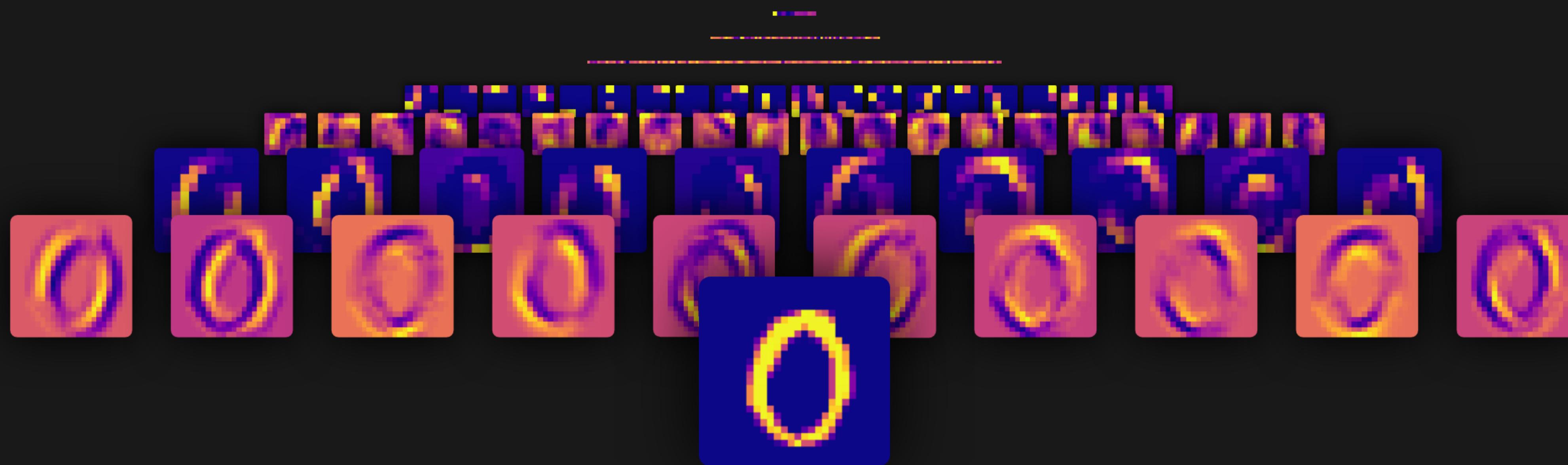


Fourier Analysis of Activations in Neural Networks

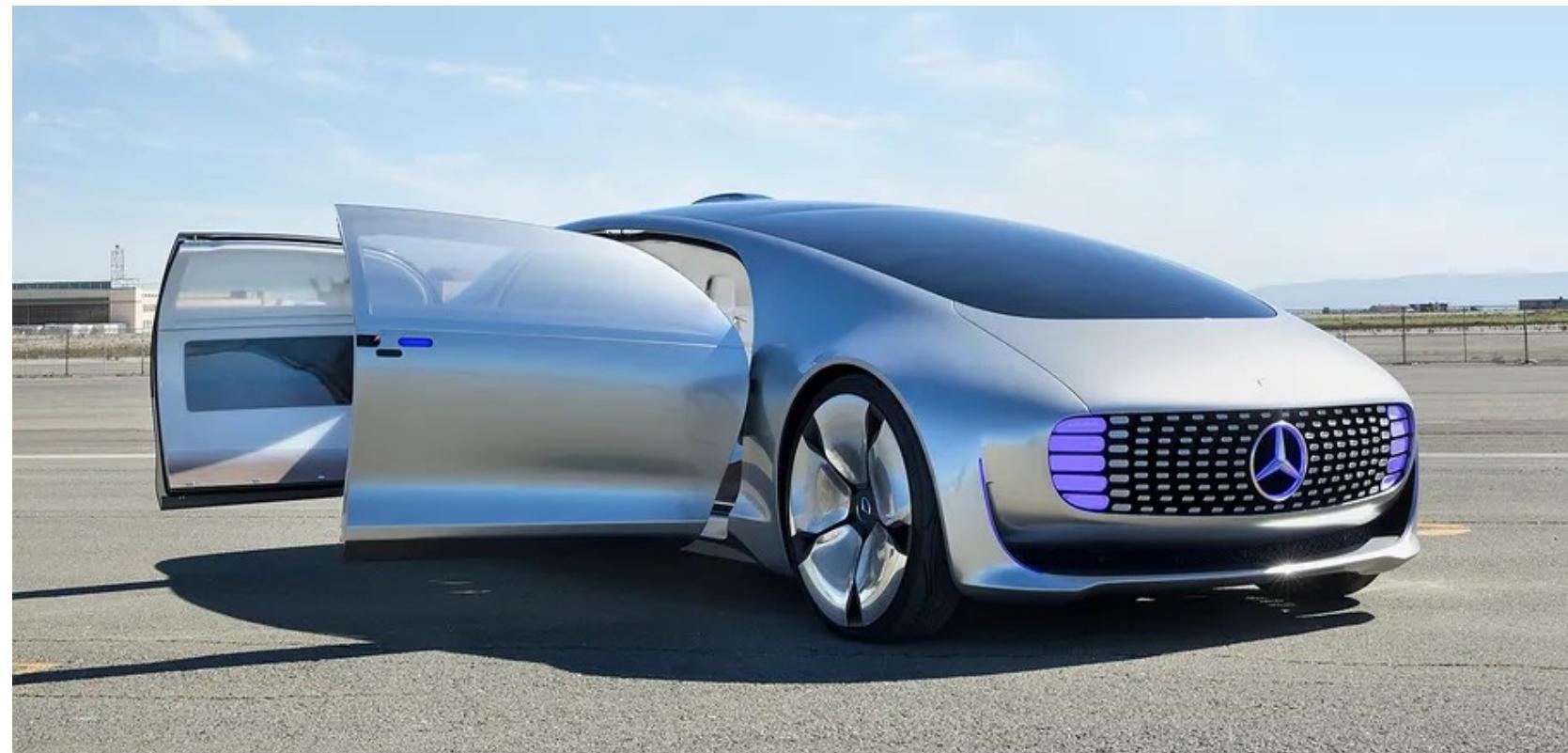
Bachelor-Thesis



Analysis on Neural Networks

Motivation

Self-driving cars



<https://www.wired.com/2017/04/mercedes-promises-self-driving-taxis-just-three-years/>

Healthcare



<https://www.medgadget.com/2019/08/healthcare-analytics-market-2019-industry-analysis-size-share-upcoming-trends-segmentation-forecast-to-2025-cagr-of-26-48.html>

Cyber-security



<https://www.news18.com/news/tech/hacker-group-darkside-suspected-of-carrying-cyber-attack-on-top-us-pipeline-operator-colonial-3723686.html>

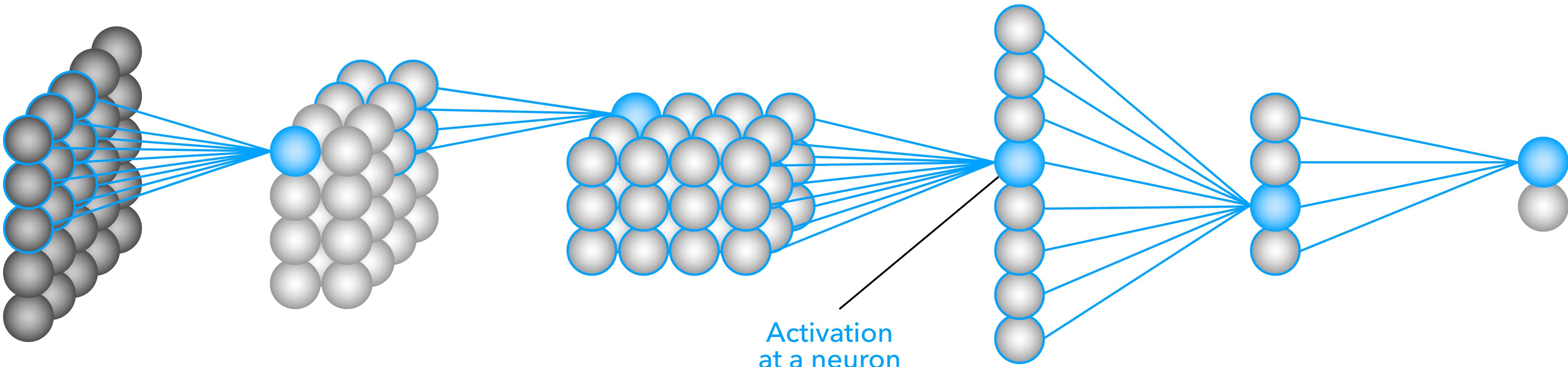
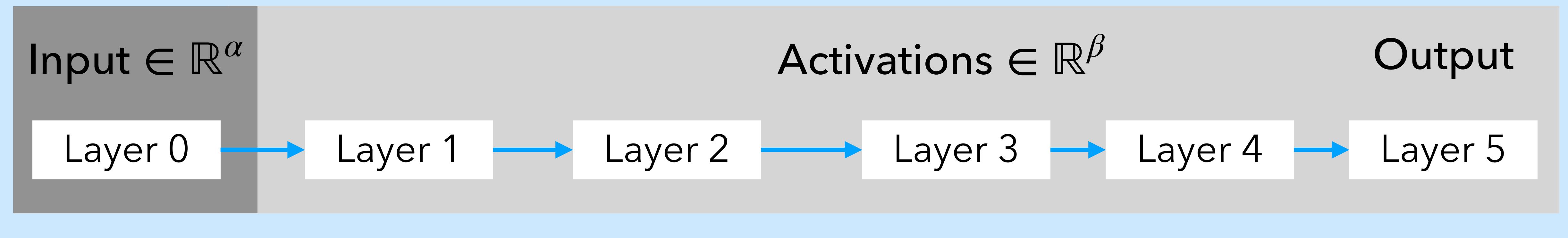
Ethics



<https://todaysveterinarybusiness.com/practice-without-prejudice/>

Neural Networks (NN)

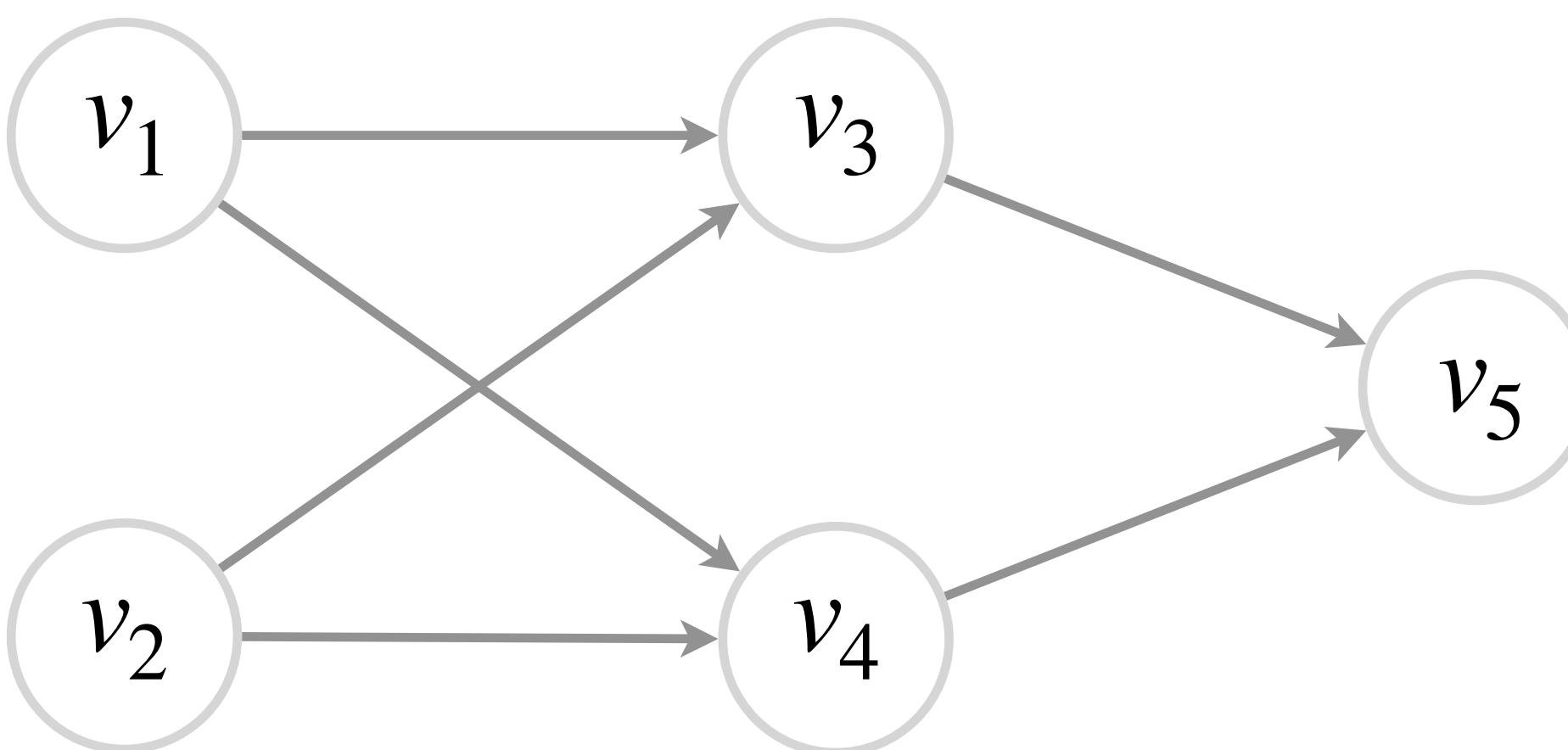
Activation pattern $\in \mathbb{R}^{\alpha+\beta}$



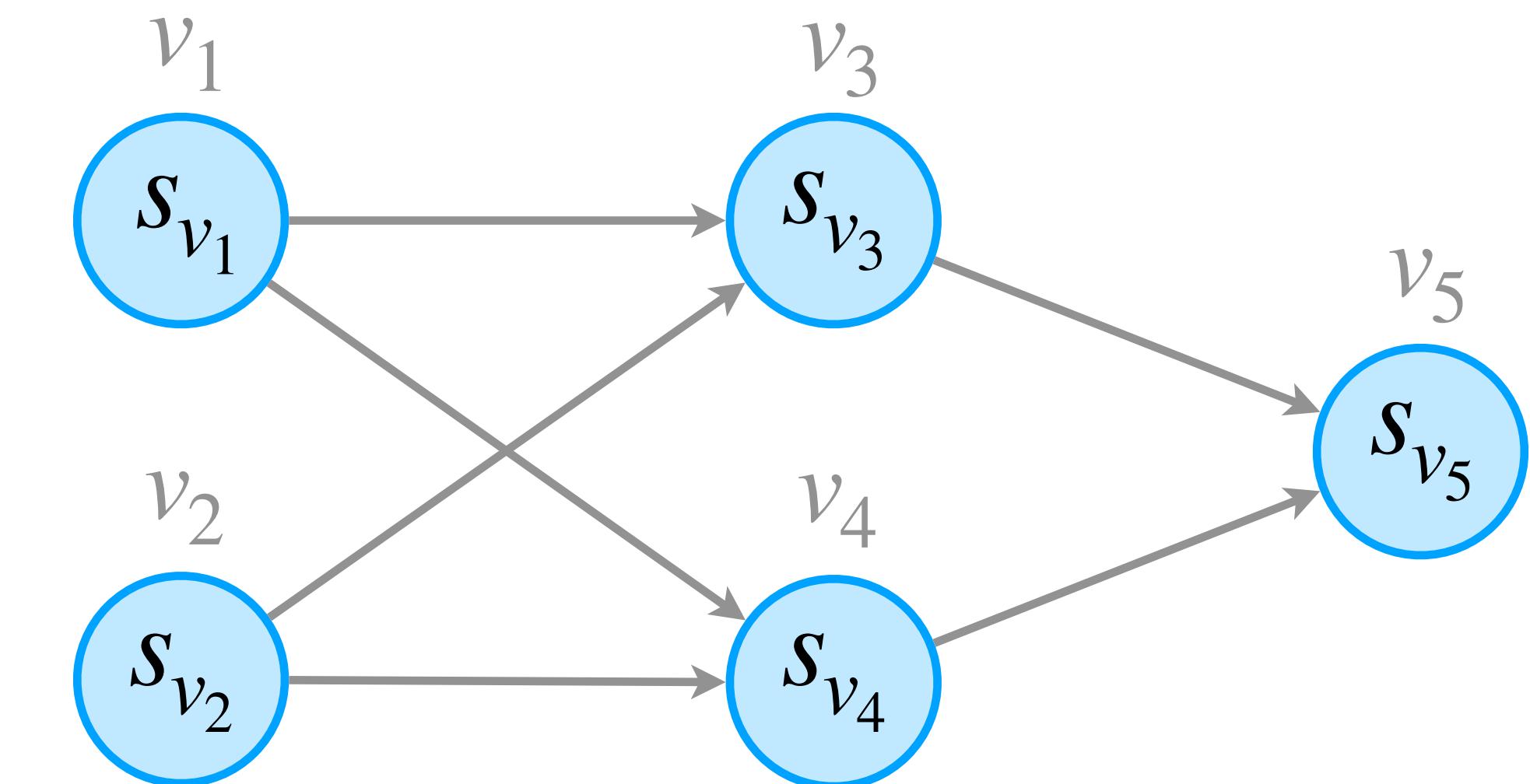
Causal Signal Processing (CSP)

Data-domain

Directed acyclic graph $G = (V, E)$



Signal $s \in \mathbb{R}^{|V|}$



Intuition:

$$s_x = \sum_{y \leq x, y \in V} \hat{s}_y$$

Causes

$$s_{v_3} = \hat{s}_{v_1} + \hat{s}_{v_2} + \hat{s}_{v_3}$$

Unobserved weight

Causal Signal Processing (CSP)

Data-domain

Signal s

$$s_x = \sum_{y \leq x, y \in V} \hat{s}_y$$

Zeta

$$\hat{s}_y = \sum_{y \in V} \zeta(x, y) \hat{s}_y$$

$$s = F^{-1} \hat{s}$$

Inverse Fourier transform F^{-1}

$$F_{x,y}^{-1} = \zeta(x, y) = \begin{cases} 1 & \text{if } y \leq x , \\ 0 & \text{otherwise.} \end{cases}$$

Fourier-domain

Fourier coefficients \hat{s}

$$\hat{s}_x = \sum_{y \in V} \mu(x, y) s_y$$

Moebius

$$\hat{s} = F s$$

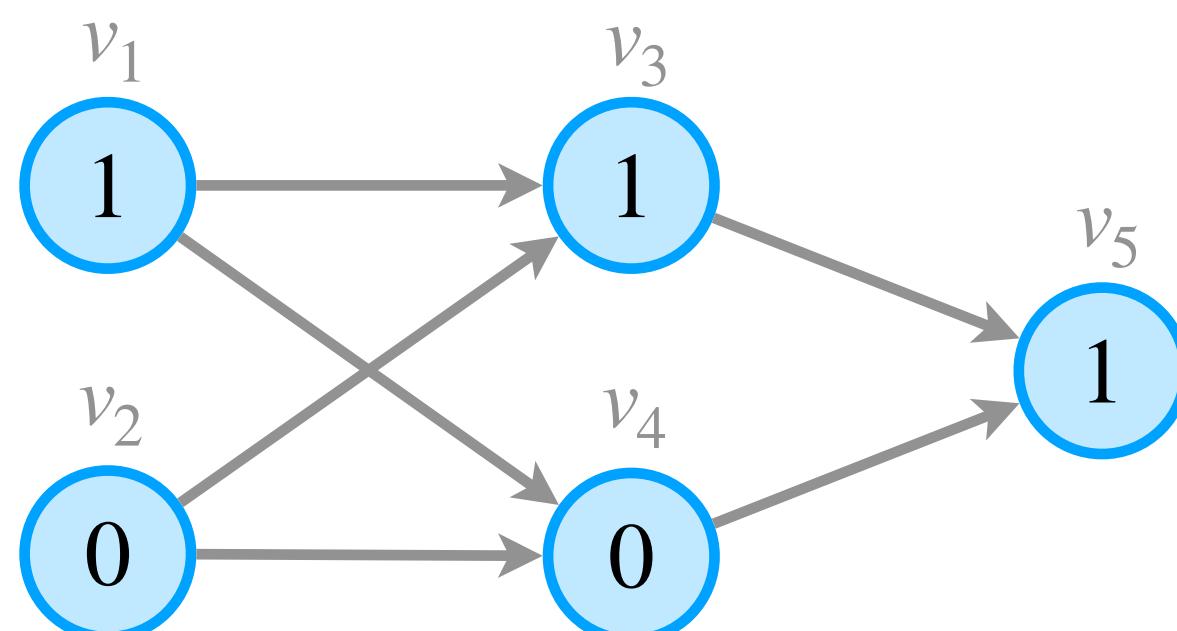
Fourier transform F

$$F_{x,y} = \mu(y, x) = \begin{cases} 1 & \text{if } x = y , \\ -\sum_{y \leq z < x} \mu(y, z) & \text{otherwise.} \end{cases}$$

Causal Signal Processing (CSP)

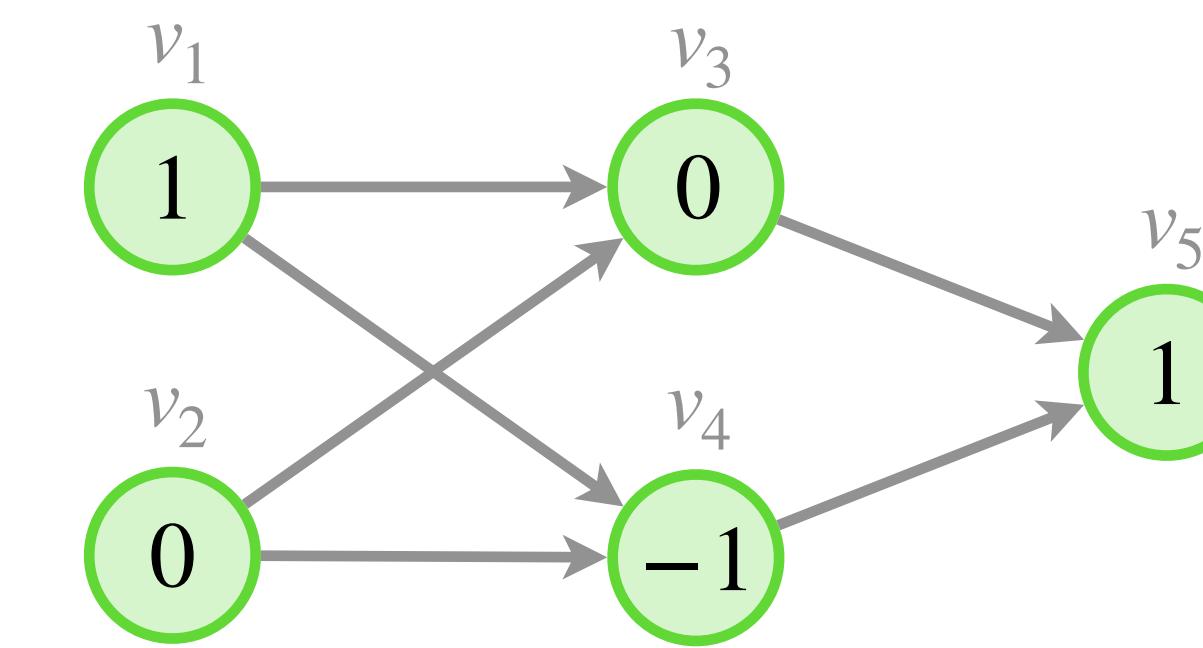
Data-domain

Signal s



Fourier-domain

Fourier coefficients \hat{s}



$$s = F^{-1} \hat{s}$$

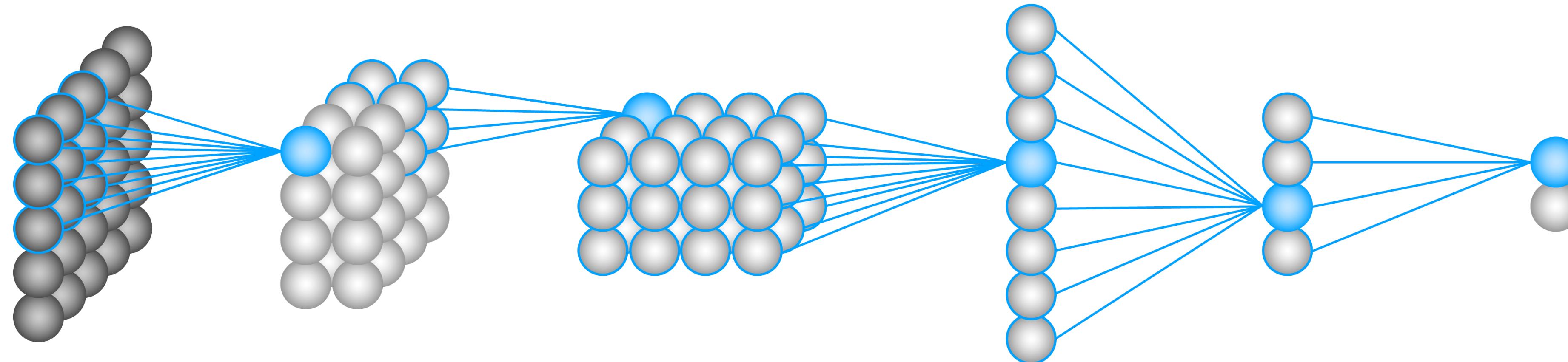
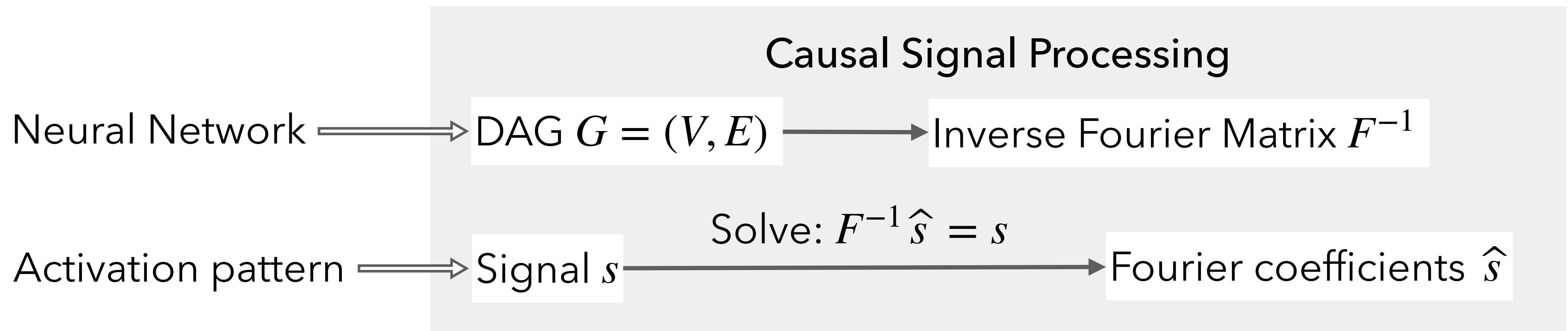
$$\begin{bmatrix} 1 \\ 0 \\ 1 \\ 0 \\ 1 \end{bmatrix} = \begin{bmatrix} 1 & 0 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 & 0 \\ 1 & 1 & 1 & 0 & 0 \\ 1 & 1 & 0 & 1 & 0 \\ 1 & 1 & 1 & 1 & 1 \end{bmatrix} \begin{bmatrix} 1 \\ 0 \\ 0 \\ -1 \\ 1 \end{bmatrix} \quad \begin{matrix} v_1 \\ v_2 \\ v_3 \\ v_4 \\ v_5 \end{matrix}$$

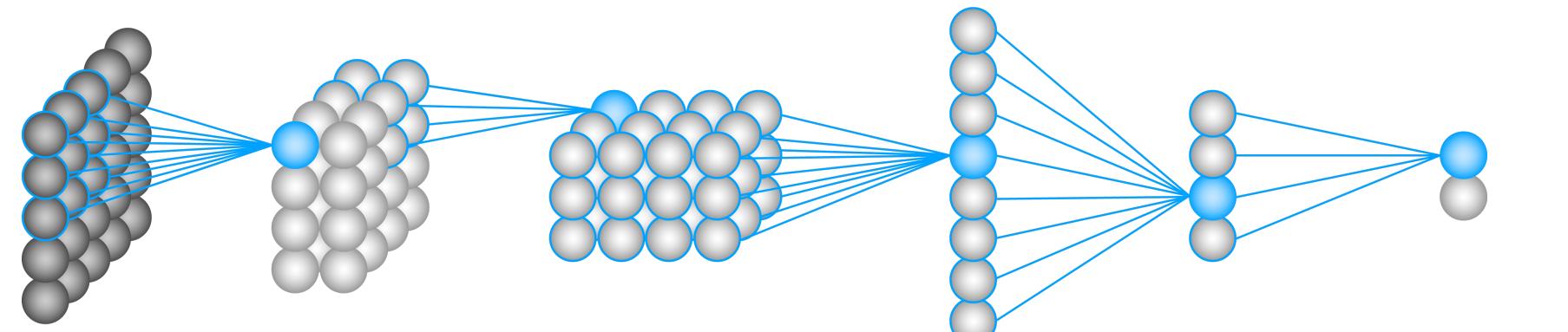
$$\hat{s} = F s$$

$$\begin{bmatrix} 1 \\ 0 \\ 0 \\ -1 \\ 1 \end{bmatrix} = \begin{bmatrix} 1 & 0 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 & 0 \\ -1 & -1 & 1 & 0 & 0 \\ -1 & -1 & 0 & 1 & 0 \\ 1 & 1 & -1 & -1 & 1 \end{bmatrix} \begin{bmatrix} 1 \\ 0 \\ 1 \\ 0 \\ 1 \end{bmatrix} \quad \begin{matrix} v_1 \\ v_2 \\ v_3 \\ v_4 \\ v_5 \end{matrix}$$

CSP on Neural Networks

Thesis Project





$$F_{x,y}^{-1} = \begin{cases} 1 & \text{if } y \leq x \\ 0 & \text{otherwise} \end{cases}$$

Layer 0: Input

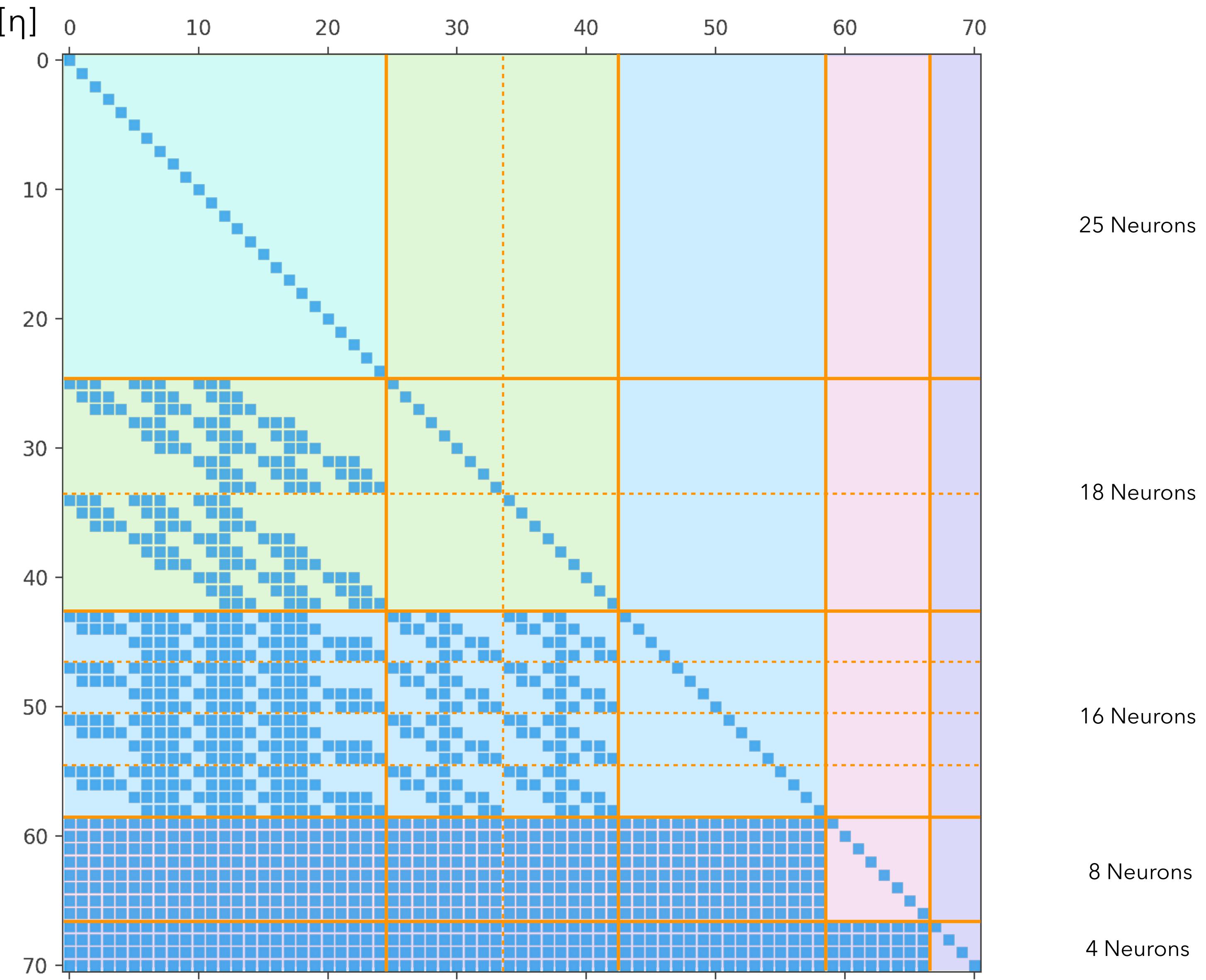
Layer 1: Convolution

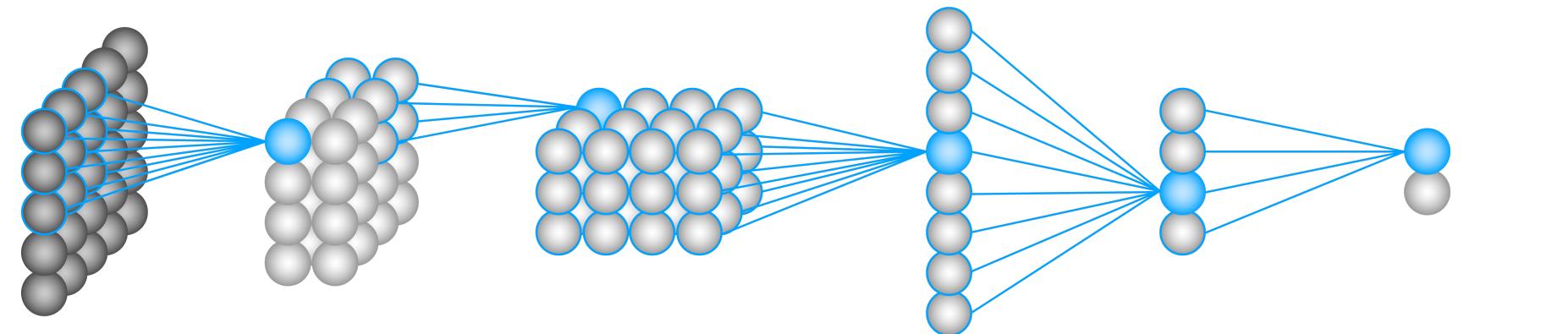
Layer 3: Convolution

Layer 4: Linear

Layer 5: Linear

Inverse Fourier Matrix F^{-1}





$$F_{x,y}^{-1} = \begin{cases} 1 & \text{if } y \leq x \\ 0 & \text{otherwise} \end{cases}$$

Layer 0: Input

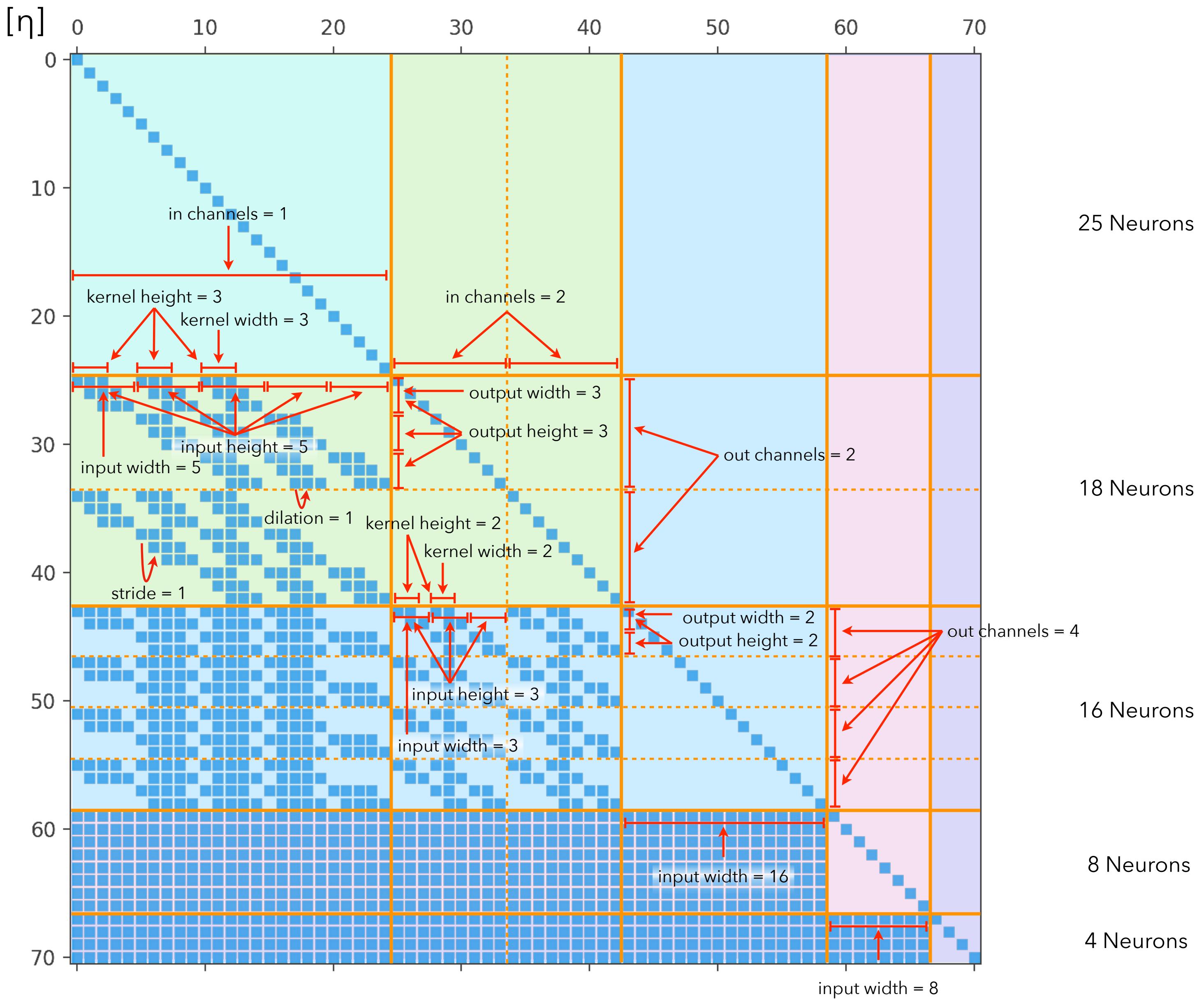
Layer 1: Convolution

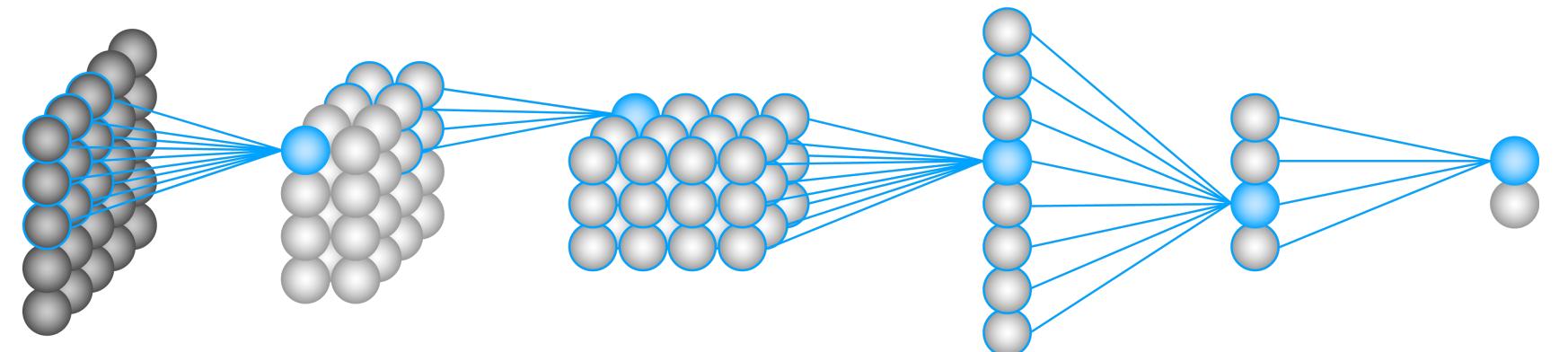
Layer 3: Convolution

Layer 4: Linear

Layer 5: Linear

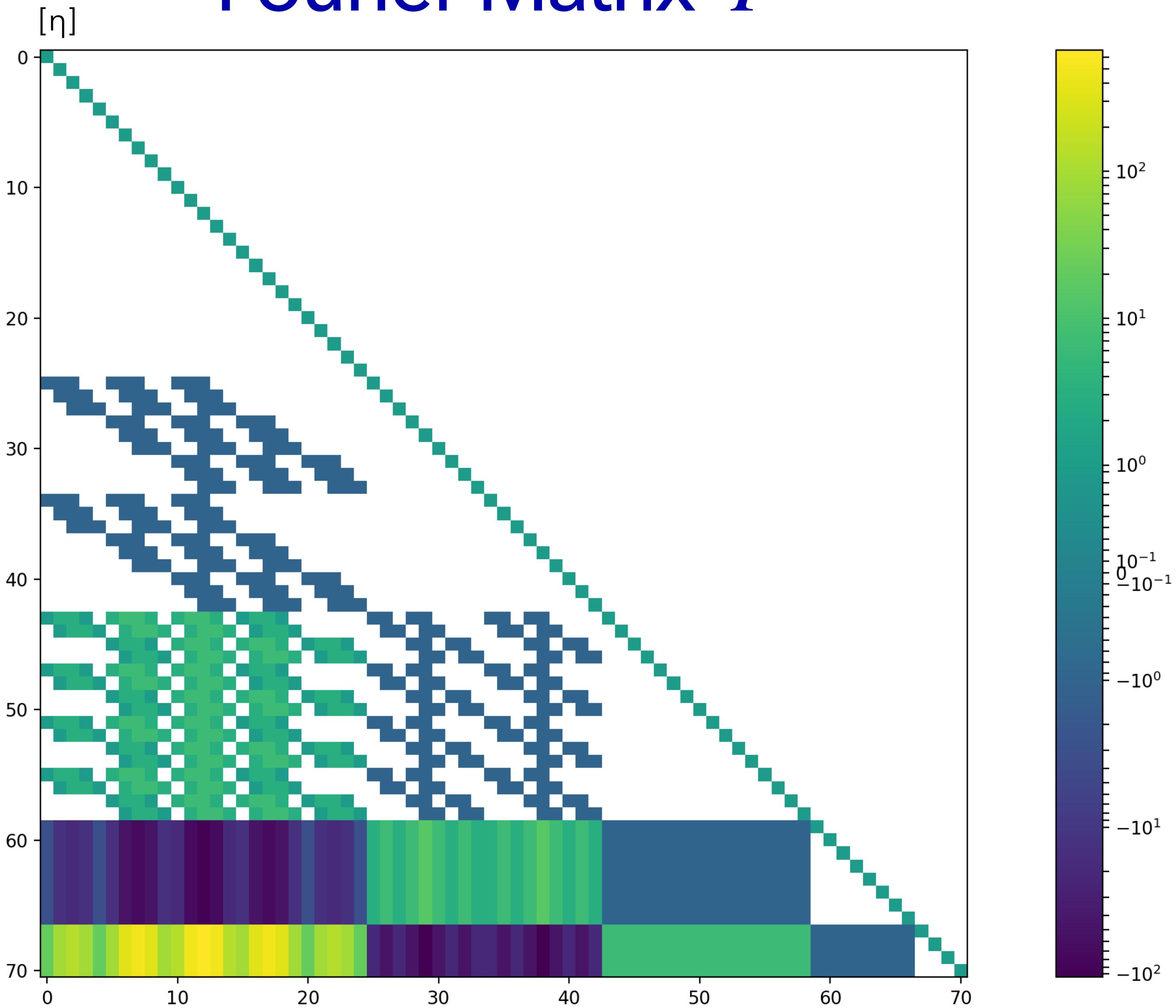
Inverse Fourier Matrix F^{-1}



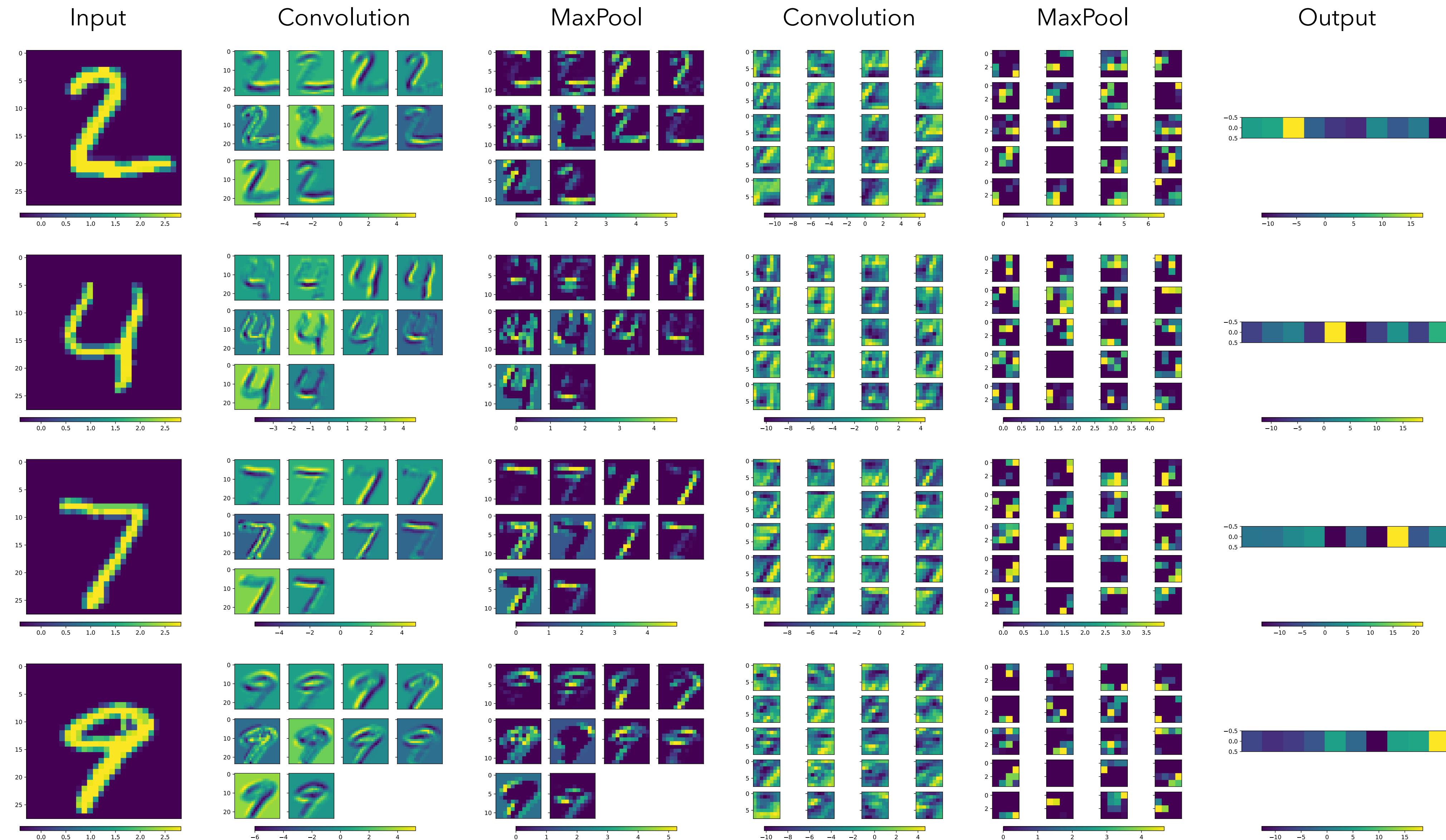


$$F_{x,y} = \begin{cases} 0 & \text{if } y \not\leq x \\ 1 & \text{if } x = y \\ -1 & \text{if } (y,x) \in E \\ -\sum_{\substack{y \leq w \leq z \\ (z,x) \in E}} \mu(y,w) & \text{otherwise} \end{cases}$$

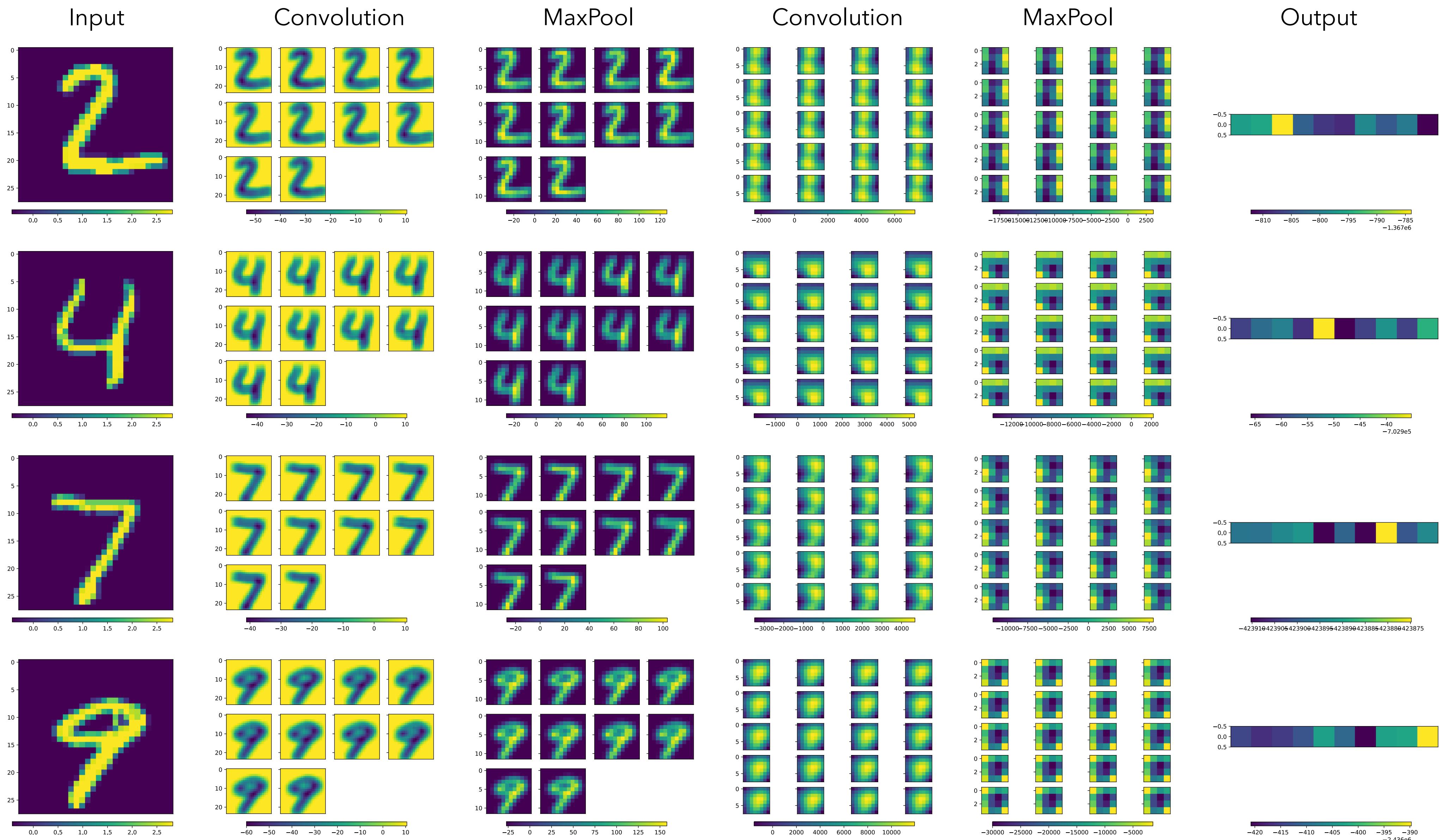
Fourier Matrix F



Activations S



Fourier Coefficients \hat{s}



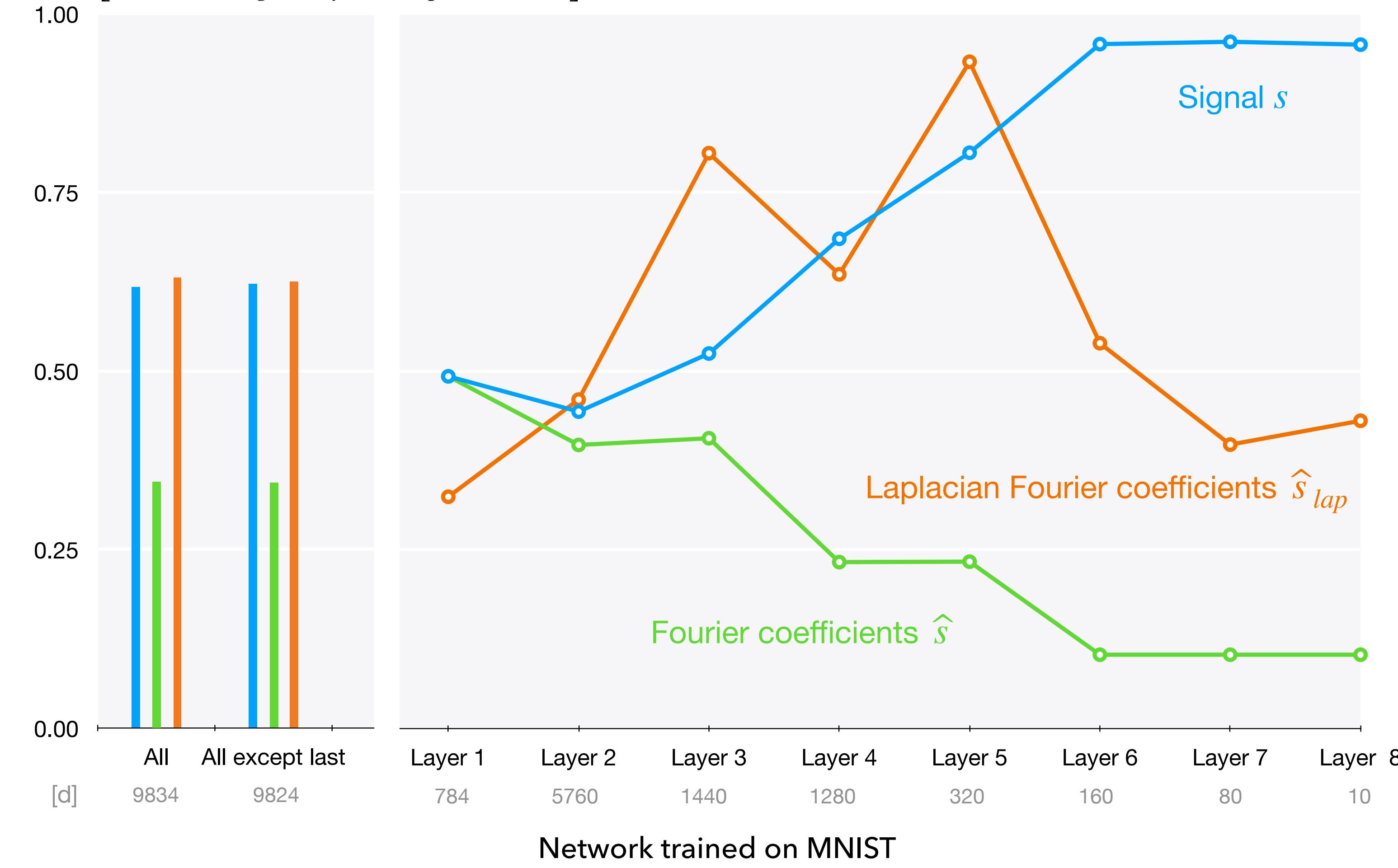
Clustering

K-means clustering on the Euclidean distance: K = 10, N = 1000

$$V = \frac{2 \times \text{homogeneity} \times \text{completeness}}{\text{homogeneity} + \text{completeness}}$$

$$\hat{s}_x = F_{x,:} s$$

$$\hat{s}_x = c_{s,i} + s_x$$



pip install nsp

