

# **Dynamic Routing Model for Visuospatial Attention**

# Introduction

- Existing models of visuospatial attention typically deal with determining where to direct attention [1]
- With this location selected, how does attention affect the processing of visual information through cortex?
- Problems with existing models:
  - $-\mbox{Lack}$  details of neuronal representations, transformations and dynamics
  - -Weight matrices are recomputed for each focus of attention
  - -Implausible number of pulvinar neurons (e.g. [2])



#### Empirical Observations

- Stronger and earlier attentional modulation in higher cortical areas
- · Stimuli surrounding the FOA are suppressed
- A cell's response is primarily driven by the attended stimulus in its RF
- Patients with pulvinar lesions exhibit localization deficits and more often report illusionary conjunctions



## Model Architecture

- Layers composed of neuronal columns with similar receptive fields
  Columns control neurons that signal where to sample within the
  receptive field
- Object-centred reference frame in top most layer
- At each layer, minimize the loss of information from the FOA



# **Deriving Attentional Control Signals**

- The pulvinar projects to control neurons in each layer, a signal indicating the FOA's location
- Starting at the top layer (*l*), control neurons determine the size of the FOA in layer *l*-1:  $\sum_{\substack{size_{l-1} \\ size_{l}}} \max_{i} \frac{maxRF_i}{i} \quad if \quad size_l > maxRF_i \\ otherwise}$

Control neurons determine spatial sampling frequency to resize the FOA for layer *l*:  $sf = \frac{(size_{i,i}) - 1}{(size_{i-1}) - 1}$ 

- The *i*<sup>th</sup> control neuron determines the location  $(\mu_{l,l})$  within its RF from which to sample visual information:  $\mu_{l,l} = pos_{l-1} + (l pos_l) \times sf$
- Feedforward signals  $x_{j,l-l}$  and control signals  $\mu_{l,l}$  project to intermediate neurons which compute the function:  $f(\mu_{l,l}x_{l,l-1}) = \exp^{-(\mu_{l,l}-x_{l,l-1})^2/2\sigma^2}$



# Waterloo

### **Neural Implementation**

- Implemented using the Neural Engineering Framework [3]
- 7 input and 3 output columns 150 LIF neurons per population
- Intermediate neurons compute a non-linear combination of control and feedforward visual signals



- Input held constant while control signal (μ) is varied across output column's RF
- Control signal indicates where to sample within column's RF
- When control signal is outside of neuron's RF, default routing is used (i.e. entire visual field is resampled at each layer)

Effect of varying control signal on neural response given a constant input



### Results

- Detailed spiking LIF implementation of attentional routing
- Requires a plausible number of pulvinar neurons
- Static synaptic weights and low dimensional control signal
- Consistent with timing of attentional modulation of neural activity
- Scales well (tested up to 40,000 neurons)
- Accounts for empirical observations

# Predictions

Case 1 – Linear dendrites

- Intermediate cortical neurons that are responsive to both cortical afferents and indirect pulvinar signals
- Non-linear dendrites are not required, however intermediate neurons are required
- Case 2 Non-linear dendrites
- Far fewer neurons would be needed
- In either case, cortical neurons in lamina 4 receiving direct pulvinar projections need not be sensitive to visual stimuli

Shipp, S – Trends Cogn Sci, 8(5), 223-230 (2004)
 Olshausen et al.– J Comp Neurosci, 2, 45-62 (1995)
 Eliasmith and Anderson – Neural Engineering – MIT press (2003)