

Effect of Diverse Spiking Patterns of Granule Cells on Optokinetic Response in The Cerebellum

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- **Cerebellar Motor Learning**

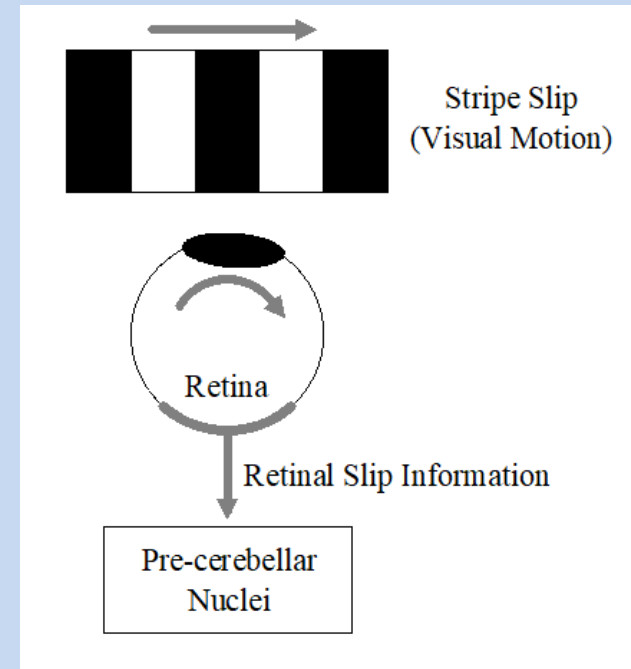
- Fine motor control for coordinating voluntary movements such as posture, balance, and locomotion, resulting in smooth and balanced muscular activity

- **Optokinetic Response (OKR)**

- Eye tracks successive stripe slip with the stationary head
- Composed of two consecutive slow and fast phases (i.e., slow tracking eye-movement and fast reset saccade).

- **Purpose of Our Study**

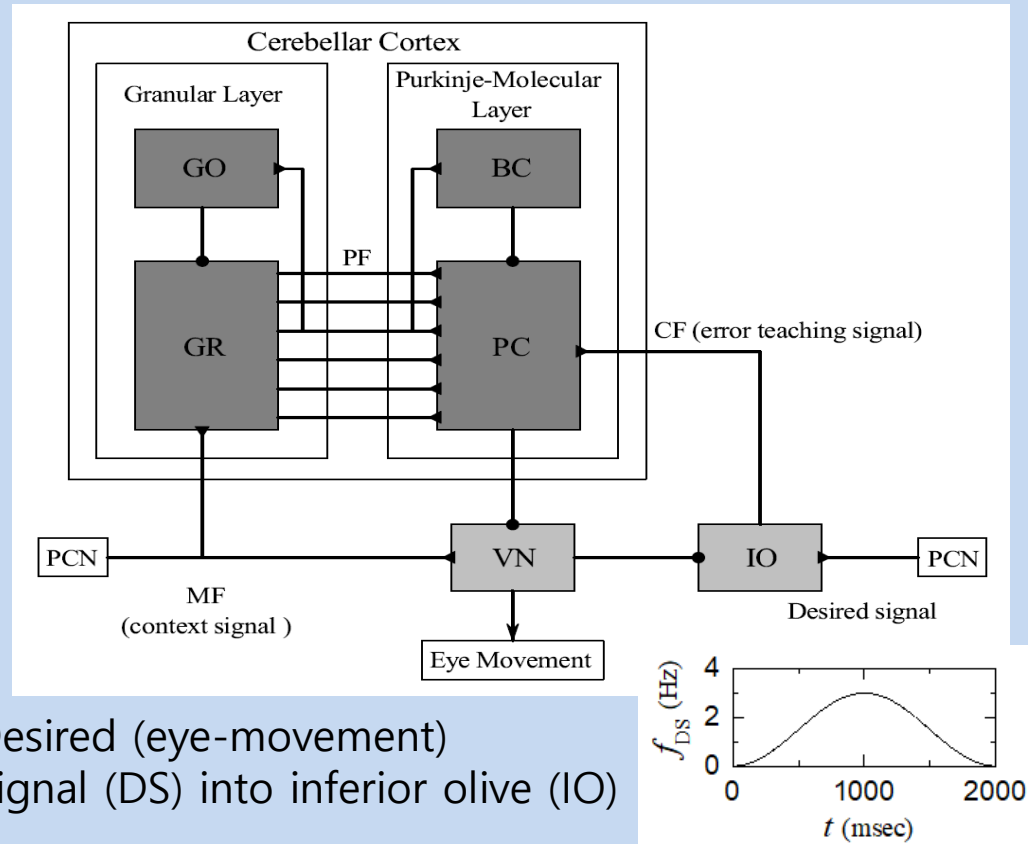
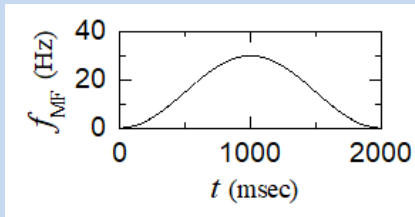
Investigation of Effect of Diverse Recoding of Granule Cells on Gain of OKR in A Cerebellar Ring Network by varying the connection probability p_c from the GO cell to the granule cells



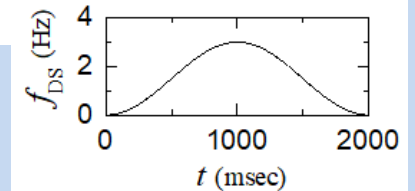
Cerebellar Network for OKR

• Cerebellar Network for OKR

- Granular Layer: Input layer
Excitatory granule (GR) cells
& Inhibitory Golgi (GO) cells
- Purkinje-Molecular Layer:
Output Layer
Inhibitory Purkinje cells (PCs)
& basket cells (BCs)
- Two External Sensory Signals:
Context signal for the post-eye-
-movement via mossy fiber (MF)



Desired (eye-movement)
Signal (DS) into inferior olive (IO)



• Recoding Process in The Granular Layer

- Recoding of MF inputs into more sparse and less similar (i.e., more orthogonal) patterns
- Recoded inputs are fed into the PCs via the parallel fibers (PFs)

• Synaptic Plasticity at The PF-PC Synapses

- PCs receive both the recoded PF signals (from GR cells) and the error-teaching climbing fiber (CF) signals (from IO neurons)
- Change in the synaptic weight of PF-PC synapses

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

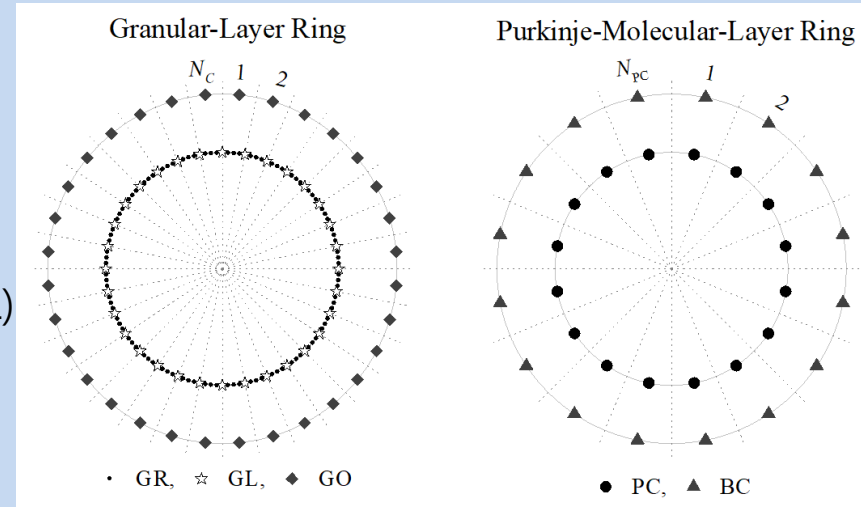
$N_C (= 2^{10})$ GO cells

Each GR cluster bounded by two glomeruli (GL)

Each GL: One MF & ~ 5 GO cells ($p_C = 0.06$)

- Purkinje-Molecular-Layer Ring Network

$N_{PC} (= 16)$ PCs and basket cells (BCs)



• Refined Rule for Synaptic Plasticity

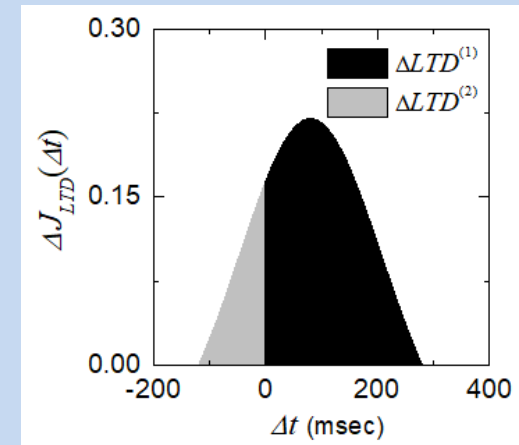
$$J_{ij}^{(PC,PF)}(t) \rightarrow J_{ij}^{(PC,PF)}(t) + \Delta LTD_{ij}^{(1)} + \Delta LTD_{ij}^{(2)} + \Delta LTP_{ij}$$

- 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



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

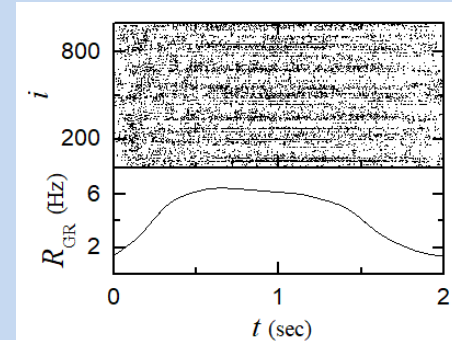
$$A = -0.12, B = 0.4, t_0 = 80, \sigma = 180 \quad 3$$

Diverse Spiking Patterns in The GR Clusters

Optimal case: $p_c^* = 0.06$

• Firing Activity in The Whole GR Cells

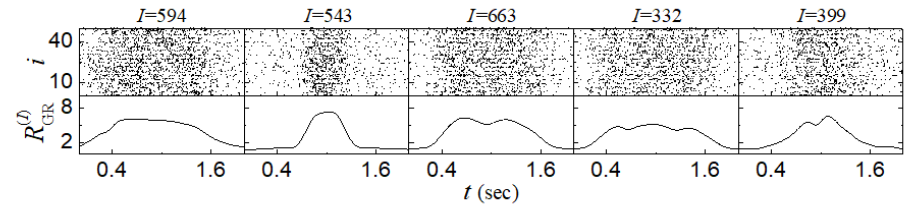
- Raster plot of spikes of 10^3 randomly chosen GR cells:
Initial & final stages of the cycle: Sparse and uniform firing
Middle stage: Dense and non-uniform firing
- Instantaneous whole-population spike rate $R_{GR}(t)$:
Basically in proportion to the MFs.



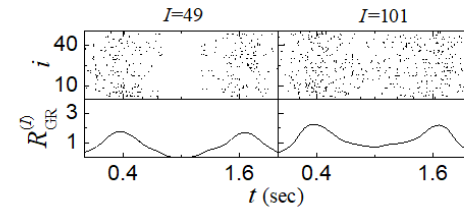
• Diverse Spiking Patterns in GR clusters

- Diverse spiking patterns $R_{GR}^{(I)}(t)$; in-phase, anti-phase, or complex out-of-phase with respect to their population averaged firing activity $R_{GR}(t)$.

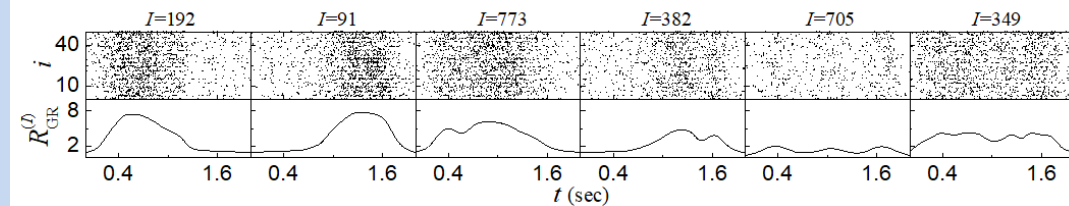
In-phase Spiking Patterns



Anti-phase Spiking Patterns

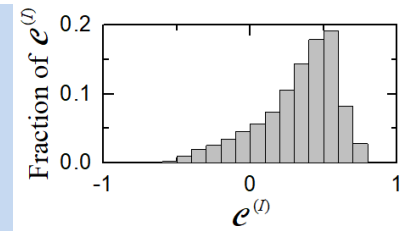


Complex Out-of-phase Spiking Patterns



• Characterization of Diverse Spiking Patterns

- Conjunction index $e^{(I)}$: Cross-correlation between $R_{GR}^{(I)}(t)$ and $R_{GR}(t)$ at the zero-time lag
- Diversity Degree \mathcal{D} : Relative standard deviation of $\{e^{(I)}\}$
 $\mathcal{D}=1.613$



Change in PF-PC Synaptic Weights & PCs' Activity during Learning

• Effective Depression at PF-PC Synapses

- Distribution of Synaptic Weights of Active PF Signals

In-phase PF signals: Strongly

depressed (i.e., strong LTD) by the in-phase CF signals

Out-of-phase PF signals: Weakly depressed (i.e., weak LTD) due to the phase difference between the PF and the CF signals.

Middle stage of cycle: Strong LTD via dominant contributions of in-phase PF spikes

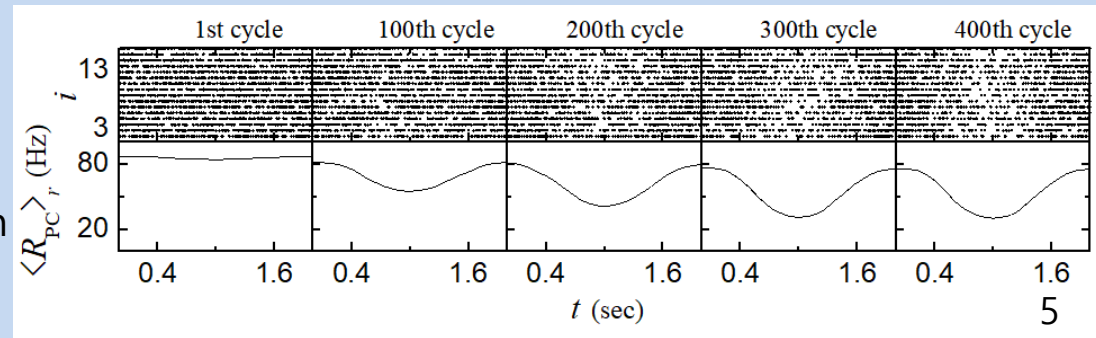
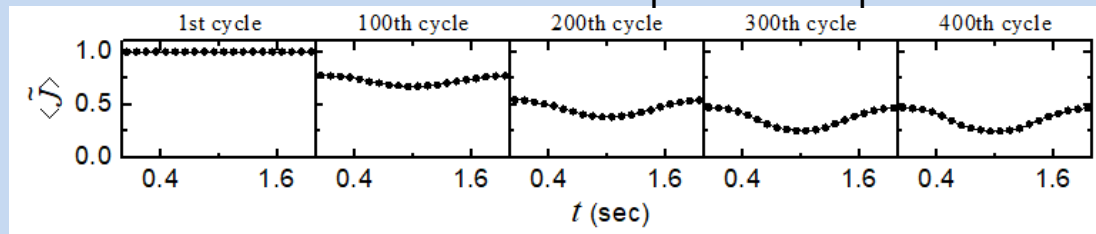
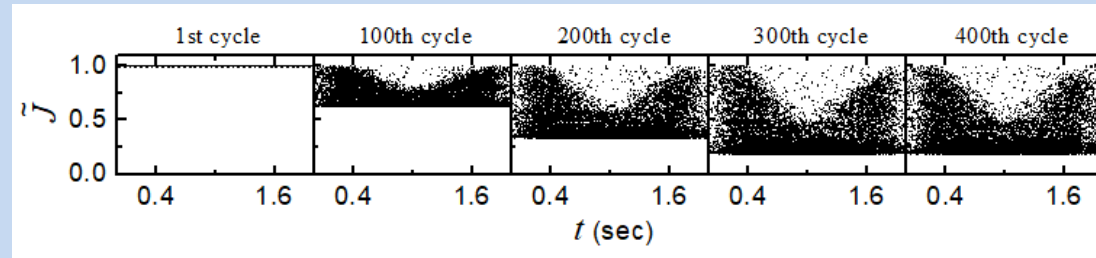
Initial & final stages: Weak LTD via contributions of both in- & out-of-phase PF spikes

- Bin-averaged synaptic weights of active PF signals: Well-shaped curve
- With the cycle, the well curve comes down, increase in modulation

[=(maximum - minimum)/2], and saturation at about the 300th cycle.

• Firing Activity of PCs during Learning

- Raster plots of spikes: with the cycle, more sparse at the middle stage.
- Population spike rates of PCs: Well-shaped curve with big modulation due to effective depression of PF-PC synapse



Change in VN's Firing Activity and Learning Gain Degree

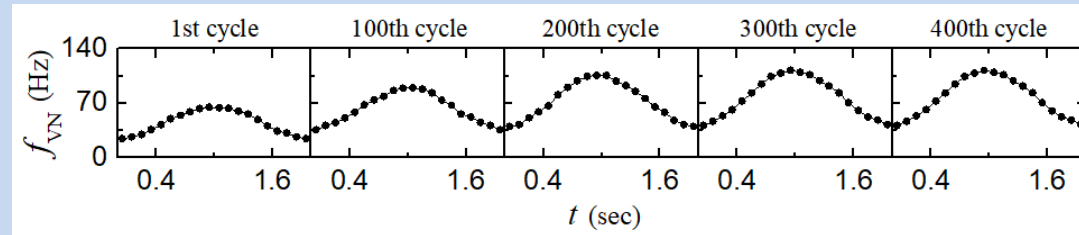
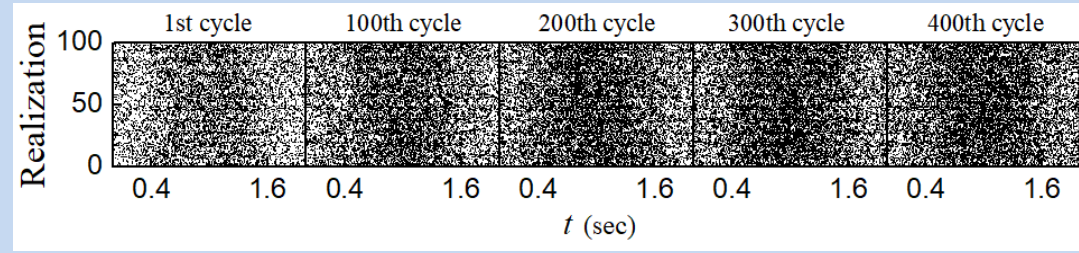
• Firing Activity of VN during Learning

- VN: Evokes the OKR eye-movement
Effective inhibitory coordination
from PCs on the VN

- Raster plots: with the cycle,
more dense at the middle stage
in contrast to the PC

- Firing Activity of VN neuron:
Bell-shaped curve with a maximum
at the middle stage.

With the cycle, the bell curve
goes up, increase in modulation,
and saturation at about the 300th cycle.

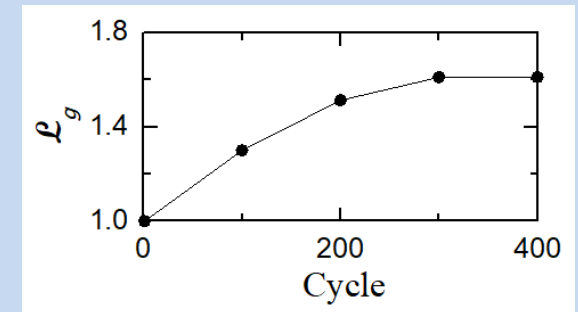


• Learning Gain Degree \mathcal{L}_g

- \mathcal{L}_g : the modulation gain ratio (i.e., normalized modulation
divided by that at the 1st cycle)

- Increase with the learning cycle and saturated at about
the 300th cycle.

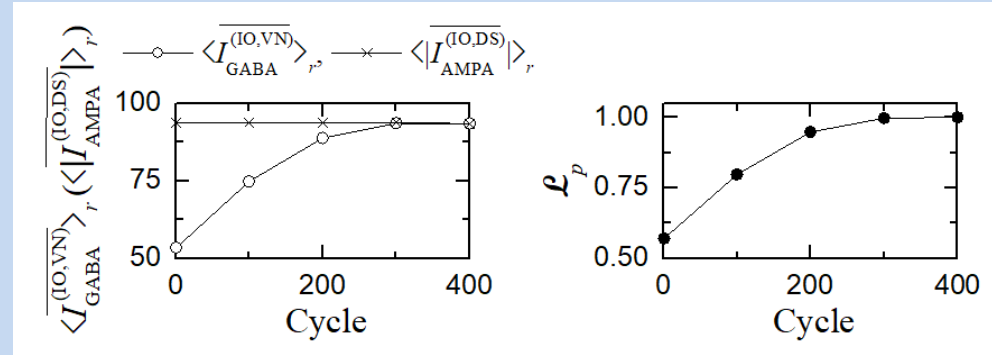
- The saturated learning gain degree \mathcal{L}_g^* ($\simeq 1.608$).



Learning Progress in The IO System

• Learning Progress

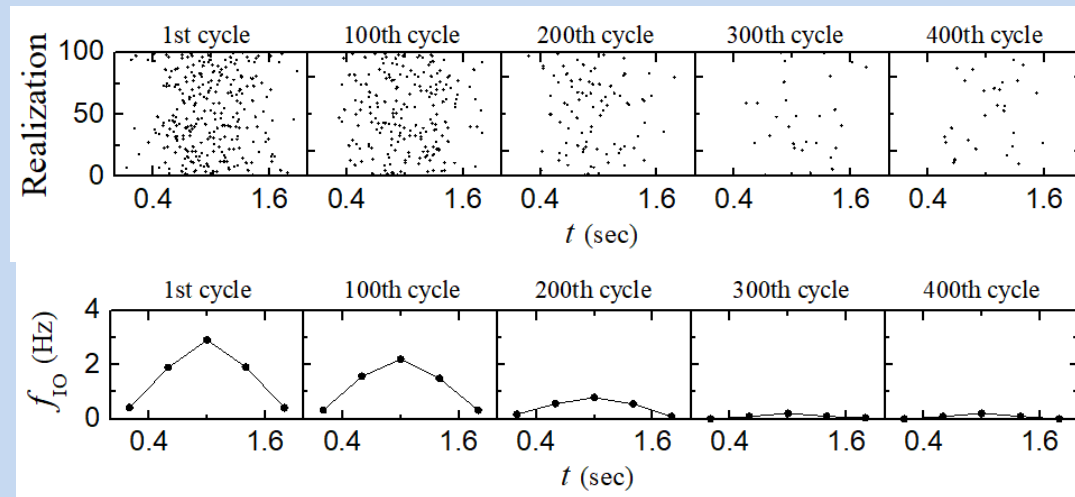
- Two inputs into IO: excitatory desired signal for a desired eye-movement and inhibitory signal from the VN neuron (denoting a realized eye-movement)
- With the cycle, increase in inhibitory input from the VN neuron, and convergence to the constant excitatory input through the IO desired signal.



- Learning progress degree $\mathcal{L}_p = \frac{\overline{I_{GABA}^{(IO,VN)}}}{\overline{I_{AMPA}^{(IO,DS)}}}$
Increase with learning cycle and saturated at 1 at about the 300th cycle.

• Firing Activity of IO neuron during Learning

- Raster plots: With the cycle, spikes at the middle stage becomes sparse due to increased inhibitory input from VN
- Firing Activity of IO neuron: Bell-shaped curve with a maximum at the middle stage. With the cycle, decrease in the amplitude, and saturated at about the 300th cycle.



Summary

- **Diverse Recoding in The GR clusters**

- Appearance of diverse in- and out-of-phase spiking patterns, due to inhibitory coordination of GO cells. Characterized in terms of conjunction index and diversity degree

- **Effect of Diverse Recoding on The OKR**

Effective depression at the PF-PC synapses

In-phase PF signals: Strong LTD by the in-phase CF signals

Out-of-phase PF signals: Weak LTD

→ Big modulation in firing of PCs & VN Neuron

- **Relation between Diverse Recoding and Learning Gain Degree**

- Diversity degree (\mathcal{D}) & Saturated learning gain degree (\mathcal{L}_g^*):

Bell-shaped curves with maximum at the same optimal value of $p_c^* = 0.06$

Strong Correlation between \mathcal{D} and \mathcal{L}_g^*

→ **The more diverse in recoding of granule cells, the more effective in motor learning for the OKR adaptation.**

