

MSTd neurons encode nonlinear combinations of retinal and extra-retinal signals

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Introduction

Neuronal activity in the dorsal Medial Superior Temporal area (MSTd) is assumed to depend on both retinal and extra-retinal input. Most of the neurons show activity when presented with a moving large-field visual stimulus (Newsome et al. 1988). This activity is modulated by extra-retinal variables as eye velocity and position (Ben Hamed et al. 2003). However, the interaction between the retinal and extraretinal input is not yet fully understood. Here we present novel insights regarding $oldsymbol{R}$ he tuning of MSTd neurons for different explanatory variables using an information-theoretic approach.

Information-theoretic approach for determining tuning functions

Neuronal tuning functions can be expressed by the conditional probability $p_{SIV}(s|v)$ of observing a spike S given some explanatory variable V. Using Bayes' theorem, this can be stated as ()

$$p_{S|V}(s|v) = \frac{p_{V,S}(v,s)}{p_V(v)}$$

where $p_{v}(v)$ is the probability of observing the explanatory variable and $p_{v}(v,s)$ the ioint probability of observing variable and spike.

 ${f R}$ or proper estimation of the tuning function it is crucial to take into account the datency between explanatory variable and spiking activity. Our approach chooses This latency in such a way, that the mutual information I(V; S) is maximized.

$$\begin{split} I(V;S) &= H(S) - H(S|V) \\ H(S|V) &= -\sum_v p_V(v) \sum_s p_{S|V}(s|v) \log p_{S|V}(s|v) \end{split}$$

Randomly moving large-field visual stimulus

We analyzed the dependence of MSTd neuronal activity in monkeys on various retinal and extra-retinal signals during Byresentation of a randomly moving large-field visual stimulus with flat velocity spectrum (white noise). The monkey's task was to fixate a target spot, which was either following the LF ostimulus or fixed at a central position. All variables refer to the Eneuron's preferred direction.



Dependence on a single explanatory variable

Across a population of 31 MSTd neurons neuronal activity depended on various retinal and extra-retinal signals during the presentation of a randomly moving largefield stimulus. The figure below shows the proportion of mutual information for each explanatory variable and neuron.



Estimated neuronal latencies of image velocity, eye velocity and eye acceleration were quite consistent across the population. For image acceleration and eve position the distribution is much wider. Dotted lines mark the average neuronal latencies of image and eve velocity.



Conclusions

· Spiking activity in MSTd neurons depends nonlinearly on various inputs. For most neurons the firing rate increases with retinal image velocity and eye velocity. However, many neurons prefer other signals, as eye and image acceleration.

 These findings support the hypothesis that MSTd uses a basis function representation for encoding nonlinear combinations of retinal and extraretinal signals as suggested by Ben Hamed et al. (2003).

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References

· Ben Hamed S, Page W, Duffy C, Pouget A. MSTd neuronal basis functions for the population encoding of heading direction

J Neurophysiol 90, 2003 Newsome WT, Wurtz RH, Komatsu H. Relation of cortical areas MT and MST to pursuit eye movements. J Neurophysiol 60, 1988

Dependence on combinations of two explanatory variables

Here we analyzed the dependence of spiking activity on a co explanatory variables. For each neuron those two variables were cho the highest mutual information in the one-dimensional analysis. Each the expected firing rate [Hz] in dependence of these variables. Estimate latencies are noted at the axis labels.



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