainly aim to
147 standardise the protocol of the proposed robotic system for measuring passive ROMs of the
148 human wrist. Its clinical efficacy in contrast with standard clinical assessment tools should be
149 also investigated.

150 **Conflict of interest statement**

151 The authors declare that there is no conflict of interest.

152 **Acknowledgement**

153 This material was based on work supported by the University of Auckland, Faculty of
154 Engineering Research Development Fund-3625057 (Physical Robot-Human Interaction for
155 Performance-Based Progressive Robot-Assisted Therapy).

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