

# Proof of DP - Bayesian Inference

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If we want the posterior distribution from the Bayesian inference is  $\epsilon$  - differential privacy, this mechanism should satisfy:

$$\frac{Pr[BayesInfer_{DP}(d_1) \in S]}{Pr[BayesInfer_{DP}(d_2) \in S]} \leq e^\epsilon,$$

where  $BayesInfer_{DP}(d)$  is a differentially private Bayesian inference mechanism, and  $d_1, d_2$  is a pair of adjacent observed data sets. Since in our model, this inference is based on a bias  $p$  with beta prior distribution:  $p \sim Beta(\alpha_0, \beta_0)$  and an observed data set  $d \sim B(n, p)$ , which result in a posterior distribution on  $p \sim Beta(\alpha_0 + k, \beta_0 + l)$  where  $k$  and  $l$  is the number of 1 and 0 in  $d$ . As a result, the requirement above can be rewrote as:

$$\frac{Pr[Beta(\alpha_0 + k_1^*, \beta_0 + l_1^*) \in S]}{Pr[Beta(\alpha_0 + k_2^*, \beta_0 + l_2^*) \in S]} \leq e^\epsilon,$$

where  $k_1^*$  and  $l_1^*$  is the number of 1 and 0 in  $d_1$  after protection, the same with  $k_2^*$  and  $l_2^*$ . The equation above is equivalent to:

$$\frac{Pr[\langle \alpha_0 + k_1^*, \beta_0 + l_1^* \rangle = S]}{Pr[\langle \alpha_0 + k_2^*, \beta_0 + l_2^* \rangle = S]} \leq e^\epsilon.$$

Under Laplace mechanism, sensitivity of  $k$  and  $l$  is both 1 we have  $k^* = k + Lap(\frac{\epsilon}{2})$ , and  $l^* = l + Lap(\frac{\epsilon}{2})$ . Suppose  $S = \langle S_1, S_2 \rangle$ , where  $S_1 - \alpha_0 - k \sim Lap(\frac{\epsilon}{2})$ ,  $S_2 - \beta_0 - l \sim Lap(\frac{\epsilon}{2})$ . Then,

$$Pr[\langle \alpha_0 + k_1^*, \beta_0 + l_1^* \rangle = S] = exp(-(\alpha_0 - k_1)\frac{\epsilon}{2})exp(-(S_2 - \beta_0 - l_1)\frac{\epsilon}{2}),$$

and

$$Pr[\langle \alpha_0 + k_2^*, \beta_0 + l_2^* \rangle = S] = exp(-(S_1 - \alpha_0 - k_2)\frac{\epsilon}{2}) * exp(-(S_2 - \beta_0 - l_2)\frac{\epsilon}{2}).$$

We can finally get:

$$\frac{Pr[\langle \alpha_0 + k_1^*, \beta_0 + l_1^* \rangle = S]}{Pr[\langle \alpha_0 + k_2^*, \beta_0 + l_2^* \rangle = S]} = exp(\epsilon),$$

i.e.

$$\frac{Pr[BayesInfer_{DP}(d_1) \in S]}{Pr[BayesInfer_{DP}(d_2) \in S]} \leq e^\epsilon,$$