

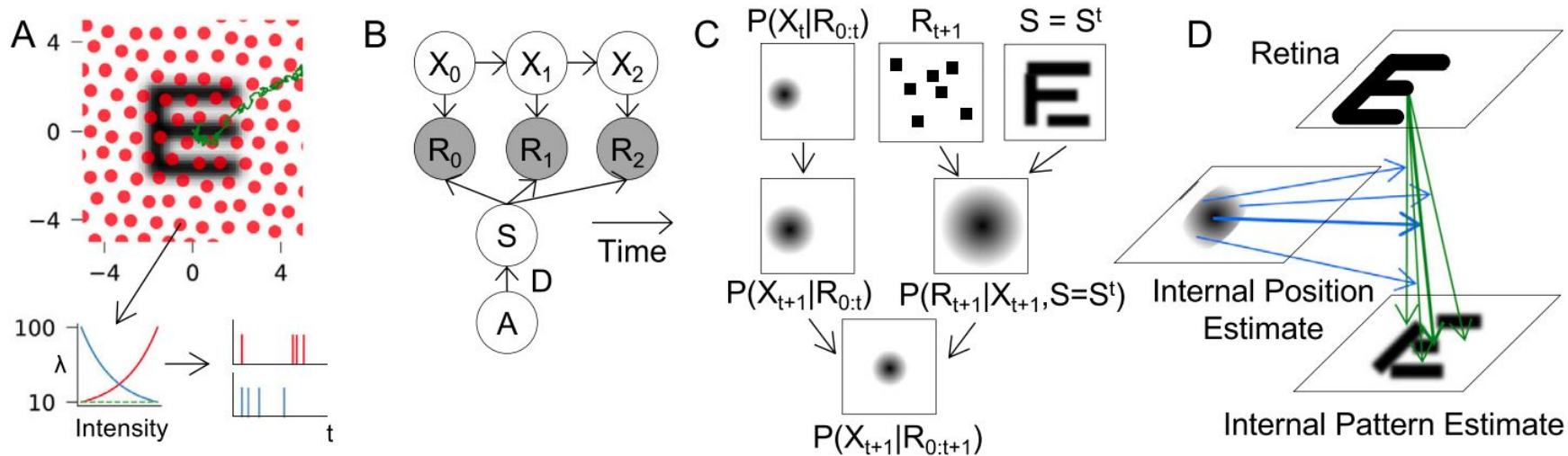
# The Effect of Lateral Inhibition on Color Perception

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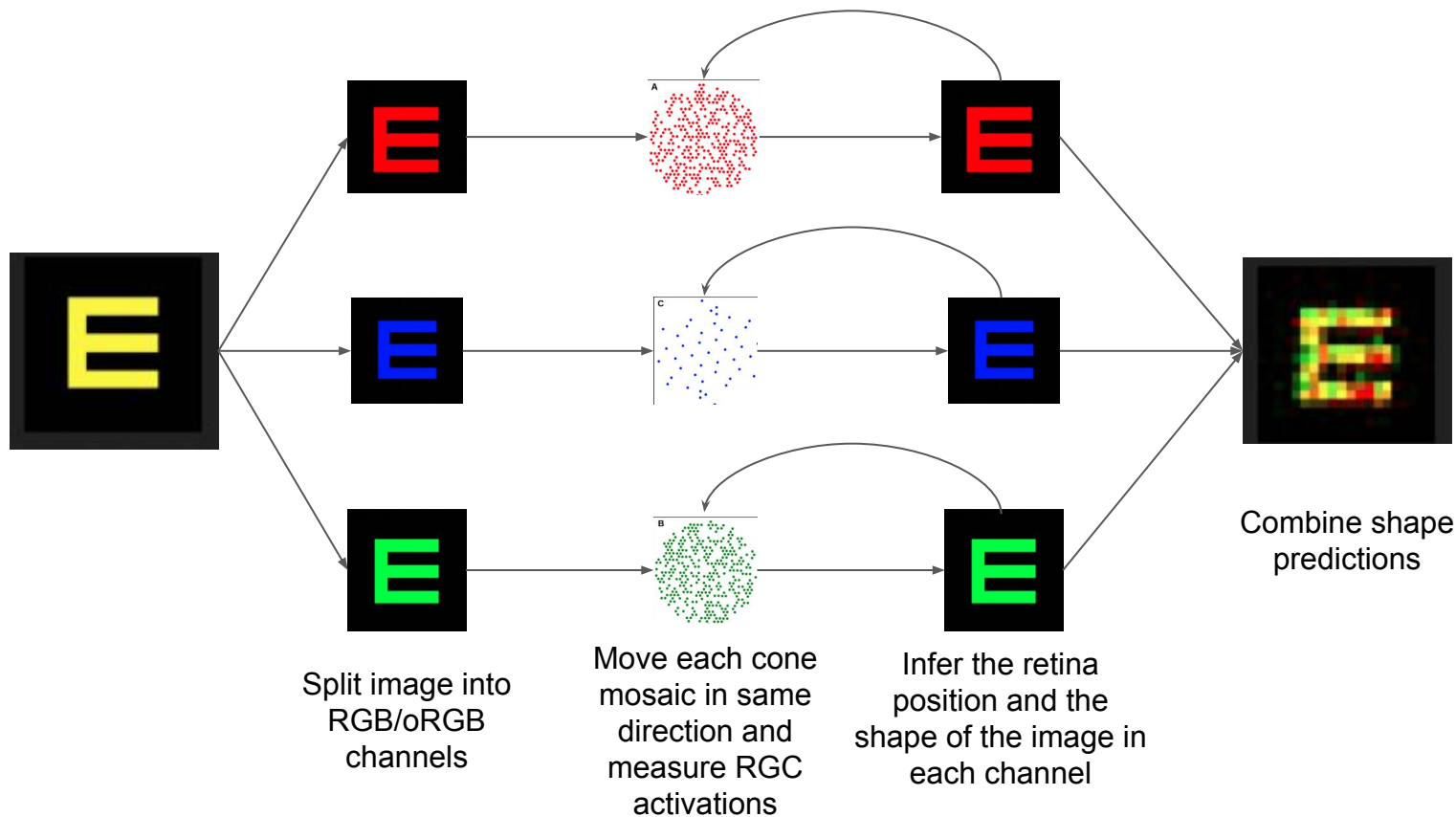
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# Recap of Anderson Model



# Recap of Last Year's Extension to Anderson Model



# Transforming Input from RGB to LMS space

- R, G, and B values do not correspond to activations of L, M, and S cones
- We want to represent the RGB image in LMS space before running the model
- To start, we can convert RGB  $\rightarrow$  XYZ  $\rightarrow$  LMS color spaces
  - Simulates the response of the cones in the eye
  - Corresponds to actual cones better
- Previous model did color opponency model with RGB  $\rightarrow$  oRGB, but that assumes that each cone corresponds directly to each opponent spaces.

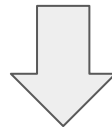
$$\begin{bmatrix} L \\ M \\ S \end{bmatrix} = \begin{bmatrix} 0.8562 & 0.3372 & -0.1934 \\ -0.8360 & 1.8327 & 0.0033 \\ 0.0357 & -0.0469 & 1.0112 \end{bmatrix} \begin{bmatrix} X \\ Y \\ Z \end{bmatrix}$$

$$\begin{bmatrix} R \\ G \\ B \end{bmatrix} = \begin{bmatrix} 0.41847 & -0.15866 & -0.082835 \\ -0.091169 & 0.25243 & 0.015708 \\ 0.00092090 & -0.0025498 & 0.17860 \end{bmatrix} \cdot \begin{bmatrix} X \\ Y \\ Z \end{bmatrix}$$

# Accounting for Lateral Inhibition

Anderson model:

$$c_{j,t} = g \cdot \sum_i S_i T(X_t^R)_{i,j}$$



Our model:

$$c_{j,t} = g \cdot \max \left( \sum_i S_i T(X_t^R)_{i,j} - \sum_{k \in \mathcal{N}} \sum_i S_i I(j, k), 0 \right)$$

$$I(j, k) = \frac{1}{\sigma \sqrt{2\pi}} e^{-\frac{1}{2} \left( \frac{\|X_j - X_k\|}{\sigma} \right)^2}$$

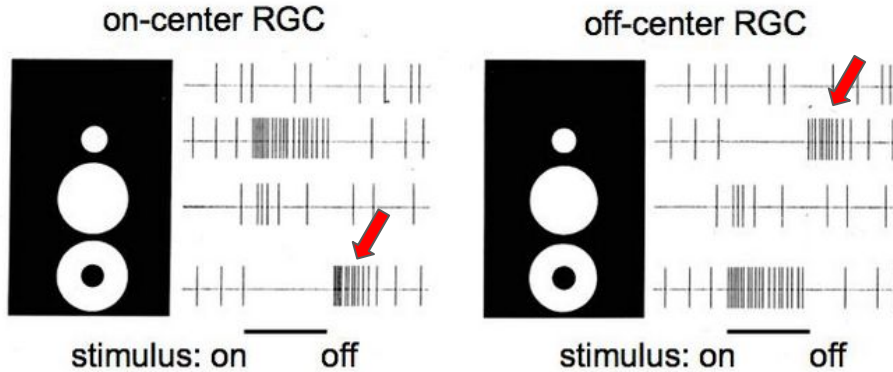
Key differences: max,  $\mathcal{N}$  (set of neighboring cones),  $I$  (distance function)

Similarities:  $g$  (gain), activation of original cone

Final value of  $c_{j,t}$  is transformed and used to calculate final frequency of firing

# Accounting for Lateral Inhibition (con't)

Issue:



Need to include a temporal aspect to activation and firing frequency

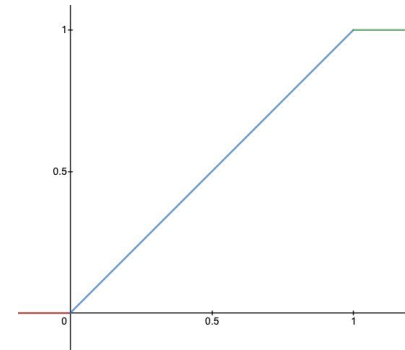
Modifications:

$$c_{j,t} = g \cdot \left( \sum_i S_i T(X_t^R)_{i,j} - \sum_{k \in \mathcal{N}} \sum_i S_i I(j,k) \right)$$

$$c'_{j,t} = c_{j,t} \text{ if } j \in \text{ON or } 1 - c_{j,t} \text{ if } j \in \text{OFF}$$

$$c''_{j,t} = c'_{j,t} + \sum_{q=0}^{t'} \max(c'_{j,t} - c'_{j,t-q}, 0)$$

$c''_{j,t}$



$c_f$

# Jointly Inferring Shape and Color

- Anderson model infers shape,  $S$ , according to  $S = DA$  where  $D$  is a dictionary of basis functions and  $A$  is the inferred latent variable
- We can modify  $D$  to represent a spatio-chromatic basis to construct colored shapes
  - For reconstruction purposes doesn't make sense to infer color and shape independently
- **Next Step:** Looking for biologically plausible dictionaries that can represent colored images and substituting such a dictionary into the model.

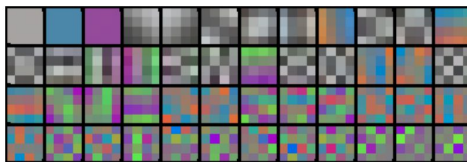
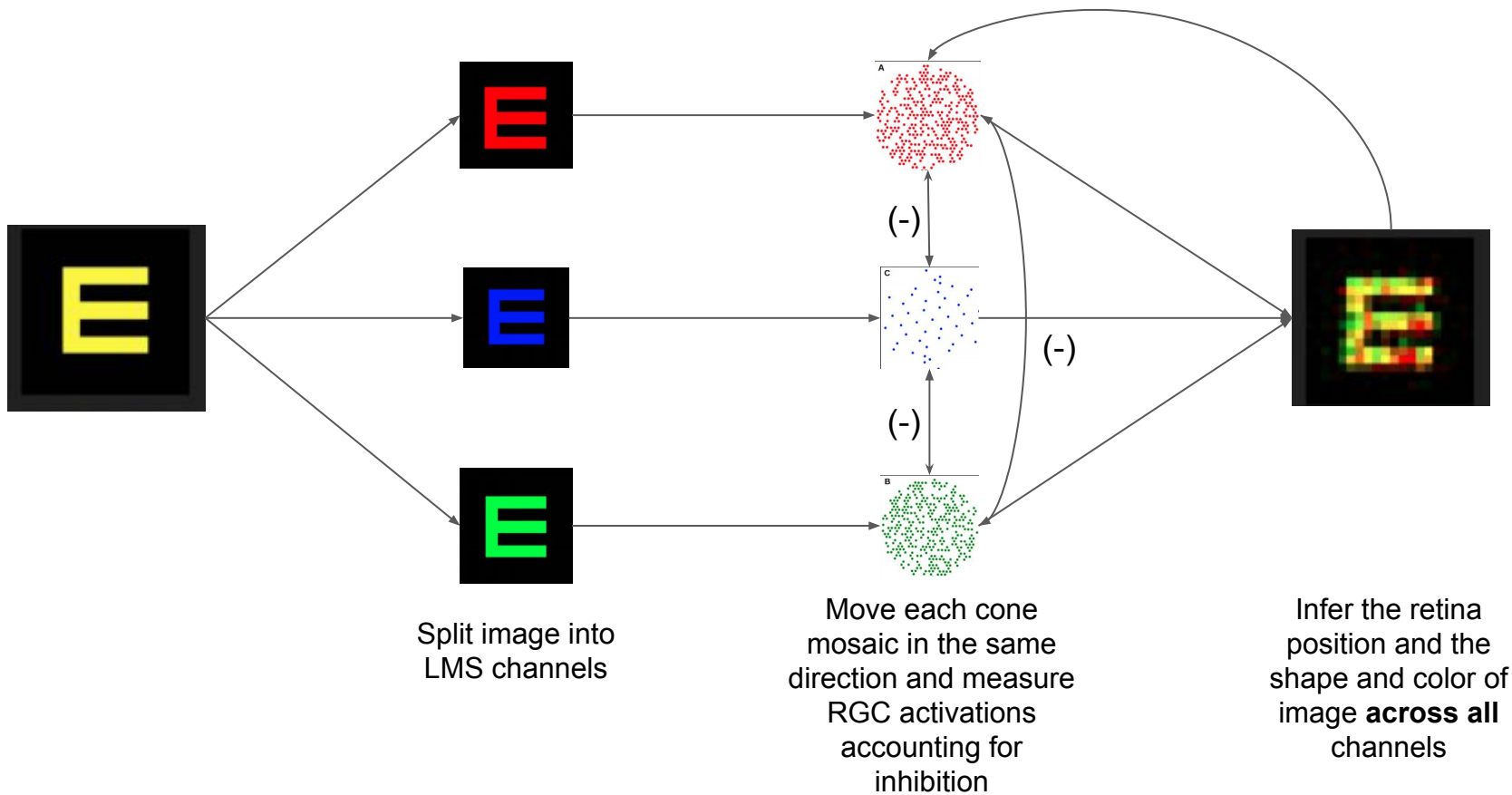


Figure 2: Spatio-chromatic basis obtained from PCA on  $4 \times 4 \times RGB$  image patches of the example in Fig. 1.





# Next Steps

- Modifying the RGC model equations in the Anderson model to incorporate lateral inhibition
- Locating and modifying the relevant code in Anderson's code in order to incorporate lateral inhibition
- Modifying the sparse dictionary to include color and shape so that they are not inferred separately