Decoding: Summary of Last Lecture

- Decoding: for neuro-prostheses and/or for understanding the relationship between the brain's activity and perception or action
- Different strategies are possible: optimal decoders (e.g. ML, MAP) vs simple decoders (e.g. winner take all, population vector), depending on what we know about the encoding model, and constraints.















Fisher information: the best possible discrimination performance for a given encoder model

* Fisher information: gives the discrimination threshold that would be obtained (asymptotically) by an optimal decoder, for eg. ML (units of var ^-1)
* is expressed in terms of the encoding model P[r|s], i.e. in terms of the tuning curves and the noise

$$I_F(s) = - < \frac{\partial^2 \ln P[r|s]}{\partial s^2} >$$

* Interpreted as a measure of 'information' in the responses;

* a useful tool to relate directly the properties of the neural responses with discrimination performance.

From Population Responses to Psychophysics From Population Responses to Psychophysics Two strategies: * Assume the decoder is optimal: Compute Fisher information from P[r|s]. We Population know that this will give us the minimal possible variance of any unbiased decoder, Response and the minimal threshold of any decoder (biased or unbiased). * Construct explicitly the decoder (e.g. population vector). Compute explicitly bias, \Rightarrow (**r**) \Rightarrow $S \Longrightarrow$ Encoder variance, and threshold of estimates. Population Response \Rightarrow (r) Encoder $S \Longrightarrow$ Decoder $\Rightarrow \hat{s}$ P[r|s]Perception sensory What changes in encoder would increase discrimination performances? stimulus ▶ Number of neurons? ▶ Tuning curves shape ? • Noise correlations ?





Optimal

Decoder

 $\Rightarrow \hat{s}$

thres(\hat{s}) \simeq

 $var(\hat{s}) \simeq \frac{1}{I_F(s)}$

 $\frac{1}{\sqrt{I_F(s)}}$



What are the factors that control performance?

* Fisher information formalizes intuition and provides a tool to explore these questions precisely.

* For Poisson noise:

$$I(s) = \sum_{i} \frac{f_i'(s)^2}{f_i(s)}$$

For independent neurons, FI of the population is the sum of each neurons' FI







[Dean, Harper & McAlpine, Nature Neuro, 2005]



Research questions

* Pooling from large populations of neurons thought to be a way to average out the noise.

* Pairs of neurons show correlations in their variability: does pooling more and more neurons increases (linearly) the accuracy of the representation?

or Is information saturating over a certain number of neurons ? [Zohary et al 1994]

Research questions

* Can the study of illusions inform us on the type of 'decoder' that is used in the brain? (cf assignment 1/ visual adaptation)