## Coverage, Polymaps and the Visual Cortex

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Retinotopy





Orientation



Spatial Frequency



**Direction Preference** 



#### Monkey Ocular Dominance Stripes



### The Retinotopic Map



2-DG activity pattern in visual cortex

from Tootell et al., (1988)



#### Monkey Orientation and Ocular Dominance Columns



1 mm

From Blasdel (1992)

## Combined Ocular Dominance, Spatial Frequency and Orientation Maps from Cat Area 17



Data from Hübener, Shoham, Bonhoeffer and Grinvald



Upper layer (light grey): ocular dominance map Middle layer (coloured lines): iso-orientation boundaries Bottom layer (dark grey): low spatial frequency domains

Adapted from Hübener et al. (1997) J. Neurosci., 17, 9270

## **TERMINOLOGY**

**Protomap:** the spatial representation across the cortex of one of the sub-features, e.g. ocular dominance

**Polymap:** the complete map of all the feature domains

#### Ocular dominance protomap



#### Spatial frequency protomap



Orientation protomap



#### Polymap representation





#### Stimulus space representation of cortical maps



The cortex in cortical coordinates



The cortex, projected into retinal space

## A dimension-reducing map



The cortex folded inside a stimulus space. The manner of folding is believed to be constrained by continuity and completeness (coverage) constraints.

#### The Traveling Salesman's Problem



## How the Kohonen Algorithm Works



The cortical neighbourhood function

The Kohonen Self-Organizing Feature Map Algorithm

$$\Delta w_j = \varepsilon h(j, j^*)(v - w_j)$$

- ε is a rate constant;
- $W_i$  is the position of cortical point *j* in stimulus space
- v is a stimulus vector;

 $j^*$  is the cortical point most responsive (i.e. nearest) to stimulus v;  $h(j, j^*)$  is a Gaussian function of the cortical distance between points j and  $j^*$ . Summary of how the Kohonen SOFM works:

For each stimulus *v*, find the cortical point *j*\* which is closest (most responsive) and move it and its neighbours towards the stimulus. Repeat this for many randomly chosen stimuli.

The width of the cortical neighbourhood function may be reduced as the map forms, a process referred to as 'annealing'.

The end result is to make the cortex fill the stimulus space as continuously as possible, satisfying completeness and continuity constraints.



## Kohonen SOFM Algorithm

orientation + ocular dominance + retinotopy

#### Data



Blasdel (1992)

#### Model



Obermayer, Ritter & Schulten (1992)

#### Model maps

Kohonen SOFM





#### Real maps

Cat 1



Elastic net



#### Calculation of Coverage Uniformity

For a particular stimulus, **v**, calculate the total activity *A* evoked over the whole map,

$$A(\mathbf{v}) = \sum_{i,j\in C} f(\mathbf{v} - \mathbf{w}_{i,j})$$

where  $w_{i,j}$  is the receptive field center of cortical point (i, j) and f() specifies the receptive field shape (typically Gaussian).

Define coverage uniformity as the ratio between the variability in *A* divided by the mean, taken over a representative set of stimuli, i.e.

 $c' = \operatorname{standard} \operatorname{deviation}(A)/\operatorname{mean}(A)$ 

 $c\,{\rm \hat{}}$  is a measure of noise in the representation of a stimulus space by a cortical area.

#### Steps in the calculation of coverage



The coverage hypothesis:

If the maps of orientation, ocular dominance and spatial frequency are optimised for uniform coverage, perturbing the spatial relations between the different maps should always lead to worse coverage. Coverage test applied to model map



Coverage uniformity (c')

#### Test 1: sliding perturbations



#### Test 1: rotations and flips



% increase in c'

Comparison of coverage between model and real maps				
	$\lambda$ od	$\lambda$ sf	λori	Coverage (c')
Real Maps				
Cat 1	28	44	27	0.412
Cat 2	37	30	26	0.446
Cat 3	69	25	26	0.537
Cat 4	35	25	27	0.432
Model Maps				
EL8.003	34	36	27	0.345
EL9.001	32	33	30	0.304
KO6.001	31	27	27	0.307
KO7.001	33	26	30	0.285

# Reasons why measured c' values might be worse in real maps:

- development does a poor job of optimising
- experimental errors in determining the locations of domain boundaries may artefactually worsen coverage
- the target stimulus distribution may not be uniform, as was assumed
- the model maps are not a structurally realistic basis for an exact comparison

# How many maps are there in visual cortex?

Given *N* binary features, each of which is represented in a periodic, stripe-like map, how many maps can be overlaid so that all 2<sup>*N*</sup> combinations get represented reasonably often (i.e. with good coverage) ? One limit on the number of maps is the number of cortical columns available to represent  $2^N$  features within the region of cortex available to each retinotopic location (the cortical point image). This region is about 1 - 2 mm in diameter.

The smallest functional columnar unit in the cortex is likely to be a mini-column, about  $30 - 50 \mu m$  in diameter. This leads to an upper limit of about 10 maps, assuming the geometrical problem can be solved. Can it be solved?

#### Simple Ways of Combining Maps to Optimise Coverage

Hubel and Wiesel's ice-cube model

The Egg-carton Model



iso-orientation domains

A solution for two maps



#### A solution for five maps



#### Protomap morphology is little affected by *N*:



#### Polymap for 8 binary features



#### **Binary Feature Maps**







from Tootell et al., Science, 1983





















Simulation showing modulus of feature vector







Simulation showing colour/luminance domains in 'thin stripes' and disparity domains in 'thick' stripes





