# Definition of Signal Density

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#### Data tensor

For a given session, let the data tensor  $X_{nbt}$  have dimensions of size [N, B, T] representing the number of neurons, number of bins, and number of trials. The value  $X_{nbt}$  is the mean DFF in neuron n when the mouse was on bin b during trial t.

## Mean response

The mean response is then

$$\mu_{nb} = \frac{1}{T} \sum_{t} X_{nbt}$$

#### Signal vector

The signal vector is then  $\Delta \mu_{nb} = \mu_{n(b+1)} - \mu_{nb}$ , keeping in mind that bins b and b+1 must actually be spatially adjacent.

#### Signal direction

The signal direction unit vector is then

$$\hat{s}_{nb} = \Delta \mu_{nb} / \sqrt{\sum_{i} (\Delta \mu_{ib})^2}$$

#### Participation rate of the signal direction

The participation rate of the signal direction unit vector is then

$$PR_b = \sum_{n} \left(\hat{s}_{nb}\right)^4$$

## Inverse participation rate of the signal direction

The inverse participation rate of the signal direction unit vector is then

$$IPR_b = 1/PR_b$$

#### Signal density

The signal density is then

$$\rho_b = IPR_b/N$$

# The expanded formula for signal density

If we expand all the terms, we can see that in terms of  $\Delta \mu_{nb}$ ,

$$\rho_b = \frac{1}{N} \left( \sum_n (\Delta \mu_{nb})^2 \right)^2 / \left( \sum_n (\Delta \mu_{nb})^4 \right)$$

## The mean signal density

We can also define the mean signal density over bins,

$$\bar{\rho} = \frac{1}{B-1} \sum_{b=1}^{B-1} \rho_b$$

This formula is easily extended to 2 directions of motion.

#### Noise density

Since I used noise density, I will mention that it is defined in the same way, just using  $\sigma_{nb}^2 = Var_t [X_{nbt}]$ .

Noise density

$$\nu_b = \frac{1}{N} \left( \sum_n \sigma_{nb}^2 \right)^2 / \left( \sum_n \sigma_{nb}^4 \right)$$

and mean noise density

$$\bar{\nu} = \frac{1}{B} \sum_{b=1}^{B} \nu_b$$