



Concurrent Composition of Differential Privacy

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TL; DR: We initiate a study of the *concurrent* composition properties of *interactive* differentially private mechanisms, and derived the *optimal* composition bound for pure interactive DP mechanisms.

Background: DP under Composition

- Goal: analyze the privacy loss under the composition of multiple different DP mechanisms on the same dataset.
- Examples of existing DP composition theorems: Basic Composition, Advanced Composition, Optimal Composition, Moment Accountant, etc.

Motivation

- Existing composition theorems: assume that the underlying DP mechanisms are "one-shot" algorithms.
- We want to compose interactive mechanisms, e.g., Sparse Vector Technique (SVT).

Interactive DP under Composition

- There could be more than one composition operations for interactive mechanisms.
- Sequential Composition:** all of the queries to the current mechanism must be completed before the session with another mechanism can be spawned.



- Concurrent Composition:** multiple interactions can be spawned and be executed simultaneously, queries to the mechanisms can be arbitrarily interleaved with each other.



- Unfortunately, none of the existing composition theorems for non-interactive DP can be directly applied to the setting of concurrent compositions.

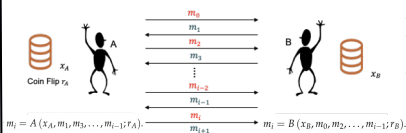
References

Kairouz, P., Oh, S., & Viswanath, P. (2015, June). The composition theorem for differential privacy. In International conference on machine learning (pp. 1376-1385). PMLR.

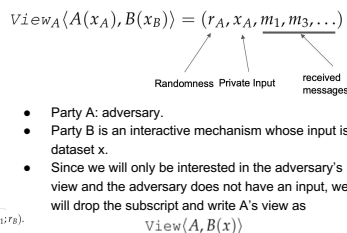
Murtagh, J., & Vadhan, S. (2016, January). The complexity of computing the optimal composition of differential privacy. In Theory of Cryptography Conference (pp. 157-175). Springer, Berlin, Heidelberg.

Interactive Protocol

- Interactive protocol between two parties A and B
 - Viewing each party as a potentially randomized function.
 - (private input, received messages, random coins) => Next message to be sent out.



View of a Party



Formalizing Interactive Differential Privacy

The interactive differentially privacy as a **type of interactive protocol** between an adversary (without any computational limitations) and an interactive mechanism of special properties.

Definition 4 (Interactive Differential Privacy). A randomized algorithm \mathcal{M} is (ϵ, δ) -differentially private interactive mechanism if for every pair of adjacent datasets x, x' , for every adversary algorithm \mathcal{A} , for every possible output set $T \subseteq \text{Range}(View(\mathcal{A}, \mathcal{M}(\cdot)))$ we have

$$\Pr[View(\mathcal{A}, \mathcal{M}(x)) \in T] \leq e^\epsilon \Pr[View(\mathcal{A}, \mathcal{M}(x')) \in T] + \delta$$

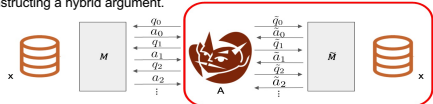
Group Privacy-like Bound for Concurrent Composition

Result: The concurrent composition of k (ϵ, δ) interactive DP mechanisms has a group privacy-like bound $(k\epsilon, k\epsilon k\delta)$.

Proof Idea: Suppose interactive mechanisms $M_0, \dots, M_{[k-1]}$ are each (ϵ_i, δ_i) -differentially private. View \mathcal{A} and $M_0, \dots, M_{[k-1]}$ as a combined adversary \mathcal{A}' , we can show that

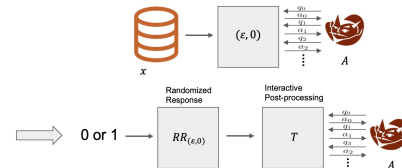
$$\Pr[View(\mathcal{A}, \text{ConComp}(M_0(x'), \dots, M_{[k-1]}(x'), M_1(x), \dots, M_{[k-1]}(x))) \in S] \leq e^{\epsilon'} \Pr[View(\mathcal{A}, \text{ConComp}(M_0(x), \dots, M_{[k-1]}(x), M_1(x), \dots, M_{[k-1]}(x))) \in S] + \delta_i$$

which can be used for constructing a hybrid argument.



Optimal Concurrent Composition Bound for Pure DP

Result: Every interactive $(\epsilon, 0)$ -DP mechanisms can be simulated by the post-processing of **randomized response** $RR_{(\epsilon, 0)}$ (a non-interactive mechanism).
=> Optimal (approx. DP, Renyi DP, f-DP, etc) bounds for concurrent composition of interactive pure DP mechanisms = optimal bounds for composition of **non-interactive** pure DP mechanisms.



- (ϵ, δ) -DP version of Randomized Response:

$$RR_{(\epsilon, \delta)} : \{0, 1\} \rightarrow \{0, 1, 'Iam0', 'Iam1'\}$$

$$\Pr[RR_{(\epsilon, \delta)}(0) = 'Iam0'] = \delta \quad \Pr[RR_{(\epsilon, \delta)}(1) = 'Iam0'] = 0$$

$$\Pr[RR_{(\epsilon, \delta)}(0) = 0] = (1 - \delta) \cdot \frac{1}{1 + e^\epsilon} \quad \Pr[RR_{(\epsilon, \delta)}(1) = 0] = (1 - \delta) \cdot \frac{1}{1 + e^{-\epsilon}}$$

$$\Pr[RR_{(\epsilon, \delta)}(0) = 1] = (1 - \delta) \cdot \frac{1}{1 + e^{-\epsilon}} \quad \Pr[RR_{(\epsilon, \delta)}(1) = 1] = (1 - \delta) \cdot \frac{1}{1 + e^\epsilon}$$

$$\Pr[RR_{(\epsilon, \delta)}(0) = 'Iam1'] = 0 \quad \Pr[RR_{(\epsilon, \delta)}(1) = 'Iam1'] = \delta$$

- Post-processing preserves differential privacy
=> To analyze the concurrent composition of arbitrary pure interactive DP mechanisms, it suffices to analyze the composition of randomized responses of the same parameters (analogous to the proof strategy in [KOV15] and [MV16]).
- Therefore, the optimal bound for concurrent composition of pure interactive DP is the same as the optimal bound for composing non-interactive counterpart.

Future Work

- We empirically test whether the Optimal Composition Theorems can be extended to the concurrent composition of approximate DP for 3-message interactive mechanisms with 1-bit message. In all our trials, we find a feasible interactive post-processing algorithm.
- We therefore conjecture that the concurrent composition of interactive DP mechanisms may still have the same bound as the composition for non-interactive DP.