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Motor unit behaviour in voluntary and evoked contractions

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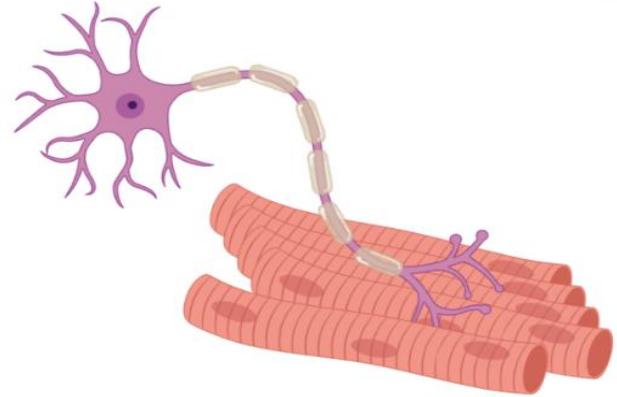
Loughborough
University



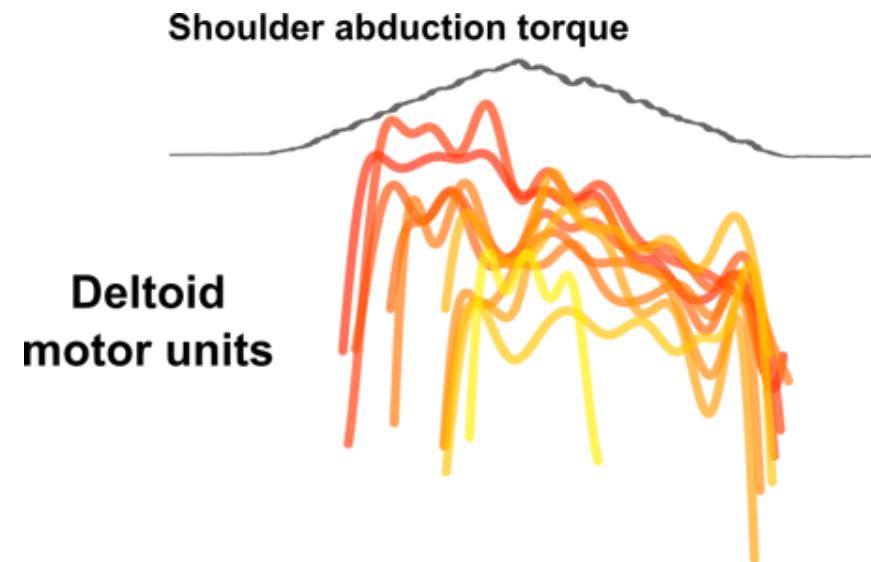
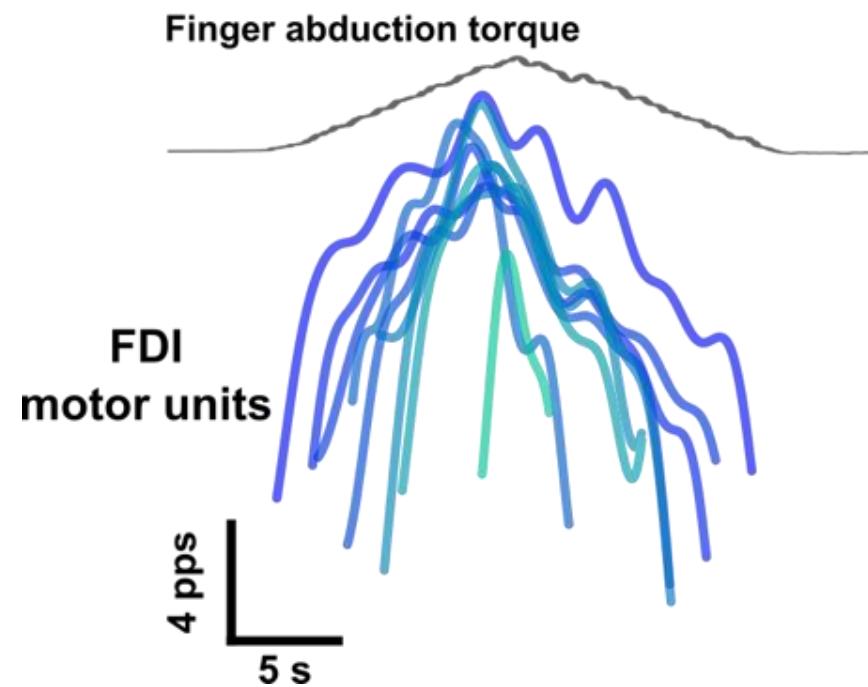
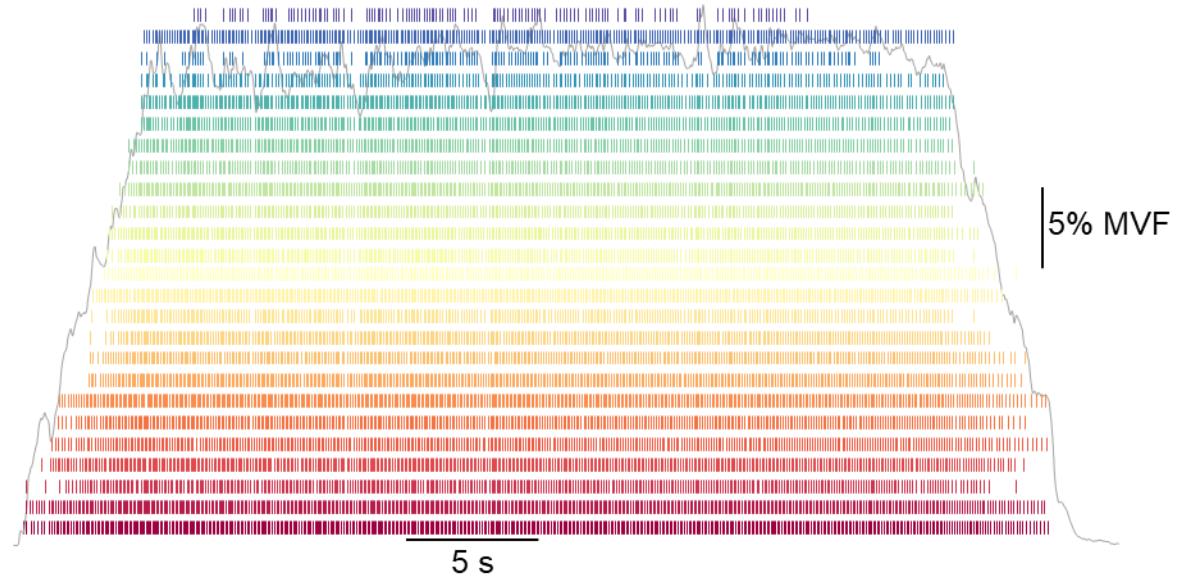
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VERSUS
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TURNING
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"Final common pathway"
(Sherrington, 1925)

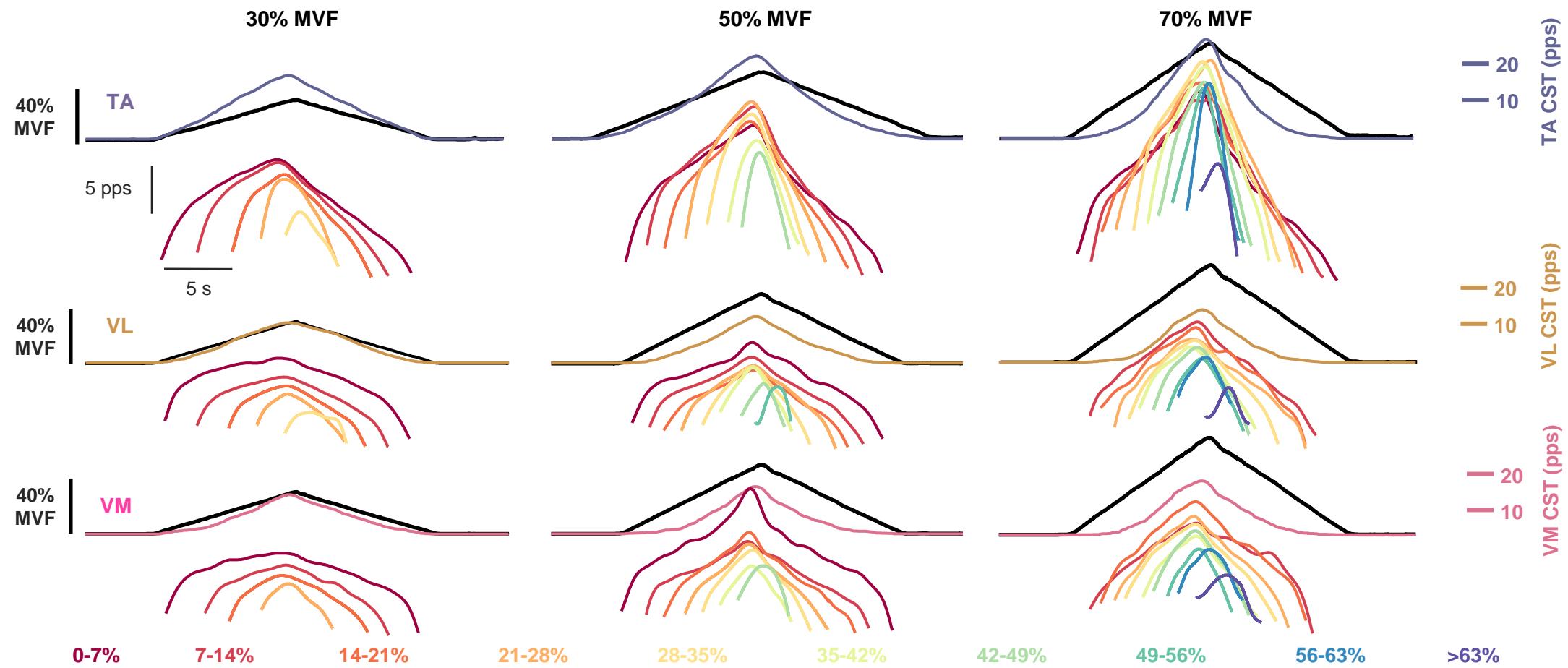


Figures courtesy of Dr Pearcey (Memorial, CAN / Northwestern, US)

PART 1

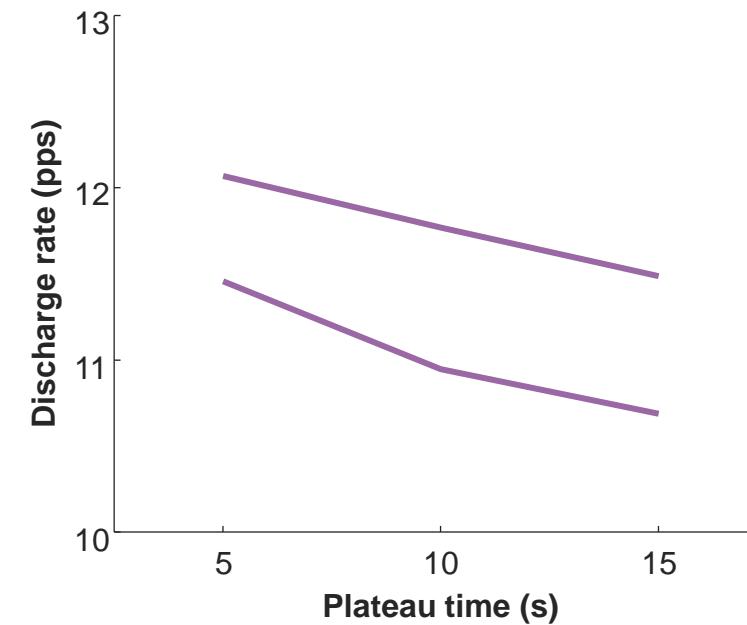
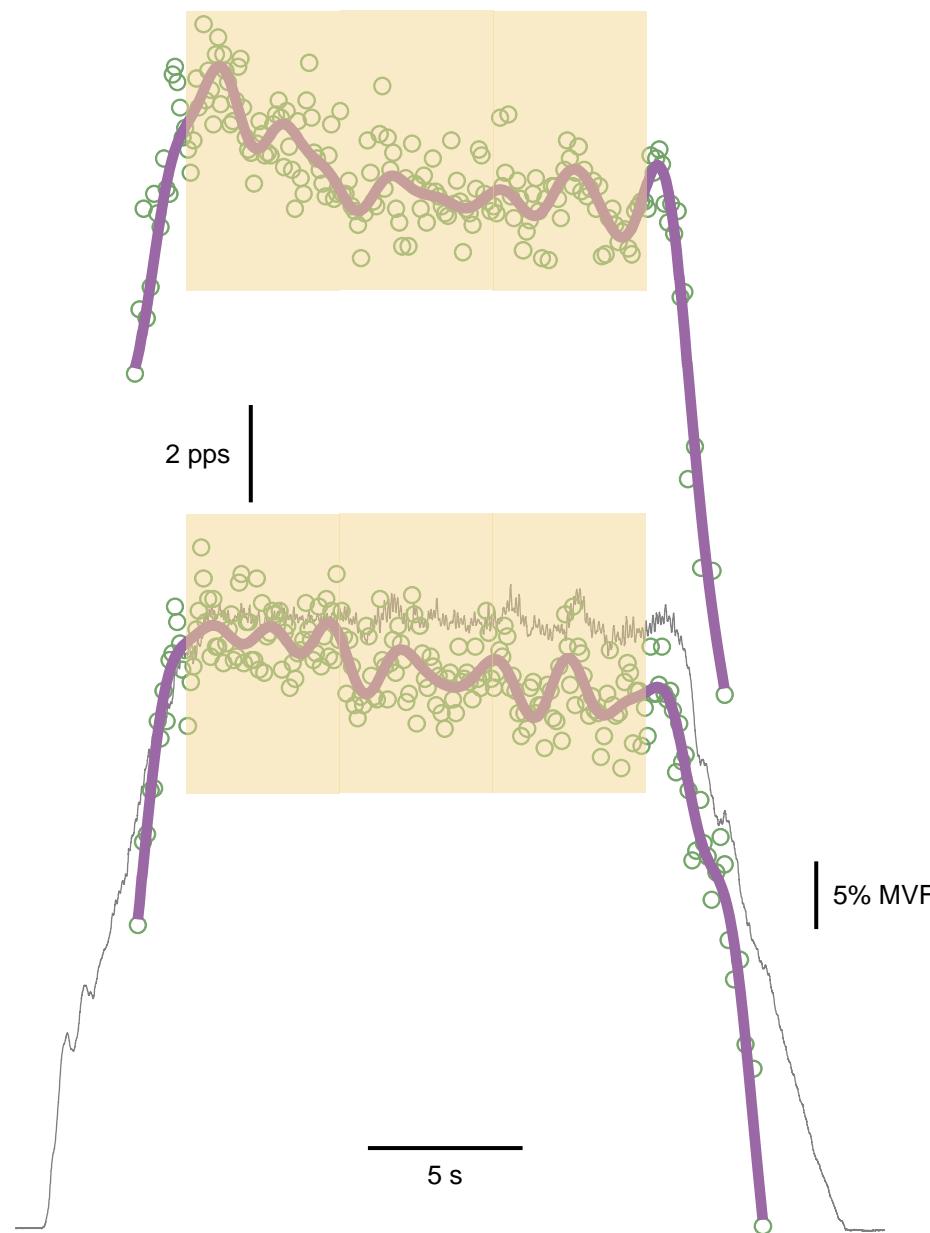
- Considerations when quantifying motor unit discharge rate during voluntary (isometric) contractions
- Different types of inputs and their effect on motor unit discharge behaviour
- Quantification of motor unit discharge behaviour in different conditions and populations

The relationship between motor unit recruitment threshold and firing rate

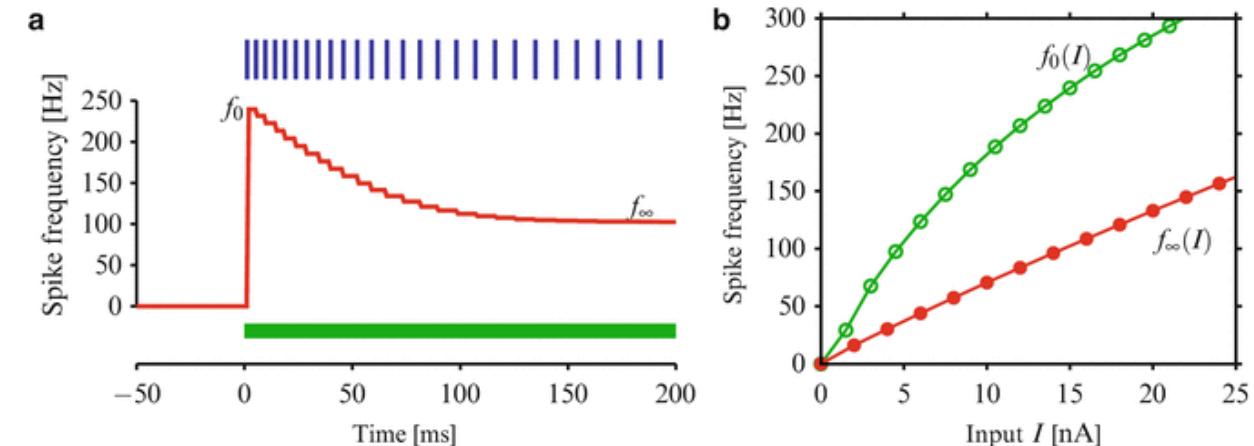


Discharge rate $\sim 1 + \text{Contraction} * \text{Muscle} + \text{Recruitment threshold} + (1 | \text{PID})$

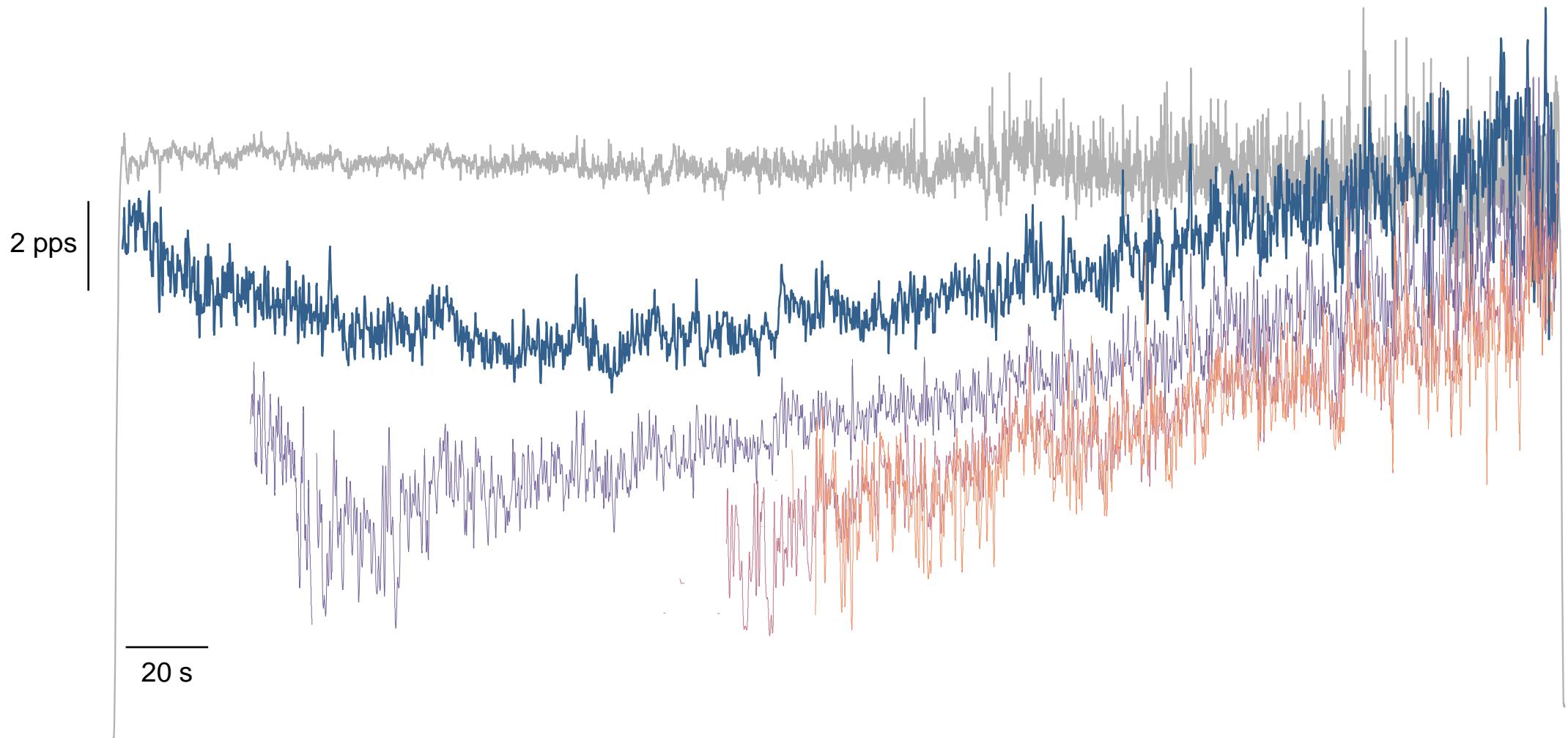
Quantifying motor unit discharge rate

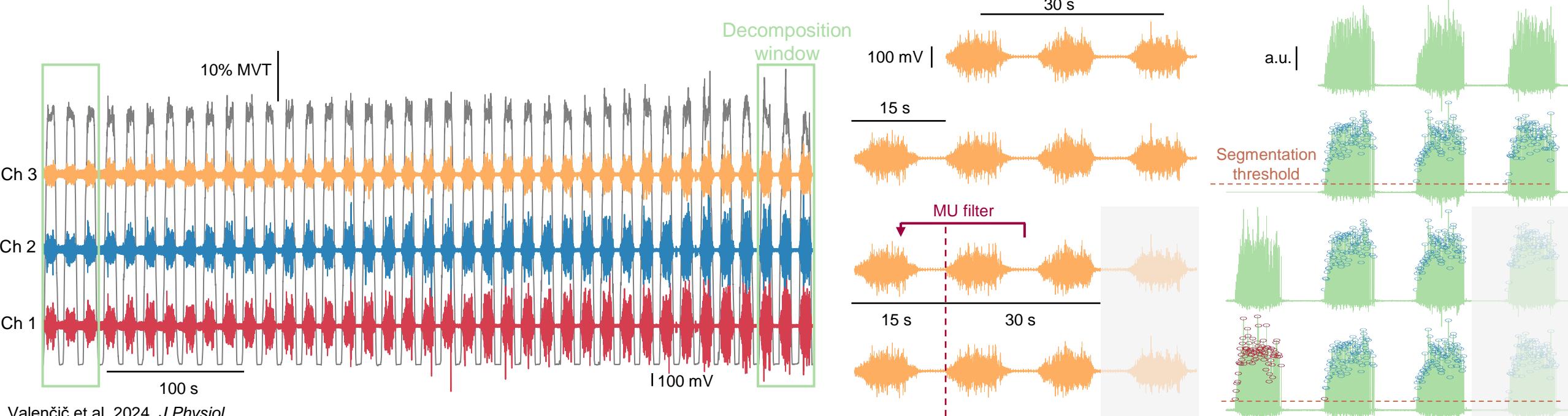
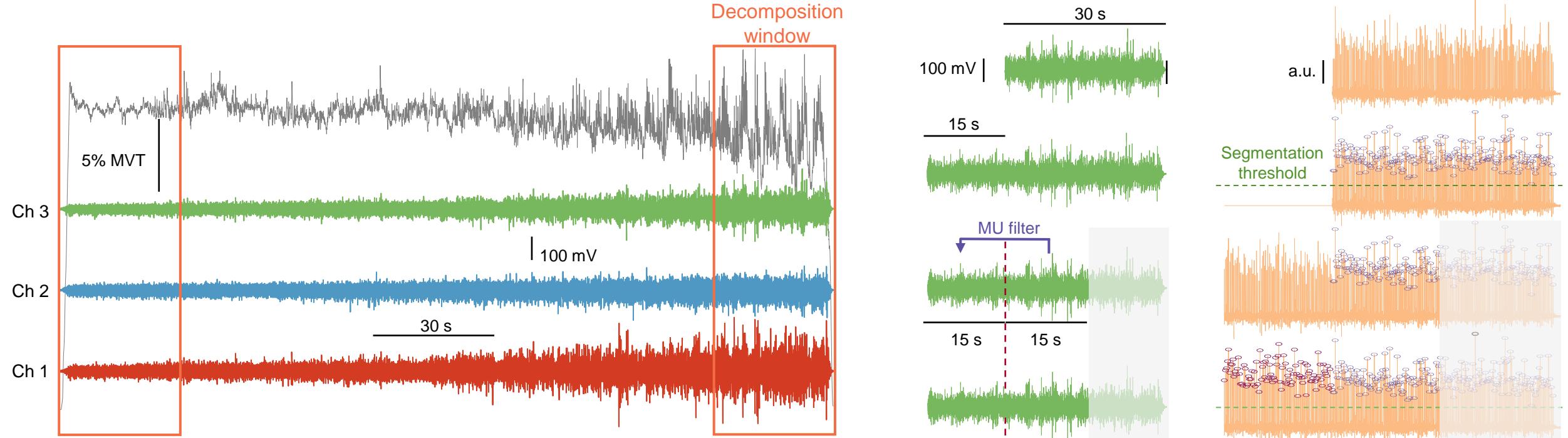


Spike frequency adaptation

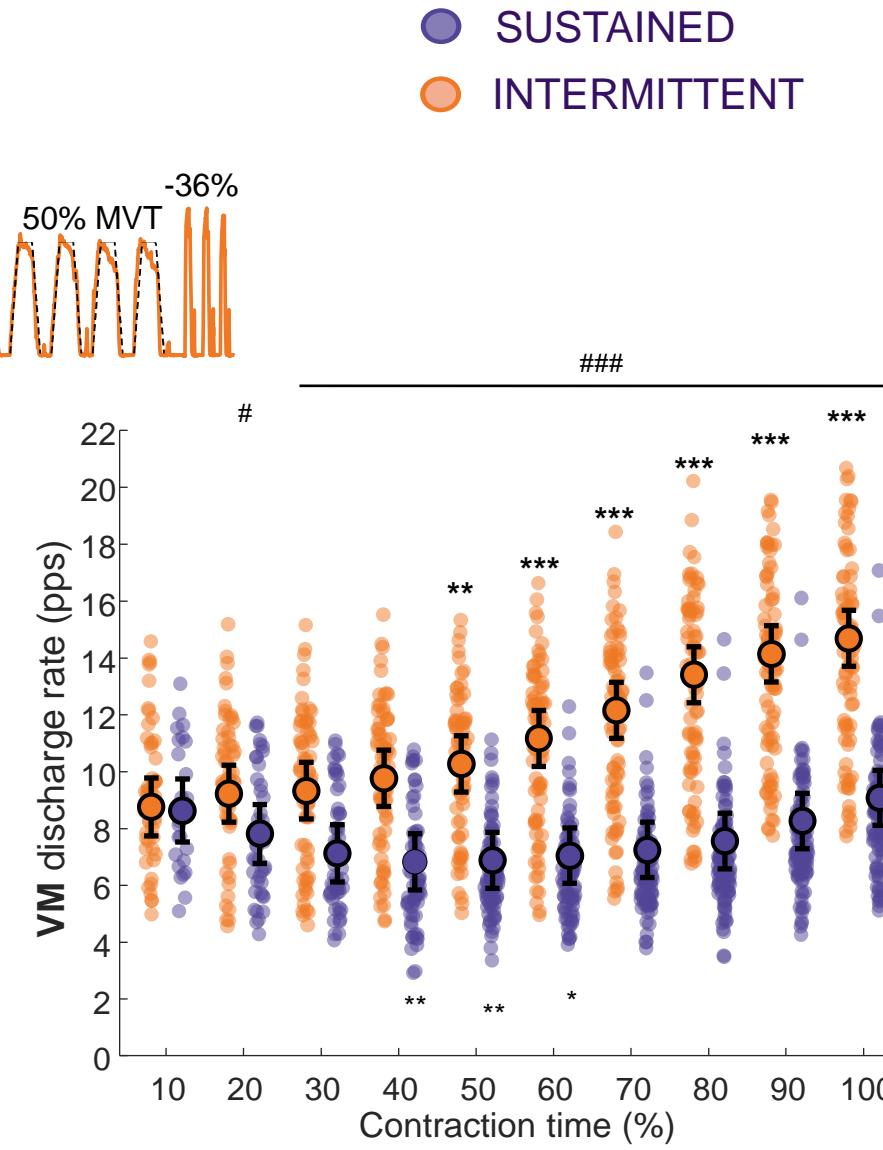
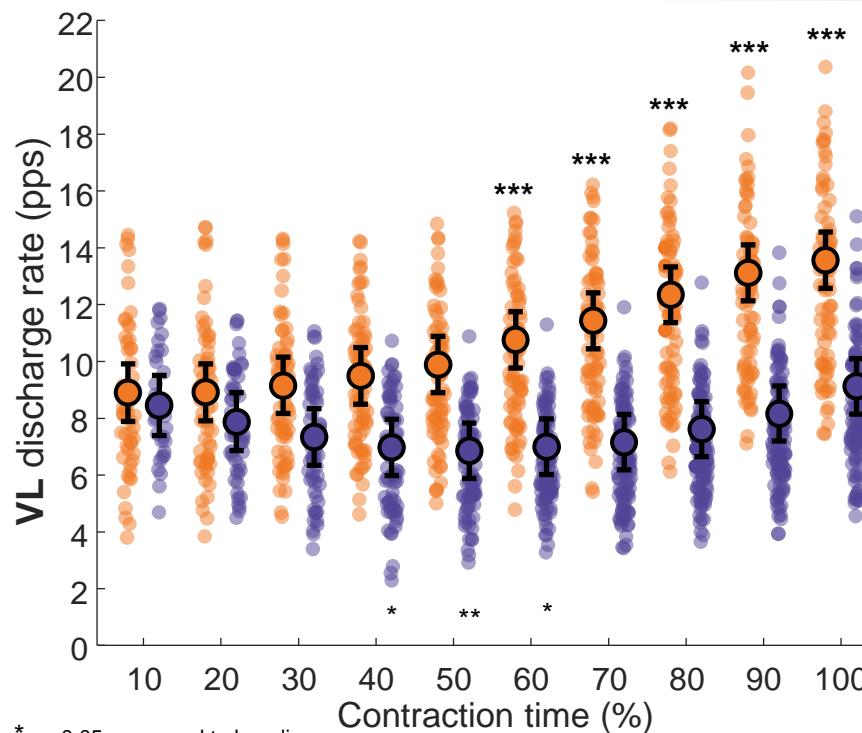
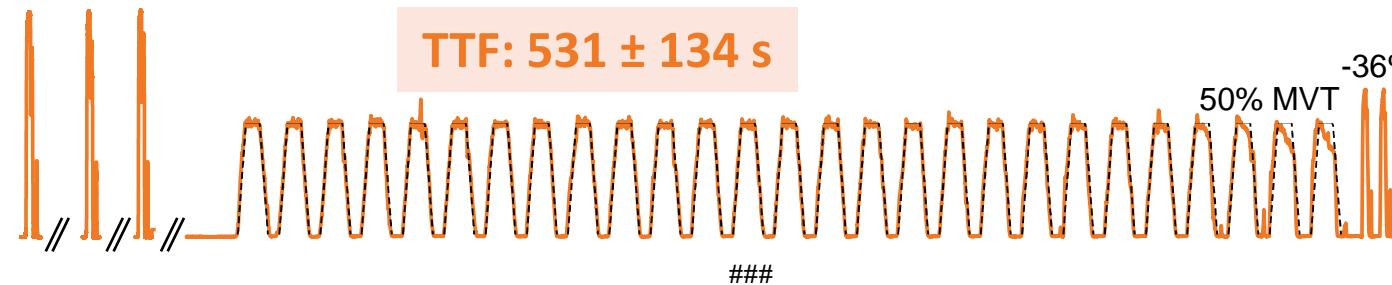
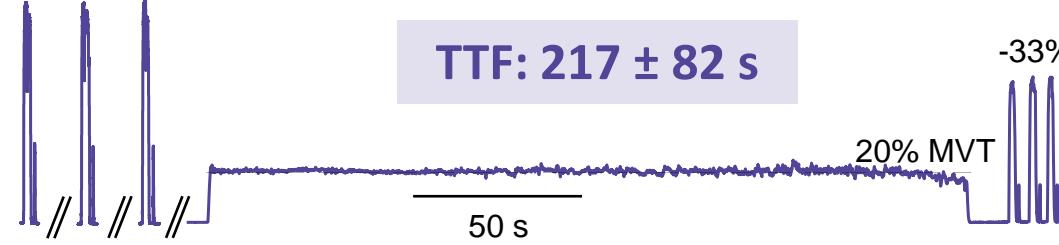


Quantifying motor unit discharge rate



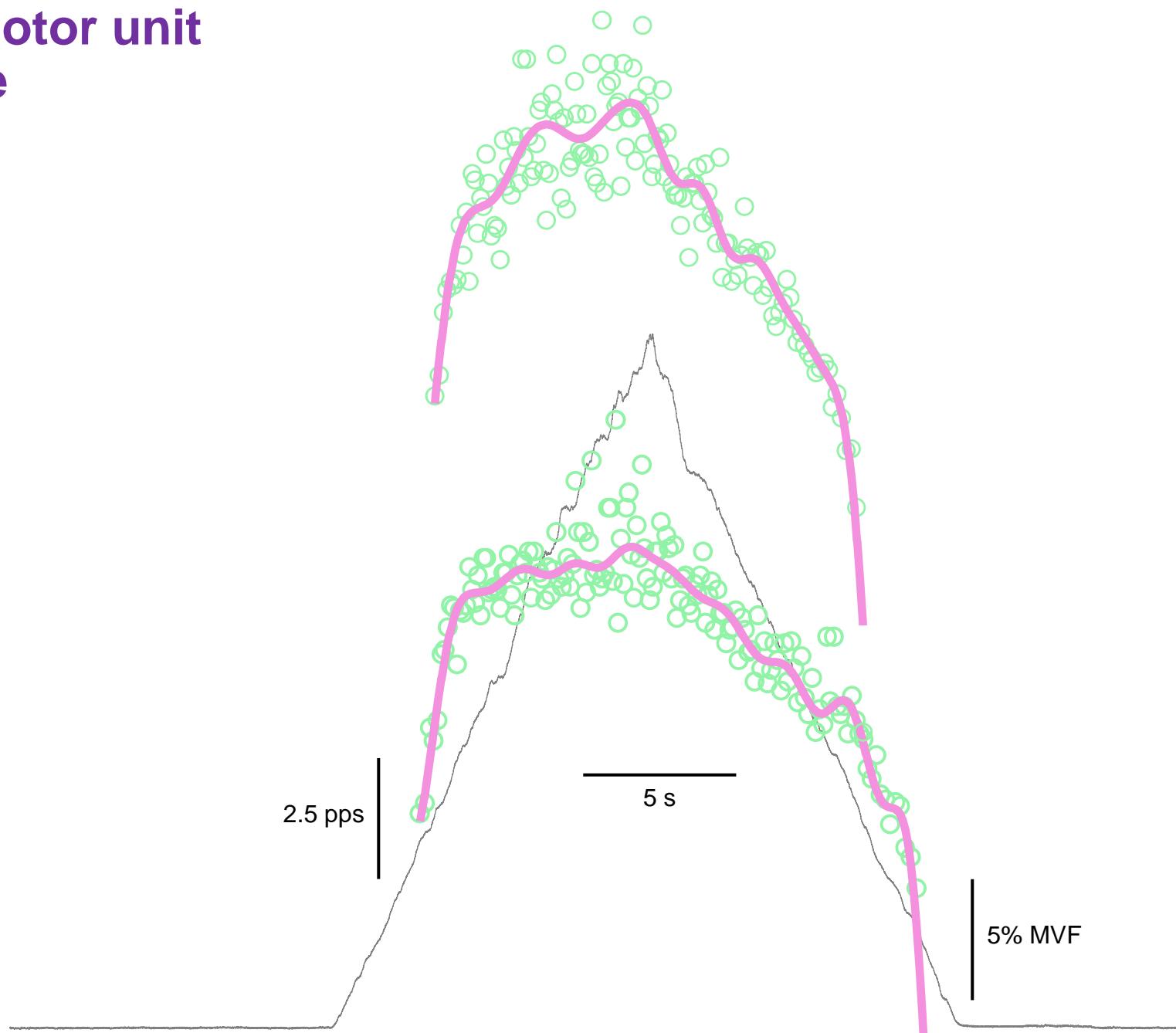


Motor unit discharge rate during contractions to failure

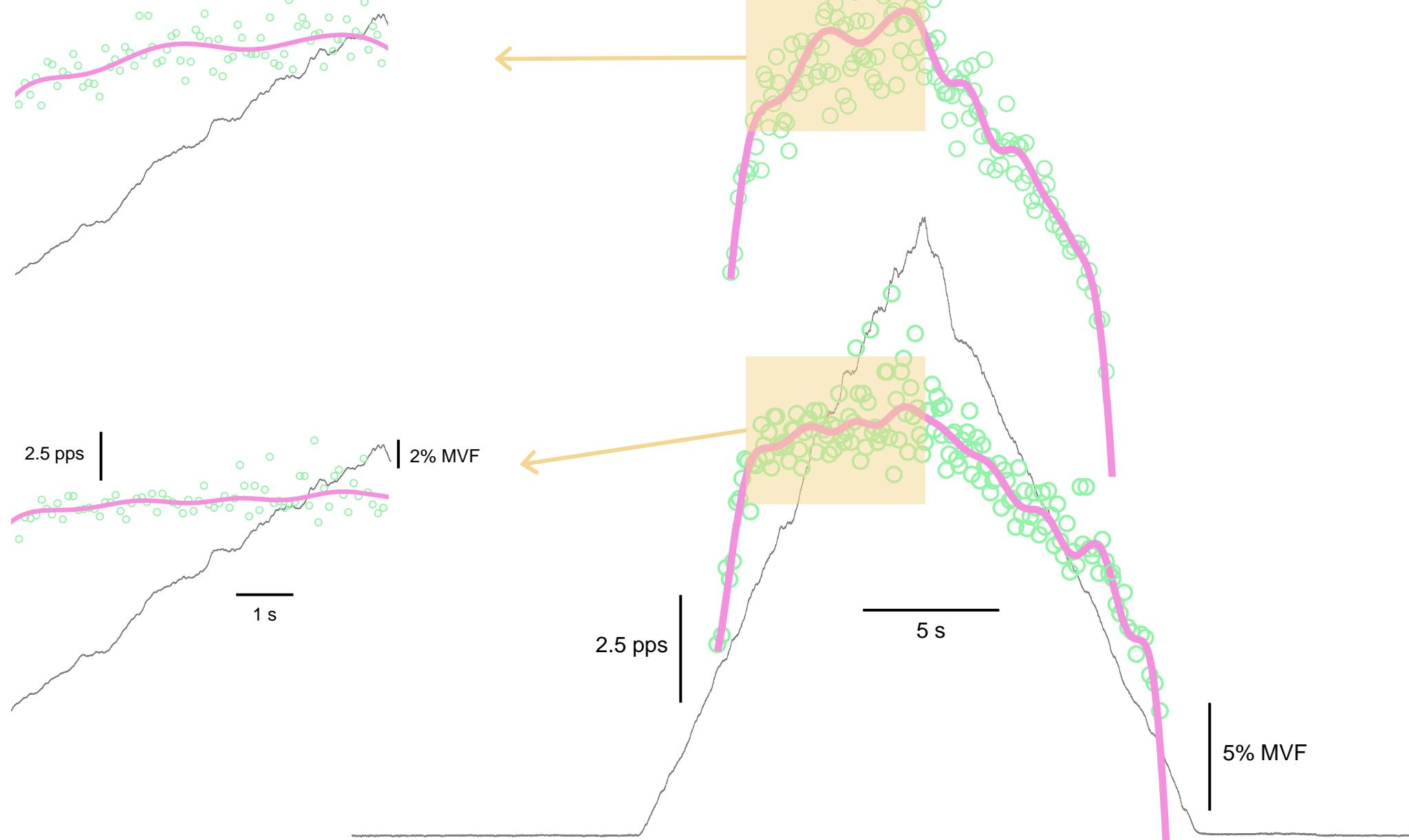


Tamara Valenčič

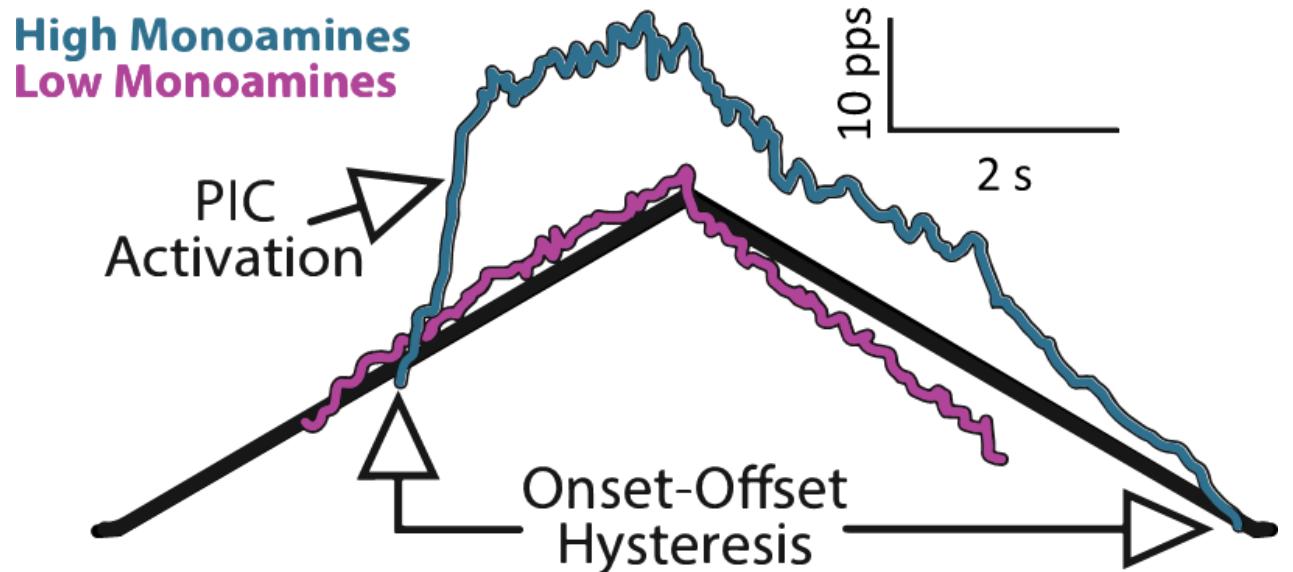
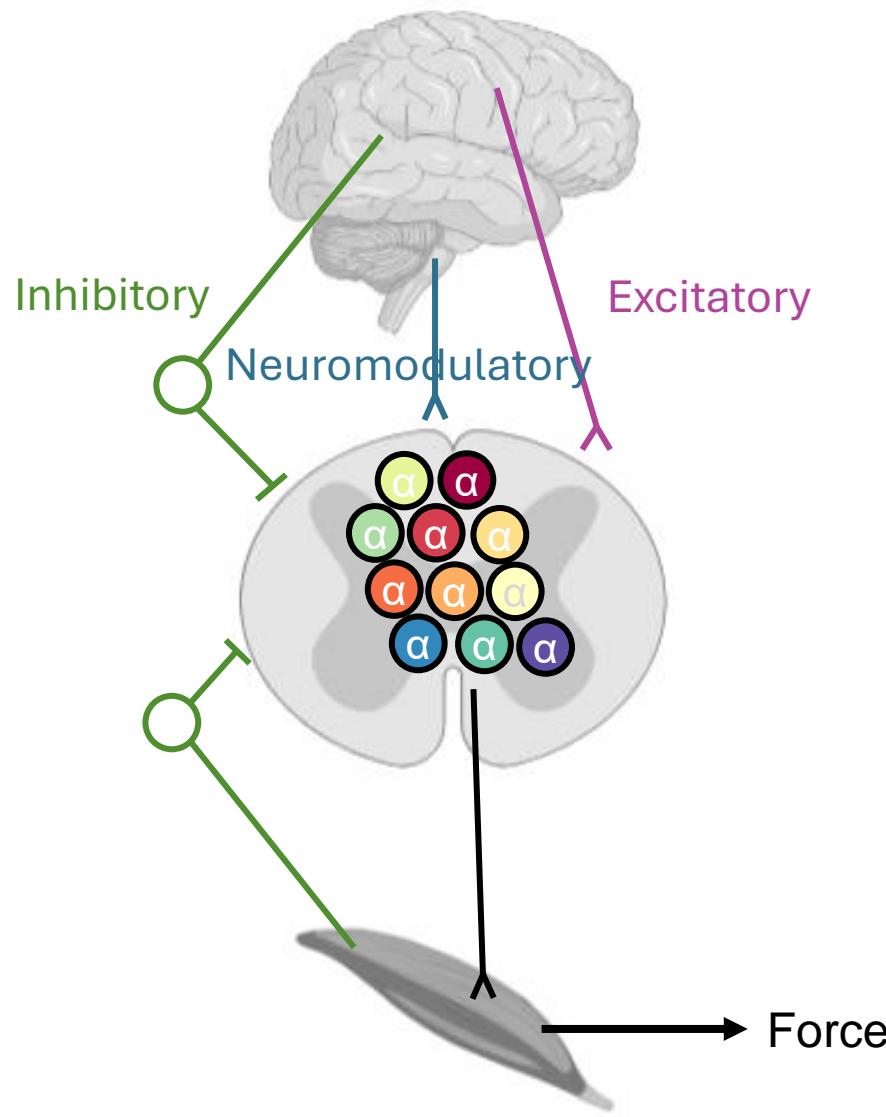
Quantifying motor unit discharge rate



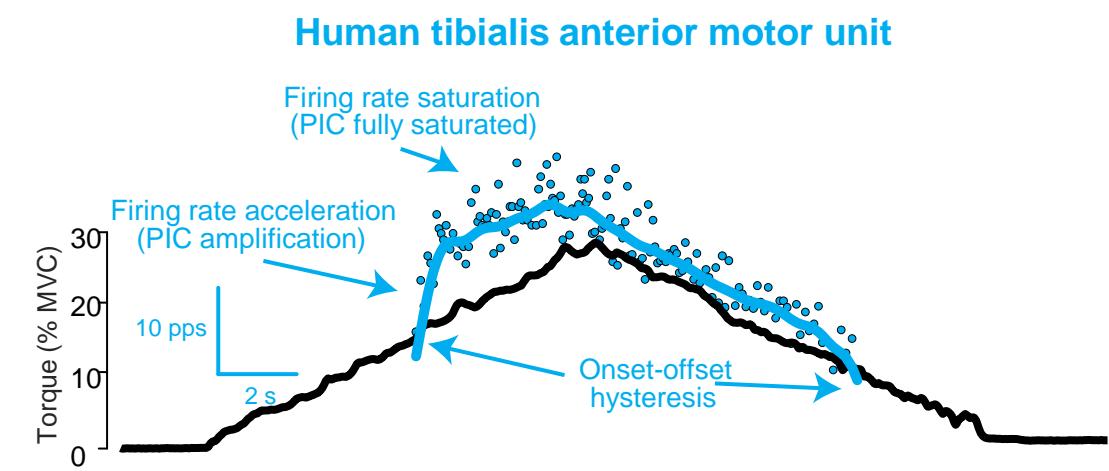
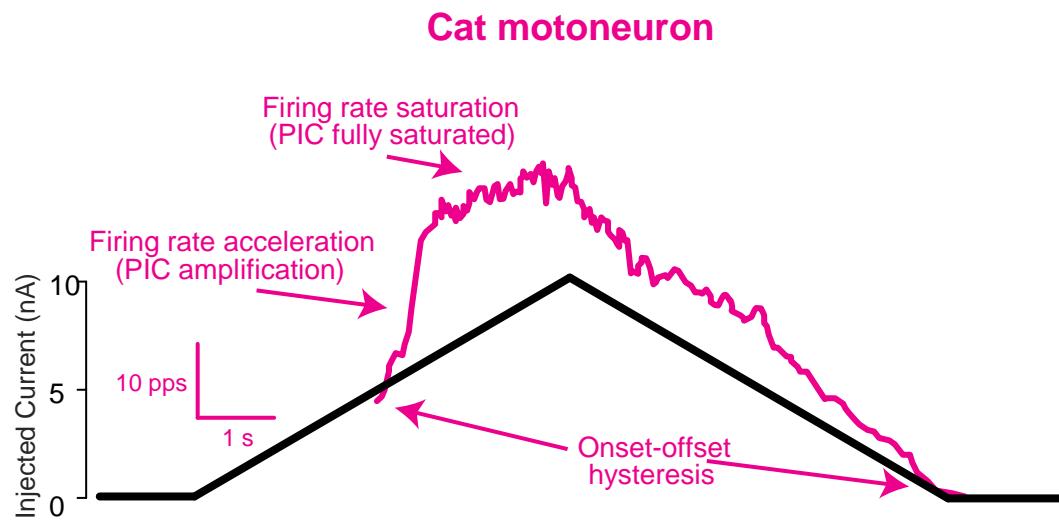
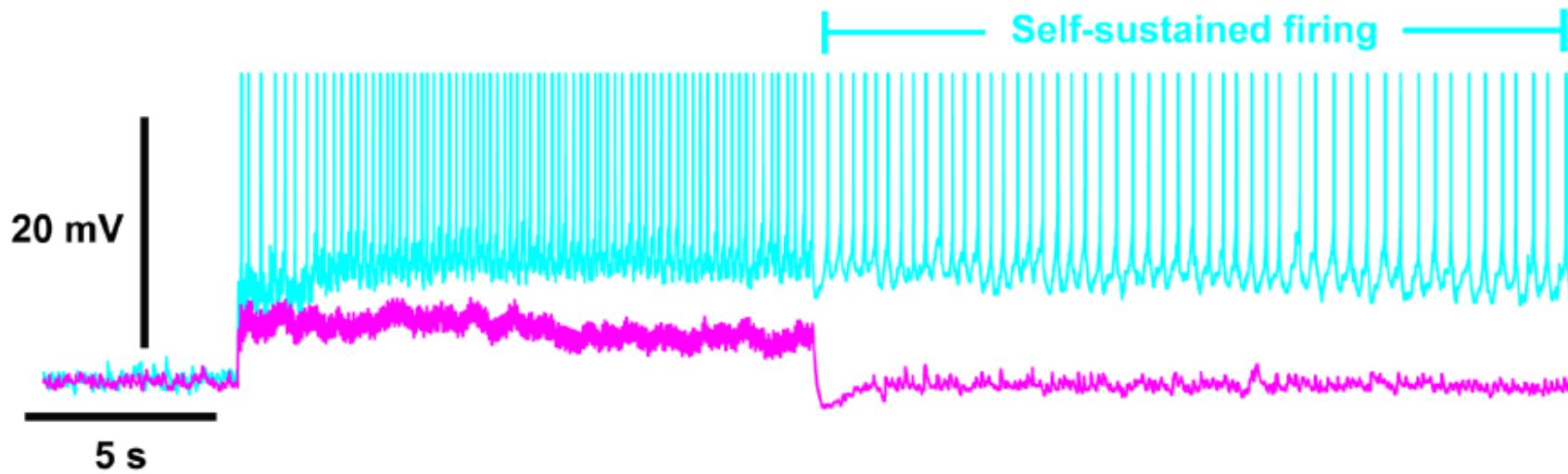
Quantifying motor unit discharge rate



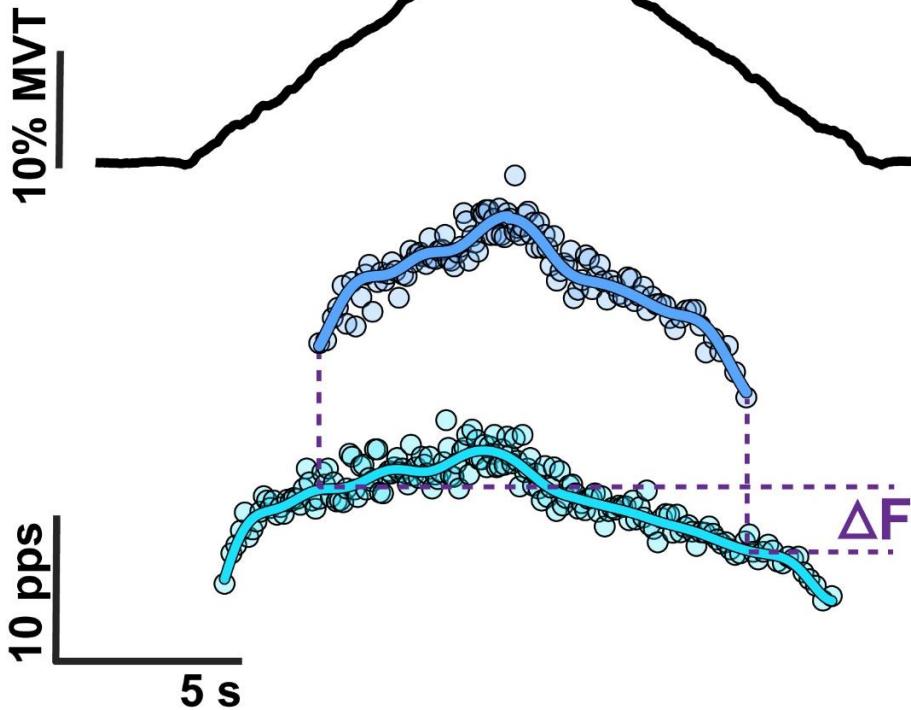
Different types of inputs to motoneurons



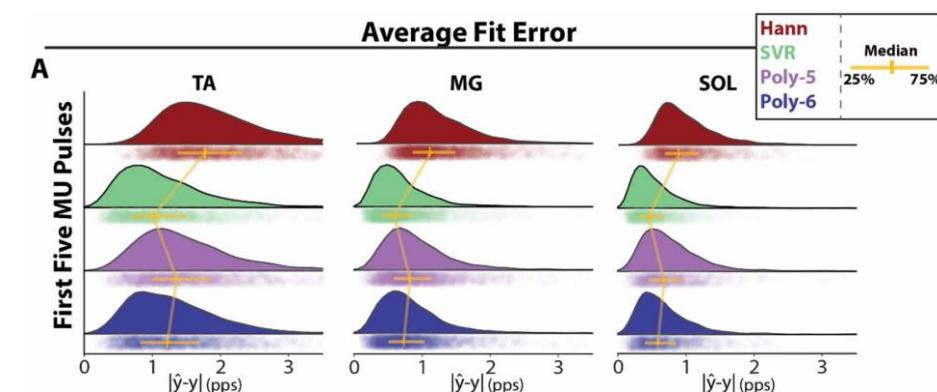
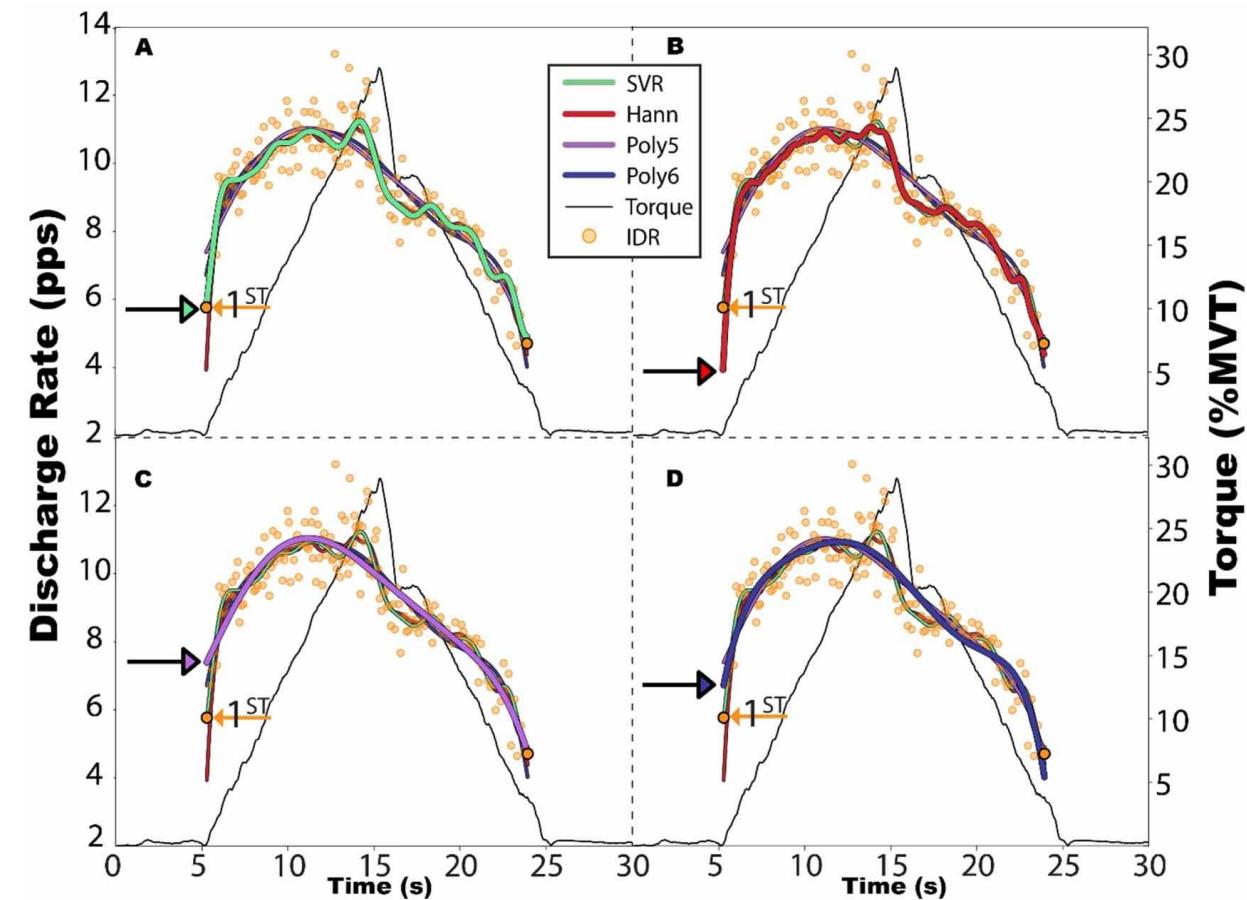
Plateau potentials – self-sustained firing of motoneurons



Prolongation – hysteresis

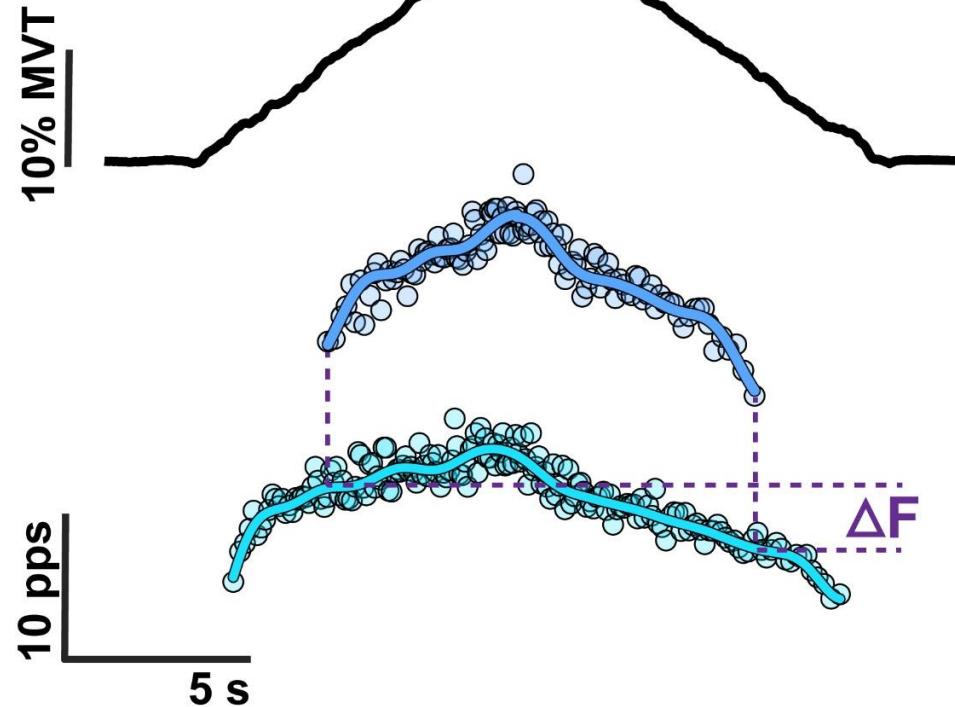


Gorassini et al. 1998, *Neurosci Lett*



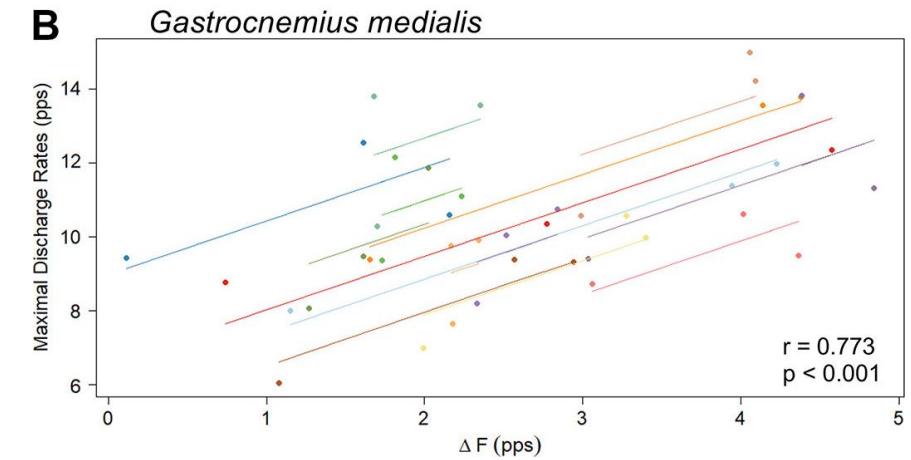
Beauchamp et al. 2022, *J Neural Eng*

Discharge rate hysteresis

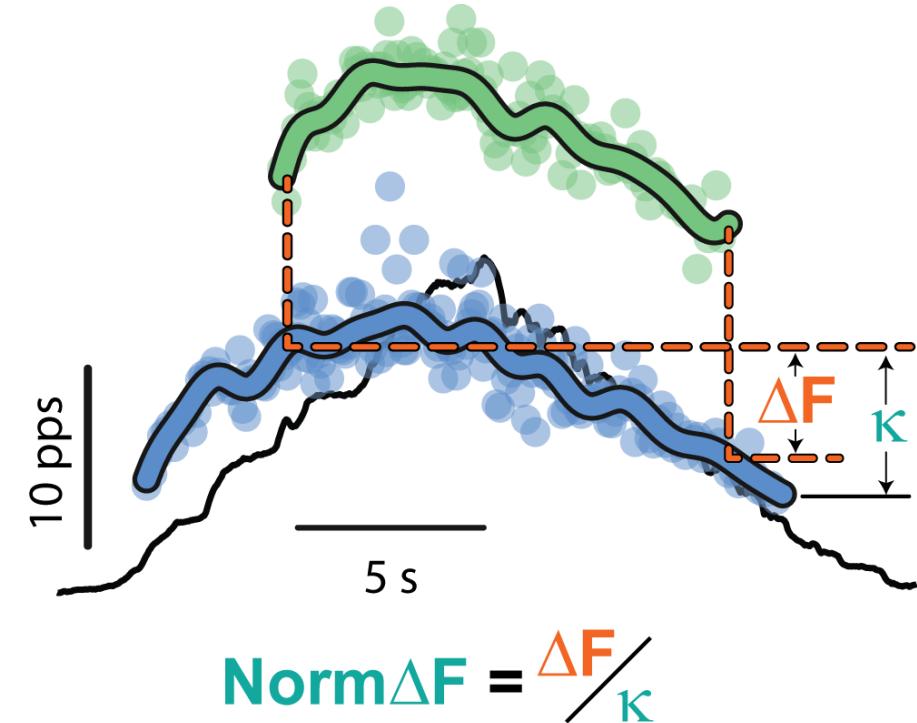


Beauchamp et al. 2023, J Neural Eng

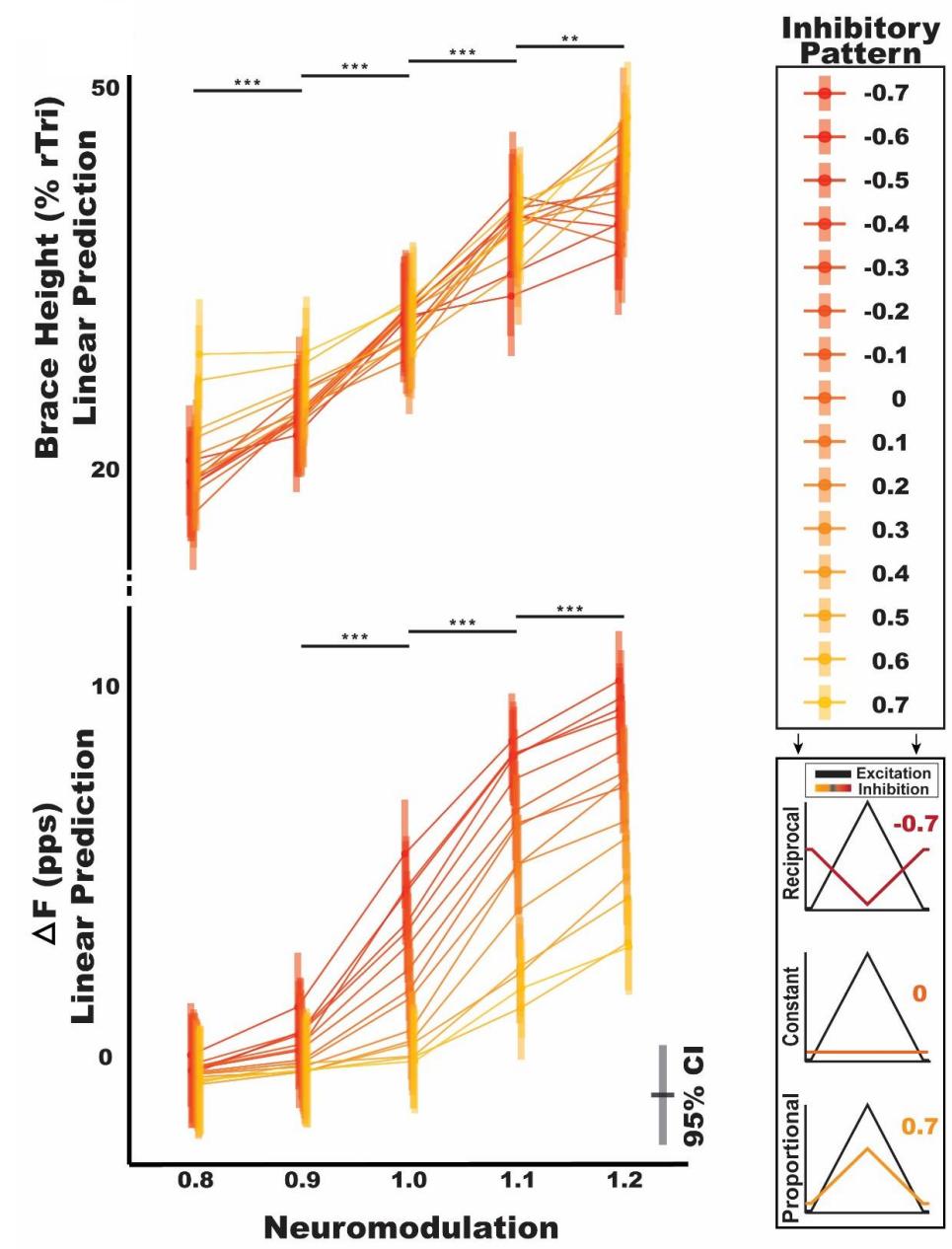
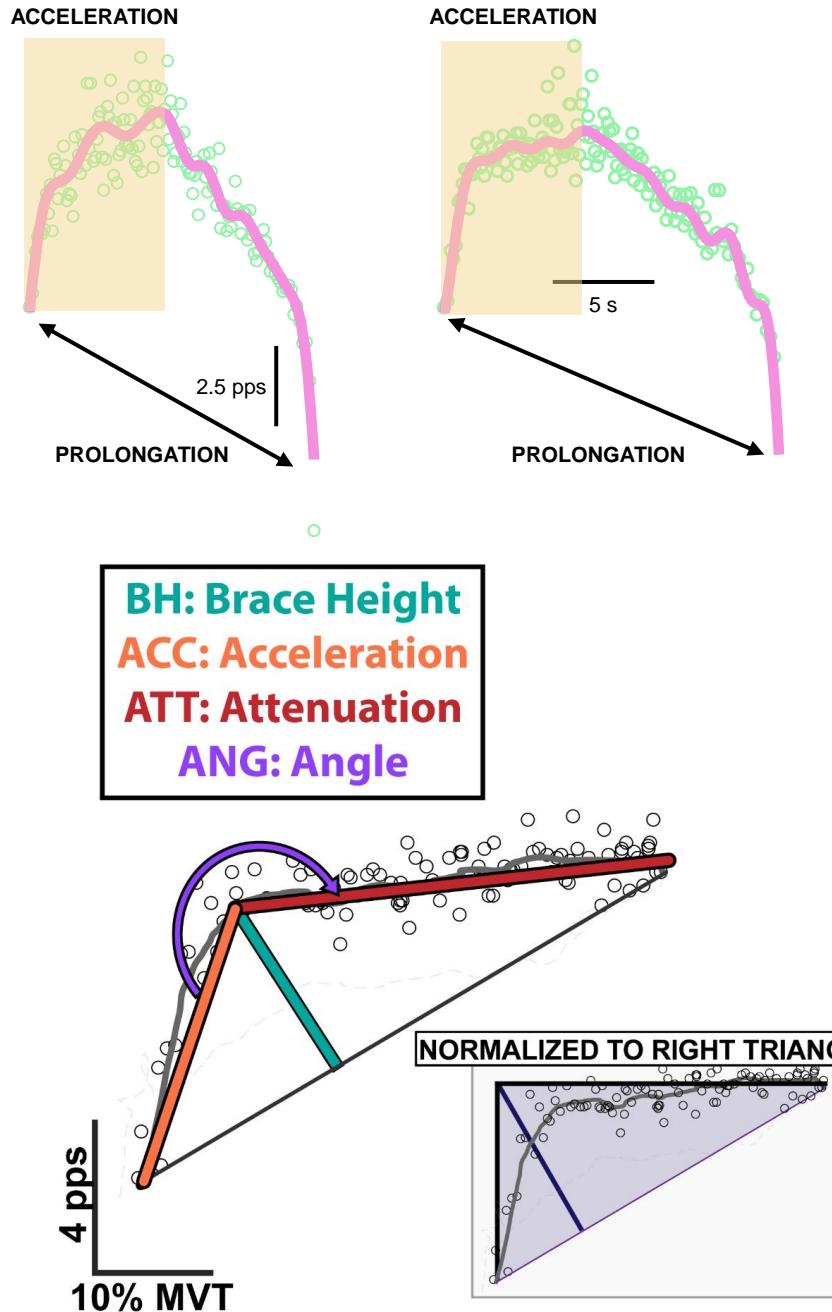
- Rate-rate correlation $R^2 > 0.7$
- Recruitment time difference $> 1 \text{ s}$
- Reporter (control) unit discharge rate modulation $> 0.5 \text{ pps}$



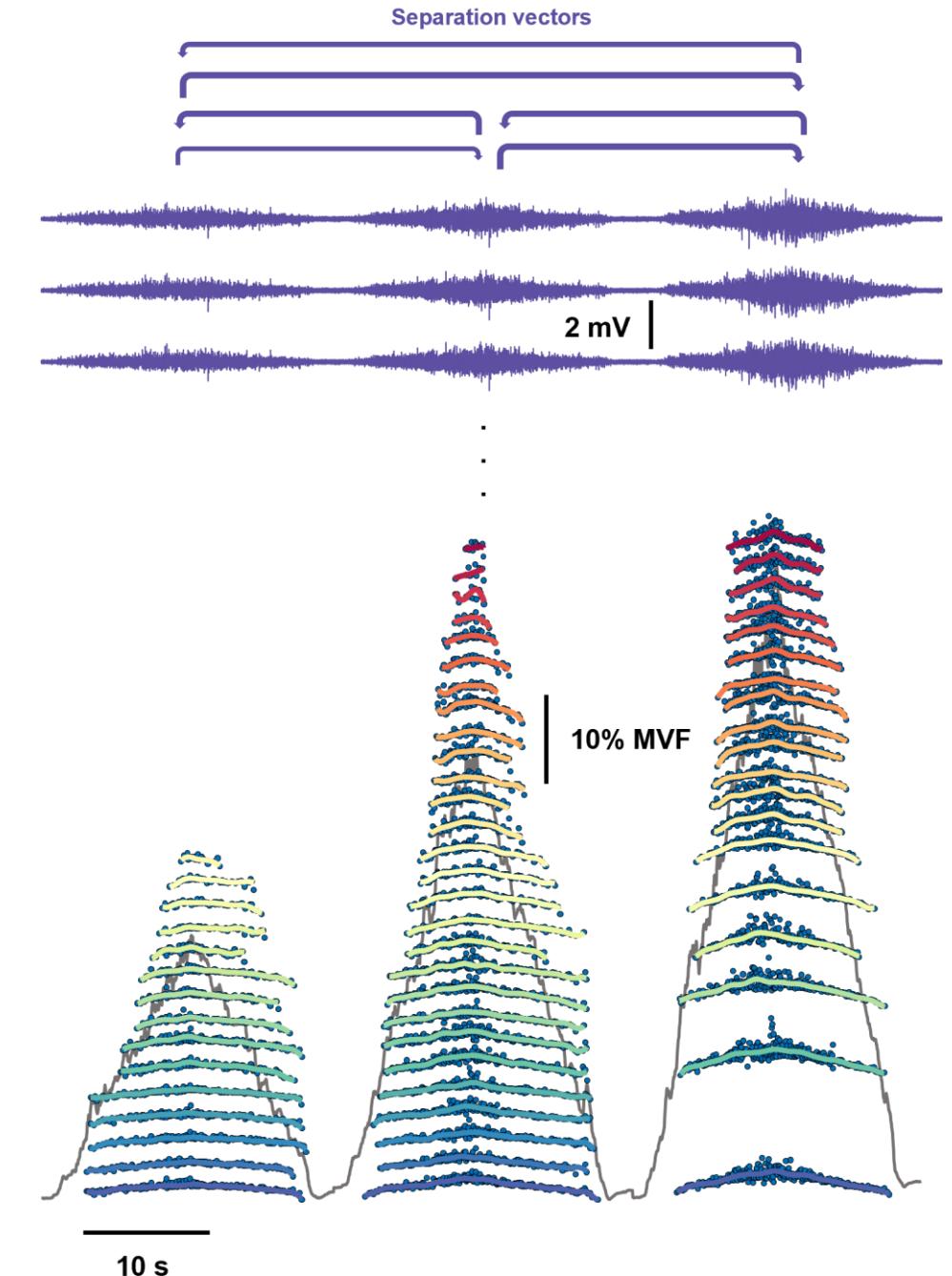
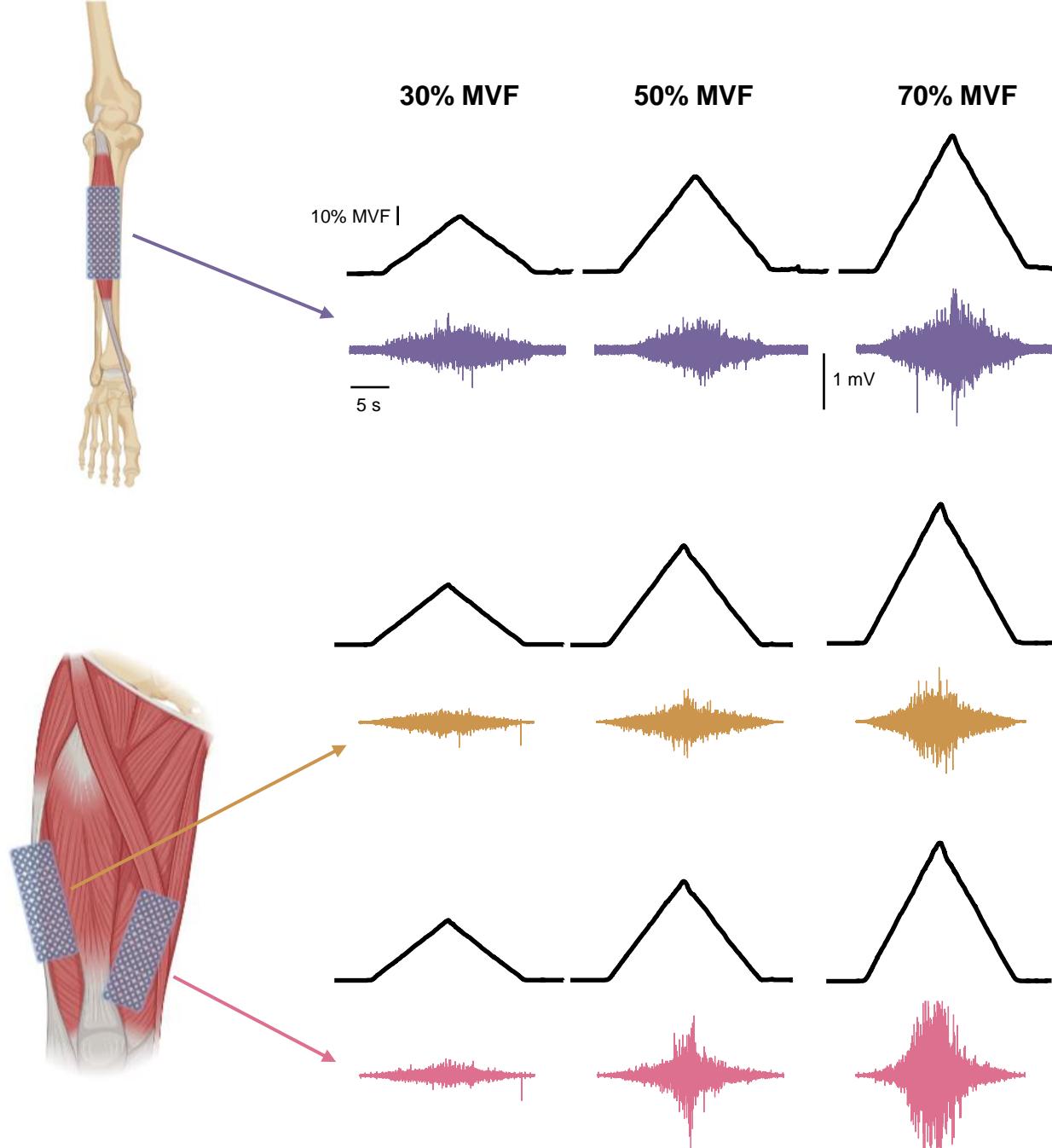
Orssatto et al. 2021, J Neurophysiol



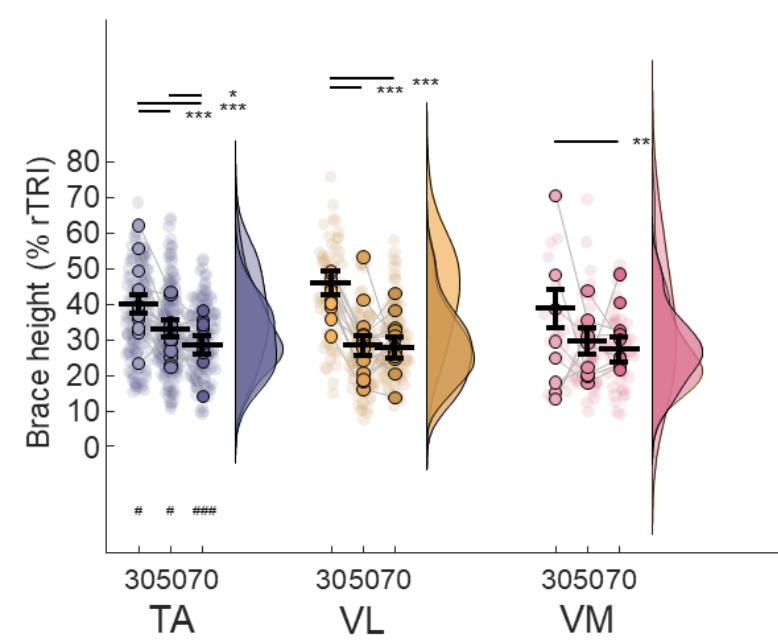
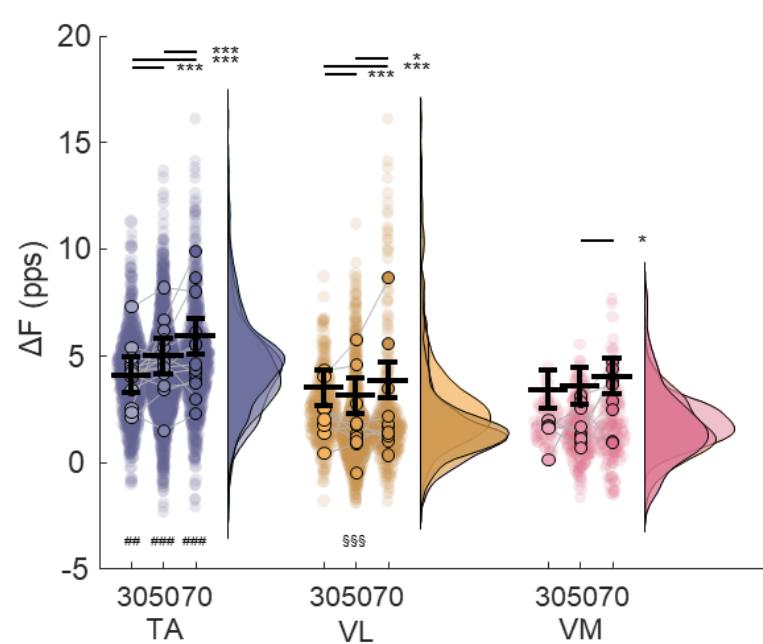
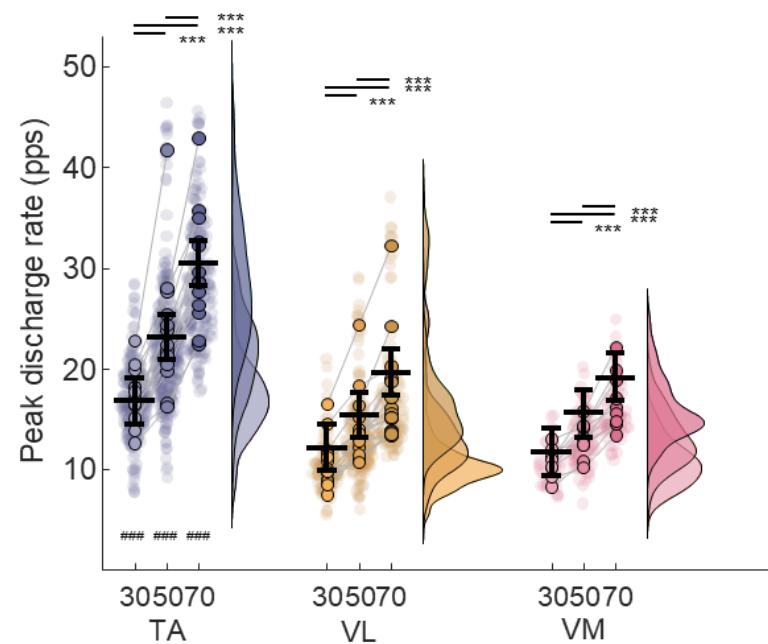
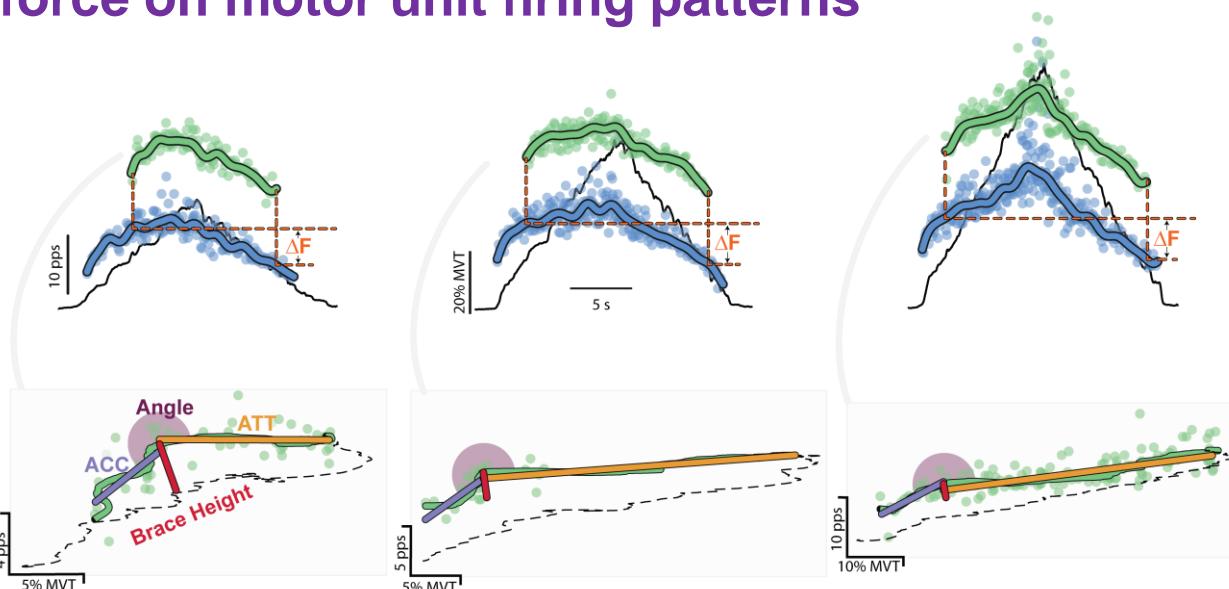
Škarabot et al. 2023, bioRxiv



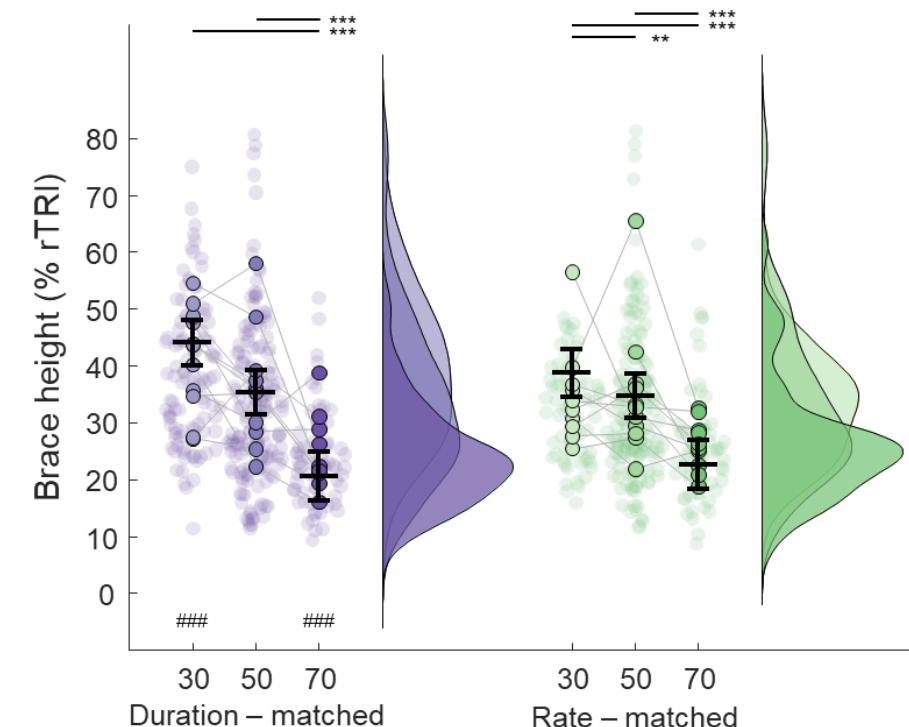
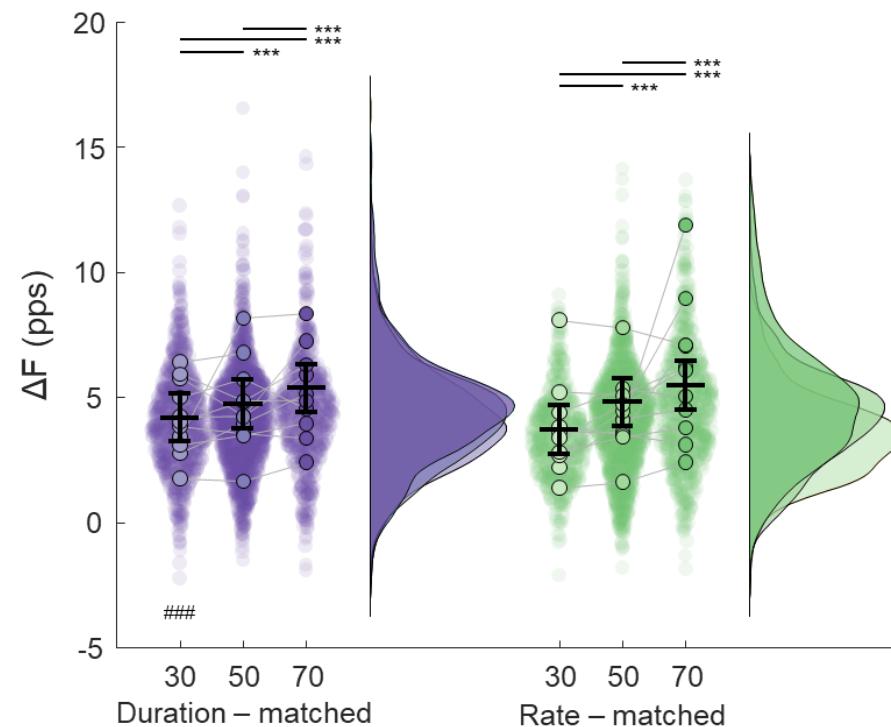
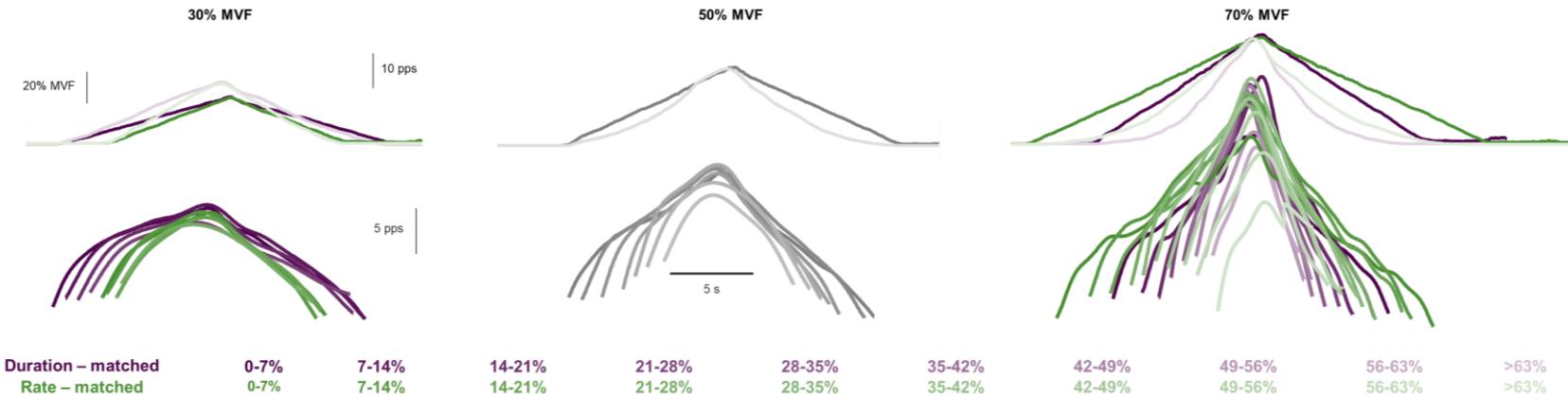
Modulation of PICs with muscle contraction force



The effect of contraction force on motor unit firing patterns

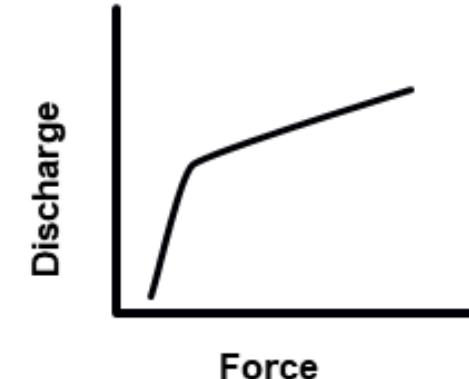
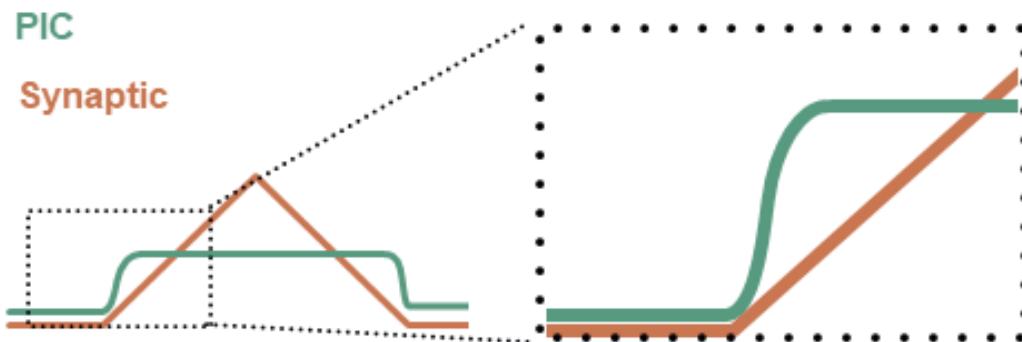


The effect of the rate of force increase on motor unit firing patterns

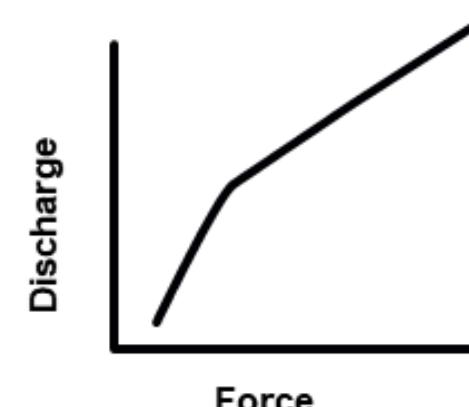
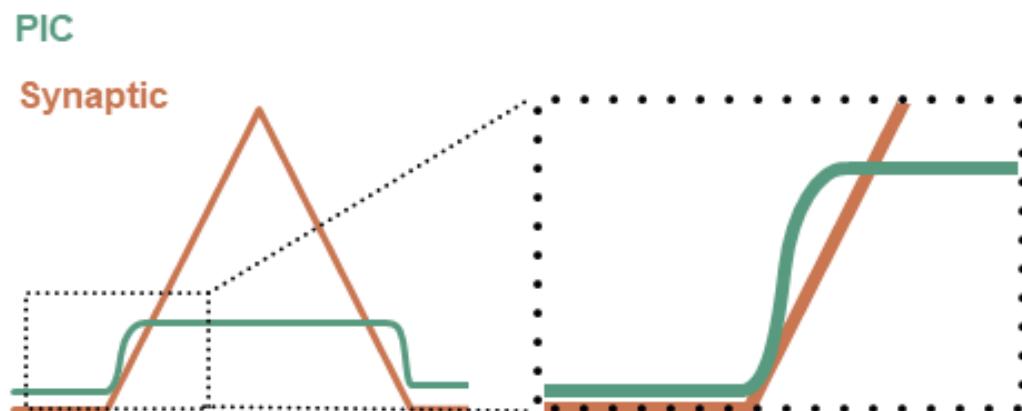


Inputs to motoneurons are uniquely shaped to support greater contraction force

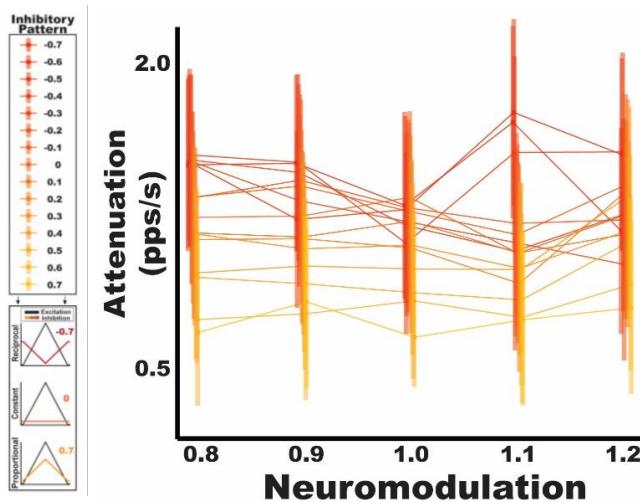
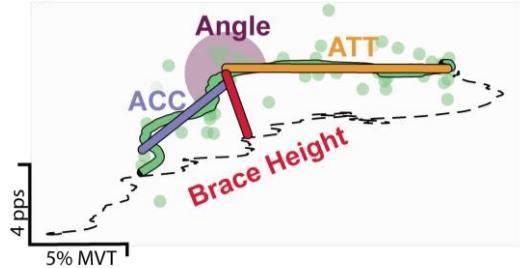
Low force or rate of synaptic input



High force or rate of synaptic input



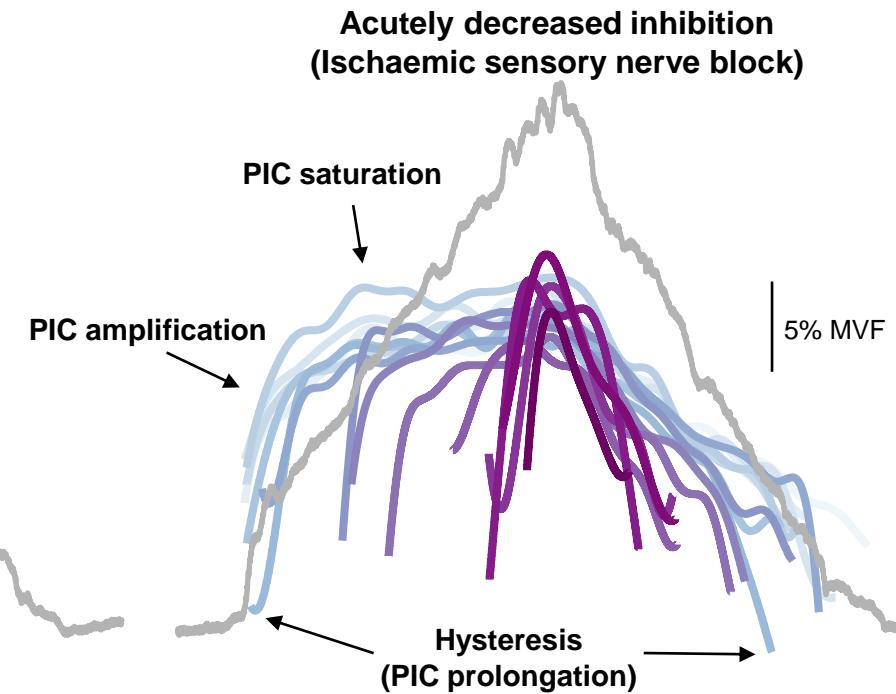
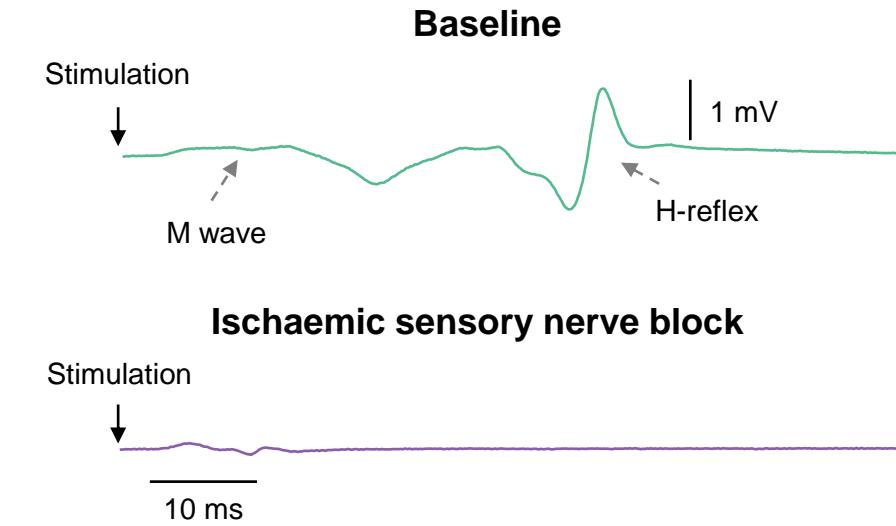
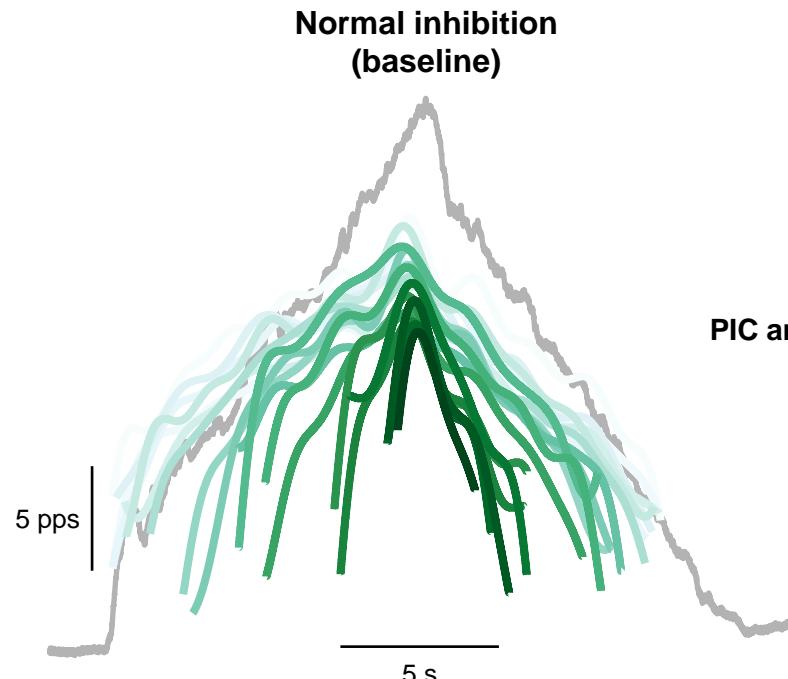
Inferring the inhibitory patterns



Beauchamp et al. 2023, J Neural Eng

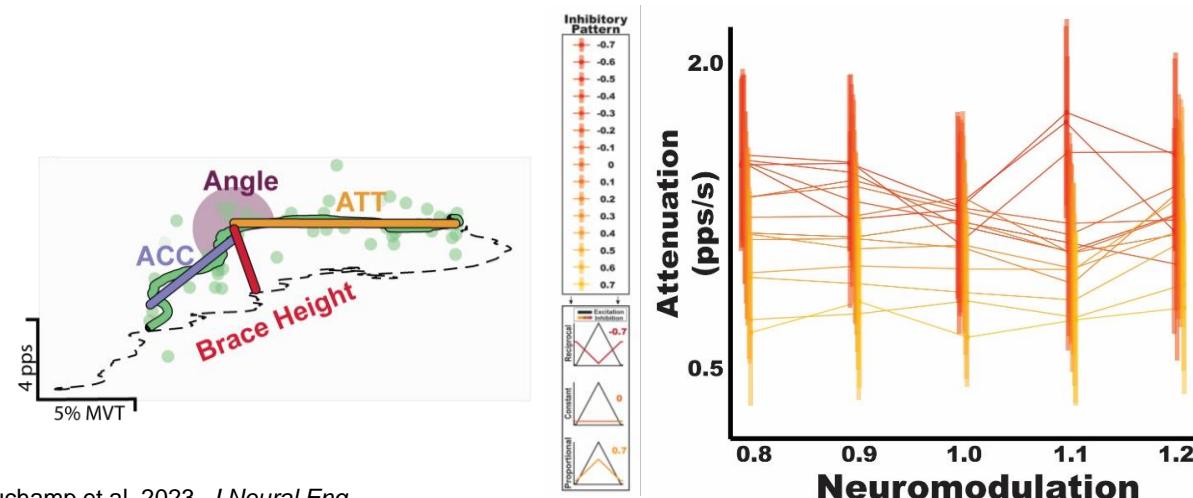


Nikki Bonett

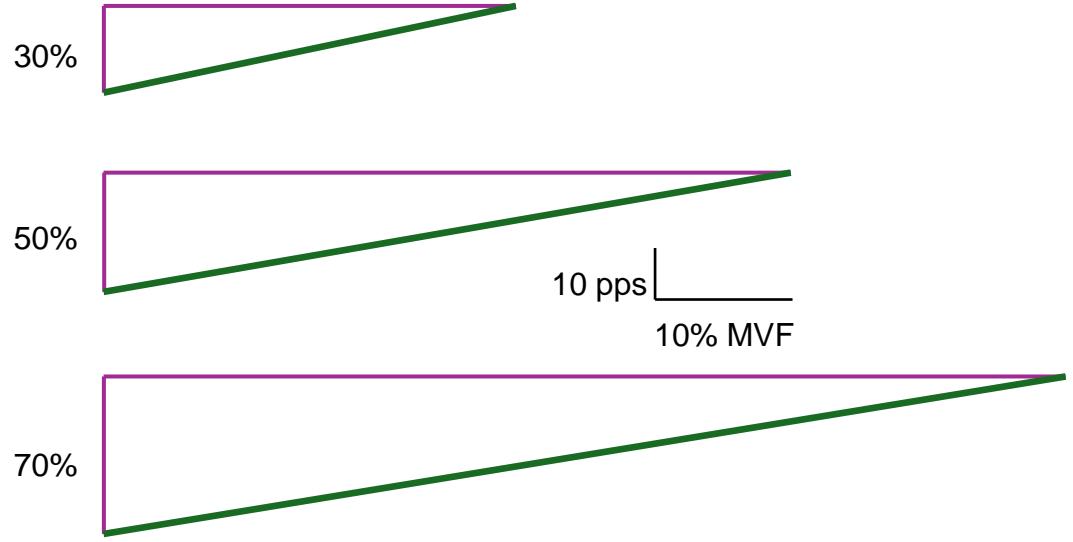


Bonett et al., *in preparation*

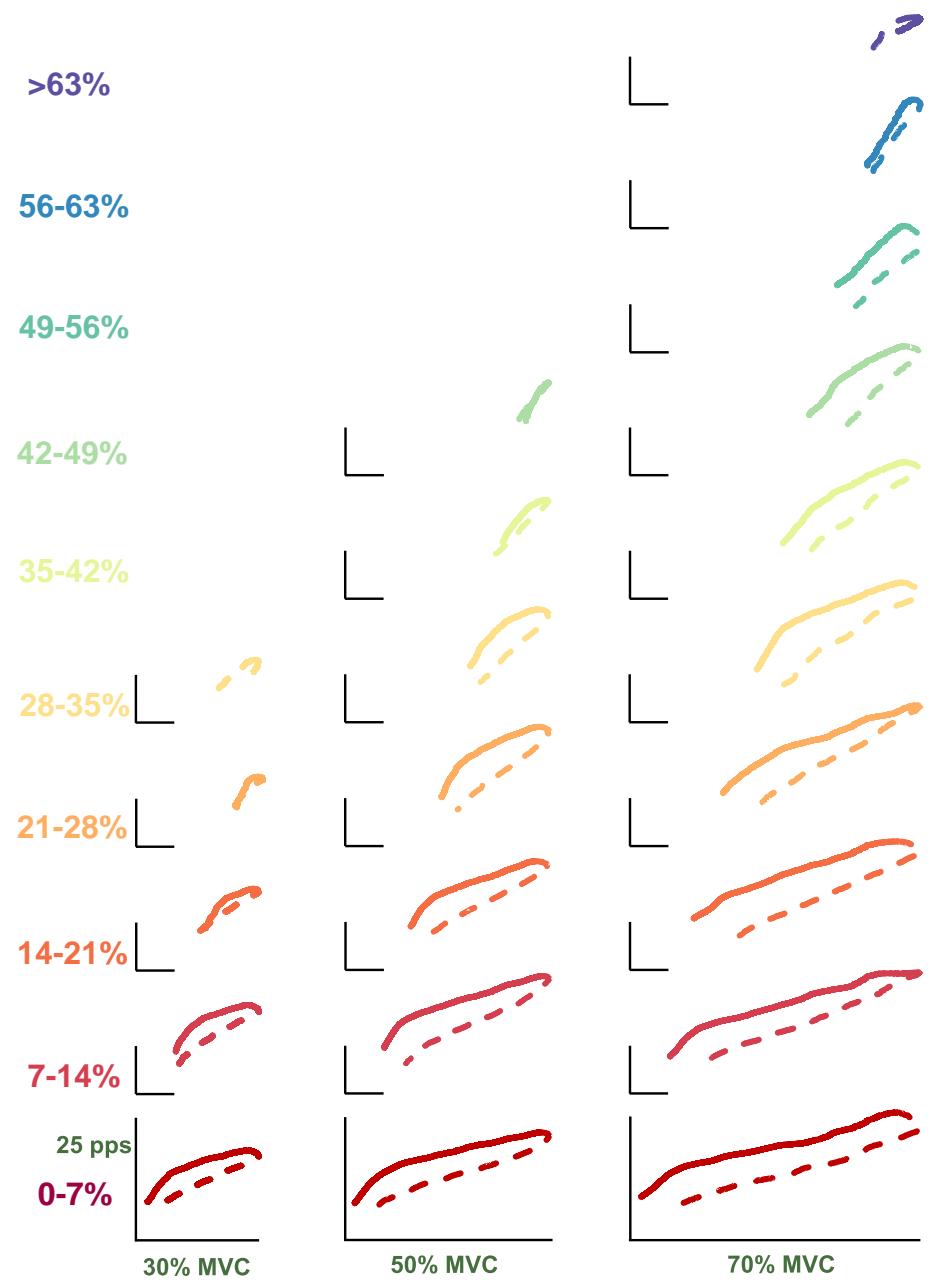
Inferring the inhibitory patterns



Beauchamp et al. 2023, J Neural Eng



Beauchamp et al. 2013, J Neural Eng

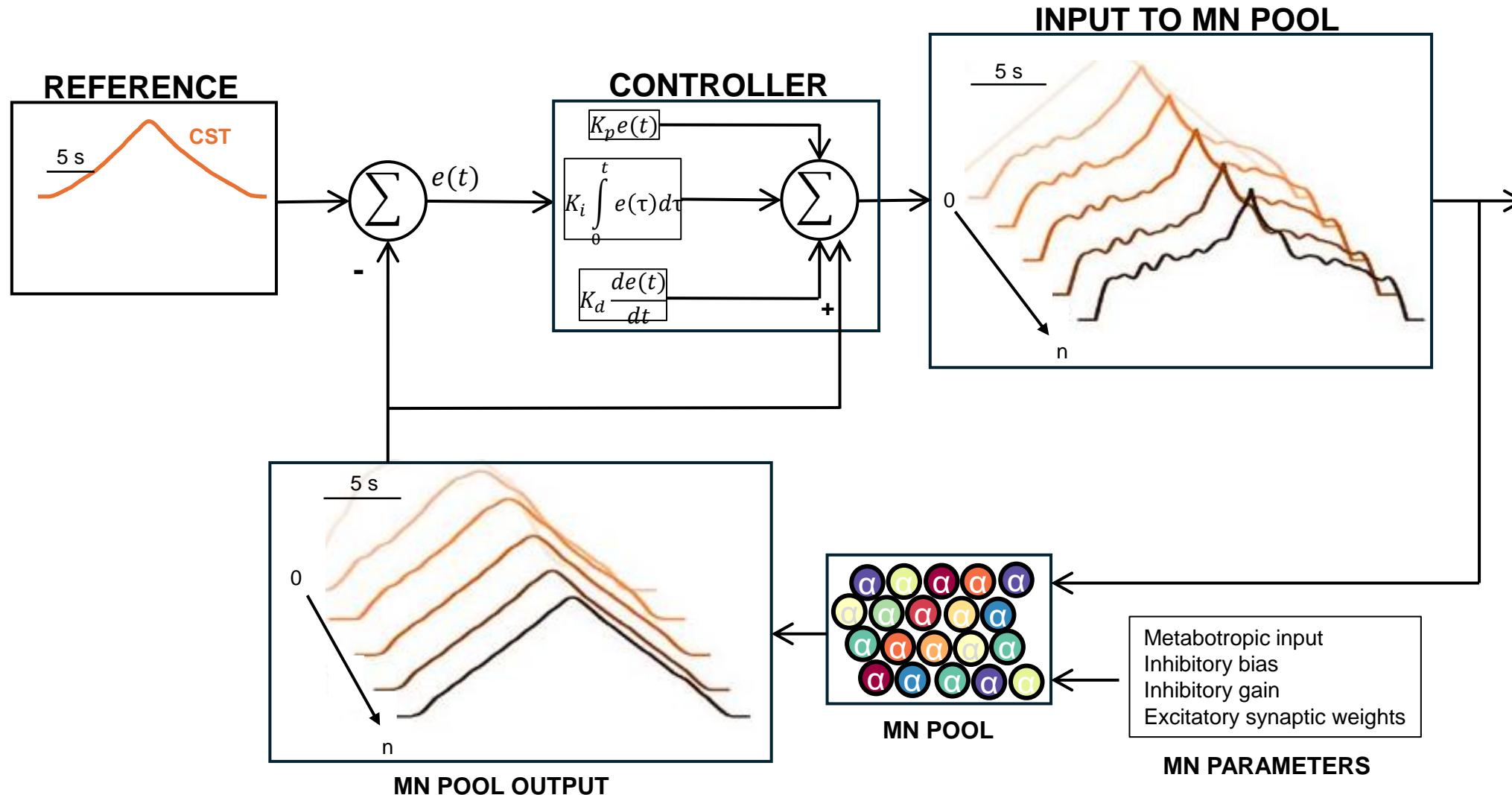


Škarabot, Beauchamp & Pearcey, *in preparation*

A realistic motoneuron model

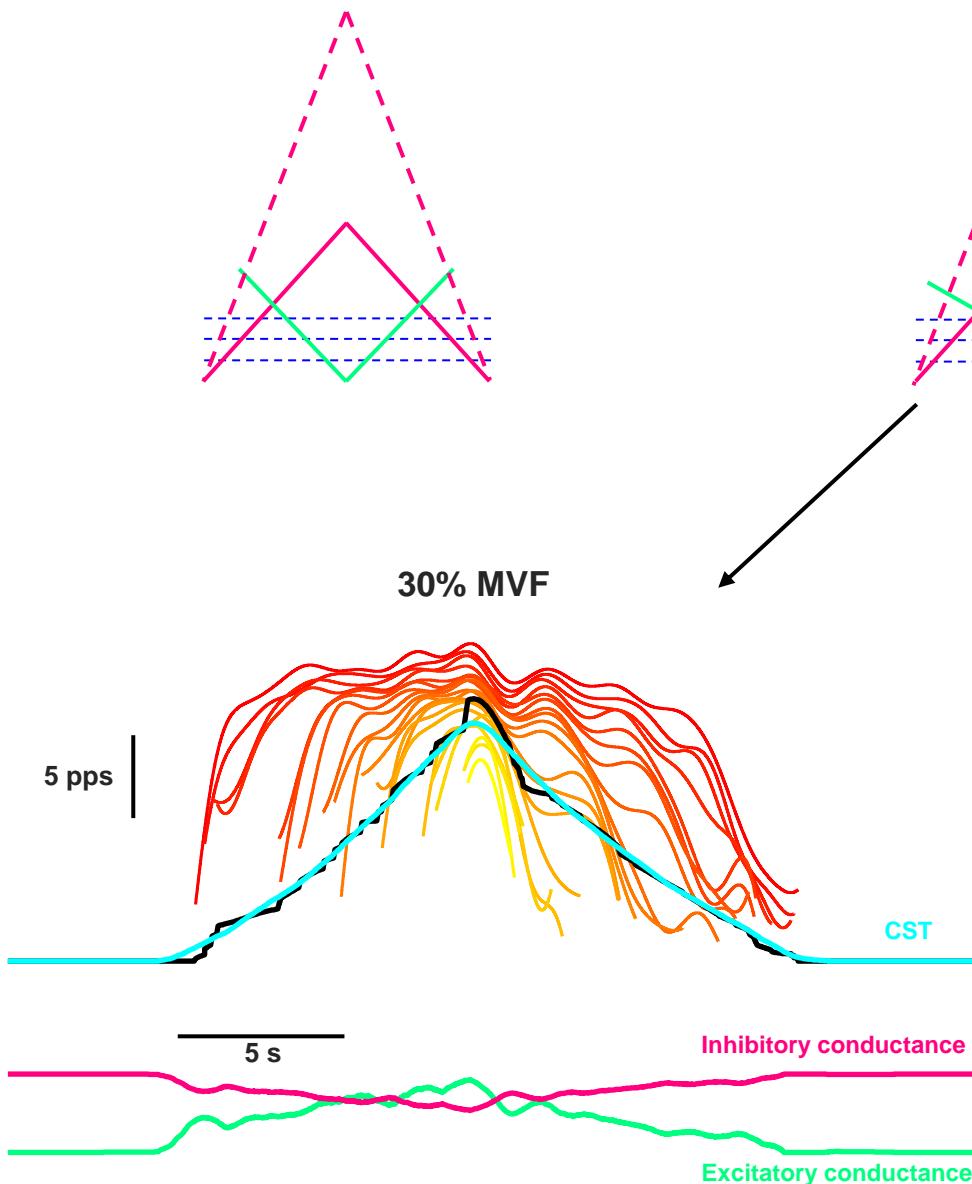


Dr Drew Beauchamp

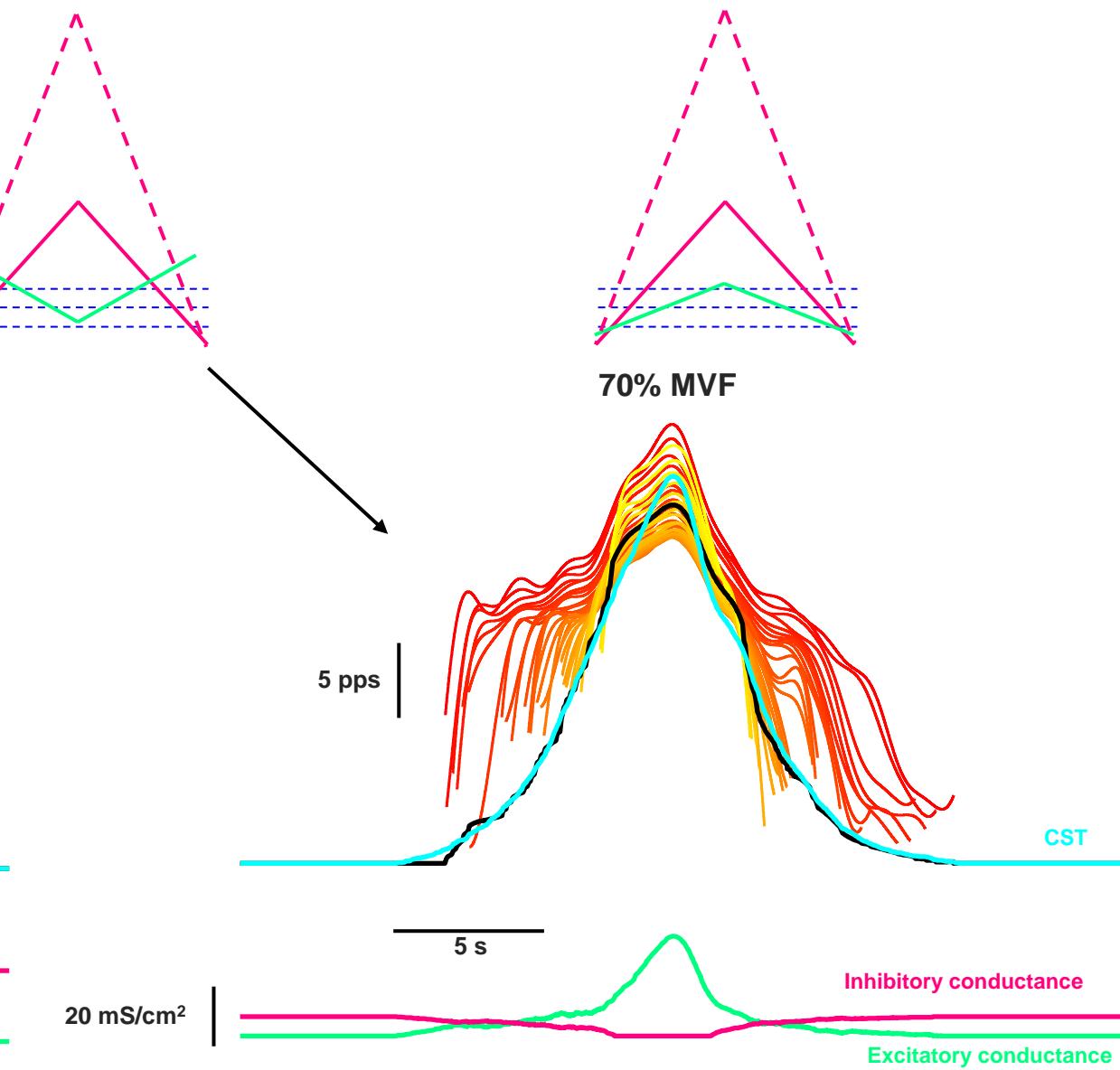


Simulated inputs to motoneuron pool

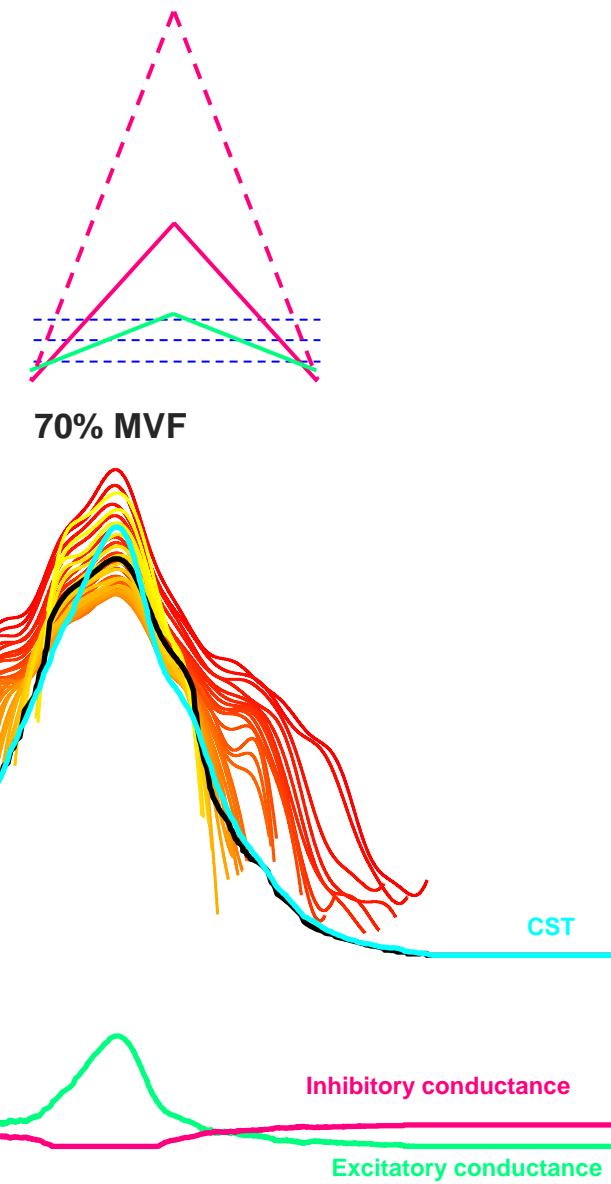
STRONG RECIPROCAL



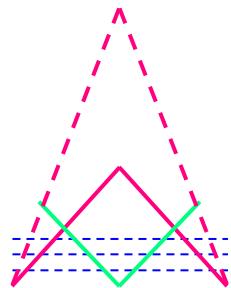
MILD RECIPROCAL



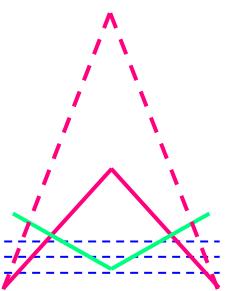
PROPORTIONAL



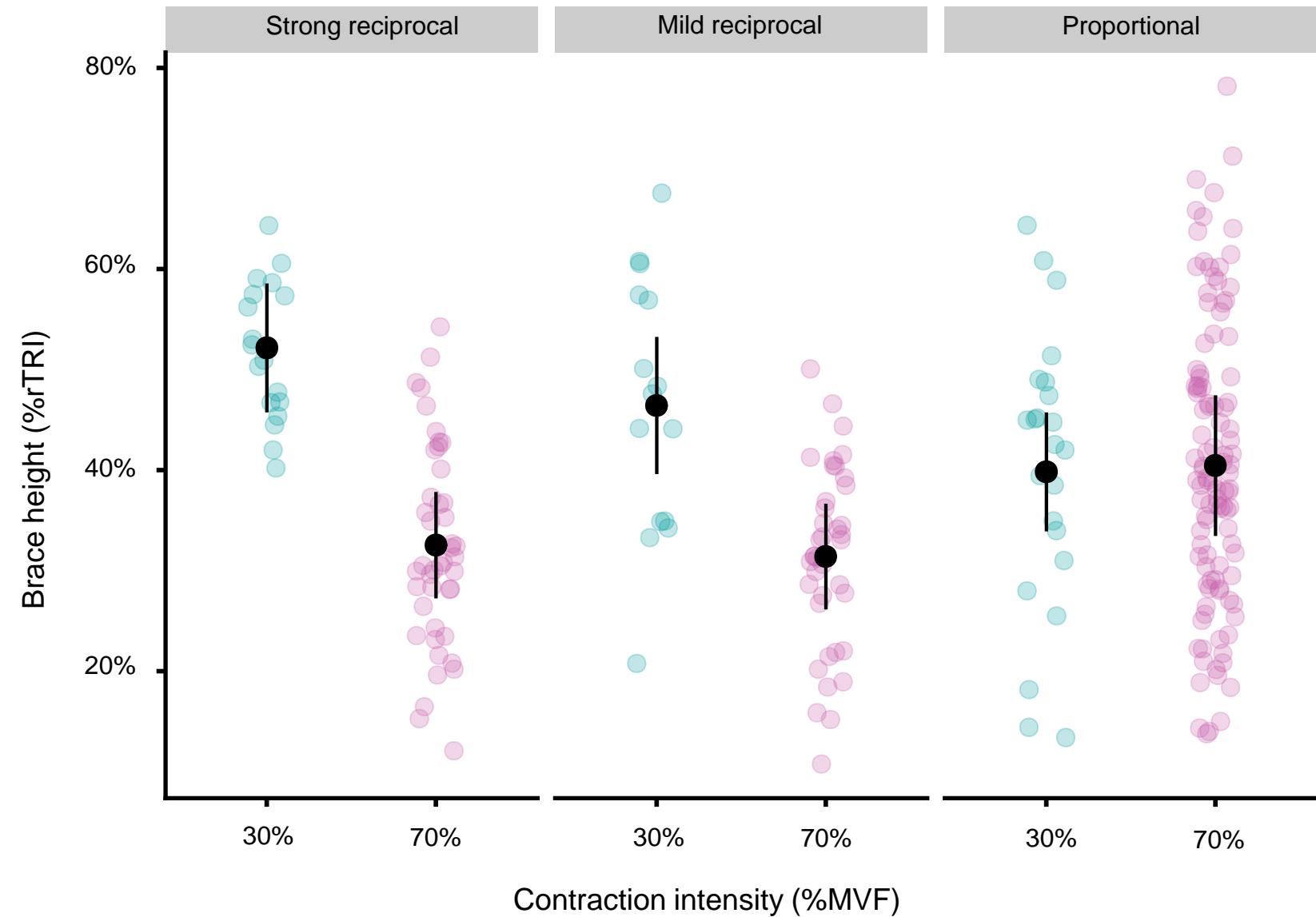
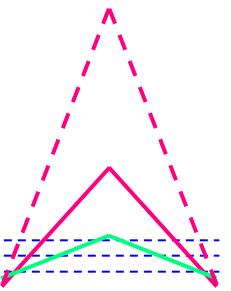
STRONG RECIPROCAL



MILD RECIPROCAL

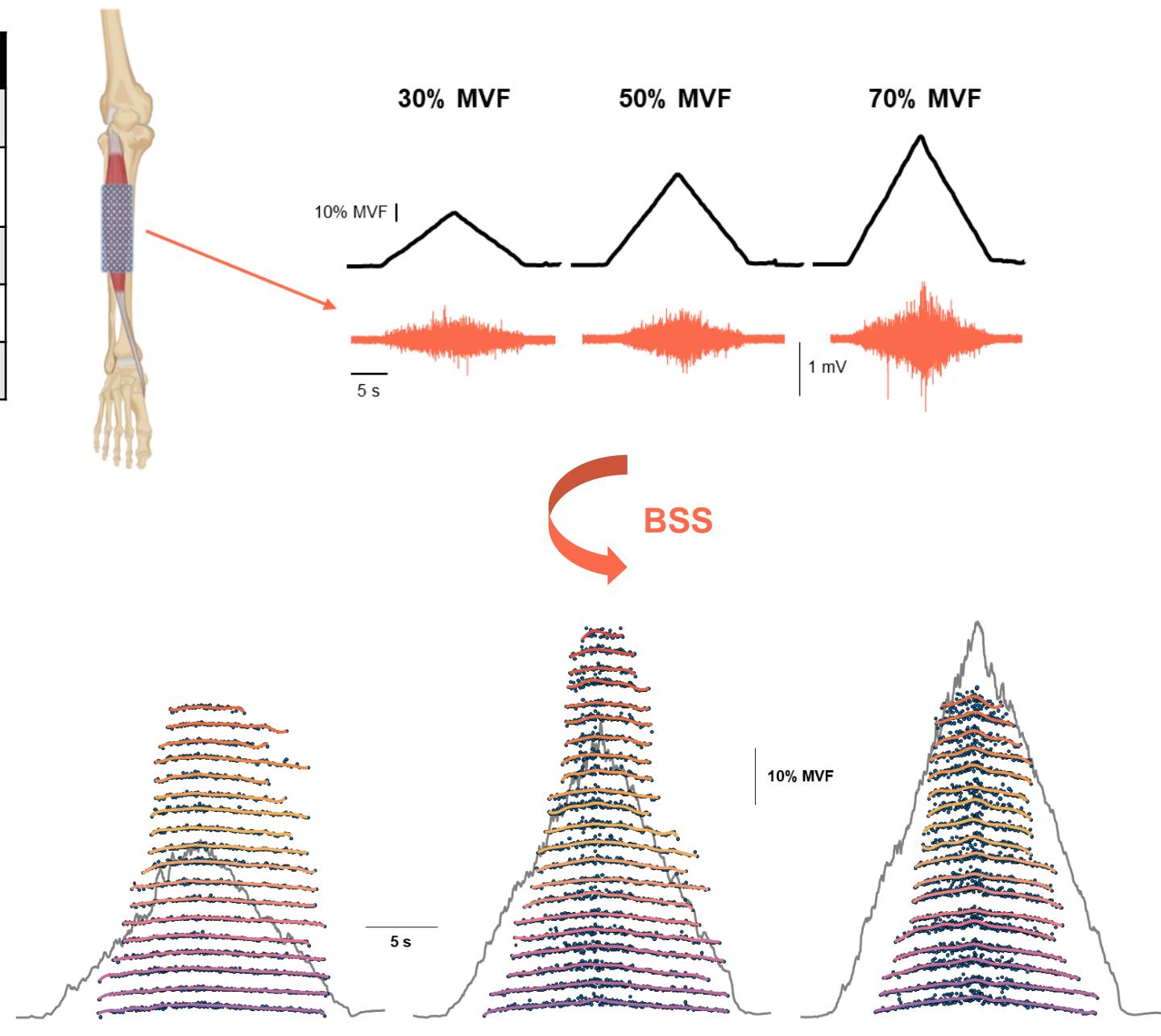
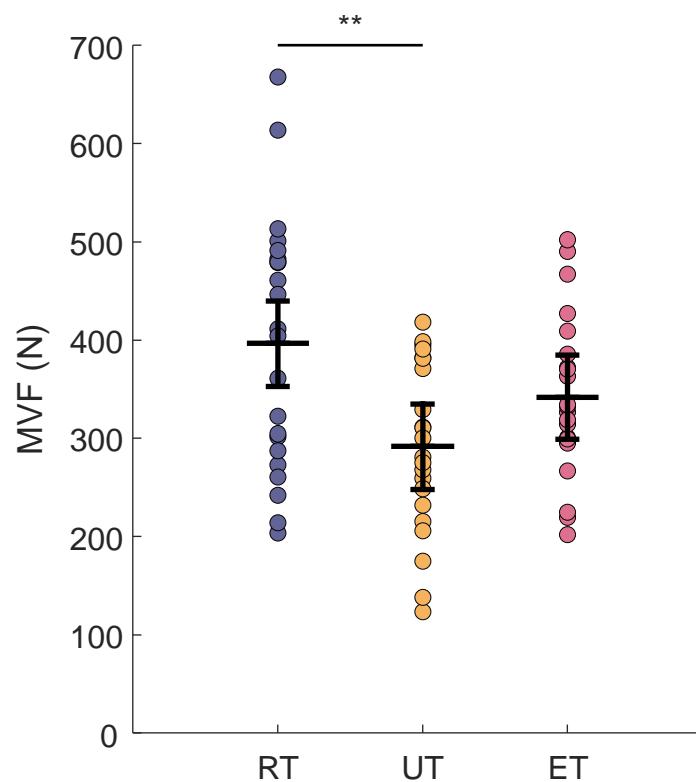


PROPORTIONAL

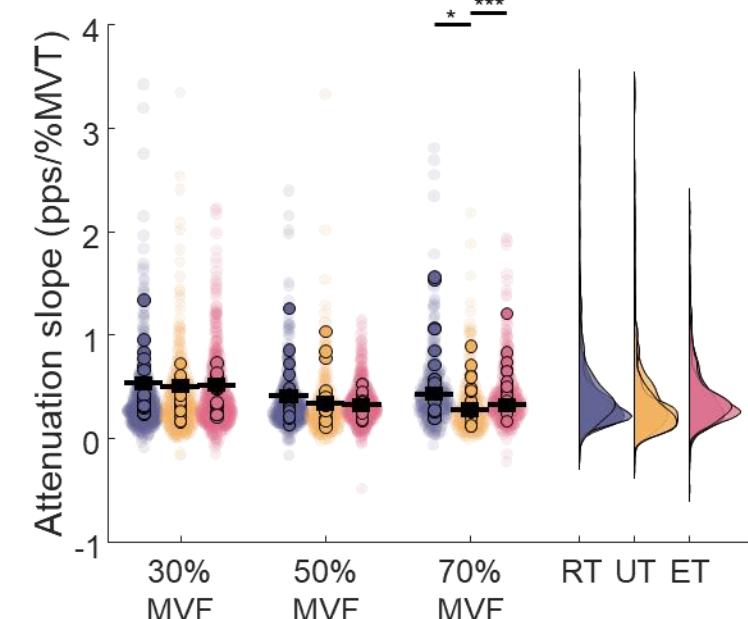
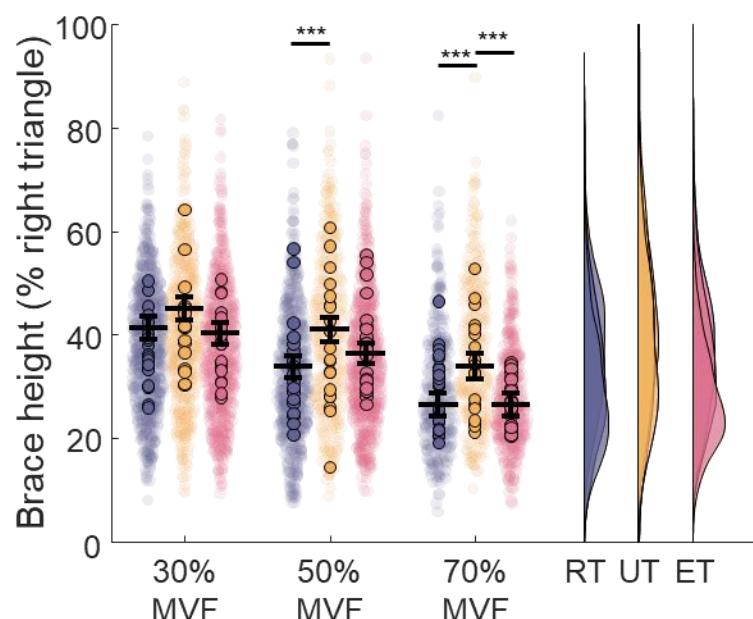
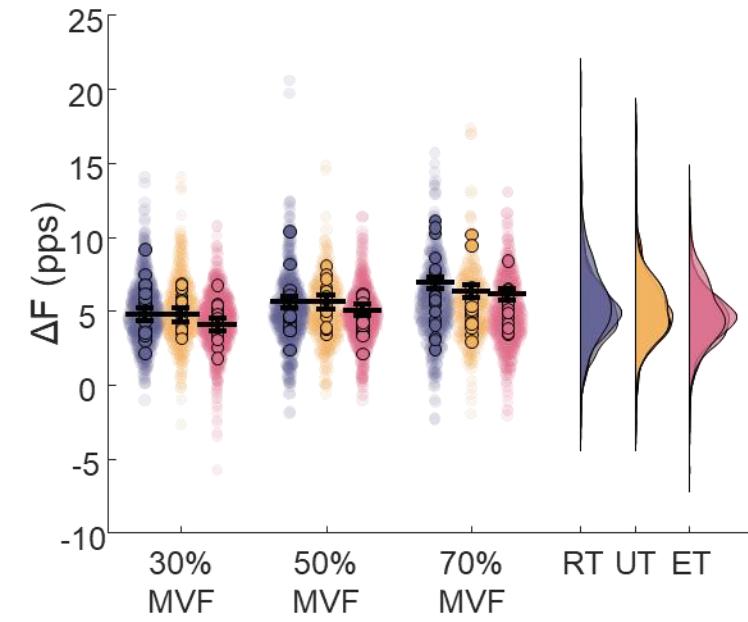
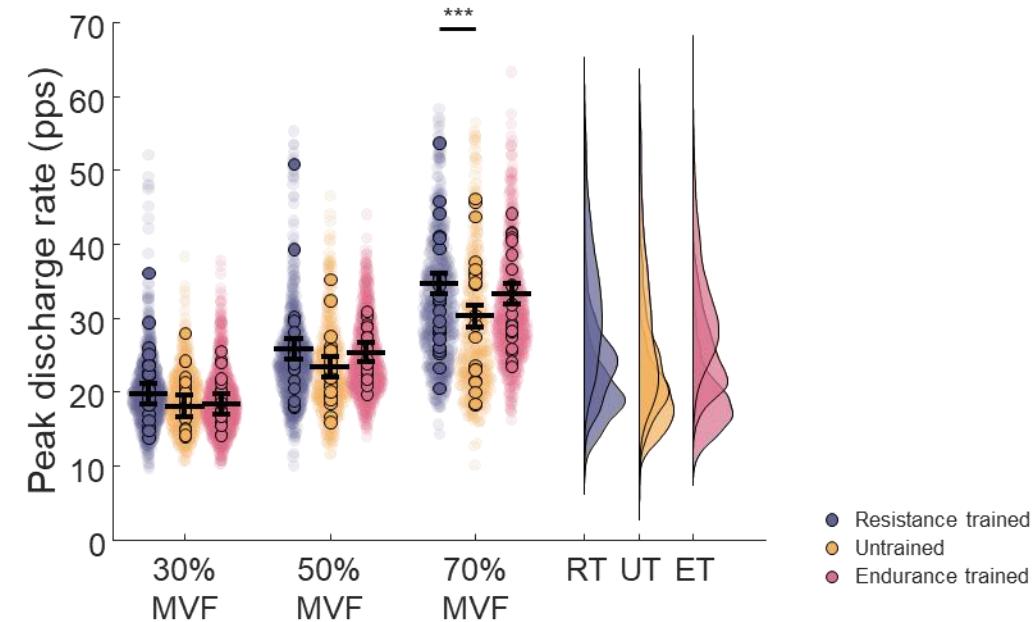


Inputs to motoneurons in chronically trained individuals

	RT (n=23)	UT (n=23)	ET (n=23)
Training Age (yrs)	9 ± 3	-	10 ± 4
IPAQ (MET-min/week)	$6401 \pm 2729^*$	4038 ± 2380	$6590 \pm 2128^{**}$
Mass (kg)	$84 \pm 17^{**}$	70 ± 12	68 ± 8
Height (m)	1.75 ± 0.08	1.73 ± 0.08	1.78 ± 0.08
Age (yrs)	23 ± 4	23 ± 3	24 ± 6

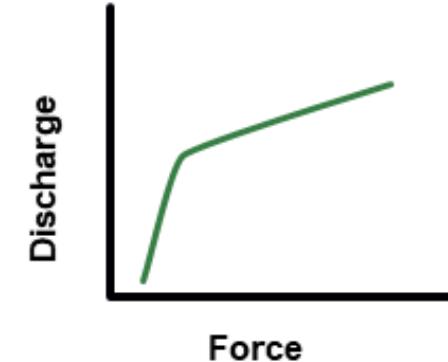
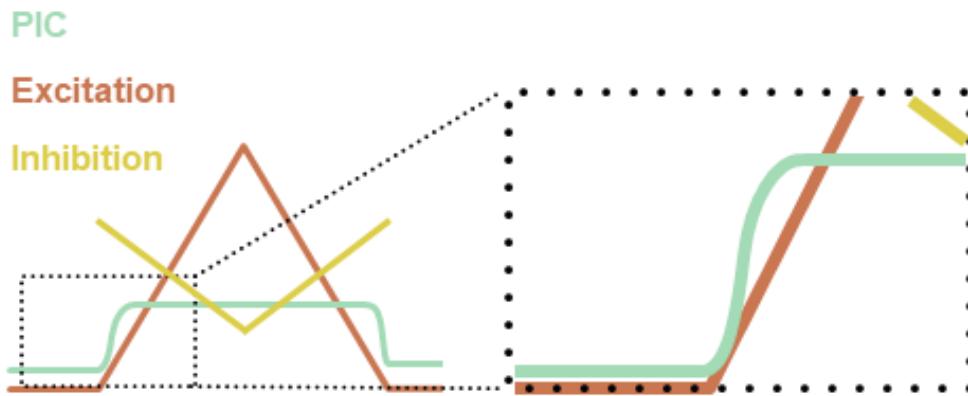


Inputs to motoneurons in chronically trained individuals

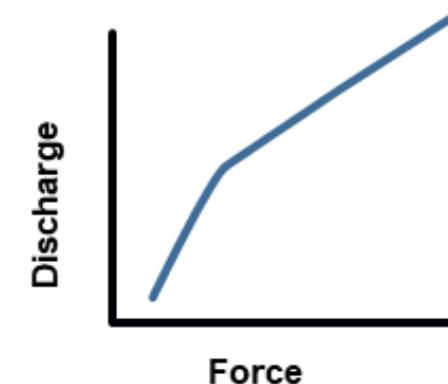
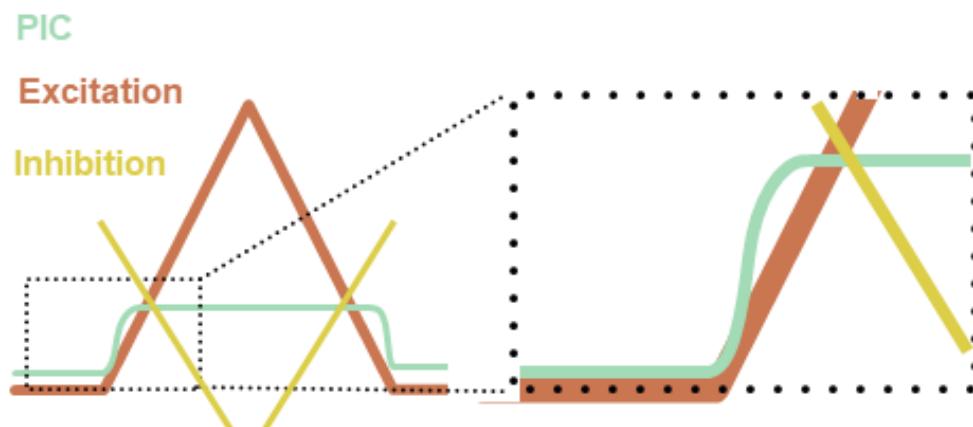


Inputs to motoneurons in chronically trained individuals

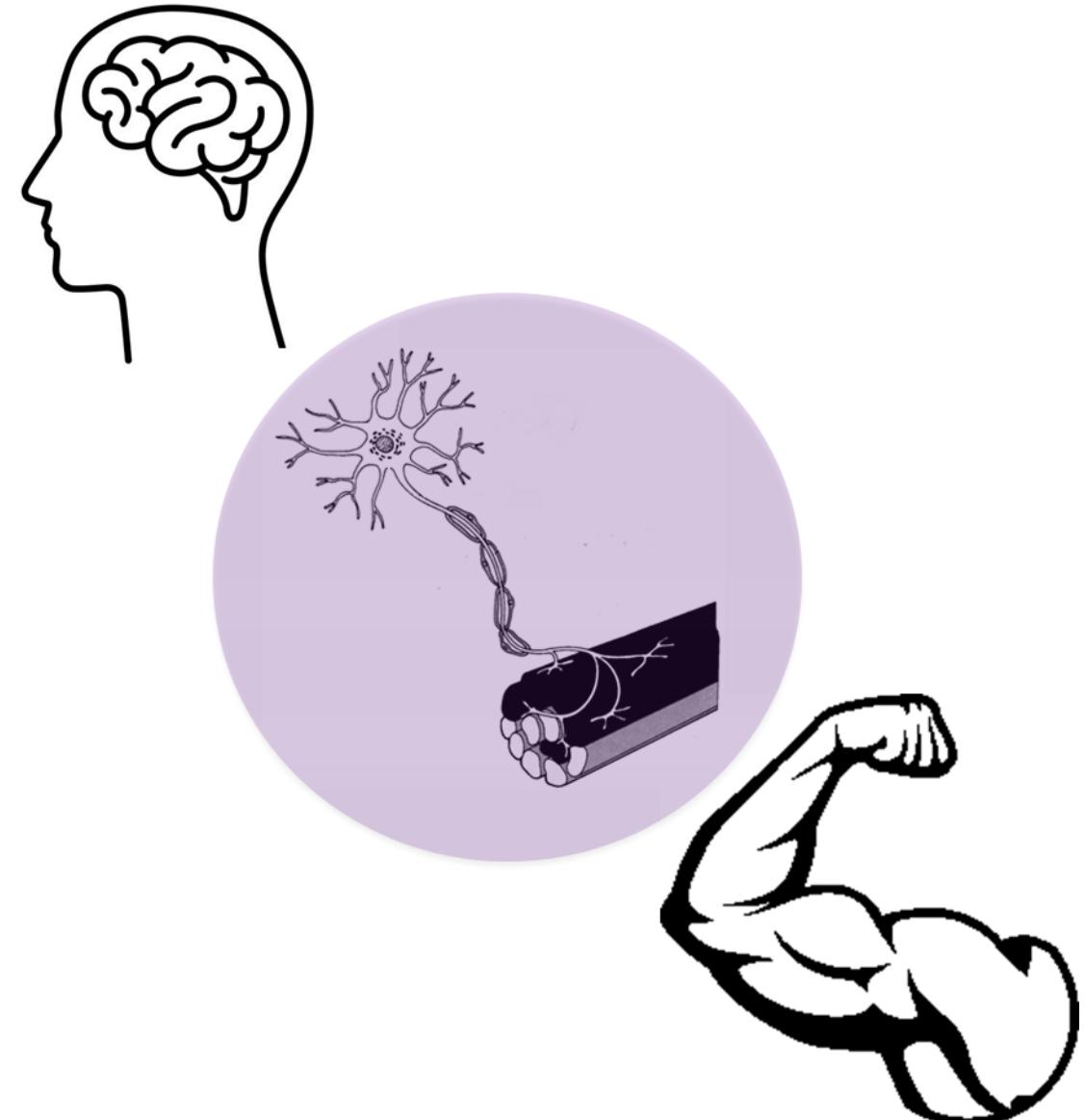
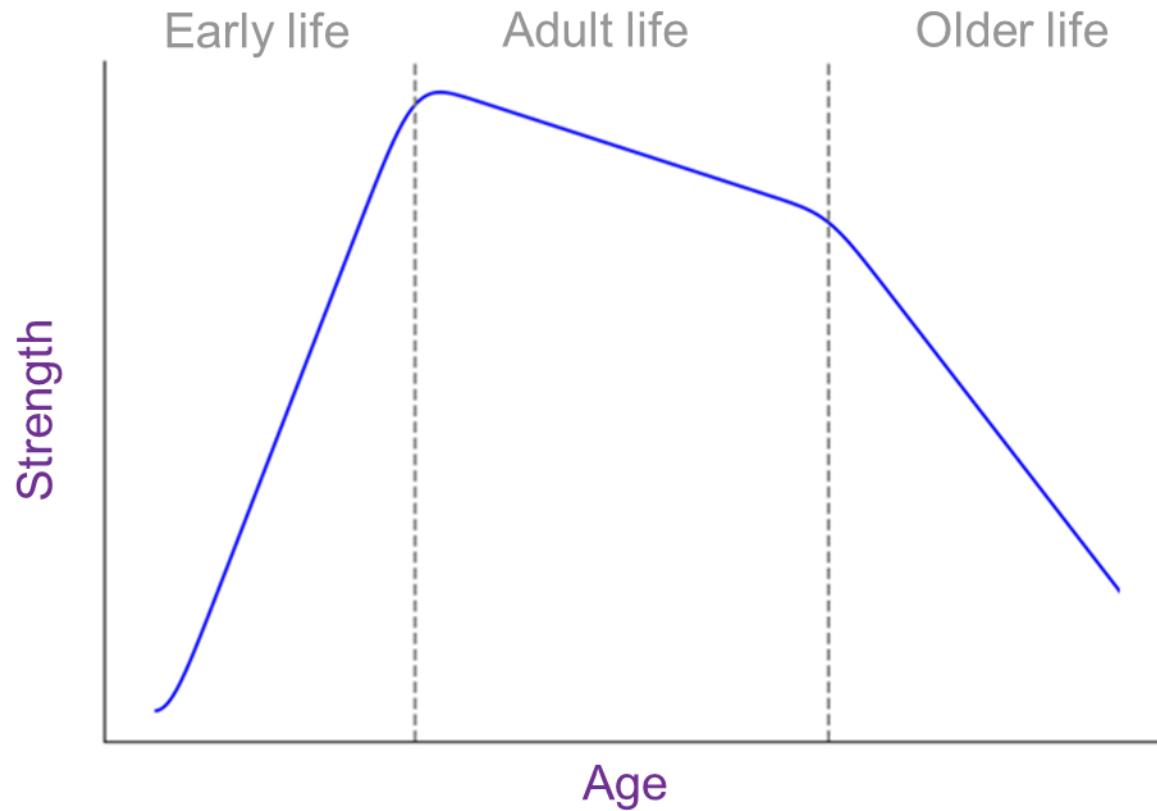
Untrained



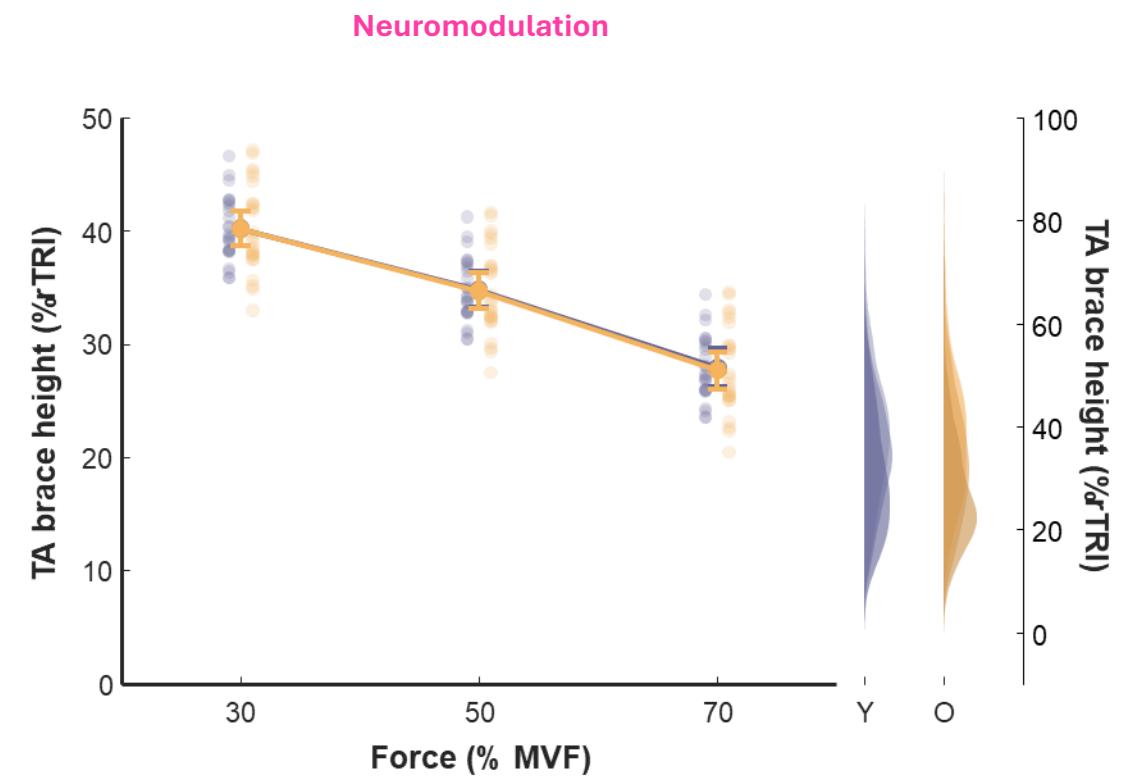
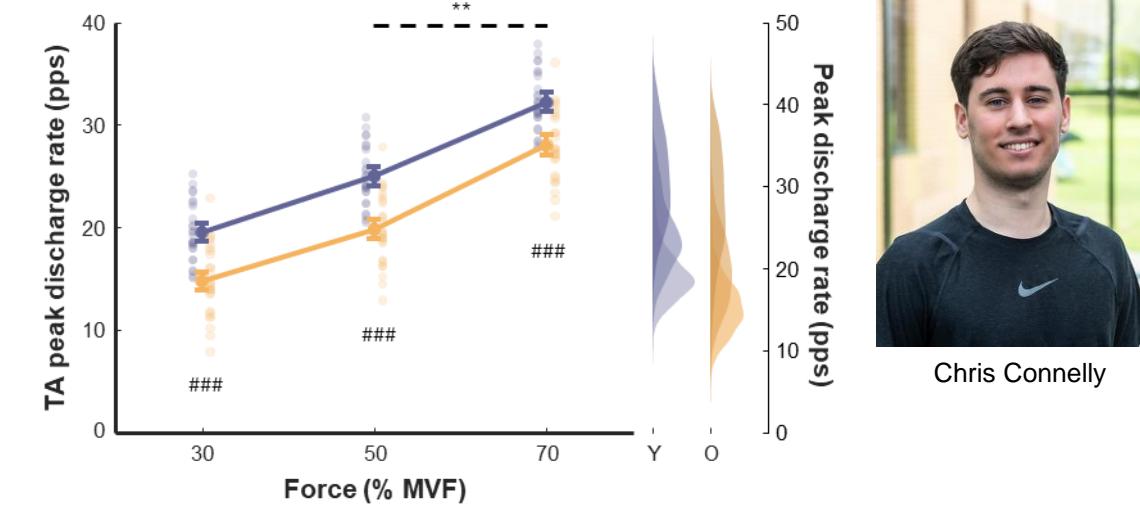
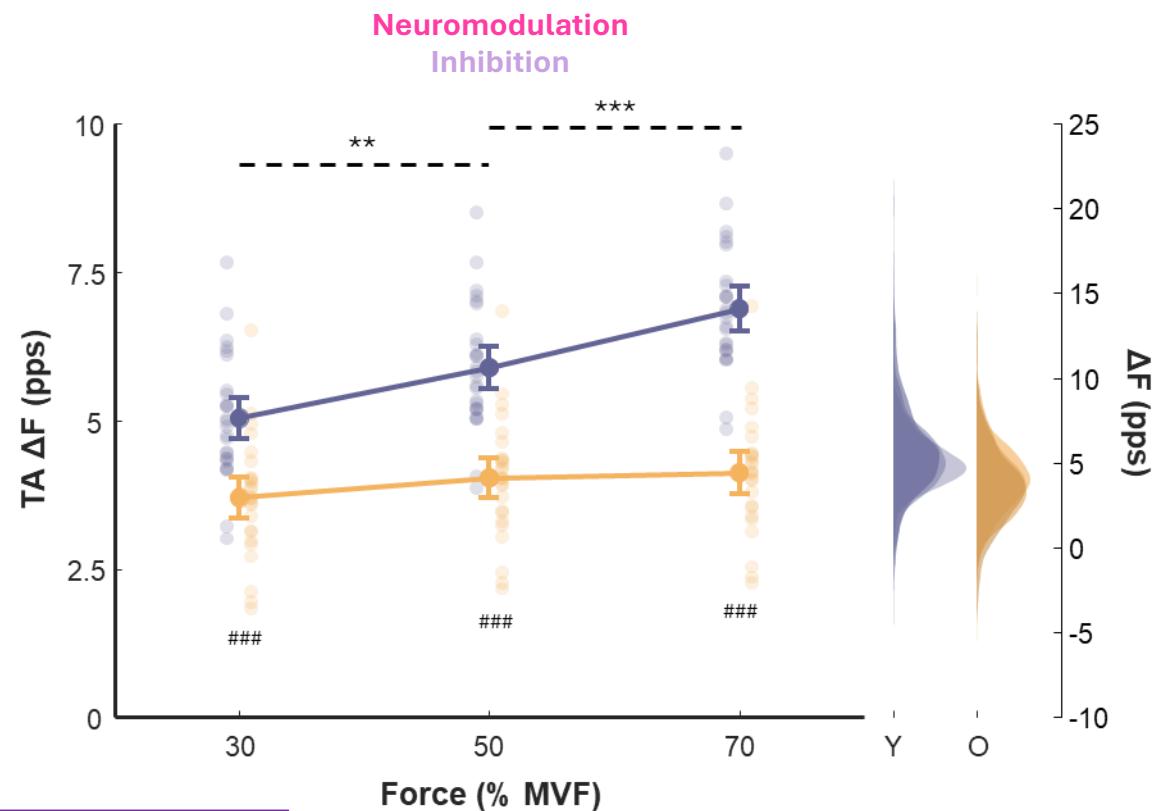
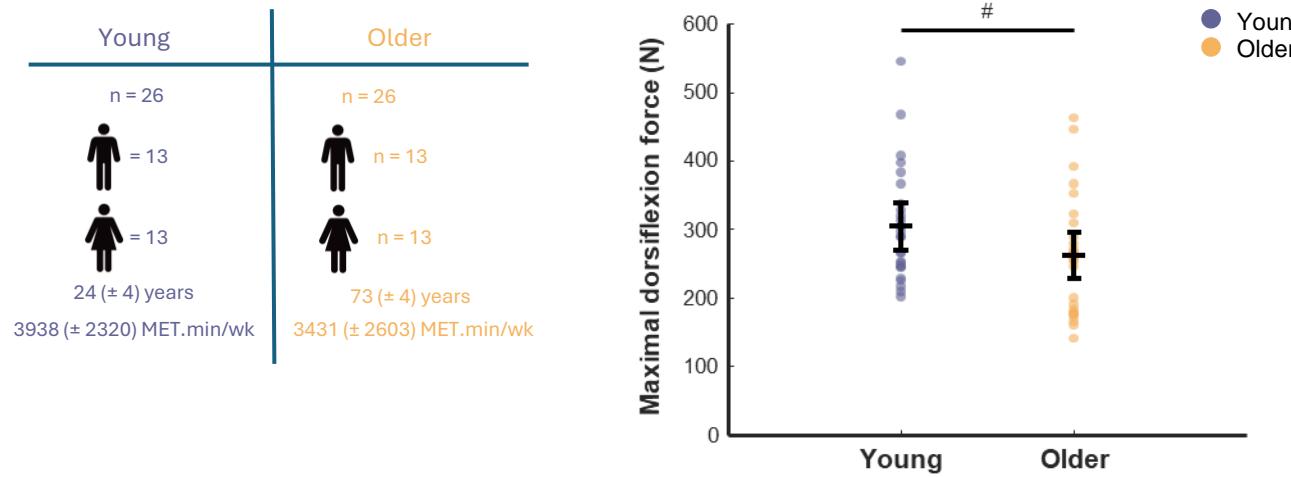
Resistance trained



Motoneuron properties in ageing individuals



(Adapted from Guyton & Hall, 1996)



Difference between groups
* Interaction contrast

Connelly et al., *in preparation*

PART 1 – RECAP

- Motor unit discharge rate should be quantified through the lens of the experimental context
 - Be mindful of selecting the epoch (spike frequency adaptation)
 - Recruitment threshold is a likely covariate to discharge rate quantification that should be considered
- Motor unit discharge rate is non-linearly related to excitation – action of persistent inward currents (PIC)
 - The effect of PICs, i.e. amplification and prolongation, may be quantified with a geometric analysis (acceleration) and a paired motor unit analysis (prolongation)
 - Be mindful of the types of inputs that may influence these metrics (as well as the experimental protocol/conditions/sample population)

Questions / break / continue?



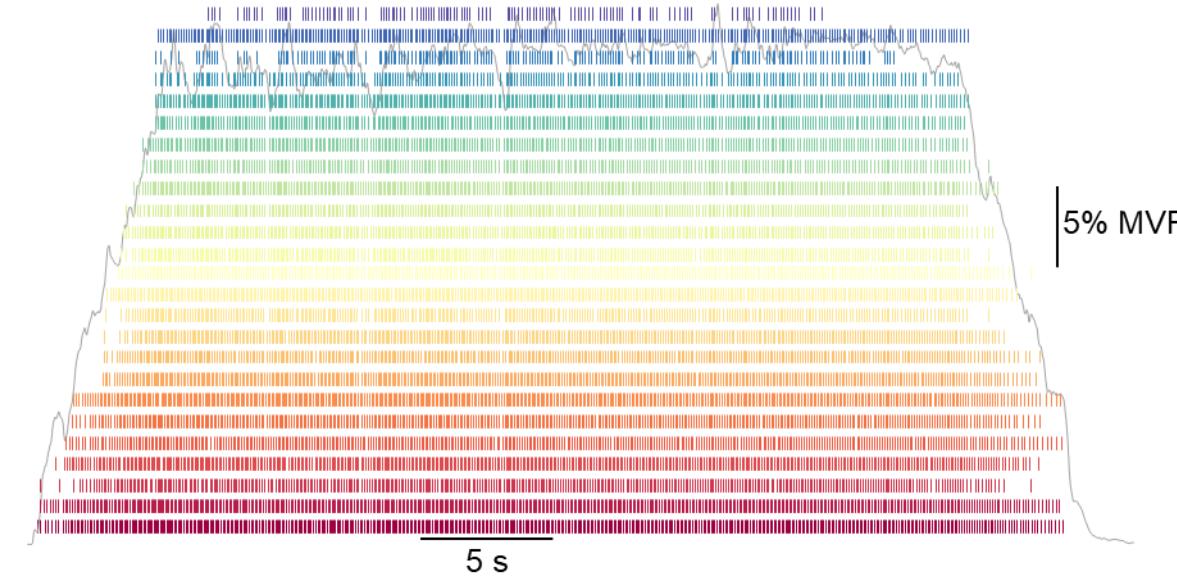
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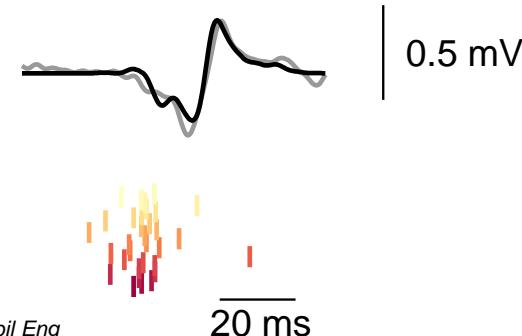
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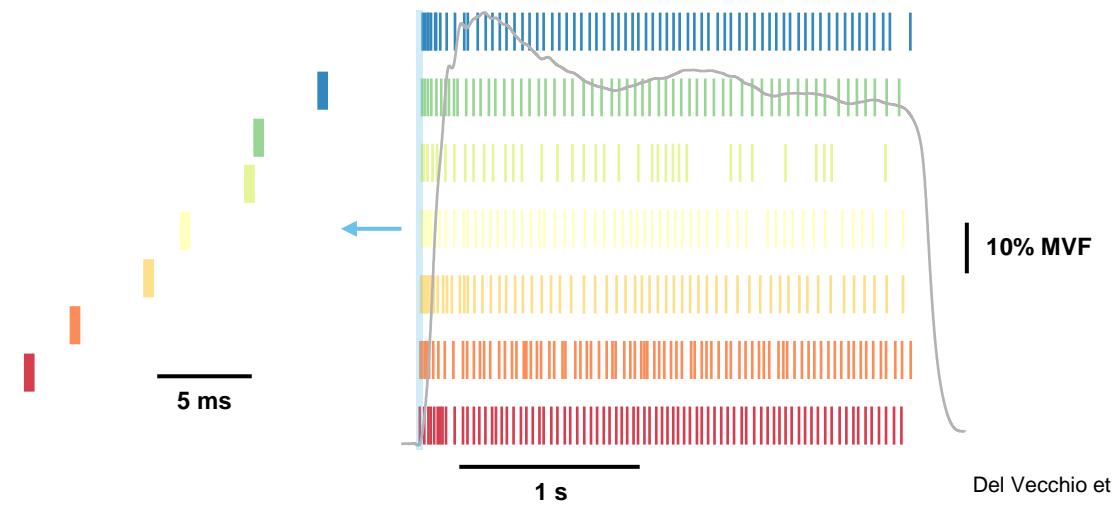
HDsEMG decomposition – Asynchronous vs. synchronous firing



Evoked contractions

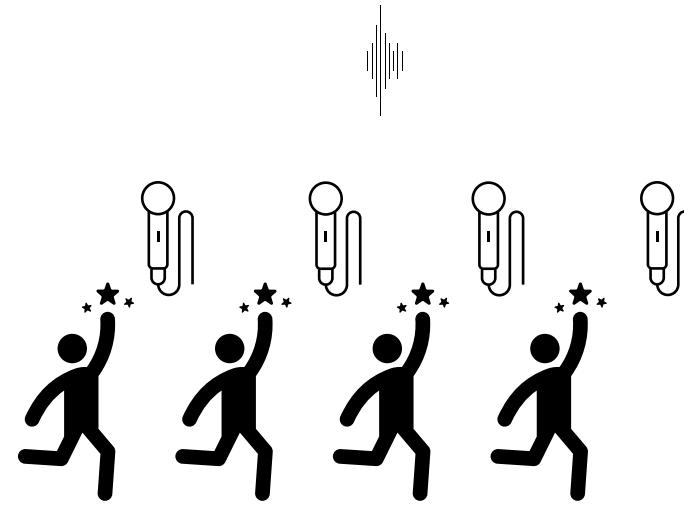
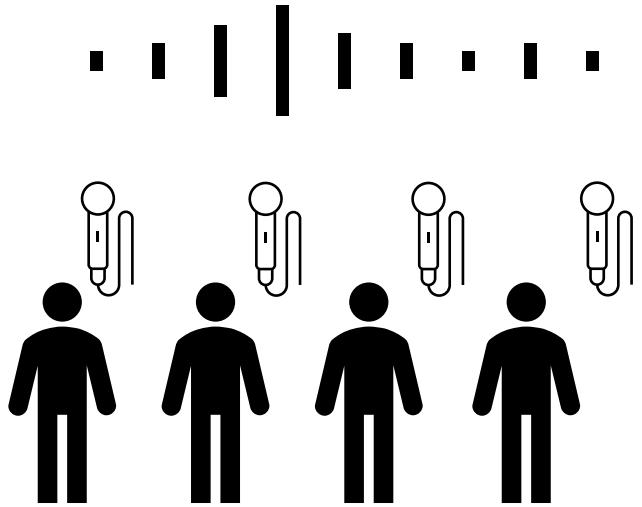


Rapid (isometric) contractions



PART 2

- Identification of motor unit discharges during evoked contractions
- Challenges in estimating motor unit recruitment thresholds in conditions of a compressed motor unit recruitment range
- Quantification of motor unit discharge behaviour during voluntary rapid contractions and maximal efforts

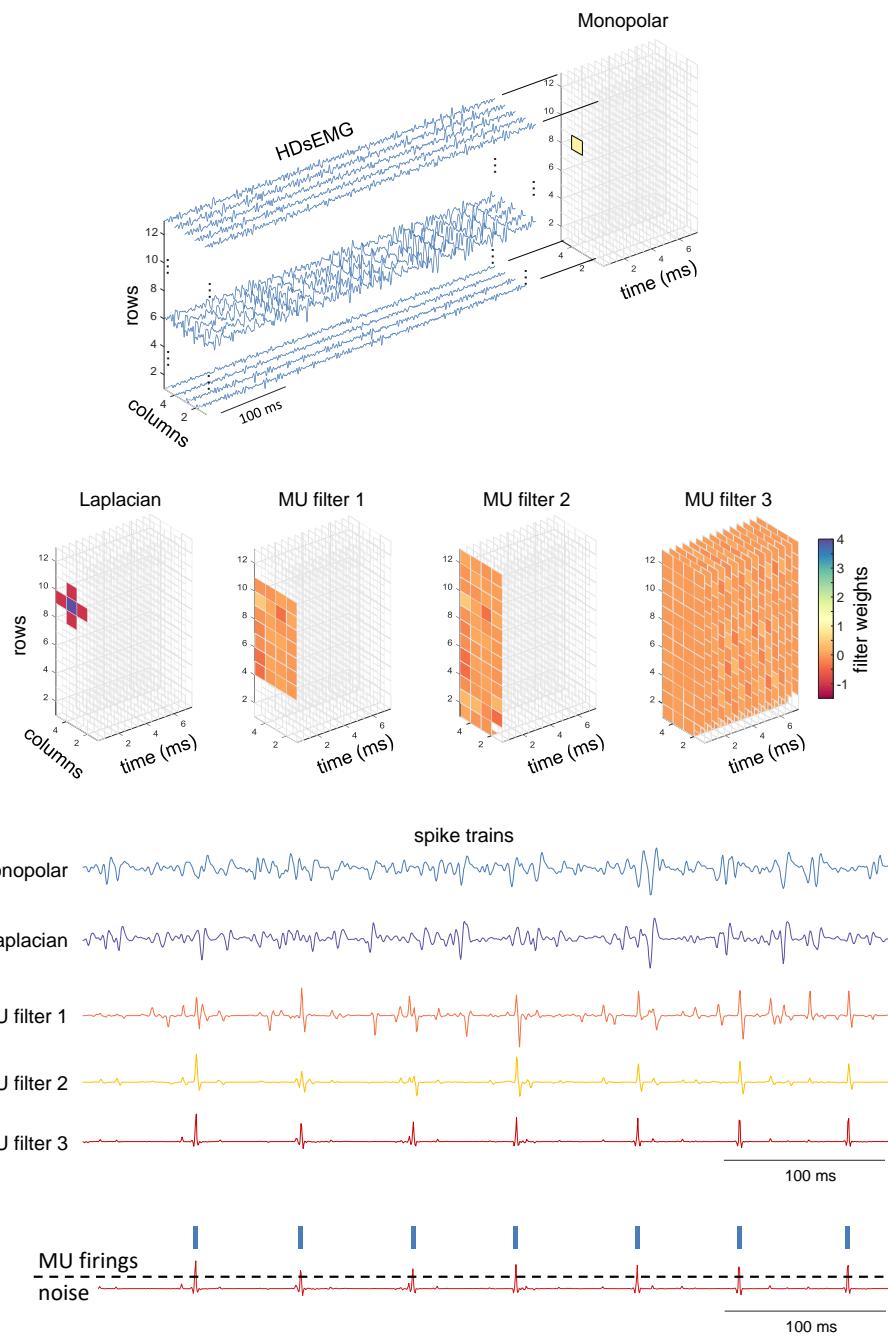
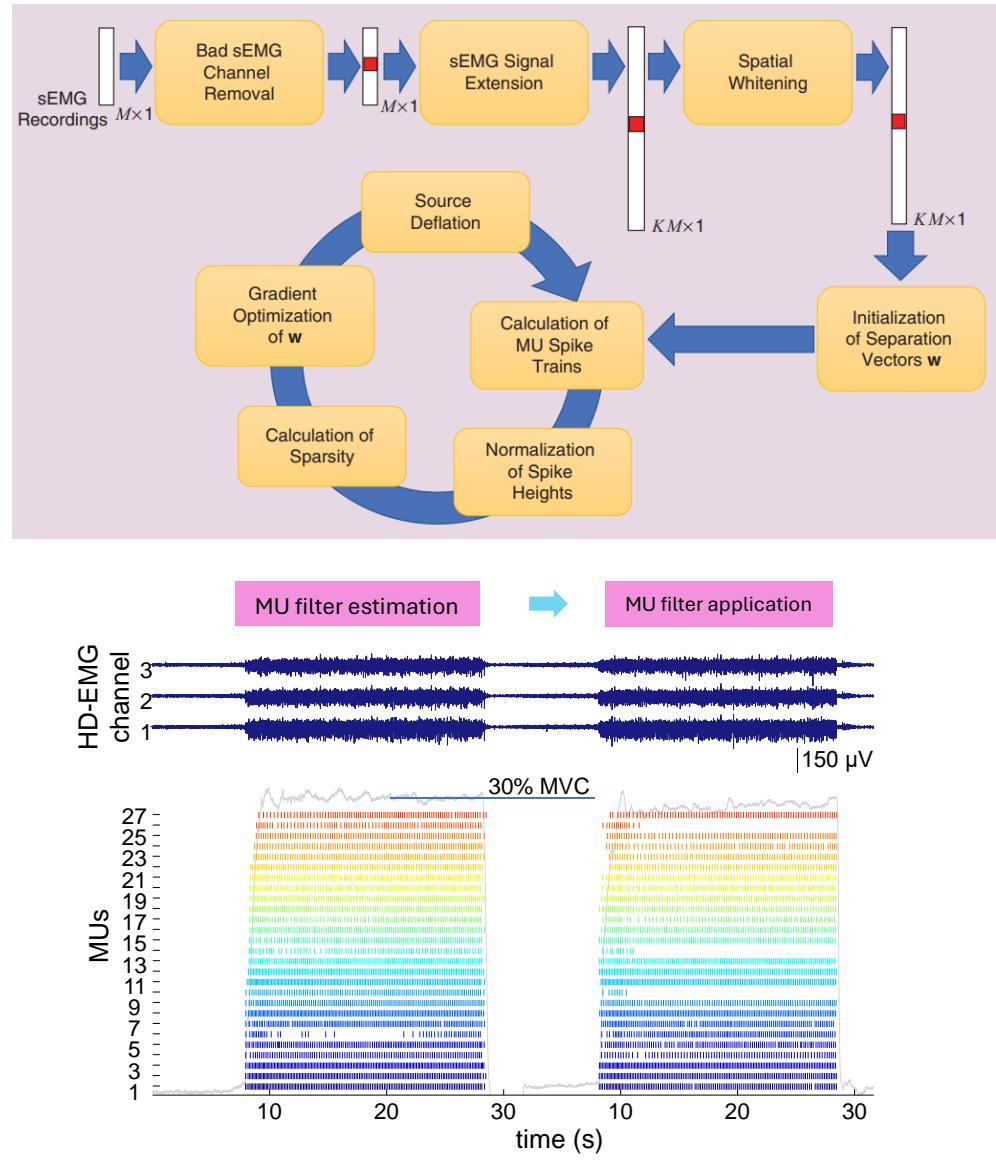


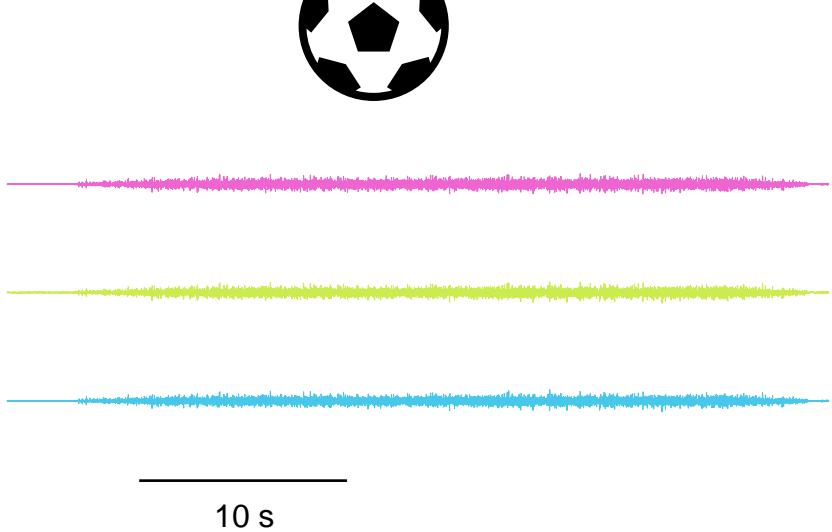
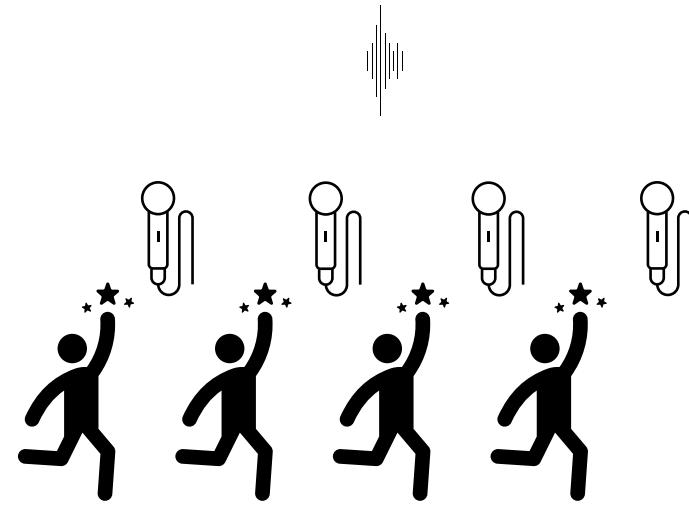
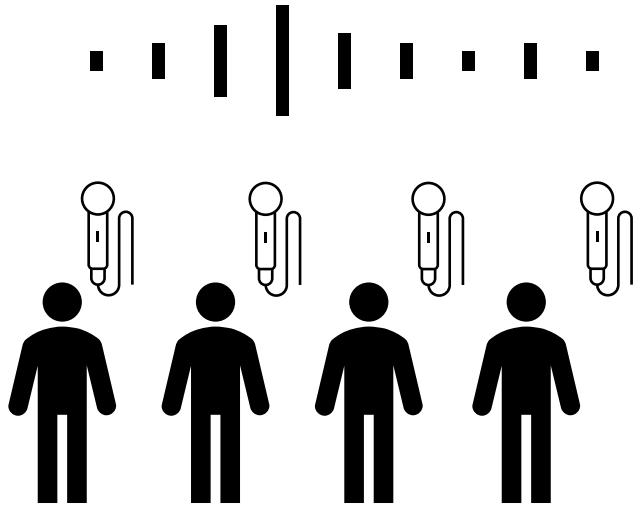
10 s



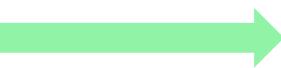
500 ms

Separation vectors – MU filters



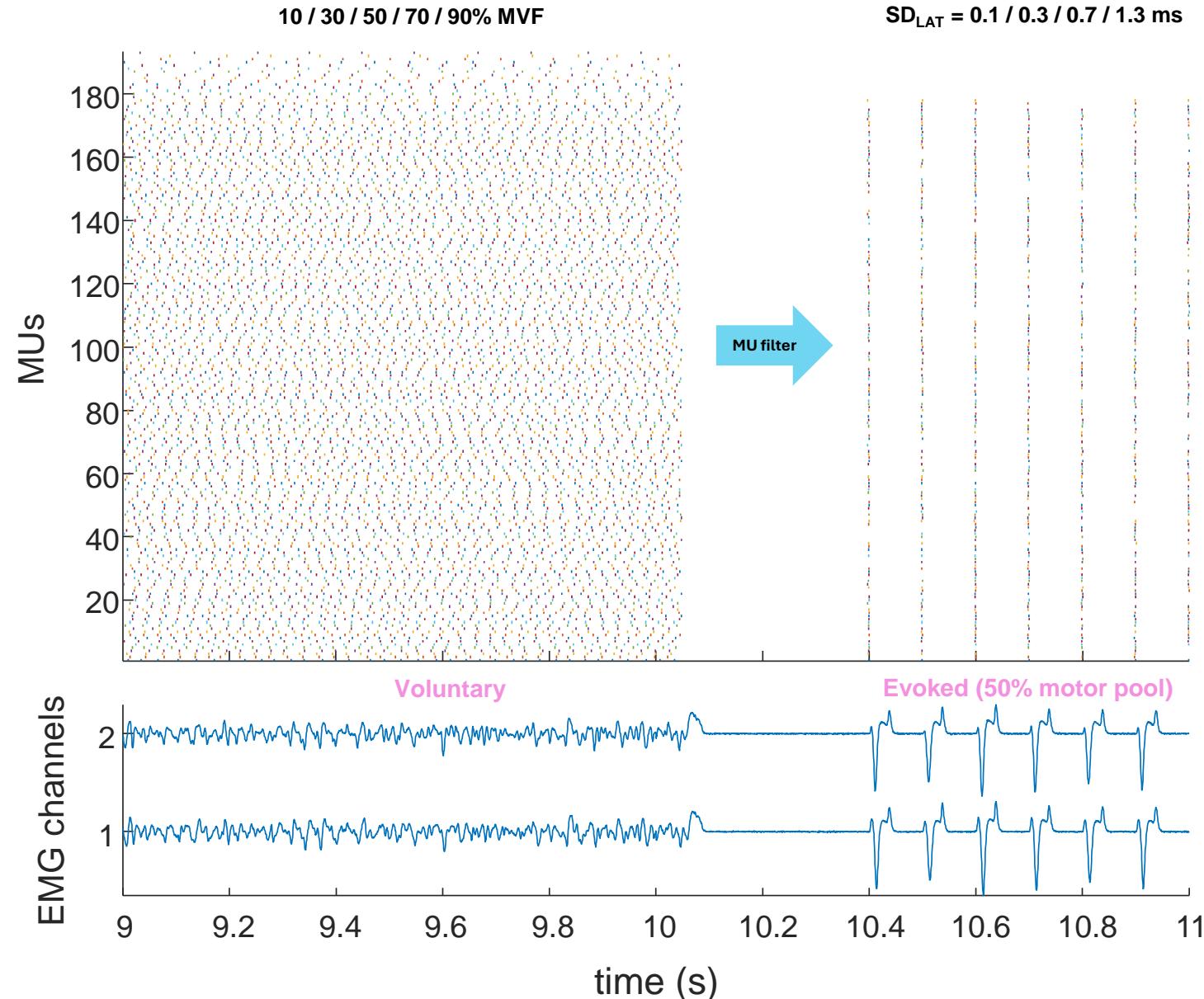


MU filter estimation

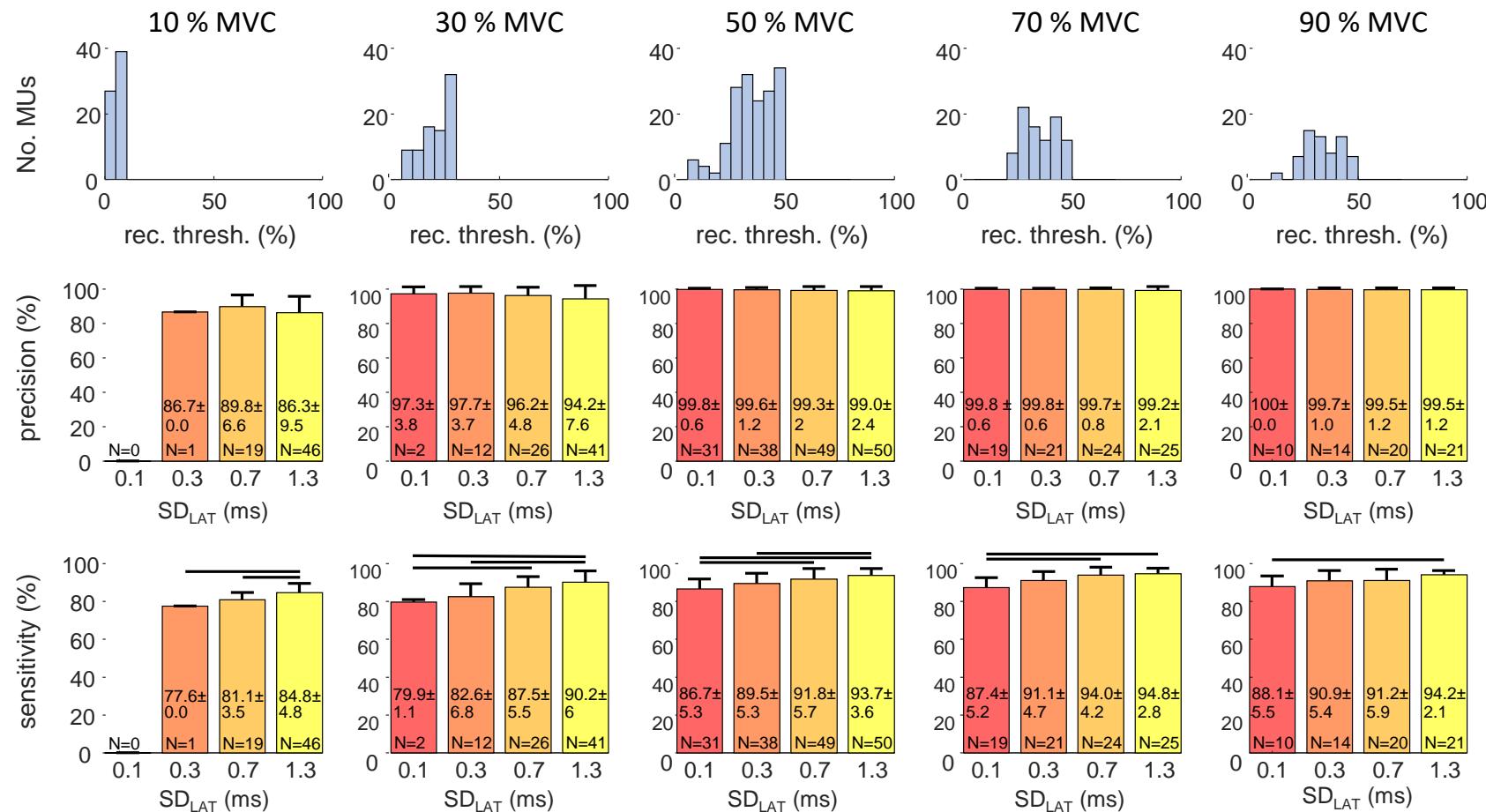


MU filter application

MU identification during high synchronisation levels

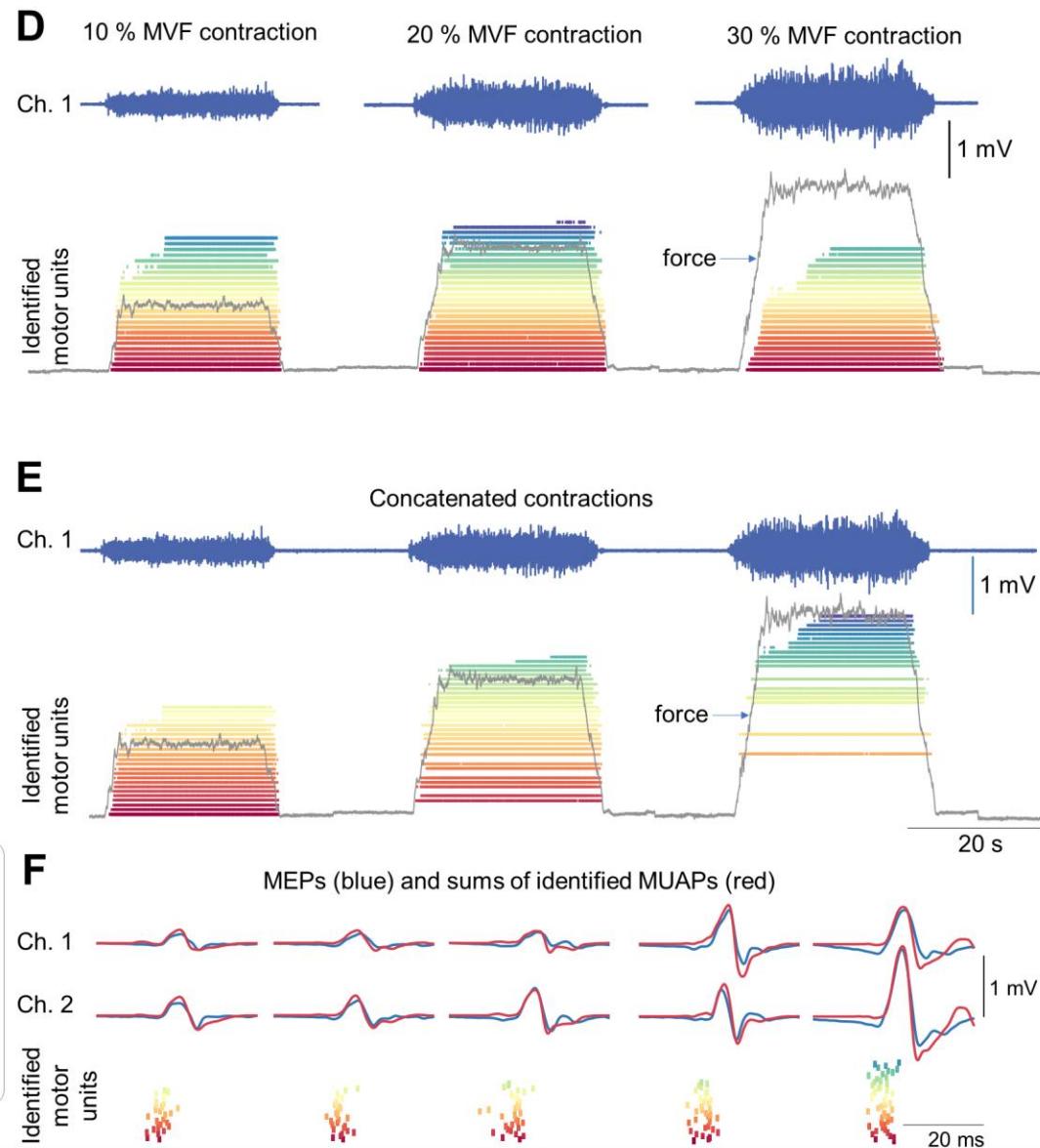
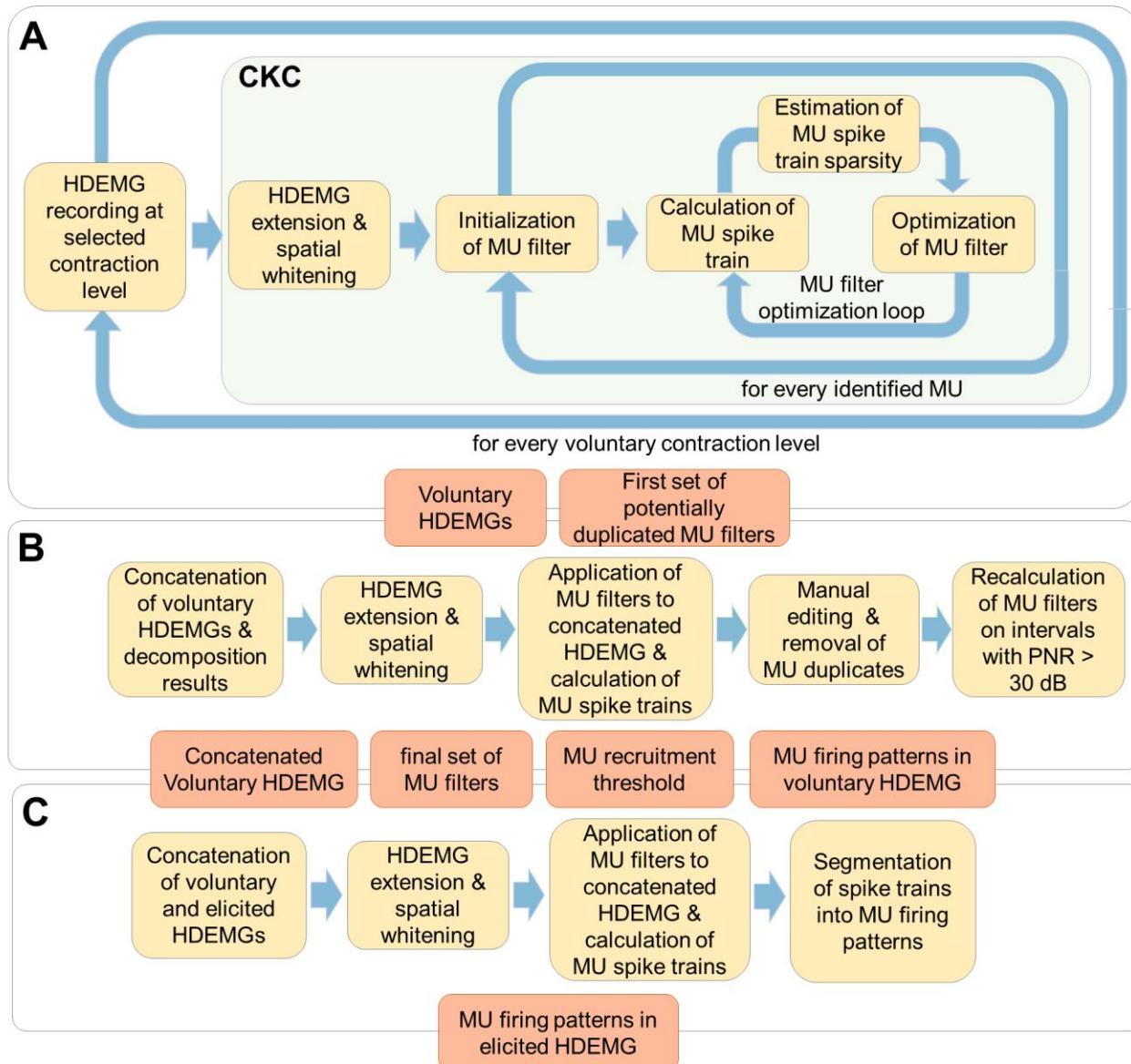


MU identification during varying synchronisation levels

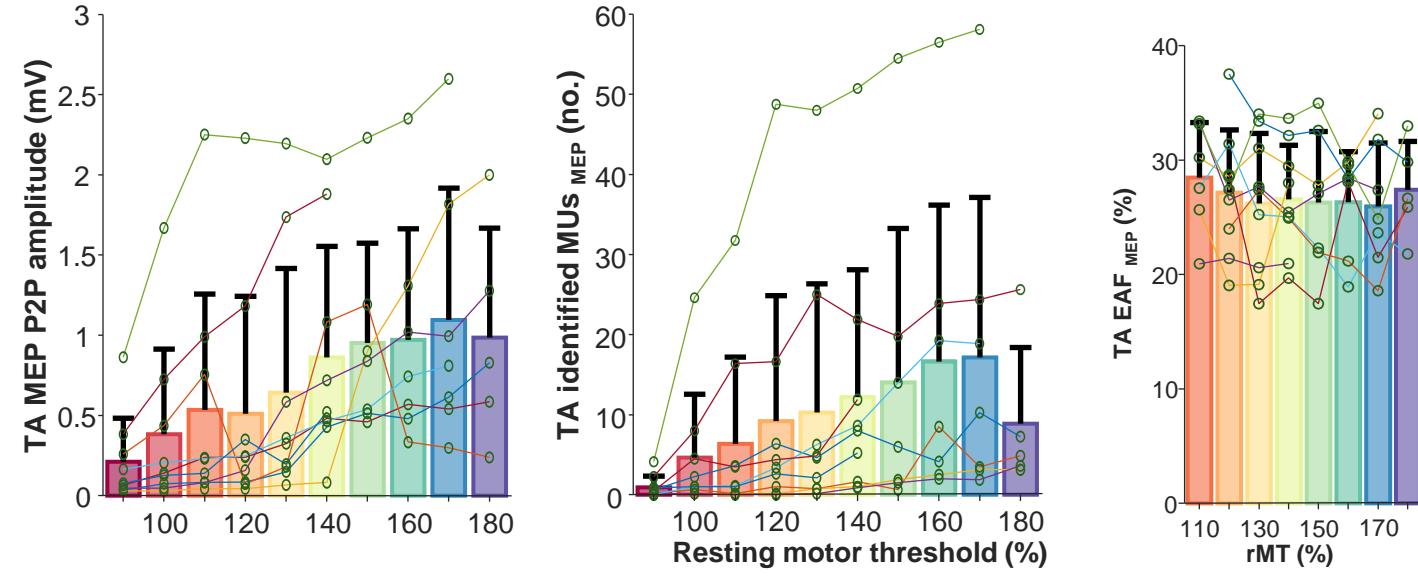
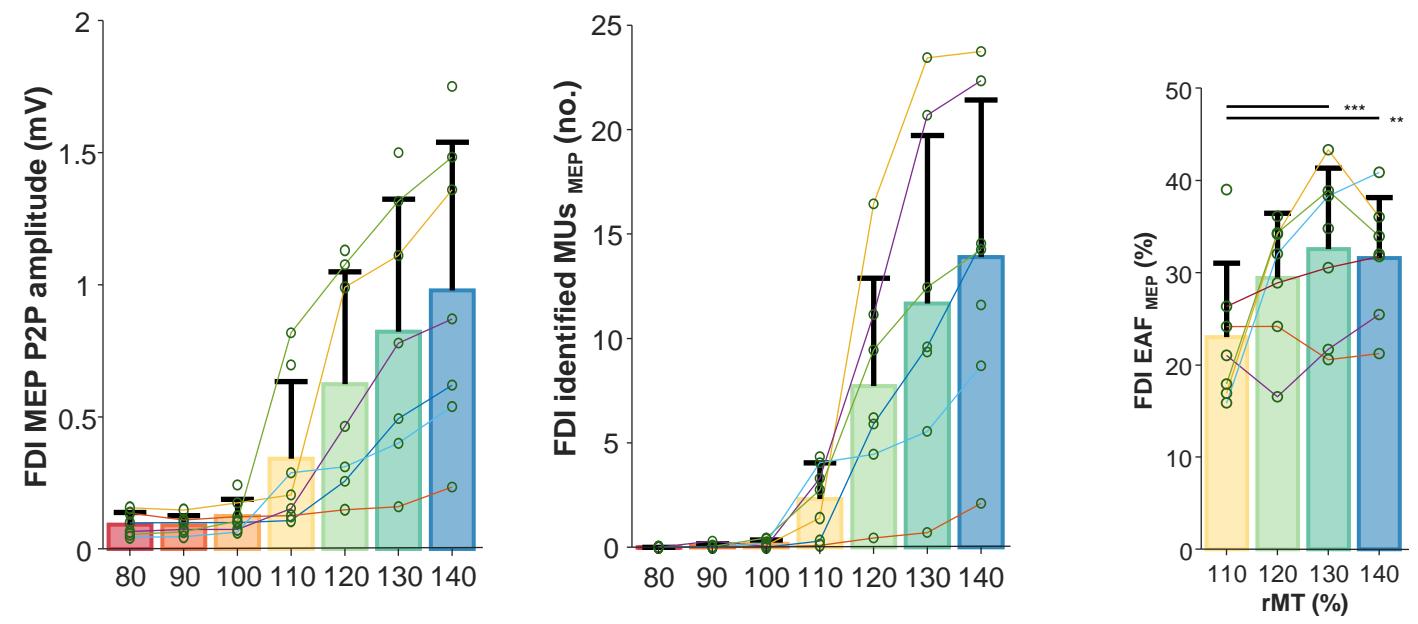
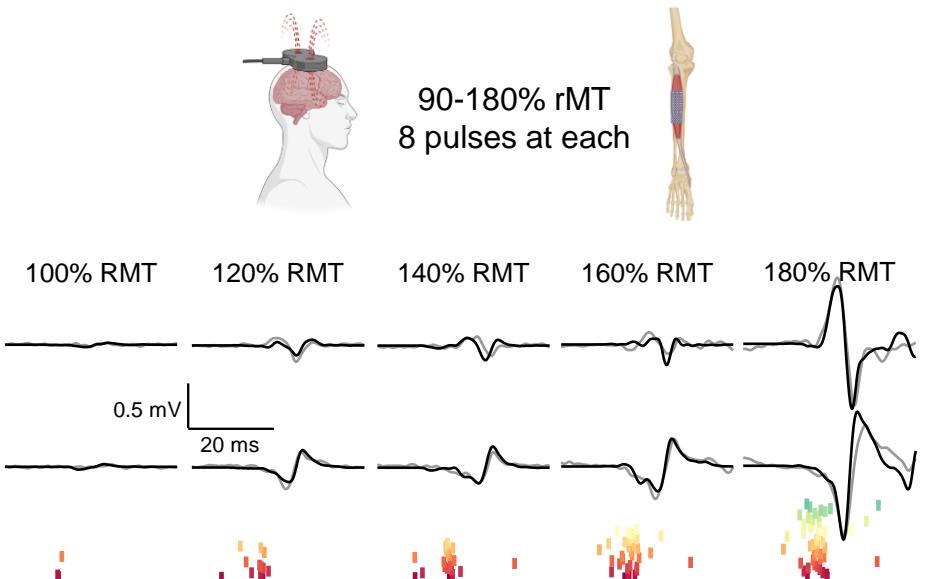
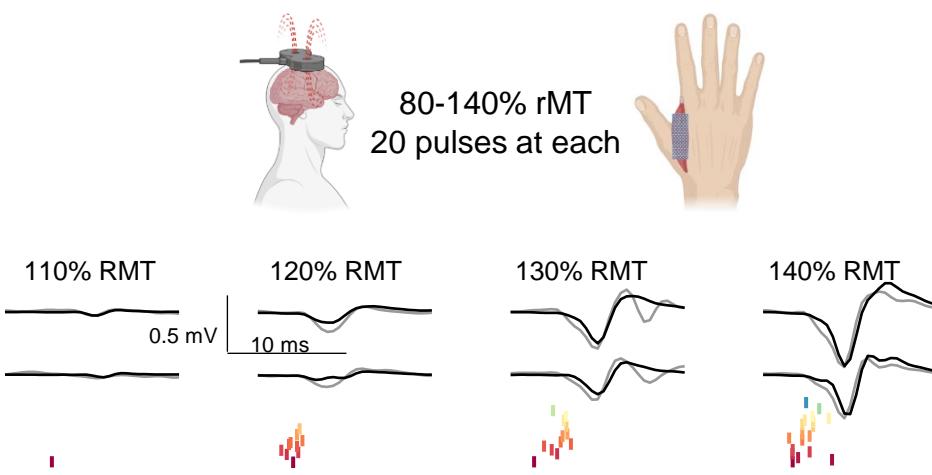


- No identified MUs with recruitment threshold >50% MVC → likely low # false negatives
- High levels of precision throughout.
- Comparatively lower sensitivity, indicating some firings might be missed.

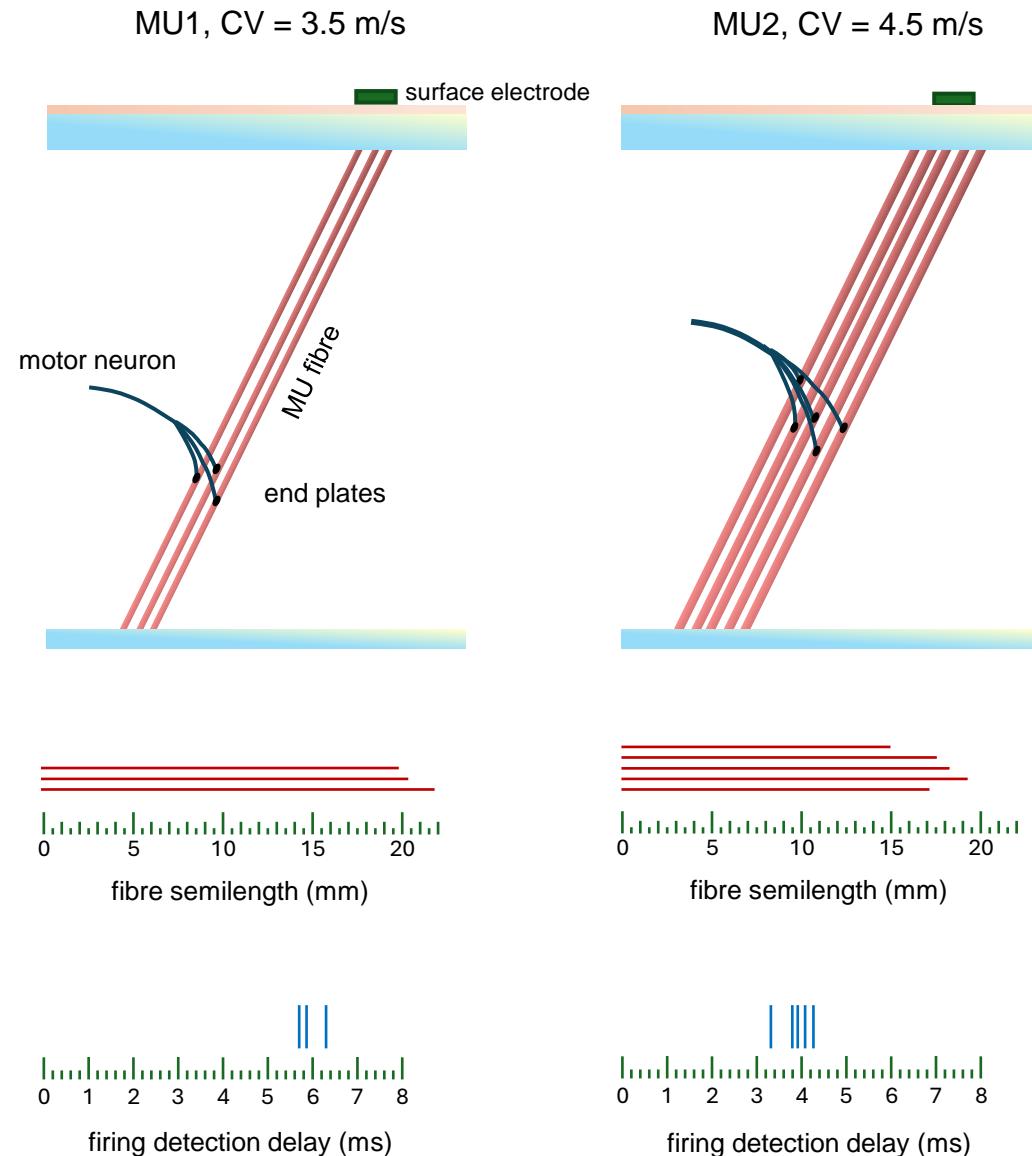
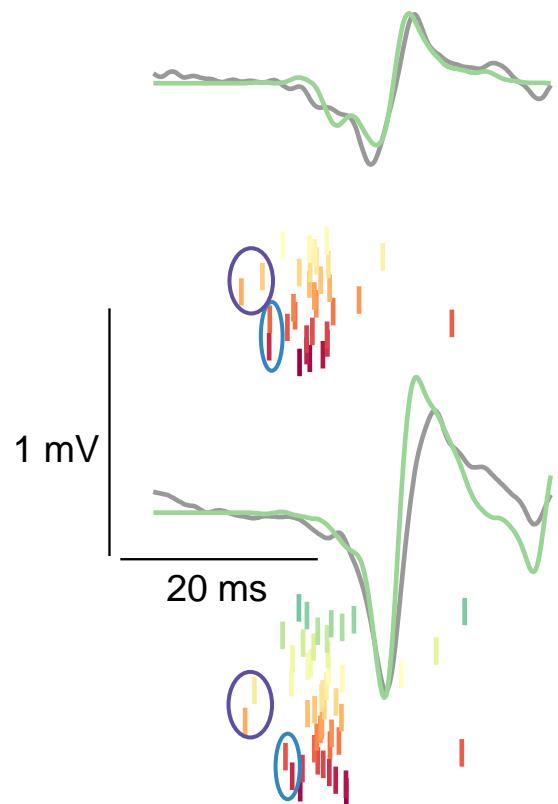
Identification of MU firings during evoked contractions – the pipeline



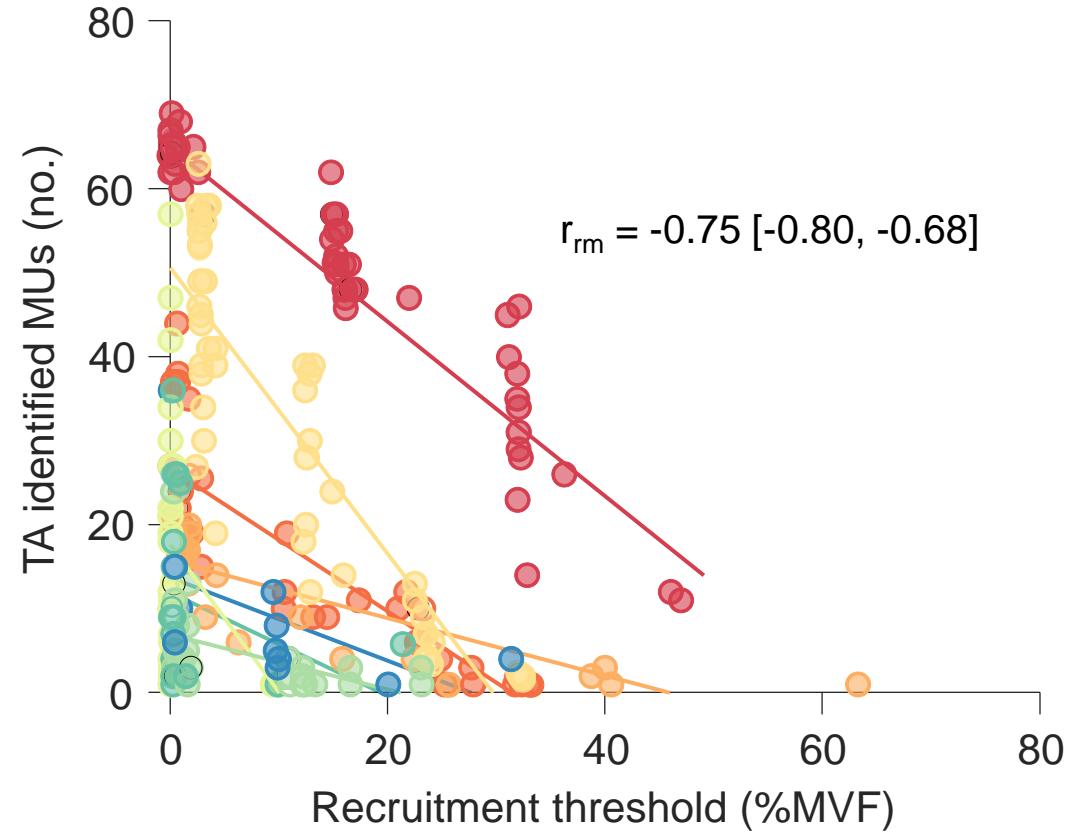
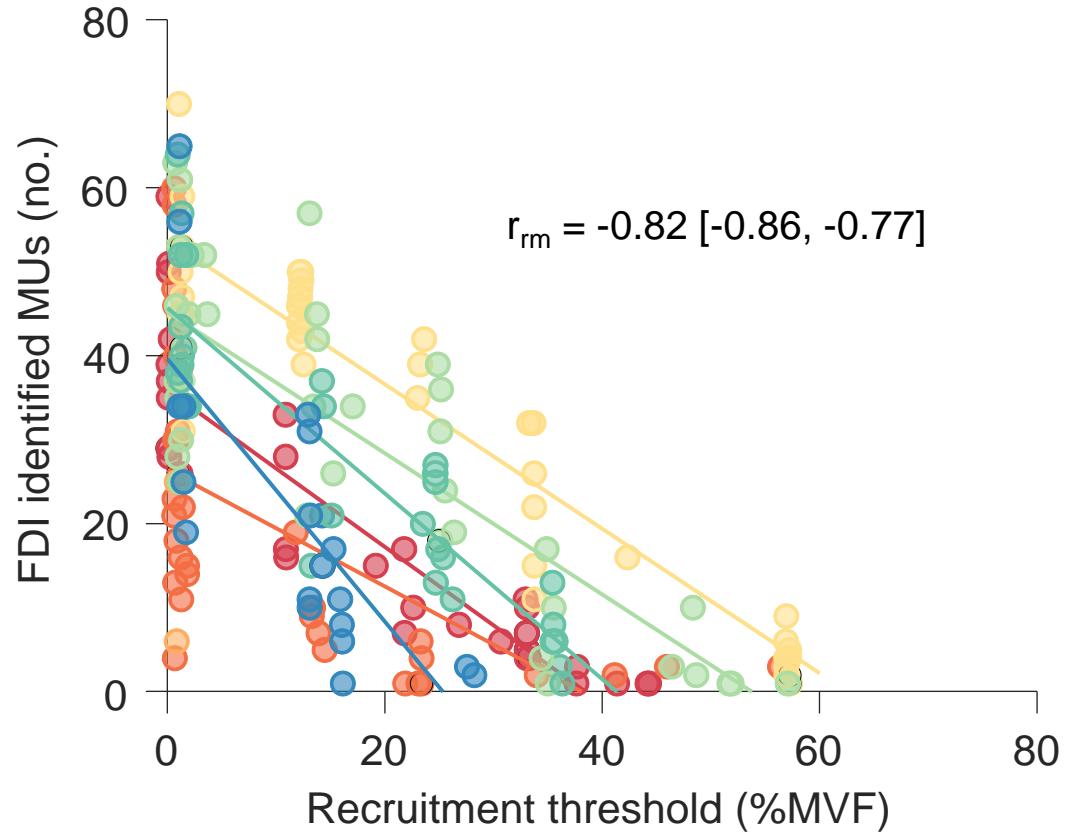
Experimental signals

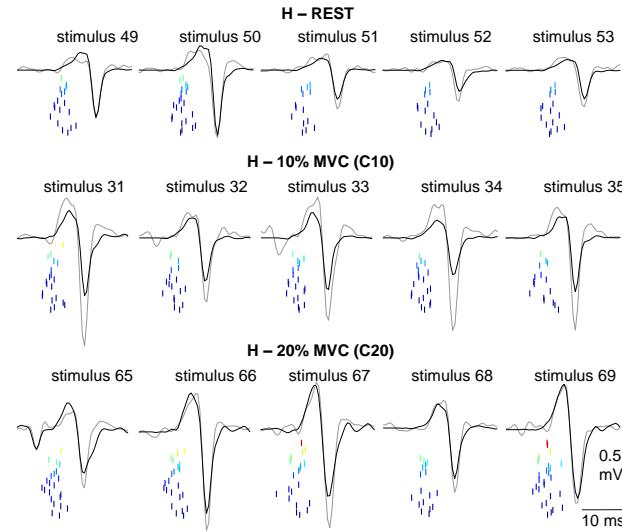
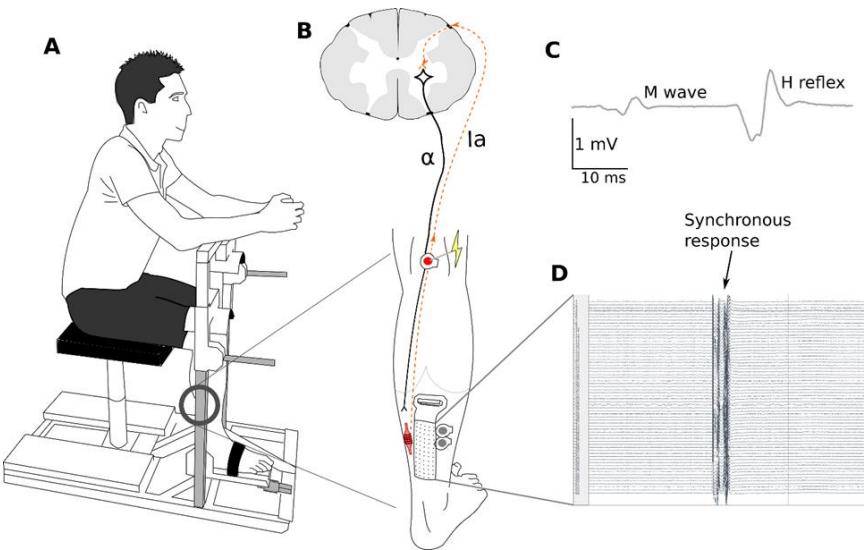


Estimation of recruitment threshold

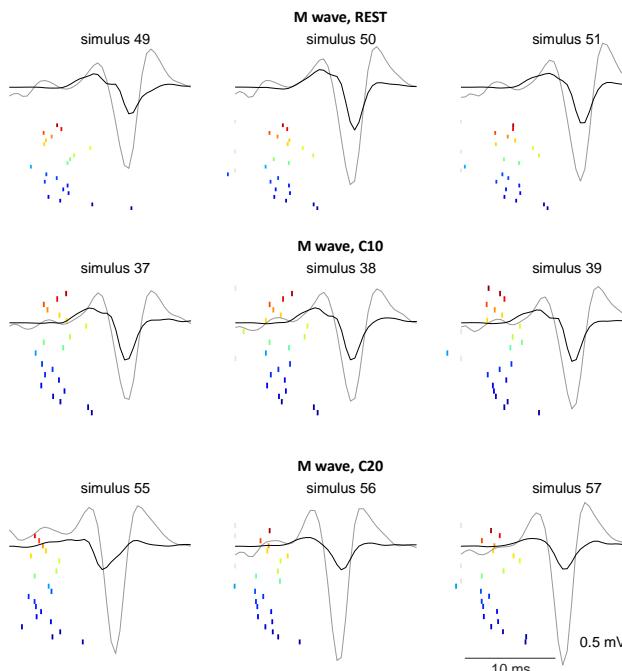
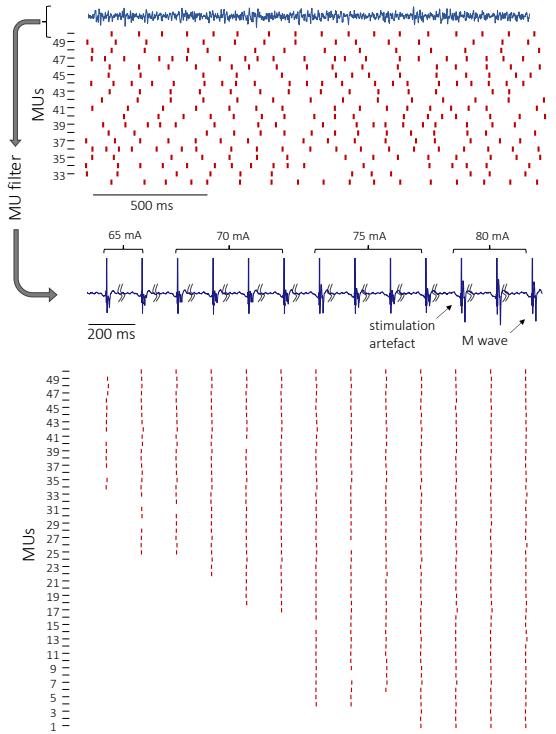


Recruitment order





Kalc et al. 2023, IEEE Trans Neural Syst Rehabilitation Eng



Kalc et al. 2023, IEEE Trans Biomed Eng

Identification of MU firings during evoked contractions from surface EMG

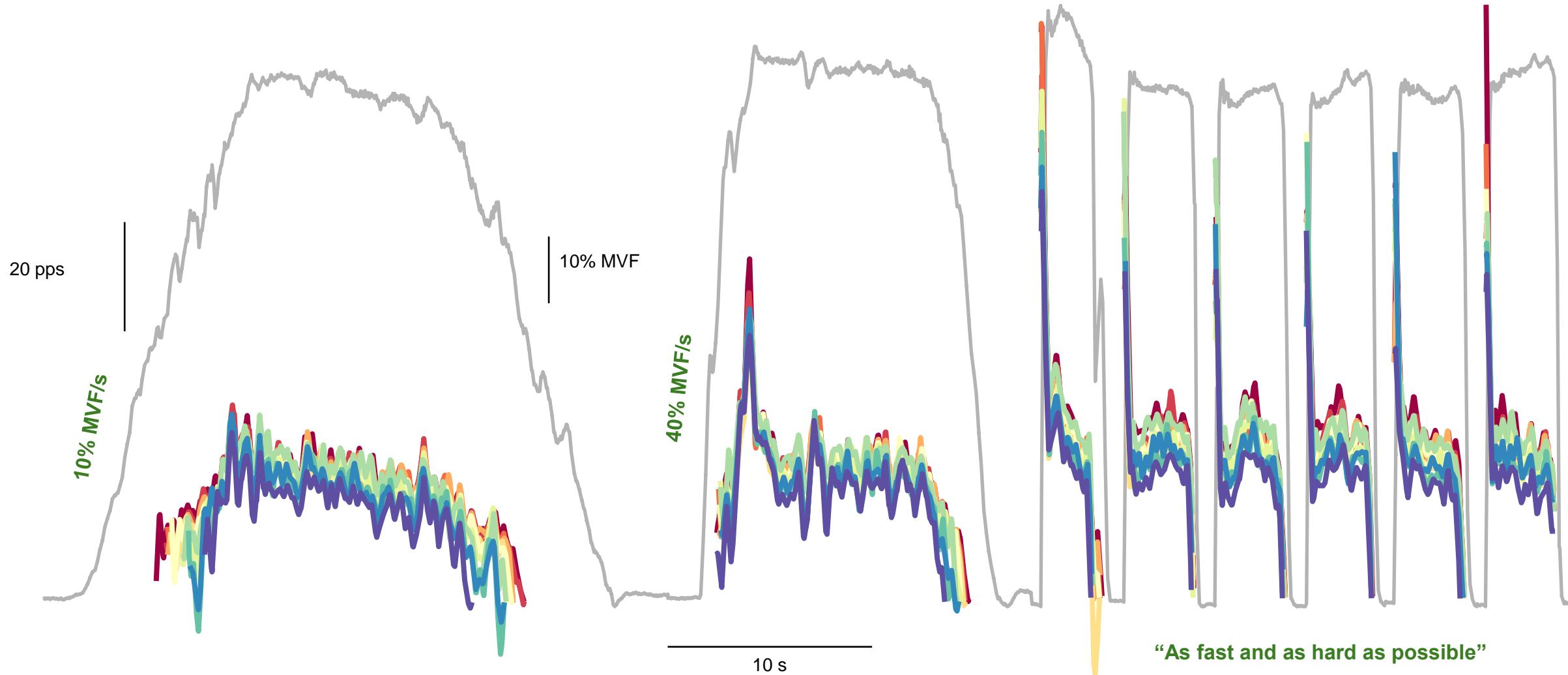
STRENGTHS

- Non-invasive investigation of responses to TMS at the level of single MUs
- Many identified MUs (wide recruitment range)
- Fewer stimuli needed
- Possibility to track units?

WEAKNESSES

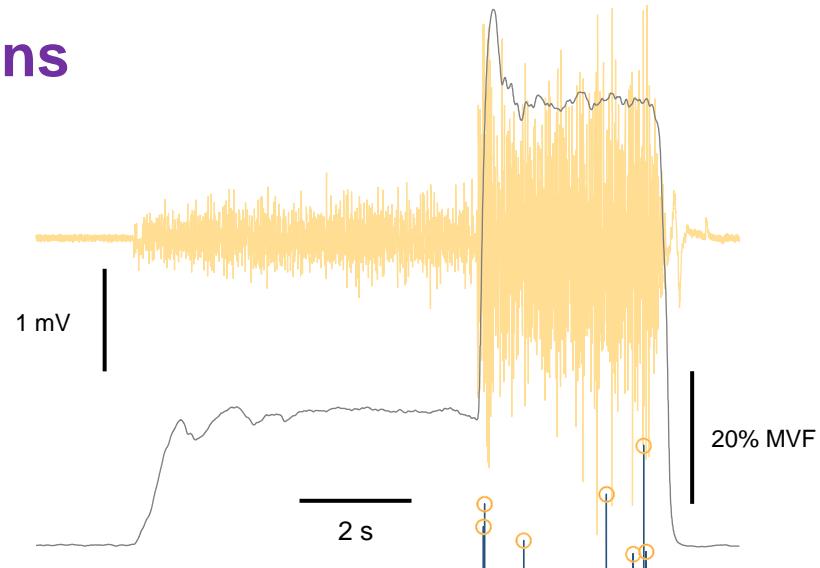
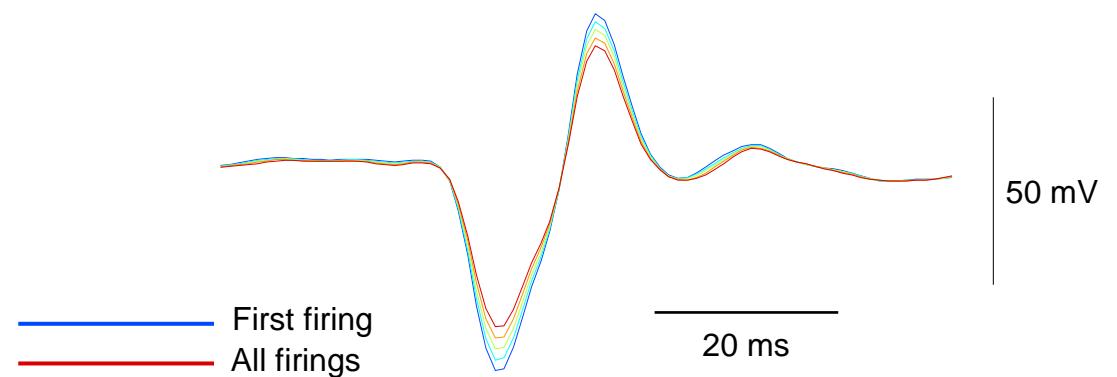
- Limited to MUs identified during voluntary efforts
- Limitations of HDsEMG decomposition with BSS still apply:
 - Inability to identify MUs further away from recording site
 - Difficulty in segmenting higher-threshold units due to superimposition of MUAPs
 - Inherent bias towards large MUs with large MUAPs

Feedback (ramp) vs. feedforward (rapid) contractions



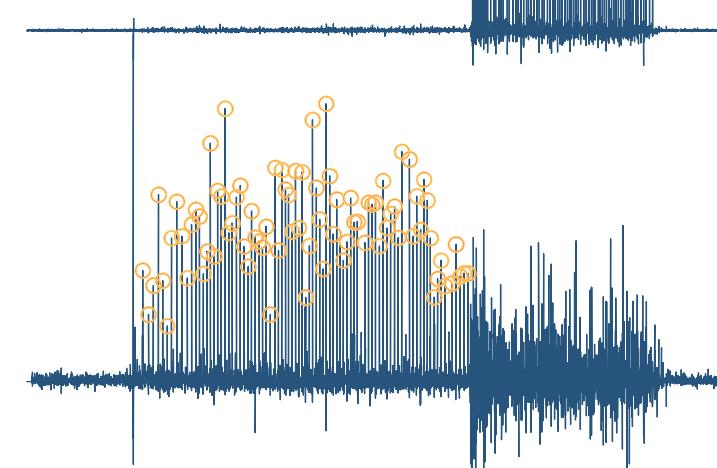
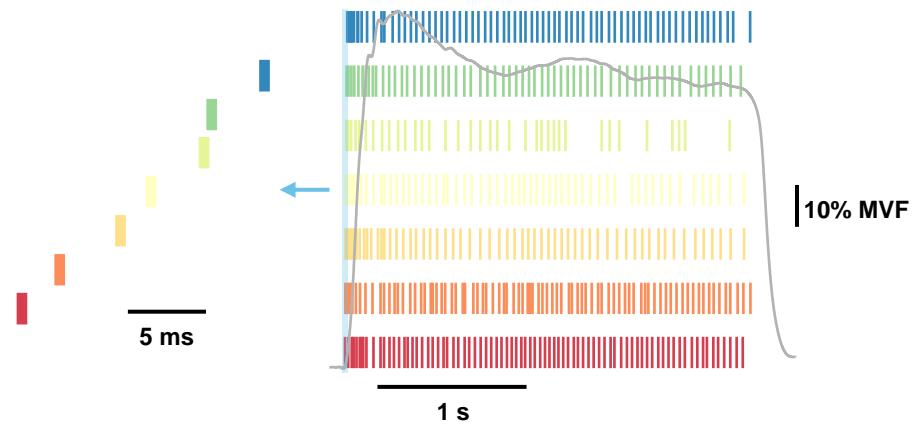
MU identification during rapid (isometric) contractions

Challenge 1: The change in MUAP waveform

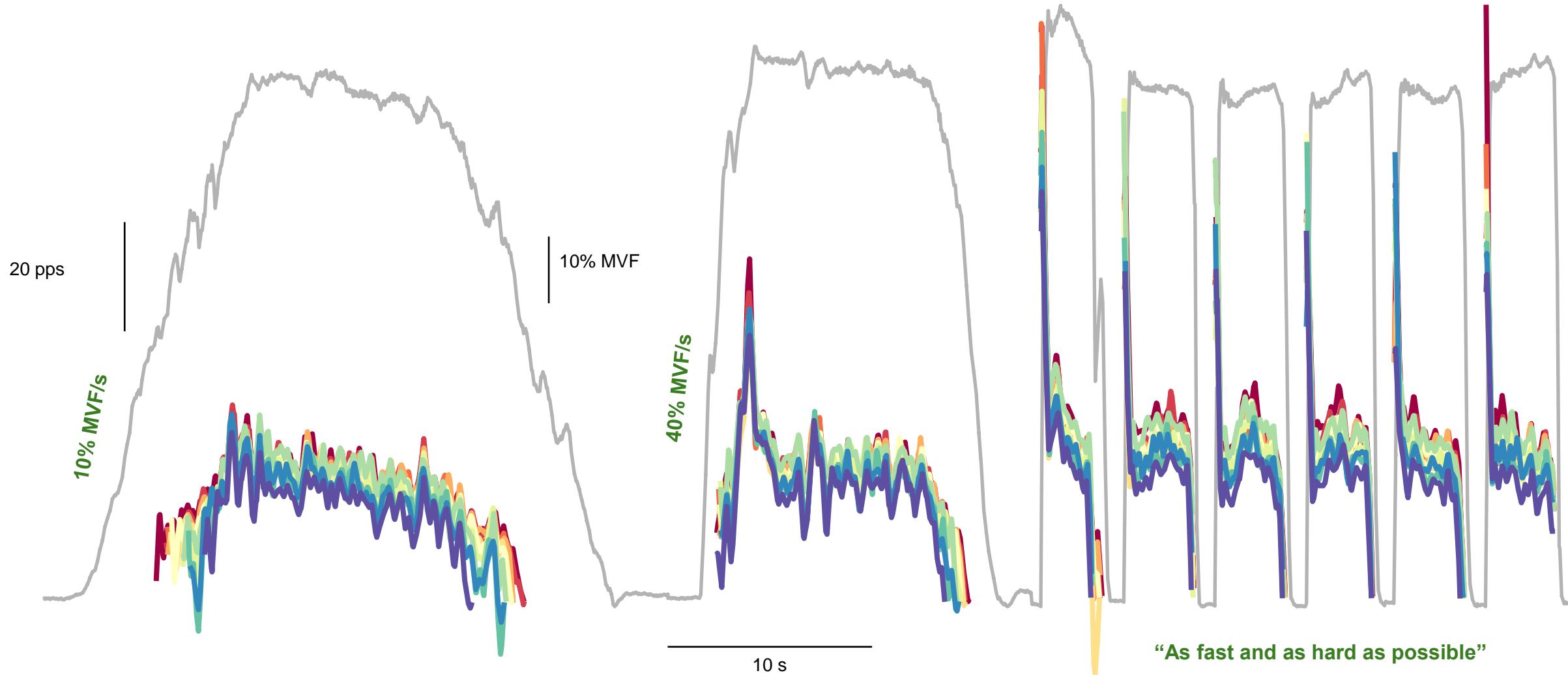


Challenge 2: Bias towards larger potentials

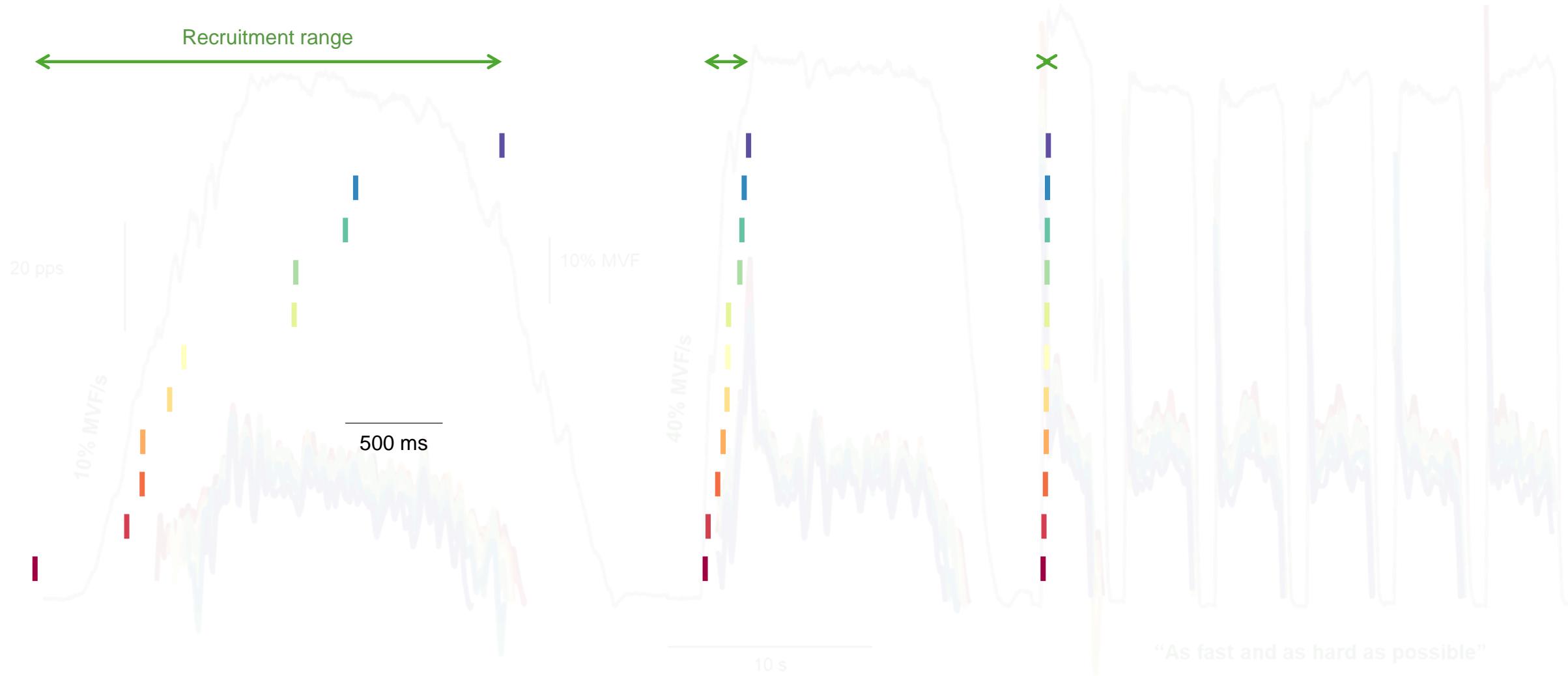
Challenge 3: Superimposition of MUAPs



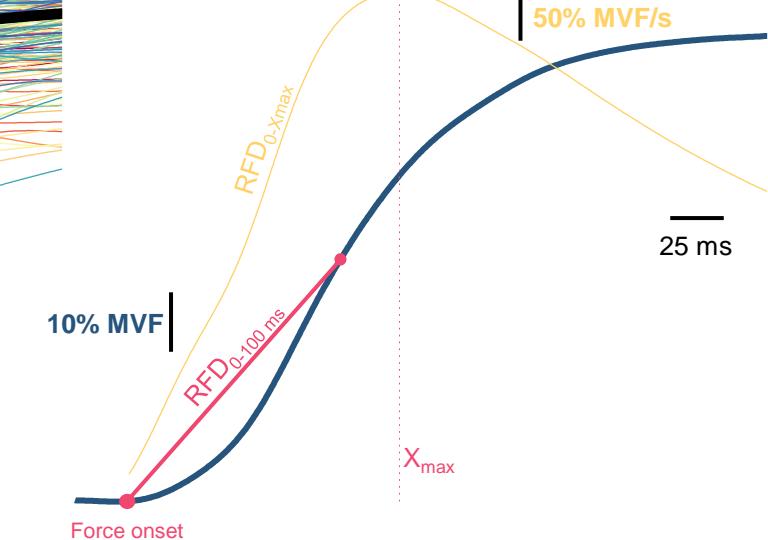
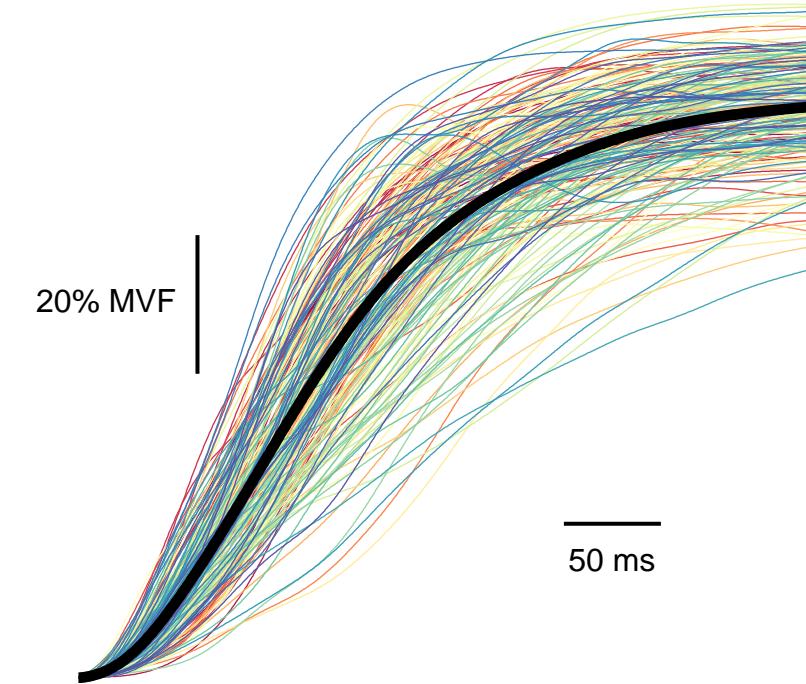
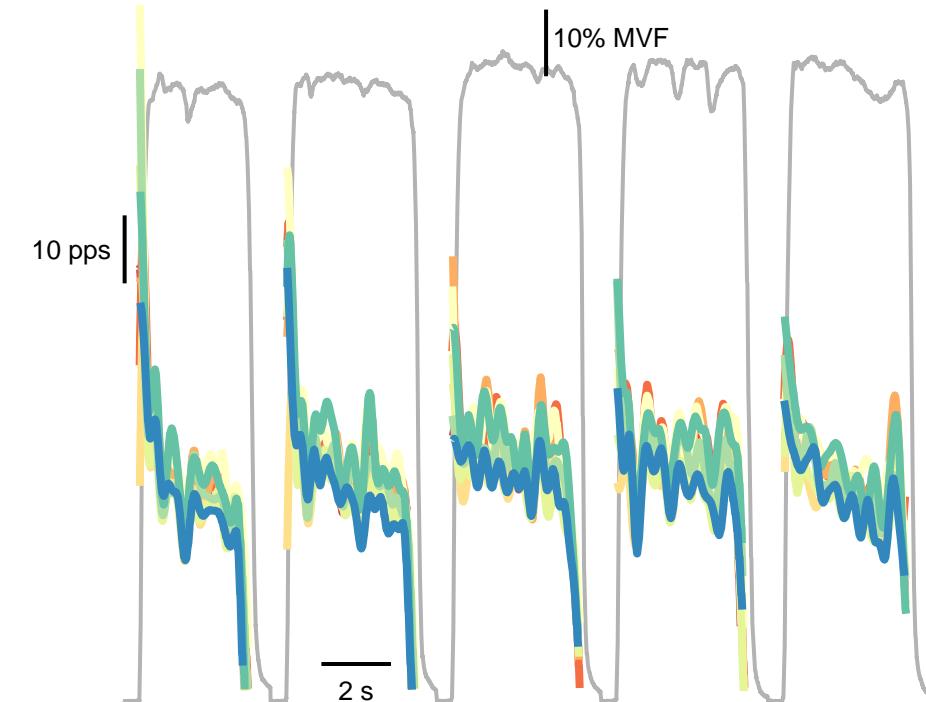
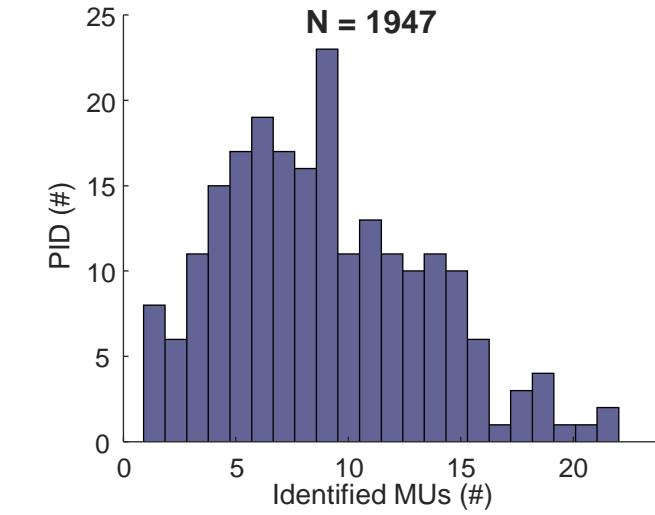
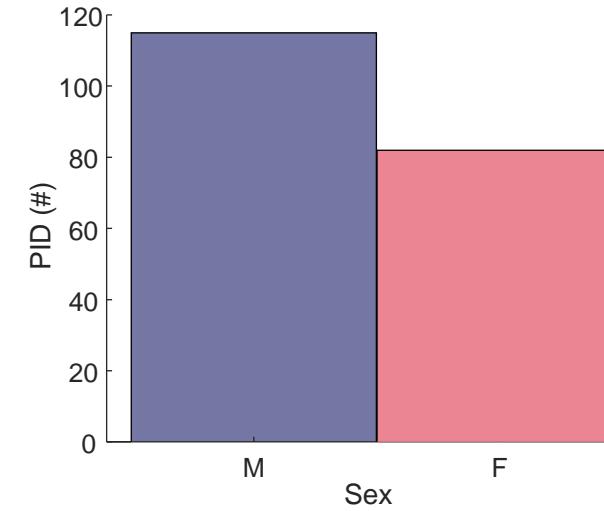
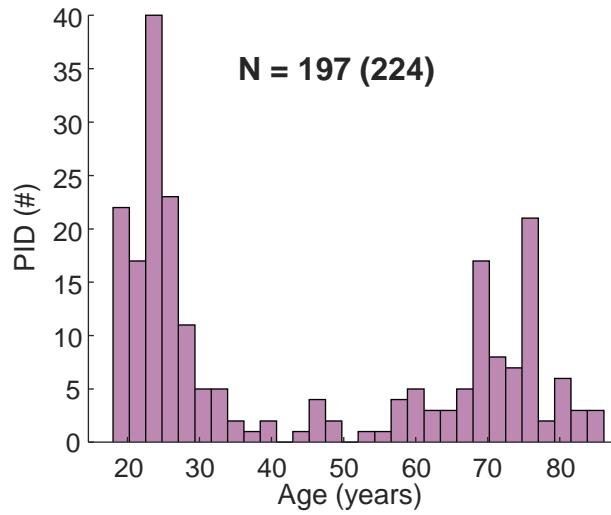
Feedback (ramp) vs. feedforward (rapid) contractions



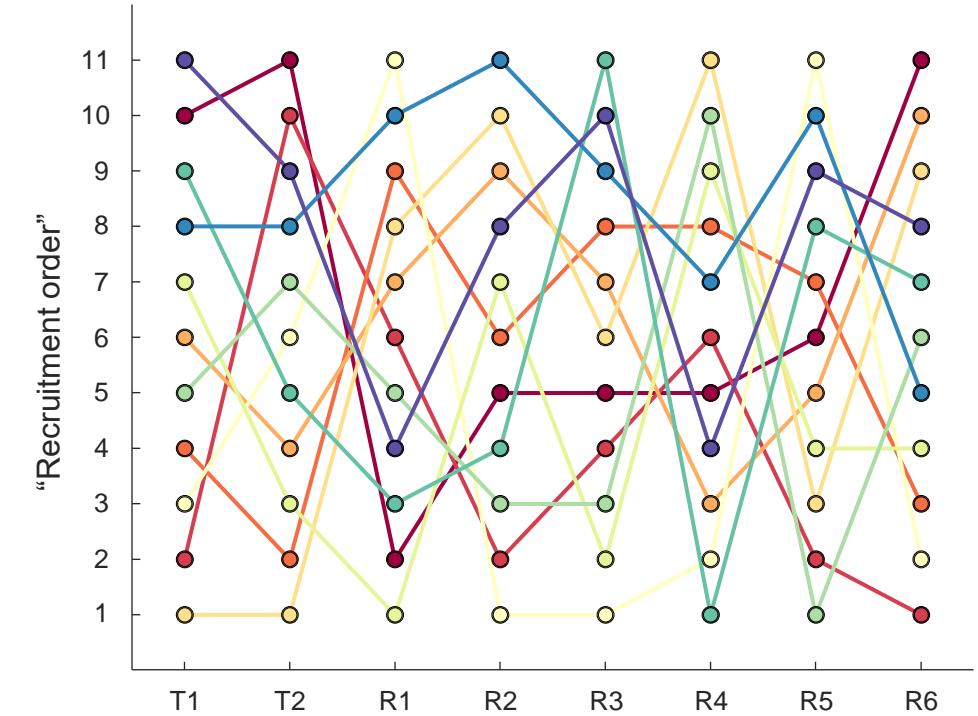
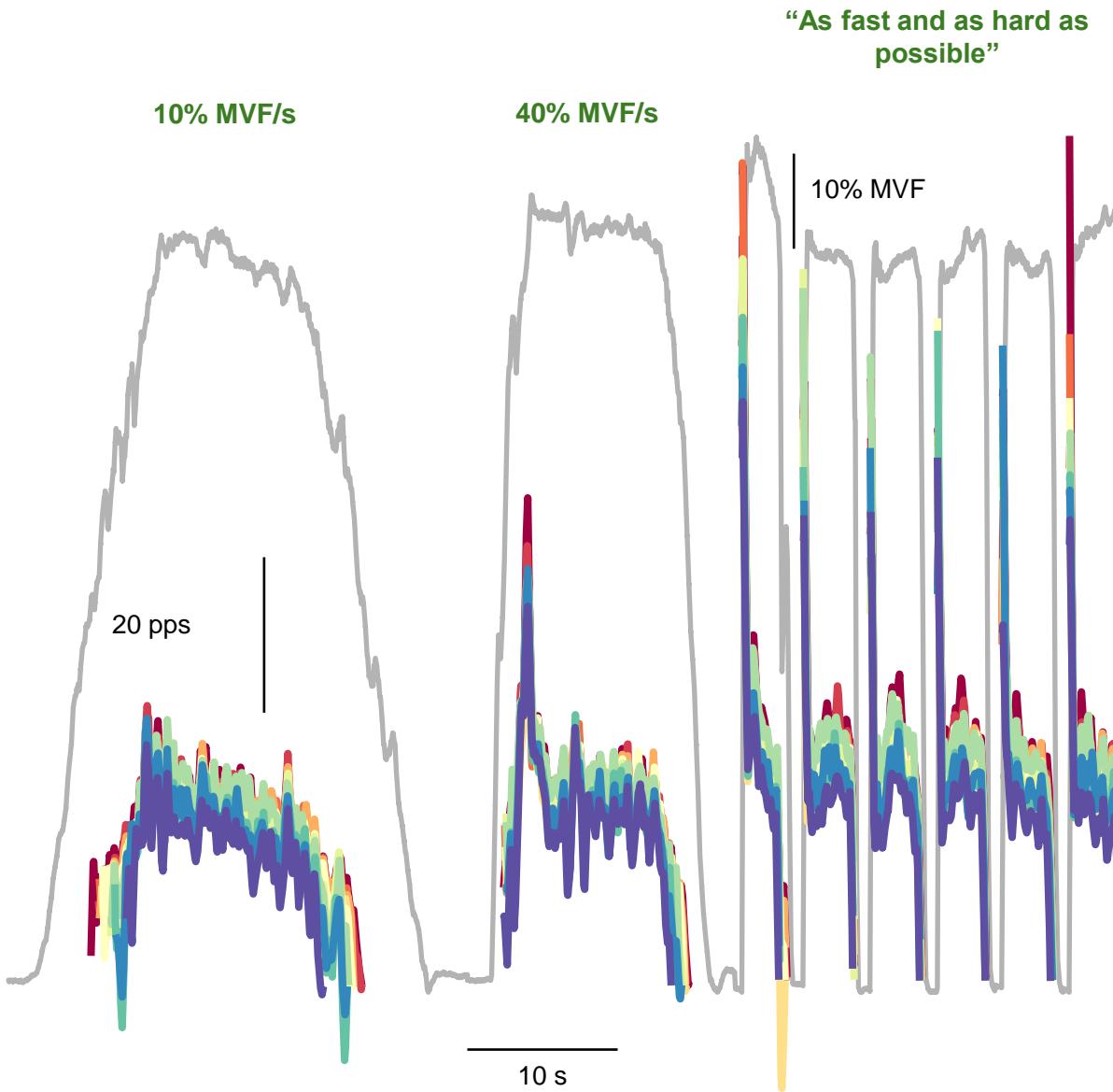
Feedback (ramp) vs. feedforward (rapid) contractions



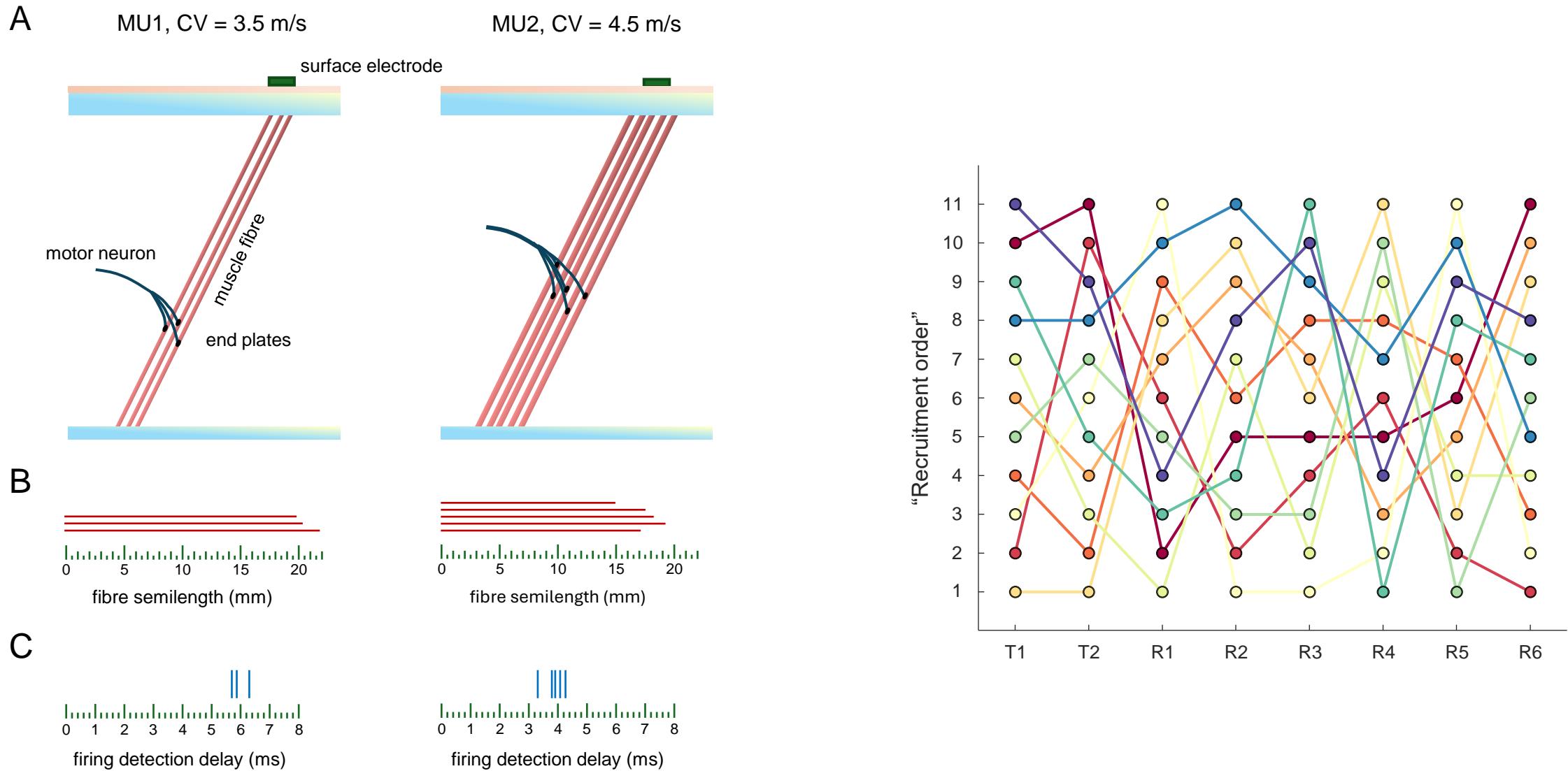
MU firing behaviour during rapid contractions



Motor unit recruitment order

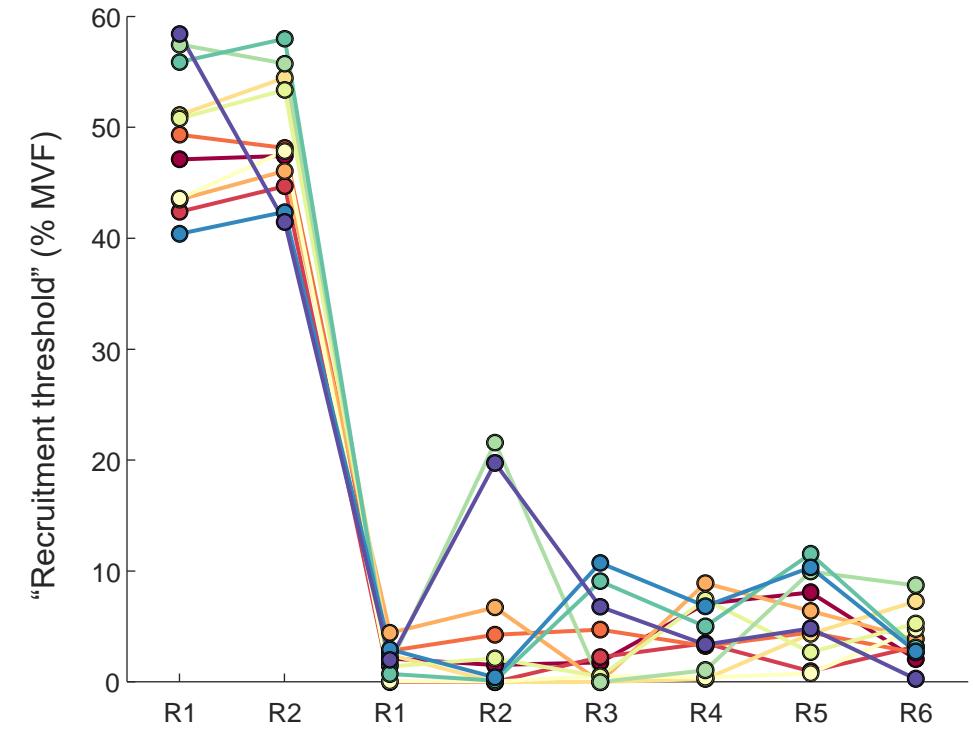
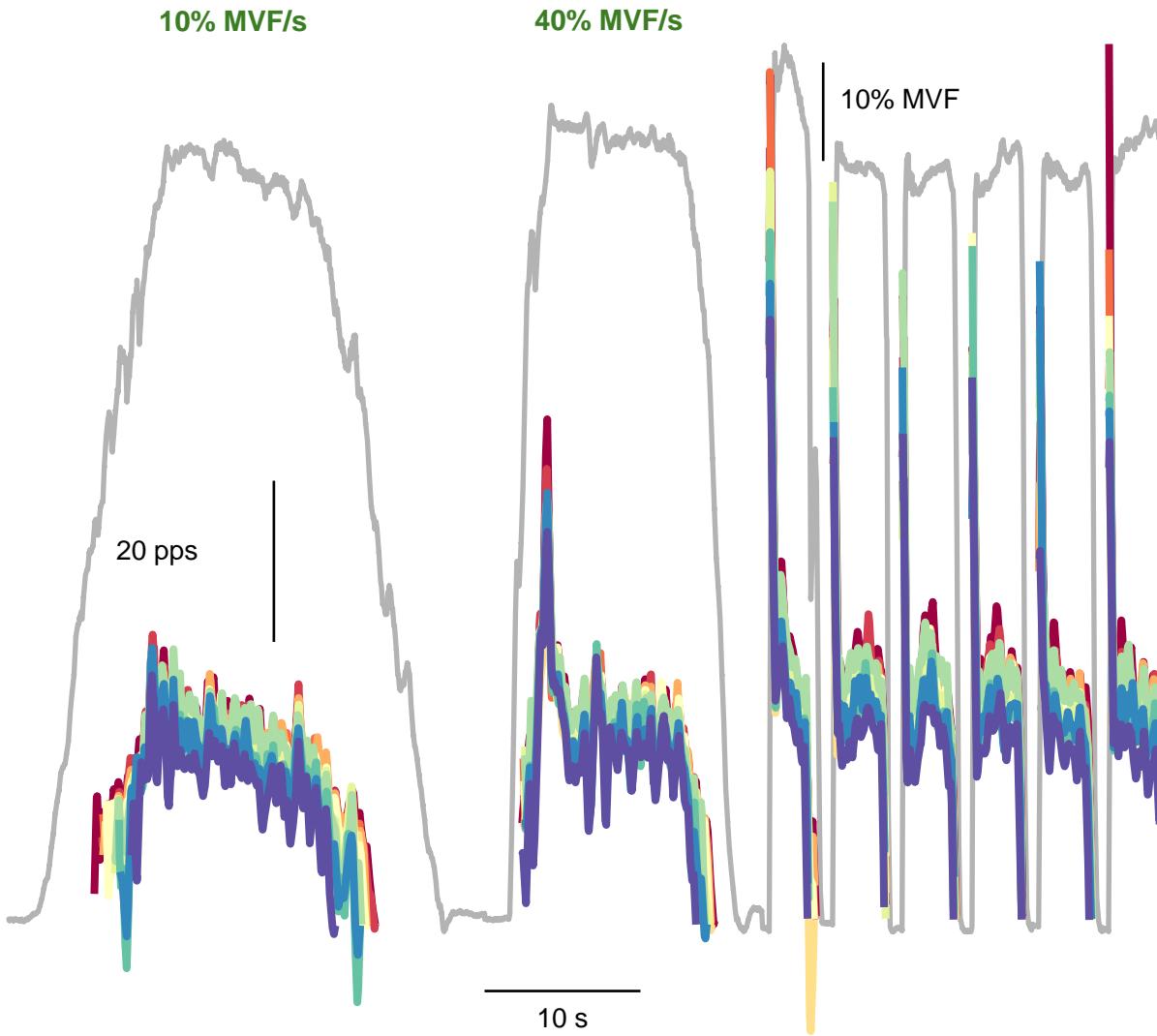


Motor unit recruitment order

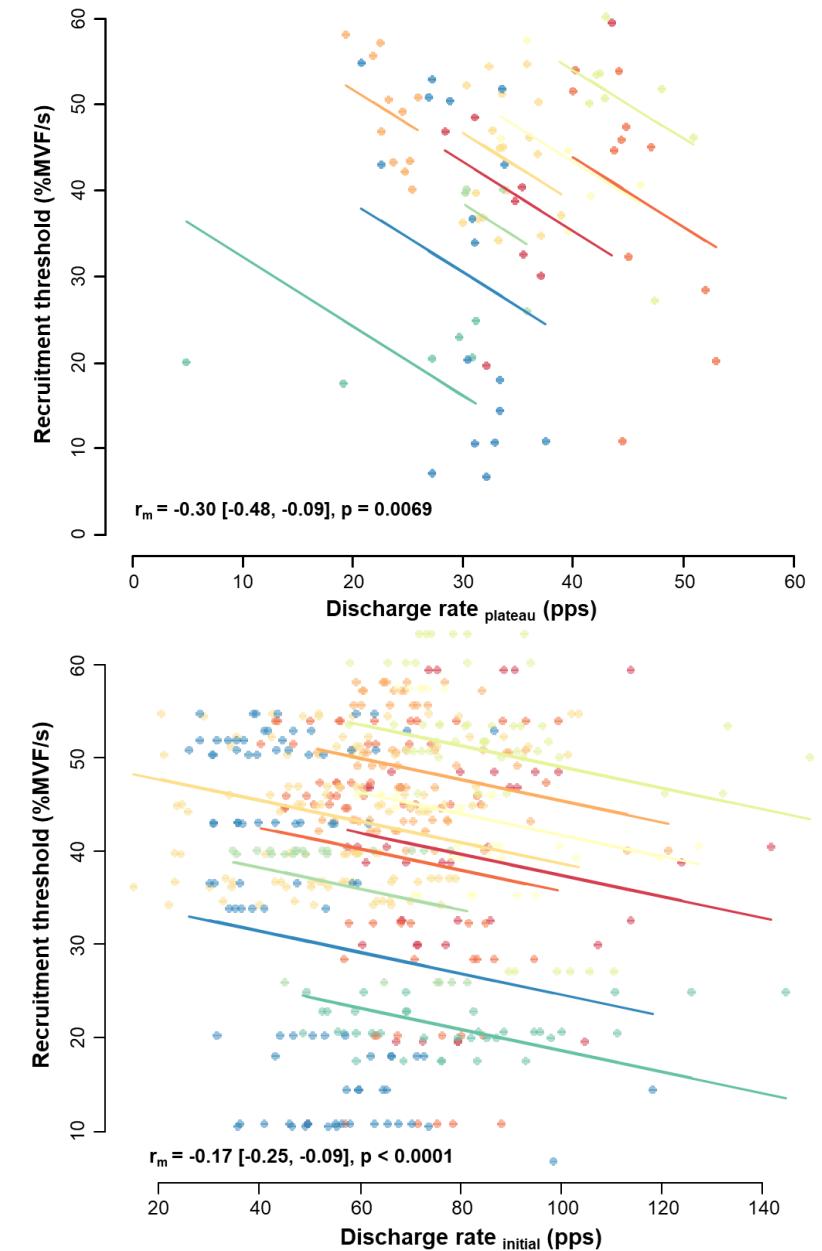
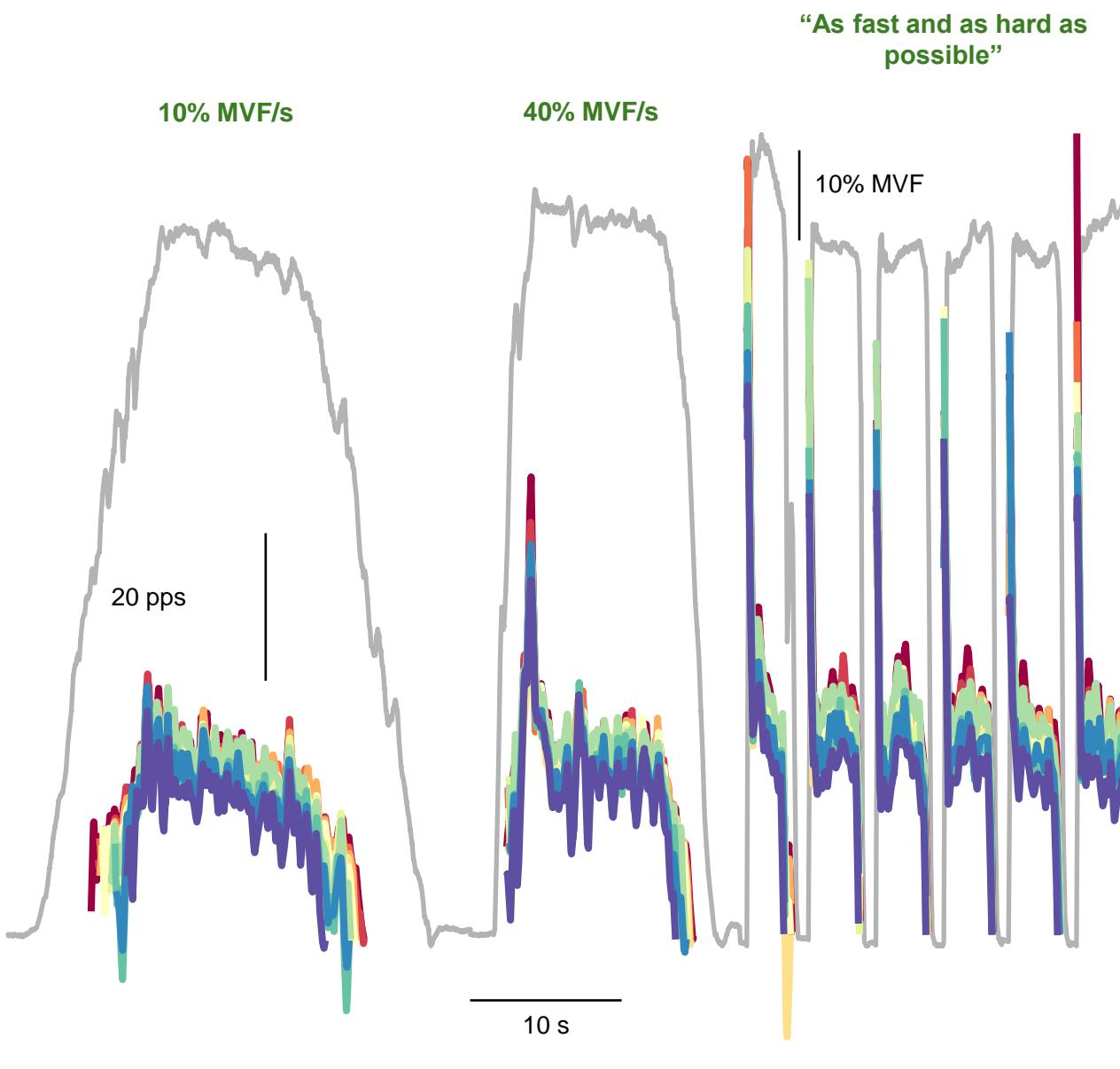


Motor unit recruitment threshold

“As fast and as hard as possible”

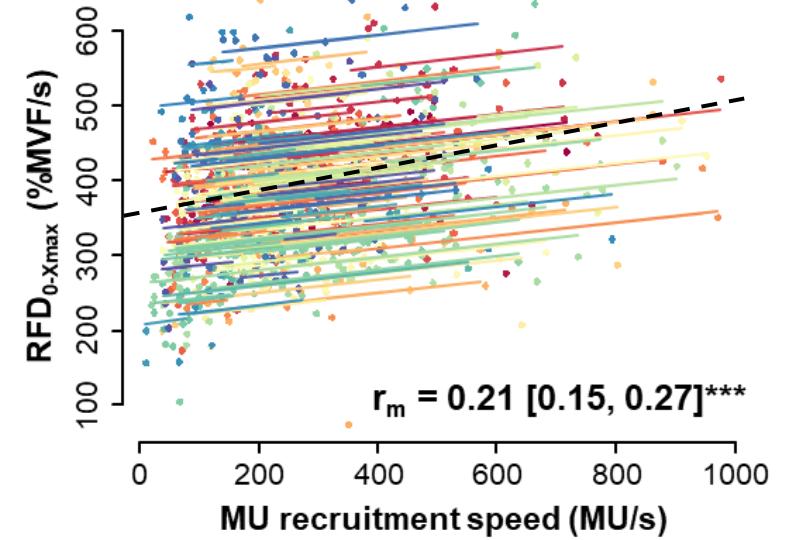
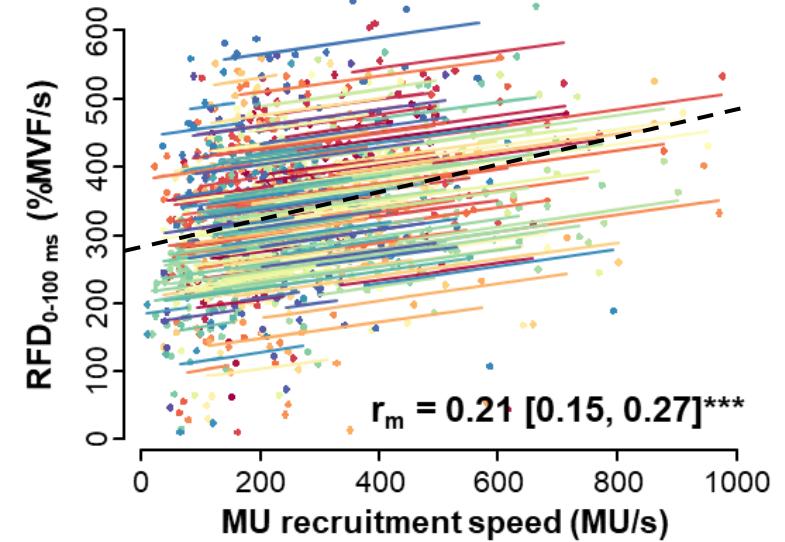
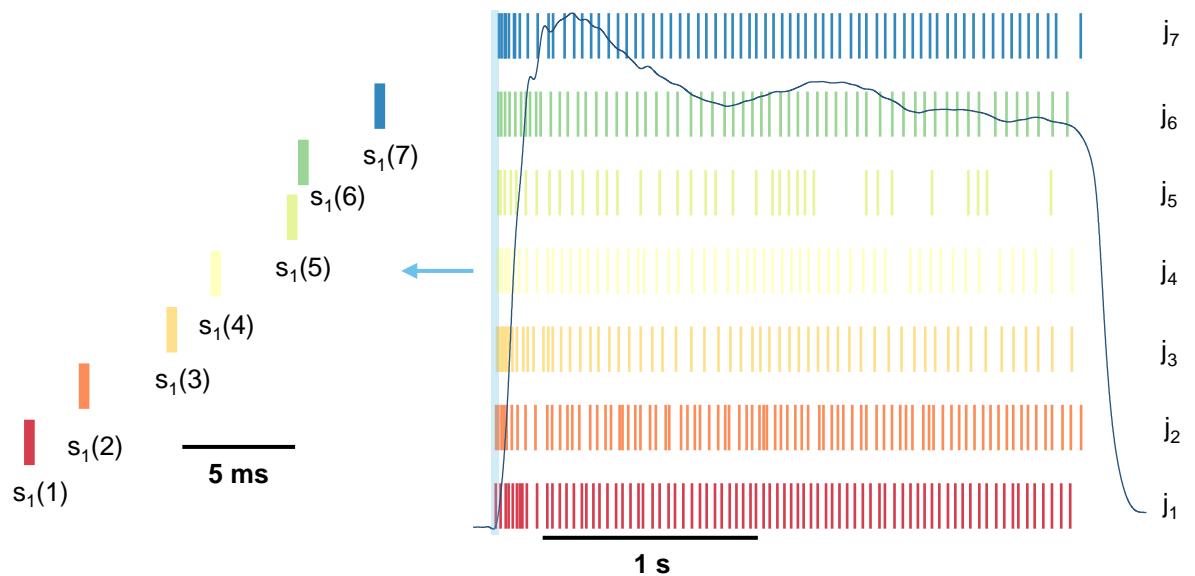


MU firing rate relative to recruitment threshold during rapid contractions

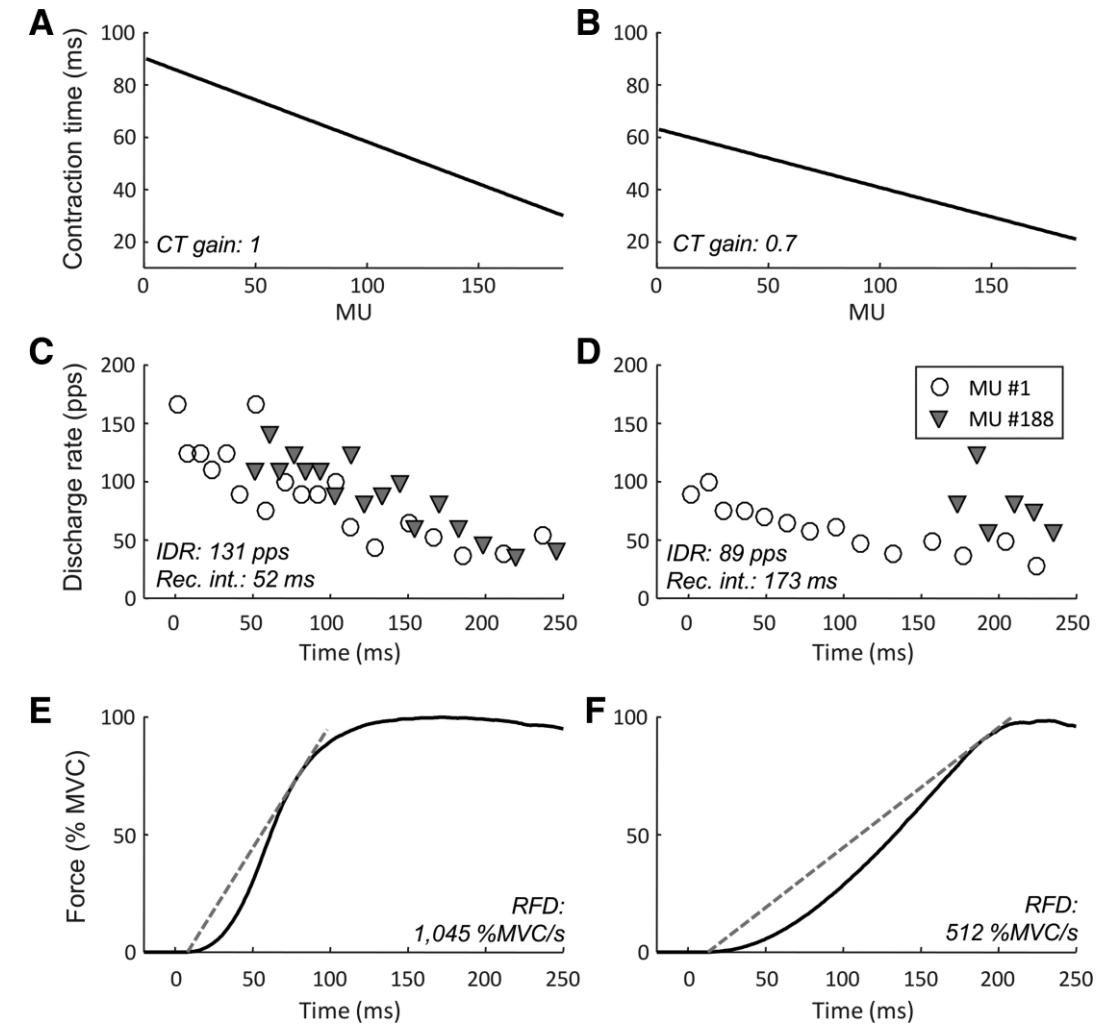
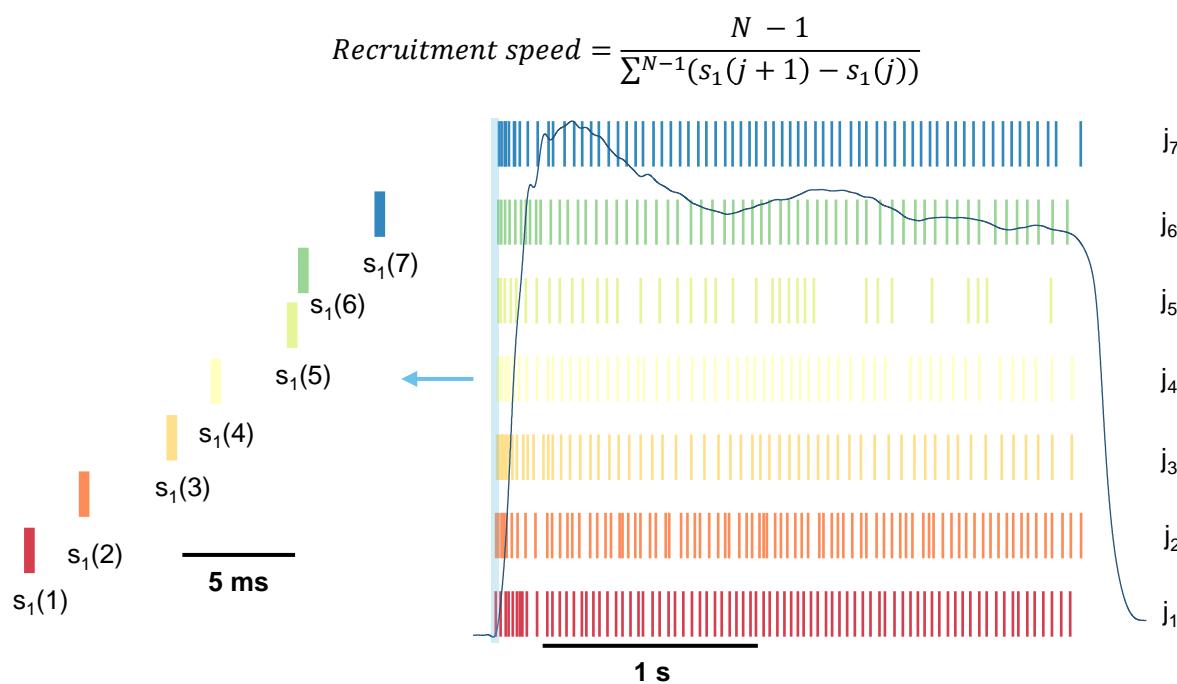


Motor unit recruitment speed

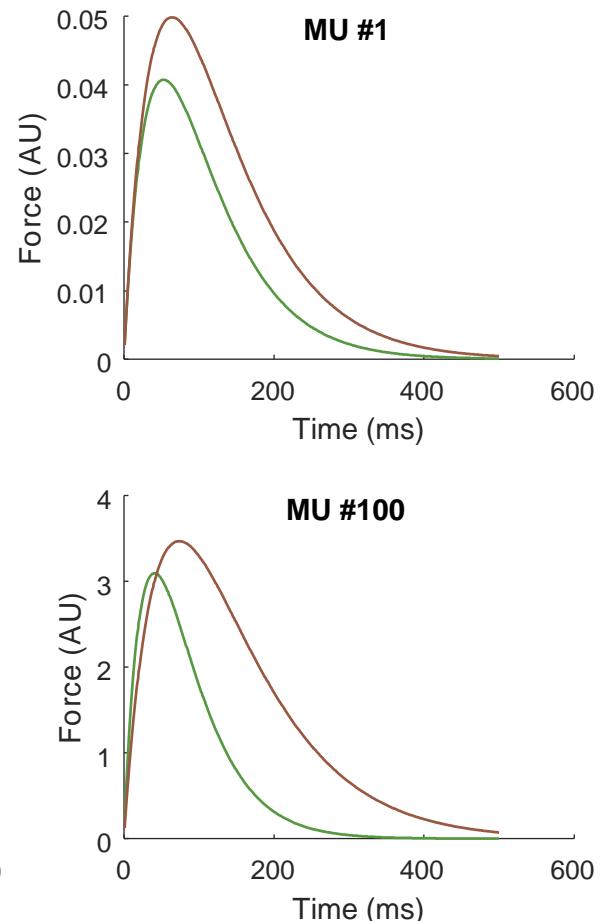
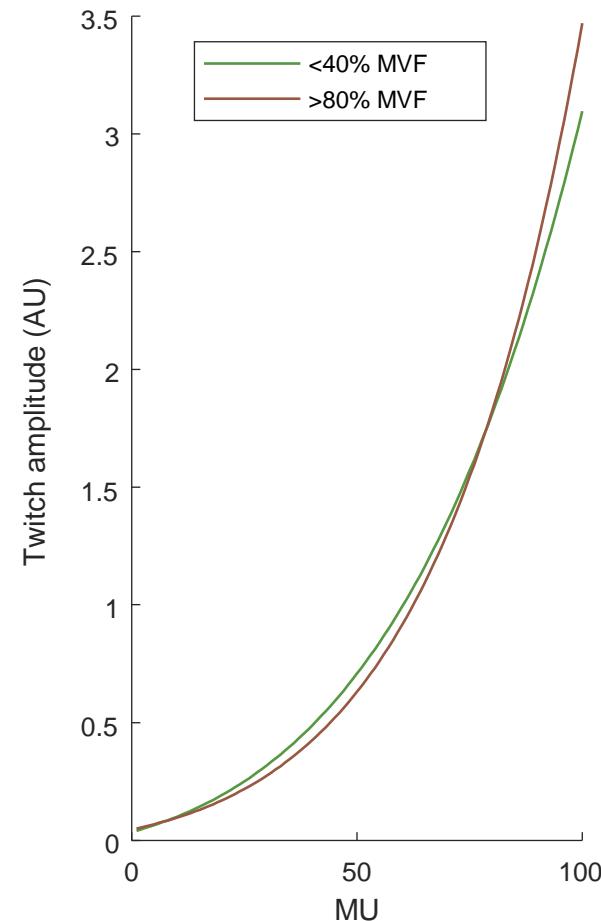
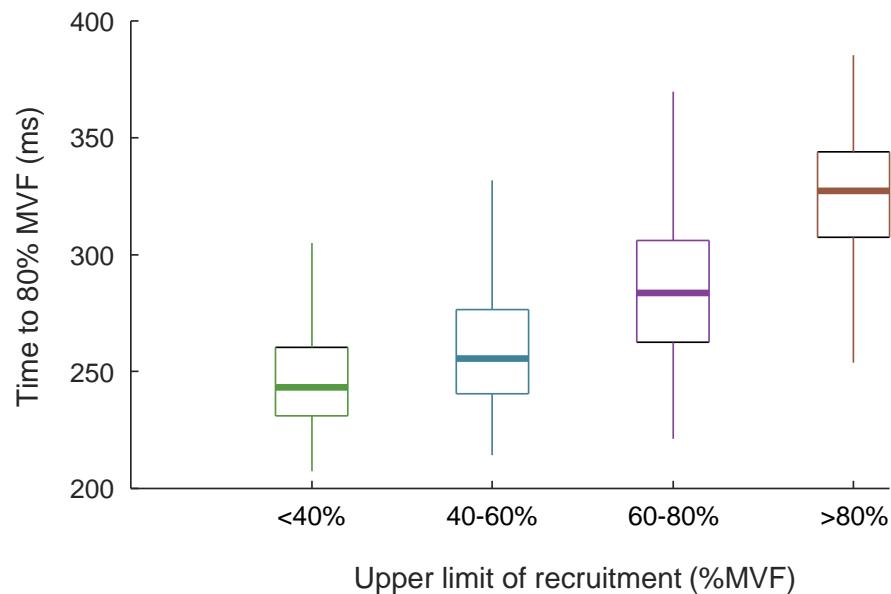
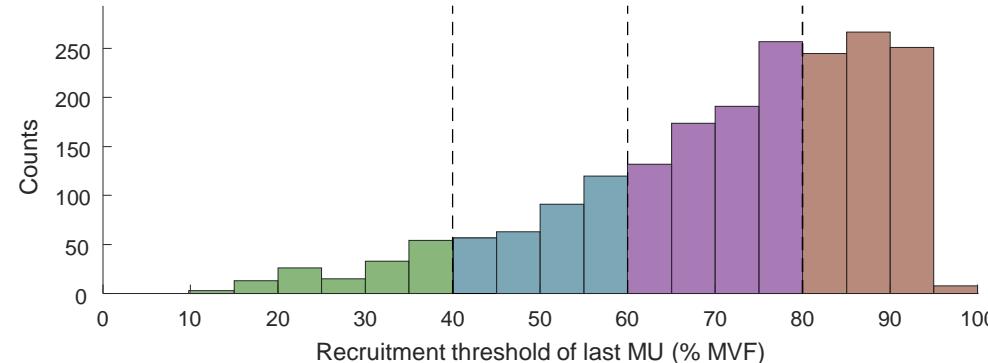
$$\text{Recruitment speed} = \frac{N - 1}{\sum^{N-1}(s_1(j + 1) - s_1(j))}$$



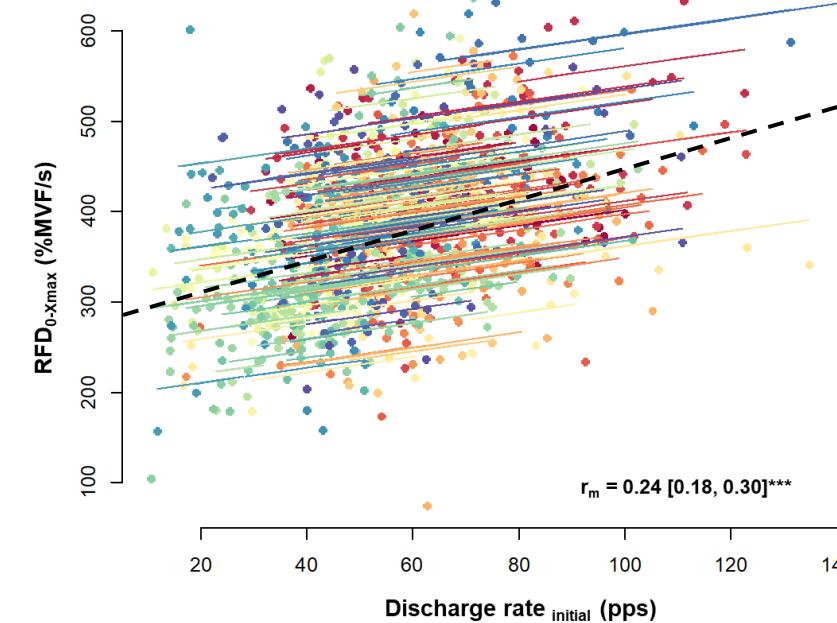
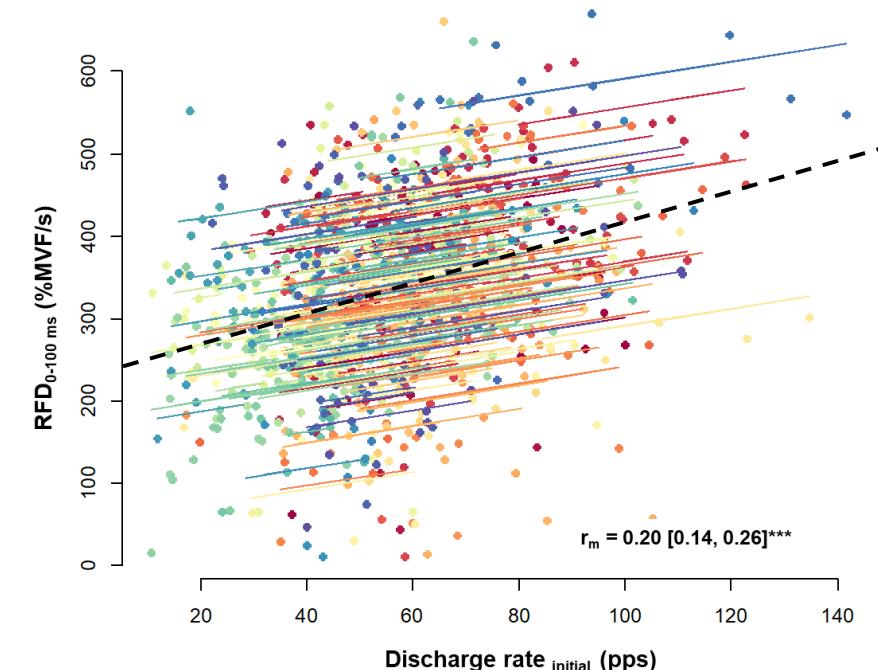
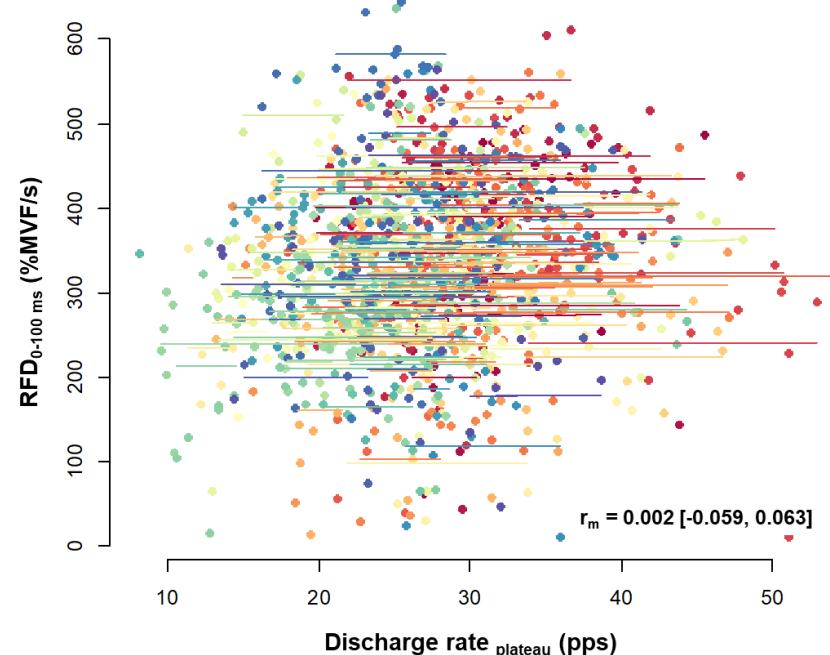
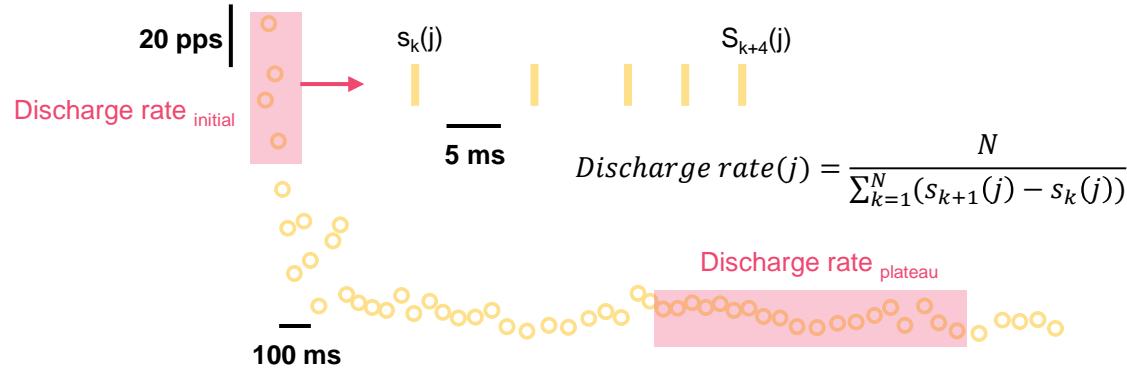
Motor unit recruitment speed



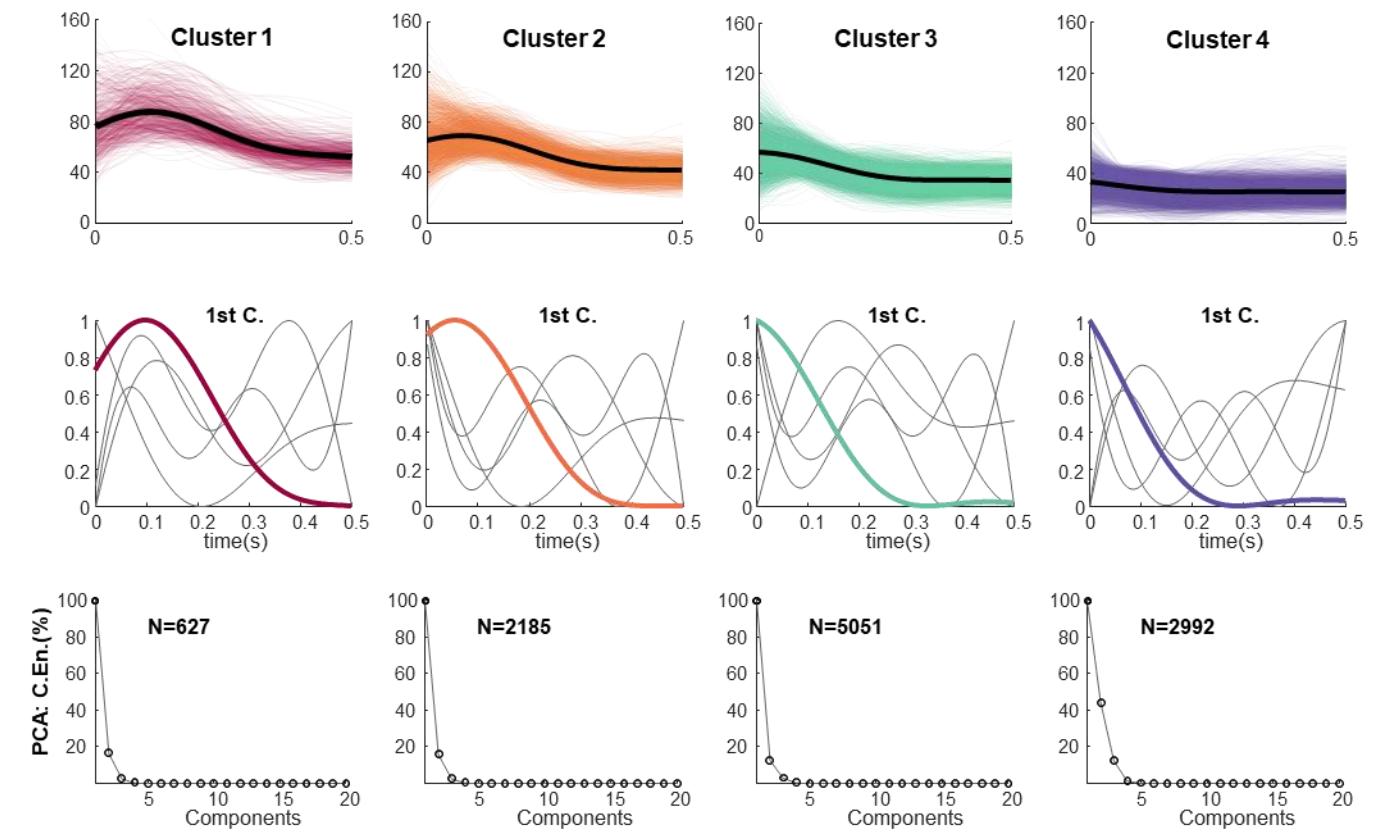
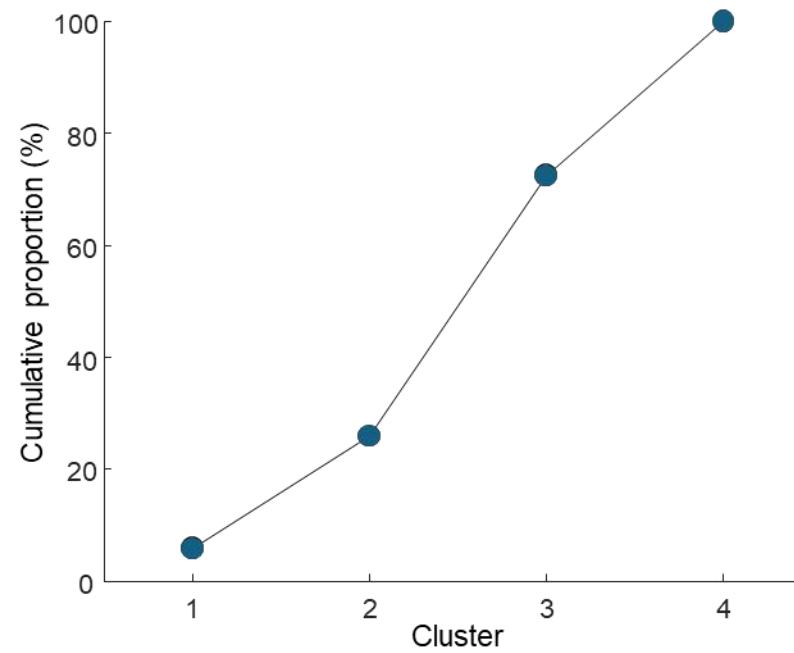
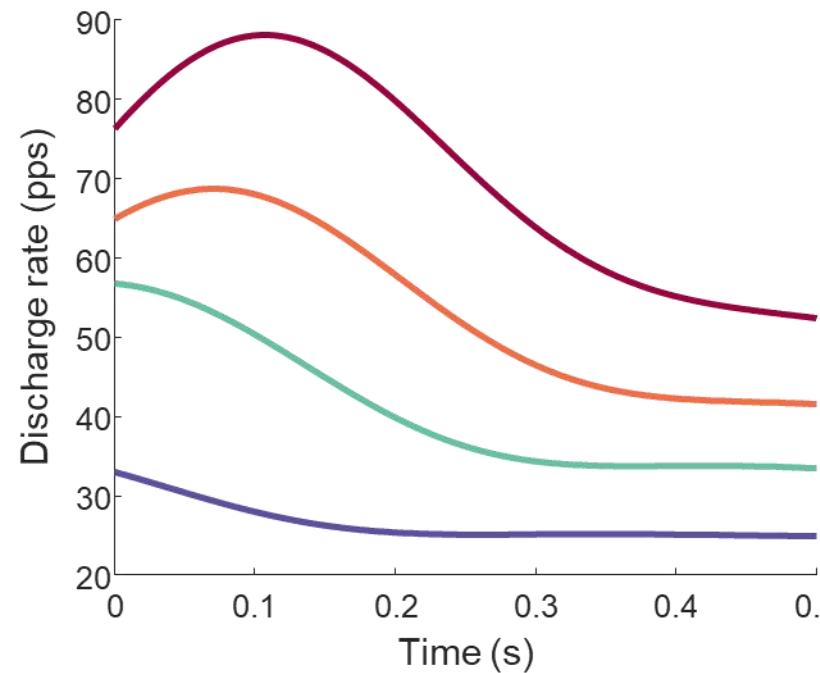
Motor unit recruitment speed as a function of the upper limit of recruitment



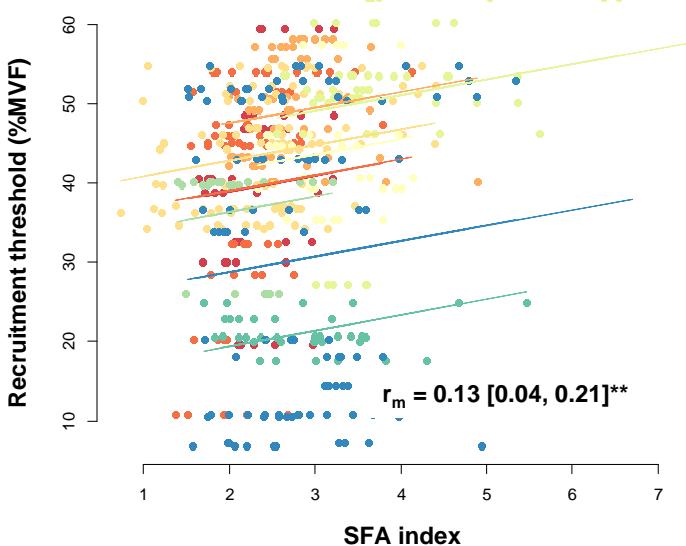
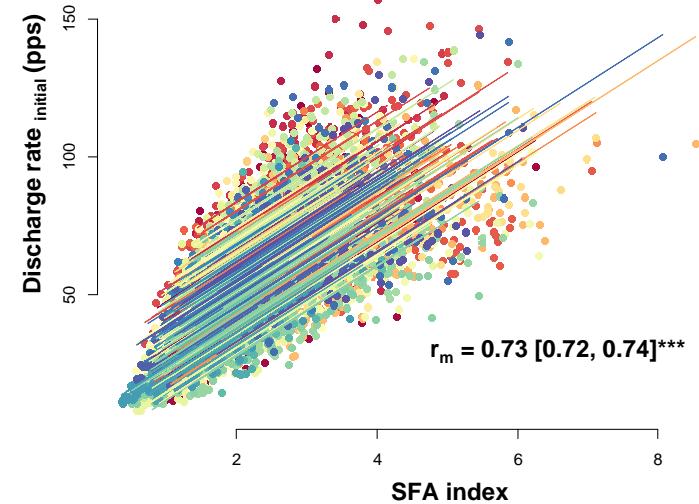
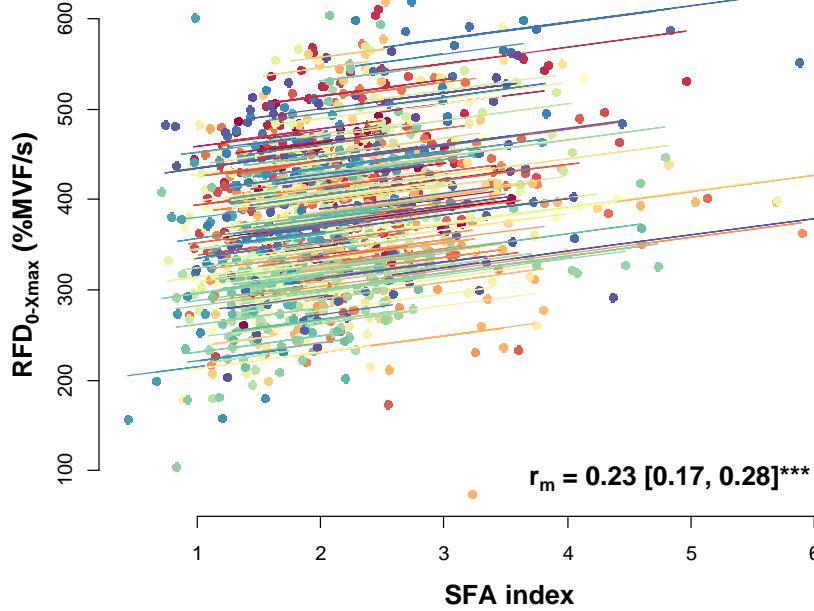
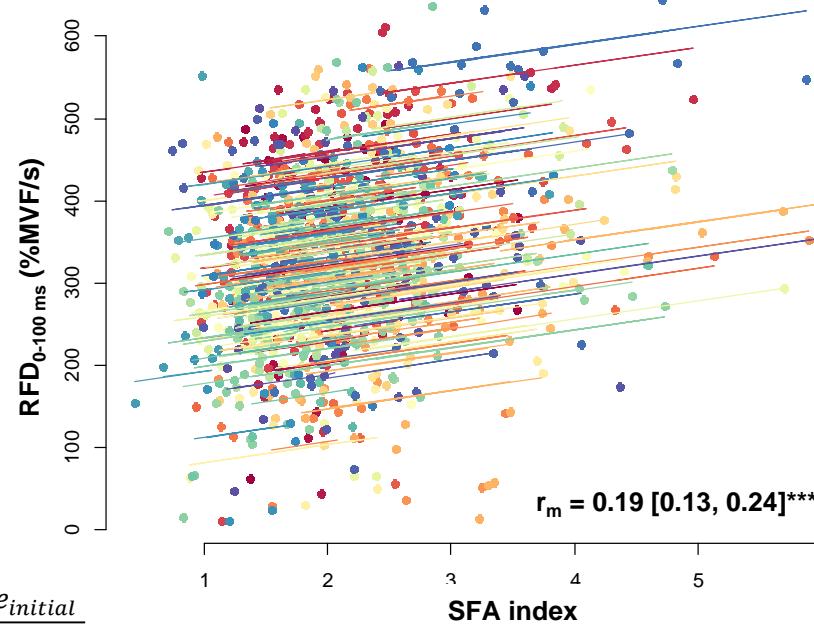
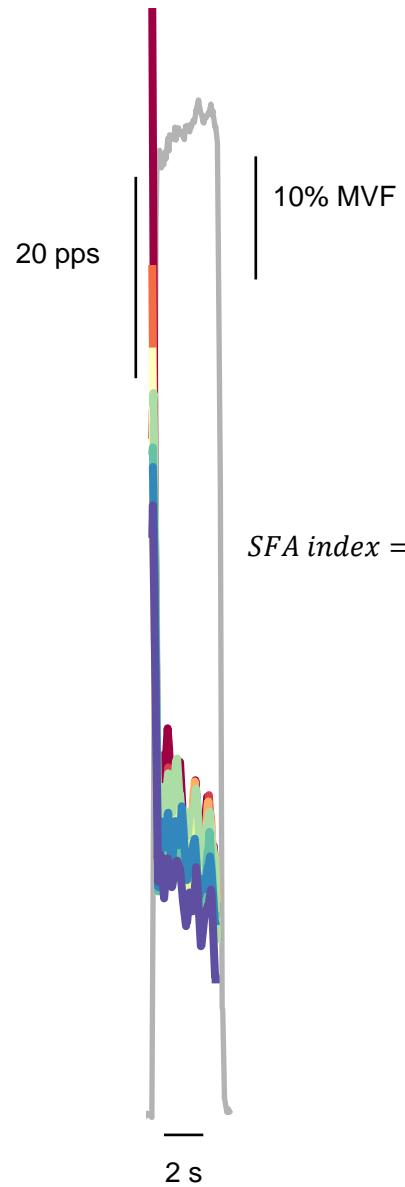
MU firing rate during rapid contractions



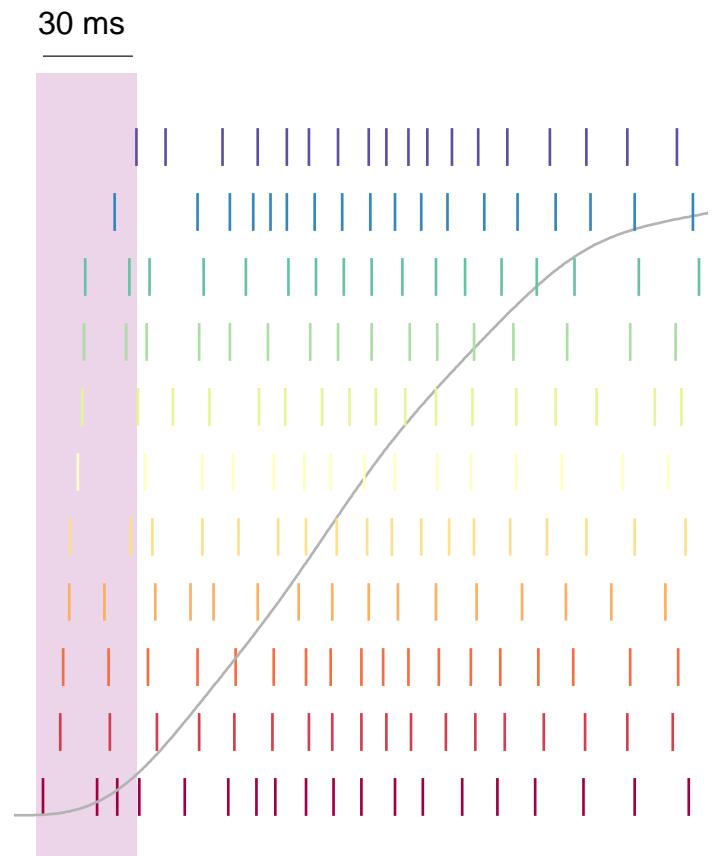
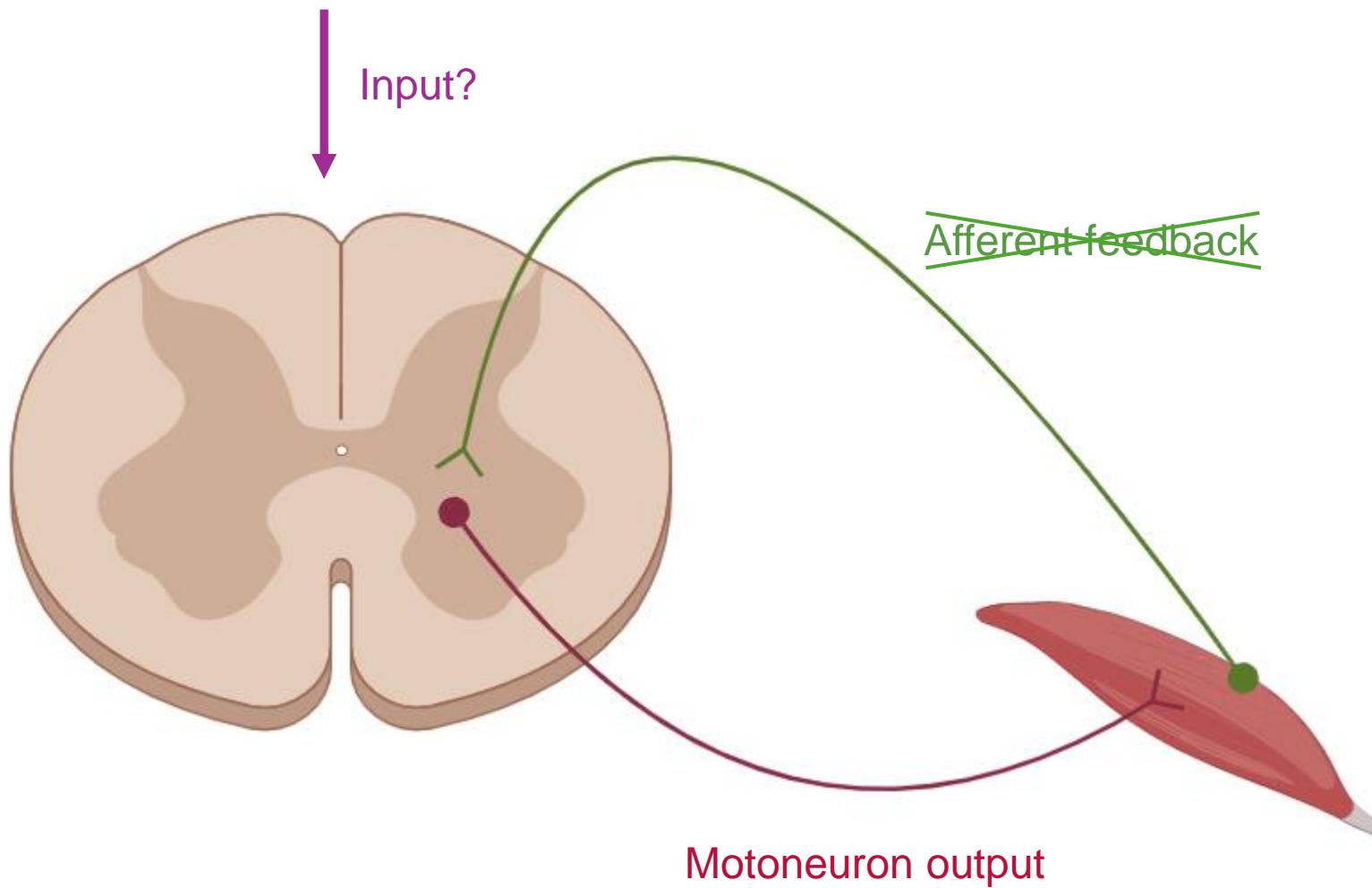
Clusters of MU firing patterns



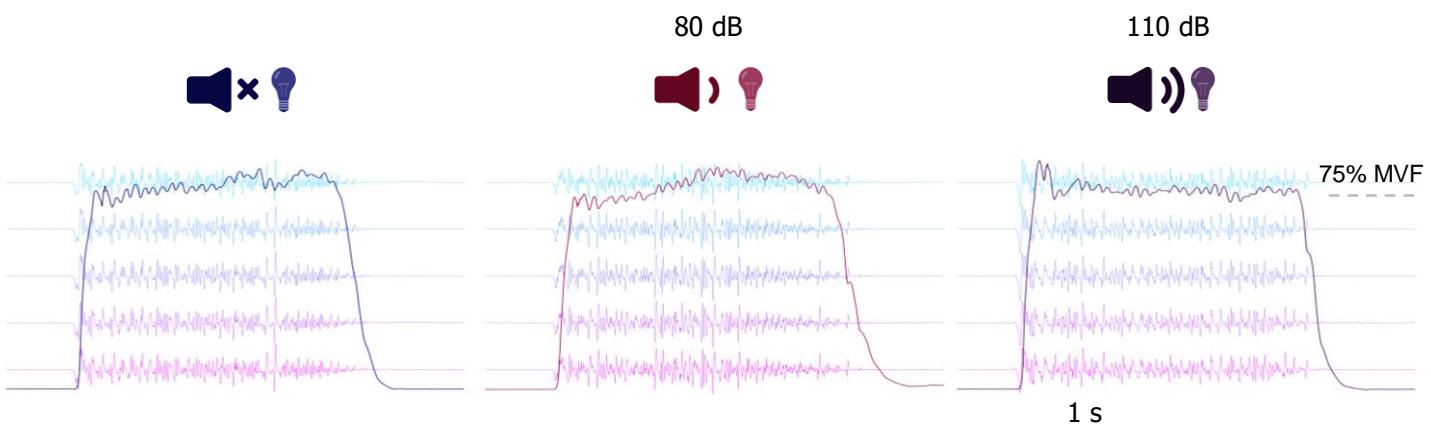
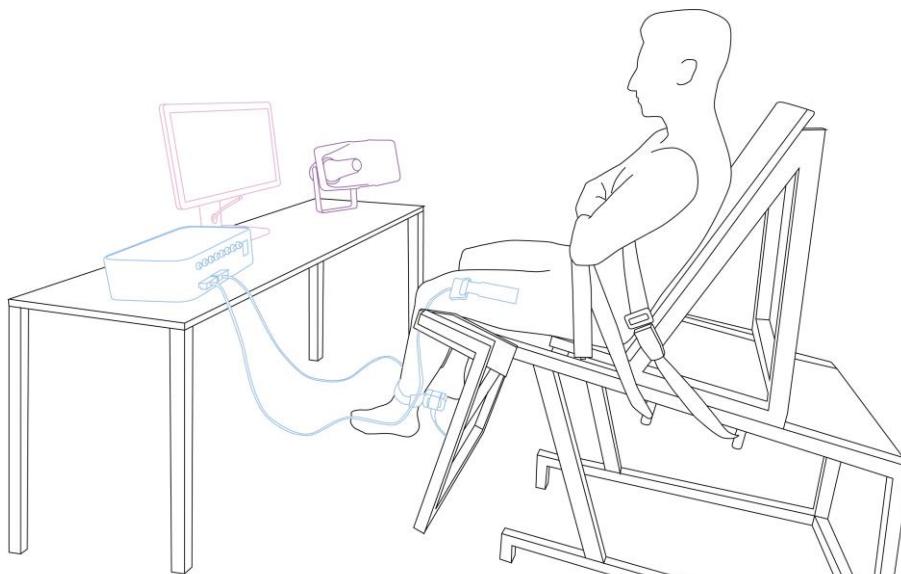
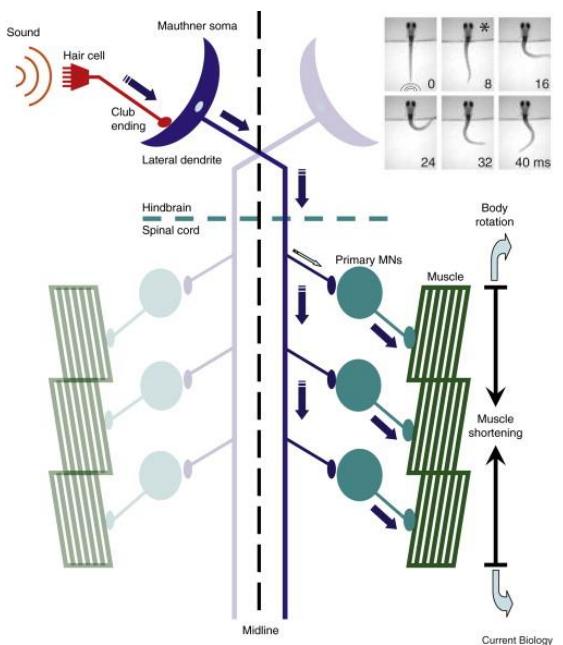
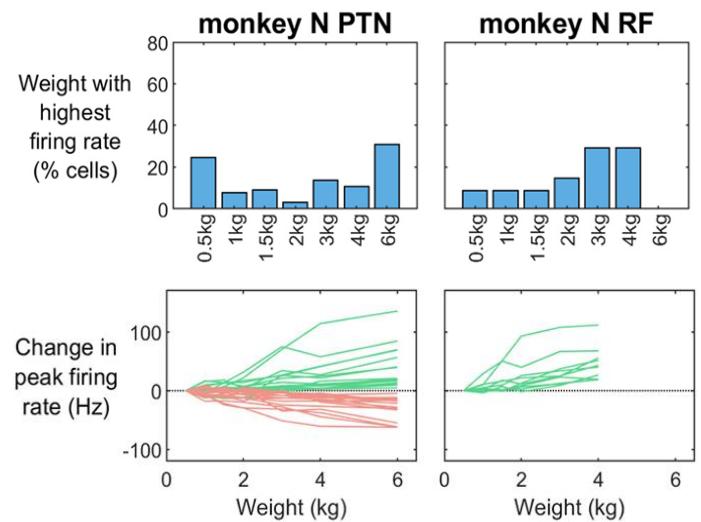
Spike frequency adaptation

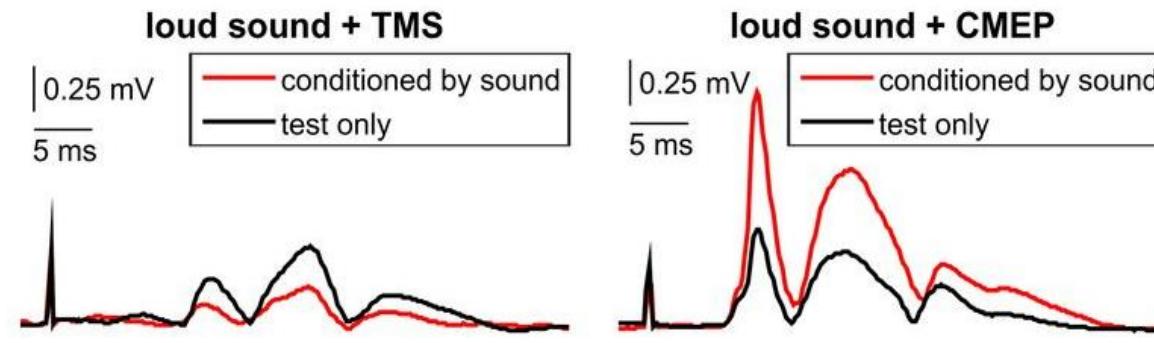
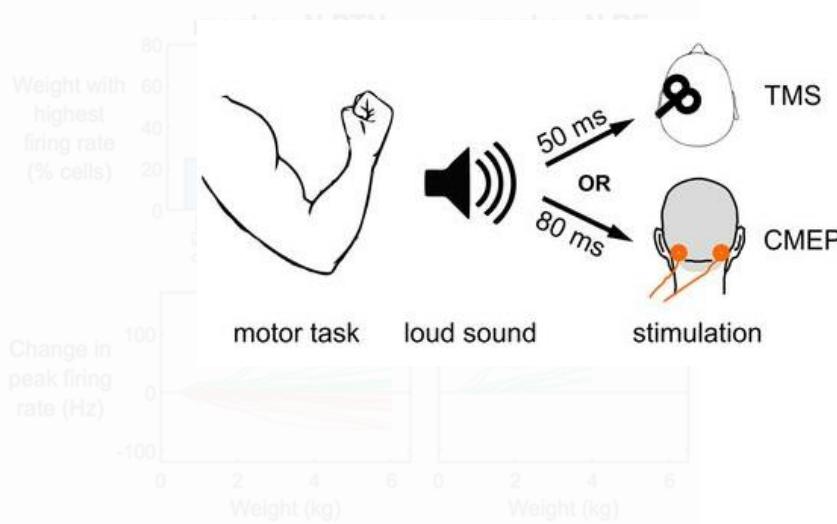


The neural substrate of motor unit behaviour during rapid contractions

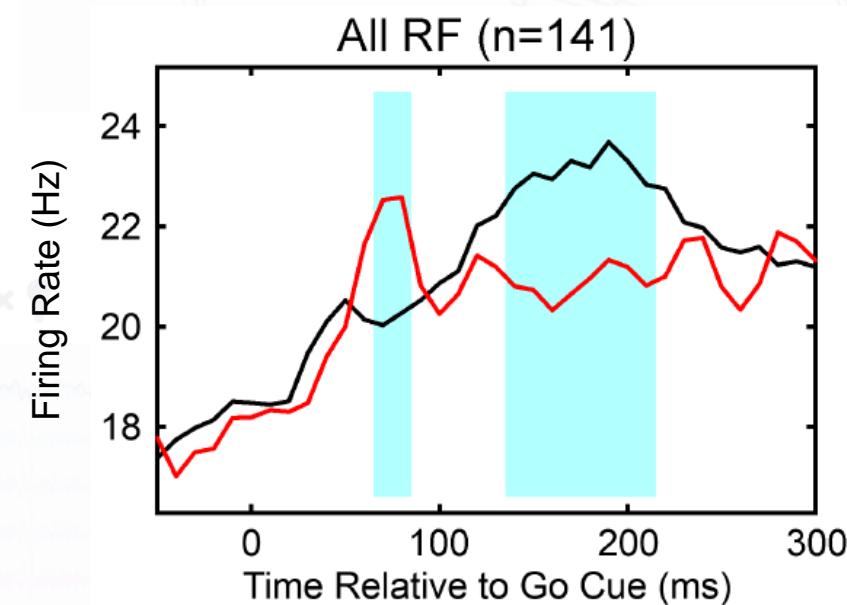
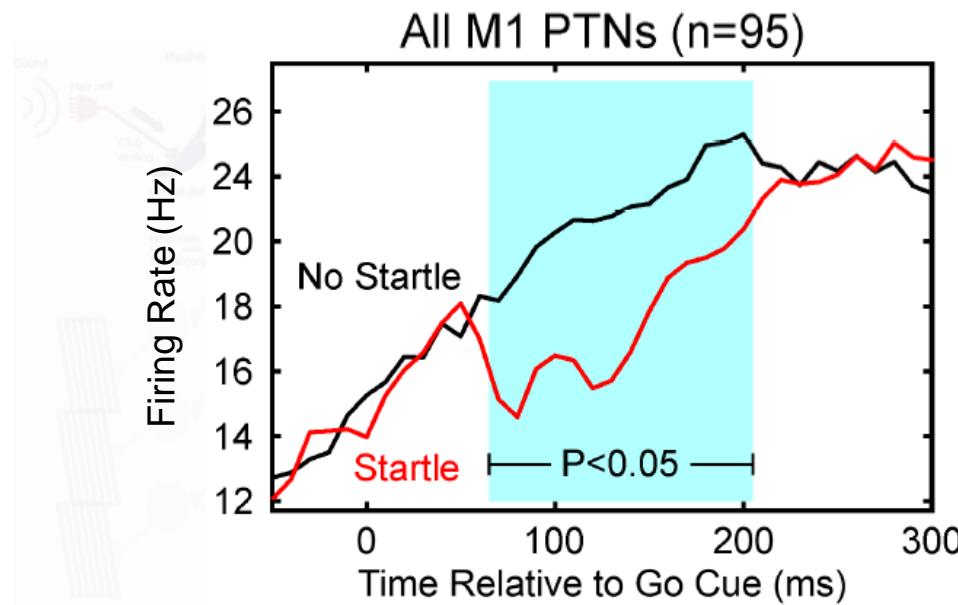


The neural substrate of motor unit behaviour during rapid contractions

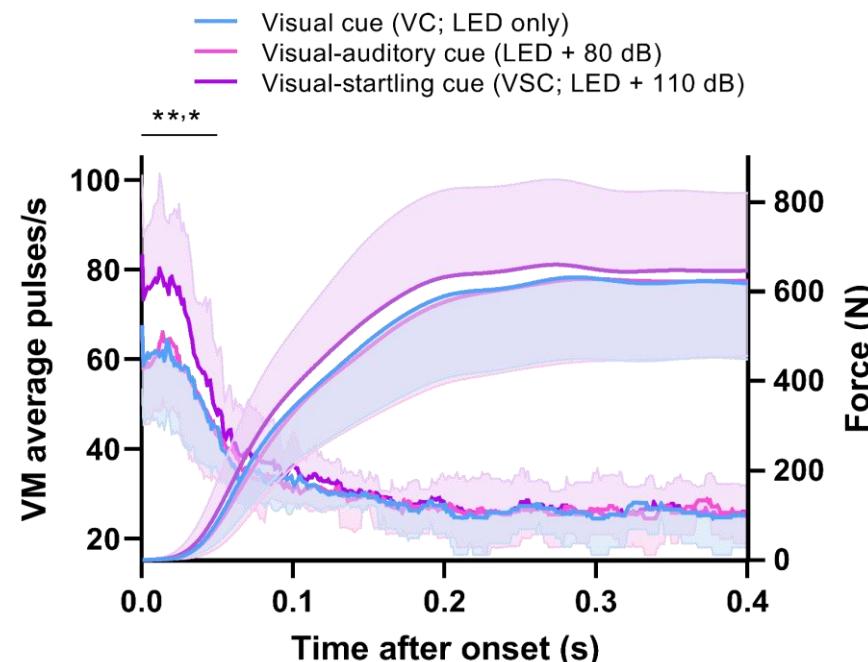
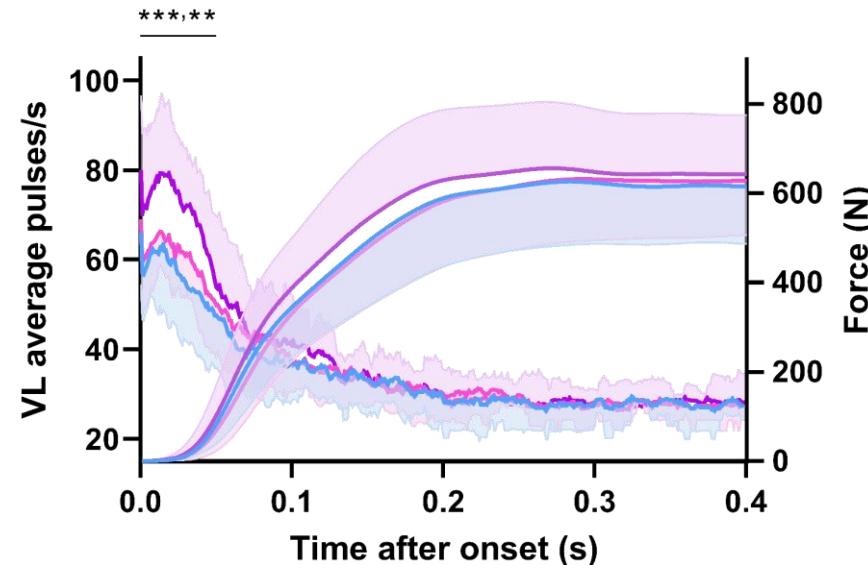
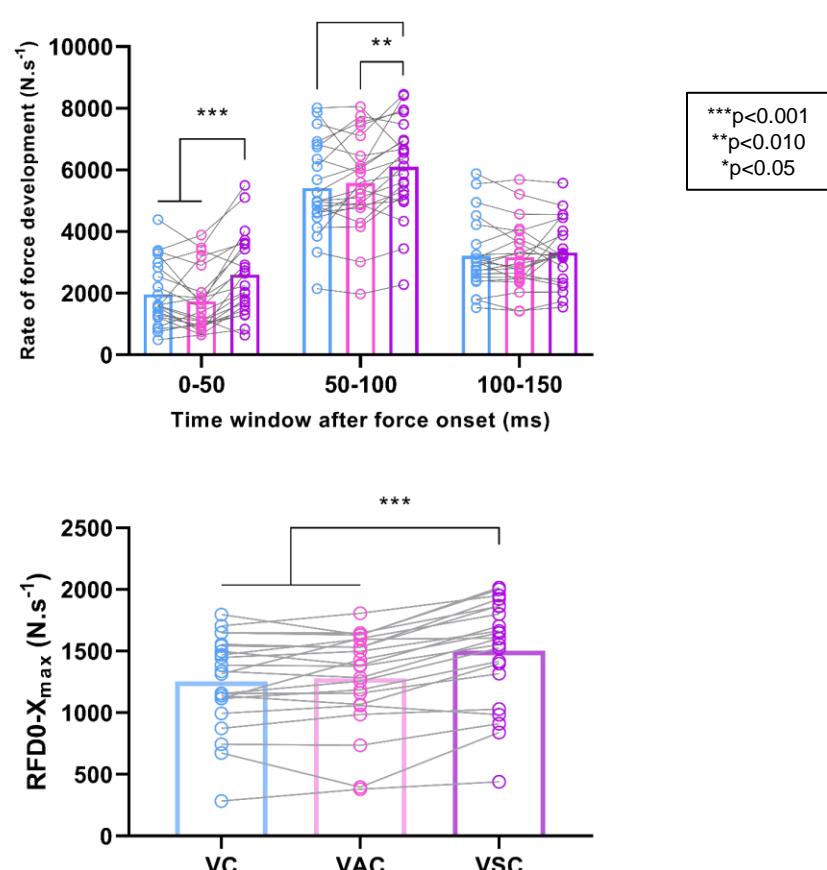
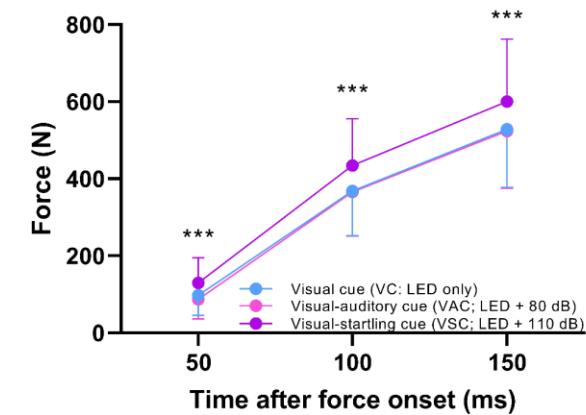




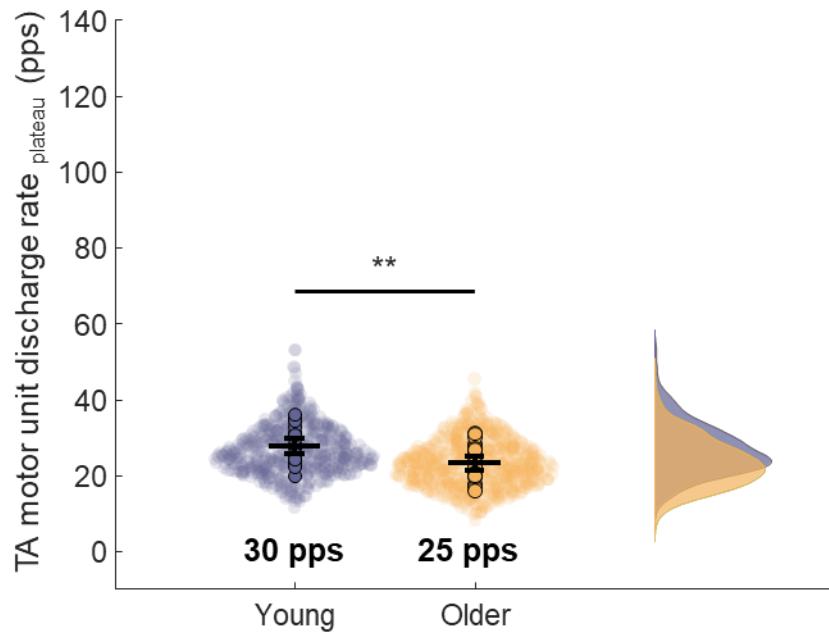
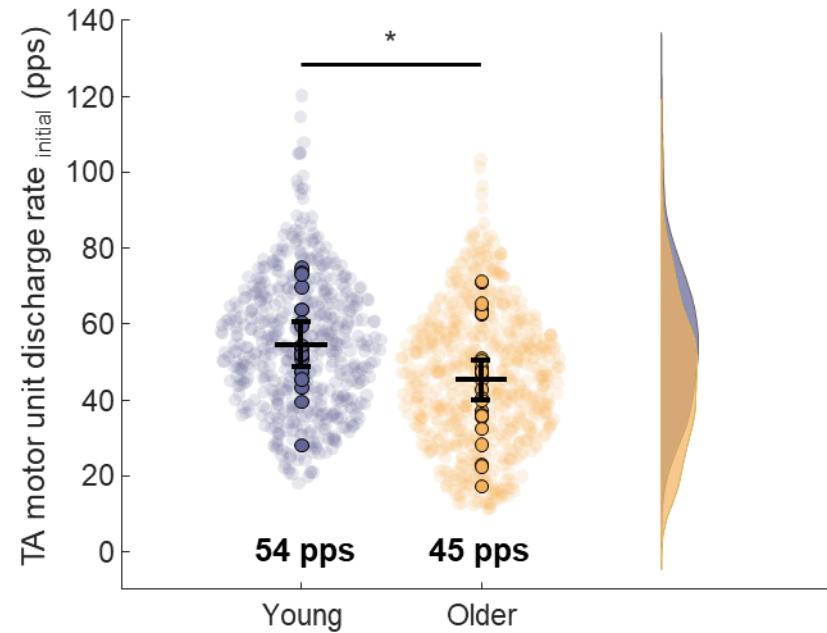
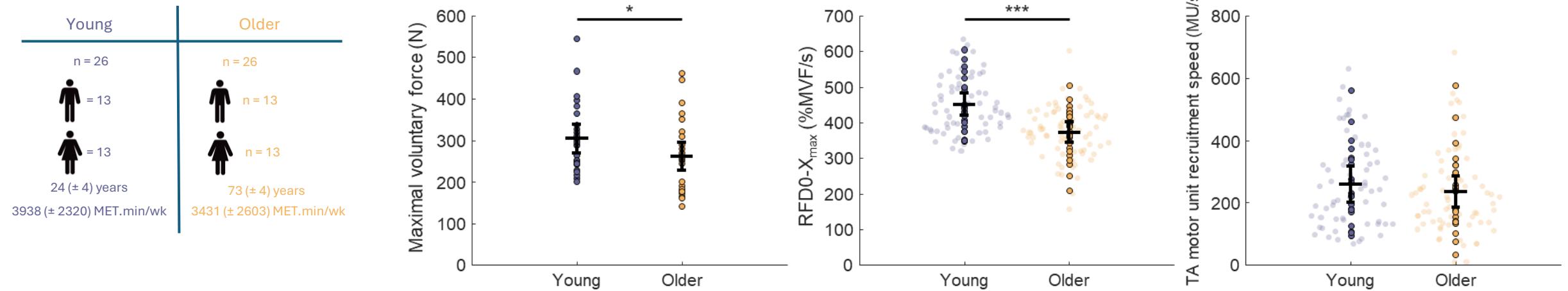
Germann & Baker 2021, *J Neurosci*



Tapia et al. 2022, *J Neurosci*



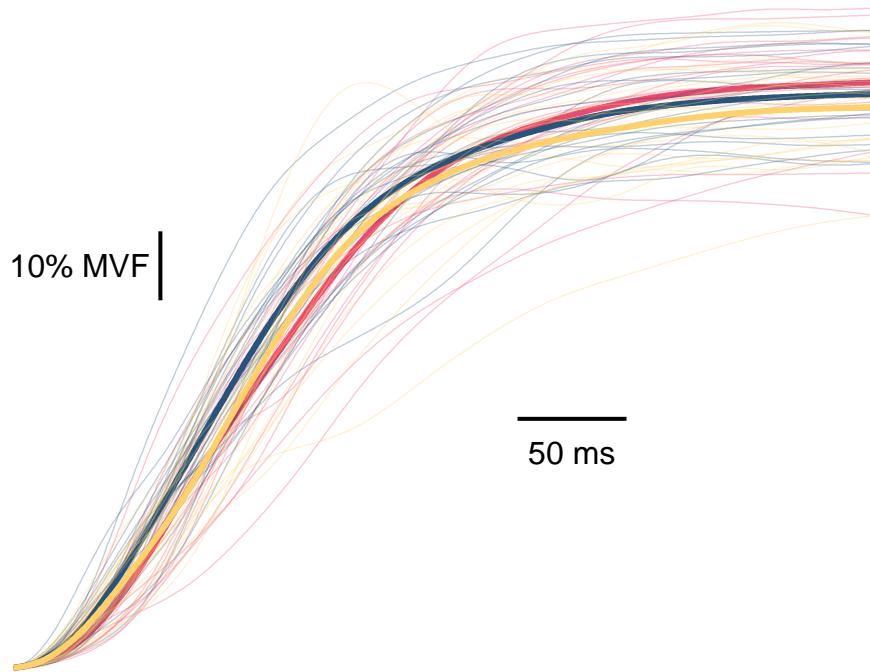
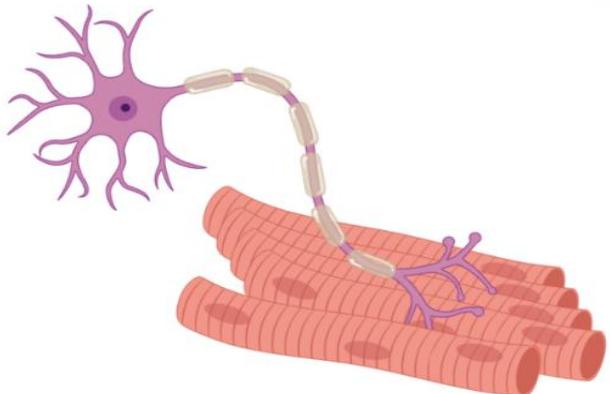
Rapid force production and MU firing in older vs. young individuals



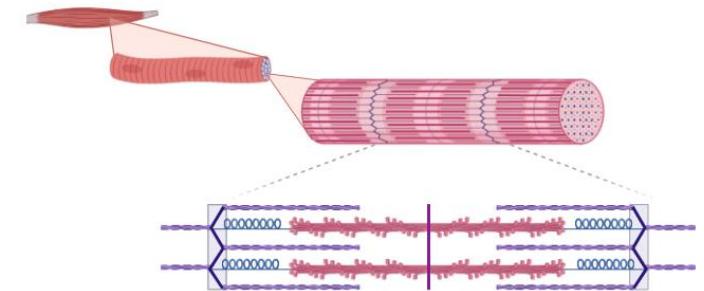
Chris Connelly

Rapid force production and MU firing in chronically trained individuals

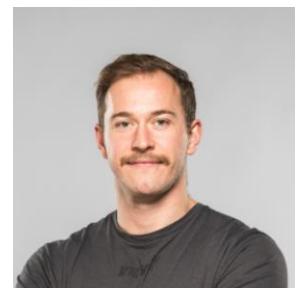
NEURAL



CONTRACTILE APPARATUS



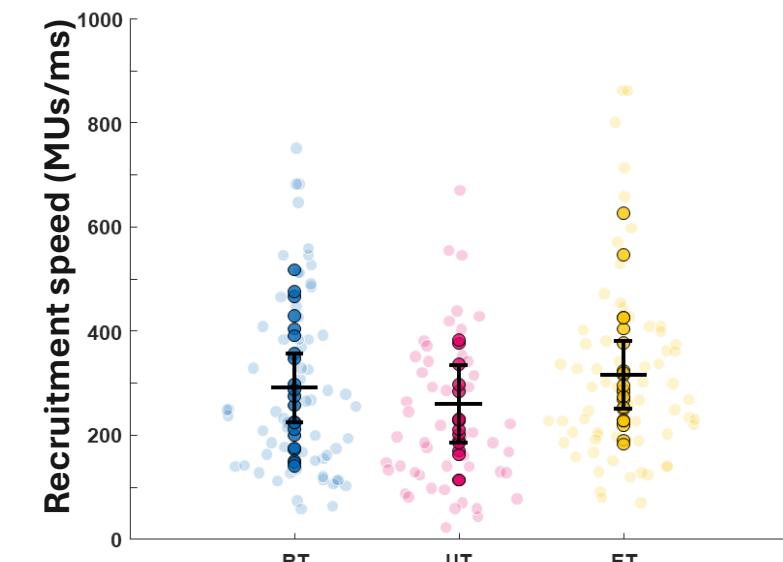
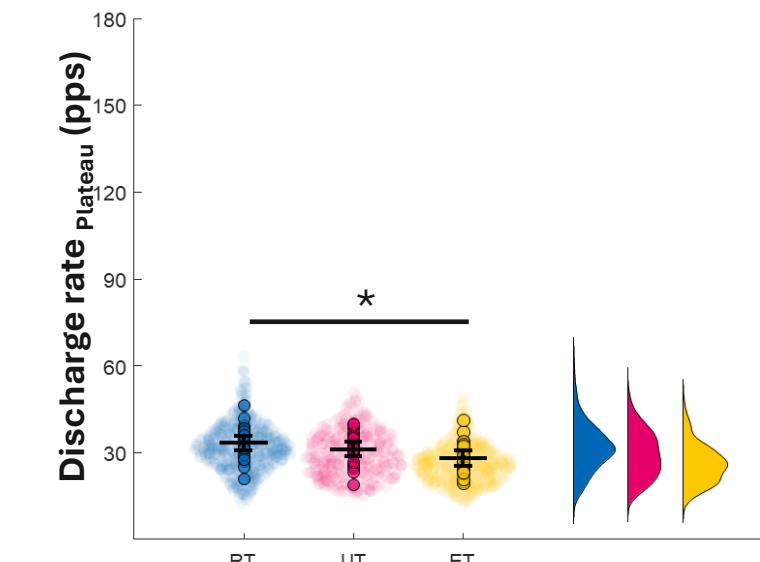
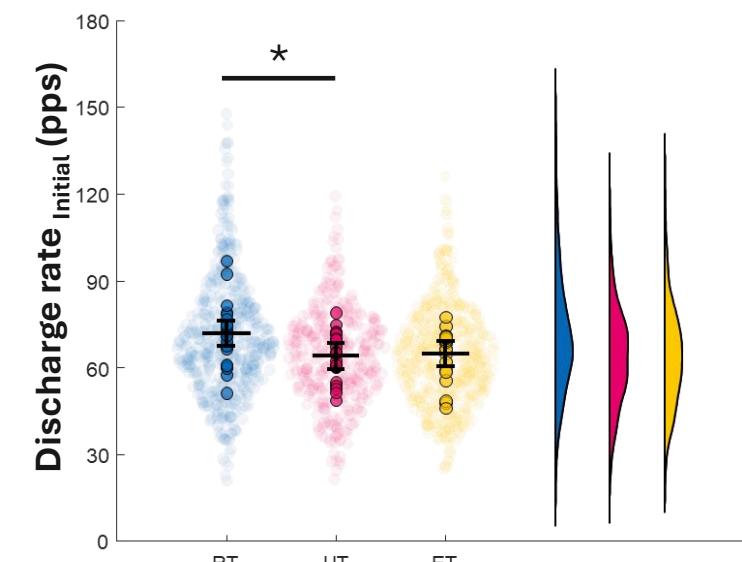
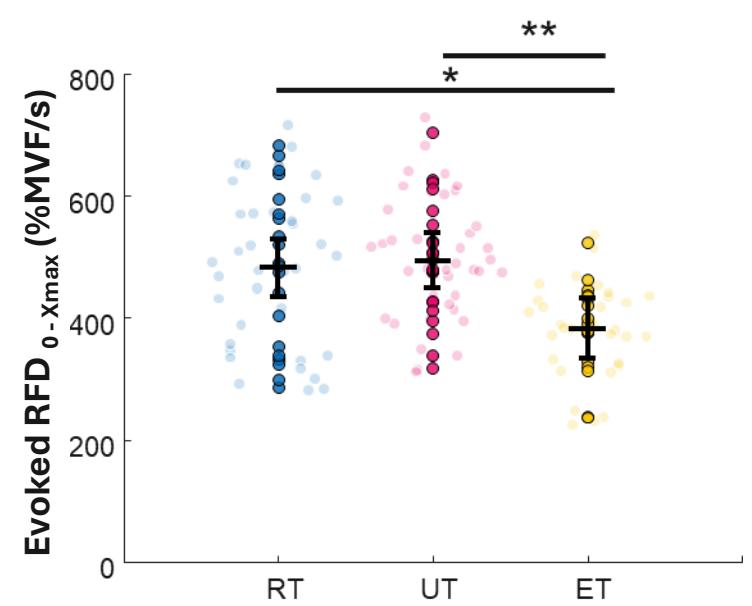
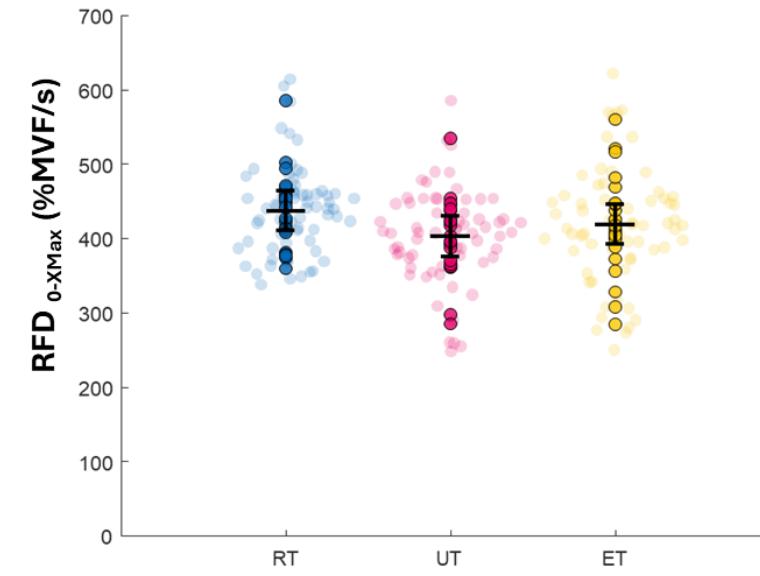
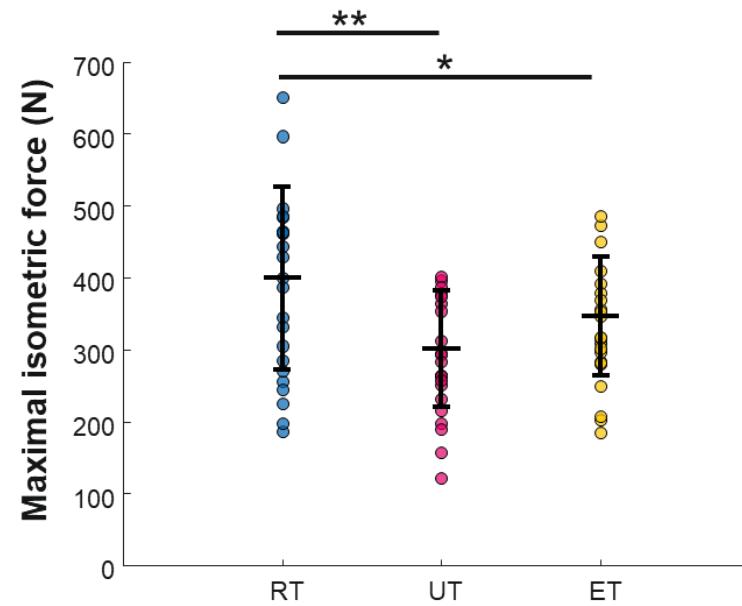
	RT (n=22)	UT (n=22)	ET (n=22)
Training Age (yrs)	9 ± 3	-	10 ± 4
IPAQ (MET-min/week)	$6401 \pm 2729^*$	4038 ± 2380	$6590 \pm 2128^{**}$
Mass (kg)	$84 \pm 17^{**}$	70 ± 12	68 ± 8
Height (m)	1.75 ± 0.08	1.73 ± 0.08	1.78 ± 0.08
Age (yrs)	23 ± 4	23 ± 3	24 ± 6



Haydn Thomason

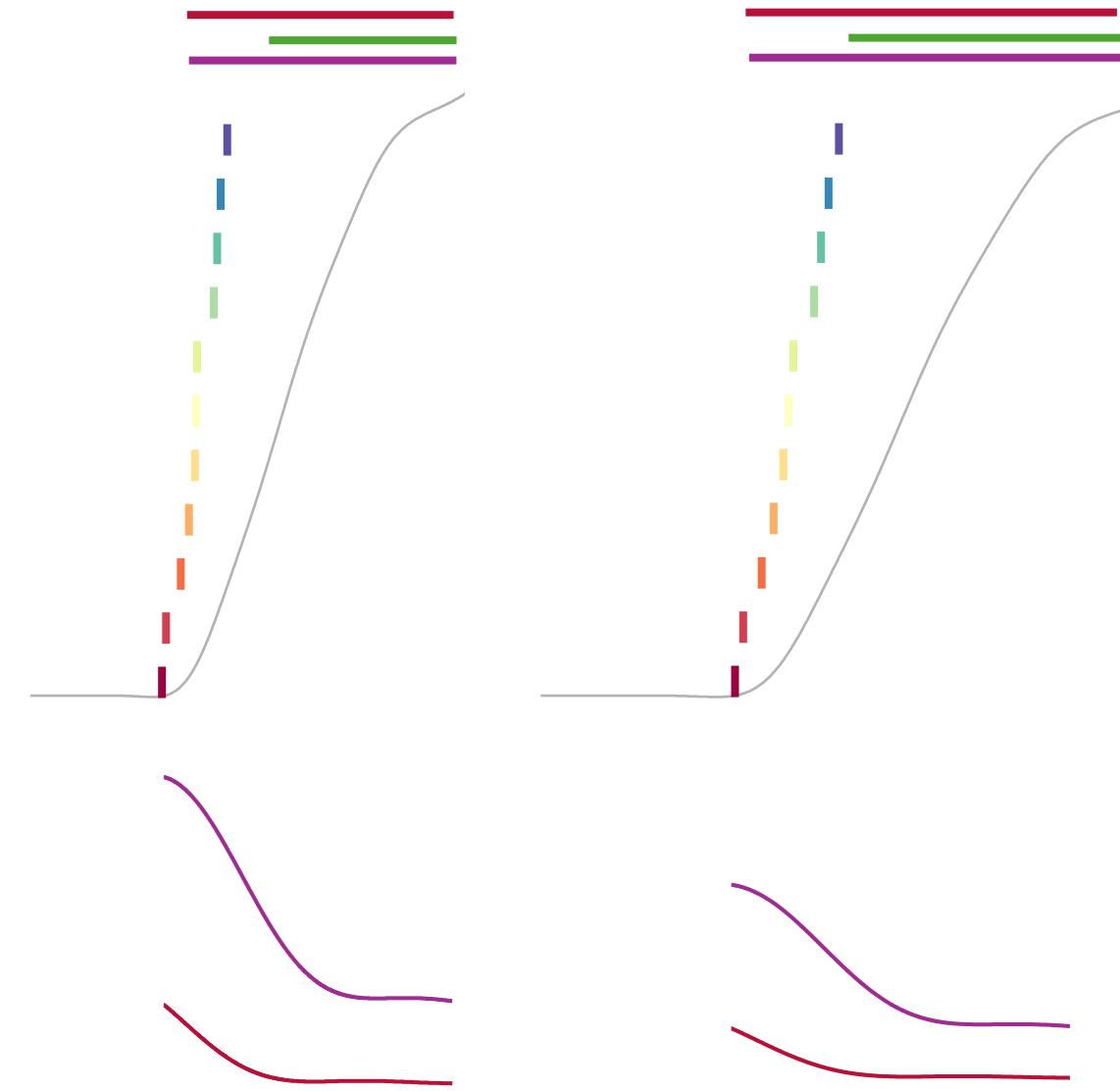
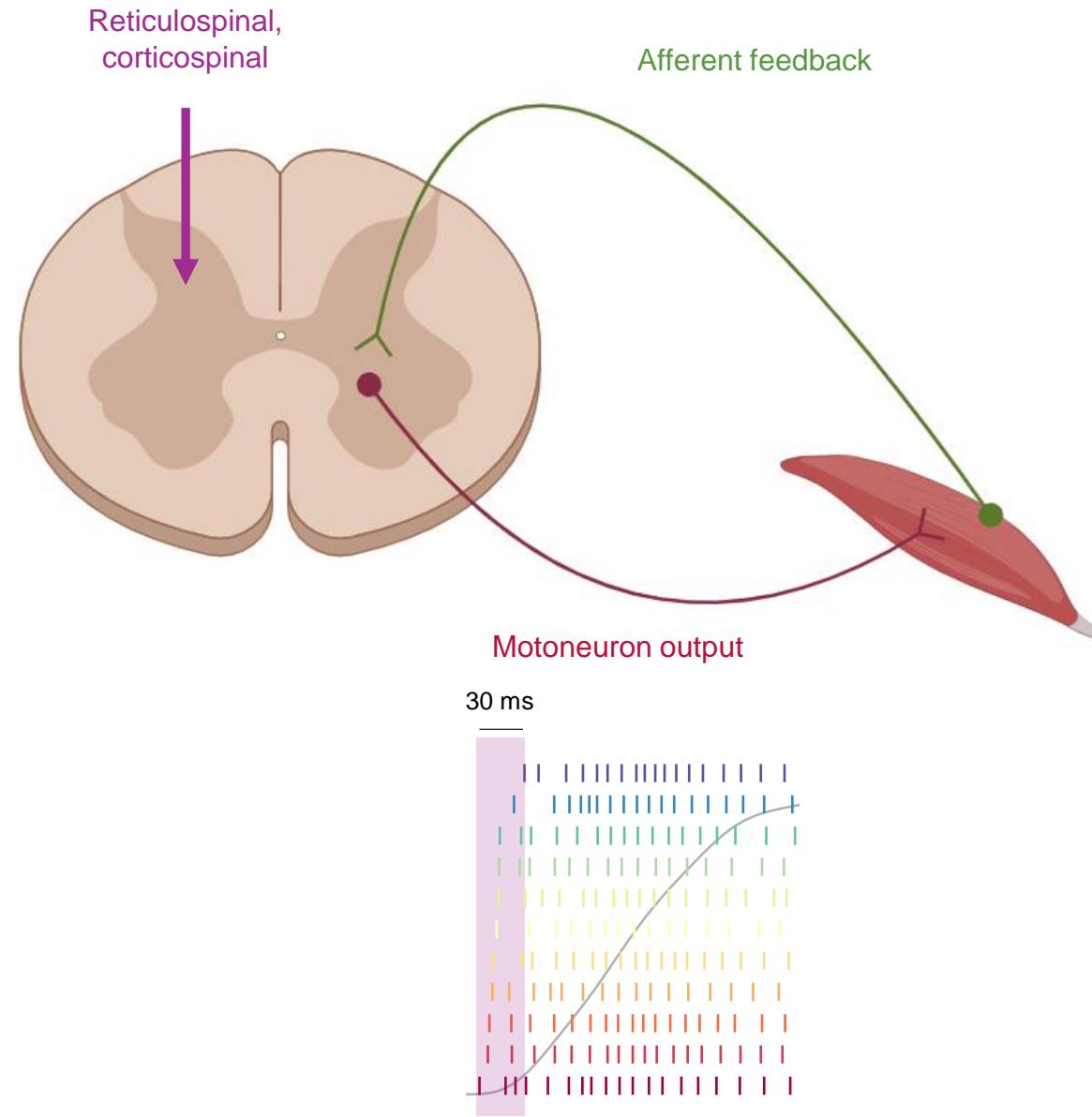
Rapid force production and MU firing in chronically trained individuals

● Resistance Trained
● Untrained
● Endurance Trained



***, p < 0.0001; **, p < 0.01; *, p < 0.05

Rapid contractions → maximal human *in vivo* motoneuron output

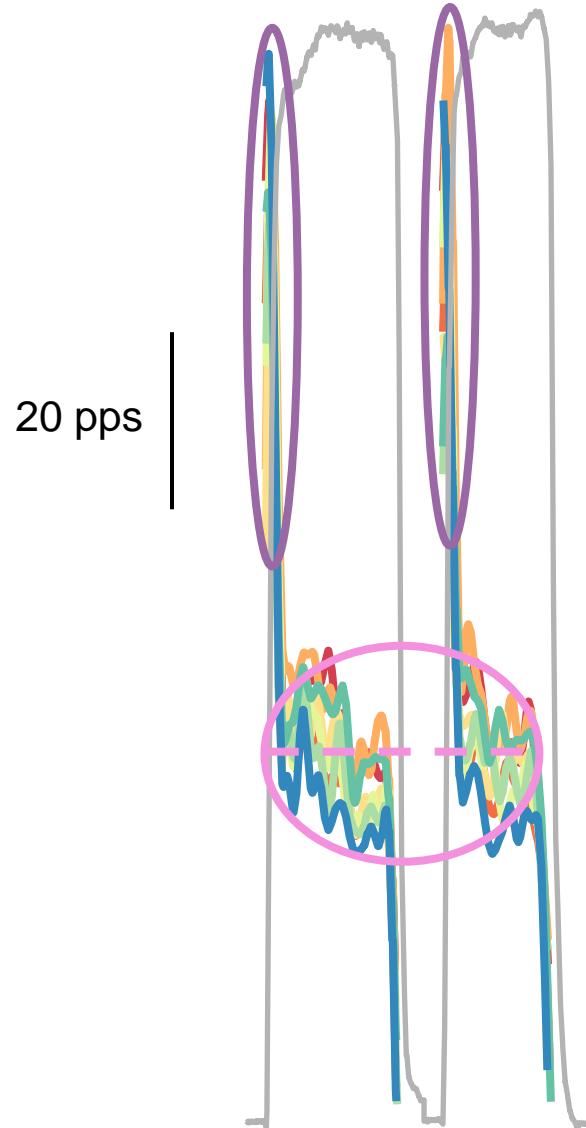


Reticulospinal,
corticospinal

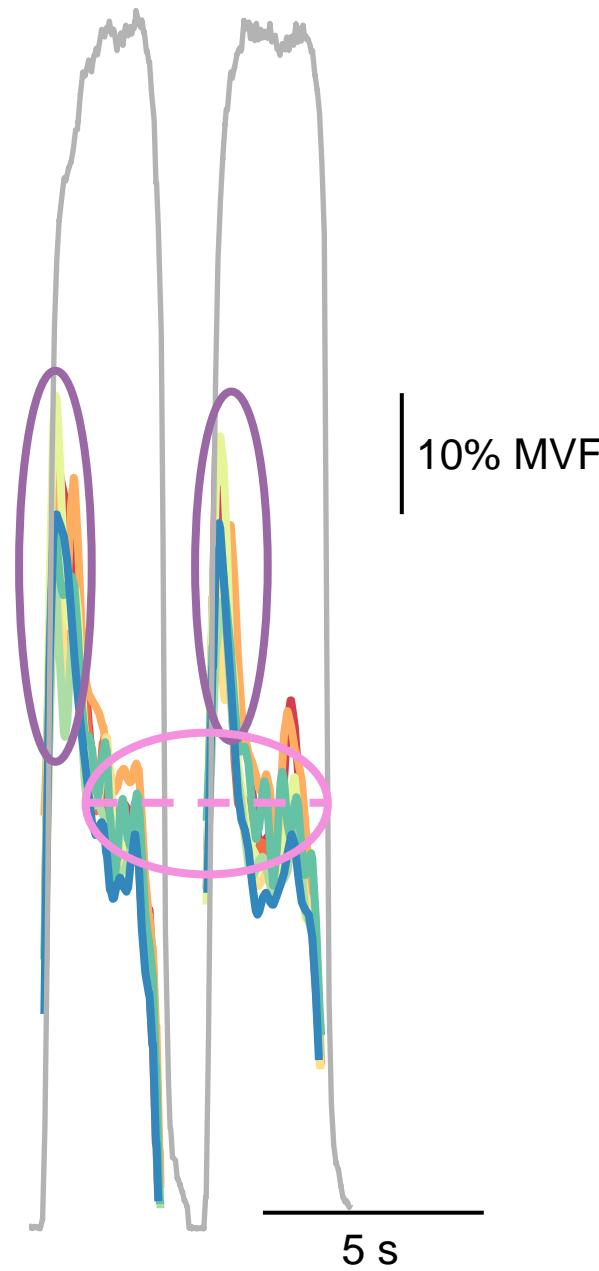
Afferent feedback

Rapid contractions → maximal human *in vivo* motoneuron output

Maximal rapid contraction
“as fast and as hard as possible”



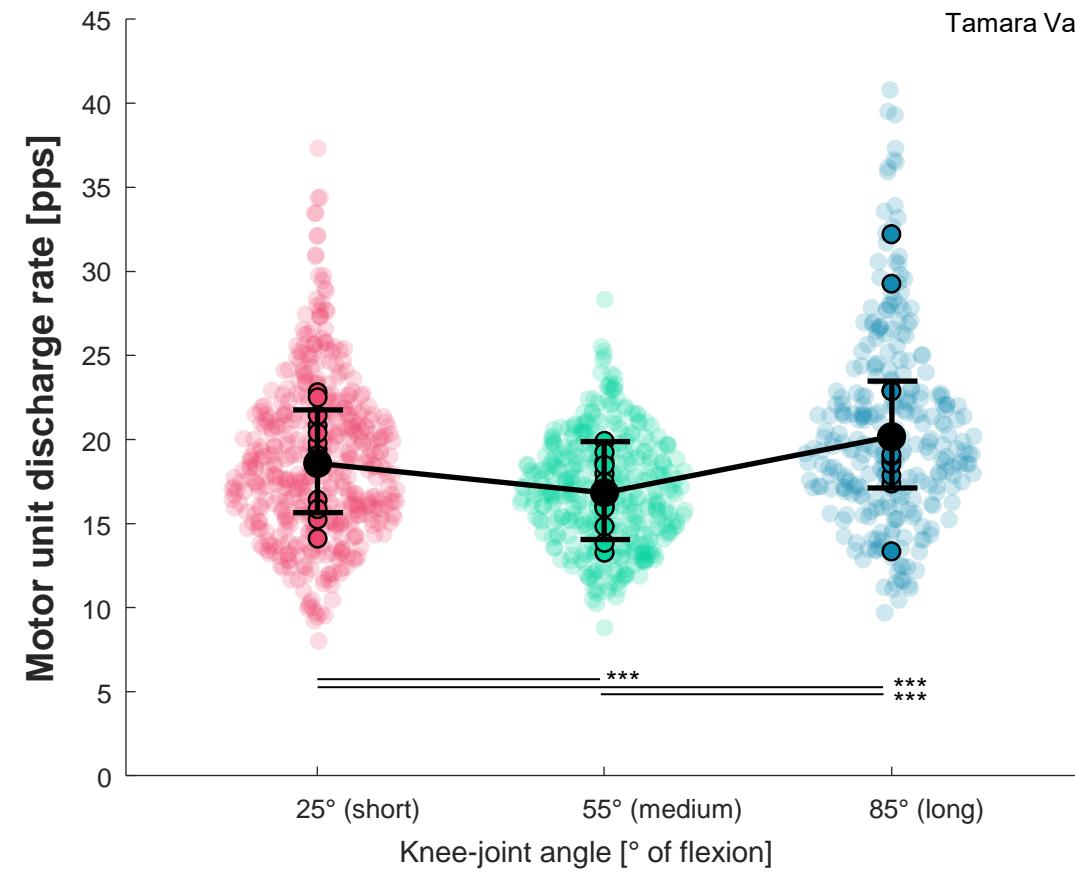
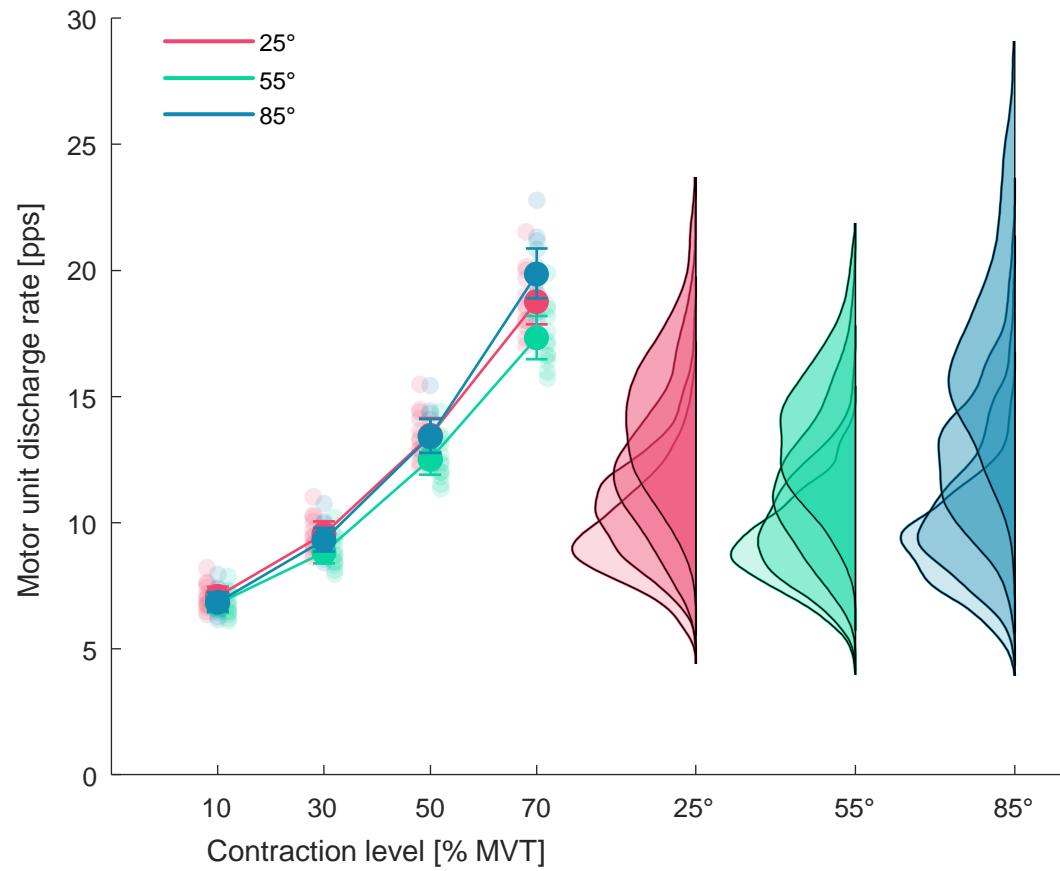
Maximal contraction
“as hard as possible”



Motor unit behaviour during maximal efforts



Tamara Valenčič



Thank you

Collaborators

- Prof Aleš Holobar, *Maribor, Slovenia*
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