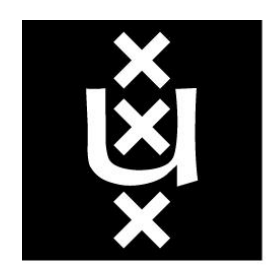


N. Müller<sup>1\*</sup>, I.I.A. Groen<sup>2</sup>, H.S. Scholte<sup>1</sup>

- Dept. of Brain and Cognition, Psychology Research Institute, University of Amsterdam
  - Video and Image Sense Lab, Informatics Institute, University of Amsterdam
- \* n.muller@uva.nl



**UNIVERSITY  
OF AMSTERDAM**

## Origins of Texture-Bias

Deep Convolutional Neural Networks (DCNNs) trained on ImageNet have been shown to exhibit a texture-bias (Geirhos et al. 2018). The origin of this texture-bias has been debated widely (e.g., Hermann et al. 2020).

Here, we show that DCNNs trained on an **ultra-high-resolution dataset** exhibit a **more human-like shape-bias**.

Further, when tested on the ImageNet texture-bias benchmark, the texture-bias drastically decreases.

## Open Amsterdam Data Set (OADS)

Ultra-high-resolution, labelled image dataset  
5691 Images  $\leftrightarrow$  **5496x3672** pixels  
98534 Object Label annotations



## Contributions

Introduction of ultra-high-resolution labelled image dataset OADS

Creation of high-resolution cue-conflict dataset

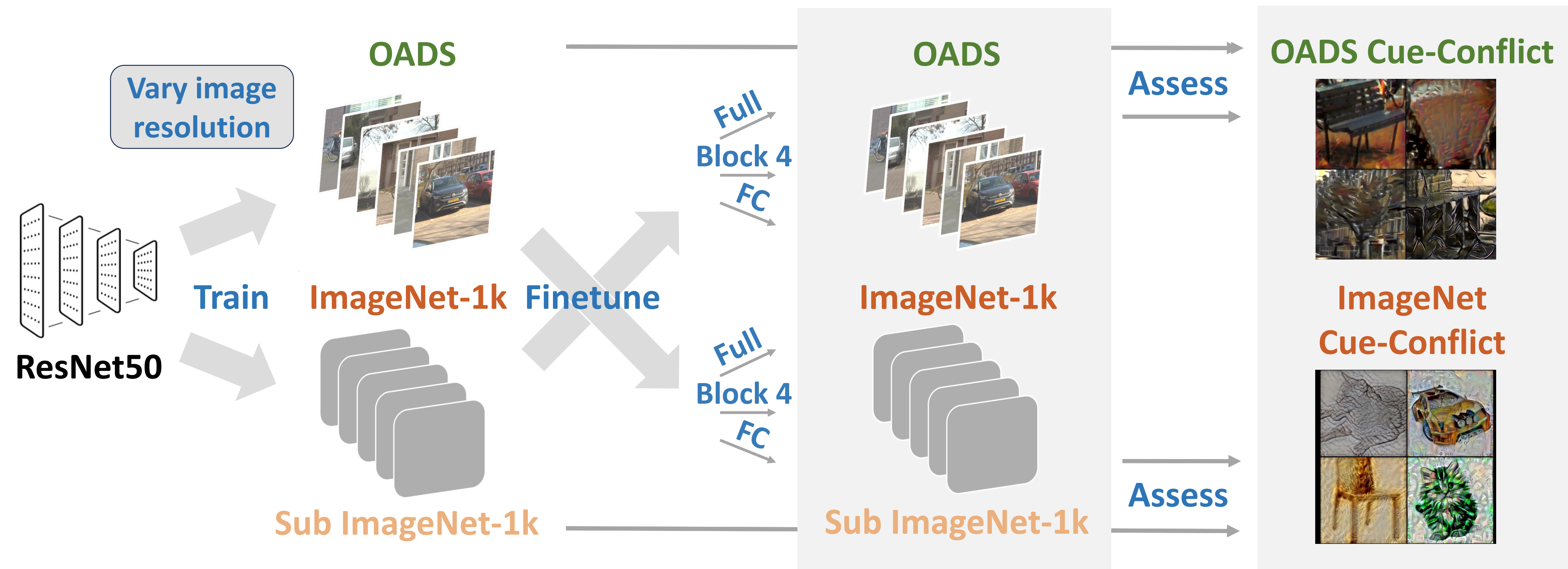
**Texture-bias arises as a function of image quality**

**Texture-bias is reflected in representational geometry**

**Interaction between image resolution during training and testing and texture-bias**

## DCNN Training & Assessment

Schematic representation of model training, cross-finetuning and texture-bias assessment

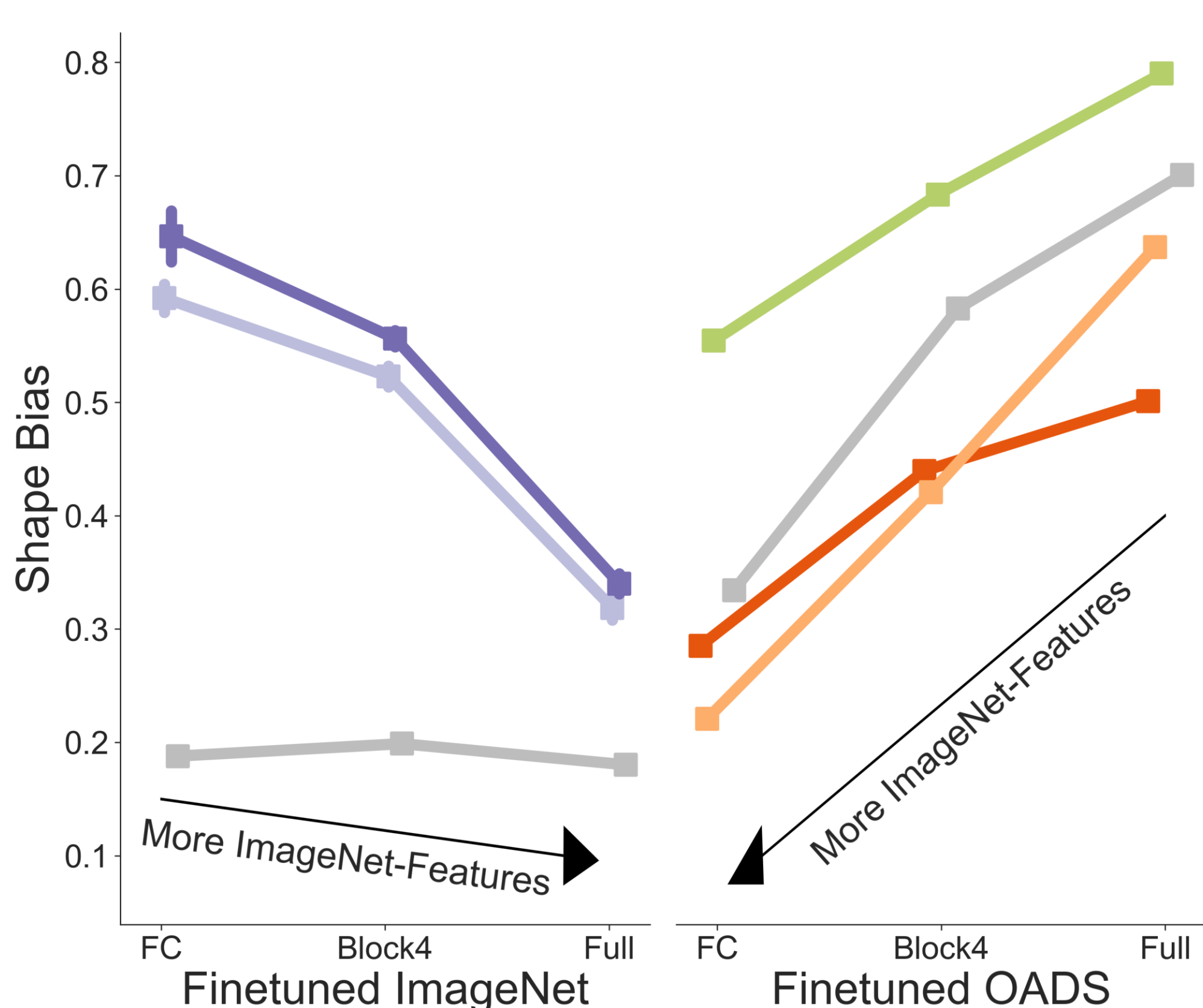


We train ResNet50 models with different image resolutions on a low-quality dataset (ImageNet) and a high-quality dataset (OADS) and assess DCNN texture-bias on both low-quality and high-quality cue-conflict benchmarks.

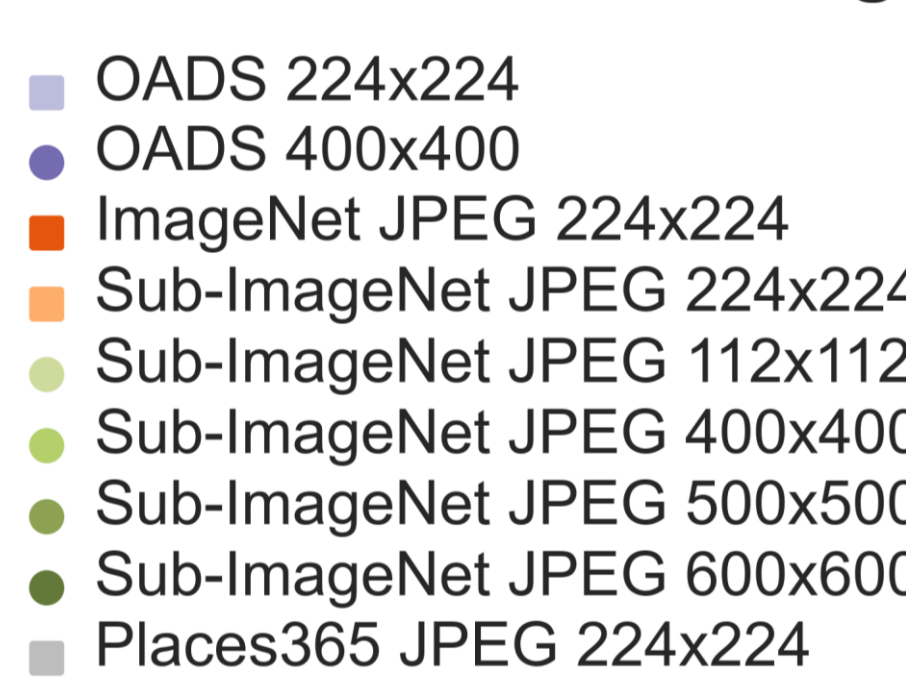
We created an **OADS** Cue-Conflict dataset using Neural Style Transfer<sup>3</sup> offering an alternative, high-resolution assessment of texture-bias in DCNNs to the ImageNet benchmark in Geirhos et al., 2018.

## Image Quality vs. Texture-Bias

### ImageNet-Features display low shape-bias



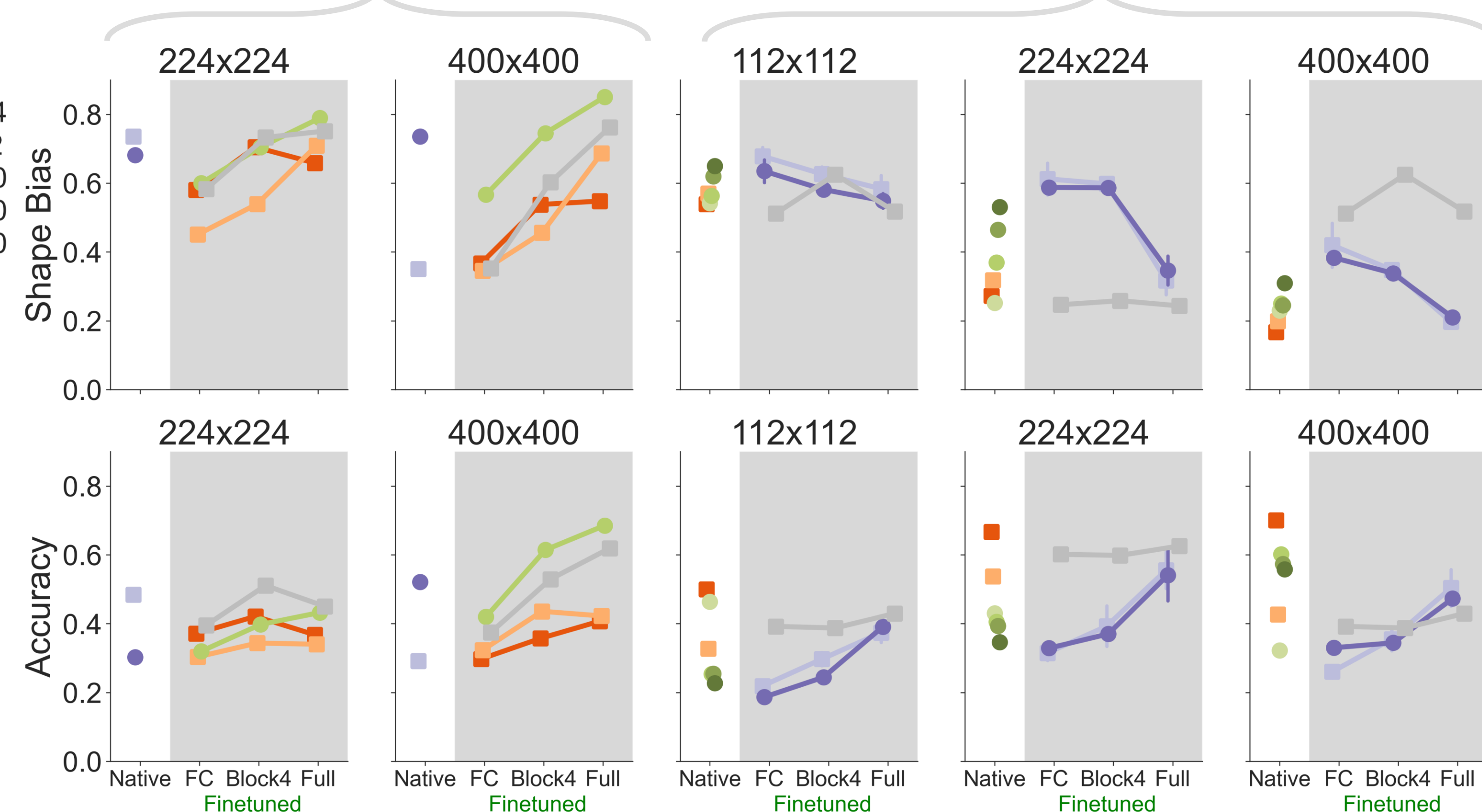
### ResNet50 Pre-Training



**Shape-bias increases with image resolution**

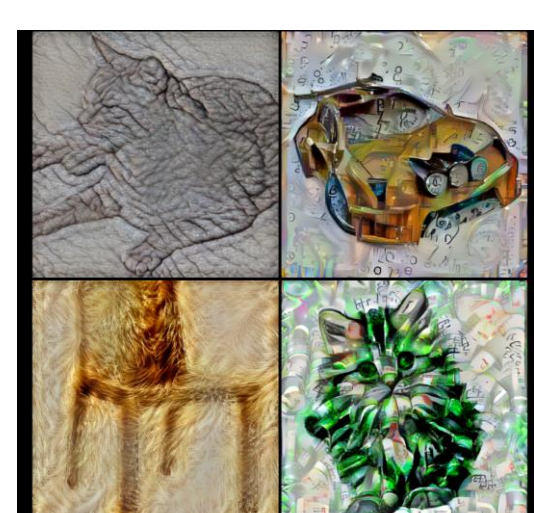
**Dataset content influences shape-bias**

### OADS Cue-Conflict

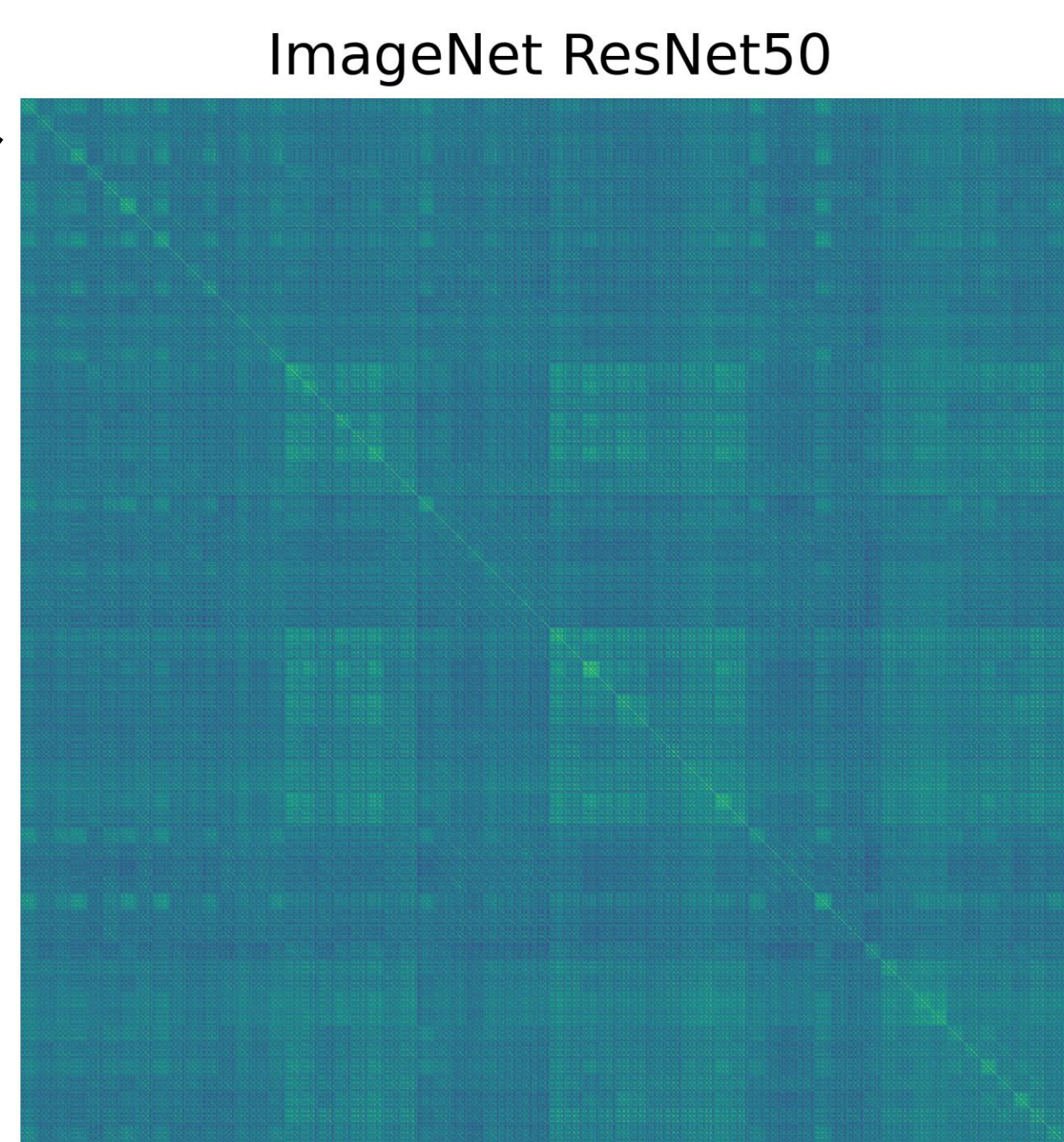
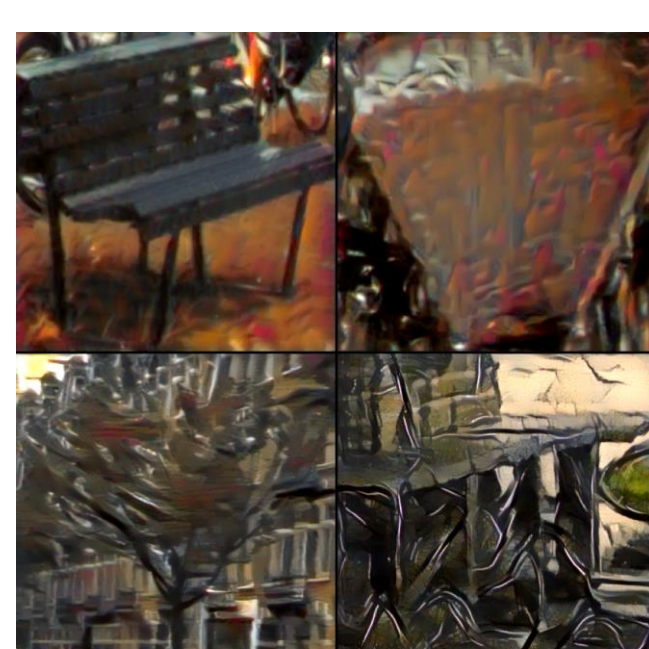


### ImageNet Cue-Conflict

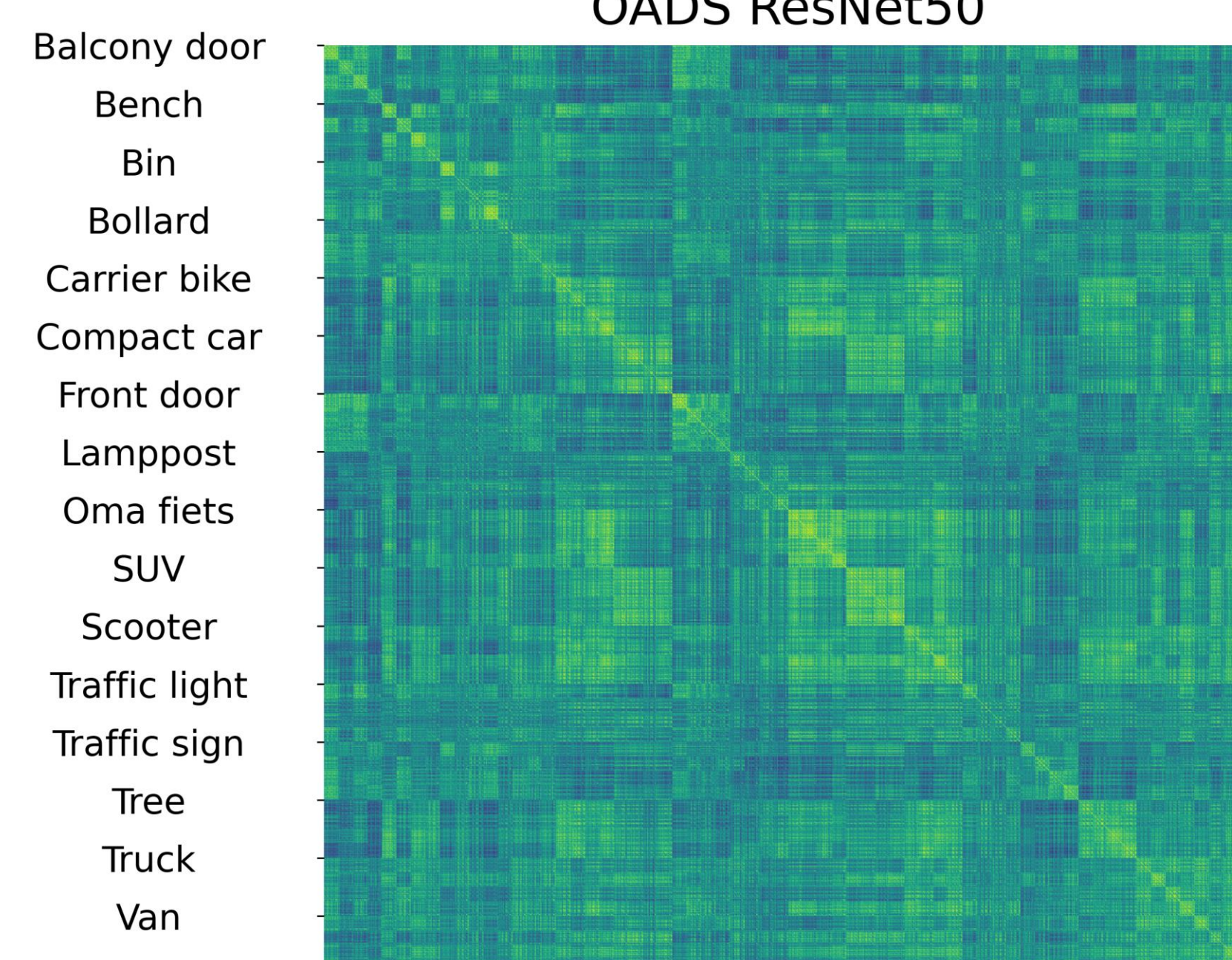
**ImageNet  
Cue-Conflict**



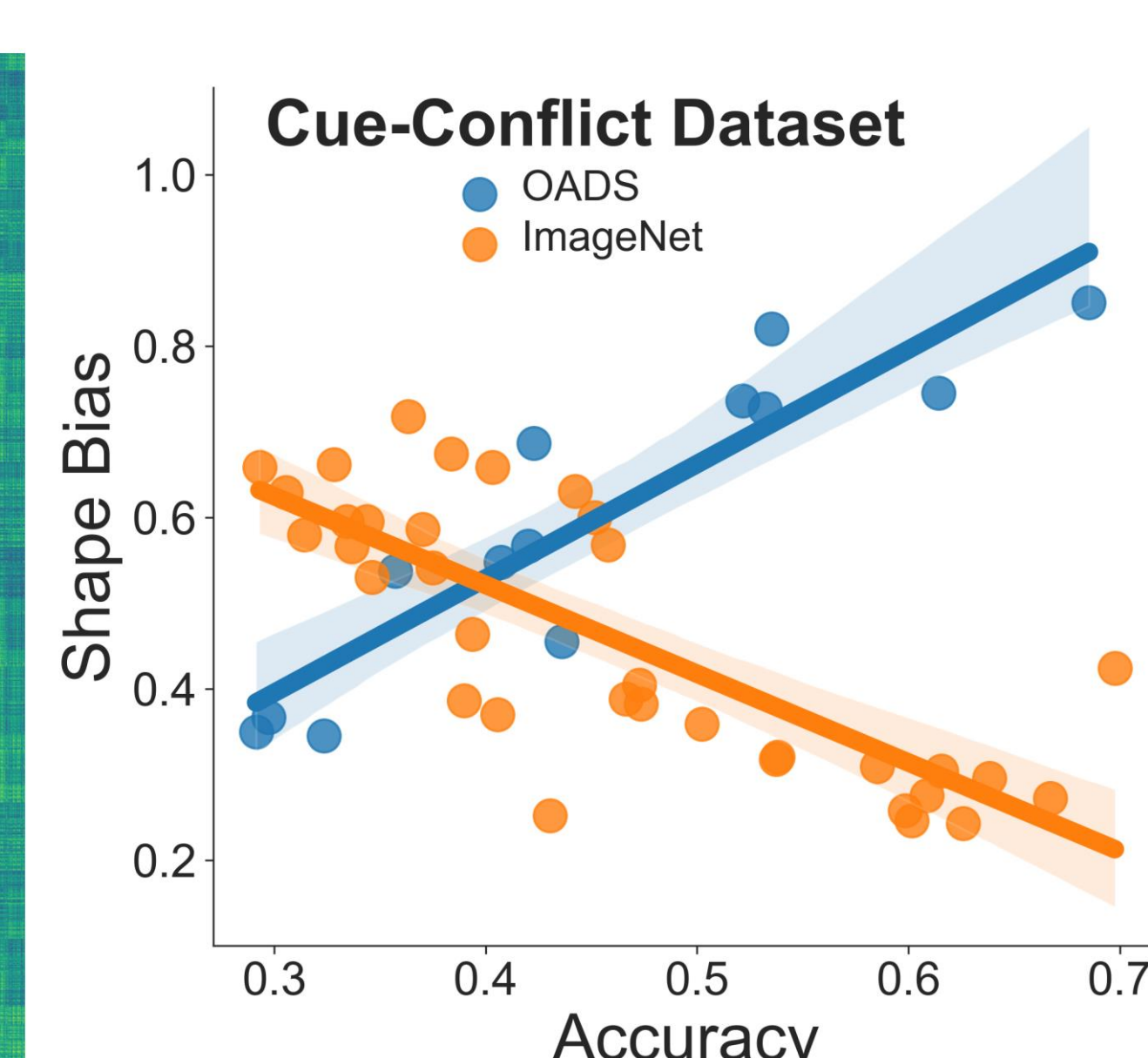
**OADS  
Cue-Conflict**



**High Texture-Similarity**



**High Shape-Similarity**



**Accuracy determines estimation precision**

**Acknowledgements** | This work was funded by the Interdisciplinary PhD Program from the UvA Data Science Center

**References** | <sup>1</sup> Geirhos et al., CVPR, 2018; <sup>2</sup> Hermann et al., NeurIPS, 2020; <sup>3</sup> Gatys et al., IEEE, 2016.