

Verifying Snapping Mechanism

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1 Formalization

Definition 1 ($\text{Snap}(\mu, a) : \text{Distr}(U) \rightarrow A \rightarrow \text{Distr}(B)$)

The ideal Snapping mechanism $\text{Snap}(\mu, a)$ is defined as:

$$u \xleftarrow{\$} \mu; y = \ln(u); s \xleftarrow{\$} \{-1, 1\}; y' = s * y; z = \frac{y'}{\epsilon}; x = f(a); w = x + z; w' = \lfloor w \rfloor_{\Lambda}; r = \text{clamp}_B(w')$$

where f is the query function over input $a \in A$, ϵ is the privacy budget and S sampled from $\{-1, +1\}$ with Bernoulli(0.5).

Definition 2

Let $\epsilon \leq 0$. The ϵ -DP divergence $\Delta_{\epsilon}(\mu_1, \mu_2)$ between two sub-distributions $\mu_1 \in$

Definition 3 (ϵ - dilation)

Let $\epsilon \geq 0$. The ϵ -dilation $D_{\epsilon}(\mu_1, \mu_2)$ between two sub-distributions $\mu_1 \in \text{Distr}(U)$, $\mu_2 \in \text{Distr}(U)$ is defined as:

$$\sup_{E \in \mathcal{U}} \left(\Pr_{x \leftarrow \mu_1} [x \in E] - \exp(\epsilon) \Pr_{x \leftarrow \mu_2} [x \in \exp(-\epsilon) \cdot E] \right)$$

Proposition 1 ((ϵ, δ) -differential privacy)

For every pair of sub-distributions $\mu_1 \in \text{Distr}(U)$, $\mu_2 \in \text{Distr}(U)$, s.t.

$$D_{\epsilon}(\mu_1, \mu_2) \leq \delta,$$

The snapping mechanism $\text{Snap}(\mu, a) : \text{Distr}(U) \rightarrow A \rightarrow \text{Distr}(B)$ is (ϵ, δ) - differentially private w.r.t. an adjacency relation Φ for every two adjacent inputs a, a' and μ_1, μ_2

Proof. Followed directly by unfolding the Snap mechanism.

$$\begin{aligned} \Pr_{x \leftarrow \text{Snap}(\mu_1, a)} [x = e] &= \Pr_{u \leftarrow \mu_1} [\lfloor f(a) + \frac{S \cdot \log(u)}{\epsilon} \rfloor_{\Lambda} = e] \\ &= \Pr_{u \leftarrow \mu_1} [u \in [\frac{\exp((e - \frac{\Lambda}{2} - f(a))\epsilon)}{S}, \frac{\exp((e + \frac{\Lambda}{2} - f(a))\epsilon)}{S}]] \\ &\leq \exp(\epsilon) \Pr_{u \leftarrow \mu_2} [u \in \exp(-\epsilon) [\frac{\exp((e - \frac{\Lambda}{2} - f(a))\epsilon)}{S}, \frac{\exp((e + \frac{\Lambda}{2} - f(a))\epsilon)}{S}]] \\ &= \exp(\epsilon) \Pr_{u \leftarrow \mu_2} [\lfloor f(a') + \frac{S \cdot \log(u)}{\epsilon} \rfloor_{\Lambda} = e] \\ &= \exp(\epsilon) \Pr_{x \leftarrow \text{Snap}(\mu_2, a')} [x = e] \end{aligned}$$

□

Definition 4 ((ϵ, δ) - Dilation lifting)

Two sub-distributions $\mu_1 \in \text{Distr}(U_1)$, $\mu_2 \in \text{Distr}(U_2)$ are related by the (ϵ, δ) - dilation lifting of $\Psi \subseteq U_1 \times U_2$, written $\mu_1 \Psi^{d(\epsilon, \delta)} \mu_2$, if there exist two witness sub-distributions $\mu_L \in \text{Distr}(U_1 \times U_2)$ and $\mu_R \in \text{Distr}(U_1, U_2)$ s.t.:

1. $\pi_1(\mu_L) = \mu_1$ and $\pi_2(\mu_R) = \mu_2$;
2. $\text{supp}(\mu_L) \subseteq \Psi$ and $\text{supp}(\mu_R) \subseteq \Psi$; and
3. $D_\epsilon(\mu_L, \mu_R) \leq \delta$.

It is easy to see that two sub-distributions μ_1 and μ_2 are related by $=^{d(\epsilon, \delta)}$ iff $D_\epsilon(\mu_1, \mu_2) \leq \delta$. Therefore, the Snap mechanism $\text{Distr}(U) \rightarrow A \rightarrow \text{Distr}(B)$ is (ϵ, δ) -differentially private w.r.t. and adjacency relation Φ and μ_1, μ_2 iff:

$$\mu_1 =^{d(\epsilon, \delta)} \mu_2$$

for every two adjacent inputs a and a' .

Definition 5 ((ϵ, δ) - lifting [1])

Two sub-distributions $\mu_1 \in \text{Distr}(U_1), \mu_2 \in \text{Distr}(U_2)$ are related by the (ϵ, δ) - dilation lifting of $\Psi \subseteq U_1 \times U_2$, written $\mu_1 \Psi^{\#(\epsilon, \delta)} \mu_2$, if there exist two witness sub-distributions $\mu_L \in \text{Distr}(U_1 \times U_2)$ and $\mu_R \in \text{Distr}(U_1, U_2)$ s.t.:

1. $\pi_1(\mu_L) = \mu_1$ and $\pi_2(\mu_R) = \mu_2$;
2. $\text{supp}(\mu_L) \subseteq \Psi$ and $\text{supp}(\mu_R) \subseteq \Psi$; and
3. $\Delta_\epsilon(\mu_L, \mu_R) \leq \delta$.

Theorem 2

if $\mu_1 \Psi^{d(\epsilon, \delta)} \mu_2$, then we can prove $\text{Snap}(\mu_1, a) \Psi^{\#(\epsilon, \delta)} \text{Snap}(\mu_2, a')$, w.r.t. an adjacent relation Φ for every $a \Phi a'$

Theorem 3

The coupling of two Snap mechanisms: $\text{Snap}(\mu_1, a_1), \text{Snap}(\mu_2, a_2)$.

$$\overline{u_1 \xleftarrow{\$} \mu \sim_{e^\epsilon u_2, 0} u_1 \xleftarrow{\$} \mu : T \Rightarrow u_1 = e^\epsilon u_2}$$

$$\overline{y_1 = \ln(u_1) \sim_{\epsilon^2, 0} y_2 = \ln(u_2) : u_1 = e^\epsilon u_2 \Rightarrow y_1 = \epsilon + y_2}$$

$$\overline{s_1 \xleftarrow{\$} \mu \sim_{0, 0} s_1 \xleftarrow{\$} \mu : T \Rightarrow s_1 = s_2}$$

$$\overline{y'_1 = s_1 * y_1 \sim_{\epsilon^2, 0} y'_2 = s_2 * y_2 : s_1 = s_2 \wedge y_1 = e^\epsilon u_2 \Rightarrow y'_1 = \epsilon + y'_2}$$

$$\overline{z_1 = \frac{y'_1}{\epsilon} \sim_{\epsilon, 0} z_2 = \frac{y'_2}{\epsilon} : y'_1 = \epsilon + y'_2 \Rightarrow z_1 = z_2 + 1}$$

$$\overline{x_1 = f(a_1) \sim_{0, 0} x_2 = f(a_2) : x_1 = x_2 + 1 \Rightarrow a_1 = a_2 + 1}$$

$$\overline{w_1 = x_1 + z_1 \sim_{0, 0} w_2 = x_2 + z_2 : x_1 + 1 = x_2 \wedge z_1 = z_2 + 1 \Rightarrow w_1 = w_2}$$

$$\overline{w'_1 = \lfloor w_1 \rfloor_\Lambda \sim_{0, 0} w'_2 = \lfloor w_2 \rfloor_\Lambda : w_1 = w_2 \Rightarrow w'_1 = w'_2}$$

$$\overline{r_1 = \text{clamp}_B(w'_1) \sim_{0, 0} r_2 = \text{clamp}_B(w'_2) : w'_1 = w'_2 \Rightarrow r_1 = r_2}$$

Figure 1: coupling of two Snap mechanisms: $\text{Snap}(\mu_1, a_1)$, $\text{Snap}(\mu_2, a_2)$

References

- [1] Gilles Barthe, Marco Gaboardi, Benjamin Grégoire, Justin Hsu, and Pierre-Yves Strub. Proving differential privacy via probabilistic couplings. In *LICS 2016*.