Synthesizing Third Normal Form Schemata that Minimize Integrity Maintenance and Update Overheads

Parameterizing 3NF by the Numbers of Minimal Keys and Functional Dependencies

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ABSTRACT

State-of-the-art relational schema design generates a lossless, dependency-preserving decomposition into Third Normal Form (3NF), that is in Boyce-Codd Normal Form (BCNF) whenever possible. In particular, dependency-preservation ensures that data integrity can be maintained on individual relation schemata without having to join them, but may need to tolerate a priori unbounded levels of data redundancy and integrity faults. As our main contribution we parameterize 3NF schemata by the numbers of minimal keys and non-key functional dependencies they exhibit. Conceptually, these parameters quantify, already at schema design time, the effort necessary to maintain data integrity, and allow us to break ties between 3NF schemata. Computationally, the parameters enable us to optimize normalization into 3NF according to different strategies. Operationally, we show through experiments that our optimizations translate from the logical level into significantly smaller update overheads during integrity maintenance. Hence, our framework provides access to parameters that guide the computation of logical schema designs which reduce operational overheads.

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The source code, data, and/or other artifacts have been made available at https://github.com/zzxhelloworld/iCONF.

1 INTRODUCTION

We will revisit classical normalization in the relational model of data, in particular Third and Boyce-Codd Normal Form (3NF, BCNF) that are based on functional dependencies (FDs) [7, 12, 18]. These topics are fundamental and taught in introductory database courses [1, 20, 26]. We will not discuss higher normal forms [9, 21, 32].

Arguably, 3NF is the most popular normal form in database practice. While BCNF guarantees that no relation can ever exhibit any

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redundant data value [7], a lossless, dependency-preserving decomposition into BCNF is not always achievable [6]. In contrast, a lossless, dependency-preserving decomposition into 3NF is always possible [7, 8, 36], but may need to tolerate unbounded levels of data redundancy and potential integrity faults [7, 16, 23]. Without dependency-preservation, maintaining data integrity is regarded as prohibitively expensive since relation schemata need to be joined before FDs can be validated [8, 23]. The use of 3NF is further promoted as it admits the fewest sources of data redundancy among all lossless, dependency-preserving decompositions [17].

State-of-the-art normalization computes a lossless, dependency-preserving decomposition into 3NF that is in BCNF whenever possible [28]. Intuitively, the algorithm checks for every critical schema (a relation schema that is in 3NF but not in BCNF) whether it is redundant. If any critical schema is non-redundant, no lossless, dependency-preserving decomposition can be in BCNF. This algorithm was optimized by breaking ties between non-critical schemata (relation schemata in BCNF) [37]. In fact, BCNF was parameterized by the number k of minimal keys the schema exhibits. Intuitively, smaller k lead to less update complexity but also speed up of fewer queries, since k represents the number of UNIQUE indices that require maintenance but facilitate query optimization. Computationally, k is used to break ties between redundant BCNF schemata. By retaining schemata with fewer minimal keys, updates can be processed faster on the resulting database designs [37].

However, we observe that not even these recent advances address the arbitrary choice between redundant critical schemata. Consequently, there are currently no systematic means to break ties between 3NF schemata. This fact is particularly important since FDs that are not implied by keys (non-key FDs) cause the biggest bottleneck for maintaining data integrity: their validation is orders of magnitude slower than that of minimal keys. Hence, biggest performance improvements are to be expected from speeding up the maintenance of non-key FDs.

Despite 3NF being one of the most fundamental topics in data-base education and practice, this opportunity has not been observed nor addressed yet in close to 50 years of research. The main technical challenges are to find parameters that adequately measure operational update overheads at the logical level, and formalizing a rigorous normalization framework that provides effective access to these parameters. As a remarkably intuitive solution, we will utilize Maier's seminal notions of minimal-reduced and optimal covers [25]. Advancing the rich knowledge on relational databases will impact modern data models, such as incomplete [19, 34], temporal [14], Web [3, 10, 35], uncertain [22] and graph data [2, 31].

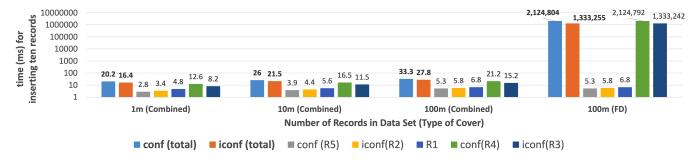


Figure 1: Update Overhead (ms) for Insertions of 10 Records on 1, 10 and 100 Million Records over *iConf*-decomposition $\mathcal{D}_1 = \{R_1, R_2, R_3\}$ and Conf-decomposition $\mathcal{D}_2 = \{R_1, R_4, R_5\}$

Our main idea is to measure the operational effort of FD maintenance already at the logical level, namely by using the magnitude required for representing the FDs. For that purpose, we separate an FD set into the set $\mathcal K$ of minimal keys it implies, and a minimal-reduced cover $\mathcal F$ [25] for the set of non-key FDs it implies. Consequently, the number k of elements in $\mathcal K$ measures how much maintenance can be shifted into minimal keys, while the number f of elements in $\mathcal F$ measures how much effort needs to be spent on maintaining non-key FDs. As a result, we can compare 3NF schemata based on k and f. There are different strategies to do that, such as prioritizing the schema with smaller f, and if a tie still persists prioritizing the schema with smaller k. Next we introduce our running example and illustrate these ideas.

Example 1.1. As running example we consider the schema $R = \{E(vent), M(anager), S(tatus), V(enue), T(ime)\}$ which records the name of events, their managers, status, venue and time they are held. The set Σ of FDs for R consists of: $VSE \to T$, $SET \to V$, $SME \to V$, $VS \to M$, $SME \to T$, $MT \to E$, and $ET \to M$. Our new normalization algorithm, called iConf, can optimize lossless, dependency-preserving decompositions into 3NF by different strategies, in this case by minimizing the number of non-key FDs. It returns the following decomposition \mathcal{D}_1 of (R, Σ) :

- $R_1 = ESTV$ and Σ_1 with 3 minimal keys EST, ESV, and STV
- $R_2 = EMT$ and Σ_2 with 2 minimal keys ET and MT
- $R_3 = EMSV$ and Σ_3 with non-key FD $VS \rightarrow M$ and 2 minimal keys ESV and EMS.

The previously best algorithm, called *Conf*, does not optimize non-key FDs. It returns the decomposition \mathcal{D}_2 of (R, Σ) :

- $R_1 = ESTV$ and Σ_1 with 3 minimal keys EST, ESV, and STV
- $R_4 = EMST$ and Σ_4 with two non-key FDs $MT \to E$, $ET \to M$ and three minimal keys EST, EMS, and MST
- $R_5 = MSV$ and Σ_5 with 1 minimal key VS.

Both decompositions contain R_1 , while \mathcal{D}_1 has BCNF-schema R_2 with 2 minimal keys, \mathcal{D}_2 has BCNF-schema R_5 with just 1 minimal key, and while \mathcal{D}_1 has 3NF-schema R_3 with 1 non-key FD, \mathcal{D}_2 has 3NF-schema R_4 with 2 non-key FDs. \mathcal{D}_1 is preferred over \mathcal{D}_2 if schemata with fewer non-key FDs are favored, while \mathcal{D}_2 is preferred over \mathcal{D}_1 if schemata with more minimal keys are targeted.

The parameters enable us to break ties between redundant 3NF schemata, based on the target strategy we provide as input to normalization algorithms. Indeed, we utilize this ability to compute

lossless, dependency-preserving decompositions into 3NF that are optimized for the strategy we target. Critically, these optimizations transcend from the logical schema level to the operational level where update overheads are minimized. For our running example, Figure 1 shows a study we conducted to measure the update overhead incurred by the decompositions in Example 1.1. As underlying data sets we used 1, 10, and 100 million records of synthetic data, respectively, each of which satisfies the given set of FDs but violates every FD not implied by the set. Hence, the data set is a perfect representation of the given constraint set. These data sets are then projected onto each element of the decompositions \mathcal{D}_1 and \mathcal{D}_2 , obtained by applying Conf (iConf). The blue (orange) bars show total times taken to insert 10 records into the projected data sets of \mathcal{D}_1 and \mathcal{D}_2 , which measures the effort of maintaining integrity used by outputs of the algorithms. The remaining bars break down these total times on relation schemata of the two decompositions (both share the schema R_1 , so it is only reported once). All times reported are averaged over 30 runs. Bars on the right group use an FD cover that enforces all constraints (non-key FDs and minimal keys) uniformly as FDs using triggers, while a combined cover of non-key FDs and minimal keys (and their UNIQUE indices) is used elsewhere. There are two main observations: (1) Combined covers facilitate integrity maintenance that is orders of magnitude faster than FD covers. (2) Update overheads on \mathcal{D}_1 , achieved by our new normalization algorithm iConf, are significantly smaller than those on \mathcal{D}_2 , resulting from the previously best algorithm *Conf*, and this is true across all sizes of the data set and both types of covers. This provides quantitative evidence that our normalization strategies at the logical level result in reductions of update overheads at the operational level. Record deletions cannot result in violations of FDs, so updates are a sequence of record deletion and insertion. Our contributions can be summarized as follows.

- (1) Fundamentally, we introduce parameters that quantify the effort of maintaining data integrity for FDs, making it possible to break ties between 3NF schemata. For that purpose, we parameterize classical relational normalization by separating non-key FDs from minimal keys in minimal-reduced and optimal covers [25].
- (2) Algorithmically, our framework enables us to optimize lossless, dependency-preserving decompositions into 3NF according to targets set for our parameters. Essentially, this improves state-ofthe-art since the target strategy declares how ties between redundant 3NF schemata are broken. Experiments with real and synthetic

FD sets showcase the quality advancement of schema designs returned by our algorithms over those in previous work.

(3) Operationally, we evaluate how much update overhead is reduced by our optimizations. We reveal which strategy works best in minimizing update overheads, illustrating performance gains over previous work, and how well optimizations at logical level transcend to integrity maintenance at operational level.

Overall, we establish the first parameterized framework for relational schema design with FDs, intrinsically linking schema optimizations to performance gains at operational level.

Organization. We summarize previous work in Section 2 before introducing a parameterized schema design based on the set of minimal keys and covers of non-key FDs in Section 3. Section 4 develops algorithms for parameterized 3NF normalization, while Section 5 presents and analyzes the results of our experiments. We conclude in Section 6 where we also comment on future work. Artifacts are available at https://github.com/zzxhelloworld/iCONF.

2 FUNDAMENTALS AND PREVIOUS WORK

We summarize the background necessary for our framework, starting with concepts from relational databases design [1, 20, 26].

FDs, Normal Forms, and Normalization. A *relation schema* is a finite set R of attributes A that have an associated domain dom(A) that contains the set of possible values for the attribute. A *relation* over R is a finite subset r of tuples from the Cartesian product $\prod_{A \in R} dom(A)$. For a tuple t and subset $S \subseteq R$ we use t[S] to denote the projection of t onto S. Intuitively, relations formalize tables of records over column names.

A functional dependency (FD) is an expression $X \to Y$ with attribute sets $X, Y \subseteq R$. A relation satisfies $X \to Y$ if every pair of records with values matching on all the attributes in X have also values matching on all the attributes in Y. An FD $X \to Y$ is trivial if $Y \subseteq X$. Example 1.1 showcases several FDs.

For an FD set $\Sigma \cup \{\varphi\}$, Σ *implies* φ if every relation that satisfies all FDs in Σ also satisfies φ . The *semantic closure* of Σ is the set Σ^* that contains all FDs implied by Σ , which equals the *syntactic closure* Σ^+ of Σ that contains the FDs that can be inferred from Σ by using an axiomatization such as Armstrong's axioms [4]. FD sets Σ and Θ are *covers* of one another if $\Sigma^+ = \Theta^+$.

For an FD set Σ over $R, X \to Y \in \Sigma^+$ is *minimal* if there is no proper subset $Z \subseteq X$ such that $Z \to Y \in \Sigma^+$ holds. X is a *key* of R, if the FD $X \to R$ is implied by Σ . In this case, the FD $X \to R$ is also called a *key dependency*. An FD $X \to Y$ is *non-key* if $X \to R$ is not implied by Σ . A key X of R is *minimal* if $X \to R$ is a minimal key dependency, that is, there is no proper subset $Y \subset X$ that is also a key of R. We use \mathcal{K}_{Σ} to denote the set of all minimal keys implied by Σ . An attribute $A \in R$ is said to be *prime* for Σ when it is contained in some minimal key of R. Example 1.1 features (minimal) keys.

FDs encode business rules of the underlying domain, but may also cause data value occurrences that are redundant. Indeed, for a tuple t of a relation r that satisfies an FD set Σ , the data value occurrence v=t[A] is redundant whenever every change of v to a different value $v'\neq v$ results in a relation that violates some FD in Σ . For a non-trivial FD there is some relation with a redundant data value occurrence if and only if it is not a key dependency. For an FD set Σ over relation schema R, (R, Σ) is in Boyce-Codd Normal Form

(BCNF) if and only if for every non-trivial FD $X \to A \in \Sigma^+$, X is a key of R. Hence, BCNF characterizes the absence of redundant data values caused by FDs. Data redundancy causes update inefficiency as updates to data values that occur redundantly need be applied to every redundant occurrence. If such values are not updated consistently, integrity faults will occur as violations of FDs.

A decomposition of R is a set $\mathcal{D} \subseteq 2^R$ such that $\bigcup_{S \in \mathcal{D}} S = R$. The decomposition \mathcal{D} is lossless if for every relation r over R that satisfies Σ , it is true that $r = \bowtie_{S \in \mathcal{D}} r[S]$, that is, r is the lossless join over its projections $r[S] = \{t[S] \mid t \in r\}$. \mathcal{D} is dependency-preserving if and only if $\Sigma^+ = (\bigcup_{S \in \mathcal{D}} \Sigma[S])^+$ where $\Sigma[S] = \{X \to Y \mid X \cup Y \subseteq S \land X \to Y \in \Sigma^+\}$. A decomposition \mathcal{D} is in BCNF (3NF, see next paragraph) if for every $S \in \mathcal{D}$, $(S, \Sigma[S])$ is in BCNF (3NF). Indeed, if a decomposition is not dependency-preserving, integrity of some FD can only be validated on the join of some relations, which is prohibitively expensive. While lossless decompositions into BCNF always exist, there are some (R, Σ) for which no lossless, dependency-preserving decomposition into BCNF exists.

For this reason, Third Normal Form was introduced. We say that (R, Σ) is in *Third Normal Form* (3NF) if and only if for every non-trivial FD $X \to A \in \Sigma^+$, X is a key of R or A is prime. Indeed, lossless, dependency-preserving decompositions into 3NF are always possible. However, unless it is a BCNF decomposition, data redundancy needs to be tolerated. Indeed, 3NF cannot guarantee any bounded level of data redundancy, which is achievable with cardinality constraints [23], but beyond our scope. Classical algorithms [28] compute for input (R, Σ) a lossless, dependency-preserving decomposition into 3NF that is in BCNF whenever possible. 3NF was shown to guarantee the fewest sources of data redundancy among all lossless, dependency-preserving decompositions [17]. However, they did not aim at breaking ties between 3NF schemata.

Complexity Results. It is important to highlight lower complexity bounds associated with computational problems in normalization. For example, the problem of deciding if a given schema (R, Σ) satisfies 3NF is *NP*-complete, as is the problem of deciding if a given attribute $A \in R$ is prime for (R, Σ) [5].

The problem of finding the number of minimal keys for (R, Σ) is #P-complete [13], but there is an algorithm for computing the set of minimal keys in time linear in the output [24]. While the number of minimal keys can be exponential in the number of given FDs, this case occurs rarely in practice, so the set of minimal keys can often be computed efficiently [24].

Parameterizing BCNF. Recent work has parameterized BCNF by the number k of minimal keys [37]. Here, k represents a measure of both update and query complexity. The larger k, the more UNIQUE indices need to be maintained during updates but the more queries may benefit from these indices.

Fundamentally, the set of k minimal keys forms a composite object of level k, characterized by k-uniqueness and k-dependence. A schema (R, Σ) is in *Composite Object Normal Form* (CONF) of level k (k-CONF) whenever every minimal left-hand side of an FD in Σ^+ is an element of a composite object. While BCNF means that every constraint can be enforced by minimal keys, k-CONF means that every constraint can be enforced by k minimal keys. Hence, the parameter k leads to a hierarchy of BCNF schemata in which ties between them may be broken by k [37].

Computationally, the best classical algorithm [28] was optimized by eliminating redundant BCNF schemata with larger (smaller, respectively) numbers of minimal keys first, with the aim of minimizing update complexity (maximizing query efficiency, respectively) on those schemata of the decomposition that are in BCNF [37].

However, non-key FDs, which cause the biggest bottleneck for integrity maintenance, have not been considered. This is the contribution of our current work. Indeed, we will parameterize 3NF using k and the number f of FDs in a suitable cover, resulting in (k, f)-3NF, where the special case of f = 0 captures k-CONF.

3 FOUNDATIONS FOR PARAMETERIZED 3NF

We will equip the classical 3NF framework with access to parameters that can optimize the output of normalization. As a byproduct, the difference between 3NF and BCNF becomes quantifiable, too. In particular, we will characterize the classical 3NF definition by the effort necessary to maintain minimal keys and non-key FDs during updates. Isolating non-key FDs enables us to minimize their overhead during normalization. In turn, this minimization at the logical level will reduce overheads for integrity maintenance at the operational level.

The intuition of this section can be summarized as follows. The classical definition of 3NF is naturally tied to the set of minimal keys as the left-hand side X of each non-trivial FD $X \rightarrow A$ either needs to be a superset of some minimal key, or the RHS A needs to be an element of some minimal key (that is, A is prime). Hence, the set of minimal keys is useful for at least two reasons: 1) Each of them gives rise to a UNIQUE index, and 2) the set enables us to isolate those FDs that cannot be enforced by keys. Separating an FD input into the set K of minimal keys it implies, and a suitable cover \mathcal{F} for the set of non-key FDs will formally lead us to defining the 3NF-core of the input FD set. If that core forms a cover for the input FD set, then it is in 3NF. Based on the cardinalities k of Kand f of \mathcal{F} we may then compare schemata that are in 3NF. In particular, if f = 0, then the schema will be in BCNF, or more precisely in k-CONF. Hence, f quantifies the overhead of a schema in (k, f)-3NF over that in k-CONF.

3.1 Intransitive Composite Objects

In the special case of BCNF, the set \mathcal{K} forms a composite object for some k. Hence, BCNF schemata only require k minimal keys for integrity maintenance during updates. In the general case of 3NF, non-key FDs are also required, but only for RHS attributes that are prime. Hence, we will generalize composite objects to intransitive composite objects, which we define as 3NF-substructures sufficient for maintaining the integrity of their input FD set under updates.

Formally, let (R, Σ) denote a relation schema R with a set Σ of FDs over R. Let Ω denote a set of keys and FDs over R such that

$$\Omega \subseteq \{X \subseteq R \mid X \to R \in \Sigma^+\} \cup \{X \to Y \in \Sigma^+ \mid (X \to R \notin \Sigma^+) \land (Y - X \subseteq P_{\Sigma})\}$$

where

$$P_{\Sigma} = \{A \in R \mid \exists K \to R \in \Sigma^{+} \land \forall K' \subset K(K' \to R \notin \Sigma^{+}) \land A \in K\}$$

denotes the set of prime attributes for Σ . We call Ω a *3NF-substructure* of (R, Σ) . While 3NF-substructures meet the requirements of 3NF, they may not enforce all FDs of the input set. This additional feature is special and defined as follows.

Definition 3.1 (intransitive composite object). Let (R, Σ) denote a relation schema R with a set Σ of FDs over R. Let Ω denote a 3NF-substructure of (R, Σ) . We call Ω an intransitive composite object for Σ if and only if the following holds:

• (3NF update completeness) For all relations r over R that satisfy Σ , for all $t \in dom(R)$, if $r \cup \{t\}$ satisfies Ω , then $r \cup \{t\}$ satisfies Σ .

Hence, integrity for Σ is retained when all FDs in an intransitive composite object for Σ are valid. Next, we illustrate the definitions.

Example 3.2. Consider $R = \{E, M, S, T\}$ and $\Sigma = \{ET \rightarrow MS, M \rightarrow E\}$. The set of minimal keys is $\{ET, MT\}$, P = EMT, and the only non-prime attribute is R - P = S. Hence, (R, Σ) is in 3NF. However, $\Omega' = \{ET, MT, MS \rightarrow E\}$ is not an intransitive composite object for Σ (essentially because the FD $M \rightarrow E$ is not implied by Ω'). However, $\Omega = \Omega' \cup \{M \rightarrow E\}$ is an intransitive composite object for Σ. Indeed, every relation that satisfies ET must also satisfy $ET \rightarrow MS$. However, $\{ET, MT\}$ is not a composite object. This can be observed on the following records.

	Event	Time	Manager	Status
t':	Workshop	21/11/2024	Sophie	approved
t:	Symposium	19/12/2025	Sophie	approved

Indeed, $r = \{t'\}$ satisfies Σ , and $r \cup \{t\}$ satisfies ET and MT, but not $M \to E$. Hence, $\{ET, MT\}$ is not a composite object. \square

3.2 Intransitive CONF

Intransitive composite objects are not unique. In fact, 3NF-substructures may contain non-minimal keys or non-minimal FDs. Similar to BCNF where the unique composite object is given by the set of minimal keys, we will now define a unique 3NF-substructure by isolating the set of minimal keys for input FD set Σ and the set of all non-key FDs $X \to A \in \Sigma^+$ where the RHS A is prime and X is minimal such that no $Y \subset X$ exists where $Y \to A \in \Sigma^+$ holds.

Definition 3.3. (3NF-core) For an FD set Σ over relation schema R, we use $\mathcal{K}_{\Sigma} = \{X \to R \in \Sigma^+ \mid \forall Z \subset X(Z \to R \notin \Sigma^+)\}$ to denote the set of minimal keys implied by Σ , and

$$\begin{array}{ll} \mathcal{F}_{\Sigma} & = & \{Z \rightarrow A \in \Sigma^{+} \mid & (Z \rightarrow R \notin \Sigma^{+}) \wedge (A \in P_{\Sigma} - Z) \wedge \\ & (\forall Y \subset Z(Y \rightarrow A \notin \Sigma^{+}))\} \end{array}$$

to denote the set of non-key minimal FDs with RHS prime attribute implied by Σ . We call $(\mathcal{K}_{\Sigma} \cup \mathcal{F}_{\Sigma}$ the *3NF-core* of Σ . We omit indices and write \mathcal{K} and \mathcal{F} when Σ is fixed.

The following continues our previous example by removing a non-minimal non-key FD.

Example 3.4. For $R = \{E, M, S, T\}$ and $\Sigma = \{ET \to MS, M \to E\}$ from Example 3.2, $\Omega_c = \mathcal{K} \cup \mathcal{F}$ forms the 3NF-core of Σ where $\mathcal{K} = \{ET, MT\}$ and $\mathcal{F} = \{M \to E\}$.

The 3NF-core is unique and provides access to parameters that can optimize classical normalization algorithms such as 3NF synthesis. We will show soon how to minimize the representation of the 3NF-core further. For now, however, we will generalize a recent characterization of BCNF by CONF [37] to a characterization of 3NF by 3NF-cores. We have already done all the work since 3NF-cores only need to satisfy 3NF update completeness to capture 3NF.

Definition 3.5. (intransitive composite object normal form) Let Σ denote an FD set over relation schema R. Then (R, Σ) is in *intransitive Composite Object Normal Form (iCONF)* if and only if the 3NF-core of Σ is an intransitive composite object for Σ .

Definition 3.5 is independent of how Σ is represented. In fact, for every cover Θ of Σ , (R, Σ) is in iCONF iff (R, Θ) is in iCONF. We will now illustrate the definition of iCONF on our running example.

Example 3.6. For $R = \{E, M, S, T\}$ and $\Sigma = \{ET \to MS, M \to E\}$ from Example 3.4, the 3NF-core Ω_c satisfies 3NF update completeness, so (R, Σ) is in iCONF.

3.3 3NF and iCONF

We will now establish the equivalence between 3NF and iCONF. The main idea is realizing that a 3NF-substructure Ω is 3NF update complete if and only if the input FD set is in 3NF and covered by Ω .

LEMMA 3.7. (Main Lemma)

Let (R, Σ) denote a set Σ of FDs over relation schema R. Let Ω denote a 3NF-substructure of (R, Σ) . Then Ω is 3NF update complete if and only if all of the following hold:

- (1) $\Sigma \subseteq \Omega^+$
- (2) (R, Σ) is in Third Normal Form.

We now conclude that the 3NF-core of a schema in 3NF must necessarily be an intransitive composite object.

COROLLARY 3.8. If (R, Σ) is in 3NF, then the 3NF-core $\mathcal{K}_{\Sigma} \cup \mathcal{F}_{\Sigma}$ is an intransitive composite object for (R, Σ) .

As targeted we can characterize 3NF by iCONF.

THEOREM 3.9. For all relation schemata R and sets Σ of FDs over R, (R, Σ) is in 3NF if and only if (R, Σ) is in iCONF.

State-of-the-art results from the literature [37] now emerge as special cases of our new framework. In particular, k-CONF is the special case of iCONF where the input FD set is covered by the set of k minimal keys.

COROLLARY 3.10. Let (R, Σ) be in 3NF. Then the following statements are equivalent:

- (1) The 3NF-core of Σ over R is covered by \mathcal{K}_{Σ}
- (2) (R, Σ) is in BCNF with $|\mathcal{K}_{\Sigma}|$ minimal keys
- (3) (R, Σ) is in CONF of level $|\mathcal{K}_{\Sigma}|$.

Next we illustrate how to break ties between different 3NF schemata, which was not considered in previous work.

Example 3.11. Consider the following two schemata that belong to the decompositions of (R, Σ) from Example 1.1:

- $R_3 = EMSV$ and Σ_3 with non-key FD $VS \to M$ and two minimal keys ESV and EMS
- $R_4 = EMST$ and Σ_4 with two non-key FDs $MT \rightarrow E$, $ET \rightarrow M$ and three minimal keys EST, EMS, and MST.

Either one of (R_3, Σ_3) and (R_4, Σ_4) is redundant, but not both, so we can choose which one to return as output. Σ_3 contains one minimal non-key FD and two minimal keys, while Σ_4 contains two minimal non-key FDs and three minimal keys. If we prefer to have fewer FDs, we may pick (R_3, Σ_3) over (R_4, Σ_4) , but if we prefer to have more minimal keys, then we may pick (R_4, Σ_4) over (R_3, Σ_3) .

The last example illustrates how classical normalization can be optimized by using parameters, such as the numbers of non-key FDs or minimal keys, or even a combination of them to break further ties. Hence, the resulting algorithms target a specific strategy rather than making arbitrary choices between redundant schemata. The next section formalizes parameterized normalization.

4 PARAMETERIZED 3NF NORMALIZATION

3NF admits the fewest sources of data redundancy among all loss-less, dependency-preserving decompositions [17]. However, not all schemata in 3NF are the same and non-key FDs cause serious update overheads. This presents a great opportunity for normalization, and it is surprising that no previous work has addressed it yet. Recently, BCNF schemata have been classified by the number of minimal keys they exhibit, but 3NF schemata have not received attention yet despite non-key FDs causing the biggest overheads.

We have already shifted as much application semantics of a given FD set into minimal keys, and then minimized the set of non-key FDs. We will now transfer these concepts into an algorithmic solution. After introducing parameterized variants of 3NF, we will study the computational complexity of parameterized normalization, introduce an algorithm suite and illustrate it on our example.

4.1 Parameterized 3NF

The first step is minimizing the set of non-key FDs, that is, the set of FDs not implied by the set of minimal keys. Maier [25] introduced minimal-reduced and optimal covers for a given set of FDs, minimizing their cardinality and size, respectively. While optimal covers provide the smallest total number of attributes occurring in any representation possible for any given FD set, their underlying decision problem is NP-complete [25]. In contrast, minimal-reduced covers provide the fewest FDs with no superfluous attribute occurrences in any representation and their computation is quadratic [25].

For a schema (R, Σ) in 3NF, the set \mathcal{K} of minimal keys for Σ , and a minimal-reduced cover \mathcal{F}_c for the set of non-key FDs implied by Σ , we call $(\mathcal{K}, \mathcal{F}_c)$ a minimal-reduced cover for the 3NF-core of (R, Σ) . For an optimal cover \mathcal{F}_s for the set of non-key FDs implied by Σ , we call $(\mathcal{K}, \mathcal{F}_s)$ an optimal cover for the 3NF-core of (R, Σ) .

Definition 4.1. Let Σ be an FD set over R, and k_c , k_s be positive integers, and f_c , f_s be non-negative integers. (R, Σ) is in (k_c, f_c) -3NF if and only if (R, Σ) is in 3NF and there is some minimal-reduced cover $(\mathcal{K}, \mathcal{F}_c)$ for the 3NF-core of (R, Σ) where \mathcal{K} has cardinality k_c , and \mathcal{F}_c has cardinality f_c . (R, Σ) is in (k_s, f_s) -3NF if and only if (R, Σ) is in 3NF and there is some optimal cover $(\mathcal{K}, \mathcal{F}_s)$ for the 3NF-core of (R, Σ) where \mathcal{K} has size k_s , and \mathcal{F}_s has size f_s .

When we write (k, f) we mean $(k, f) \in \{(k_c, f_c), (k_s, f_s)\}$, so we treat the two cases of minimal-reduced and optimal covers simultaneously. The property of (R, Σ) being in (k, f)-3NF is independent of how the semantic constraints are represented. That is, if Σ' and Σ

are covers of one another, then (R, Σ) is in (k, f)-3NF if and only if (R, Σ') is in (k, f)-3NF. All minimal-reduced covers have the same cardinality, and all optimal covers have the same size. Hence, finding some minimal-reduced or optimal cover, respectively, gives assurance that f is the best achievable.

The parameters $k \in \{k_c, k_s\}$ and $f \in \{f_c, f_s\}$ lead to different finite orders o such as i) minimizing or maximizing the ii) cardinality or size of the iii) set of minimal keys or set of non-key FDs. By motivation, we want to minimize f and will not consider $>_f$.

Based on a given order O, we fix a ranking $<_O$ that ranks the elements in order of preference by O. In $<_O$, the least element is always regarded "best" since our normalization framework aims at minimizing integrity maintenance. For example, if $O = <_f$, then $<_O = O$ is the natural order $0 < \ldots < n$ on some prefix of the nonnegative integers; and, if $O = >_k$, then $<_O$ is the reverse natural order $n < n - 1 < \ldots < 1 < 0$ on the prefix $0, \ldots, n$.

We may choose a primary parameter $l \in \{k, f\}$, and secondary parameter $m \in \{k, f\} - \{l\}$. We then have the following possible orders combining primary and secondary parameters $O = (O_l, O_m) \in \{(<_f, <_k), (<_f, >_k), (<_k, <_f), (>_k, <_f)\}$. The next example makes the ranking $<_O$ explicit for such orders O.

Example 4.2. The rankings $<_O$ we obtain from different orders O with primary and secondary parameters are:

0	<0
$(<_f,>_k)$:	$(0, k_0) < \ldots < (0, 1) < \ldots < (f, k_f) < \ldots < (f, 1)$
$(<_f,<_k)$:	$(0,1) < \ldots < (0,k_0) < \ldots < (f,1) < \ldots < (f,k_f)$
	$(1,0) < \ldots < (1,f_1) < \ldots < (k,0) < \ldots < (k,f_k)$
	$(k,0) < \ldots < (k,f_k) < \ldots < (1,0) < \ldots < (1,f_1)$

For the first ranking, a schema with FD set of cardinality/size 0, and set of minimal keys with cardinality/size k_0 is ranked first, while for the second ranking, a schema with FD set of cardinality/size 0, and set of minimal keys with cardinality/size 1 is ranked first.

It makes sense to apply different orders for parameter k regarding critical (f>0) and non-critical (f=0) schemata. For instance, we may want to minimize k on BCNF schemata using $<_k$, while maximizing k as secondary parameter on critical schemata by using $>_k$. In such cases, we write $<_{O^-BCNF}$ for the ranking resulting from the elements of an order O^-BCNF where f=0, and $<_{O^-3NF}$ for the ranking resulting from the elements of an order O^-3NF where f>0. We may merge $<_{O^-BCNF}$ and $<_{O^-3NF}$ into the ranking $<_{O^-BCNF/3NF}$ by defining the largest element (the least preferred) of the former to precede the smallest element (the most preferred) of the latter. For $O'-BCNF = <_k$ and $O''-3NF = (<_f,>_k)$ we may obtain the following merged ranking $<_{O-BCNF/3NF}$:

$$1 < \ldots < k < (1, k_1) < \ldots < (1, 2) < \ldots < (f, k_f) < \ldots < (f, 2)$$

where the k smallest elements represent ranking $<_{O^*\text{-}BCNF}$ and the remaining elements represent ranking $<_{O^*\text{--}3NF}$. Critical schemata have at least two different minimal keys. We now lift any ranking $<_O$ to a ranking $<_O^R$ of critical schemata, thereby breaking ties.

Definition 4.3. For a schema (R, Σ) in 3NF and a ranking $<_O$ for a finite order O, we define the 3NF-rank r_O^R of (R, Σ) as the smallest rank of any (k, f) in the ranking $<_O$ for which (R, Σ) is in (k, f)-3NF. We further define the order $<_O^R$ on 3NF schemata by $(R, \Sigma) <_O^R (R', \Sigma')$ if and only if $r_O^R <_O r_O^{R'}$.

Next we illustrate on our running example how different 3NF schemata can be compared with respect to given orders.

Example 4.4. Consider the two 3NF schemata from Example 3.11: $R_3 = \{E, M, S, V\}$ and Σ_3 with non-key FD $VS \to M$ and 2 minimal keys ESV and EMS; $R_4 = \{E, M, S, T\}$ and Σ_4 with 2 non-key FDs $MT \to E$, $ET \to M$ and 3 minimal keys EST, EMS, and MST. First, we select the order $O = <_{f_c}$ with ranking 1 < 2 as a strategy to minimize the number of non-key FDs. Since we have $r_O^{R_3} = 1$ and $r_O^{R_4} = 2$, we obtain $(R_3, \Sigma_3) <_O^R (R_4, \Sigma_4)$. Even if $O = (<_{f_c}, >_{k_c})$, we obtain (1, 2) < (2, 3) and the same result. Second, we select the order $O' = >_{k_c}$ with ranking 3 < 2 as a strategy to maximize the number of minimal keys. Since the 3NF-rank of R_3 for that order is $r_{O'}^{R_3} = 2$ and that of R_4 is $r_{O'}^{R_4} = 1$, we obtain $(R_4, \Sigma_4) <_{O'}^R (R_3, \Sigma_3)$. Even if $O' = (>_{k_c}, <_{f_c})$, we obtain ranking (3, 2) < (2, 1) and the same result. Since the two schemata are in 3NF, but not in BCNF, merging the rankings with any $<_{O-BCNF}$ will not have any impact. □

4.2 Computational Complexity

We have already discussed that classical 3NF synthesis suffers from likely computational intractability in general. However, it needs to be stressed that the underlying variables are defined at schema level, which means they are typically small. Despite the worst-case bounds, computation in practice is typically efficient. We will illustrate this later by experiments for our parameterized framework. However, it is still important to analyze the computational complexity to understand fundamental limits and set expectations.

The first fundamental problems are to decide whether a given schema is in (k_c, f_c) -3NF, or in (k_s, f_s) -3NF, respectively.

```
Parameterised 3NF Input: (R, \Sigma), non-negative integers (k, f) \in \{(k_c, f_c), (k_s, f_s)\} Problem: Decide whether (R, \Sigma) is in (k, f)-3NF
```

As the problem of computing the set of minimal keys is outputpolynomial and typically efficient in practice, we may regard this set as additional input, particularly in our framework where we separate this set from that of non-key FDs.

```
PARAMETERISED 3NF WITH SET OF MINIMAL KEYS

Input: (R, \Sigma), non-negative integer f \in \{f_c, f_s\}
Set \mathcal{K} of minimal keys of cardinality k_c and size k_s

Problem: Decide whether (R, \Sigma) is in (k, f)-3NF
```

The previous problems validate whether a schema meets some normal form condition. However, the design problem asks for loss-less, dependency-preserving decompositions of the schema. Hence, we need to consider dependencies on subsets of the schema. Consequently, we arrive at the following problem.

```
Parameterised 3NF Design  \begin{aligned} &\text{Input: } (R, \Sigma), \text{ non-negative integers } (k, f) \in \{(k_c, f_c), (k_s, f_s)\} \\ &\text{Set } S \subseteq R \end{aligned}  Problem: Decide whether (S, \Sigma[S]) is in (k, f)-3NF
```

Next we summarize the results associated with the computational complexity for the decision problems above.

Theorem 4.5. The parameterised variants associated with Third Normal Form have the following computational complexity.

(1) PARAMETERISED 3NF is NP-complete for each the cardinality-based and size-based variant.

- (2) PARAMETERISED 3NF WITH THE SET OF MINIMAL KEYS is polynomial for the cardinality-based variant, but NP-complete for the size-based variant.
- (3) PARAMETERISED 3NF DESIGN is NP-complete for each the cardinality-based and size-based variant, even if the set of minimal keys on the given schema is given, and even if the input is its own atomic cover.

These results set expectations for the following sections that will develop an algorithm suit and conduct experiments.

4.3 Algorithm Suite

We will now channel our ideas into an algorithm that uses a specific strategy to optimize state-of-the-art normalization.

Previous algorithms guarantee a lossless, dependency-preserving decomposition into 3NF that is in BCNF whenever possible [28]. BCNF is returned iff every critical schema is redundant. Recently [37], this was improved by parameterizing schemata in BCNF by the number of minimal keys they exhibit. This made it possible to break ties between BCNF schemata. However, no work has attempted to break ties between critical schemata. Hence, outputs are not optimized for integrity maintenance of non-key FDs.

We will address this opportunity for optimization by introducing a co-lexical order $<_{O\text{-}BCNF/3NF}^{D}$ on lossless, dependency-perserving 3NF decompositions, using rankings < O-BCNF/3NF of orders O-BCNF and O-3NF for parameters k and f. Our algorithm will return a lossless, dependency-preserving decomposition into 3NF that is optimal for the target order O-BCNF/3NF.

Put simply, we minimize the number of critical schemata that are less preferred according to $<_{O-3NF}$. For example, based on $(<_{f_c},>_{k_c})$, we first keep redundant 3NF schemata with smaller f_c and break further ties by keeping those with larger k_c .

We will now introduce the co-lexical order. Intuitively, a decomposition is better than another when the former's worst rank (its least preferred) is better than the latter's worst rank.

Definition 4.6. Let D be a set of lossless, dependency-preserving decompositions of a schema (R, Σ) into 3NF. For $\mathcal{D} \in \mathbf{D}$, ranking $<_{O extit{BCNF/3NF}}$ that spans all values of parameters that occur in any decomposition in D, rank r in $<_{O\text{-}BCNF/3NF}$, let $\mathcal{S}_r^{\mathcal{D}}$ denote the set of schemata $(S, \Sigma[S]) \in \mathcal{D}$ with rank $r_Q^S = r$, that is,

$$\begin{split} \mathcal{S}_r^{\mathcal{D}} &= \{ (S, \Sigma[S]) \in \mathcal{D} \mid (S, \Sigma[S]) \text{ has rank } r_O^S = r \text{ in } <_{O\text{-}BCNF/3NF} \} \\ \text{and } n_r^{\mathcal{D}} \text{ its cardinality, that is, } n_r^{\mathcal{D}} &= |\mathcal{S}_r^{\mathcal{D}}|. \text{ For } \mathcal{D}', \mathcal{D}'' \in \mathbf{D}, \mathcal{D}' \text{ is } \\ D\text{-}better \text{ than } \mathcal{D}'' \left(\mathcal{D}' <_{O\text{-}BCNF/3NF}^D \mathcal{D}'' \right) \text{ if and only if for the worst } \\ \text{rank } r \text{ in } <_{O\text{-}BCNF/3NF} \text{ where } \mathcal{S}_r^{\mathcal{D}'} \neq \mathcal{S}_r^{\mathcal{D}''}, \mathcal{S}_r^{\mathcal{D}'} = \emptyset. \end{split}$$

Alternatively, \mathcal{D}' is n-better than \mathcal{D}'' $(\mathcal{D}' <^n_{\textit{O-BCNF/3NF}}, \mathcal{D}'')$ if and only if for the worst rank r in $<_{O-BCNF/3NF}$ where $n_r^{\mathcal{D}'} \neq 0$ or $n_r^{\mathcal{D}''} \neq 0, n_r^{\mathcal{D}'} = 0$. However, if \mathcal{D}' is D-better than \mathcal{D}'' , then \mathcal{D}' is also *n*-better than \mathcal{D}'' , but not vice versa. As our algorithm will ensure $<_{O\text{-}BCNF/3NF}^{D}$ -optimality, it will also be optimal for $<_{O\text{-}BCNF/3NF}^{n}$

For illustration of our definitions, we compare the decompositions of our running example with respect to different orders.

Example 4.7. Consider $D = \{\mathcal{D}_1, \mathcal{D}_2\}$ with \mathcal{D}_1 and \mathcal{D}_2 from Example 1.1. First, let O-BCNF = $<_{k_c}$ and O-3NF = $<_{f_c}$, resulting in the following ranking $<_{O-BCNF/3NF}$ where the two sub-rankings

Algorithm 1 $ICONF(R, \Sigma, O-3NF, O-BCNF)$

Require: (R, Σ) with FD set Σ over schema R, 3NF-order O-3NF, BCNF-order O-BCNF

Ensure: Lossless, FD-preserving 3NF decomposition \mathcal{D} of (R, Σ) that is $<_{O\text{-}BCNF/3NF}^{D}$ -optimal

```
1: Compute the atomic closure \overline{\Sigma}_a of \Sigma [15, 28]
 2: \Sigma_a \leftarrow \overline{\Sigma}_a, \Sigma_c \leftarrow \emptyset
 3: for all X \to A \in \Sigma_a do
          for all Y \to B \in \Sigma_a(YB \subseteq XA \land XA \nsubseteq Y_{\Sigma}^+) do
               Compute k_{XA} and f_{XA} in target 3NF-core of \Sigma_a[XA]
               \Sigma_c \leftarrow \Sigma_c \cup \{(X \to A, k_{XA}, f_{XA})\}
 7: \mathcal{D} \leftarrow \emptyset;
 8: for all (X \to A, k_{XA}, f_{XA}) \in \Sigma_c in reverse <_{O\text{-}3NF}-ranks do
          if \overline{\Sigma}_a \setminus \{X \to A\} \not\models X \to A then
               \mathcal{D} \leftarrow \mathcal{D} \cup \{(XA, \Sigma_a[XA], k_{XA}, f_{XA})\}
10:
11:
               \overline{\Sigma}_a \leftarrow \overline{\Sigma}_a \setminus \{X \to A\}
13: for all X \to A \in \overline{\Sigma}_a \backslash \Sigma_c do
          k_{XA} \leftarrow |\{K \mid K \rightarrow XA \in \Sigma_a[XA]\}^+\}| \text{ using } [24]
          \Sigma_a \leftarrow (\Sigma_a \setminus \{X \to A\}) \cup \{(X \to A, k_{XA})\}
16: for all (X \to A, k_{XA}) \in \overline{\Sigma}_a \backslash \Sigma_c in reverse <_{O\text{-}BCNF}-ranks do
          if \overline{\Sigma}_a \setminus \{X \to A\} \not\models X \to A then
               \mathcal{D} \leftarrow \mathcal{D} \cup \{(XA, \Sigma_a[XA], k_{XA})\}
18:
19:
               \overline{\Sigma}_a \leftarrow \overline{\Sigma}_a \setminus \{X \to A\}
21: Remove all (S, \Sigma_a[S]) \in \mathcal{D} if \exists (S', \Sigma_a[S']) \in \mathcal{D}(S \subseteq S')
22: if there is no (R', \Sigma') \in \mathcal{D} where R' \to R \in \Sigma_{\mathfrak{N}}^+ then
          Choose a minimal key K for R with respect to \Sigma
           \mathcal{D} \leftarrow \mathcal{D} \cup \{(K, \Sigma_a[K], 1)\}
25: Return(D)
```

are separated by <<:1<2<3<<1<2. The table below shows $S_r^{\mathcal{D}}$ for every $\mathcal{D} \in \mathbf{D}$ and every rank r.

\mathcal{D}	$\mathcal{S}_1^{\mathcal{D}}$	$\mathcal{S}_2^{\mathcal{D}}$	$\mathcal{S}_3^{\mathcal{D}}$	$\mathcal{S}_4^{\mathcal{D}}$	$\mathcal{S}_5^{\mathcal{D}}$
\mathcal{D}_1	Ø	$\{R_2\}$	$\{R_1\}$	$\{R_3\}$	Ø
\mathcal{D}_2	$\{R_5\}$	Ø	$\{R_1\}$	Ø	$\{R_4\}$

The worst rank on which \mathcal{D}_1 and \mathcal{D}_2 have different schemata is rank 5. As $n_5^{\mathcal{D}_1} = 0 < 1 = n_5^{\mathcal{D}_2}$, we have $\mathcal{D}_1 <_{O\text{-}BCNF/3NF}^D \mathcal{D}_2$. Second, let $O\text{'-}BCNF = <_{k_c}$ and $O\text{'-}3NF = >_{k_c}$, resulting in the following ranking $<_{O'-BCNF/3NF}$ where the two sub-rankings are separated by <<: 1 < 2 < 3 << 3 < 2. The table below shows $\mathcal{S}_r^{\mathcal{D}}$ for every $\mathcal{D} \in \mathbf{D}$ and every rank r.

\mathcal{D}	$\mathcal{S}_1^{\mathcal{D}}$	$\mathcal{S}_2^{\mathcal{D}}$	$\mathcal{S}_3^{\mathcal{D}}$	$\mathcal{S}_4^{\mathcal{D}}$	$\mathcal{S}_5^{\mathcal{D}}$
\mathcal{D}_1	Ø	$\{R_2\}$	$\{R_1\}$	Ø	$\{R_3\}$
\mathcal{D}_2	$\{R_5\}$	Ø	$\{R_1\}$	$\{R_4\}$	Ø

The worst rank on which \mathcal{D}_1 and \mathcal{D}_2 have different schemata is rank 5. As $n_5^{\mathcal{D}_1} = 1 > 0 = n_5^{\mathcal{D}_2}$, we have $\mathcal{D}_2 <_{O'\text{-}BCNF/3NF}^D \mathcal{D}_1$.

Algorithm 1 computes a lossless, dependency-preserving 3NF decomposition that is $<_{O\text{-}BCNF/3NF}^{D}$ -optimal for its input. It starts with the *atomic closure* $\overline{\Sigma}_a$ of input Σ in line 1, that is, the unique set of minimal FDs $X \to A$ implied by Σ [15, 28]. Line 2 creates a copy of the atomic closure from which redundant FDs may be removed later, while the original closure checks for critical FDs (line 4).

The part in lines (3-6) is called *Critical*. It assembles the set Σ_c of critical schemata (line 6) and values of their parameters k and f (line 5) that determine the ranking $<_{O-3NF}$ for input order O-3NF.

The part in lines (7-12) is called *Opt-3NF* where FDs that cause critical schemata are evaluated in reverse ranks of $<_{O-3NF}$ (line 8). If an FD is redundant (line 12), we do not need the schema. Processing these FDs from worst to best ensures we eliminate remaining worst schemata when possible. If the FD is not redundant (line 9), the schema is added (line 10). *Opt-3NF* ensures $<_{O-3NF}$ -optimality.

The part in lines (13-15) is called *Key*. It computes the number of minimal keys for non-critical schemata. The computation is done in output-polynomial time [24].

The part in lines (16-20) is called Opt-BCNF. It loops through noncritical FDs in reverse ranks of $<_{O$ -BCNF</sub> (line 16), eliminates the FD when redundant (line 20) or add its schema otherwise (lines 17/18). This eliminates redundant BCNF schemata following O-BCNF.

The part in line 21 is called *Subset*. Here, we remove any schema that is a subset of another one.

Finally, the part in lines (22-24) is called *Lossless*. It adds a minimal key to the output that ensures the decomposition is lossless when returned in line 25. This concludes the explanation of Algorithm 1.

The following result improves the state-of-the-art [37], which returns a lossless, dependency-preserving decomposition into 3NF that is in BCNF if possible and $<_{k_c}$ -optimal in that case.

Theorem 4.8. On input $(R, \Sigma, O\text{-}3NF, O\text{-}BCNF)$, Algorithm 1 returns a lossless, dependency-preserving decomposition into 3NF that is $<_{O\text{-}BCNF/3NF}^O$ -optimal.

Unlike any previous work, Algorithm 1 optimizes critical schemata and also does so with respect to any given target strategy.

4.4 Running Example

We illustrate Algorithm 1 by showing how the decompositions in Example 1.1 were derived. Firstly, we choose $O_{3NF} = <_{f_c}$ and $O_{BCNF} = <_{k_c}$ as input. Hence, our strategy is to minimize the number of non-key FDs on critical schemata and the number of minimal keys on BCNF schemata.

We start with the atomic closure $EMS \rightarrow V$, $SV \rightarrow M$, $EMS \rightarrow T$, $MT \rightarrow E$, $ET \rightarrow M$, $MST \rightarrow V$, $SET \rightarrow V$, $ESV \rightarrow T$, $STV \rightarrow E$.

Critical schemata are (R_3, Σ_3) with 1 non-key FD and 2 minimal keys, (R_4, Σ_4) with 2 non-key FDs and 3 minimal keys, and $(R_6 = MSTV, \Sigma_6)$ with 1 non-key FD $SV \rightarrow M$ and 2 minimal keys STV and MST. BCNF schemata are (R_5, Σ_5) with 1 minimal key, (R_2, Σ_2) with 2 minimal keys, and (R_1, Σ_1) with 3 minimal keys.

For $O_{3NF} = <_{f_c}$, the critical schemata are ordered as: $(R_3, \Sigma_3) =_{f_c}^R (R_6, \Sigma_6) <_{f_c}^R (R_4, \Sigma_4)$. As the R_4 -generating FD $EMS \to T$ is redundant, (R_4, Σ_4) is not required. The R_3 -generating FD $EMS \to V$ is not redundant now, so the schema (R_3, Σ_3) is added to the decomposition. Next, the R_6 -generating FD $MST \to V$ is still redundant, so the schema (R_6, Σ_6) is not required. For $O_{BCNF} = <_{k_c}$, the BCNF schemata are ordered as: $(R_5, \Sigma_5) <_{k_c}^R (R_2, \Sigma_2) <_{k_c}^R (R_1, \Sigma_1)$ but the R_1 -generating FD $EST \to V$ is not redundant, and neither the R_2 -generating FDs $ET \to M$ or $MT \to E$, so (R_1, Σ_1) and (R_2, Σ_2) are added to the decomposition. Indeed, the same is true for the R_5 -generating FD $SV \to M$, so (R_5, Σ_5) is added,

too. However, $R_5 \subseteq R_3$, so (R_5, Σ_5) is removed from the decomposition. The final decomposition $\mathcal{D}_1 = \{(R_1, \Sigma_1), (R_2, \Sigma_2), (R_3, \Sigma_3)\}$ is $<_{O\text{-}3NF/BCNF}^D$ -optimal.

Using $O'_{3NF} = >_{k_c}$ as input, the critical schemata are ordered as: $(R_4, \Sigma_4) <_{k_c}^R (R_3, \Sigma_3) =_{k_c}^R (R_6, \Sigma_6)$, the R_3 -generating FD $EMS \rightarrow V$ is redundant and thus removed, and the R_6 -generating FD $MST \rightarrow V$ is redundant and also removed. However, the R_4 -generating FD $EMS \rightarrow T$ is not redundant, and the schema (R_4, Σ_4) is added to the decomposition. Based on $O'_{BCNF} = <_{k_c}$, the BCNF schemata are ordered as before: $(R_5, \Sigma_5) <_{k_c}^R (R_2, \Sigma_2) <_{k_c}^R (R_1, \Sigma_1)$, but the R_1 -generating FD $EST \rightarrow V$ is not redundant, and neither the R_2 -generating FDs $ET \rightarrow M$ or E0 or E1 or E2 are added to the decomposition. Indeed, the same is true for the E3-generating FD E3 or E4, so E5 is added, too. However, E8 or E4, so E5 is removed. The final decomposition E6 or E7 or E8 or E9 is removed. The final decomposition E9 or E1 or E1 or E2 or E3 is removed. The final decomposition E9 or E1 or E2 or E3 is removed. The final decomposition E9 or E1 or E3 is removed. The final decomposition E9 or E1 or E1 or E2 or E3 is removed. The final decomposition E9 or E1 or E1 or E2 or E3 is removed. The final decomposition E1 or E2 or E3 is removed.

5 EXPERIMENTS

Through experiments we seek answers to the following questions.

- (Q1) How do keys and non-key FDs affect performance?
- (Q2) How much can we improve state-of-the-art algorithms?
- (Q3) How much update overheads can we save?

Answers to (Q1) will motivate our research more and further inform the strategies we report on. (Q2) focuses on the efficacy of our normalization algorithms over previous work at the logical level, in terms of meeting their design goals and the time they need to do that. (Q3) investigates how well our optimizations transcend from logical to operational level. In particular, we will see which strategy works best for lowering overheads of integrity maintenance.

5.1 Experimental set up

Our algorithms were implemented in Java, Version 17.0.7, and run on a 12th Gen Intel(R) Core(TM) i7-12700, 2.10GHz, with 128GB RAM, 1TB SSD, and Windows 10. We used MySQL 8.0.29.

Data sets. Apart from perfect synthetic data for realistic schemata and sets of FDs, we used FDs mined from 12 real-world benchmark data plus TPC-H ¹. These have served as benchmarks for profiling data dependencies [29, 30, 33], but also for experiments in previous work [37]. Hence, we cannot only compare our algorithms to state-of-the-art, but also analyze them on instances with tens, hundreds, and thousands of FDs to test scalability.

Algorithms. We implemented our new algorithms, and used the mining of FDs [33] and generation of Armstrong relations [27]. In line with previous work, we used the cardinality of key sets and FDs sets. We will denote by iConf-fk (A1) and iConf-f (A2) our Algorithm 1 where O-3NF is $(<_{f_c},>_{k_c})$ and $<_{f_c}$, respectively, and where $O\text{-}BCNF\text{=}<_{k_c}$ in both cases. We will compare performance of these to our implementations of previous work: Conf (A3) [37] (which is based on $O\text{-}BCNF\text{=}<_{k_c}$), BC-Cover (A4) [28], and Synthesis (A5) [8].

5.2 How do keys and FDs affect performance?

We ran the TPC-H benchmark (scaling factor 0.1) with 22 queries, 7 refresh and 3 insert (adding 1k, 2k, and 3k of records) operations

 $^{^1}hpi.de/naumann/projects/repeatability/data-profiling/fds.html\\$

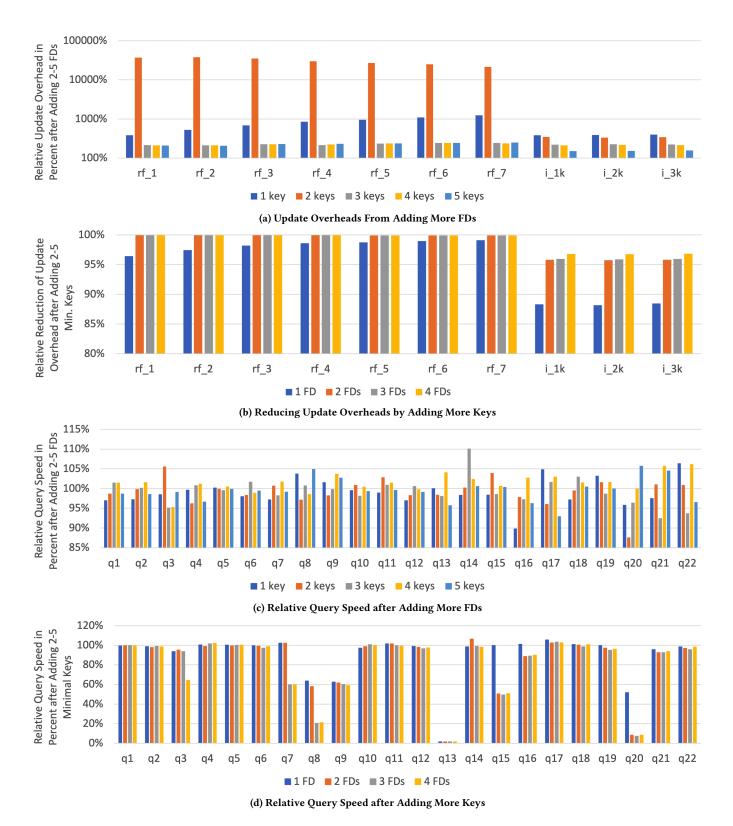


Figure 2: Impact of More FDs and More Keys on Update and Query Performance over TPC-H

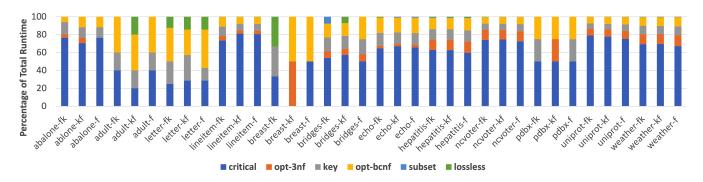


Figure 3: Breakdown of (A1-5) in Percent of Total Runtime

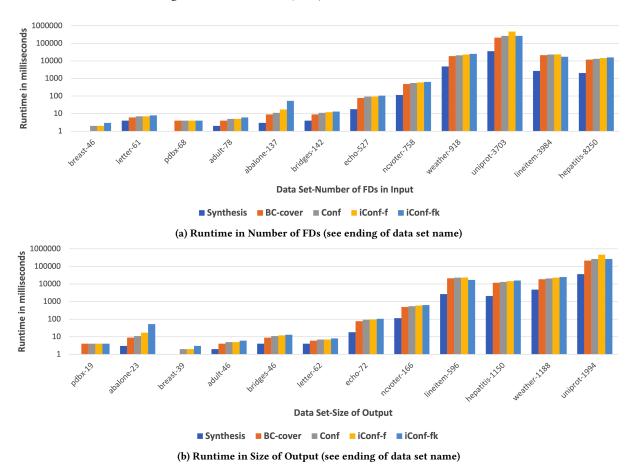


Figure 4: Runtime in Input Size and Output Size

after declaring 1-5 minimal keys as UNIQUE constraints and enforcing 1-5 non-key FDs by triggers on each table. The constraints were mined [33]. Hence, we compare performance of TPC-H across 25 constraint sets. Times reported for an operation are averaged over 30 runs. We study how adding FDs and keys to the schema affects performance, primarily for refresh and inserts. Each bar in Figure 2 represents, in percent, the relative speed of an operation (update or query) run after adding 1 to 4 FDs (or keys), when compared to the

same operation run with just 1 FD (or key), in each case of having 1-5 keys (or FDs) present.

Figure 2a shows how adding FDs to a number of minimal keys incurs update overheads. The average overhead across the 7 refresh operations and all constraint sets is more than 6400%, and across the 3 inserts it is more than 264%. Having sufficiently many keys does scale update performance when more non-key FDs are present.

	Charact	erist	ics	Time of Algorithms (in ms)				
Data set	#R	#C	#FD	Synthesis	BC-Cover	Conf	iConf-f	iConf-fk
abalone	4,177	9	137	3	9	11	17	53
adult	48,842	14	78	2	4	5	5	6
breast	699	11	46	1	1	2	2	3
bridges	108	13	142	4	9	11	12	13
echo	132	13	527	18	77	93	94	105
hepatitis	155	20	8,250	2064	11,797	13,134	14,551	15,865
letter	20,000	17	61	4	6	7	7	8
lineitem	6,001,215	16	3,984	2,698	21,269	23,056	23,696	17,364
ncvoter	1,000	19	758	115	489	547	595	640
pdbx	17,305,799	13	68	1	4	4	4	4
uniprot	512,000	30	3,703	36,238	213,958	266,825	468,059	266,257
weather	262,920	18	918	4,796	18,824	20,925	23,184	25,140

Table 1: Data Sets and Runtimes (in ms) by Data Sets

Figure 2b shows how adding minimal keys to a number of non-key FDs reduces update overheads. The average reduction across the 7 refresh operations and all constraint sets is more than 99.4%, and across the 3 inserts it is more than 94%. Adding a few more FDs incurs huge update overheads, but having a few more minimal keys can scale integrity maintenance well.

Figure 2c shows how adding FDs to a number of minimal keys affects query performance. The average speed across the 22 queries is just below 99.8%. So while some queries are affected, on average there is little impact on query performance resulting from FDs.

Figure 2d shows how adding minimal keys to a given number of non-key FDs impacts query performance. The average speed is just below 83.6%, so a speed up of over 14.4%. This quantifies our expectations that the UNIQUE index resulting from keys does improve query speed. Selecting FDs arbitrarily does not affect query performance much, but UNIQUE indices resulting from arbitrary selections of minimal keys do have a noticeable impact.

Conclusion. Our mini and TPC-H studies quantify the need for developing a framework that separates non-key FDs from minimal keys to access the parameters, and that aligns schema design closer with integrity maintenance at the operational level. The experiments show that (1) the number of non-key FDs is a valuable parameter to minimize, (2) the number of minimal keys is a valuable parameter that affects update and query performance, and (3) relying exclusively on FDs for integrity maintenance is infeasible.

5.3 How good are our algorithms?

All algorithms return a lossless, dependency-preserving (LD-) decomposition into 3NF. If an LD-decomposition into BCNF exists, BC-Cover will find one. Conf improves BC-Cover by returning an LD-decomposition into k-CONF for the lowest k possible. iConf-f guarantees that the LD-decomposition into 3NF is optimized for O-3NF=< f_c and O-BCNF=< f_c . iConf-f guarantees that the maximum number of non-key FDs across all output schemata is minimized; and if an LD-decomposition into BCNF exists (that is, if that maximum number is zero), then iConf-f returns the same result as Conf. iConf-fk breaks remaining ties between 3NF schemata with the same number of non-key FDs by prioritizing larger numbers of minimal keys. Later, we will discuss other variants where O-3NF=(f_c , f_c) or O-3NF=(f_c , f_c).

5.3.1 Runtime analysis. Part of Table 1 shows for each of the 12 data sets, their name, numbers of rows ($\sharp R$) and columns ($\sharp C$), and

the number of FDs in the atomic closure for the set of FDs exhibited by the data sets ($\sharp FD$). In line with previous work, we uniformly regard two missing values as a match. Regarding them as no match leads to different FDs, but overall observations do not change.

In addressing O2, Table 1 reports the total run time of Algorithm 1 and the time spent on its steps indicated before, for each data set. Based on the different characteristics of our data sets, run times differ quite significantly, too. Foremost, all algorithms run efficiently despite some large input sizes. In fact, the longest run times were exhibited on uniprot, with less than 5 minutes except for iConf-f which took less than 8 minutes. All other FD sets did not take longer than 30 seconds. There is an order of magnitude difference between the run time of Synthesis and the remaining algorithms, which quantifies their effort to find an LD-decomposition into BCNF, not attempted by Synthesis. Unsurprisingly, there are runtime overheads for optimizations, in particular for iConf-f over Conf. While the difference is mostly insignificant, it is less than 1.5 seconds on hepatitis, less than a second on lineitem, less than 3 seconds on weather, and less than 3.5 minutes in the most extreme case uniprot. Schema design is a critical task, and our experiments quantify the runtime overheads incurred by optimizing schema design algorithms for the target strategy.

Figure 3 shows how much percent each step of the algorithm contributes to the overall run time. *Critical* consumes most of the time due to computing critical schemata, minimal covers for FDs projected on their schemata, and finding all minimal keys. The optimizations for 3NF and BCNF consume significant time due to large numbers of FDs and keys, and so does *Keys* when computing all minimal keys on the BCNF schemata.

Figure 4 quantifies the expected exponential dependence of the run time of all algorithms on the number of input FDs, and the number of schemata in the output, respectively.

5.3.2 Output analysis. For Q3, Table 2 reports the results of applying (A1)-(A5) to the FDs mined from the 12 data sets. For each data set, we list the Alg orithms for which we report results (joining algorithms with the same results), the total number Size of relation schemata in the output, split into the number BCNF of BCNF schemata and the number 3NF of critical schemata, and the average numbers of minimal keys #Keys (non-key #FDs) across schemata in BCNF (3NF). Under Distribution, n_i denotes how many BCNF (3NF) schemata exhibit precisely s_i minimal keys (non-key FDs).

By design, BC-Cover (A4) generates never more and usually fewer critical schemata than Synthesis (A5), in percent of 3NF schemata. By design, Conf and BC-Cover produce the same schemata in 3NF, but Conf optimizes BCNF schemata for $<_{k_c}$. In fact, Conf produces decompositions with better D-ranks than BC-Cover on abalone, echo, hepatitis, lineitem, nevoter and uniprot; and fewer schemata at the lowest ranks where they differ on bridges, pdbx, and weather.

As our main target, *iConf-f* (A2) optimizes the 3NF distribution over *Conf* (A3). This is most visible on *lineitem* where (A2) has eliminated 3NF schemata with 10, 9 and 5 FDs in them. Similarly, (A2) produces decompositions that are *D*-better than those generated by (A3) on ncvoter, uniprot, and weather; and fewer schemata at the lowest rank where they differ on abalone and hepatitis.

Variant iConf-fk (A1) is an optimization that retains those redundant 3NF schemata that are tied using iConf-f (A2) but exhibit

		Dec	omposi	tion		Schema in BCNF		Schema in 3NF
Data set	Alg	Size	BCNF	3NF	#Keys	Distribution	#FDs	Distribution
abalone	A1	26	22	4	1.64	[3:1,2:12,1:9]	2	[4:1,2:1,1:2]
	A2	23	18	5	1.61	[2:11,1:7]	1.8	[4:1,2:1,1:3]
	A3	21	16	5	1.81	[3:2,2:9,1:5]	2.4	[4:2,2:1,1:2]
	A4	20	15	5	2.07	[5:1,3:2,2:8,1:4]	2.4	[4:2,2:1,1:2]
	A5	21	14	7	1.93	[3:2,2:9,1:3]	2.29	[4:2,3:1,2:1,1:3]
adult	A1-5	46	46	0	1.02	[2:1,1:45]		
breast	A1-5	39	37	2	1.03	[2:1,1:36]	1	[1:2]
bridges	A1-3	46	39	7	1.15	[3:1,2:4,1:34]	1	[1:7]
	A4	44	37	7	1.19	[3:1,2:5,1:31]	1	[1:7]
	A5	43	34	9	1.21	[3:1,2:5,1:28]	1	[1:9]
echo	A1-3	72	65	7	1.46	[3:6,2:18,1:41]	1.14	[2:1,1:6]
	A4-5	72	65	7	1.58	[4:1,3:9,2:17,1:38]	1.14	[2:1,1:6]
hepatitis	A1-2	1150	842	308	1.12	[3:9,2:82,1:751]	1.88	[10:1,9:1,8:1,7:3,6:4,5:12,4:12,3:31,2:63,1:180]
	A3	1130	826	304	1.12	[3:10,2:82,1:734]	1.97	[10:1,9:1,8:3,7:4,6:5,5:12,4:13,3:27,2:66,1:172]
	A4	1123	819	304	1.13	[4:1,3:10,2:80,1:728]	1.97	[10:1,9:1,8:3,7:4,6:5,5:12,4:13,3:27,2:66,1:172]
	A5	1113	784	329	1.1	[4:1,3:6,2:66,1:711]	1.93	[10:1,9:1,8:3,7:4,6:6,5:13,4:14,3:27,2:67,1:193]
letter	A1-5	62	62	0	1	[1:62]		
lineitem	A1	590	560	30	1.39	[15:1,10:1,6:3,5:3,4:5,3:23,2:105,1:419]	1.4	[4:1,2:9,1:20]
	A2	596	567	29	1.39	[15:1,10:1,6:4,5:3,4:5,3:23,2:105,1:425]	1.41	[4:1,2:9,1:19]
	A3	587	558	29	1.38	[15:1,10:1,6:3,5:3,4:5,3:22,2:104,1:419]	2.62	[10:2,9:1,5:1,4:1,3:2,2:10,1:12]
	A4	562	533	29	2.32	$[15{:}1{,}11{:}1{,}10{:}2{,}9{:}9{,}8{:}9{,}7{:}19{,}6{:}26{,}5{:}26{,}4{:}26{,}3{:}26{,}2{:}46{,}1{:}342]$	2.62	[10:2,9:1,5:1,4:1,3:2,2:10,1:12]
	A5	531	466	65	2.29	[11:1,10:1,9:8,8:9,7:19,6:21,5:25,4:24,3:18,2:30,1:310]	2.28	[10:3,9:1,5:1,4:4,3:6,2:20,1:30]
ncvoter	A1	166	145	21	1.19	[2:27,1:118]	1.19	[3:1,2:2,1:18]
	A2	166	145	21	1.2	[3:1,2:27,1:117]	1.19	[3:1,2:2,1:18]
	A3	168	147	21	1.2	[3:1,2:28,1:118]	1.29	[4:1,2:3,1:17]
	A4	162	141	21	1.28	[4:2,3:5,2:23,1:111]	1.29	[4:1,2:3,1:17]
	A5	154	123	31	1.24	[4:1,3:3,2:20,1:99]	1.35	[4:1,2:8,1:22]
pdbx	A1-3	19	14	5	1.21	[2:3,1:11]	1	[1:5]
	A4-5	18	13	5	1.31	[2:4,1:9]	1	[1:5]
uniprot	A1	1992	1576	416	1.13	[4:1,3:5,2:187,1:1383]	1.48	[14:1,8:1,7:2,5:1,4:13,3:17,2:91,1:290]
	A2	1994	1578	416	1.13	[4:1,3:5,2:187,1:1385]	1.48	[14:1,8:1,7:2,5:1,4:13,3:17,2:91,1:290]
	A3	1981	1564	417	1.13	[4:1,3:5,2:186,1:1372]	1.56	[14:1,11:1,8:1,7:2,5:3,4:17,3:19,2:91,1:282]
	A4	1946	1529	417	1.16	[5:1,4:2,3:10,2:207,1:1309]	1.56	[14:1,11:1,8:1,7:2,5:3,4:17,3:19,2:91,1:282]
	A5	1923	1443	480	1.13	[5:1,4:1,3:8,2:169,1:1264]	1.53	[14:1,11:1,8:1,7:2,5:3,4:17,3:23,2:103,1:329]
weather	A1	1186	796	390	1.2	[6:1,5:1,4:3,3:11,2:120,1:660]	2.47	[7:5,6:10,5:27,4:46,3:68,2:112,1:122]
	A2	1188	796	392	1.2	[6:1,5:1,4:3,3:11,2:119,1:661]	2.46	[7:5,6:10,5:27,4:46,3:68,2:112,1:124]
	A3	1162	770	392	1.21	[6:1,5:1,4:3,3:11,2:119,1:635]	2.56	[9:1,7:7,6:11,5:27,4:50,3:67,2:115,1:114]
	A4	1154	762	392	1.25	[6:2,5:3,4:3,3:16,2:126,1:612]	2.56	[9:1,7:7,6:11,5:27,4:50,3:67,2:115,1:114]
	A5	1127	702	425	1.19	[4:1,3:12,2:104,1:585]	2.53	[9:1,7:8,6:12,5:27,4:53,3:74,2:120,1:130]

Table 2: Properties of Output Decomposition based on FD sets mined from Data Sets

more minimal keys. Compared to (A2), (A1) eliminates a few more redundant 3NF schemata due to this strategy, such as on abalone and weather. For nevoter, the 3NF distributions coincide but the schemata differ, making it possible to eliminate the BCNF schema with the most minimal keys compared to (A2) and (A3). On lineitem, however, (A1) uses an additional 3NF schema over (A2), but has fewer BCNF-schemata on some ranks.

Conclusion. Our algorithms optimize the logical design of database schemata based on a target strategy. The experiments illustrate what our algorithms achieve over state-of-the-art, such as reducing the number of non-key FDs in critical schemata. Considering the computational barriers to overall efficiency, our algorithms achieve their goals efficiently in practice.

5.4 How much overhead do we save?

We will study now how our optimizations on the logical level transcend to integrity maintenance at the operational level. For that purpose, we insert 10k, 20k, and 30k of records into abalone, hepatitis, lineitem, ncvoter, and weather. These insertions are done for the projections of these records onto the output schemata of our decompositions, resulting from *iConf-f, Conf, BC-Cover*, and *Synthesis*. Operations are repeated 10 times and the average runtime reported. We report the total times where all constraints are enforced by FDs and where FDs are separated into non-key FDs and minimal keys.

Firstly, integrity maintenance with only FDs is inefficient, if not infeasible. There are orders of time units difference (hours over minutes, or minutes over seconds) between the uniform use of FDs and the combined use of non-key FDs and minimal keys. In fact, non-key FDs require triggers while minimal keys are supported by

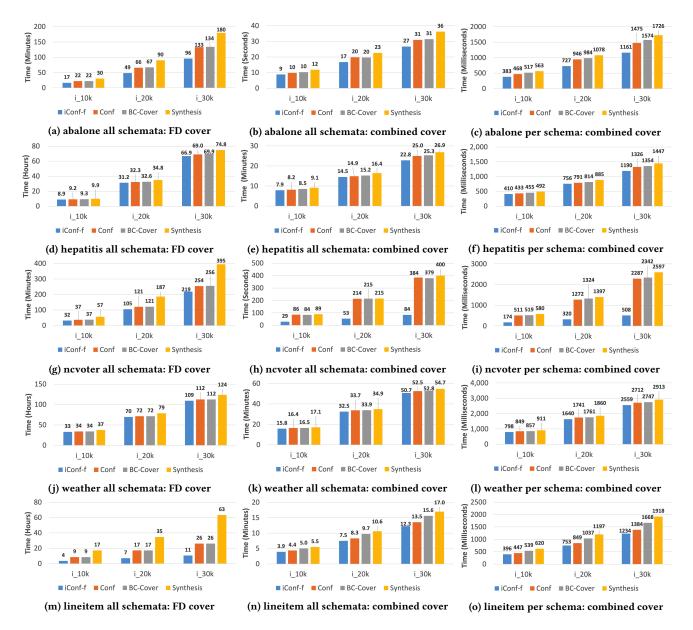


Figure 5: Overheads for maintaining integrity with FD and combined covers when inserting 10k, 20k, 30k of records on schemata obtained by different normalization methods

UNIQUE indices. This further motivates our parameterized framework that inherently links representations of constraints at the schema level with integrity maintenance at the operational level.

Secondly, our solution really does address the bottleneck of update inefficiency. By minimizing non-key FDs we do not just reduce update overheads further, but our reduction comes at a larger scale than optimizations from previous work. This is quantified in Table 3 that compares the algorithms in their ability to reduce update overheads when FDs are separated into non-key FDs and minimal keys. We report the average reductions in total and per schema (in

Comparison	total	per schema
iConf-f over Conf	20.0%	23.5%
Conf over BC-Cover	3.0%	6.2%
BC-Cover over Synthesis	5.7%	8.7%

Table 3: Average reduction of overheads across algorithms

percent), across all update operations and data sets for each of the two algorithms we compare.

Figure 5 details update overheads on all 5 data sets, including total times using (1) minimal-reduced covers with FDs only, and (2)

constraint sets combining all minimal keys with a minimal-reduced cover for all non-key FDs. Indeed, *iConf-f* outperforms the previously best algorithm *Conf*, across all scenarios with different input sizes for schemata, constraints and records. Hence, the optimizations do translate from logical to operational level. The magnitudes of reduction differ between scenarios but are significant.

Optimizations. We may use secondary parameters to break some ties that persist to hold after using primary parameters. Table 4 lists properties of decompositions resulting from these strategies on some data sets where they differ. We note small differences, and sometimes few schemata may be eliminated or added. Figure 6 illustrates the update performance on these decompositions. In line with the small differences at logical level, there are small differences at operational level. Overall, breaking further ties by maximizing the number of minimal keys, that is strategy (A1), appears to result in further small reductions of update overheads.

Conclusion. Experiments at operational level demonstrate that our framework does address the bottleneck of previous normalization efforts. By minimizing non-key FDs we achieve reductions at a scale larger than optimizations from previous work. Without separating non-key FDs from minimal keys, integrity maintenance degrades by orders of magnitude. Selecting redundant critical schemata with fewer FDs (and more keys if ties persist) results in the largest reduction of update overheads we were able to demonstrate.

6 CONCLUSION AND FUTURE WORK

The impact of schema design quality on integrity maintenance and update overheads has not been addressed systematically by previous work. In fact, as the bottleneck of integrity maintenance results from FDs, it makes sense to shift as much semantics of FDs into keys as possible and minimize remaining non-key FDs. Based on these ideas, we have developed a 3NF schema design framework that rigorously separates minimal keys and non-key FDs. As this provides access to these parameters, they can inform strategies that optimize 3NF normalization at the logical level. Despite the likely intractability of the underlying design problem in general, our algorithms perform efficiently in practice, especially considering that schema design happens rarely compared to frequent updates at operational level. In fact, our optimizations translate into reductions of costly update overheads. Our improvements are larger than previous optimizations using minimal keys only, confirming that FDs constitute the larger bottleneck for integrity maintenance compared to minimal keys. This justifies the development of our logical normalization framework that is tightly coupled to integrity maintenance in practice.

Future work will address optimum covers that use sizes of keys and FDs [25], rather than their numbers. Here, the trade-off between expensive computations of optimum covers and additional performance gains will be interesting to analyze. For future work, it will also be interesting to investigate how schema designs can be optimized further after the database has become operational and information about workload patterns of updates and queries are available. Such knowledge is not assumed to be input for classical schema design, such normalization into BCNF or 3NF. Higher normal forms [9], such as 4NF [11], 5NF [32] and Inclusion Dependency Normal Form [21], will also be investigated. Interestingly,

notions of minimal and optimal covers do not exist for the dependencies they are based on, such as multivalued, join and inclusion dependencies. It is also important to extend the work to other data models, including incomplete [19, 34], temporal [14], Web [3, 10, 35], uncertain [22] and graph data [2, 31].

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Data	Alg	Size	ВС	3NF	BCNF distribution	3NF distribution
abalone	iConf-fk	26	22	4	[3:1,2:12,1:9]	[4:1,2:1,1:2]
	iConf-f <k< td=""><td>23</td><td>18</td><td>5</td><td>[2:11,1:7]</td><td>[4:1,2:1,1:3]</td></k<>	23	18	5	[2:11,1:7]	[4:1,2:1,1:3]
	iConf->kf	26	22	4	[3:1,2:12,1:9]	[4:1,2:1,1:2]
	iConf-f	23	18	5	[2:11,1:7]	[4:1,2:1,1:3]
ncvoter	iConf-fk	166	145	21	[2:27,1:118]	[3:1,2:2,1:18]
	iConf-f <k< td=""><td>164</td><td>143</td><td>21</td><td>[3:1,2:28,1:114]</td><td>[3:1,2:2,1:18]</td></k<>	164	143	21	[3:1,2:28,1:114]	[3:1,2:2,1:18]
	iConf->kf	163	142	21	[2:28 1:114]	[5:1,3:1,2:3,1:16]
	iConf-f	166	145	21	[3:1,2:27,1:117]	[3:1,2:2,1:18]
lineitem	iConf-fk	590	560	30	[15:1,10:1,6:3,5:3,4:5,3:23,2:105,1:419]	[4:1, 2:9, 1:20]
	iConf-f <k< td=""><td>602</td><td>573</td><td>29</td><td>$[15{:}1,10{:}1,6{:}4\ 5{:}3,4{:}6,3{:}23,2{:}106,1{:}429]$</td><td>[4:1, 2:9, 1:19]</td></k<>	602	573	29	$[15{:}1,10{:}1,6{:}4\ 5{:}3,4{:}6,3{:}23,2{:}106,1{:}429]$	[4:1, 2:9, 1:19]
	iConf->kf	558	528	30	[15:1,10:1,6:3,5:3,4:5,3:21,2:95,1:399]	[10:1,8:2,6:1,5:1,4:1,3:2,2:9,1:13]
	iConf-f	596	567	29	$[15{:}1,\!10{:}1,\!6{:}4,\!5{:}3,\!4{:}5,\!3{:}23,\!2{:}105,\!1{:}425]$	[4:1,2:9,1:19]

Table 4: Properties of Designs for Optimized Strategies

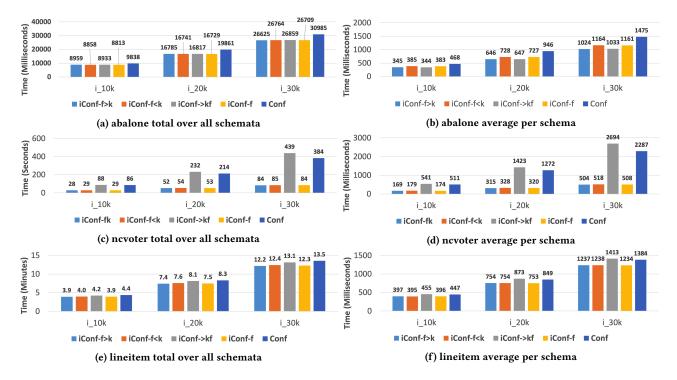


Figure 6: Optimizations of insertion overheads: total time for all schemata and average time per schema with combined covers

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A PROOFS

LEMMA A.1 (LEMMA 3.7 RESTATED).

Let (R, Σ) denote a set Σ of FDs over relation schema R. Let Ω denote a 3NF-substructure of (R, Σ) . Then Ω is 3NF update complete if and only if all of the following hold:

(1) $\Sigma \subseteq \Omega^+$

(2) (R, Σ) is in Third Normal Form

PROOF. "\(\Lefta\)": We show first that the two conditions (1) and (2) are sufficient for 3NF update completeness to hold.

Let r denote a relation over R that satisfies Σ , and let $t \in dom(R)$ and $t' \in r$ be such that for all $X \to Y \in \Omega$, i) if $X \to R \in \Omega^+$, then $X \nsubseteq ag(t,t')$ and ii) if $Y - X \subseteq P_\Omega$ and $X \subseteq ag(t,t')$, then $Y \subseteq ag(t,t')$. We need to show that $r \cup \{t\}$ satisfies Σ .

Let $X \to Y \in \Sigma$. Assume that $X \subseteq ag(t,t')$, otherwise there is nothing else to show. Since (2) holds it follows that $X \to R \in \Sigma^+$ or $Y - X \subseteq P_{\Sigma}$. Since (1) holds we have $\Omega^+ \subseteq (\Sigma^+)^+ = \Sigma^+ \subseteq (\Omega^+)^+ = \Omega^+$ and therefore $\Omega^+ = \Sigma^+$. It follows that a) $X \to R \in \Omega^+$ or b) $Y - X \subseteq P_{\Omega}$. Since $X \subseteq ag(t,t')$, our assumption dictates that $Y \subseteq ag(t,t')$, which is what we needed to show. Consequently, $r \cup \{t\}$ satisfies Σ and we have shown that 3NF udpate completeness holds.

"⇒": We show now that 3NF update completeness entails (1) and (2). We show first that not (2) implies that 3NF update completeness does not hold.

Assume that (R, Σ) is not in Third Normal Form. Then there is some $X \to Y \in \Sigma^+$ such that $X \to R \notin \Sigma^+$ and $Y - X \nsubseteq P_\Sigma$. In particular, $X \to Y \notin \Omega$. Let $r := \{t'\}$ be a single tuple relation and $t \in dom(R)$ such that for all $A \in R$, t'(A) = t(A) if and only if $A \in X_\Omega^+$. Let $Z \to Y \in \Omega$. Assume $Z \to R \in \Omega^+$. If $Z \subseteq X_\Omega^+$, then $X \to Z \in \Omega^+$, and thus $X \to R \in \Omega^+$. Hence, $X \to R \in \Sigma^+$, a contradiction. Consequently, $Z \nsubseteq X_\Omega^+$, which means $Z \nsubseteq ag(t, t')$. Assume now that $Y - Z \subseteq P_\Omega$ and $Z \subseteq ag(t, t')$. We show this case cannot occur. Indeed, this case would mean that $Z \subseteq X_\Omega^+$ by construction. This meant $X \to Z \in \Omega^+$ and therefore $X \to Y \in \Omega^+$, a contradiction.

We show secondly that not (1) implies that 3NF update completeness does not hold. Assume that there is some $X \to Y \in \Sigma - \Omega^+$.

Let $r := \{t'\}$ be a single tuple relation and $t \in dom(R)$ such that for all $A \in R$, t'(A) = t(A) if and only if $A \in X_{\Omega}^+$. We show that

(i) $r \cup \{t\}$ satisfies Ω and (ii) $r \cup \{t\}$ does not satisfy Σ because $X \to Y$ is violated. Consequently, Ω would not satisfy 3NF update completeness.

Firstly, we show (i). Let $Z \to Y \in \Omega$. If t,t' do not match on all attributes in Z, then $\{t,t'\}$ satisfies $Z \to Y$. Otherwise, t,t' match on all attributes in Z, which means that $Z \subseteq X_{\Omega}^+$, and therefore $X \to Z \in \Omega^+$. Consequently, $X \to Y \in \Omega^+$, and thus also $Y \subseteq X_{\Omega}^+$. By construction, $\{t,t'\}$ must have matching values on all the attributes in Y. Hence, $T \cup \{t\}$ satisfies $T \to Y$. As $T \to Y$ was chosen as an arbitrary FD from $T \to X$ 0, $T \to X$ 1 satisfies $T \to Y$ 2. This shows (i).

Secondly, we show (ii). Indeed, $X \subseteq X_{\Omega}^+$ but $Y \not\subseteq X_{\Omega}^+$ since $X \to Y \not\in \Omega^+$. Consequently, by construction, t, t' have matching values on all attributes in X but have a non-matching value on some attribute in Y. Hence, $r \cup \{t\}$ violates $X \to Y$. This concludes the proof.

COROLLARY A.2 (COROLLARY 3.8 RESTATED). If (R, Σ) is in 3NF, then the 3NF-core $\mathcal{K}_{\Sigma} \cup \mathcal{F}_{\Sigma}$ is an intransitive composite object for (R, Σ) .

PROOF. The 3NF-core is always a 3NF-substructure. If (R, Σ) is in 3NF, then $\Sigma \subseteq (\mathcal{K}_{\Sigma} \cup \mathcal{F}_{\Sigma})^+$. Consequently, (1) and (2) of Lemma 3.7 are satisfied by $\Omega := \mathcal{K}_{\Sigma} \cup \mathcal{F}_{\Sigma}$. Hence, Ω is an intransitive composite object for (R, Σ) by Lemma 3.7.

THEOREM A.3 (THEOREM 3.9 RESTATED). For all relation schemata R, and all sets Σ of FDs over R the following holds: (R, Σ) is in 3NF if and only if (R, Σ) is in iCONF.

PROOF. " \Rightarrow ": Assume that (R, Σ) is in 3NF. We show that (R, Σ) is in iCONF.

Indeed, Corollary 3.8 shows that $\Omega := \mathcal{K}_{\Sigma} \cup \mathcal{F}_{\Sigma}$ is an intransitive composite object. By definition, (R, Σ) is in iCONF.

" \Leftarrow ": We show the contraposition. Hence, we assume that (R, Σ) is not in 3NF, and need to show that (R, Σ) is not in iCONF.

The 3NF-core $\Omega := \mathcal{K}_{\Sigma} \cup \mathcal{F}_{\Sigma}$ is a 3NF-substructure, but since (R, Σ) is not in 3NF, Lemma 3.7 shows that Ω is not 3NF update complete. Consequently, Ω is not an intransitive composite object, so (R, Σ) is not in iCONF.