

A IoT-benchmark Data Generation Algorithm

Algorithm 1 describes the data generation strategy. For parameter details, please refer to Table 1.

Table 1: Parameters Description of Data Generator

Notation	Data Features
μ_v	Mean of values
μ_d	Mean of deltas
θ_d	Variance of deltas
γ	Repeat Rate
η	Increase rate
l	Series length

Algorithm 1: Numerical data generator [45]

Input: $\mu_v, \mu_d, \theta_d, \gamma, \eta$, length n
Output: TS
 $DS := \text{empty_list}();$
while $|DS| < n$ **do**
 $isRepeat := \text{random_index}(\gamma);$
 if $isRepeat$ **then**
 else
 $repeat_len := \text{random}(8, T);$
 $DS.append(0, repeat_len);$
 end
 $isPositive := \text{random_index}(\eta);$
 $delta := 0;$
 if $isPositive$ **then**
 else
 while $delta \leq 0$ **do**
 $delta := \text{random_gauss}(\mu_d, \theta_d);$
 end
 end
 while $delta \geq 0$ **do**
 $delta := \text{random_gauss}(\mu_d, \theta_d);$
 end
 $DS.append(delta);$
 end
 $TS := \text{prefix_sum}(DS);$
 $TS.zoom(\mu_v);$
return $TS;$

B Proof of Proposition 4.11

The proof of Proposition 4.11 is as follows.

PROOF. Let $Cost(I_u(c))$ denote the cost of the traditional query process of a compressed database, i.e. decompressing first, then restore and query on uncompressed data, where

$$\begin{aligned}
 Cost(I_u(c)) &= Cost((Q_u \circ R_u \circ U)(c)) = Cost(Q_u(R_u(U(c)))) \\
 &= Cost\left(\left\{u : \begin{array}{l} u_1 \leftarrow op_1(u_0), u_2 \leftarrow op_2(u_1), \\ \dots, u \leftarrow op_n(u_{n-1}) \end{array} \mid \begin{array}{l} i \in \{1, \dots, n\} \\ op_i \in \Pi \end{array} \right\}\right) \\
 &\quad + Cost(R_u(u)) + Cost(U(c)) \\
 &= Cost(U(c)) + Cost(R_u(u)) + Cost(\{op_1, op_2, \dots, op_{n-1}\}(u_1)).
 \end{aligned}$$

Let $Cost(I_c(c))$ denote the cost of the partial homomorphic query process of a compressed database, where

$$\begin{aligned}
 Cost(I_c(c)) &= Cost((I'_u \circ Q_c \circ R_c)(c)) = Cost(I'_u(Q_c(R_c(c)))) \\
 &= Cost(I'_u(c_j)) + Cost(Q_c(c_0)) + Cost(R_c(c)) \\
 &= Cost\left(\left\{u : \begin{array}{l} u_j \leftarrow op_j(u_{j-1}), u_{j+1} \leftarrow op_{j+1}(u_j), \\ \dots, u \leftarrow op_n(u_{n-1}) \end{array} \mid \begin{array}{l} i \in \{j, \dots, n\} \\ op_i \in \Pi \end{array} \right\}\right) \\
 &\quad + Cost\left(\left\{c_j : \begin{array}{l} c_1 \leftarrow op'_1(c_0), c_2 \leftarrow op'_2(c_1), \\ \dots, c_j \leftarrow op'_j(c_{j-1}) \end{array} \mid \begin{array}{l} i \in \{1, \dots, j\} \\ op'_i \in \Theta \end{array} \right\}\right) \\
 &\quad + Cost(U(c_j)) + Cost(R_c(c)) \\
 &= Cost(R_c(c)) + Cost(\{op'_1, op'_2, \dots, op'_j\}(c_0)) + Cost(U(c_j)) \\
 &\quad + Cost(\{op_{j+1}, op_{j+2}, \dots, op_{n-1}\}(u_{j-1})),
 \end{aligned}$$

with $\varphi(u_i) = c_i, j \in \{1 \dots n-1\}$. Then we have
 $Cost(I_u(c)) - Cost(I_c(c))$

$$\begin{aligned}
 &= Cost(U(c)) + Cost(R_u(u)) + Cost(\{op_1, op_2, \dots, op_{n-1}\}(u_0)) \\
 &\quad - (Cost(R_c(c)) + Cost(\{op'_1, op'_2, \dots, op'_j\}(c_0)) + Cost(U(c_j)) \\
 &\quad + Cost(\{op_{j+1}, op_{j+2}, \dots, op_{n-1}\}(u_{j-1}))) \\
 &= (Cost(U(c)) - Cost(U(c_j))) + (Cost(R_u(u)) - Cost(R_c(c))) \\
 &\quad + (Cost(\{op_1, op_2, \dots, op_{n-1}\}(u_0)) - Cost(\{op'_1, op'_2, \dots, op'_j\}(c_0)) \\
 &\quad - Cost(\{op_{j+1}, op_{j+2}, \dots, op_{n-1}\}(u_{j-1}))) \\
 &= (Cost(U(c)) - Cost(U(c_j))) + (Cost(R_u(u)) - Cost(R_c(c))) \\
 &\quad + (Cost(\{op_1, op_2, \dots, op_j\}(u_0)) - Cost(\{op'_1, op'_2, \dots, op'_j\}(c_0)))
 \end{aligned}$$

Note that when $j = n-1$, query Q is a fully homomorphic query.
 By Lemma 4.10, $Size(c) \geq Size(c_j)$, thus, $Cost(U(c)) \geq Cost(U(c_j))$.
 And by Definition 4.9, we have

$$Cost(\{op_1, op_2, \dots, op_j\}(u_0)) \geq Cost(\{op'_1, op'_2, \dots, op'_j\}(c_0)).$$

And by Definition 4.8, we have

$$Cost(R_u(u)) \geq Cost(R_c(c))$$

Thus, we have $Cost(I_u(c)) - Cost(I_c(c)) \geq 0$, i.e.,

$$Cost(I_c(c)) \geq Cost(I_u(c)). \quad \square$$