SYDE 556/750

Simulating Neurobiological Systems Lecture 3: Representations

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January 14 & 16, 2020





Visua'**,** ∩ortex

Mapping receptive fields

cell activity





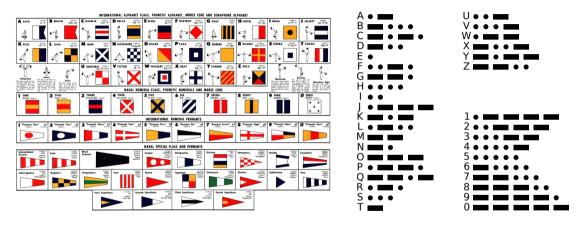


NEF Principle 1: Representation

NEF Principle 1 – Representation

Groups ("populations", or "ensembles") of neurons *represent* represent values via nonlinear encoding and linear decoding.

Lossless Codes



Encoding: $\mathbf{a} = f(\mathbf{x})$ Decoding: $\mathbf{x} = f^{-1}(\mathbf{a})$

▶ Represent a natural number between 0 and $2^n - 1$ as n binary digits.

- ▶ Represent a natural number between 0 and $2^n 1$ as n binary digits.
- ► Nonlinear encoding

$$a_i = (f(x))_i = \begin{cases} 1 & \text{if } x - 2^i \left\lfloor \frac{x}{2^i} \right\rfloor > 2^{i-1}, \\ 0 & \text{otherwise}. \end{cases}$$

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▶ Linear decoding

ecoding
$$x = f^{-1}(\mathbf{a}) = \sum_{i=0}^{n-1} 2^i a_i = \mathbf{F} \mathbf{a} = \begin{pmatrix} 1 & 2 & \dots & 2^{n-1} \end{pmatrix} \begin{pmatrix} a_0 \\ a_1 \\ \vdots \\ a_{n-1} \end{pmatrix}.$$

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► This is a **distributed code**.

- ▶ Represent a natural number between 0 and $2^n 1$ as n binary digits.
- ► Nonlinear encoding

$$\mathbf{a}_i = \left(f(\mathbf{x})\right)_i = \begin{cases} 1 & \text{if } \mathbf{x} - 2^i \left\lfloor \frac{\mathbf{x}}{2^i} \right\rfloor > 2^{i-1}, \\ 0 & \text{otherwise}. \end{cases}$$

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 .

This is a distributed code. But, not robust against additive noise!

Lossy codes

► Lossy code

Inverse f^{-1} does not exist, instead approximate the represented value

Encoding: $\mathbf{a} = f(\mathbf{x})$

Decoding: $\mathbf{x} \approx g(\mathbf{a})$

Lossy codes

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Examples

- Audio, image, and video coding schemes (MP3, JPEG, H.264)
- ightharpoonup Basis transformation onto first n principal components (PCA)

Lossy codes

► Lossy code

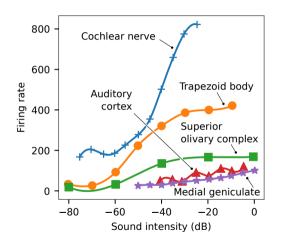
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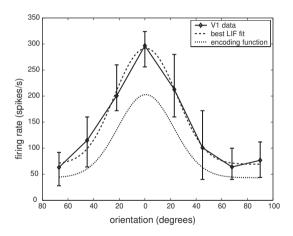
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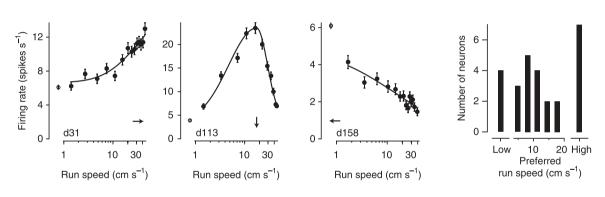
- Examples
 - Audio, image, and video coding schemes (MP3, JPEG, H.264)
 - ightharpoonup Basis transformation onto first n principal components (PCA)
 - Neural Representations

Tuning curves (I)

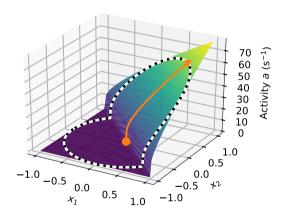




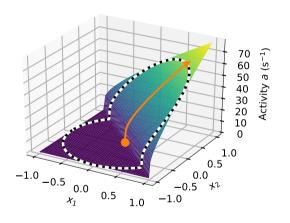
Tuning curves (II)

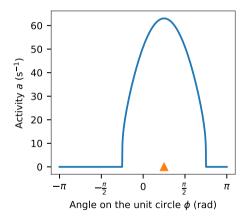


Preferred Directions in Higher Dimensions: Representing 2D Values

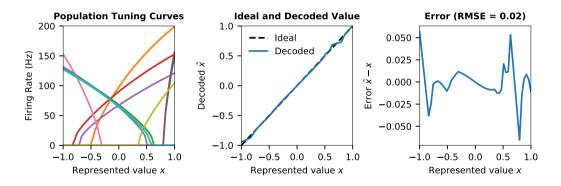


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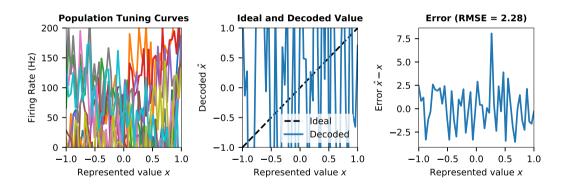




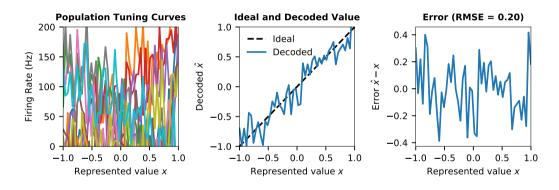
Decoding Without Taking Noise Into Account



Decoding Noisy A Without Taking Noise Into Account



Decoding Noisy A Accounting for Noise



Administration

► Assignment 1 has been released.

The due date has been adjusted to January, 30.

► Some new potential times for office hours

Mon 15:30-16:30, Mon 16:30-17:30, Tue 15:00-16:00,

Thu 11:30-12:30 (current slot), Thu 12:30-13:30

Image sources

Title slide

"The Ultimate painting." Author: Clark Richert. From Wikimedia.