

# Why is attention limited? Implications from neural coding

## Mike W. Oram

School of Psychology, University of St Andrews, Fife, KY16 9JU 1994; Treue and Maunsell, 1996) but otherwise leaving the neural code unchanged

Imagine the responses to two stimuli A & B. Under passive viewing conditions the

mean spike count elicited by stimulus A could be higher than that elicited by

stimulus B. With no signal to indicate the behavioural condition, a high spike count

could be due to stimulus A being viewed under passive conditions or stimulus B

This ambiguity could be addressed by the presence of a signal from a second source

indicating whether the responses should be decoded using the mean responses

Different population codes are examined to see if the absence of a signal of

Gaussian tuning curve (SD=15, see upper panels of Figure 3 & 4). The variance of the responses

to each instance was set as twice the mean spike count. Spike counts were either simulated as

being in an unattended (gain-control=1.0, thin lines) or attended state (gain-control=1.0 - 2.5).

Assuming an equal probability of each stimulus instance,  $p(s_1)=p(s_2)...=p(s_n)$ , the probability of

each of the possible 100 instances of the stimulus was assessed for each simulated trial for each

neurone using  $p(s|r) = p(r|s)/\Sigma_{i=1} {}_{100}p(r|s_i)$  where p(s|r) is the probability of stimulus s being

present given the trial response of r spikes, p(r|s) is the probability of observing r spikes given

that stimulus s was present and  $\Sigma_{i=1}$  100p(r|s<sub>i</sub>) is the sum of the p(r|s) over all n stimulus instances

(see Oram et al., 1998 for review). The p(r|s) were obtained from the truncated Gaussian

distributions given by the mean and variance. Assuming independence, the probability of

stimulus instance s being present is given by the product of the p(s|r) from all neurones. The error

in estimating the most likely input stimulus from the responses of a simulated population of

(1) The ideal observer who "knows" the response distributions to each stimulus/attention

(2) Constant decoding assuming spike-count distributions derive from unattended stimuli

(4) Decoding assuming the spike-count distributions come from 50% attended and 50%

Constant decoding assuming the spike-count distributions derive from attended stimuli

neurones was calculated from 10000 simulated trials using four decoding strategies:

The mean spike counts of 10 simulated neurones to 100 stimuli were taken as having

(Oram et al., 1999; Wiener et al., 2001).

associated with the attended or unattended state.

viewed when attended.

Methods

condition (circles).

unattended state (diamonds).

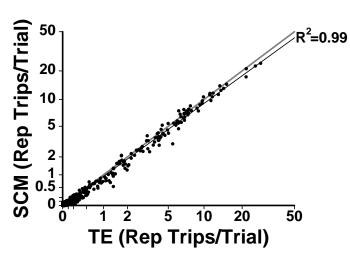
attention state "makes sense".

### Abstract

Neurophysiological recordings from visual system suggest that the evidence for a role for precisely timed spikes relative to other spike times (~1-10ms resolution) is inconclusive. This suggests that the visual system does not carry a signal that identifies whether the responses were elicited when the stimulus was attended or not. Simulations show that the absence of such a signal reduces but does not eliminate the increased discrimination between stimuli that are attended compared to when the stimuli are unattended. The increased accuracy asymptotes with increased gain-control, however, suggesting limited benefit from increasing attention. Paradoxically, the absence of a signal identifying the attentional state under which stimuli were viewed can produce the greatest discrimination between attended and unattended stimuli. Furthermore, the greatest reduction in discrimination errors occurs for a limited range of gain-control. These coding issues suggest attention maximises the discrimination of attended from unattended stimuli that is "target detection" and that as such, the effects of attention should be limited.

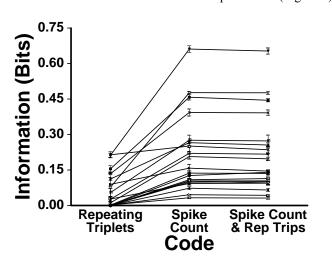
#### Introduction

• The number of precisely timed spikes of TE neuronal responses during a DMS task indicates are predictable from the changes in the coarse temporal response measures (Figure 1).



predicts the precisely timed spike patterns in task. The number of timed spike patterns (repeating triplets) found neurones during the different phases of a delayed match to sample task are predicted using the number of precisely timed spike patterns (Oram et al. 1999, 2002)

The information available from the precisely timed spike patterns is redundant with the information available from the spike count (Figure 2).



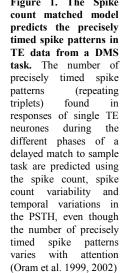
carried by precisly timed spike patterns is redundant with the information carried by spike count. The information about the stimulus-identity/DMS-phase combination was calculated from the number of repeating triplets, the spike counts and from the dual code of the spike count and the number of repeating triplets. Each line represents the results from a different neuron. The redundancy of information from repeating triplets with the information from spike count was observed for all individual cells (p > 0.05

the non-human primate suggests that attention acts as a "gain-control", scaling the responses to each stimulus by a constant (Desimone and Duncan, 1995; Motter,

Figure 1. The Spike count matched model TE data from a DMS of single TE the spike count, spike variability and temporal variations in the PSTH, even though varies with attention

Figure 2. The information 30 20 од 3.5 Ш 3.5 Decoding 1.5 0.5 0.5

The influence of behavioural significance on stimulus elicited neural responses in



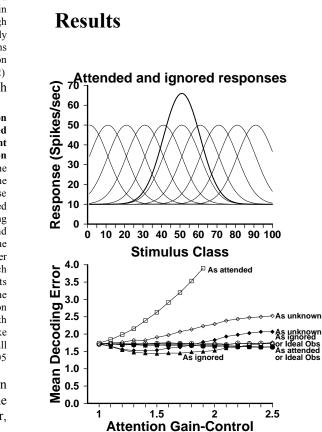
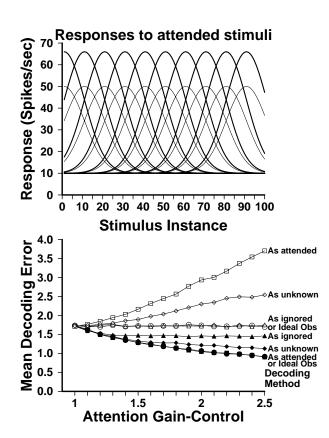
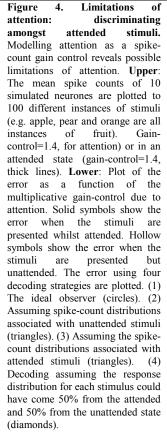


Figure 3. Limitations of attention: discriminating attended unattended stimuli. Upper: The mean spike counts of 10 simulated neurones are plotted to 100 different classes of stimuli (e.g. fruit, vegetable, furniture). The variance of the responses to each instance was set as twice the mean spike count. Spike counts were either simulated as being in an unattended state (gaincontrol=1.0, thin lines) or in an attended state (gain-control=1.4, thick lines) when attention was selective towards only one stimulus class (class 50) Lower: Plot of the error as a function of the multiplicative gain-control due to attention. Solid symbols show the error when the stimuli are presented whilst attended. Hollow symbols show the error when the stimuli are presented but unattended. The error using four decoding strategies are plotted (1) The ideal observer (circles). (2) distributions spike-count with unattended stimuli associated (triangles). (3) Assuming the spike-count distributions associated with attended stimuli (triangles). (4) Decoding assuming the response distribution for each stimulus could have come 50% from the attended and 50% from the unattended state (diamonds).





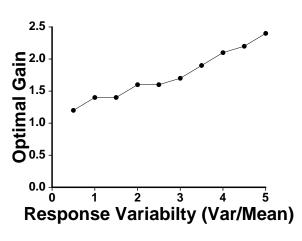


Figure 5. Optimal gain co-varies with response variability. The gain control that produces the minimum decoding error from any of the four methods is plotted as a function of the fano factor used in the simulated neural population (fano factor = variance / mean).

## **Conclusion**

- A context-specific signal indicating the state of attention is not evident in the neural codes of inferotemporal cortical neurones.
- Discriminating attended from unattended stimuli should <u>not</u> involve a contextspecific signal (Figure 3)
- Maximising the discrimination between attended stimuli involves the presence of a context-specific signal (Figure 4).
- Thus, the neural codes suggest that attention maximises the discrimination of attended from unattended stimulus classes.
- It is predicted that taking the response variability of individual neurones into account will explain some of the variability in the attention gain control observed for different neurones, both within and between different brain areas.