How many neurons does it take to classify a lightbulb?

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(Joint work with Yuval Benjamini)

Overview

Background and motivation

- Information theory and network information theory
- Entropy, conditional entropy, and mutual information
- Studying the neural code
- Functional fMRI study of face recognition

Questions

- Can random stimuli samples be used to estimate mutual information?
- Can we obtain mutual information from the Bayes error?
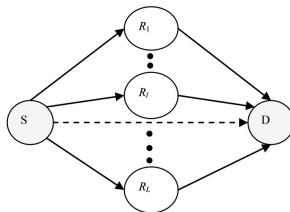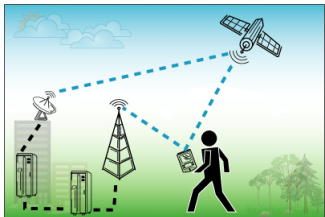- Can we obtain mutual information from observed classification rates?

Methods

- Using Fano's inequality
- Using low-SNR universality

Results

Information theory

The complexity of modern communications system is made possible by Shannon's theory of information.

A information-processing network can be analyzed in terms of interactions between its components (which are viewed as random variables.)   Image credit

CartouCHe, Aziz et al. 2011

Entropy and mutual information

Let X and Y be real-valued random variables with a joint density $p(x, y)$ with respect to a measure μ, and marginals $p_X(x)$ wrt μ_X and $p_Y(y)$ wrt μ_Y. The entropy, conditional entropy, and mutual information are nonlinear measures of spread, conditional spread, and dependence.

Quantity	Definition	Linear analogue
Entropy	$H(X) = -\int \log p(x) p(x) \mu_X(dx)$	$\text{Var}(X)$
Conditional entropy	$H(X\|Y) = \mathbf{E}[H(X\|Y)]$	$\mathbf{E}[\text{Var}(X\|Y)]$
Mutual information	$I(X; Y) = H(X) - H(X\|Y)$	$\text{Cor}(X, Y)$

The above definition includes both *differential* entropy and *discrete* entropy. Information theorists tend to use log base 2, we will use natural logs in this talk.

Properties of mutual information
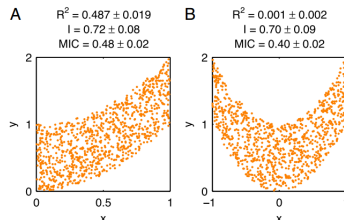
- Nonnegative: $I(X; Y) \geq 0$
- Symmetric: $I(X; Y) = I(Y; X)$
- Bijection-invariant: $I(\phi(X); \psi(Y)) = I(\psi(Y); \phi(X))$.
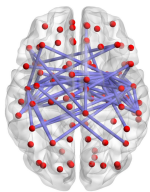- Additivity. If $(X_1, Y_1) \perp (X_2, Y_2)$, then

$$I((X_1, X_2); (Y_1, Y2)) = I(X_1; Y_1) + I(X_2; Y_2)$$

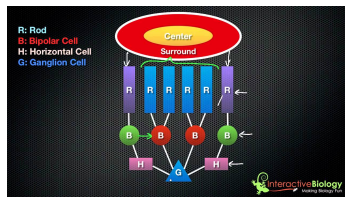- Relation to KL divergence $\mathbb{D}$.

$$\mathbb{D}(p(x, y) || p(x)p(y)) = I(X; Y)$$

Image credit Kinney et al. 2014

Studying the neural code

The brain is the *most complex* information processing system we know!

Neural network inferred from data
(Hong et al.)

Human retinal cells

How do neurons encode, process, and decode sensory information?

Image credit: Hong et al., Interactive Biology

References

- Cover and Thomas. Elements of information theory.
- Muirhead. Aspects of multivariate statistical theory.
- van der Vaart. Asymptotic statistics.