

**Theorem** for privacy parameters  $\epsilon, \delta$ , database  $X$ , domain  $R$  and sensetivity-1 utility function  $u$  :

if the domain  $R$  is such that  $|R| \leq \sqrt{\frac{\beta}{\delta}}$  and the gap between  $OPT_u(x)$  and the second highest score is less than  $\frac{\log(\frac{1}{\delta\beta})}{\epsilon}$  then the probability that the exponential mechanism will fail to output  $OPT_u(x)$  is less than  $\beta$

**proof** we want to use the exponential mechanism in a way that in every case that A-dist outputs a value the exponential mechanism will output that value with at least the same probability  
ergo from the theorem about exponential mechanism

$$Pr \left[ u(M_E(x, u, R)) \leq OPT_u(x) - \frac{2\Delta u}{\epsilon} \left( \ln \left( \frac{|R|}{|R_{OPT}|} \right) + t \right) \right] \leq e^{-t}$$

when  $f$  is  $\frac{\log(\frac{1}{\delta\beta})}{\epsilon}$ -stable, in our case meaning that this is the minimum real-gap between  $OPT_u(x)$  to the second highest score, then A-dist fails to return a value in probability  $\beta$

so we want to calculate the probability that if that is the case the exponential mechanism will output something other then  $OPT_u(x)$

we can assume that  $|R_{OPT}| = 1$  and that  $\Delta u = 1$  and set  $e^{-t} = \beta$

so we want to get

$$Pr \left[ u(M_E(x, u, R)) \leq OPT_u(x) - \frac{\log \left( \frac{1}{\delta\beta} \right)}{\epsilon} \right] \leq Pr \left[ u(M_E(x, u, R)) \leq OPT_u(x) - \frac{2}{\epsilon} (\ln(|R|) + t) \right] \leq$$

or in other words

$$-\log \left( \frac{1}{\delta\beta} \right) \leq -2 (\ln(|R|) + t) \Rightarrow 2t \leq \log \left( \frac{1}{\delta\beta} \right) - 2\ln(|R|) = \ln \left( \frac{1}{\delta\beta|R|^2} \right)$$

$$\Rightarrow e^{2t} \leq \frac{1}{\delta\beta|R|^2} \Rightarrow |R|^2 \leq \frac{1}{\delta\beta e^{2t}} = \frac{e^{-t}e^{-t}}{\delta\beta} = \frac{\beta}{\delta} \Rightarrow |R| \leq \sqrt{\frac{\beta}{\delta}} \quad \blacksquare$$