


$\underline{y} = \underline{r} \psi \in$ recurrent weight strength
 $\underline{c} \in$ external input strength

$$\tau \frac{d\underline{y}}{dt} = -\underline{y} + \alpha (\underline{J}\underline{y} + \underline{g}) \cdot \psi \quad \underline{J} = \frac{\underline{W}}{\underline{\psi}}$$

$\underline{\psi} = (|W|)$
 $\underline{c} = (|g|)$

$$\psi c^{n-1}$$

$$g = \frac{h}{c}$$

recurrently



feed forward



$$\underline{r} = \frac{\underline{c}}{\underline{\psi}} \underline{y}$$

$$\frac{c}{\psi}$$

When balanced?

$$\underline{r} = f(\underline{W}\underline{c} + \underline{c}g)$$

$$f(x) = 0 \text{ if } x \leq 0$$

$$\psi \in W$$

$$\frac{c\underline{y}}{\underline{\psi}} = f\left(\frac{\psi \underline{J} c \underline{y}}{\psi} + \underline{c}g\right)$$

$$\frac{c\underline{y}}{\underline{\psi}} \in r$$

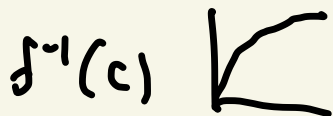
$$= f(\underline{c}(\underline{J}\underline{y} + \underline{g}))$$

$$-\underline{g} + \frac{1}{c} f^{-1}\left(\frac{c\underline{y}}{\underline{\psi}}\right) = (\underline{J}\underline{y})$$

$$\underline{y} = \underline{J}^{-1} \underline{g} + \underline{J}^{-1} \frac{1}{c} f^{-1}\left(\frac{c\underline{y}}{\underline{\psi}}\right)$$

(1) if scale $c \in \psi$ together

(2) if scale $c \propto \frac{1}{c} f^{-1}(c) \rightarrow 0$



SSN

① Explains all behaviours described previously

- Isobars symmetric
⇒ no linear for weak inputs

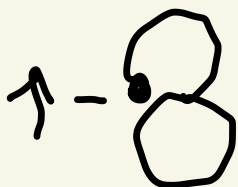
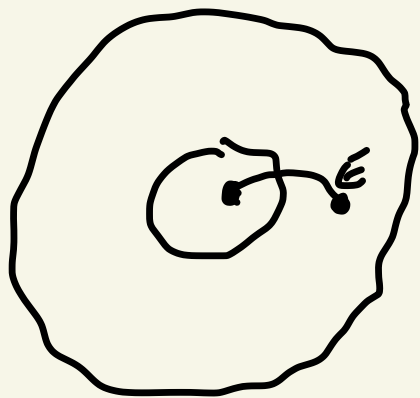
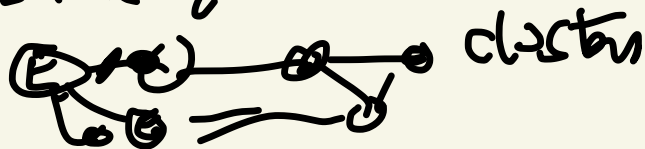
Suzuki, ... Buzaki, Histed

- Connectivity decreases w/ distance
surround suppression / facilitation

shared in variability of neural response decreases w/ stimulus

- chaotic stimulus → suppress chaos

Litwin-kumar - Ashok & Doiron



Akowitz 2017

Attractors

fixed input
 $t \rightarrow \infty$

fixed point
oscillations
chaos

multiple fixed point

Hopfield \underline{u}^i $i, 1 \dots p$ uncorrelated
(0, 0, 1, ...)

"batch"
$$\underline{W} = \sum_i \underline{u}^i \underline{u}^{iT}$$

$$\underline{W} \underline{r} = \sum_i \underbrace{\underline{u}^i \underline{u}^{iT}}_{S_{ij}} \underline{u}_j \rightarrow \underline{u}_j \quad N = \# \text{ neurons}$$

memories $\propto N$

"online" Stefano \rightarrow # memories $\propto N$
assuming $\log N$
finite # levels forget e^{-kt}
synaptic strength

Lyapunov function
"Energy" function

$$f(\xi) \\ \frac{df(\xi)}{dt} < 0$$

$$\tau \frac{dr}{dt} = -r + \underline{W} \underline{r} + \underline{h}$$

ξ has a minimum value

$$L = \frac{1}{2} \underline{r}^T \underline{W} \underline{r} + \underline{r}^T \underline{h} - \frac{\tau}{2} \underline{r}^T \underline{r}$$

$$\nabla_r L \quad \frac{dL}{dr} = \frac{1}{2} \underline{W} \underline{r} + \frac{1}{2} \underline{r}^T \underline{W} + \underline{h} - \underline{r}$$

if $W_{ij} = W_{ji}$

$$(\underline{W} \underline{r})_i = \sum_j W_{ij} r_j \\ (\underline{r}^T \underline{W})_i = \sum_j r_j W_{ji}$$

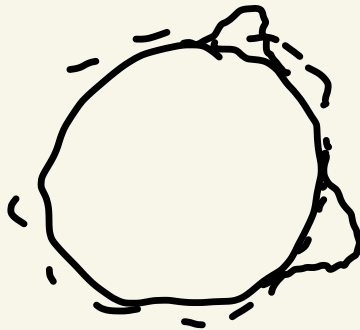
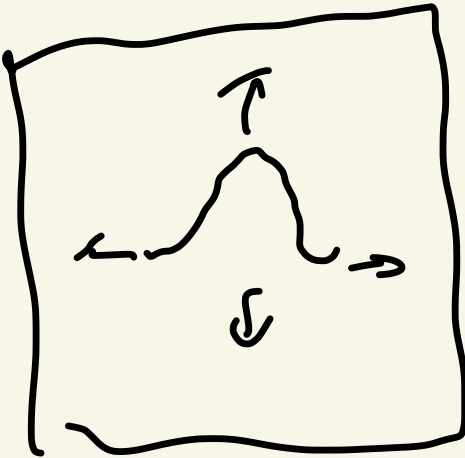
$$\frac{dc}{dr} = \frac{dr}{dt} \approx \underline{W} \underline{r} + \underline{h} - \underline{r} \\ \frac{dL}{dt} = \frac{dL}{dr} \frac{dr}{dt} = \left(\frac{dr}{dt} \right)^2 \rightarrow = \sum W_{ij} r_j$$

Attractors

"Bump" attractor
"Ring" attractor



Continuous attractor



$$\frac{dr}{dt} = -r + \underbrace{Wr}_{\lambda \omega = 1} + h$$

$$\frac{dr_i}{dt} = h_i$$

- working memory
- integration ←

path integration
place-grid

heading direction

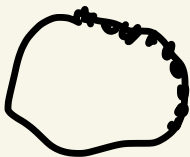
Oculomotor integrator

Problem: fragile

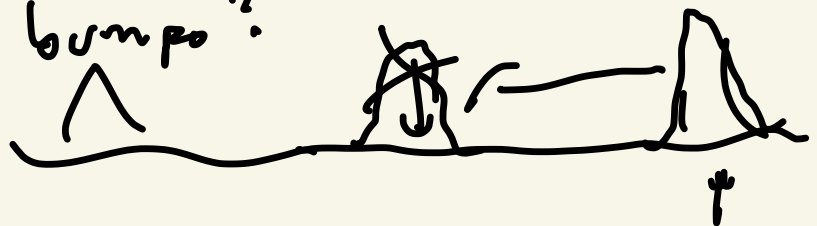
$$\lambda = 0$$

$\lambda > 0 \Rightarrow$ drift, explode

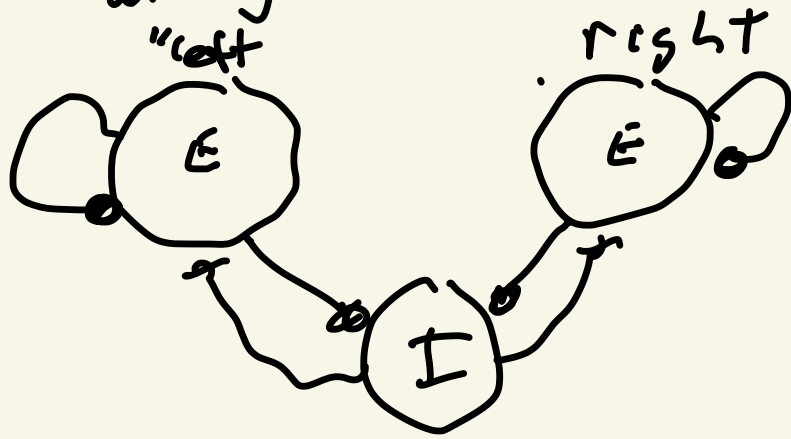
$\lambda < 0 \Rightarrow$ drift, diminish



two bumps?



~ decision making



Models of activity-dependent development
~ "learning"

Von der Malsburg '73

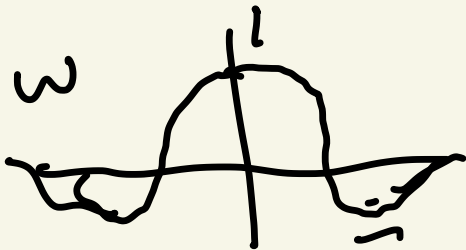
① Hebbian rule "synapses that fire together wire together"

Correlation-based: causal corr between pre & post

Competitive: if one corr pattern \uparrow others must go \downarrow

② Cortical Activity spatially clustered \Rightarrow post all selective

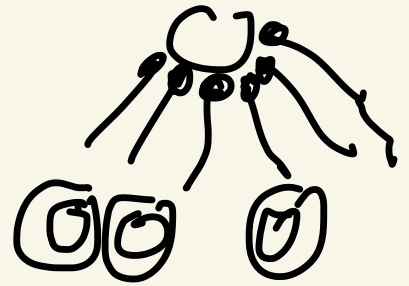
\Rightarrow nearby cells develop correlated RFS



Two approaches:

(1) "High D"

(2) "Feature-based"



Assume cells represent a feature



Fred Wolf

pattern formation

- pattern forming instability
select period

- near instability

$$\underline{z} = r e^{i2\theta}$$

θ = pref or

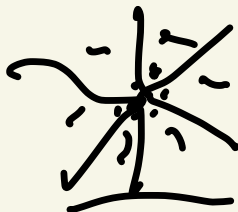
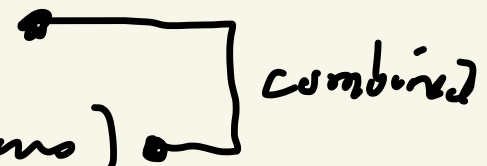
r = strength of ori selectivity

"universality classes" based on symmetries

- translation

- rotation (spatial)

- rotation (orientations)



when does solution arise w/ stable pinwheels?



Long-range suppressive interaction

↓ > 1 cycle of ORC period
x

density of pinwheels $\mu \lambda^2$
 $\approx \pi$

20,000 pinwheels $\sim 10^7$

100 maps

$\pi \pm 2\%_0$

Two primates

Carnivore