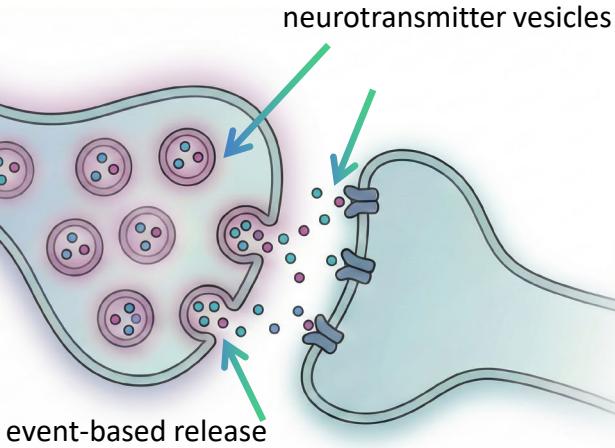
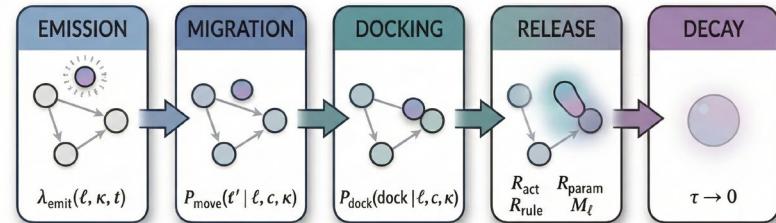


BIOLOGICAL NEUROMODULATION



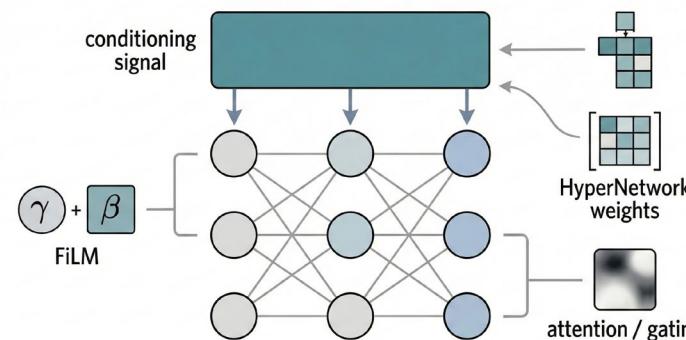
In biology, neuromodulators are carried in discrete vesicles that are emitted, diffuse, bind, and are cleared.

VESICLE LIFE CYCLE AS A STOCHASTIC PROCESS



Modulation now arises from a population process: vesicles are emitted in response to signals, migrate over G , probabilistically dock, locally modify computation, and then decay.

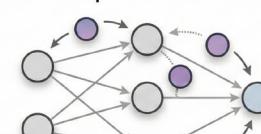
CLASSICAL 'NEUROMODULATION' IN DEEP LEARNING



Most mechanisms collapse neuromodulation into additional tensors: scaling factors, masks, or generated weights, applied everywhere in every forward pass.

COUPLED DYNAMICS AND DENSITY RELAXATION

Explicit vesicles

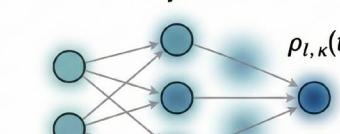


$$S_{t+1} \sim K(S_t)$$

$$S_t = (\theta_t, \{h_t^{(l)}\}, V_t, \{M_l\})$$

The same framework admits both a particle view (explicit vesicles) and a field view (density relaxation), enabling end-to-end training while preserving event-based semantics.

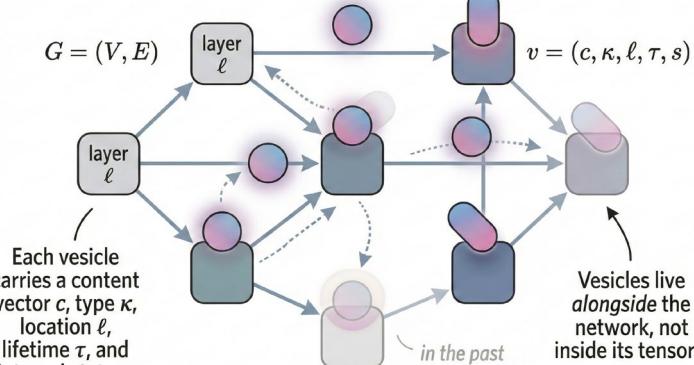
Density relaxation



$$\rho_{l,k}(t+1) = \rho_{l,k}(t) + E_{l,k}(t) - D_{l,k}(t) + M_{l,k}(t)$$

continuous
differentiable reaction-diffusion dynamics on the graph

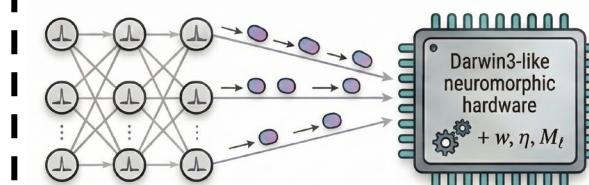
NEURO-VESICLES AS FIRST-CLASS DYNAMICAL ENTITIES



Each vesicle carries a content vector c , type κ , location ℓ , lifetime τ , and internal state s .

Vesicles live alongside the network, not inside its tensors.

SPIKING NETWORKS, NEUROMORPHIC HARDWARE, AND NESTED LEARNING



On neuromorphic hardware, vesicles correspond to routed event packets that program local learning rules and memory on chip.

NESTED LEARNING	NEURO-VESICLES
optimizes parameters at multiple time scales	extends the state with mobile entities
no explicit particles	explicit vesicle population V_t
-	orthogonal: can meta-learn emission and migration policies