1 Proof

Lemma 1.1 (The modified Heuristic Miner is $n\epsilon_1 \cdot \frac{cap}{\tau \# traces}$ Differential Private).

Proof. We apply n times the AboveThreshold-Mechanism with noise parameter ϵ_1 , to all elements in the Event-Log, so that we can exclude edges that were visited in less than τ number of traces in the Event-log D.

For the *i*th invocation of AboveThreshold, we have a sequence $[(\bot)_{i=1,...,m_i-1}, \top]$ (for some $m_i \in \mathbb{N}$). We then compute BelowThreshold for the same queries and noise parameter ϵ_1 , until we get a \bot , so a sequence looks like $[(\top)_{i=1,...,m_{i+1}-1}, \bot]$. So if we get \top from the AboveThreshold-Mechanism, we have to get \bot for that same query with the BelowThreshold-Mechanism. The BelowThreshold-Mechanism provides the same privacy as the AboveThreshold-Mechanism, which can be proven like the AboveThreshold-Mechanism in [1]. These results can be used without any additional leakage, due to the post-processing theorem.

From [1] we know that the AboveThreshold-Mechanism is ϵ -DP, with ϵ being the noise parameter. With the use of the basic composition theorem we also know that executing a ϵ -DP algorithm n_1 times, we get that it is $n_1\epsilon$ -DP. Therefore executing the AboveThreshold-Mechanism n times, with noise parameter ϵ_1 , we get that $n\epsilon_1$ -DP holds.

Next, the algorithm compute the frequency for every edge that evaluated with \top . If an edge has a frequency greater than cap, then we will cap it at cap. Hence, we minimize leakage about which processes are carried out the most. Additionally, we filter out edges that occur in less than a $\tau \in [0,1]$ fraction of traces in the Event-Log. We consider the queries of the following form: $q_e(D) :=$ "In which frequency does edge e to occur in the Event-log D?" Each of these queries has sensitivity $cap/(\tau \# traces)$.

Therefore $n\epsilon_1 \cdot \frac{cap}{\tau \# traces}$ -DP holds, due to the basic composition theorem. Everything that is done with this graph has no additional leakage due to the post-processing theorem.

References

[1] Cynthia Dwork, A.R.: The Algorithmic Foundations of Differential Privacy (2014), https://www.cis.upenn.edu/aaroth/Papers/privacybook.pdf