

# Deep Learning Service for Efficient Data Distribution Aware Sorting

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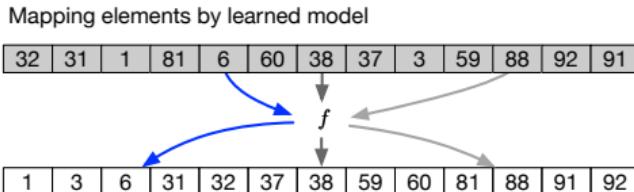
## Background

- ▶ Fundamental CS problem
- ▶ Traditional sorting algorithm: Quick Sort, Merge Sort, QuickX Sort ...
- ▶ ML-enhanced algorithms: Learned Data Structures & Algorithms, SageDB Sort ...
- ▶ Sorting is a well-studied topic, but we argue that leveraging ML models offers a way to further accelerate it.

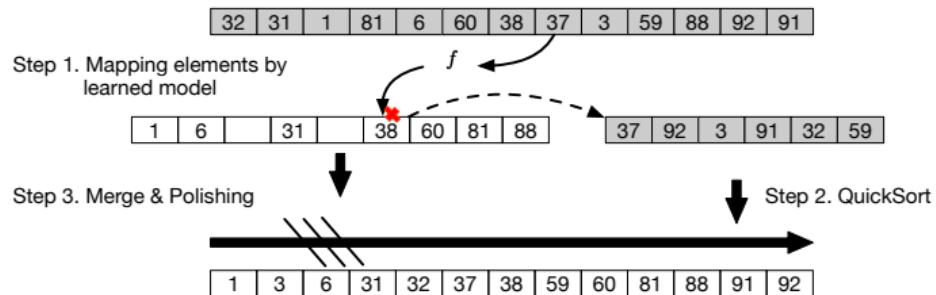
# Motivation

## Learned Sorting

- ▶ Train a model  $f$ .
- ▶ Use  $f$  to predict the final position of each key in the sorted output.



## SageDB Sort



## Limitations of SageDB Sort

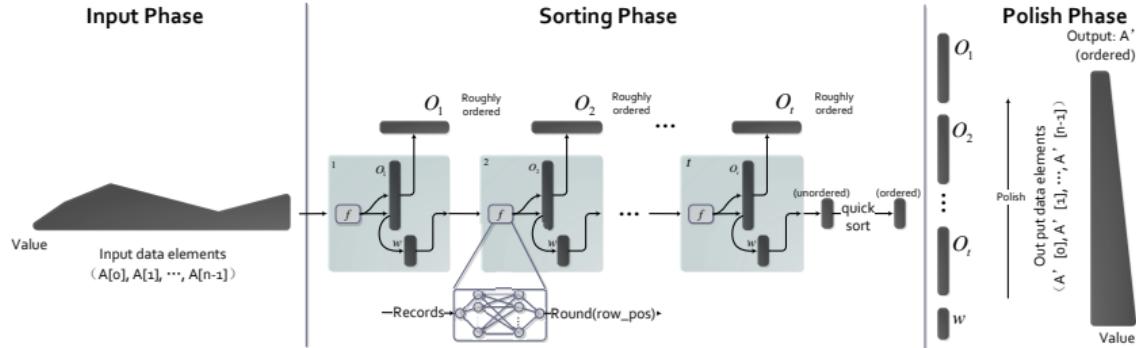
### A toy example

- ▶ QuickDraw game dataset Google Creative Lab
  - 50,426,265 records
  - schema: 'key-id', 'word', 'country code', 'timestamp', 'recognized'
- ▶ Near 10% elements are conflicts
- ▶ SageDB Sort addresses collisions using traditional sorting algorithms, which incurs computational overhead when collisions is too large

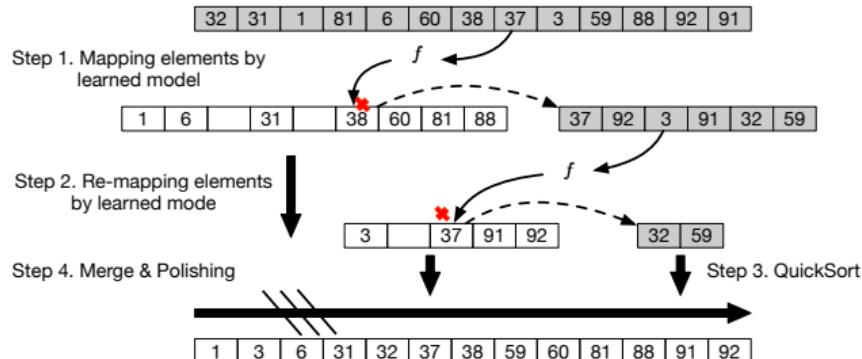
Algorithm name	Time (sec.)	Sorting Rate (elements/sec.)	Conflicting rate(%)
std::heap sort	13.46	3746.44	-
std::sort	23.71	2127.19	-
SageDB Sort	10.53	4790.125	9.16

# NN-sort

## Architecture



## Example



## The complexity

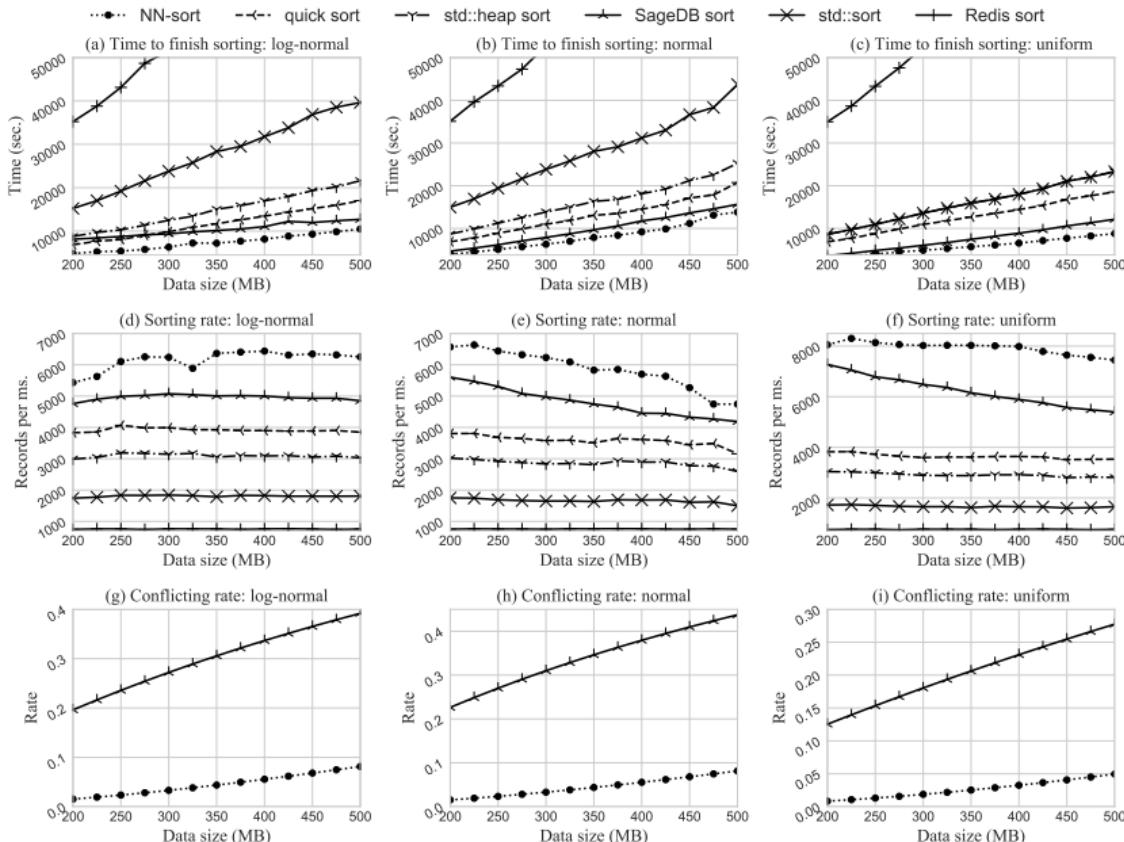
symbols	notations
$n$	the amount of data elements to be sorted
$\sigma_i$	collision rate per iteration
$e_i$	the number of data elements that were out-of-order in the $i$ -th iteration
$\epsilon$	the predefined limit of iterations
$t$	the number of completed iterations
$\theta$	The operations required for data to pass through $f$

$$T(n, e, \sigma, t, \theta) = \begin{cases} 1, & \text{if } n = 1 \\ C_1 n^2 + C_2 n \log n + C_3 n, & \text{if } n > 1 \end{cases}$$

$$C_1 = \left[ \frac{1}{2} \sum_{i=1}^t e_i (1 - \sigma_i) \left( \prod_{j=1}^{i-1} \sigma_j \right)^2 \right], C_2 = \prod_{j=1}^t \sigma_j$$

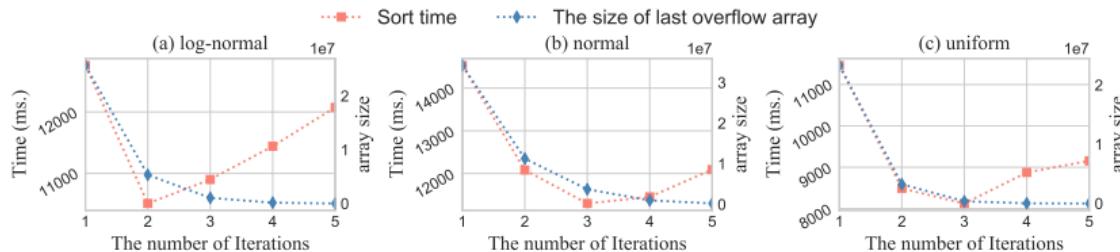
$$C_3 = \sum_{i=1}^t \left[ \theta \sum_{j=1}^i \sigma_j + (1 - e_i) (1 - \alpha_i) \prod_{j=1}^{i-1} \sigma_j + \prod_{j=1}^i \sigma_j \right] + \left( \prod_{j=1}^t \sigma_j \right) \log \left( \prod_{j=1}^t \sigma_j \right)$$

# Experimental Results: overall performance

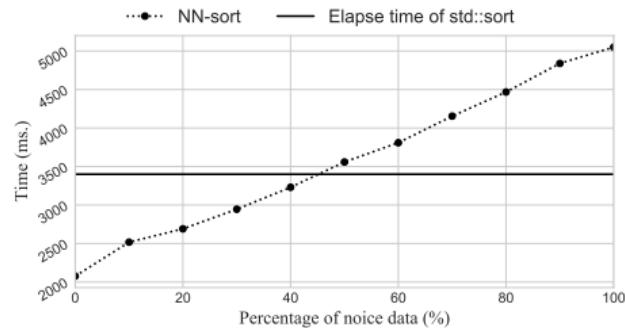


# Experimental Results: other results

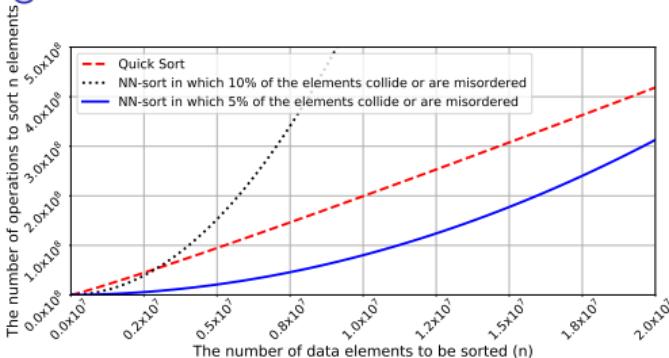
## Impact of Iterations



## Impact of data distribution



## Operations between traditional sorting algorithm and NN-sort



Thanks