

A More Details of the LFP Experiment

In this section, we give some detail on the Sync-RKM model. In order to incorporate SyncNet model [21] into our framework, the weight $W^{(x)} = [W^{(x_0)}, W^{(x-1)}, \dots, W^{(x-n+1)}]$ defined in Eq. (12) is parameterized as wavelet filters. If there is a total of K filters, then $W^{(x)}$ is of size $K \times C \times n$. Specifically, suppose the n -gram input data at time t is given as $X_t = [x_{t-n+1}, \dots, x_t] \in \mathbb{R}^{C \times n}$ with channel number C and window size n . The k -th filter for channel c can be written as

$$W_{kc}^{(x)} = \alpha_{kc} \cos(\omega_k t + \phi_{kc}) \exp(-\beta_k t^2) \quad (21)$$

$W_{kc}^{(x)}$ has the form of the Morlet wavelet base function. Parameters to be learned are α_{kc} , ω_k , ϕ_{kc} and β_k for $c = 1, \dots, C$ and $k = 1, \dots, K$. t is time grid of length n , which is a constant vector. In the recurrent cell, each $W_{kc}^{(x)}$ is convolved with the c -th channel of X_t using 1-d convolution. Figure 1 gives the framework of this Sync-RKM model. For more details of how the filter works, please refer to the original work [21].

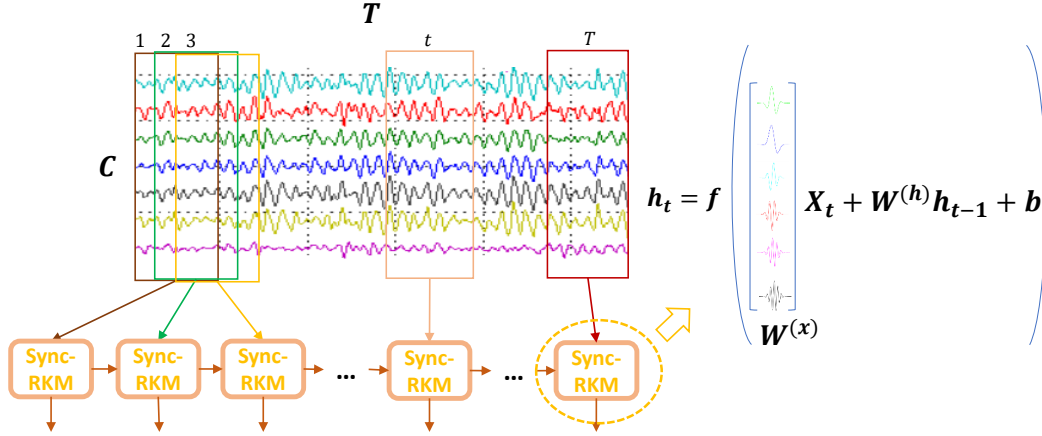


Figure 1: Illustration of the proposed model with SyncNet filters. The input LFP signal is given by the $C \times T$ matrix. The SyncNet filters (right) are applied on signal chunks at each time step.

When applying the Sync-RKM model on LFP data, we choose the window size as $n = 40$ to consider the time dependencies in the signal. Since the experiment is performed by treating each mouse as test iteratively, we show the subject-wise classification accuracy in Figure 2. The proposed model does consistently better across nearly all subjects.

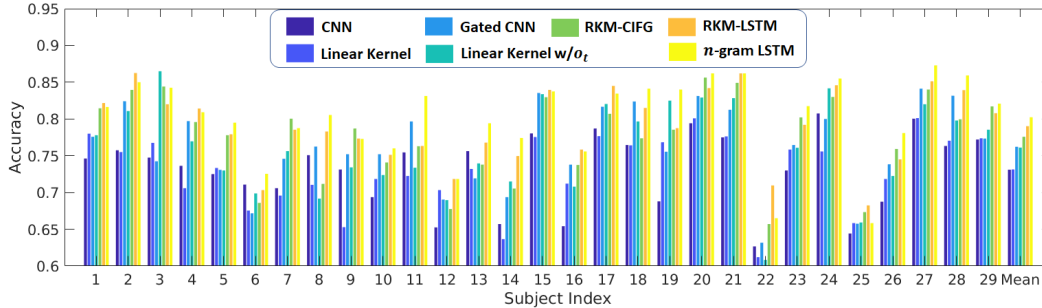


Figure 2: Subject-wise classification accuracy comparison for LFP dataset.