# Notes on the series of articles on the topic Deconstructing Neural Networks using Circuits

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## Introductory notes

Given the popularity of the Neural Networks for Computer Vision it is clear that the scientific community needs *mechanistic interpretability* for those neural network architectures.

Speculative claims about Neural Networks operations and explainability of the latter

**Features**: *Features* are the fundamental unit of Neural Networks. They correspond to *Directions*. By *Directions* the authors mean linear combination of neurons in a layer. One can think of this as a direction vector in the vector space of activations of neurons in a given layer. Often individual neurons will be discussed and analyzed but there are cases where analyzing combinations of neuronal output is the best way to understand the functioning of the neural net. This becomes even more important when we will be dealing with polysemantic neurons.

## References

[Zoom In: An Introduction to Circuits, Chris Olah et al, OpenAI, 2020](https://distill.pub/2020/circuits/zoom-in)

[An Overview of Early Vision in InceptionV1, Chris Olah et al, OpenAI, 2020](https://distill.pub/2020/circuits/early-vision/)

[Curve Detectors, Nick Cammarata et al, OpenAI, 2020](https://distill.pub/2020/circuits/curve-detectors/)

[Naturally Occurring Equivariance in Neural Networks, Chris Olah et al, OpenAI, 2020](https://distill.pub/2020/circuits/equivariance/)

[High-Low Frequency Detectors, Ludwig Schubert et al, OpenAI, 2021](https://distill.pub/2020/circuits/frequency-edges/)

[Curve Circuits, Nick Cammarata et al, OpenAI, 2021](https://distill.pub/2020/circuits/curve-circuits/)

[Visualizing Weights, Chelsea Voss et al, OpenAI, 2021](https://distill.pub/2020/circuits/visualizing-weights/)

[Branch Specialization, Chelsea Voss et al, OpenAI, 2021](https://distill.pub/2020/circuits/branch-specialization/)