# What are deep neural networks and what are they good for?

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# Outline

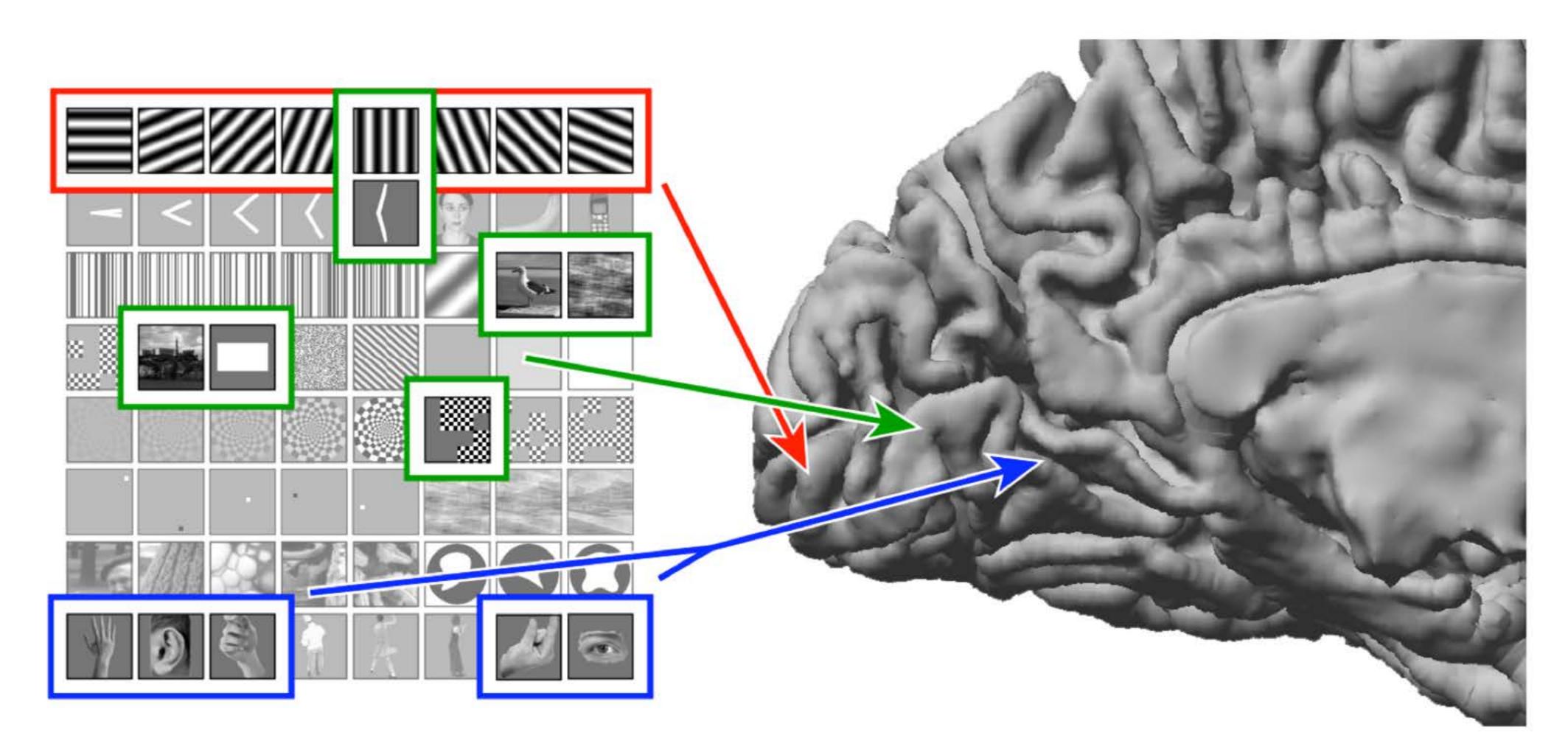
### What are deep neural networks (DNNs)?

### How should we evaluate DNNs?



# **Context for DNNs**

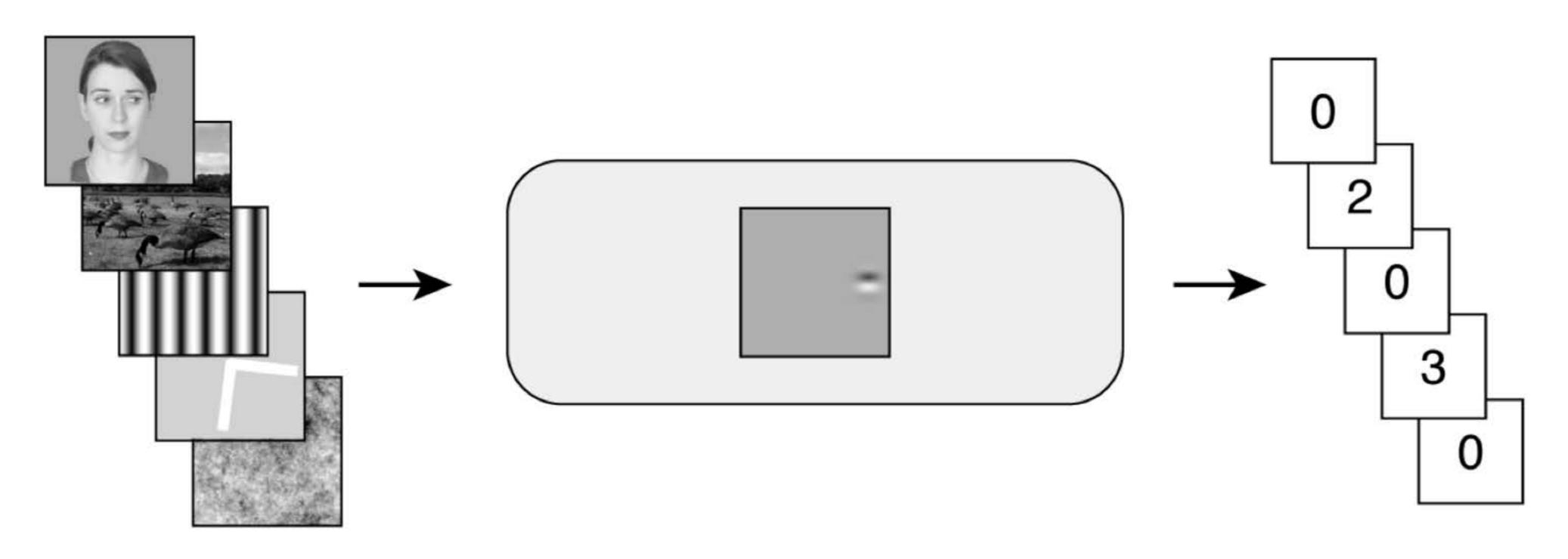
Observation: visual areas show stimulus selectivity



# **Context for DNNs**

# **Observation:** visual areas show stimulus selectivity Goal: develop image processing models

(receptive-field model, forward model, encoding model, representational model)



### Stimuli

Model

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### Responses

# **Context for DNNs**

- Observation: visual areas show stimulus selectivity
- Goal: develop image processing models (receptive-field model, forward model, encoding model, representational model)
- **Details:** 
  - individual units or similarity matrices
  - experimental design, cross-validation, noise analysis

### **Brief overview of DNNs**

### History:

- Neural networks
- Improvements in training procedures

### Why exciting?

- Powerful (very good performance on computer vision tasks)
- Possibly a good model of the brain?

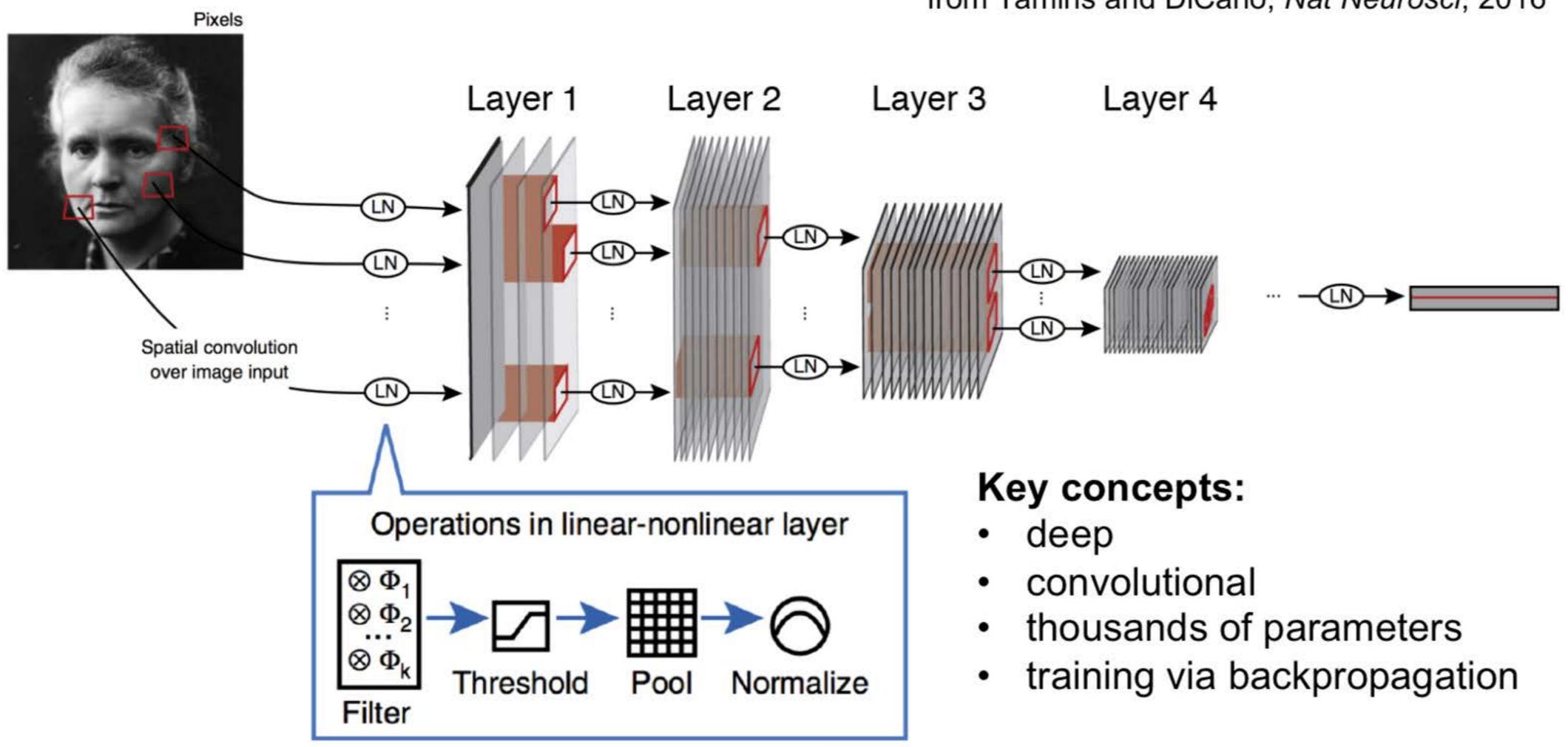
Yamins et al., PNAS, 2014 Khaligh-Razavi and Kriegeskorte, PLoS Comp Bio, 2014 Cadieu et al., PLoS Comp Bio, 2014 Agrawal et al., arXiv, 2014 Güçlü and van Gerven, J Neurosci, 2015 Kubilius et al., PLoS Comp Bio, 2016 Cichy et al., arXiv, 2016

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More information:

 Yamins and DiCarlo, Nat Neurosci, 2016 Kriegeskorte, Annual Reviews, 2015

### Brief overview of DNNs

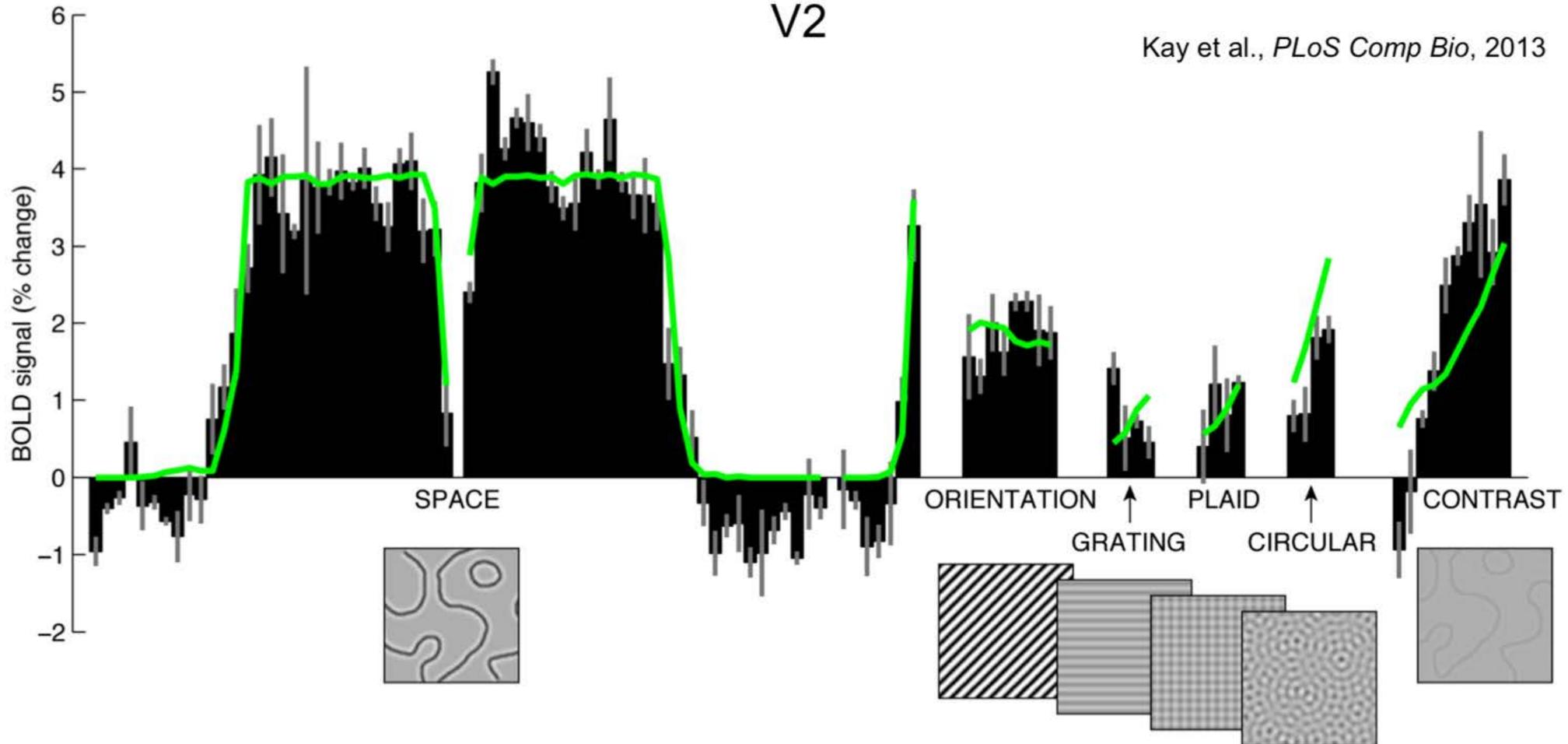


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from Yamins and DiCarlo, Nat Neurosci, 2016

### What makes a good model?

### Accuracy • Cross-validated predictions of experimental data



### What makes a good model?

### • Accuracy

- Cross-validated predictions of experimental data
- Caution: NOT circuits. We are just observing responses.

### Inhibition, Spike Threshold, and Stimulus Selectivity in Primary Visual Cortex

Nicholas J. Priebe<sup>1</sup> and David Ferster<sup>2,\*</sup>

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### experimental data just observing responses.

Neuron 57, February 28, 2008

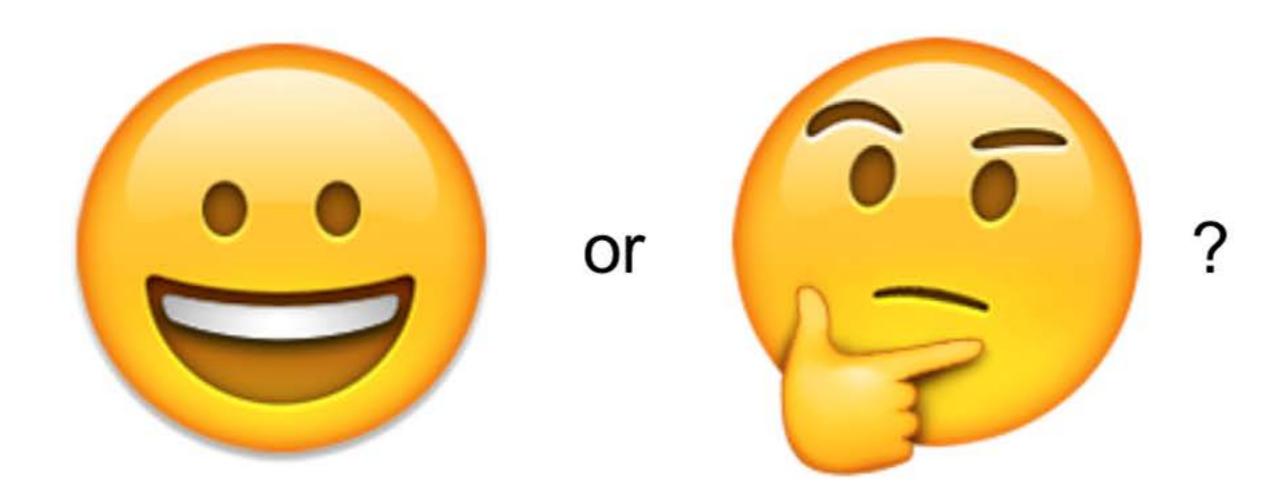
# What makes a good model?

### Accuracy

- Cross-validated predictions of experimental data
- Caution: NOT circuits. We are just observing responses.

### Understanding

– "I have a model that perfectly simulates your data."

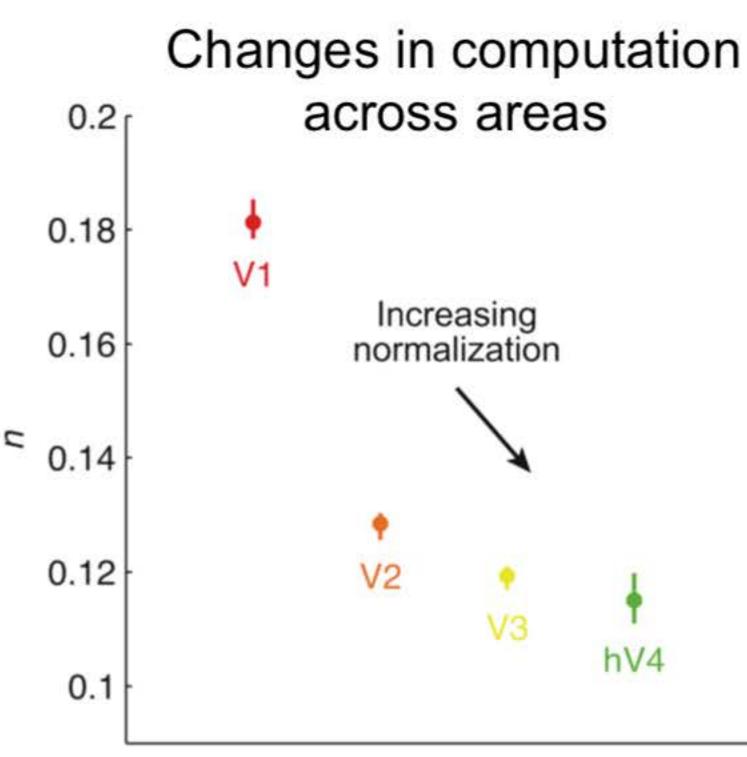


# What does it mean to understand?

- TUNING Do you know how the model behaves?
- PARAMETERS Do you know what happens if you perturb the parameters?
- ARCHITECTURE Have you done model surgery to identify important parts?
  - At a deeper level: Can you predict responses in your head? Can you implement the model from scratch?

# Why do we want to understand?

- Understanding enables simplification
  - Smaller, more efficient models
  - More insightful comparison across visual areas, cognitive states, individuals, groups



Kay et al., PLoS Comp Bio, 2013

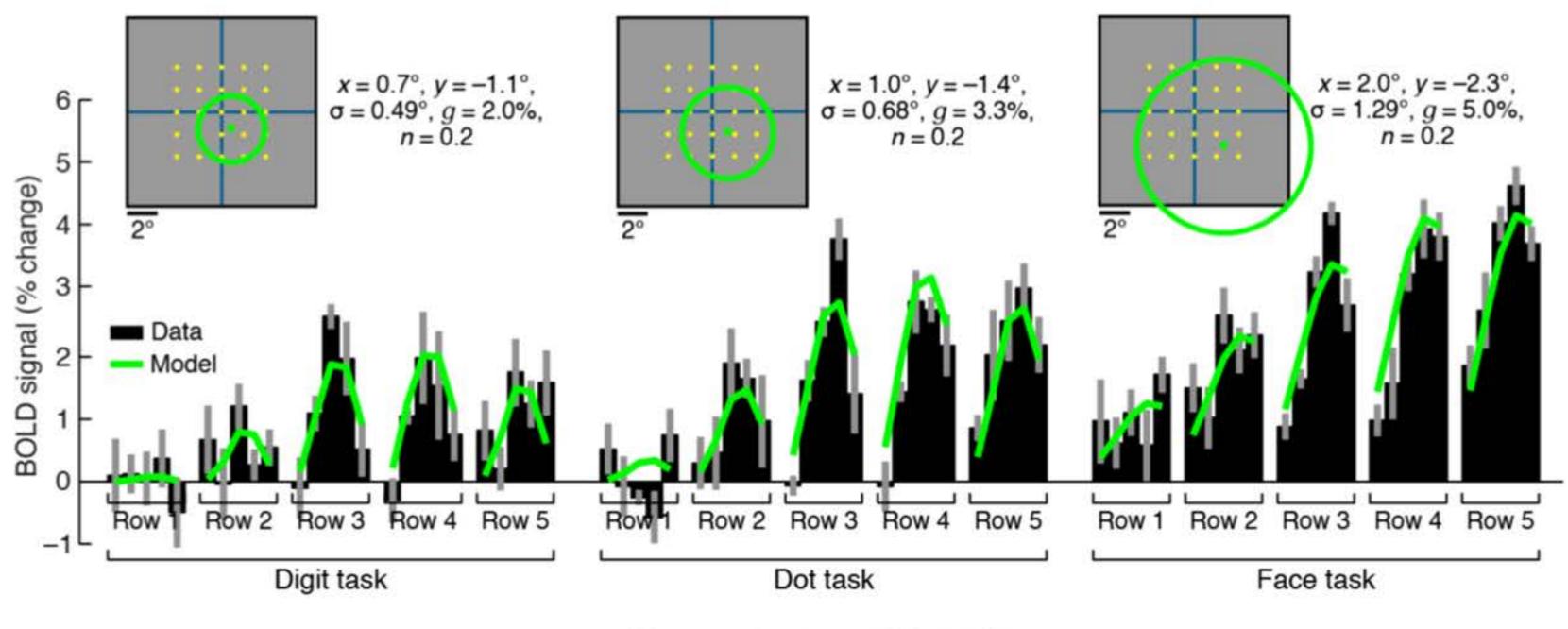
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Changes in computation across cognitive states



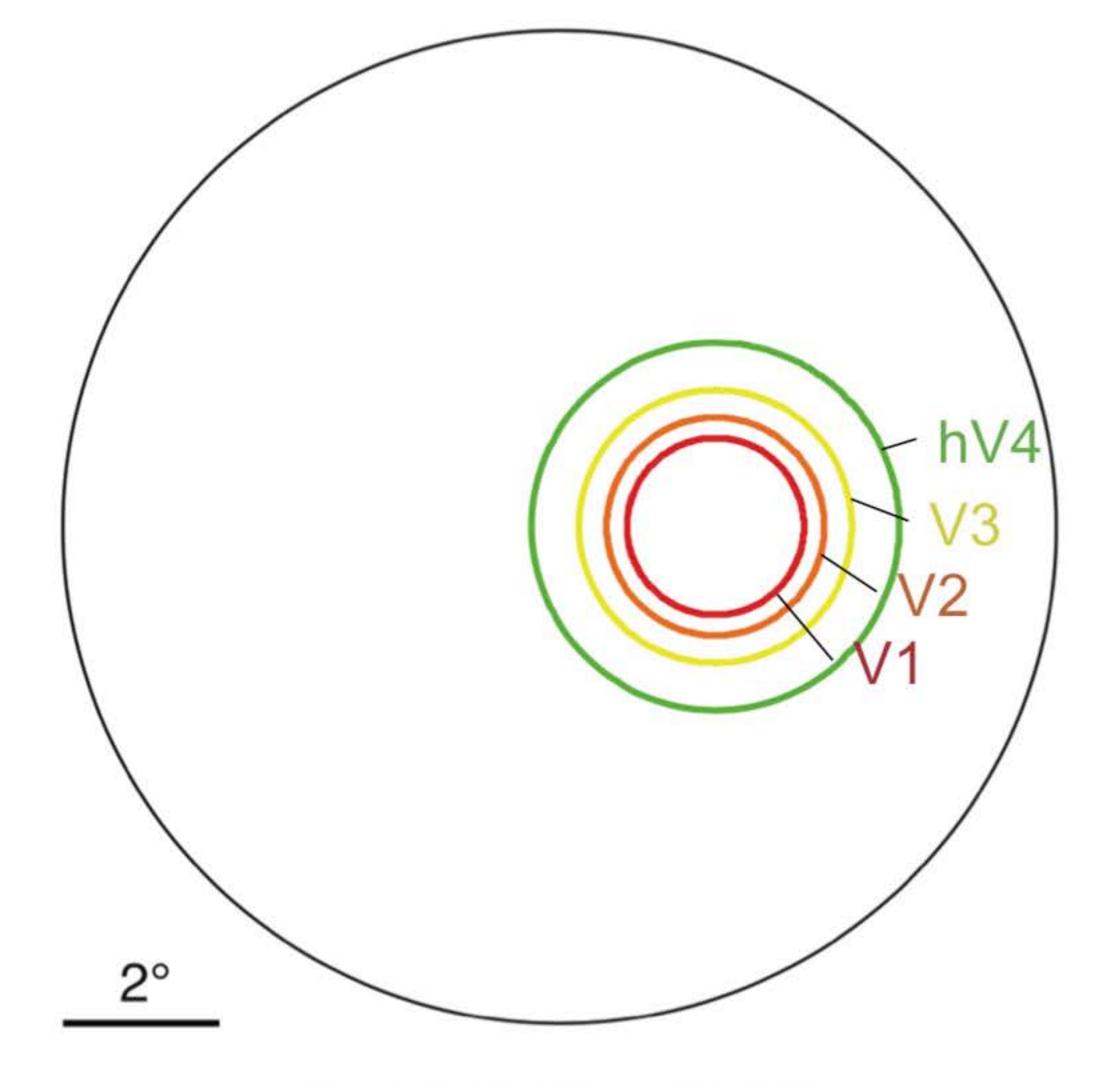
Kay et al., Curr Biol, 2015

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Observe it

Kay et al., PLoS Comp Bio, 2013

Observe it



Kay et al., PLoS Comp Bio, 2013

- **Observe it**
- Manipulate it •

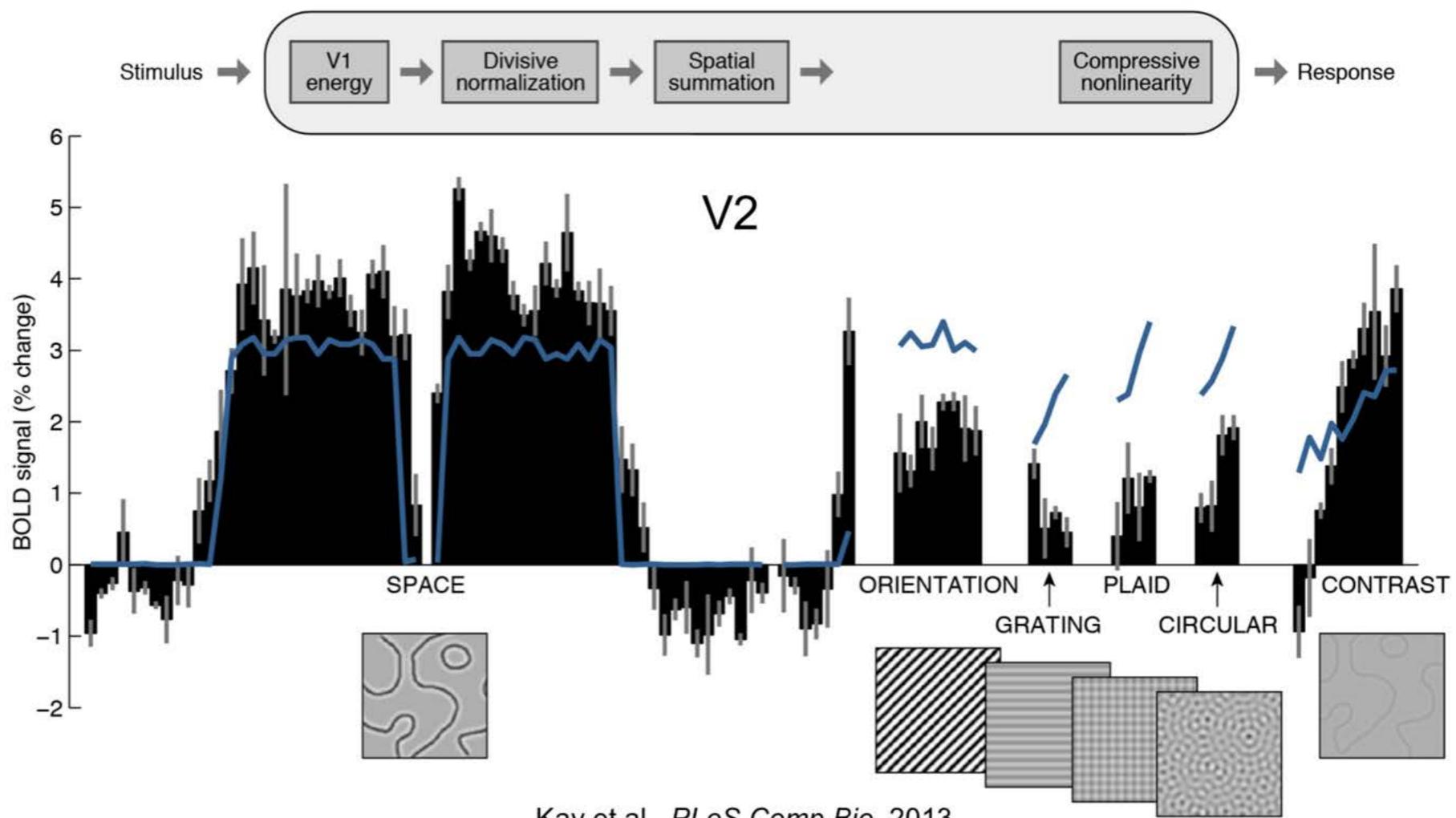
### If we change Architecture | Parameters - Accuracy | Tuning



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# Is there an effect?

- Observe it
- Manipulate it

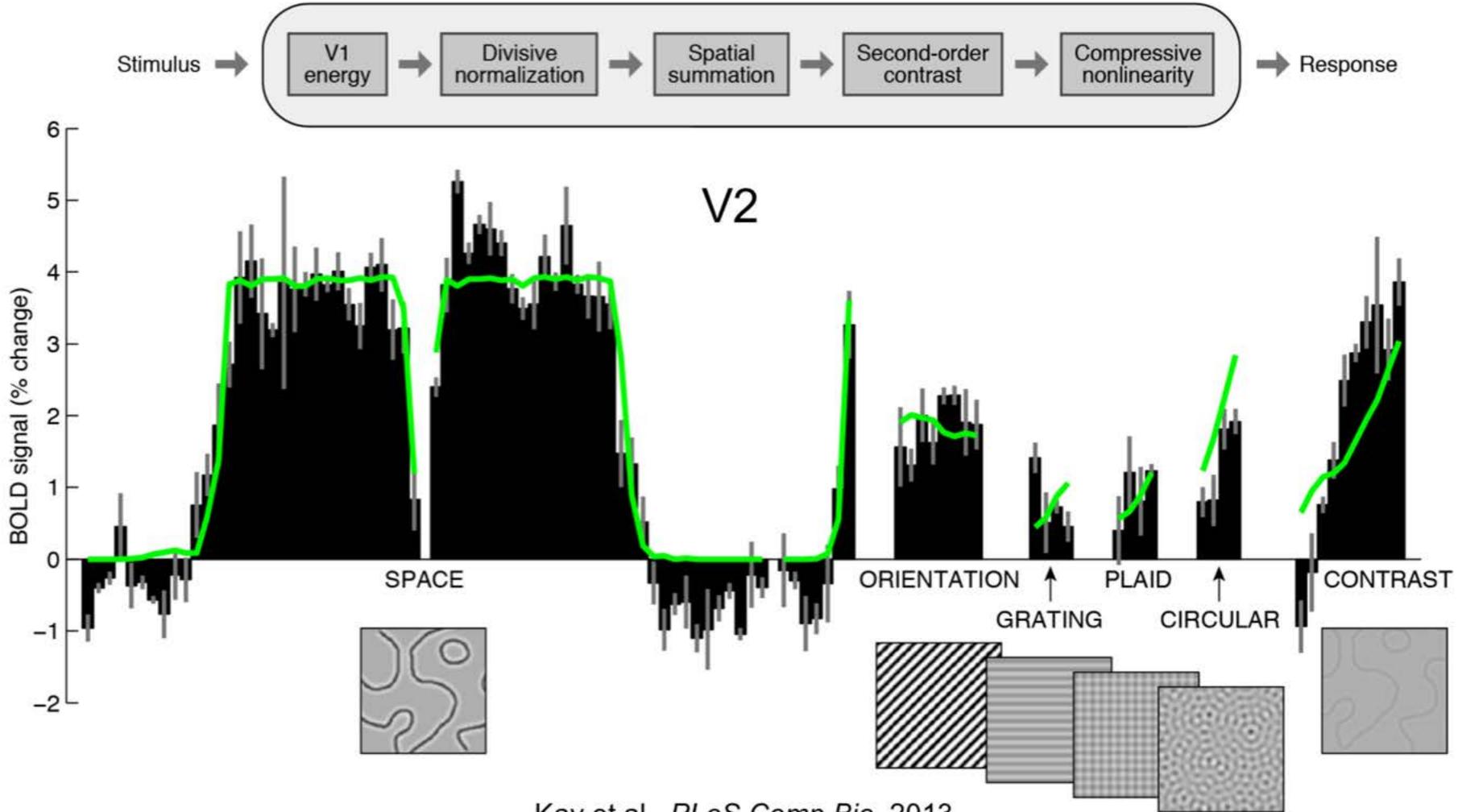


Kay et al., PLoS Comp Bio, 2013

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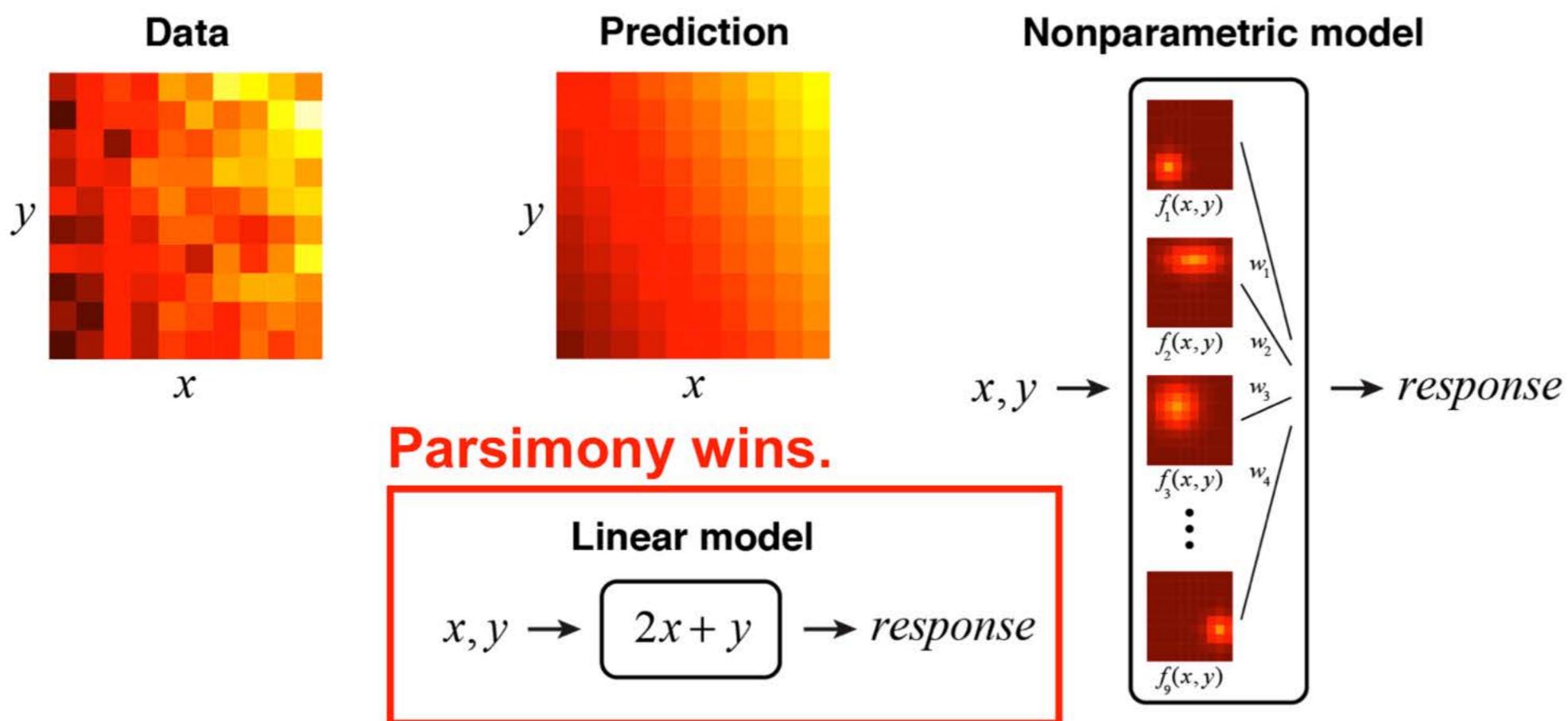
- **Observe it**
- Manipulate it ٠

Also see Nishimoto and Gallant, J Neurosci, 2011



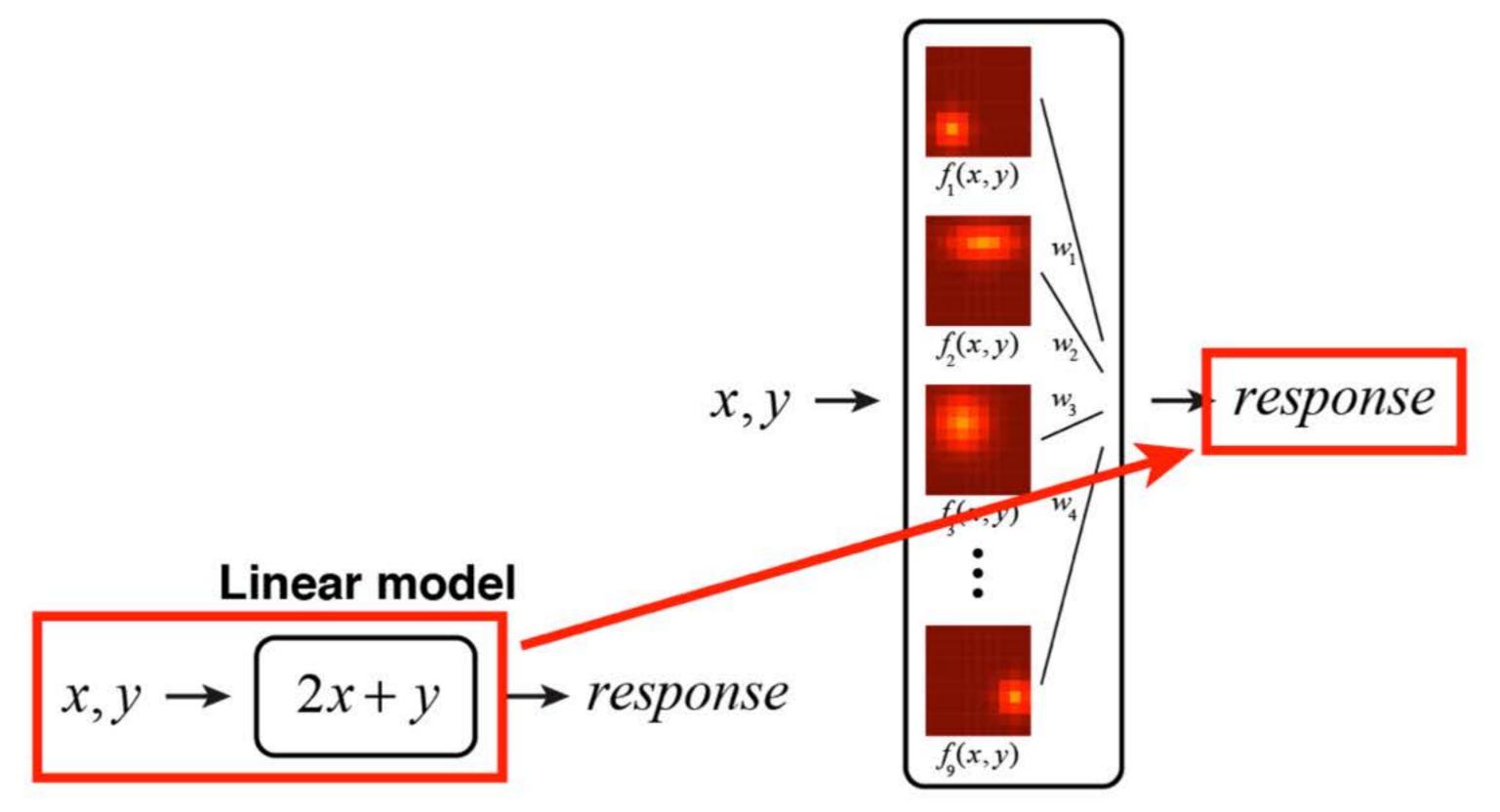
Kay et al., PLoS Comp Bio, 2013

- **Observe it**
- Manipulate it ٠
- Model it



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- **Observe it**
- Manipulate it ٠
- Model it



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### Nonparametric model

### Take-home points

- Criteria:
  - ONN strength
    Accuracy
- What we can do:
  - Observe the model
  - Manipulate the model —
  - Model the model
- Understand and simplify
  - Which filter weights are actually important?
  - Do you need all those layers?
  - Are your effects just driven by receptive field size?

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### TUNING **PARAMETERS, ARCHITECTURE** PARSIMONY

- Sunday 8:30am talk (31.22)
- Sunday AM poster (33.4071)

See also:

### Take-home points

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# In praise of shallow networks

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### **PARAMETERS, ARCHITECTURE** PARSIMONY

# TUNING

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- See also: