

Supplementary Materials

Algorithm 1 Implementation of the adaptive model

Recording and pre-processing

- Record resting data and training data
- Calibrate activity detection
- Initialize inducing points using k-Means and train sGPs w.r.t. their hyperparameters σ_n, σ_s, l

Prediction steps

Data: WL EMG-feature of eight electrodes
Result: Velocity vector \mathbf{v}^*

- 1: **while** New data points arrive **do**
- 2: **if** Activity detected **then**
- 3: Computation of \mathbf{v}^*
- 4: Add sample \mathbf{x}^* to active batch (X^*)
- 5: **if** size of $X^* > k^*$ **then**
- 6: Pass X^* to the *update branch*
- 7: **else**
- 8: $\mathbf{v}^* = \mathbf{0}$
- 9: Add sample to resting batch (X^o)
- 10: **if** size of $X^o > k^o$ **then**
- 11: Pass X^o to the *update branch*
- 12: **if** No activity detected for t seconds **then**
- 13: Reset active batch

Update rules

	δ	ϵ	$\bar{\sigma}^{*2}$	$\bar{\mathbf{v}}^*$	label
1)	-	+	$-, (> \lambda_{\sigma 1})$	$v_1 > v_2$	class 1
2)	-	-	$++, (> \lambda_{\sigma 1})$	$v_1 > v_2$	class 1
3)	+	o	$o, (> \lambda_{\sigma 2})$	$v_1 \gg v_2$	class 1
4)	+	o	o	$v_1 \approx v_2$	diag. label
5)	(-)		(++)		
6)		(+)	(+)		

Here, - indicates small values, o medium values, + large values and ++ is defined as almost one. $\lambda_{\sigma 1} := 0.2$ and $\lambda_{\sigma 2} := 0.3$. An inducing point is added to X_u at $\tilde{\mathbf{x}}^*$ if all conditions in () in one row are fulfilled.

Update steps

Data: Active batch or resting batch
Result: Modified data structure of the sGPs

- 1: **if** X^* is passed **then**
- 2: Calculate $\tilde{\mathbf{x}}^*, \bar{\mathbf{v}}^*$ and $\bar{\sigma}^{*2}$
- 3: Calculate δ and ϵ using $\tilde{\mathbf{x}}^*$
- 4: Pull $\mu_{1,j+1}^m$ towards $\tilde{\mathbf{x}}^*$
- 5: **if** $\delta > 0.55$ **then**
- 6: Pull $\mu_{2,j+1}^m$ towards $\tilde{\mathbf{x}}^*$
- 7: **if** Update rule is fulfilled **then**
- 8: Add rand subset of X^* to \mathcal{D}_d and remove rand subset of \mathcal{D}_d with equivalent label
- 9: **if** Update rule to update X_u is fulfilled **then**
- 10: Add $\tilde{\mathbf{x}}^*$ to X_u of all sGPs
- 11: **if** Number of updates $> t_{u1}$ **or** update rule 5 or 6 **then**
- 12: Optimize X_u of each sGP
- 13: **if** Number of updates $> t_{u2}$ **then**
- 14: Recalculate μ_c^0 and Σ_c^0 using \mathcal{D}_{sub}
- 15: **else if** X^o is passed **then**
- 16: Add rand subset of X^o to \mathcal{D}_d and remove rand subset of \mathcal{D}_d
- 17: **if** Data structure of sGPs is modified **then**
- 18: Recomputation of model structures within the library which are effected by changes of \mathcal{D}_d or X_u
- 19: Overwrite adapted model structures in *prediction branch*

Parameterization

Parameter	Value	Parameter	Value
k^*	50	$\alpha 1$	0.1
k^o	1000	$\alpha 2$	0.025
t	0.1	$\lambda_{\sigma 1}$	0.2
t_{u1}	15	$\lambda_{\sigma 2}$	0.3
t_{u2}	20		

Table 1: Results of the Friedman test

Friedman:	Chi-Square Value (χ_R^2)(2)	p-Value (p)	Effect Size (ω)
Position 1	33.917	< 0.0001	0.485
Position 2*	64.825	< 0.0001	0.671
Position 2**	104.039	< 0.0001	0.850
Main Axes	68.474	< 0.0001	0.563
Diagonal Axes	123.293	< 0.0001	0.756

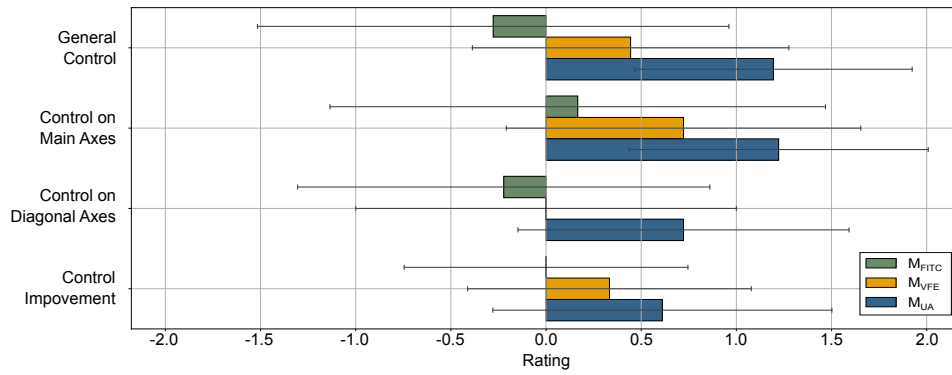


Figure 1: Results of the questionnaire on the satisfaction of the control with the range from -2.5 to +2.5. Subjects answered the questionnaire at the end of each test session.

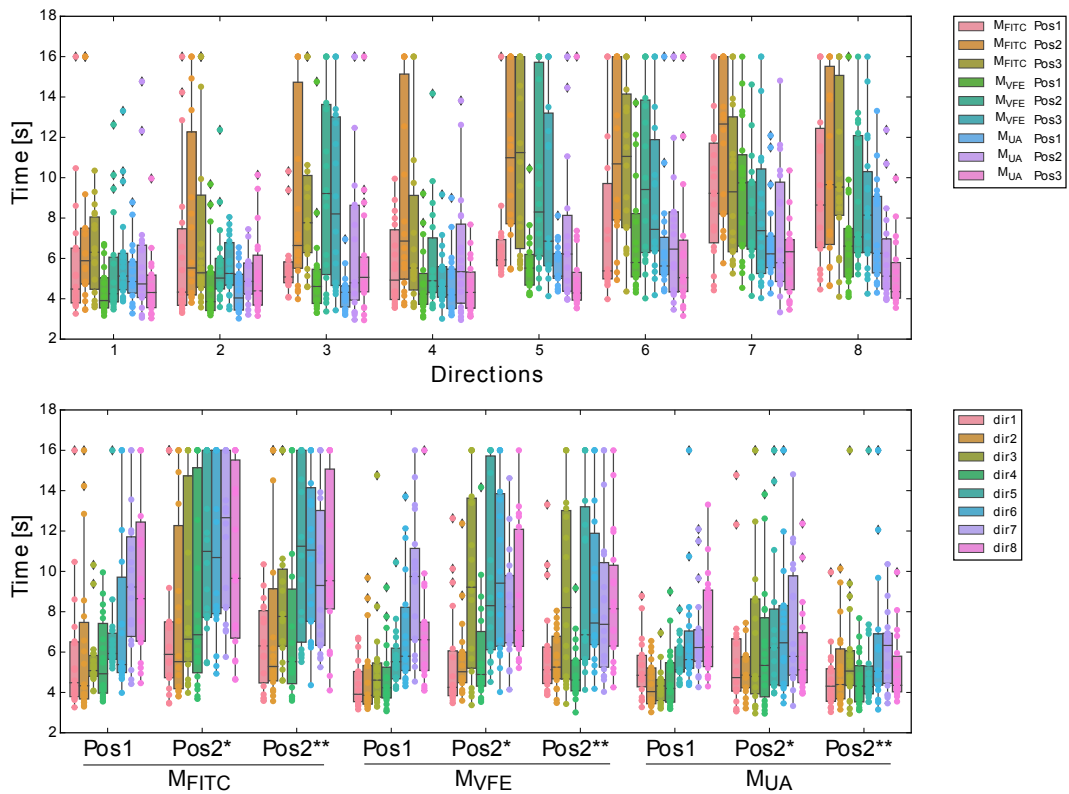


Figure 2: Results of the Fitts analysis; upper part: results for the different directions, direction 1-4 refer to the main axes and direction 5-8 to the diagonal axes; lower part: results for the different positions.

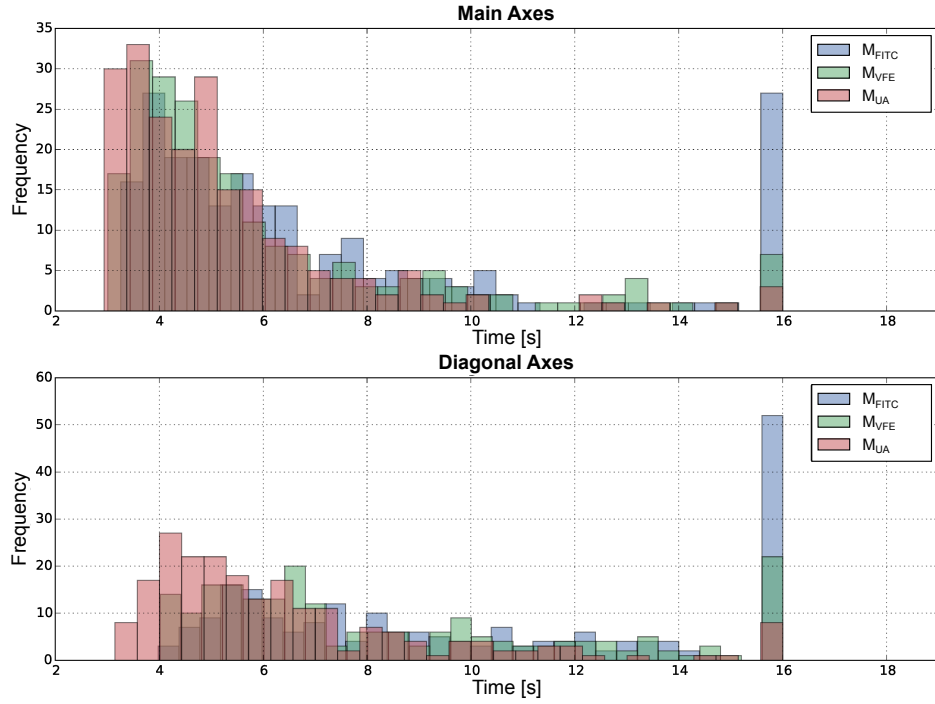


Figure 3: Distribution of time with respect to navigation along the main axes (top subplot) or diagonal axes (bottom subplot). Note that a time of 16s corresponds to failed trials.

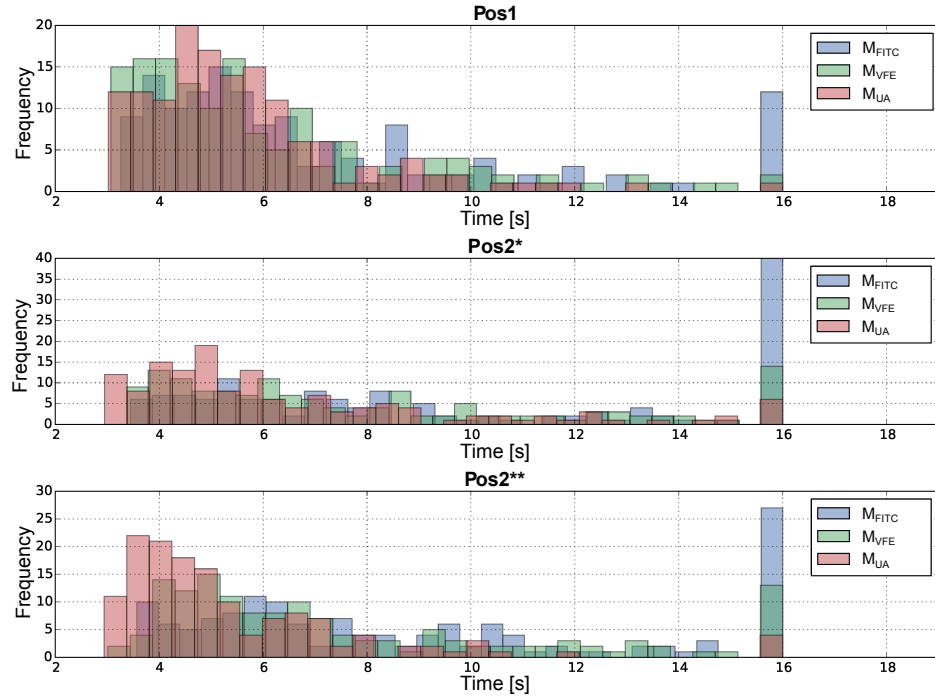


Figure 4: Distribution of time with respect to the different positions. Note that a time of 16s corresponds to failed trials.