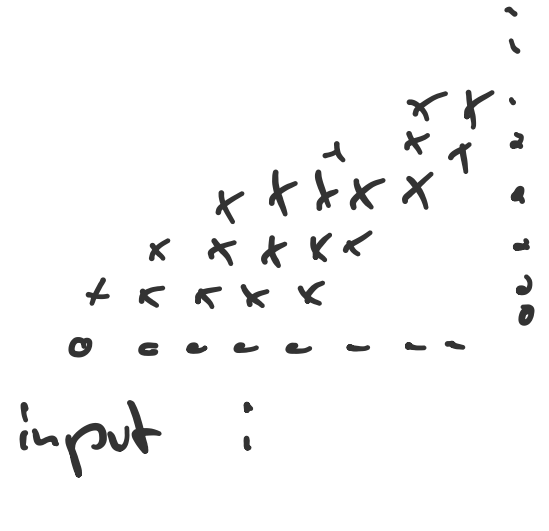


About sheet 1:

Exercise 1: Dynamic Routing



$$y_i = \sum_j u_{ij} x_j$$

$$w_{ij} = \sum_k c_k \Gamma_{ijk}$$

1. $y_i := x_{i+l} = \sum_j w_{ij} x_j$

if $u_{ij} = \delta_{(i+l)j} = \begin{cases} 1 & j = i+l \\ 0 & \text{otherwise} \end{cases}$

Then $y_i = \sum_j u_{ij} x_j =$

$$\sum_j \delta_{(i+l)j} = 0 + \dots + x_{i+l} + \dots + 0 + \dots$$

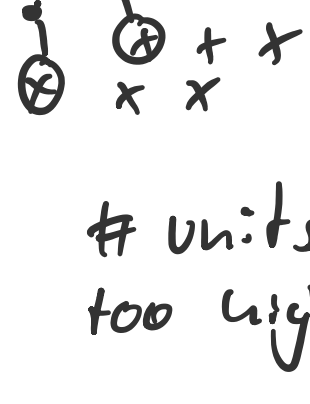
$$= x_{i+l} = y_i$$

2. scaling up/down

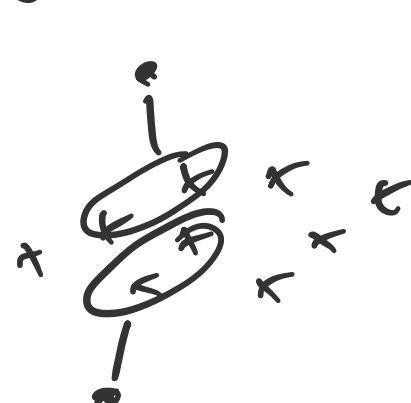


3. • 10 shifts, 10 neurons \rightarrow 100

• real world unrealistic



Way out:



\rightarrow Chlshausen et. al. 93.

4. $E = E_{\text{match}} + \beta E_{\text{cons}}$

$$E_{\text{match}} = - \sum_i p_i y_i \quad E_{\text{cons}} = \sum_{k, e \neq k} c_e c_k$$

\rightarrow double sum

$$y_i = \sum_j u_{ij} x_j = \sum_j x_j \sum_k c_k \Gamma_{ijk}$$

$$E = - \sum_i p_i \sum_j (x_j \sum_k c_k \Gamma_{ijk}) + \beta \sum_i \sum_{k \neq k} c_k c_k$$

$$\frac{\partial E}{\partial c_k} = - \sum_i \sum_j p_i x_j \Gamma_{ijk} + 2 \cdot \beta \cdot \sum_{e \neq k} c_e$$

$$\tau \frac{\partial}{\partial t} c_k + c_k = R \left(- \sum_i \sum_j p_i x_j \Gamma_{ijk} + 2 \beta \sum_{e \neq k} c_e \right)$$

• E_{match} : max match between template and output

• E_{cons}

Exercise 2: optimal visual stimuli

simple cell: $y_s = \sigma(f \cdot x)$

complex cell: $y_c = [f^{(1)} \cdot x]^2 + [f^{(2)} \cdot x]^2$

$$\|x\| = 1$$

1. $\vec{f} \cdot \vec{x} = \underbrace{\|\vec{f}\|}_{=1} \cdot \underbrace{\|\vec{x}\|}_{=1} \cdot \underbrace{\cos \varphi}_{\text{Lmax R's}}$

$$\vec{x} = \vec{f} \cdot b = \vec{f} \cdot \frac{1}{\|\vec{f}\|}$$

2. $\frac{\partial L}{\partial x} = 2 (f^{(1)} \cdot x) f^{(1)} + 2 (f^{(2)} \cdot x) f^{(2)} - 2x = 0$

$$2x = \underbrace{(f^{(1)} \cdot x) f^{(1)}}_{f^{(1)T} x} + \underbrace{(f^{(2)} \cdot x) f^{(2)}}_{f^{(2)T} x} \quad (\text{because it's a scalar})$$

$$= f^{(1)} f^{(1)T} x + f^{(2)} f^{(2)T} x =$$

$$= \underbrace{[f^{(1)} f^{(1)T} + f^{(2)} f^{(2)T}]}_A x$$

A: Projection matrix

$$2x = Ax \rightarrow \text{eigenvalue problem}$$

$\rightarrow x$ is an eigenvector of A

$\rightarrow x$ is in the plane of $f^{(1)} f^{(2)}$

$$x = \alpha f^{(1)} + \beta f^{(2)} \quad \text{with } \|x\|^2 = 1$$

$$\hookrightarrow \alpha^2 + \beta^2 = 1$$

• stimulus unique for simple cell

• not for complex cell

Outlook for the sheet 2

$$f(x) = x$$

$\mathcal{F} = \{1, x\}$ $\rightarrow \sum a x + b$
"let \mathcal{F} be the set of linear functions"
bases of all the functions with form $ax + b$

$$\Delta_i = \langle y_i^2 \rangle_+$$

Δ_i smaller $\rightarrow y_i$ slower

Average over time:

$$\langle \cdot \rangle_t = \frac{1}{b-a} \int_a^b \cdot dt \quad \text{if } t \in [a, b]$$

average: $\langle x \rangle_t = \frac{1}{b-a} \int_a^b x dt$ (1st moment)

Variance: $\text{Var}(x) = \langle x^2 \rangle_t - \langle x \rangle_t^2$

$$= \frac{1}{b-a} \int_a^b x^2 dt \quad (2^{\text{nd}} \text{ moment})$$

proof by induction

\rightarrow proof that sth. holds for $n = 1 \dots N$

• base case: $n=1$ show that "if" holds for $n=1$

• assumption: assume $n=k$ (*)

• inductive step: show that if $n=k$ holds $\Rightarrow n=k+1$ does for

example:

$$\text{show: } \sum_{i=1}^n i = \frac{n(n+1)}{2}$$

base case: $n=1 \quad \sum_{i=1}^1 i = 1 = \frac{1(1+1)}{2} \quad \checkmark$

assumption: $\sum_{i=1}^k i = \frac{k(k+1)}{2} \quad (*)$

induction: $\sum_{i=1}^{k+1} i = \sum_{i=1}^k i + (k+1) \stackrel{(*)}{=} \frac{k(k+1)}{2} + (k+1)$

$$= \frac{k(k+1) + 2(k+1)}{2} = \frac{(k+1)(k+1+1)}{2}$$

$$\square \quad \dot{x} = M \dot{z}$$

$$\dot{z} = \frac{\partial z}{\partial t}$$

$$\dot{x} \dot{x}^T = M \dot{z} \dot{z}^T M^T$$

\square trigonometric formulas

$$\cdot \cos(x) \cdot \cos(y) = \frac{1}{2} (\cos(x-y) + \cos(x+y))$$

$$\cdot \sin(x) \cdot \sin(y) = \frac{1}{2} (\cos(x-y) - \cos(x+y))$$

$$\cdot \cos^2(x) = \frac{1}{2} (1 + \cos(2x))$$

$$\cdot \sin^2(x) = \frac{1}{2} (1 - \cos(2x))$$