

## *Supplementary Material*

### 1 TRAINING HYPERPARAMETERS

Table S1 summarizes the training hyperparameters used for each dataset, including learning rate, batch size, and early-stopping patience. Unless otherwise noted, hyperparameters were kept fixed across architectures within each dataset.

**Table S1.** Training hyperparameters used for each dataset. Patience refers to the number of epochs without validation improvement before early stopping is triggered. “N/A” indicates that early stopping was not applied.

Dataset	Learning rate	Batch size	Patience
mnist	0.001	128	5
fashion_mnist	0.001	128	5
sincos	0.001	64	5
uci_har	0.01	32	10
uci_har_raw	0.01	32	10
mnist_shuffled	0.001	128	N/A
random	0.001	128	N/A

### 2 COMPLETE EXPERIMENTAL RESULTS

In Table S2, we report the complete set of experimental results referenced in the main text, covering all datasets, model architectures, and training configurations evaluated.

**Table S2.** Complete experimental results across datasets and architectures. Reported values are correlation intervals [95% CI] between training loss and complexity metrics. Bold values indicate cases of non-overlapping confidence intervals, with the higher correlation highlighted.  $\Delta r$  and  $\Delta \rho$  report the difference between midpoint estimates (midpoint(BDM) minus midpoint(entropy)), computed from the same rounded bounds. Accuracy is reported as mean  $\pm$  standard deviation. Epochs denotes the average number of epochs used (mean  $\pm$  std), and  $N$  denotes the number of valid runs used for bootstrap estimation.

Dataset	Hidden layers	Params	Entropy $r$	BDM $r$	$\Delta r$	Entropy $\rho$	BDM $\rho$	$\Delta \rho$	Accuracy (%)	Epochs	$N$
mnist	16	1760	[0.60, 0.68]	<b>[0.91, 0.93]</b>	0.28	[0.58, 0.68]	<b>[0.77, 0.85]</b>	0.18	79.5 $\pm$ 1.2	13.3 $\pm$ 5.2	199
mnist	80	8800	[0.84, 0.86]	<b>[0.92, 0.93]</b>	0.08	[0.83, 0.88]	<b>[0.89, 0.93]</b>	0.06	88.5 $\pm$ 0.4	20.5 $\pm$ 5.8	200
mnist	192	21120	[0.86, 0.88]	<b>[0.92, 0.92]</b>	0.05	[0.93, 0.95]	<b>[0.96, 0.97]</b>	0.03	90.3 $\pm$ 0.3	24.8 $\pm$ 6.2	200
mnist	8, 4	872	[0.47, 0.60]	<b>[0.77, 0.86]</b>	0.28	[0.52, 0.68]	<b>[0.74, 0.85]</b>	0.19	52.1 $\pm$ 5.3	9.3 $\pm$ 3.6	163
mnist	16, 8	1808	[0.55, 0.65]	<b>[0.90, 0.93]</b>	0.31	[0.58, 0.70]	<b>[0.85, 0.91]</b>	0.24	67.9 $\pm$ 2.4	11.2 $\pm$ 4.2	190
mnist	32, 16	3872	[0.74, 0.79]	<b>[0.92, 0.94]</b>	0.16	[0.73, 0.81]	<b>[0.84, 0.90]</b>	0.10	79.5 $\pm$ 1.0	14.0 $\pm$ 5.0	200
mnist	64, 32	8768	[0.84, 0.86]	<b>[0.93, 0.94]</b>	0.09	[0.85, 0.89]	[0.89, 0.92]	0.04	85.9 $\pm$ 0.5	17.5 $\pm$ 5.5	200
mnist	128, 64	21632	[0.86, 0.87]	<b>[0.92, 0.93]</b>	0.06	[0.92, 0.94]	[0.93, 0.95]	0.01	89.0 $\pm$ 0.4	20.3 $\pm$ 5.5	200
mnist	12, 24, 16	2032	[0.51, 0.63]	<b>[0.84, 0.90]</b>	0.30	[0.50, 0.63]	<b>[0.70, 0.80]</b>	0.19	69.4 $\pm$ 1.9	10.9 $\pm$ 4.1	194
mnist	40, 64, 32	8928	[0.77, 0.80]	<b>[0.91, 0.93]</b>	0.14	[0.71, 0.79]	<b>[0.80, 0.86]</b>	0.08	83.7 $\pm$ 0.7	16.1 $\pm$ 5.2	200
mnist	64, 128, 48	21216	[0.78, 0.81]	<b>[0.87, 0.89]</b>	0.08	[0.79, 0.84]	[0.81, 0.87]	0.03	87.0 $\pm$ 0.5	18.0 $\pm$ 5.7	200
mnist	12, 24, 16, 8	2080	[0.50, 0.62]	<b>[0.80, 0.87]</b>	0.27	[0.50, 0.64]	<b>[0.71, 0.81]</b>	0.19	61.3 $\pm$ 3.0	10.7 $\pm$ 4.4	191
mnist	32, 64, 44, 16	8928	[0.66, 0.73]	<b>[0.87, 0.90]</b>	0.19	[0.67, 0.76]	<b>[0.80, 0.86]</b>	0.11	79.4 $\pm$ 1.0	14.0 $\pm$ 5.0	200
mnist	64, 128, 48, 16	21664	[0.75, 0.80]	<b>[0.89, 0.91]</b>	0.12	[0.76, 0.83]	[0.82, 0.88]	0.06	83.2 $\pm$ 0.7	15.4 $\pm$ 5.2	200
fashion_mnist	8, 4	872	[0.07, 0.25]	[0.13, 0.36]	0.08	[0.14, 0.35]	[0.16, 0.39]	0.03	50.4 $\pm$ 4.9	10.5 $\pm$ 4.2	144
fashion_mnist	16, 8	1808	[-0.04, 0.11]	<b>[0.21, 0.38]</b>	0.26	[-0.06, 0.10]	<b>[0.20, 0.38]</b>	0.27	65.9 $\pm$ 2.1	13.0 $\pm$ 5.1	197
fashion_mnist	32, 16	3872	[0.03, 0.17]	<b>[0.30, 0.46]</b>	0.28	[0.04, 0.18]	<b>[0.32, 0.48]</b>	0.29	73.1 $\pm$ 0.9	15.0 $\pm$ 5.4	200
fashion_mnist	64, 32	8768	[0.09, 0.22]	<b>[0.26, 0.40]</b>	0.18	[0.09, 0.23]	<b>[0.30, 0.45]</b>	0.21	76.9 $\pm$ 0.6	17.4 $\pm$ 6.2	200
fashion_mnist	128, 64	21632	[0.20, 0.31]	[0.24, 0.37]	0.05	[0.24, 0.36]	[0.30, 0.43]	0.07	79.1 $\pm$ 0.5	18.3 $\pm$ 5.4	200
sincos	16	832	[0.18, 0.30]	<b>[0.60, 0.71]</b>	0.42	[0.25, 0.42]	<b>[0.61, 0.73]</b>	0.33	93.9 $\pm$ 0.9	17.5 $\pm$ 5.1	175
sincos	80	4160	[0.27, 0.40]	<b>[0.68, 0.75]</b>	0.38	[0.33, 0.47]	<b>[0.78, 0.85]</b>	0.41	97.0 $\pm$ 0.6	19.0 $\pm$ 7.2	200
sincos	192	9984	[0.34, 0.45]	<b>[0.66, 0.73]</b>	0.30	[0.45, 0.58]	<b>[0.80, 0.88]</b>	0.33	98.0 $\pm$ 0.5	20.2 $\pm$ 7.3	200
sincos	8, 4	456	[0.24, 0.39]	<b>[0.62, 0.75]</b>	0.37	[0.36, 0.58]	<b>[0.64, 0.78]</b>	0.24	83.3 $\pm$ 4.5	13.8 $\pm$ 4.7	125
sincos	16, 8	976	[0.23, 0.36]	<b>[0.67, 0.75]</b>	0.41	[0.31, 0.47]	<b>[0.69, 0.79]</b>	0.35	90.9 $\pm$ 1.8	15.7 $\pm$ 5.2	188
sincos	32, 16	2208	[0.25, 0.36]	<b>[0.63, 0.72]</b>	0.37	[0.30, 0.43]	<b>[0.64, 0.75]</b>	0.33	94.6 $\pm$ 1.1	19.5 $\pm$ 6.2	200
sincos	64, 32	5440	[0.36, 0.45]	<b>[0.71, 0.77]</b>	0.33	[0.45, 0.55]	<b>[0.77, 0.83]</b>	0.30	96.5 $\pm$ 0.7	22.0 $\pm$ 6.9	200
sincos	128, 64	14976	[0.46, 0.53]	<b>[0.73, 0.77]</b>	0.26	[0.58, 0.67]	<b>[0.82, 0.87]</b>	0.22	97.7 $\pm$ 0.5	23.9 $\pm$ 6.8	200
sincos	12, 24, 16	1312	[0.14, 0.28]	<b>[0.44, 0.58]</b>	0.30	[0.20, 0.37]	<b>[0.45, 0.60]</b>	0.24	92.4 $\pm$ 1.4	13.1 $\pm$ 4.9	187
sincos	40, 64, 32	6656	[0.05, 0.19]	<b>[0.39, 0.53]</b>	0.34	[0.07, 0.22]	<b>[0.42, 0.57]</b>	0.35	95.6 $\pm$ 0.8	15.2 $\pm$ 5.3	198
sincos	64, 128, 48	17600	[0.01, 0.16]	<b>[0.31, 0.44]</b>	0.29	[0.05, 0.20]	<b>[0.33, 0.47]</b>	0.28	96.8 $\pm$ 0.7	15.5 $\pm$ 5.4	200
sincos	12, 24, 16, 8	1408	[0.03, 0.20]	<b>[0.43, 0.59]</b>	0.40	[0.04, 0.23]	<b>[0.44, 0.60]</b>	0.39	89.9 $\pm$ 2.1	11.3 $\pm$ 4.7	180
sincos	32, 64, 44, 16	7168	[0.09, 0.24]	<b>[0.36, 0.50]</b>	0.27	[0.15, 0.30]	<b>[0.41, 0.55]</b>	0.26	94.2 $\pm$ 1.1	13.8 $\pm$ 5.4	199
sincos	64, 128, 48, 16	18240	[0.06, 0.21]	<b>[0.35, 0.49]</b>	0.28	[0.04, 0.19]	<b>[0.37, 0.51]</b>	0.33	95.6 $\pm$ 0.8	13.0 $\pm$ 5.3	198
uci_har	8, 4	4544	[0.68, 0.77]	[0.63, 0.74]	-0.04	[0.57, 0.68]	[0.51, 0.63]	-0.05	73.7 $\pm$ 6.5	16.1 $\pm$ 8.2	185
uci_har	16, 8	9152	[0.79, 0.84]	[0.73, 0.81]	-0.04	[0.59, 0.68]	[0.47, 0.59]	-0.10	88.4 $\pm$ 1.8	21.6 $\pm$ 10.2	200
uci_har	32, 16	18560	[0.87, 0.89]	[0.84, 0.88]	-0.02	<b>[0.69, 0.76]</b>	[0.59, 0.68]	-0.09	92.8 $\pm$ 0.9	25.0 $\pm$ 10.2	199
uci_har	64, 32	38144	[0.89, 0.90]	[0.90, 0.91]	0.01	[0.83, 0.87]	[0.78, 0.83]	-0.05	94.2 $\pm$ 0.7	25.3 $\pm$ 10.6	199
uci_har	128, 64	80384	[0.88, 0.89]	[0.89, 0.90]	0.01	[0.89, 0.92]	[0.86, 0.90]	-0.03	95.0 $\pm$ 0.6	25.8 $\pm$ 10.2	199
uci_har_raw	4, 16	4768	[0.57, 0.70]	[0.49, 0.64]	-0.07	[0.54, 0.66]	[0.46, 0.60]	-0.07	50.9 $\pm$ 2.6	17.5 $\pm$ 9.3	187
uci_har_raw	8, 16	9440	[0.81, 0.86]	[0.77, 0.83]	-0.03	[0.70, 0.78]	[0.65, 0.75]	-0.04	54.7 $\pm$ 2.4	20.4 $\pm$ 9.6	193
mnist_shuffled	16	1760	[0.80, 0.82]	[0.82, 0.84]	0.02	[0.61, 0.68]	[0.67, 0.74]	0.06	11.6 $\pm$ 4.7	14.0 $\pm$ 0.0	200
mnist_shuffled	32, 16	3872	[0.60, 0.66]	[0.65, 0.71]	0.05	[0.67, 0.75]	[0.67, 0.75]	0.00	11.2 $\pm$ 4.3	14.0 $\pm$ 0.0	200
mnist_shuffled	12, 24, 16	2032	[0.58, 0.62]	<b>[0.64, 0.68]</b>	0.06	[0.86, 0.90]	[0.83, 0.88]	-0.03	11.4 $\pm$ 3.6	14.0 $\pm$ 0.0	200
mnist_shuffled	40, 64, 32	8928	[0.29, 0.39]	<b>[0.44, 0.53]</b>	0.15	[0.51, 0.63]	[0.62, 0.73]	0.10	11.7 $\pm$ 3.4	14.0 $\pm$ 0.0	200
mnist_shuffled	32, 64, 44, 16	8928	[0.34, 0.44]	[0.40, 0.50]	0.06	[0.52, 0.64]	[0.53, 0.65]	0.01	11.0 $\pm$ 3.8	14.0 $\pm$ 0.0	200
random	16	1760	[0.81, 0.84]	[0.78, 0.82]	-0.02	[0.70, 0.78]	[0.73, 0.81]	0.03	10.3 $\pm$ 0.2	14.0 $\pm$ 0.0	200
random	32, 16	3872	[0.75, 0.80]	[0.74, 0.79]	-0.01	[0.48, 0.57]	[0.50, 0.61]	0.03	10.3 $\pm$ 0.3	14.0 $\pm$ 0.0	200
random	12, 24, 16	2032	[0.70, 0.73]	<b>[0.74, 0.77]</b>	0.04	[0.82, 0.86]	[0.79, 0.85]	-0.02	10.3 $\pm$ 0.3	14.0 $\pm$ 0.0	200
random	40, 64, 32	8928	[0.39, 0.49]	[0.44, 0.53]	0.04	[0.54, 0.66]	[0.48, 0.61]	-0.06	10.0 $\pm$ 0.3	14.0 $\pm$ 0.0	200
random	32, 64, 44, 16	8928	[0.39, 0.49]	[0.41, 0.50]	0.01	[0.48, 0.59]	[0.42, 0.54]	-0.05	10.3 $\pm$ 0.3	14.0 $\pm$ 0.0	200