Effect of various recoding of granule cells on Pavlovian eyeblink conditioning in a cerebellar network

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• Cerebellar Motor Learning

- Precise temporal and spatial motor control for coordinating voluntary movements (e.g. posture, balance, and locomotion)
- Timing: Temporal information of movements [e.g. initiation or termination]
- Gain: Spatial information of movements

• Pavlovian Eyeblink Conditioning (EBC)

- Repeated presentations of paired conditioned stimulus (CS; tone) and (eyeblink-eliciting) unconditioned stimulus (US; airpuff).



- Eyelid conditioned response (CR) via learning representation of the time passage between the onsets of CS and US

• Purpose of Our Study

Investigation of Effect of Various Recoding of Granule Cells on Pavlovian EBC in A Cerebellar Ring Network by Varying The Connection Probability p_c from The Golgi Cells to The Granule Cells

Cerebellar Network for EBC

• Cerebellar Network for Pavlovian EBC

- Granular Layer: Input layer of cerebellar cortex Excitatory granule (GR) cells & Inhibitory Golgi (GO) cells
- Purkinje-Molecular Layer:

Output Layer Inhibitory Purkinje cells (PCs) & basket cells (BCs)

- Context signal for the EBC via mossy fiber (MF): Transient Conditioned Stimulus (CS) for resetting and sustained CS (representing the tone)



- Cerebellar Cortex Purkinje-Molecular Granular Laver Layer GO BC PF CF (error-teaching signal) GR PC CS (Tone) CN IO US (Airpuff) MF (eyeblink conditioning signal) Desired timing signal CR (Eyeblink)
 - Desired timing signal Unconditioned Stimulus (US) into inferior olive (IO)



Cerebellar Ring Network & Synaptic Plasticity Rule

Cerebellar Ring Networks

- Granular-Layer Ring Network $N_C (= 2^{10})$ GR clusters $N_{GR} (= 50)$ GR cells in each GR cluster N_C GO cells Each GR cluster bounded by four glomeruli (GL): Upper & lower GLs Each GL: One MF & ~ 2 GO cells $(p_c = 0.029)$
- Purkinje-Molecular-Layer Ring Network N_{PC} (= 16) PCs and basket cells (BCs)

• Refined Rule for Synaptic Plasticity

 $J_{ij}^{(\text{PC,PF})}(t) \rightarrow J_{ij}^{(\text{PC,PF})}(t) + \Delta \text{LTD}_{ij}^{(1)} + \Delta \text{LTD}_{ij}^{(2)} + \Delta \text{LTP}_{ij}$

- Synaptic modification (LTD or LTP) depending on the relative time difference between CF & PF activation times
 - $\Delta LTD_{ij}^{(1)}$: Major LTD in the case that the CF signal is associated with earlier PF signals
 - $\Delta LTD_{ij}^{(2)}$: Minor LTD in the case that the CF signal is related to later PF signals
 - ΔLTP_{ij} : LTP in the presence of PF signals alone without association with the CF signal





$$\Delta J_{LTD}(\Delta t) = A + B \cdot e^{-[(\Delta t - t_0)/\sigma]^2}$$

$$A = -0.12, B = 0.4, t_0 = 80, \sigma = 180$$

Various Firing Patterns in The GR Clusters

Optimal case: $p_c^* = 0.029$

• Firing Activity in The Whole GR Cells

- Raster plot of spikes of 10³ randomly chosen GR cells: Beginning of trial stage: all GR cells fire due to strong transient CS signal Remaining part of the trial stage: GR cells make random repetition of transitions between active and inactive states because of sustained CS signal
- Instantaneous whole-population spike rate $R_{GR}(t)$: Basically in proportion to the transient and Well-Matched Spiking Patterns ($\mathcal{C}^{(l)} > 0$) sustained CS inputs via MFs

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- Due to the inhibitory effect of GO cells, the overall firing rates: Uniformly lowered

• Various Firing Patterns in **GR** clusters



- Various firing patterns $R_{GR}^{(l)}(t)$: Well- & ill-matched

with respect to US $[f_{\text{IIS}}(t)]$

Characterization of Various Firing Patterns

- Matching index $\mathcal{M}^{(I)}$: Cross-correlation between $R_{GR}^{(l)}(t)$ and $f_{US}(t)$ at the zero-time lag Well-matched firing group $\rightarrow \mathcal{M}^{(l)} > 0$ Ill-matched firing group $\rightarrow \mathcal{M}^{(l)} < 0$ - Variety Degree \mathcal{V} : Relative standard deviation of $\{\mathcal{M}^{(l)}\}\ \mathcal{V}=1.842$









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Change in PF-PC Synaptic Weights during Learning

• Distribution of Normalized Active PF-PC Synapses

- Combination of separate top horizontal line with a central gap and lower band
- Horizontal line: No essential change with trials

Arising from ill-matched firing group with a central gap

No LTD because of no associations with error-teaching CF signal



 Lower band: Increase in vertical width & Saturated at about the 250th trial Arising from well-matched firing group Strong LTD due to good association between the well-matched PF and CF signals

• Bin-averaged Synaptic Weights $\langle \tilde{J} \rangle$ of Active PF

- Step-well-shaped curve
- With the trial, step-well curve comes down, increase in its width & depth, and saturation at about the 250th trial



Change in Firing Activity of PCs during Learning



• Raster Plots of Spikes

Increase in trial, more sparse at the middle part due to strong LTD

• Population Spike Rates $\langle R_{PC}(t) \rangle_r$ of PCs

- Step-well-shaped curve
- With the trial, the step-well in the middle part comes down, a zero-bottom appears in the middle part at the 141th trial
 - \rightarrow CN (cerebellar nucleus) neuron which evokes CR (conditioned response: eyeblink) CN neuron fire spikes
- With increase in the trial from the 141th trial, increase in the (top) width of the step well and the width of the zero-bottom.

Change in Firing Activity of CN during Learning

Firing Activity of CN during Learning

- 140th trial: No firing due to strong inhibition from PC
- CN neuron begins to fire in the middle part at the 141th trial due to $\frac{2}{50}$ appearance of a zero-bottom in PC 🦉
- 200 800 200 800 200 200 800 t (msec) - With increasing the trial, raster plots



of spikes of the CN neuron become more dense in the middle part

- Instantaneous individual firing rates $f_{CN}(t)$: Bell-shaped curve.
- With the trial, increase in bottom-base width and peak height, and saturated at about the 250th trial

Learning Efficiency Degree

- Timing degree T_d : Matching degree between the firing activity of CN and US signal

Cross-correlation between $f_{CN}(t) \& f_{US}(t)$ at the zero-time

lag: Reflecting width of the bottom base of the bell curve With the trial, decrease in \mathcal{T}_d , and saturated at about the 250th trial

- Strength \mathcal{S} of CR: Representing the amplitude of the eyelid closure Modulation [(maximum – minimum)/2] of $f_{CN}(t)$
- With the trial, increase in S, and saturated at about the 250th trial - Learning efficiency degree \mathcal{L}_e for CR: $\mathcal{L}_e = \mathcal{T}_d \cdot \mathcal{S}$

With the trial, increase in \mathcal{L}_{e} , and saturated at about the 250th trial





Learning Progress in The IO System

Learning Progress

 Two inputs into IO: Excitatory US signal for the desired timing and the inhibitory signal from CN (representing a realized eye-movement)



- After acquisition of CR, with increasing trial, increase in inhibitory input from the CN, and convergence to the constant excitatory input through the US signal.
- Learning progress degree $\mathcal{L}_p = \overline{I_{\text{GABA}}^{(\text{IO},\text{CN})}} / \overline{\left|I_{\text{AMPA}}^{(\text{IO},\text{US})}\right|}$

With the trial, increase in \mathcal{L}_p , and saturated at 1 at about the 250th cycle.

• Firing Activity of IO neuron during Learning

- Before the 141th threshold trial, dense spikes appear in the middle part due to excitatory US signal.
- With increasing the trial from the threshold, spikes become sparse, because of increased inhibitory input



from the CN neuron, and saturated at abut the 250th trial

Summary

• Various Temporal Recoding in The GR clusters

Appearance of various well- and ill-matched firing patterns, due to inhibitory coordination of GO cells

Characterized in terms of matching index and variety degree

• Effect of Various Recoding on The EBC

Effective depression at the PF-PC synapses Well-matched PF signals: Strong LTD by the CF signals, Ill-matched PF signals: Practically no LTD

 \rightarrow Effective modulation in firing of PCs & CN Neuron

Relation between Various Recoding and Learning Efficiency Degree

- Strong Correlation between Variety Degree \mathcal{D} and Saturated Learning Efficiency Degree \mathcal{L}_e^* Bell-shaped curves with maximum at the same optimal value of $p_c^* = 0.029$

The more various in temporal recoding of granule cells

→ The more effective in motor learning for the Pavlovian EBC

