# The influence of posture, applied force and perturbation direction on hip joint viscoelasticity

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Abstract-Limb viscoelasticity is a critical factor used to 1 regulate the interaction with the environment. It plays a key 2 role in modelling human sensorimotor control, and can be used 3 to assess the condition of healthy and neurologically affected individuals. This paper reports the estimation of hip joint 5 viscoelasticity during voluntary force control using a novel 6 device that applies a leg displacement without constraining the hip joint. The influence of hip angle, applied limb force and perturbation direction on the stiffness and viscosity values 9 was studied in ten subjects. No difference was detected in the 10 hip joint stiffness between the dominant and non-dominant 11 12 legs, but a small dependency was observed on the perturbation direction. Both hip stiffness and viscosity increased monotoni-13 cally with the applied force magnitude, with posture to being 14 observed to have a slight influence. These results are in line 15 with previous measurements carried out on upper limbs, and 16 can be used as a baseline for lower limb movement simulation 17 and further neuromechanical investigations. 18

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#### I. INTRODUCTION

Muscles are characterised by their viscoelasticity, where 20 stiffness and viscosity increase with activation. By co-21 activating the muscles acting on limbs, the human nervous 22 system can control its stiffness and viscosity in magnitude, 23 shape and orientation [1]. Critically, this enables humans to 24 regulate their interaction with the environment [2] e.g. during 25 object manipulation, or for running optimally on different 26 grounds. 27

In order to understand how humans control the limb 28 viscoelasticity, a large body of experiments have estimated 29 stiffness and viscosity in the upper limb, in particular at the 30 wrist and arm [1]. Stiffness and viscosity can be measured 31 indirectly by applying a mechanical disturbance on the limb 32 and regressing the resulting changes of position and force. 33 Measurements carried out using this method showed that 34 stiffness generally increases linearly with the applied force: 35 in one deafferented muscle, in a single joint (thus including 36 reflexes), and in the arm [1]. 37

Much less is known on the viscoelasticity in the lower 38 limbs, in part due to the difficulty to carry out suitable 39 experiments involving heavy leg mass. For instance, existing 40 robotic interfaces to estimate viscoelasticity in the lower 41 limb either require a sitting or lying position [3], [4], [5], 42 [6], or are not sufficiently rigid to apply fast perturbations 43 without causing non-negligible oscillations e.g. [7], [8], [9], 44 [10]. In addition, all of these interfaces are affixed to the 45 body and thus determine the joints around which the limb 46 can move, while anatomical joints generally vary with the 47 posture (e.g. the knee joint rotates and translates during 48 locomotion). An alternative method consists of applying 49 perturbations directly on the foot, which can be used to 50 estimate ankle viscoelasticity [11], [12]. 51

In view of the limitations of previous devices to investigate the lower limb viscoelasticity, we have developed a dedicated robotic interface [13], [14]. This rigid interface can be used to investigate the hip, knee or ankle neuromechanics in a natural upright posture. It uses an endpoint-based approach to apply dynamic environments on the leg, thus does not need to impose joint movement.

Due to the difficulty to apply a mechanical disturbance on 59 the leg for estimating viscoelasticity, experiments reported 60 in the literature have been mainly restricted to a single joint, 61 i.e. at the ankle [15] and knee joints [16], [17]. In [8] the 62 LOPES exoskeleton has been used to estimate viscoelasticity 63 at the whole leg (including the hip joint), using a multi-joint 64 random torque as perturbation and an indirect measurement 65 of the resulting displacement from its series elastic actuators. 66 Random torque perturbations enable experiments to identify 67 both the stiffness and viscosity simultaneously [18], [19], but 68 may lead to identification problems as the velocity dependent 69 component is much smaller than the position dependent 70 component [20], [21], [22]. Therefore, we preferred using 71 a single position displacement to focus on accurately deter-72 mining the joint stiffness [23], and estimated viscosity in 73 a second step using the whole perturbation, including the 74 ramp up before the constant displacement phase and the 75 ramp down after it. This allowed us to examine the effect of 76 individual factors such as posture or force level separately. 77

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## II. METHODS

#### 79 A. Measurement system

The Neuromechanics Evaluation Device (NED) is a 80 powerful cable-driven robotic interface to yield computer-81 controlled dynamic testing on one leg of subjects supported 82 in a seated or upright posture ([13], Fig.1a). NED's open 83 stand support allows for conducting biomechanics identifica-84 tion experiments on various subjects including subjects with 85 impaired motor function. Used in different configurations, 86 this cable-based system can control the motion of the whole 87 leg, foreleg, or foot in order to estimate the hip, knee or 88 ankle neuromechanics. The pulley system can be adjusted 89 to keep the cable orientation approximately normal to the 90 limb's movement in different orientations for subjects of 91 various size [13]. 92

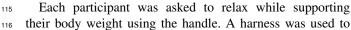
## 93 B. Experiment

The experimental protocol was approved by the Imperial 94 College Research Ethics Committee. Safety measures with 95 NED include software limits on the velocity, acceleration 96 and jerk, an optical system to check perturbation limits, 97 and emergency buttons for the subject and experimenters 98 [13]. Ten subjects (of age 21-27, with 6 females) without 99 any known lower-limb injury or medical condition were 100 recruited, they were informed on the device and experiment 101 and signed a consent form prior to participation. Subjects' 102 weight and leg length (from the anterior superior iliac spine 103 to the lateral malleolus) were then measured to estimate leg 104 inertia. These subjects' parameters are reported in Table I. 105 Bipolar electromyography (EMG) electrodes placed on 106 the rectus femoris, biceps femoris and tibialis anterior mus-107 cles were used to check when subjects are relaxed. EMG 108 signals were recorded at 2048 Hz, filtered using a [5,500]Hz 109 bandpass Butterworth filter, followed by a notch filter at 110 50 Hz (to attenuate the power frequency), then rectified. A 111 locking knee brace was used to keep the knee joint fixed 112 during the perturbations, and thus ensure that the leg is 113 straight during the whole procedure.

TABLE I BIOGRAPHICAL INFORMATION OF THE SUBJECTS

no	weight [kg]	height [m]	leg length [m]	age	sex
1	67	1.70	0.89	25	Μ
2	47	1.55	0.82	24	F
3	100	1.79	0.85	27	М
4	47	1.55	0.82	26	F
5	61	1.72	0.93	23	F
6	54	1.68	0.88	27	F
7	54	1.72	0.94	21	F
8	69	1.71	0.85	25	Μ
9	85	1.79	0.87	23	Μ
10	50	1.50	0.81	24	F

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connect the ankle of the leg under test to the cable system (Fig.1a). The subject could familiarise with the device by experiencing several perturbations, after which the system workspace safety limits were set. 120

A position perturbation was used to estimate hip joint 121 impedance. The perturbation shown in Fig.1b was used. It 122 consists of a 150ms long plateau with 20mm amplitude 123 (corresponding e.g. to an angle of 1.15° for a 90cm long 124 leg) with smooth ramps up and down. This perturbation 125 profile was determined by trial and error to ensure a force 126 measurement profile with negligible oscillations [13]. All 127 data but the EMG was measured at 1000 Hz. 128

For both legs, measurement was carried at different initial 129 postures with the hip angle (relative to vertical) at  $\{15^\circ,$ 130  $25^{\circ}$ ,  $35^{\circ}$ ,  $45^{\circ}$ ,  $55^{\circ}$ . At every posture, subjects were first 131 asked to relax (which was checked using EMG) while a 132 perturbation (with profile as in Fig.1b) was applied by the 133 system randomly in the forward or backward direction, with 134 five trials in each direction. The time of a perturbation was 135 also random so that the subject could not prepare for a 136 perturbation. 137

After experiencing the perturbation while relaxed (0N) 138 condition, each subject was asked to pull or push the leg to 139 exert a force of {-20, -10, 10, 20}N (with positive value for 140 backward kick) as was controlled by the subject using real-141 time feedback of the applied force displayed on a computer 142 screen placed in front of them. The force level was taken 143 relative to the relaxed condition of each subject, so that 144 the effect of gravity was compensated by the interface. To 145 prevent a subject from volitionally reacting to a perturbation, 146 visual feedback was not updated during the perturbation. 147

The subjects carried out two such measurement cycles (5 minutes each), with a ten minute rest during which they were detached from NED. For the two legs of the ten subjects, there were thus ten trials at each of the five postures and five force levels, using two perturbation directions (see Fig.2). The total experiment time was 100 minutes excluding the breaks.

## C. Data analysis

Linearization of the hip joint dynamics (valid for small angles  $\delta\theta$ ) yields <sup>156</sup>

$$\tau_m + \tau_g + \tau = I\delta\theta + B\delta\theta + K\delta\theta, \qquad (1)$$

where  $\tau_m$  is a torque produced by muscle tension to counteract the gravitational torque  $\tau_g$  and  $\tau$  corresponding to the external forces.  $I\delta\ddot{\theta}$  is the inertia component and  $B\delta\dot{\theta} + K\delta\theta$  the hip viscoelasticity component corresponding to a displacement angle  $\delta\theta$ , where *B* is the viscosity and *K* stiffness.

Two loadcells are used at the extremities of the ankle fixture to record the interaction forces between the robot and the subject. As the change of gravitational torque  $\delta \tau_g$  165 between the extremities was found to have less than 1% 167

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$$L\,\delta(F_1 - F_2) \equiv \delta\tau = I\delta\ddot{\theta} + B\delta\dot{\theta} + K\delta\theta\,,\tag{2}$$

where L is the leg length,  $F_1$  and  $F_2$  are the forces measured at the front and rear loadcells, respectively, and  $\delta \tau$  is the torque response to  $\delta \theta$ .

effect on the overall joint torque with the 20mm perturbation

amplitude, it is considered as negligible. Furthermore,  $\tau_m$ 

can be considered as constant, as the visual feedback was not

updated during perturbation thus there is no reaction to the

perturbation (as shown in Fig.1b). The dynamics measured

Similar to the method described in [23], a constant displacement (as shown in Fig.1b) was used to identify stiffness K using:

$$\delta \tau \equiv K \,\delta \theta \,. \tag{3}$$

For each participant, leg, posture, force level, and perturbation direction condition, the perturbation displacement  $\delta\theta$  and resulting change of torque  $\delta\tau$  in the last 100ms of the perturbation plateau of all 10 trials formed 1x1000 vectors, which were used to estimate K as the least-square solution of Eq.3.

Viscosity was determined in a second step as the leastsquare solution of the transfer function (using Matlab *tfest* command with search method set 'auto' for best fit):

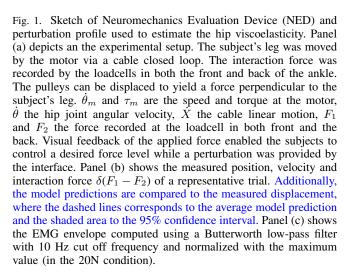
$$\frac{\Delta\Theta(s)}{\Delta T(s)} = \frac{1}{Is^2 + Bs + K},\tag{4}$$

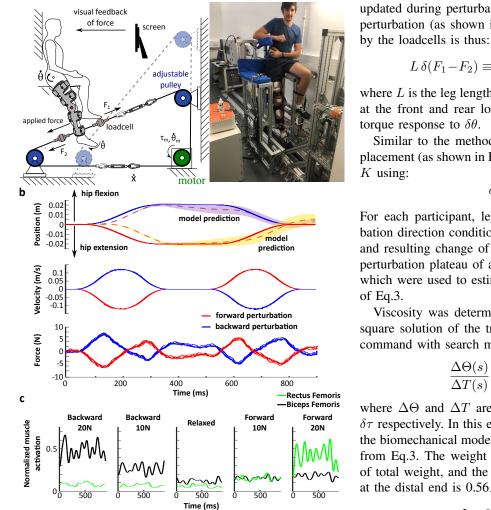
where  $\Delta\Theta$  and  $\Delta T$  are the Laplace transforms of  $\delta\theta$  and  $\delta\tau$  respectively. In this equation, inertia was computed from the biomechanical model of [24] and stiffness was estimated from Eq.3. The weight of the leg was estimated as 16.1% of total weight, and the radius of gyration of the whole leg at the distal end is 0.56*L*, thus 189

$$I = 0.161 M (L \, 0.56)^2, \tag{5}$$

with the mass M and length L parameters from Table 195 I. For each subject, the angle  $(\delta\theta)$  and torque data  $(\delta\tau)$ 196 over the whole perturbation period (900ms) for the five 197 trials corresponding to a specific (posture, force level and 198 perturbation direction) condition are used together to identify 199 viscosity using Eq.4. Instead of concatenating the data of all 200 five trials, it is "grouped" and defined as multi-experiment 201 data (using Matlab *merge*) to avoid potential prediction error 202 due to the transition period between two concatenated time 203 series data. 204

The quality of the identification was tested by predicting 205 the displacement from the force data and the identified 206 values for K, I, B. As can be seen in Fig.1b, the prediction 207 generally follows that actual displacement. The delay of 208 the predicted position probably stems from delays in the 209 force measurement due e.g. to cable compliance. The means 210 and standard deviations of the calculated coefficient of 211 determination  $(R^2)$  are listed in Table.II. 212





Neuromechanics Evaluation Device

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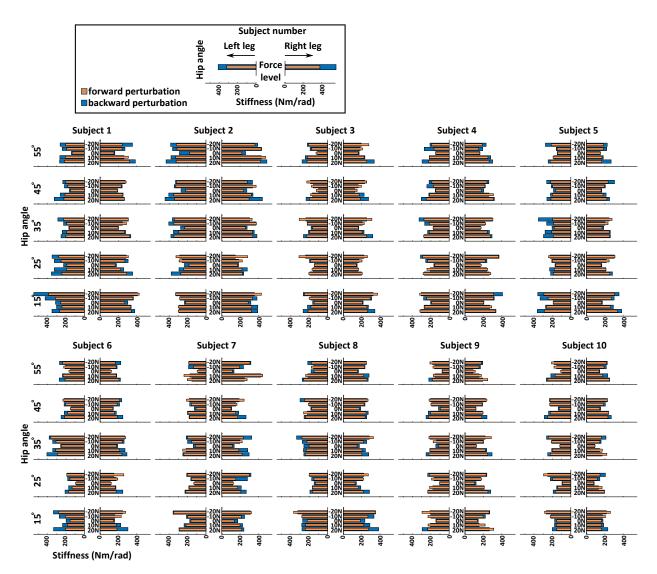


Fig. 2. Hip stiffness results for all subjects and conditions.

TABLE II Reliability of model prediction

subject number	mean R-square	SD of R-square
1	0.58	0.14
2	0.55	0.19
3	0.59	0.18
4	0.57	0.11
5	0.54	0.24
6	0.49	0.16
7	0.67	0.15
8	0.53	0.14
9	0.39	0.19
10	0.47	0.39

# III. RESULTS

Fig.2 summarizes the stiffness estimation results of all ten subjects. These results were obtained with the two perturbation directions, for their two legs, at the selected

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five postures and the five force levels. Hip joint overall 217 stiffness changes with the perturbation direction, applied 218 limb force level and hip angle (as was tested by separate 219 Friedman's tests with p<0.05). No difference was detected 220 between stiffness values in the dominant and non-dominant 221 legs (as was tested using both Friedman's test and paired t-222 test). In the following, we will investigate how stiffness and 223 viscosity depend on the perturbation direction, force level 224 and hip posture. 225

Perturbation direction dependency. Fig.3 shows how the 226 stiffness values of all subjects, at all postures and force 227 levels, depend on the perturbation direction. We see that a 228 larger portion of the stiffness values is below the identity 229 line, suggesting that the backward perturbation results in 230 larger stiffness values than the forward perturbation. This 231 was confirmed by a paired t-test indicating that the difference 232 between the estimation was different with the two different 233

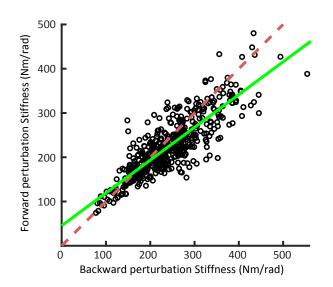


Fig. 3. Hip stiffness measurement depends on the perturbation direction. Each dot represents the stiffness at a specific subject leg, posture and force level, with stiffness measured with backward perturbation in the abscissa and with forward perturbation in the ordinate. The linear regression (green solid line) below the (dashed red) diagonal indicates larger values with perturbations in backward as in forward directions.

directions (p<0.05). The linear regression result (green solid line, with R<sup>2</sup>=0.72) described in Table III exhibits a difference of 26% between the estimation in the two directions. On the other hand, the estimated viscosity values showed no clear perturbation direction dependency, with regression close to identity line but R<sup>2</sup> <0.1 for the best linear regression model.

*Force-level dependency.* To investigate the relationship
between measured viscoelasticity, applied limb force level
and hip angle, we performed three steps of mixed effect
modelling to examine the stiffness change due to the selected
parameter. Firstly, stiffness was assumed to vary linearly
with applied limb force while posture may influence this
linear relation, modeled as:

where K are the stiffness values of one subject's leg mea-248 sured at n = 5 different hip angles and m = 6 force and 249 perturbation factors. m = 6 corresponds to 3 (either positive, 250 or negative) force levels, and 2 perturbation directions. 251  $\{\mathbf{X}, \boldsymbol{\beta}\}\$  are to capture the fixed effect and  $\{\mathbf{Z}, \boldsymbol{\mu}\}\$  to test 252 the influence of hip angle upon the identified force-stiffness 253 relation.  $\varepsilon$  is the error. By estimating mixed effect models 254 for each subject's leg, it was found that stiffness increases 255 monotonically with applied force amplitude in all subjects 256 (presented in Fig.4a). The estimated force-level dependency 257 weight  $(a_0)$  has a mean value of 5.15Nm/rad per applied 258 Newton force and a standard deviation of 0.98Nm/rad. This 259 finding indicates a positive relationship between applied limb 260 forces and hip joint stiffness, which is further confirmed by 261 F-tests (p<0.05 for all subjects' legs). 262

Furthermore, Friedman tests showed that the hip angle 263 would change both fixed-effect parameters, namely the re-264 laxed stiffness  $(a_1)$  and force-level dependency  $(a_0)$  (with 265 p=0.0006 and p<0.0001, respectively). To further emphasize 266 stiffness change due to hip angle, random effects are pre-267 sented as the relative percentage of fixed effects  $(b_{0i}/a_0)$  and 268  $b_{1i}/a_1$ ). Furthermore, the acquired percentages were further 269 subtracted by random effect percentages estimated at 55° 270 hip angle in order to present stiffness change with respect 271 to 55° hip angle. As shown in Fig.4d, relaxed stiffness  $(a_1)$ 272 changes with posture and reached statistically significance at 273  $15^{\circ}$  degree hip angle (tested with two tailed Wilcoxon rank 274 sum test with Bonferroni correction). On the other hand, 275 Fig.4c shows that the force-level dependency  $(a_0)$  changed 276 inconsistently due to posture and does not reach statistical 277 significance at any specific hip angle. 278

The same investigation was carried out on the estimated 279 viscosity. All subjects had an increased viscosity with ap-280 plied force (with a mean slope of 0.19Nm s/rad, presented 281 in Fig.4b). However, only 42% cases passed the F-test, 282 indicating that the viscosity change due to the applied limb 283 force may be insignificant. Additionally, the identified mixed 284 effect models showed low prediction accuracy and a limited 285 data variance explained by the model (with mean  $R^2=0.35$ 286 over the subjects lower than the stiffness model prediction 287 with  $R^2=0.79$ ) despite the inclusion of random effects. It is, 288 therefore, unclear whether hip joint viscosity exhibits similar 289 force-level dependency or whether the identified trend was 290 merely due to noise. 291

*Posture dependency.* To better catch the larger stiffness at 25° and 15°, a second investigation used a model assuming that stiffness changes quadratically with hip angle. We tested how the applied limb force may influence this quadratic relation by using the following model: 296

$$\mathbf{K} = \begin{bmatrix} \theta_{11}^2 & \theta_{11} & 1 \\ \vdots & \vdots & \vdots \\ \theta_{nm}^2 & \theta_{nm} & 1 \end{bmatrix} \begin{bmatrix} a_2 \\ a_3 \\ a_4 \end{bmatrix} + \mathbf{Z} \boldsymbol{\mu} + \boldsymbol{\varepsilon}$$
(7)  
$$\boldsymbol{\mu} \equiv \begin{bmatrix} b_{21} & b_{31} & b_{41} & b_{22} & b_{32} & b_{42} & \dots & b_{2n} & b_{3n} & b_{4n} \end{bmatrix}^T$$

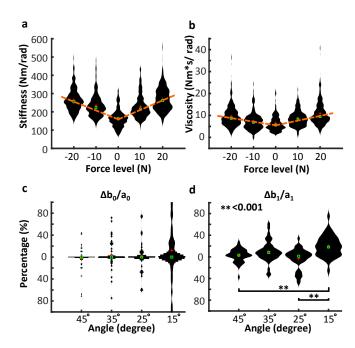


Fig. 4. Violin plots showing the probability density of force-level dependency and how it changes due to hip angle. The dashed lines indicate the least square fitted force dependency. Panels (a) and (b) show how hip stiffness and viscosity changes with the applied force. Panel (c) and panel (d) shows the influence of hip angle upon forcelevel dependency. The influence is presented as random effects ( $b_{0i}$ ) and  $b_{1i}$ ) and specifically in the percentage of fixed effects ( $a_0$  and  $a_1$ ). Additionally, it is presented as changes with respect to hip angle  $55^{\circ}$  in order to examine changes from a specific hip angle. Within each violin plots, a cross indicates the median value of the respective violin plot and a square the mean value. Random effects are found to change the identified force-level dependency  $(a_0)$  inconsistently and does not reach statistically significant at any hip angle. On the other hand, relaxed stiffness  $(a_1)$  is found to change with hip angle and confirmed to be statistical significant by two tailed Wilcoxon rank sum test and corrected by Bonferroni correction.

where K are the stiffness values measured at n = 5 different 297 force and m = 10 angle and perturbation factors. m = 10298 corresponds to 5 hip angles and 2 perturbation directions. 299 The random-effect design matrix  $\mathbf{Z}$  is built based on ele-300 ments of the fixed-effect design matrix X, and can be found 301 in the Appendix A.  $\mu$  is the random-effect vector indicating 302 the influence of the ith applied force on the stiffness and 303 hip angle relation, where  $\{b_{2i}\}$  show the influence on the 304 quadratic angle term and  $\{b_{3i}\}$  on the linear angle term, and 305  $\{b_{4i}\}$  demonstrate the constant effect. 306

The identified fixed effect parameters indicated that most 307 legs exhibit an inverse relationship between measured stiff-308 ness and hip angle, as presented in Table III in combination 309 with F-test results. In other words, it was found within our 310 experiment range  $\{15-55^\circ\}$  that hip joint stiffness would 311 increase with the decrease of hip angle. However, the 312 identified posture dependency was less influential compared 313 to the previously identified force-level dependency, as the 314

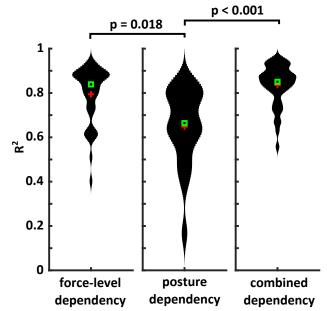


Fig. 5. Model prediction accuracy comparison. Prediction accuracy is presented as  $R^2$  and compared between all three models. It is shown that both models that considers force-level dependency performed a better prediction (tested with two tailed Wilcoxon rank sum test). On the other hand, the combined model improves estimation accuracy, however, did not reach a statistical significant (with p = 0.3579).

model without random effects showed a low estimation 315 accuracy (mean over the subjects  $\overline{R^2} = 0.16$ ) and required 316 random effects that consider applied limb forces (mean 317 over the subjects  $\overline{R^2}$  = 0.65). The importance of force-318 level dependency was consolidated by theoretical likelihood 319 tests (where 95% cases passed with p < 0.05), and suggested 320 that applied limb force is a stronger influencing factor in 321 comparison with hip angle. 322

The same process was repeated on estimated viscosity with Eq.7. The identified models showed poor prediction accuracy and explained limited variance of data (mean over the subjects  $\overline{R^2}$ =0.28) with 65% of the models failed the F-tests (indicating no posture dependency). 327

*Force and posture dependency.* Based on the aforementioned test results, we further hypothesised that stiffness changes according to both applied limb force and hip angle, with each factor possibly affecting the other one: 331

$$\mathbf{K} = \begin{bmatrix} F_{11} & \theta_{11}^{2} & \theta_{11} & 1 \\ \vdots & \vdots & \vdots & \vdots \\ F_{nm} & \theta_{nm}^{2} & \theta_{nm} & 1 \end{bmatrix} \begin{bmatrix} a_{0}' \\ a_{2}' \\ a_{3}' \\ a_{5} \end{bmatrix} + \mathbf{Z}_{1}\boldsymbol{\mu}_{1} + \mathbf{Z}_{2}\boldsymbol{\mu}_{2} + \boldsymbol{\varepsilon}$$
$$\boldsymbol{\mu}_{1} \equiv \begin{bmatrix} b_{01}' & b_{11}' & b_{02}' & b_{12}' & \dots & b_{0n}' & b_{1n}' \end{bmatrix}^{T}$$
(8)
$$\boldsymbol{\mu}_{2} \equiv \begin{bmatrix} b_{21}' & b_{31}' & b_{41}' & b_{22}' & b_{32}' & b_{42}' & b_{33}' & b_{43}' \end{bmatrix}^{T}$$

where K are the stiffness values measured at n = 5 different 332 angle and m = 6 force and perturbation factors. m = 6 333

corresponds to 3 (either positive, or negative) force levels 334 and 2 perturbation directions. The random-effect design 335 matrices  $\mathbf{Z}_1$  and  $\mathbf{Z}_2$  are built based on elements of the fixed-336 effect design matrix  $\mathbf{X}$  and can also be found in Appendix A. 337  $\mu_1$  and  $\mu_2$  are the random-effect vectors with  $\mu_1$  indicating 338 the influence of the *i*-th hip angle on the stiffness and force 339 relation, and  $\mu_2$  indicating the influence of applied force on 340 the angle relation. 341

Interestingly, the newly identified fixed effects exhibited 342 values similar to previous findings. Stiffness was again 343 found to increase with applied limb force, with slopes 344 (mean  $a'_0=4.98$ ) close to previous values (mean  $a_0=5.15$ ). 345 By calculating the differences between both values, 83% 346 cases showed differences less than 10% (calculated by 347  $(a'_0 - a_0)/a_0$ ). Meanwhile, most subjects were again found 348 to exhibit a negative relation between stiffness and hip angle, 349 and are presented in Table III along with F-test results. These 350 findings imply that the identified force-level and posture 351 dependencies coexist. 352

The estimated generalised linear models, which refers 353 to models without random effects, were shown to predict 354 hip joint stiffness of all subjects' legs with acceptable 355 variance being explained (mean over the subjects  $\overline{R^2}=0.68$ , 356 with standard deviation of 0.16). The model can be further 357 improved by including random effects (mean over the sub-358 jects  $\overline{R^2}$ =0.84, with standard deviation 0.09, 92.5% cases 359 passed F-tests). This finding demonstrates the importance 360 of correlation among parameters (e.g. hip angle changing 361 force-level dependency). On the other hand, random effects 362  $(b'_{1i} \text{ and } b'_{4i})$  which affect the constant value  $(a_5)$  are shown 363 to decrease since both posture and force-level dependencies 364 are considered in this model. 365

The model prediction accuracy of all three models is presented in Figure 5.

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### IV. DISCUSSION

We performed a systematic experimental investigation 369 of hip viscoelasticity using NED, a novel rigid robotic 370 interface dedicated to lower limb neuromechanics studies. 371 A position displacement was used as a mechanical per-372 turbation, that enabled us to obtain an accurate estimation 373 of hip stiffness. Viscosity was computed in a second step 374 using a least-square minimization of the linear second order 375 model. The relatively large perturbation amplitude ensured 376 a reliable estimation despite large force measurement noise. 377 We also analysed the influence of the leg, posture, force 378 level and perturbation direction on stiffness and viscosity 379 estimates. The dominant and non-dominant legs exhibited 380 similar values of viscoelasticity, which may not be surprising 38 as the legs are mostly used for the symmetric walking. 382 Sports activities such as playing football might induce some 383 asymmetry, although this could not be studied with the 384 available population. 385

 TABLE III

 STATISTICS OF LINEAR REGRESSION AND MIXED EFFECT MODELS

[						
	expected value	standard deviation				
Stiffness: perturbation direction dependency						
$Y = 0.74X + 45.29,  R^2 = 0.717$						
intercept	45.29	5.26				
slope	0.74	0.02				
Stiffness: force lev						
a <sub>0</sub> [m/rad]	5.15	0.98				
a <sub>1</sub> [Nm/rad]	169.39	39.61				
$b_{0i}/a_0$	0	19.93%				
b <sub>1<i>i</i></sub> /a <sub>1</sub>	0	14.24%				
Identified dependencies: 1	100% cases foun	d force-level dependency				
Stiffness: posture	e dependency (I	Eq.7, $\overline{R^2} = 0.84$ )				
a <sub>2</sub> [Nm/rad <sup>3</sup> ]	137.12	240.90				
a <sub>3</sub> [Nm/rad <sup>2</sup> ]	-212.96	303.65				
a <sub>4</sub> [Nm/rad]	302.75	104.06				
b <sub>2i</sub> /a <sub>2</sub>	0	21.41%				
b <sub>3i</sub> /a <sub>3</sub>	0	6.1%				
b <sub>4i</sub> /a <sub>4</sub>	0	14.19%				
Identified dependencies: 20% cases failed F-tests, showing no posture dependency 5% cases showed positive posture dependency 75% cases showed negative posture dependency						
Stiffness: posture and fo	rce-level depen	<b>dency</b> (Eq.8, $\overline{R^2} = 0.84$ )				
a <sub>0</sub> [m/rad]	4.98	1.34				
$a'_2$ [Nm/rad <sup>3</sup> ]	117.58	230.51				
$a'_{3}$ [Nm/rad <sup>2</sup> ]	-186.59	292.69				
a <sub>5</sub> [Nm/rad]	234.00	102.24				
$b'_{0i}/a'_0$	0	15.97%				
$b_{1i}'/a_5$	0	13.11%				
$b'_{2i}/a'_{2}$	0	31.46%				
$b_{3i}^{7}/a_{3}^{7}$	0	5.02%				
$b_{4i}^{\prime}/a_5$	0	0.74%				
Identified dependencies: 100% cases found force-level dependency 7.5% cases failed F-tests, showing no posture dependency 20% cases showed positive posture dependency 72.5% cases showed negative posture dependency						

Stiffness was found to be slightly larger when estimated 386 from displacement applied in the posterior direction than 387 in the anterior direction. This is probably due to stronger or 388 larger muscles since stiffness is known to vary proportionally 389 to the cross-sectional area of a stretched muscle [25], and 390 the quadriceps femoris may be larger than the biceps femoris 391 [26]. The study [8] estimated hip and knee multi-joint 392 viscoelasticity using an exoskeleton, but could not study 393 the influence of applied force systematically. Using the 394 dedicated NED interface, we could systematically analyse 395 the influence of posture and applied force on the single-joint 396 viscoelastic parameters in a controlled manner. We found 397 that stiffness increases monotonically with the applied limb 398 force, with a relation consistent with previous measurements 399 in the upper limb [1]. The stiffness value was found to 400 be slightly influenced by the hip angle, as was previously 401 found in the ankle [15]. The viscosity exhibited no clear 402

dependency upon perturbation direction or hip angle, and 403 slightly increases with the applied limb force. The difficulty 404 in identifying viscosity dependencies may originate from its 405 low value relative to stiffness. 406

The obtained viscoelasticity values we have observed with 407 our subjects population are in the same order as reported in 408 previous studies, although such comparison is limited by the 409 fact that viscoelasticity depends on the individuals. In [27], it 410 was found that knee joint stiffness in the relaxed condition is 411 around 75Nm/rad and viscosity is about 2Nm s/rad, and both 412 of these factors increase with muscle contraction. The values 413 we obtained for the hip joint are larger (with stiffness values 414 between 75-318Nm/rad and viscosity 2-21Nm s/rad under 415 relaxed condition), as expected as larger muscles are in-416 volved. Using the LOPES exoskeleton perturbing the whole 417 leg and indirect position measurement from the serial elastic 418 actuators used in LOPES, [8] found stiffness values between 419 50-220Nm/rad and viscosity between 0.5-10Nm s/rad. While 420 being in the same order of magnitude, the difference with 421 the values we have obtained may be in part due to the older 422 population of that study with ages between 67-72 while our 423 young adults were between 21-27. 424

#### REFERENCES

[1] E. Burdet, D. W. Franklin, and T. E. Milner, Human robotics : 426 neuromechanics and motor control. MIT Press, 2013. 427

425

- [2] N. Hogan, "The mechanics of multi-joint posture and movement 428 control." Biological Cybernetics, vol. 52, no. 5, pp. 315-31, 1985. 429
- K. Amankwah, R. J. Triolo, and R. F. Kirsch, "Effects of spinal [3] 430 cord injury on lower-limb passive joint moments revealed through 431 a nonlinear viscoelastic model." Journal of Rehabilitation Research 432 433 and Development, vol. 41, no. 1, pp. 15-32, 2004.
- [4] P. J. Sinclair, G. M. Davis, and R. M. Smith, "Musculo-skeletal 434 modelling of NMES-evoked knee extension in spinal cord injury," 435 436 Journal of Biomechanics, vol. 39, no. 3, pp. 483-92, 2006.
- [5] K. Perell, A. Scremin, O. Scremin, and C. Kunkel, "Quantifying mus-437 cle tone in spinal cord injury patients using isokinetic dynamometric 438 439 techniques." Paraplegia, vol. 34, no. 1, pp. 46-53, 1996.
- [6] M. N. Akman, R. Bengi, M. Karatas, S. Kilinç, S. Sözay, and 440 R. Ozker, "Assessment of spasticity using isokinetic dynamometry 441 in patients with spinal cord injury." Spinal Cord, vol. 37, no. 9, pp. 442 638-43, 1999. 443
- L. Lünenburger, G. Colombo, R. Riener, and V. Dietz, "Clinical as-444 [7] sessments performed during robotic rehabilitation by the gait training 445 446 robot Lokomat," in Proceedings of the IEEE International Conference on Rehabilitation Robotics, 2005, pp. 345-8. 447
- [8] B. Koopman, E. H. F. van Asseldonk, and H. van der Kooij, "Estima-448 tion of human hip and knee multi-joint dynamics using the LOPES 449 gait trainer," IEEE Transactions on Robotics, vol. 32, no. 4, pp. 920-450 32, 2016. 451
- [9] J. Meuleman, E. Van Asseldonk, G. Van Oort, H. Rietman, and H. Van 452 453 Der Kooij, "LOPES II - Design and evaluation of an admittance controlled gait training robot with shadow-leg approach," IEEE Trans-454 455 actions on Neural Systems and Rehabilitation Engineering, vol. 24, 456 no. 3, pp. 352-63, 2016.
- [10] I. Farkhatdinov, J. Ebert, G. Van Oort, M. Vlutters, E. Van Asseldonk, 457 and E. Burdet, "Assisting human balance in standing with a robotic 458 exoskeleton," IEEE Robotics and Automation Letters, vol. 4, no. 2, 459 460 2019
- H. Lee, P. Ho, M. A. Rastgaar, H. I. Krebs, and N. Hogan, "Mul-[11] 461 462 tivariable static ankle mechanical impedance with relaxed muscles," Journal of Biomechanics, vol. 44, no. 10, pp. 1901-08, 2011. 463

- [12] E. J. Rouse, L. J. Hargrove, E. J. Perreault, and T. A. Kuiken, 464 "Estimation of human ankle impedance during the stance phase of 465 walking," IEEE Transactions on Neural Systems and Rehabilitation 466 Engineering, vol. 22, no. 4, pp. 870-8, 2014. 467 468
- Farkhatdinov, [13] H.-Y. Huang, I. A. Arami, M. Bouri. "Cable-driven robotic interface for and E. Burdet, lower 469 limb neuromechanics identification," 2019. [Online]. Available: 470 http://arxiv.org/abs/1908.02689
- [14] H.-Y. Huang, "Development of the neuromechanics evaluation device (NED) for subject-specific lower limb modelling of spinal cord injury," Ph.D. dissertation, Imperial College London, 2019.
- [15] M. Mirbagheri, H. Barbeau, and R. Kearney, "Intrinsic and reflex contributions to human ankle stiffness: variation with activation level and position," Experimental Brain Research, vol. 135, no. 4, pp. 423-36, 2000.
- [16] S. Pfeifer, H. Vallery, M. Hardegger, R. Riener, and E. J. Perreault, "Model-Based Estimation of Knee Stiffness," IEEE Transactions on Biomedical Engineering, vol. 59, no. 9, pp. 2604-12, 2012.
- [17] D. Ludvig, M. Plocharski, P. Plocharski, and E. J. Perreault, "Mechanisms contributing to reduced knee stiffness during movement," Experimental Brain Research, vol. 235, no. 10, pp. 2959-70, 2017.
- [18] E. Perreault, R. Kirsch, and P. Crago, "Effects of voluntary force generation on the elastic components of endpoint stiffness," Experimental Brain Research, vol. 141, no. 3, pp. 312-23, 2001.
- [19] E. J. Perreault, R. F. Kirsch, and P. E. Crago, "Voluntary control of static endpoint stiffness during force regulation tasks," Journal of Neurophysiology, vol. 87, no. 6, pp. 2808-16, 2002.
- [20] H. Gomi and M. Kawato, "Equilibrium-point control hypothesis examined by measured arm stiffness during multijoint movement," Science, vol. 272, no. 5258, pp. 117-120, 1996.
- [21] H. Lee and N. Hogan, "Time-Varying Ankle Mechanical Impedance During Human Locomotion," IEEE Transactions on Neural Systems and Rehabilitation Engineering, vol. 23, no. 5, pp. 755-64, 2015.
- [22] H. Lee, E. J. Rouse, and H. I. Krebs, "Summary of Human Ankle Mechanical Impedance During Walking," IEEE Journal of Translational Engineering in Health and Medicine, vol. 4, pp. 1-7, 2016.
- [23] E Burdet R Osu D W Franklin T Yoshioka T E Milner and M. Kawato, "A method for measuring endpoint stiffness during multijoint arm movements." Journal of Biomechanics, vol. 33, no. 12, pp. 1705-9 2000
- [24] D. Winter, "Human balance and posture control during standing and walking," Gait and Posture, vol. 3, no. 4, pp. 193-214, 1995.
- [25] R. V. Gonzalez, T. S. Buchanan, and S. L. Delp, "How muscle architecture and moment arms affect wrist flexion-extension moments," Journal of Biomechanics, vol. 30, no. 7, pp. 705-12, 1997
- [26] T. L. Wickiewicz, R. R. Roy, P. L. Powell, and V. R. Edgerton, "Muscle architecture of the human lower limb." Clinical Orthopaedics and Related Research, no. 179, pp. 275-83, 1983
- L. Q. Zhang, G. Nuber, J. Butler, M. Bowen, and W. Z. Rymer, "In vivo human knee joint dynamic properties as functions of muscle contraction and joint position." Journal of Biomechanics, vol. 31, no. 1, pp. 71-6, 1998.

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Appendix

517 The full expansion of Eq.7 and Eq.8 are, respectively:

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$$\mathbf{K} = \mathbf{X} \boldsymbol{\beta} + \mathbf{Z} \boldsymbol{\mu} + \boldsymbol{\varepsilon}$$
(9)  
$$\begin{bmatrix} K_{11} \\ \vdots \\ K_{1,10} \\ K_{21} \\ \vdots \\ K_{2,10} \\ \vdots \\ K_{51} \\ \vdots \\ K_{5,10} \end{bmatrix} = \begin{bmatrix} \theta_{11}^2 & \theta_{11} & 1 \\ \vdots & \vdots & \vdots \\ \theta_{1,10}^2 & \theta_{1,10} & 1 \\ \theta_{21}^2 & \theta_{21} & 1 \\ \vdots & \vdots & \vdots \\ \theta_{21}^2 & \theta_{21} & 1 \\ \vdots & \vdots & \vdots \\ \theta_{21}^2 & \theta_{2,10} & 1 \\ \vdots & \vdots & \vdots \\ \theta_{2,10}^2 & \theta_{2,10} & 1 \\ \vdots & \vdots & \vdots \\ \theta_{2,10}^2 & \theta_{2,10} & 1 \\ \vdots & \vdots & \vdots \\ \theta_{21}^2 & \theta_{2,10} & 1 \\ \vdots & \vdots & \vdots \\ \theta_{21}^2 & \theta_{21} & 0 \\ \theta_{1,10} & 1 & 0 & 0 & 0 \\ 0 & 0 & 0 & \theta_{21}^2 & \theta_{21} & 1 \\ \vdots & \theta_{21}^2 & \theta_{21} & 1 \\ \theta_{1,10}^2 & \theta_{1,10} & 1 \\ \theta_{1,10} & 0 & 0 & 0 \\ 0 & 0 & 0 & \theta_{21}^2 & \theta_{21} & 1 \\ \theta_{1,10}^2 & \theta_{2,10} & 1 \\ \theta_{1,10}^2 & \theta_{2,10} & 0 \\ \theta_{1,10} & \theta_{1,10} & 1 \\ \theta_{1,10} & \theta_{1,10} & 0 \\ \theta_{1,10} & \theta_{1,10} & \theta_{1,10} & \theta_{1,10} & \theta_{1,10} & \theta_{1,10} \\ \theta_{1,10} & \theta_{1,10} & \theta_{1,10} & \theta_{1,10} & \theta_{1,10} \\ \theta_{1,10} & \theta_{1,10} & \theta_{1,10} & \theta_{1,10} & \theta_{1,10} \\ \theta_{1,10} & \theta_{1,10} & \theta_{1,10} & \theta_{1,10} & \theta_{1,10} & \theta_{1,10} \\ \theta_{1,10} & \theta_{1,10} & \theta_{1,10} & \theta_{1,10} & \theta_{1,10} & \theta_{1,10} \\ \theta_{1,10} & \theta_{1,10} & \theta_{1,10} & \theta_{1,10} & \theta_{1,10} & \theta_{1,10} & \theta_{1,10} \\ \theta_{1,10} & \theta_{1,10$$

 $\mathbf{K} = \mathbf{X}\boldsymbol{\beta} + \mathbf{Z}_1\boldsymbol{\mu}_1 + \mathbf{Z}_2\boldsymbol{\mu}_2 + \boldsymbol{\varepsilon}$ 

(10)

$\mathbf{N} = \mathbf{X}\boldsymbol{\rho} + \mathbf{Z}_1\boldsymbol{\mu}_1 + \mathbf{Z}_2\boldsymbol{\mu}_2 + \boldsymbol{\varepsilon} \tag{1}$							
$ \begin{bmatrix} K_{11} \\ \vdots \\ K_{16} \\ K_{21} \\ \vdots \\ K_{26} \\ \vdots \\ K_{51} \end{bmatrix} = \begin{bmatrix} F_{11} \theta_{11}^2 \theta_{11} 1 \\ \vdots & \vdots & \vdots \\ F_{16} \theta_{16}^2 \theta_{16} 1 \\ F_{21} \theta_{21}^2 \theta_{21} 1 \\ \vdots & \vdots & \vdots \\ F_{26} \theta_{26}^2 \theta_{26} 1 \\ \vdots & \vdots & \vdots \\ F_{51} \theta_{51}^2 \theta_{51} 1 \\ \vdots & \vdots & \ddots \end{bmatrix} \begin{bmatrix} a'_0 \\ a'_2 \\ a'_3 \\ a'_4 \end{bmatrix} + $	$\begin{bmatrix} F_{11} 1 & 0 & 0 & \dots & 0 & 0 \\ \vdots & \vdots & & & & \\ F_{16} 1 & 0 & 0 & & & \\ 0 & 0 & F_{21} 1 & & \vdots & & \\ \vdots & & & & & \\ F_{26} 1 & & & & \\ \vdots & 0 & 0 & \ddots & 0 & 0 \end{bmatrix} \begin{bmatrix} b'_{01} \\ b'_{11} \\ b'_{02} \\ b'_{12} \\ \vdots \\ b'_{1} \end{bmatrix} + \begin{bmatrix} b'_{01} \\ b'_{11} \\ b'_{02} \\ b'_{12} \\ \vdots \\ b'_{1} \end{bmatrix} + \begin{bmatrix} b'_{01} \\ b'_{11} \\ b'_{02} \\ b'_{12} \\ \vdots \\ b'_{1} \end{bmatrix} + \begin{bmatrix} b'_{01} \\ b'_{11} \\ b'_{02} \\ b'_{12} \\ \vdots \\ b'_{1} \end{bmatrix} + \begin{bmatrix} b'_{01} \\ b'_{11} \\ b'_{02} \\ b'_{12} \\ \vdots \\ b'_{1} \end{bmatrix} + \begin{bmatrix} b'_{01} \\ b'_{11} \\ b'_{02} \\ b'_{12} \\ \vdots \\ b'_{1} \end{bmatrix} + \begin{bmatrix} b'_{01} \\ b'_{11} \\ b'_{12} \\ \vdots \\ b'_{12} \\ \vdots \\ b'_{12} \end{bmatrix} + \begin{bmatrix} b'_{01} \\ b'_{11} \\ b'_{12} \\ \vdots \\ b'_{12} \\ \vdots \\ b'_{12} \\ \vdots \\ b'_{12} \end{bmatrix} + \begin{bmatrix} b'_{01} \\ b'_{11} \\ b'_{12} \\ \vdots \\ b'_{12} \\ b'_{12} \\ \vdots \\ b'_{12} \\ \vdots \\ b'_{12} \\ \vdots \\ $	$ \begin{array}{c} \left( \begin{array}{c} 10 \right) \\ \theta_{11}^{2} \theta_{11} 1 & 0 & 0 & 0 & 0 & 0 & 0 \\ \theta_{12}^{2} \theta_{12} 1 & 0 & 0 & 0 & \vdots \\ 0 & 0 & 0 & \theta_{13}^{2} \theta_{13} 1 & \vdots \\ \vdots & \theta_{14}^{2} \theta_{14} 1 & 0 & 0 & 0 \\ \vdots & 0 & 0 & 0 & \theta_{15}^{2} \theta_{15} 1 \\ \theta_{21}^{2} \theta_{21} 1 & 0 & 0 & 0 & 0 & 0 \\ \theta_{22}^{2} \theta_{22} 1 & 0 & 0 & 0 & 0 & 0 \\ \theta_{22}^{2} \theta_{22} 1 & 0 & 0 & 0 & 0 & 0 \\ \end{array} \right) + \varepsilon $					
	$\begin{array}{cccc} \vdots & F_{51} 1 \begin{bmatrix} b_{15} \end{bmatrix} \\ \vdots & \vdots \\ 0 & 0 & 0 \end{bmatrix}$	$\begin{bmatrix} b_{33} \\ b_{43} \end{bmatrix} \\ 0 & 0 & 0 & 0 & 0 & 0 & \theta_{55}^2 \\ 0 & 0 & 0 & 0 & 0 & \theta_{56}^2 \\ 0 & 0 & 0 & 0 & 0 & \theta_{56}^2 \\ \theta_{56} & \theta_{56} & 1 \end{bmatrix} \begin{bmatrix} b_{33} \\ b_{43} \end{bmatrix}$					