

Dynamic Ensemble Bayesian Filter for Robust BCI Control of a Human with Tetraplegia

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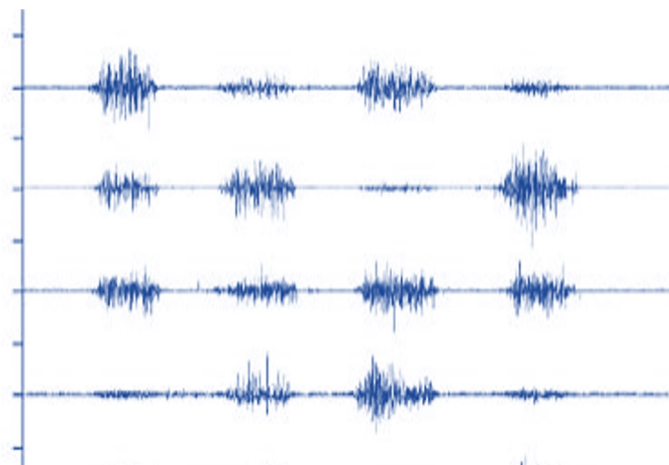
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Zhejiang University

The **Robustness** of BCI control is an important problem

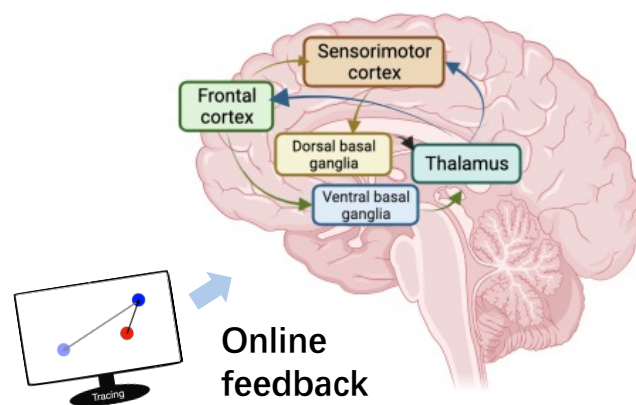
Challenge: the **variability** in neural signals

Noises

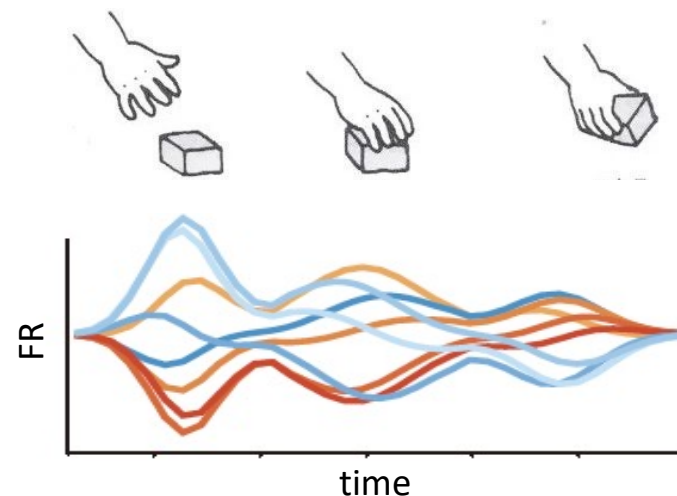


External or Biological

Dynamic Neural Pathways



Neural Functional Changes



Neural function changes when control speed changes or error occurs [\[Shenoy et al., 2017; Schwartz et al., 2018\]](#).

Static decoders makes inaccurate predictions

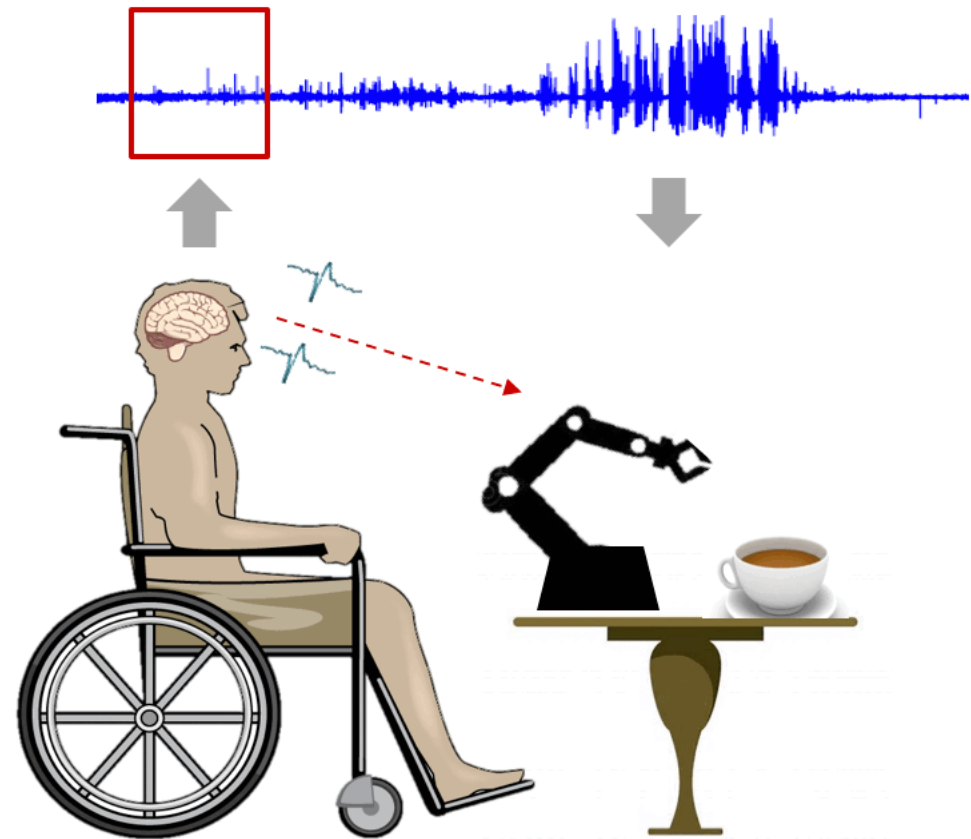
Variability in neural signals



Static neural decoders

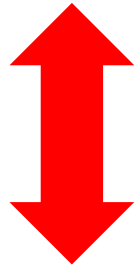


Instabilities in BCI control



Dynamic neural decoders to cope with neural variability

Variability in neural signals

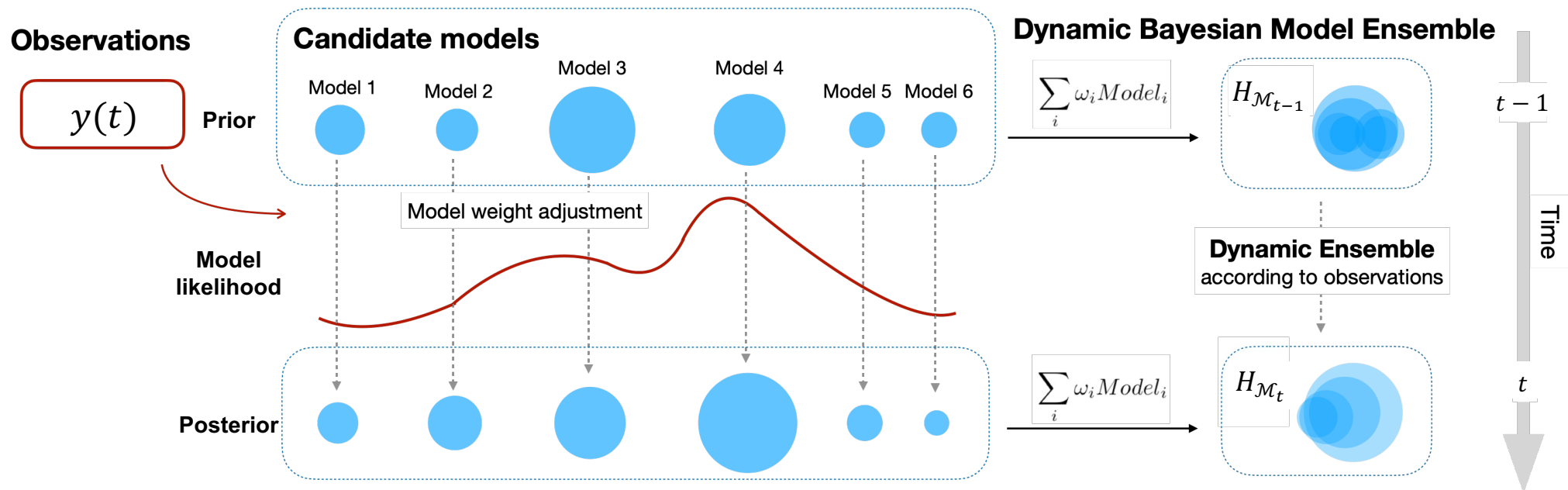


Dynamic neural decoders



But how???

The Dynamic Ensemble Bayesian Filter model (DyEnsemble)



Model weights estimation with Bayesian model averaging function

$$p(\mathcal{H}_{\mathcal{M}_t} = m_k | \mathbf{y}_{0:t}) = \frac{p(\mathcal{H}_{\mathcal{M}_t} = m_k | \mathbf{y}_{0:t-1}) p_k(\mathbf{y}_t | \mathbf{y}_{0:t-1})}{\sum_{j=1}^q p(\mathcal{H}_{\mathcal{M}_t} = m_j | \mathbf{y}_{0:t-1}) p_j(\mathbf{y}_t | \mathbf{y}_{0:t-1})}$$



Enabling model switching with changes in signals.

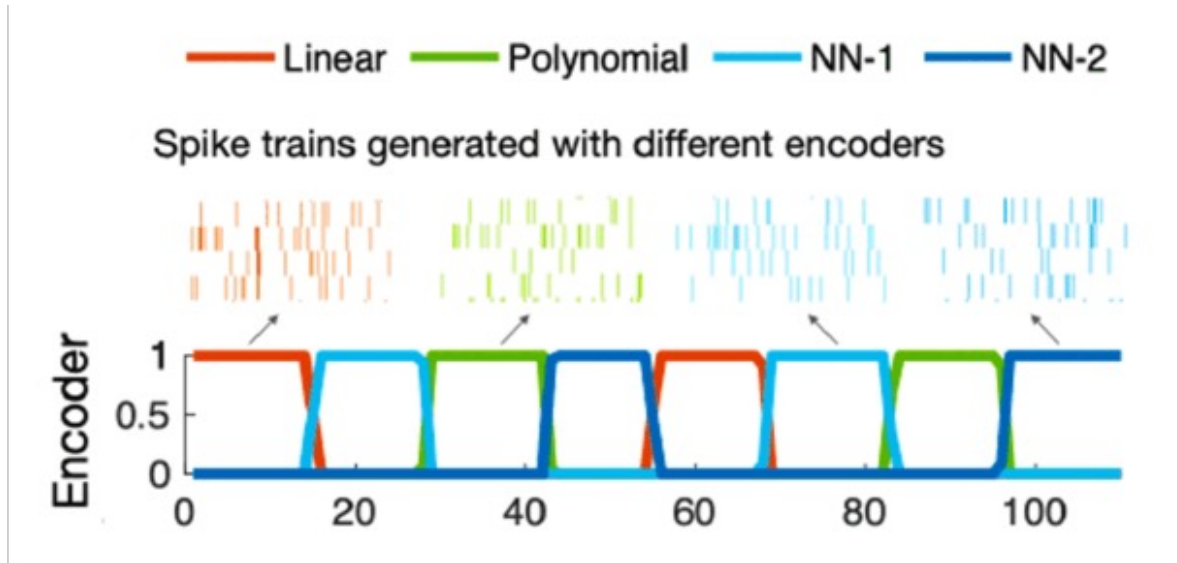
Neural signals



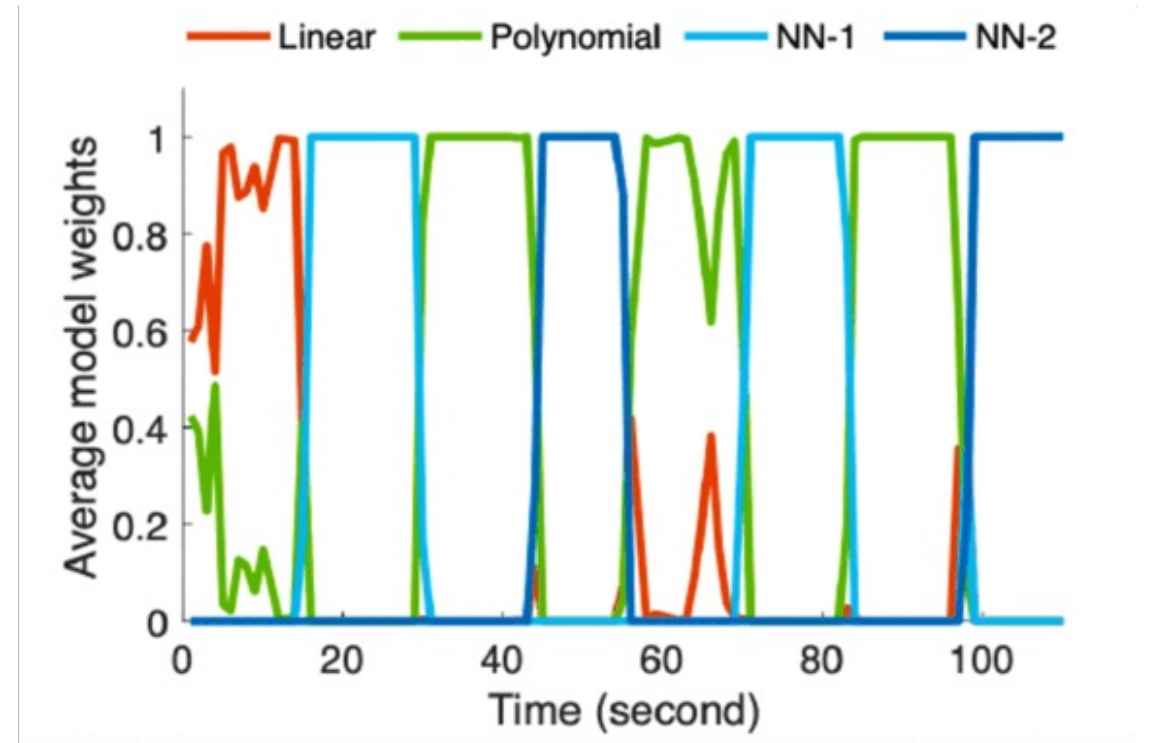
Model space



Simulation performance

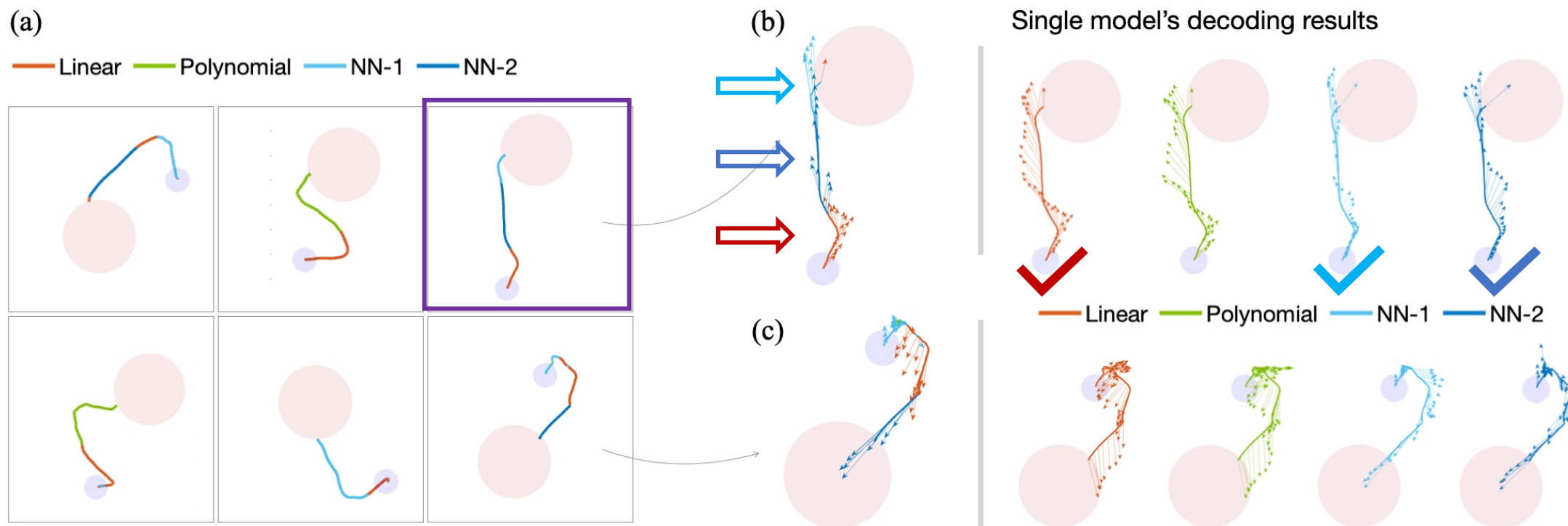


- Simulated neural signals with encoding model changes in time



- DyEnsemble **correctly assigns** model weights in simulations

Visualization of online BCI control



DyEnsemble **switches** to the '**optimal**' model for each time slot to achieve more accurate online control

Online performance comparison with KF

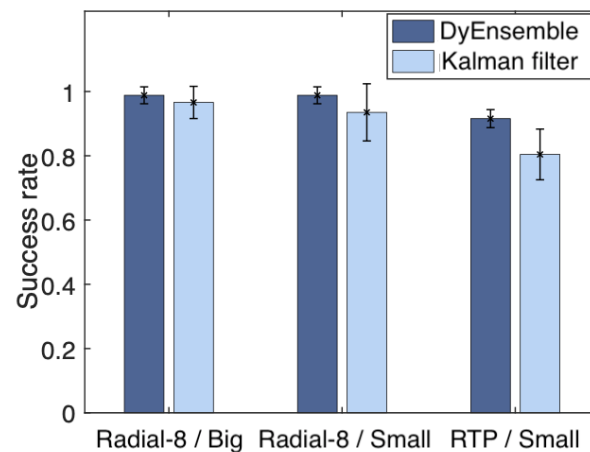
Improve the accuracy

max **13.9%**

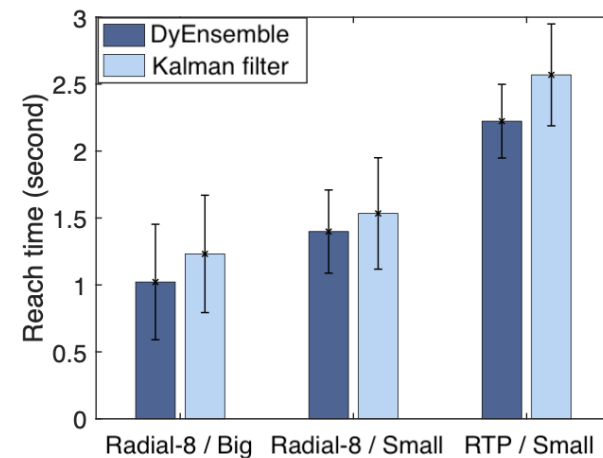


Improve the stability

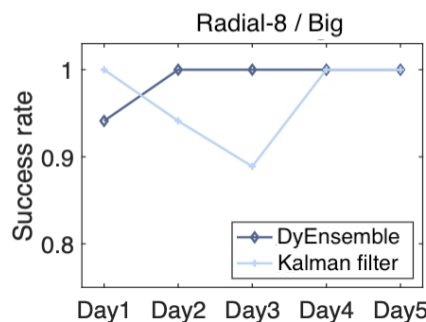
max **91%**



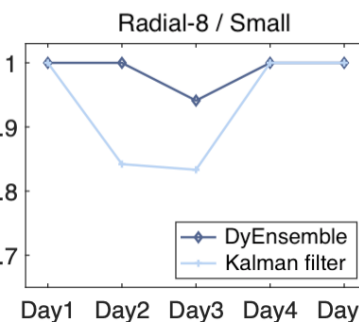
(a)



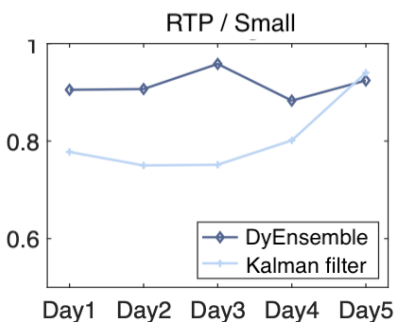
(b)



(c)



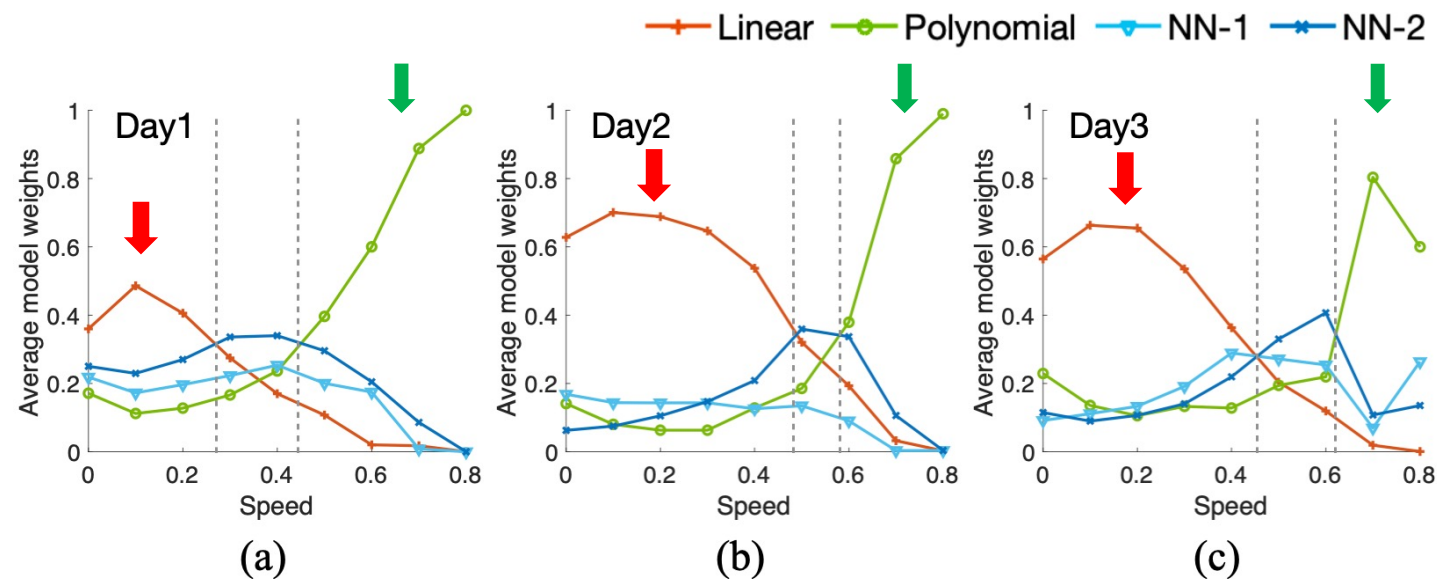
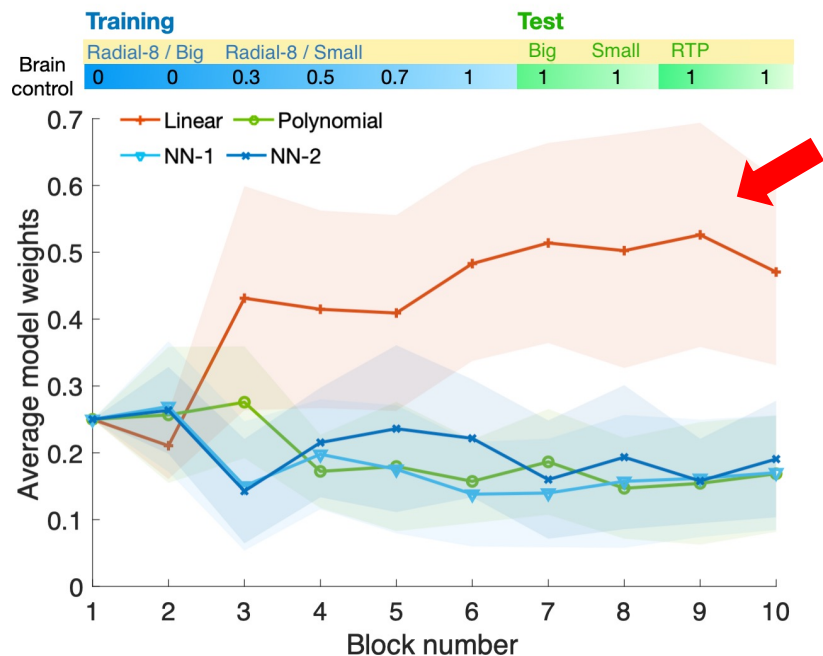
(d)



(e)

DyEnsemble significantly improves the control accuracy (increases the success rate by 13.9% in the random target pursuit task) and robustness (performs more stably over different experiment days).

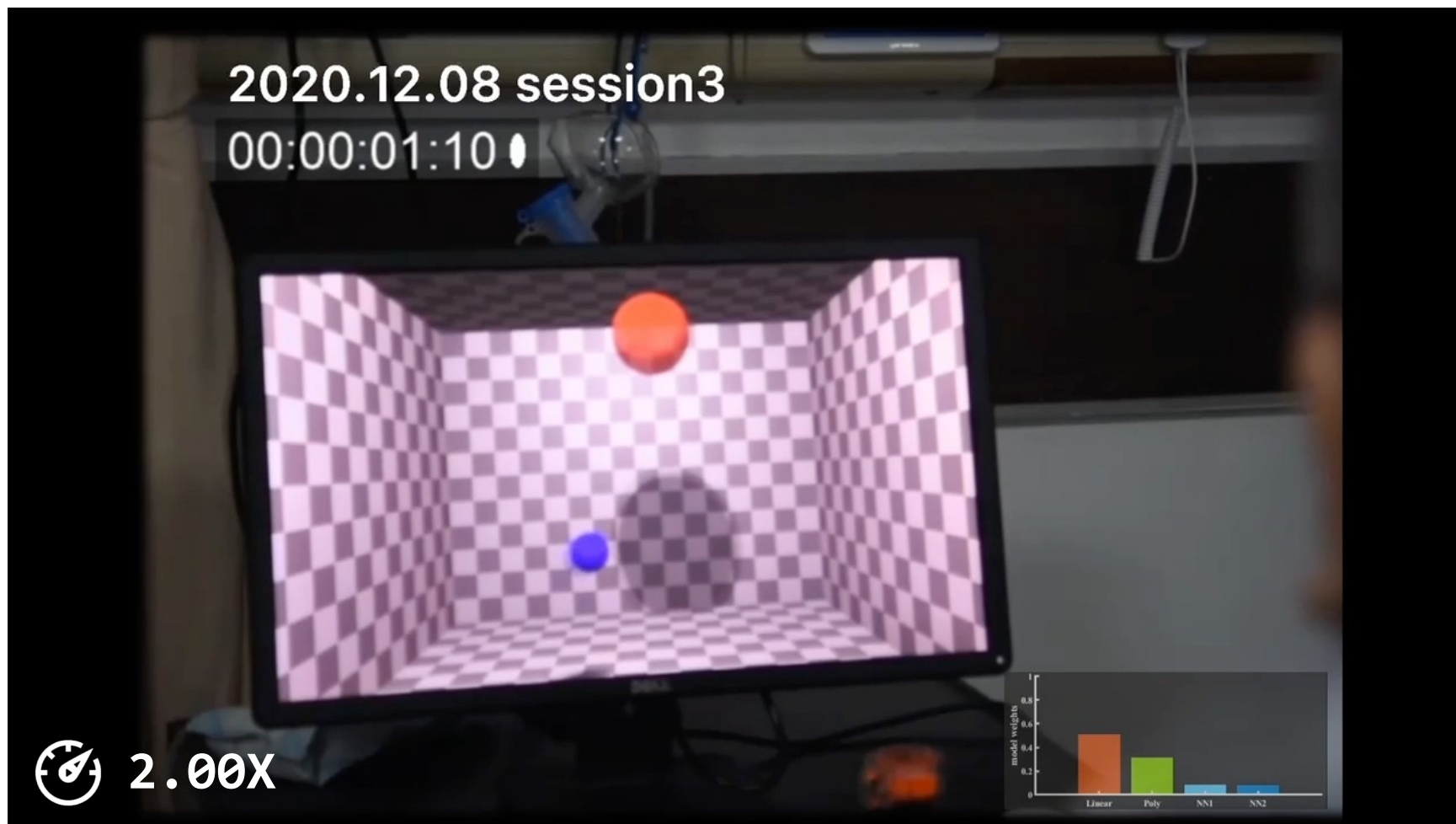
Analysis of the dynamic ensemble process



- Model ensemble changes during the closed-loop calibration, converging to a **linear dominant model** at full brain control stages.

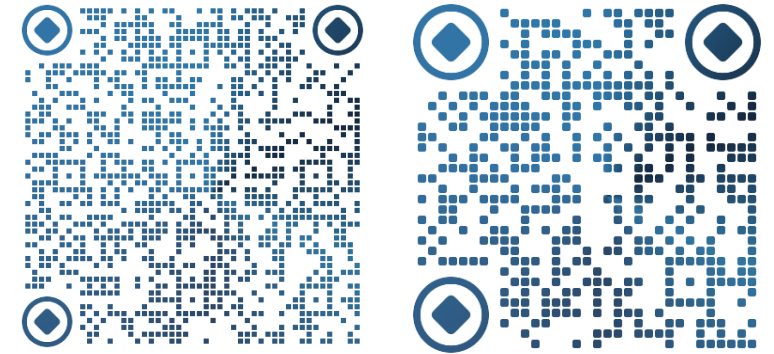
- Linear models are favored at **low speeds**
- Polynomial models cover the **high speeds**.

Now Uncle Bobo can happily play his favorite mahjong with BCI !



Thanks!

Scan me to download the papers:



Please find details in our papers:

- [1] Dynamic Ensemble Bayesian Filter for Robust Control of a Human Brain-machine Interface, *Arxiv* 2022 (***Trans. BME*** minor)
- [2] Dynamic Ensemble Modeling Approach to Nonstationary Neural Decoding in Brain-Computer Interfaces, ***NeurIPS*** 2019