

# Learning and Memorization

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# Motivation

- ▶ [Zhang et al. 2017]: neural networks have the capacity to memorize their training set
  - ▶ Train AlexNet on CIFAR-10 with randomly permuted labels
  - ▶ Training error goes to 0
- ▶ What is the link between **memorization** and **generalization**?
  - ▶ Why don't NNs just memorize their training set?
- ▶ [Chatterjee 2018]: Is it possible to generalize by memorizing alone?



# Basic idea of paper

- ▶ What is a simple form of memorization? → a table

Lives in water	Has eyes	Has limbs	Vertebrate
0	1	1	1
1	1	0	1
1	0	0	0

Model for classification of animals into vertebrates/invertebrates

- ▶ We must binarize the dataset
- ▶ We must limit the complexity
  - ▶ 28x28 images →  $28 \cdot 28 = 784$  →  $2^{784} \propto 10^{236}$



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Network of lookup tables ("luts")

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# Preprocessing data

- ▶ MNIST dataset: 28x28 images of handwritten digits (0-9)
- ▶ We unroll the images:  $28 \cdot 28 = 784$
- ▶ We scale the numerical values to the range  $[0, 1]$
- ▶ We binarize the data using the operator  $> 0.5$
- ▶ Labels (to be predicted): class 0-4 vs. 5-9
  
- ▶ We end up with
  - ▶ Features: matrix of shape  $(N, 784)$ , boolean entries
  - ▶ Labels: matrix of shape  $(N, 1)$ , boolean entries



# A single lut

- ▶ Reminder: every example is an instance of a “bit pattern” (e.g.  $x^i = 010$ ) and has a label (e.g.  $y^i = 1$ )
- ▶ For each bit pattern, we count how many times  $y = 0$  and  $y = 1$

$$\hat{f}(\text{bit pattern}) = \begin{cases} 0 & \text{if } \sum_{y=0} > \sum_{y=1} \\ 1 & \text{if } \sum_{y=0} < \sum_{y=1} \\ \text{rand}(0, 1) & \text{if } \sum_{y=0} = \sum_{y=1} \end{cases}$$



# A single lut: example

$x$	$y$	bit pattern	$\sum_{y=0}$	$\sum_{y=1}$	bit pattern	$\hat{f}$
000	0	000	1	2	000	1
000	1	001	0	1	001	1
000	1	010	0	0	010	0*
001	1	011	0	0	011	1*
100	0	100	1	0	100	0
100	0	101	0	0	101	1*
110	0	110	1	1	110	1*
110	1	111	0	0	111	0*



# A single lut applied on MNIST

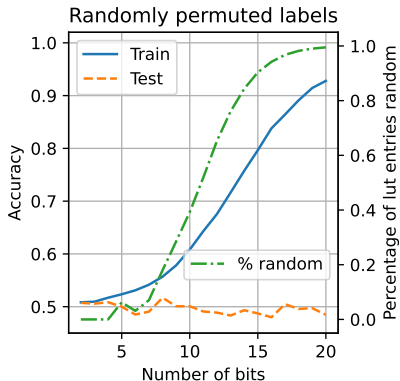
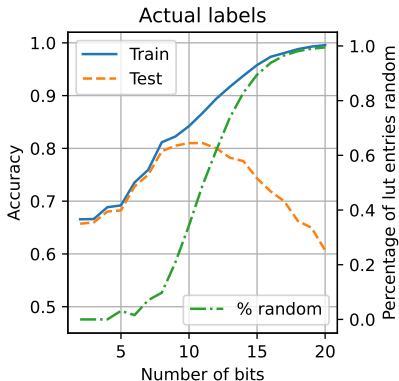
- ▶ MNIST features: matrix of shape  $(N, 784)$
- ▶ We perform PCA and obtain a matrix of shape  $(N, k)$ , varying  $k$  from 2 to 20
- ▶ A single lut is able to handle this dataset





# A single lut applied on MNIST

Performance of a single lut on 0-4 vs. 5-9 MNIST classification  
(PCA used to reduce dimensions to corresponding bit size)



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# Network

- ▶ As we've seen, a single lut is not very powerful



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# How to go from there



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# Recap

Hello there, this is empty :)





Any Questions?





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# References lala



Chatterjee, Satrajit (2018). “Learning and memorization”. In: *International Conference on Machine Learning*. PMLR, pp. 755–763.



Zhang, Chiyuan et al. (2017). *Understanding deep learning requires rethinking generalization*. arXiv: 1611.03530 [cs.LG].

