

# The Neural Code

Summer term 2022, Christian Leibold, Yuk-Hoi Yiu

Problem Set 2

—Tuning Curves—

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1. (MMSE) Consider the mean square error of an estimator

$$\text{MSE}(\hat{x}) = \langle (\hat{x}(y) - x)^2 \rangle = \sum_{i,j} p_{ij} (\hat{x}(y_i) - x_j)^2$$

in a discrete input-output space  $(y_i, x_j)$  with probabilities  $p_{ij}$ .

- (a) Compute the derivative  $\partial_{\hat{x}(y_k)} \text{MSE}$  and show that it equals

$$2 p_k (\hat{x}(y_k) - \langle x \rangle_k)$$

with  $p_k = \sum_j p_{kj} = \sum_i p_{ik}$ .

- (b) Interpret the extremal solution  $\hat{x}(y_k)$  and show it is a minimum!

2. (Fisher information of Poisson Process)

Show that the Fisher information of tuning curve with Poisson statistics is

$$\frac{f'(x)^2}{f(x)}$$

3. (MASE)

- (a) Compute the Fisher Information matrix for a Gaussian process

$$p(y|x_1, x_2) = \frac{1}{\sqrt{2\pi C}} \exp[-(y - f)^2/(2C)]$$

on a 2-d stimulus space with tuning curve  $f(x_1, x_2)$  and Covariance  $C(x_1, x_2)$ .

- (b) Apply the above result to a Gaussian processes with tuning curves on a cylinder ( $\phi \in [0, 2\pi)$ ,  $z \in [0, 1]$ )

$$f(\phi, z) = f_0 \exp[\cos(\phi - 2\pi z)]$$

and covariance matrix

$$C(\phi, z) = \sigma_0^2 + \sigma_1^2 z$$

- (c) Write a python function that computes the Fisher information matrix  $\mathcal{I}$  for given parameters  $\sigma_{0/1}$  at given  $\phi, z$ . Plot the MASE  $\text{tr}\mathcal{I}^{-1}$  for the given example as a function of  $\phi$  and  $z$  for varying choices of  $\sigma_{0/1}$ .