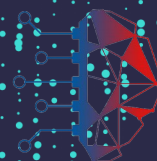


Predictive cost functions in the neocortex

Blake Richards



Institut-Hôpital
néurologique de Montréal
Montreal Neurological
Institute-Hospital



HIBALL

HELMHOLTZ International BigBrain
Analytics & Learning Laboratory

June 26th, 2020
HIBALL Launch Workshop

A potential general principle of intelligence:

In order to successfully predict future inputs an agent must have a model of the world that captures important latent variables, variables that will then be useful for guiding actions.

If so, learning to predict stimuli is a means of learning the structure of the world that is relevant for decision making.

Thus, agents can learn good representations in an unsupervised (or self-supervised) manner by predicting future stimuli.

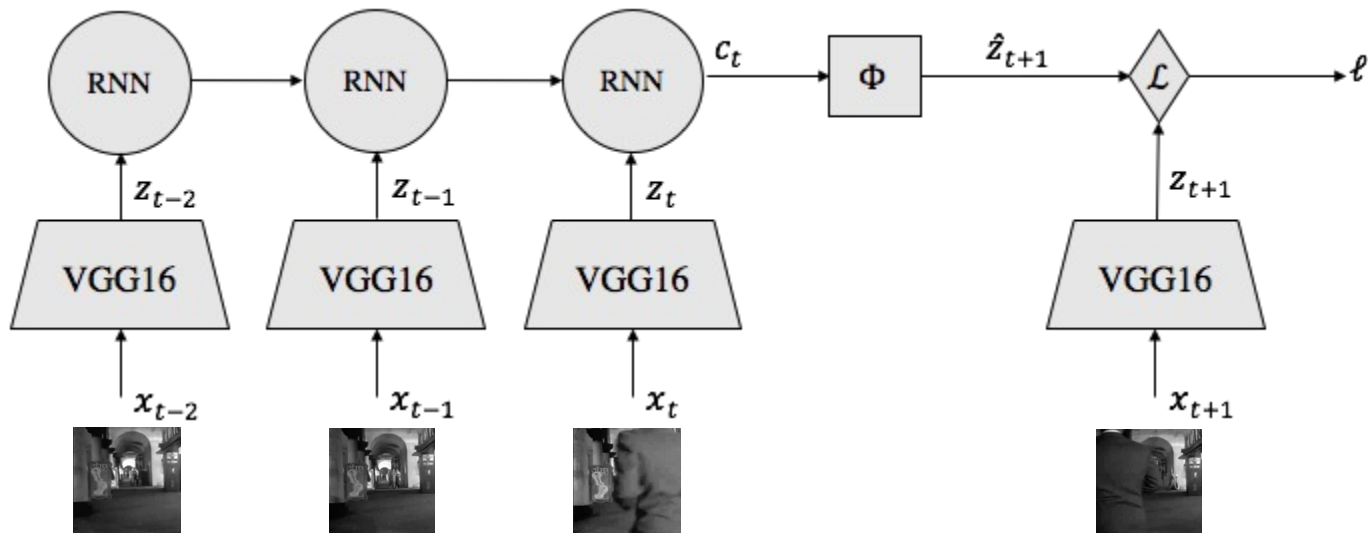


We wanted to investigate whether different regions of the mouse visual cortex may be using an unsupervised predictive cost function.

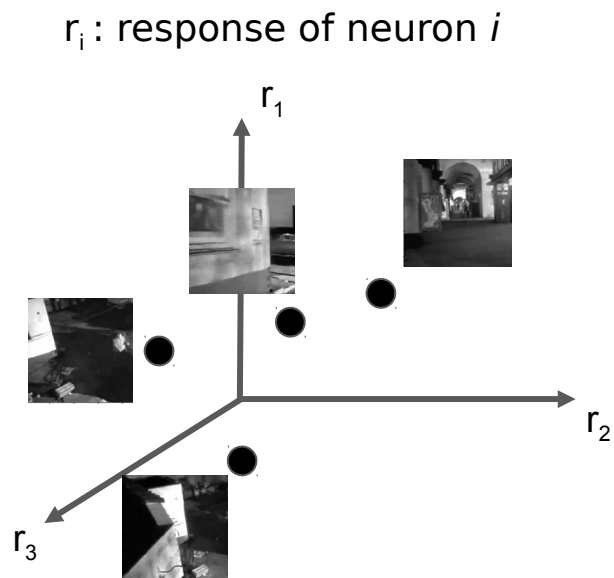
To explore this we used open 2-photon calcium data from the Allen Brain institute and ran representational similarity analyses to artificial neural networks trained in a self-supervised manner.

Brain mapping with the **contrastive predictive coding cost function**

$$\mathcal{L} = - \sum_{i,k} \left[\log \frac{\exp(\hat{z}_{i,k}^\top \cdot z_{i,k})}{\sum_{j,m} \exp(\hat{z}_{i,k}^\top \cdot z_{j,m})} \right]$$



Representation Similarity Analysis

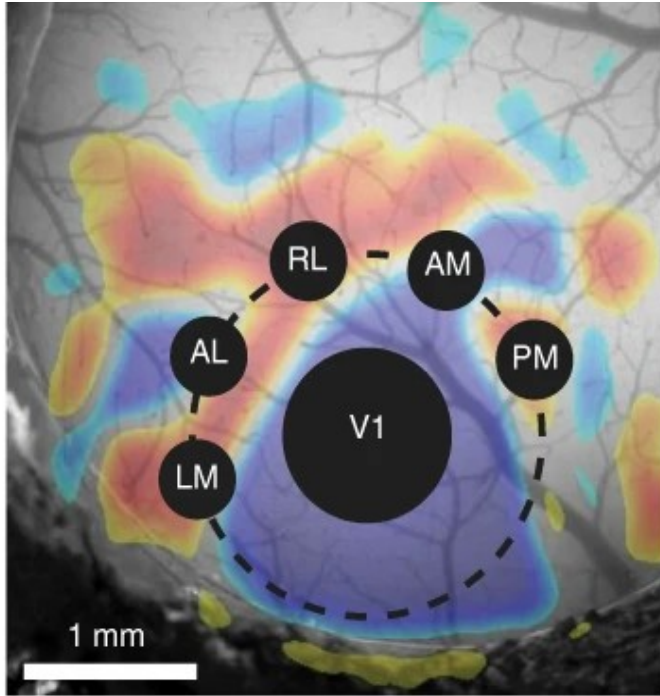


$$R^{V1} = \begin{bmatrix} \\ \\ \\ \end{bmatrix} \begin{matrix} \downarrow \\ \# \\ \text{frames} \end{matrix}$$

$\xrightarrow{\# \text{ neurons}}$

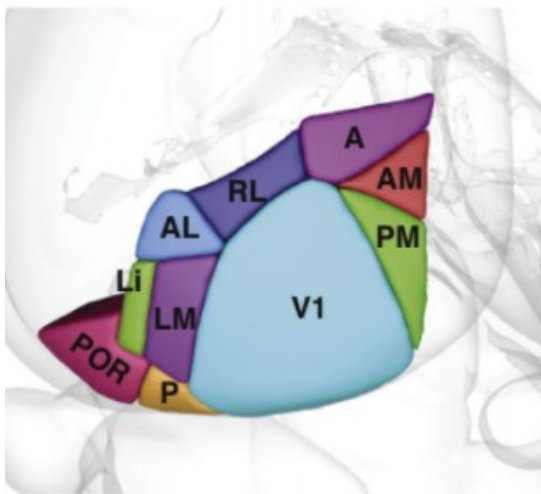
e.g. RR^T

$$RSM^{V1} = \begin{bmatrix} \\ \\ \\ \end{bmatrix}$$

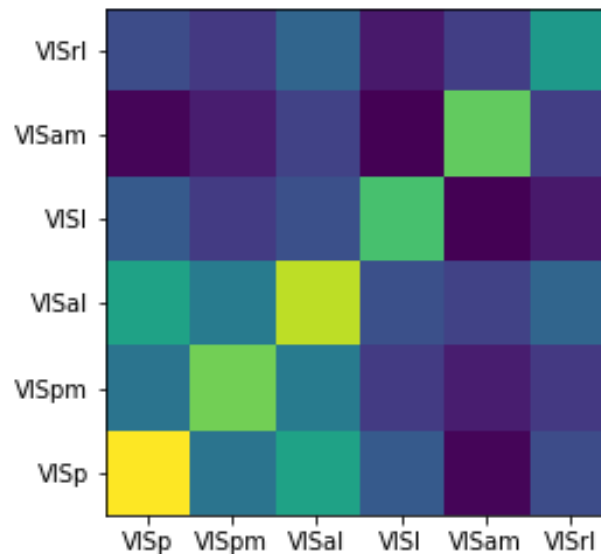


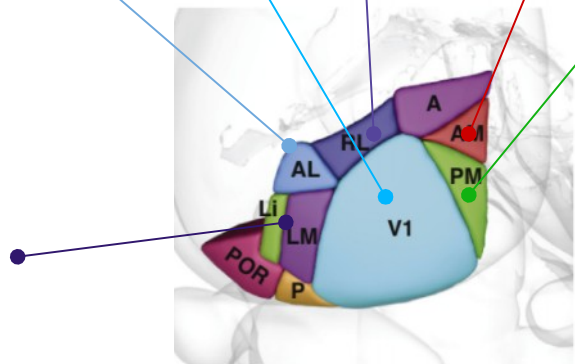
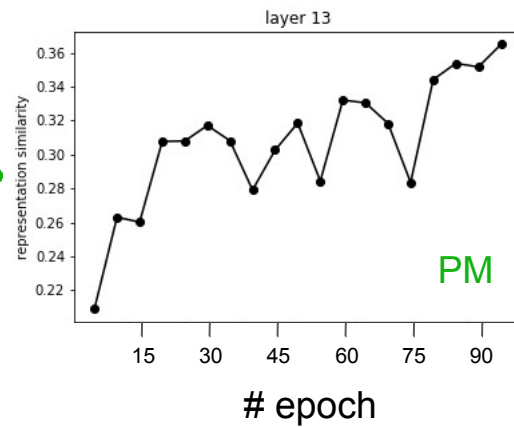
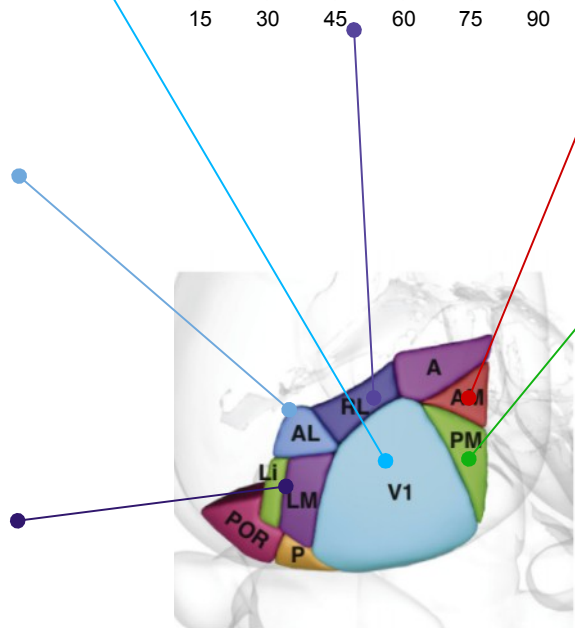
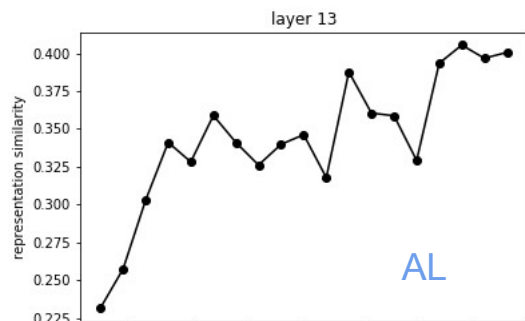
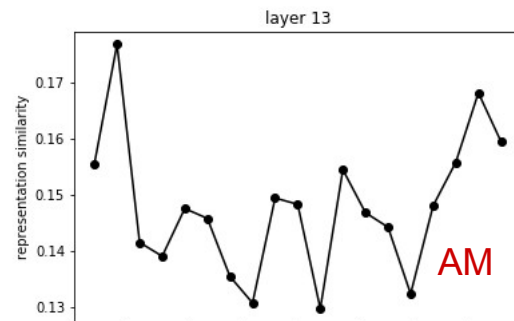
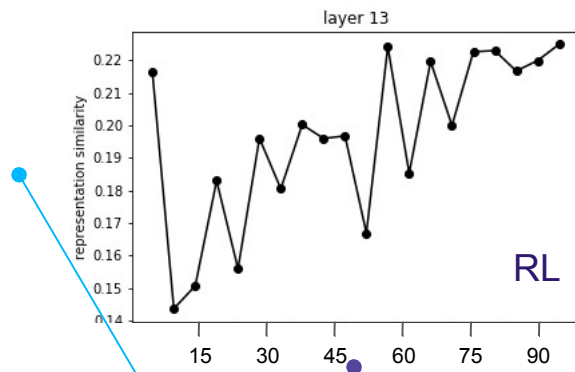
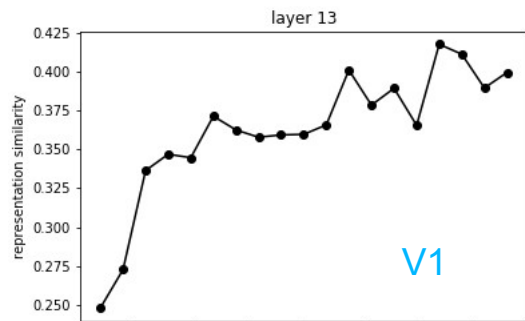
De Vries et al. (2020), *Nature Neuroscience*, 23: 138

Sanity check: brain regions map to themselves with RSM



RSM similarity

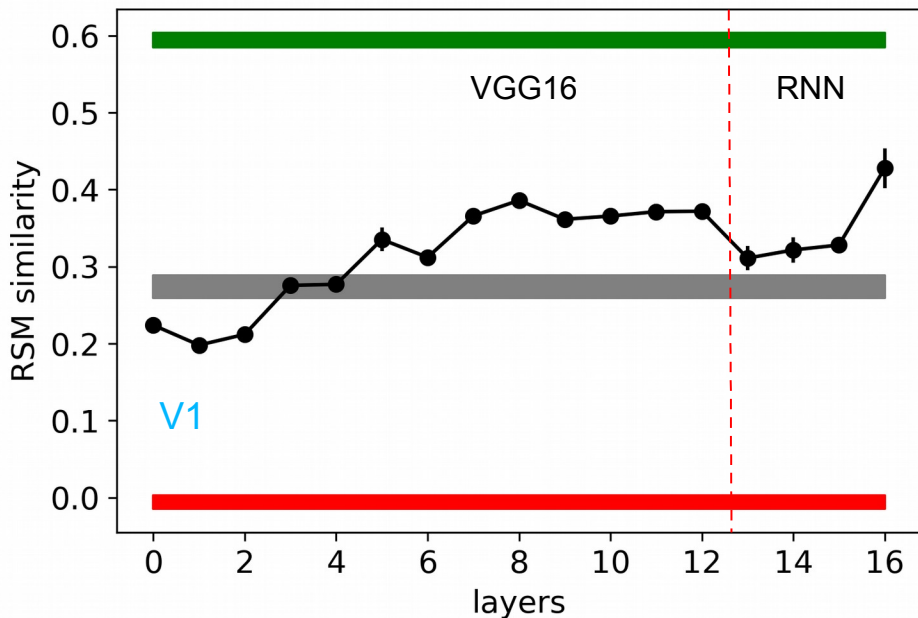


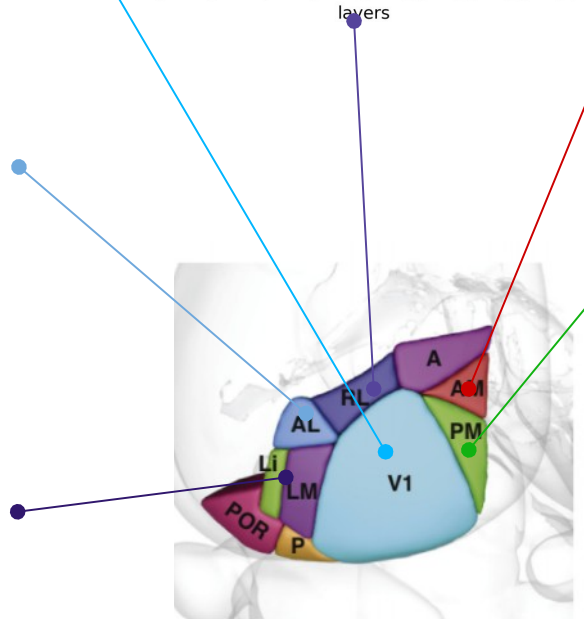
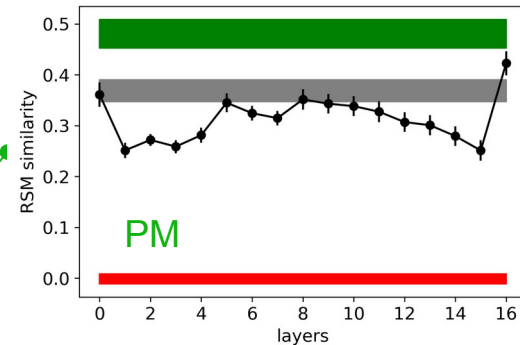
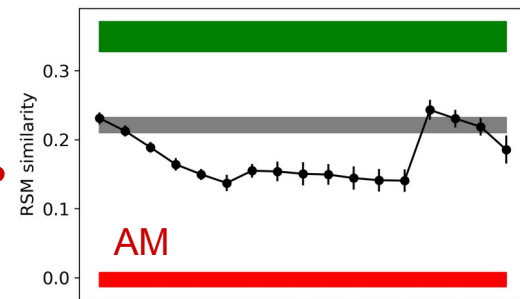
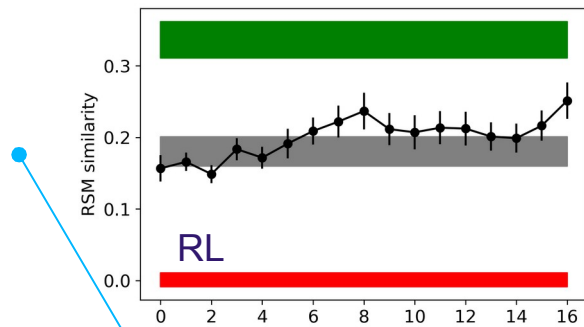
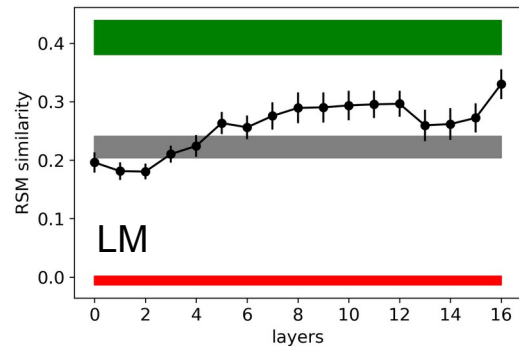
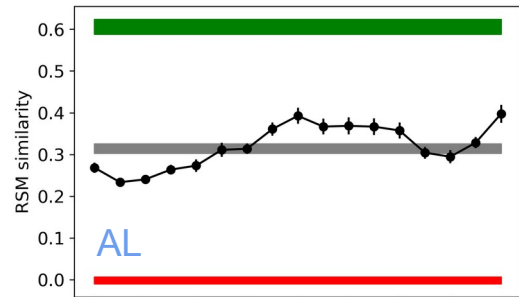
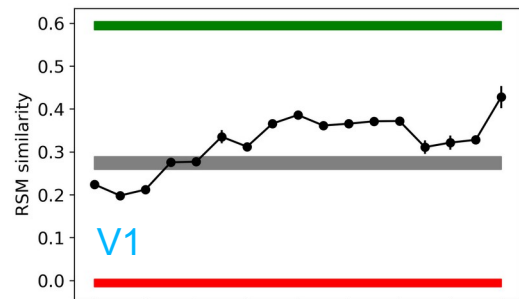


When is RSM similarity high?

- Noise ceiling
- Pixel RSM
- Shuffled V1 RSM

$$\text{RSM similarity} = \langle \text{RSM}_{\text{cpc}}, \text{RSM}^{\text{V1}} \rangle$$





@ Epoch 100

Two recent papers have similar results in macaque and human!

Unsupervised Neural Network Models of the Ventral Visual Stream

Chengxu Zhuang, Siming Yan, Aran Nayebi, Martin Schrimpf, Michael C. Frank, James J. DiCarlo, Daniel L. K. Yamins

doi: <https://doi.org/10.1101/2020.06.16.155556>

Instance-level contrastive learning yields human brain-like representation without category-supervision

 Talia Konkle,  George A. Alvarez

doi: <https://doi.org/10.1101/2020.06.15.153247>

Next step: use HIBALL data to do this in ANNs with human-like architectures



Can we get better RSA matches to the human brain?

Can we get networks with good inductive biases for downstream RL?

Thanks to Shahab Bakhtiari!



Funding:



HEALTHY BRAINS
FOR **HEALTHY LIVES**



NSERC
CRSNG

CIFAR
—