

Differential Privacy using Quasi-Borel Spaces

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Abstract

This entry formalizes differential privacy using quasi-Borel spaces. In general, differential privacy is discussed using measurable spaces. Sato and Katsumata showed that quasi-Borel spaces are also applied to formulate differential privacy [1]. We formalize basic definitions and properties of differential privacy using quasi-Borel spaces, and show two examples: randomized response and the naive report noisy max algorithm.

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theory *DP-QBS*

imports *Differential-Privacy.Differential-Privacy-Divergence*
Differential-Privacy.Differential-Privacy-Standard
S-Finite-Measure-Monad.Monad-QuasiBorel

begin

declare *qbs-morphism-imp-measurable*[*measurable-dest*]

1 Definitions

Details of differential privacy using quasi-Borel spaces are found at [1]

1.1 Divergence for Differential Privacy using QBS

definition $DP\text{-}qbs\text{-}divergence :: 'a\ qbs\text{-}measure \Rightarrow 'a\ qbs\text{-}measure \Rightarrow real \Rightarrow ereal$
 $(DP'\text{-}divergence_Q)$ **where**
 $DP\text{-}qbs\text{-}divergence\text{-}qbs\text{-}l: DP\text{-}divergence_Q\ p\ q\ e \equiv DP\text{-}divergence\ (qbs\text{-}l\ p)\ (qbs\text{-}l\ q)\ e$

abbreviation $DP\text{-}qbs\text{-}inequality\ (DP'\text{-}inequality_Q)$ **where**
 $DP\text{-}qbs\text{-}inequality\ p\ q\ \varepsilon\ \delta \equiv DP\text{-}divergence_Q\ p\ q\ \varepsilon \leq ereal\ \delta$

lemmas $DP\text{-}qbs\text{-}divergence\text{-}def = DP\text{-}qbs\text{-}divergence\text{-}qbs\text{-}l[simplified\ DP\text{-}divergence\text{-}SUP]$

lemma $DP\text{-}qbs\text{-}divergence\text{-}nonneg[simp]: 0 \leq DP\text{-}divergence_Q\ p\ q\ e$
 $\langle proof \rangle$

lemma $DP\text{-}qbs\text{-}divergence\text{-}le\text{-}ereal\text{-}iff:$
 $DP\text{-}divergence_Q\ p\ q\ \varepsilon \leq ereal\ \delta \longleftrightarrow (\forall A \in sets\ (qbs\text{-}l\ p). measure\ (qbs\text{-}l\ p)\ A -$
 $exp\ \varepsilon * measure\ (qbs\text{-}l\ q)\ A \leq \delta)$
 $\langle proof \rangle$

corollary $DP\text{-}qbs\text{-}divergence\text{-}le\text{-}ereal\text{-}dest:$
assumes $DP\text{-}divergence_Q\ p\ q\ \varepsilon \leq ereal\ \delta$
shows $measure\ (qbs\text{-}l\ p)\ A \leq exp\ \varepsilon * measure\ (qbs\text{-}l\ q)\ A + \delta$
 $\langle proof \rangle$

corollary $DP\text{-}qbs\text{-}divergence\text{-}le\text{-}erealI:$
assumes $\bigwedge A. A \in sets\ (qbs\text{-}l\ p) \implies measure\ (qbs\text{-}l\ p)\ A \leq exp\ \varepsilon * measure\ (qbs\text{-}l\ q)\ A + \delta$
shows $DP\text{-}divergence_Q\ p\ q\ \varepsilon \leq ereal\ \delta$
 $\langle proof \rangle$

lemma $DP\text{-}qbs\text{-}divergence\text{-}zero:$
assumes $p \in monadP\text{-}qbs\ X$
and $q \in monadP\text{-}qbs\ X$
and $DP\text{-}inequality_Q\ p\ q\ 0\ 0$
shows $p = q$
 $\langle proof \rangle$

lemma $DP\text{-}qbs\text{-}divergence\text{-}antimono: a \leq b \implies DP\text{-}divergence_Q\ p\ q\ b \leq DP\text{-}divergence_Q\ p\ q\ a$
 $\langle proof \rangle$

lemma $DP\text{-}qbs\text{-}divergence\text{-}refl[simp]: DP\text{-}divergence_Q\ p\ p\ 0 = 0$
 $\langle proof \rangle$

lemma $DP\text{-}qbs\text{-}divergence\text{-}refl'[simp]: 0 \leq e \implies DP\text{-}divergence_Q\ p\ p\ e = 0$
 $\langle proof \rangle$

lemma *DP-qbs-divergence-trans'*:
assumes $DP\text{-inequality}_Q\ p\ q\ \varepsilon\ \delta$
and $DP\text{-inequality}_Q\ q\ l\ \varepsilon'\ 0$
shows $DP\text{-inequality}_Q\ p\ l\ (\varepsilon + \varepsilon')\ \delta$
 $\langle proof \rangle$

lemmas $DP\text{-qbs-divergence-trans} = DP\text{-qbs-divergence-trans}'[\mathbf{where}\ \delta=0]$

proposition *DP-qbs-divergence-compose*:
assumes $[qbs, measurable]: p \in monadP\text{-qbs}\ X\ q \in monadP\text{-qbs}\ X\ f \in X \rightarrow_Q monadP\text{-qbs}\ Y\ g \in X \rightarrow_Q monadP\text{-qbs}\ Y$
and $dp1: DP\text{-divergence}_Q\ p\ q\ \varepsilon \leq ereal\ \delta$
and $dp2: \bigwedge x. x \in qbs\text{-space}\ X \implies DP\text{-divergence}_Q\ (f\ x)\ (g\ x)\ \varepsilon' \leq ereal\ \delta'$
and $[arith]: 0 \leq \varepsilon\ 0 \leq \varepsilon'$
shows $DP\text{-divergence}_Q\ (p \ggg f)\ (q \ggg g)\ (\varepsilon + \varepsilon') \leq ereal\ (\delta + \delta')$
 $\langle proof \rangle$

corollary *DP-qbs-divergence-dataprocessing*:
assumes $[qbs]: p \in monadP\text{-qbs}\ X\ q \in monadP\text{-qbs}\ X\ f \in X \rightarrow_Q monadP\text{-qbs}\ Y$
and $dp: DP\text{-divergence}_Q\ p\ q\ \varepsilon \leq ereal\ \delta$
and $[arith]: 0 \leq \varepsilon$
shows $DP\text{-divergence}_Q\ (p \ggg f)\ (q \ggg f)\ \varepsilon \leq ereal\ \delta$
 $\langle proof \rangle$

lemma *DP-qbs-divergence-additive*:
assumes $[qbs]: p \in monadP\text{-qbs}\ X\ q \in monadP\text{-qbs}\ X\ p' \in monadP\text{-qbs}\ Y\ q' \in monadP\text{-qbs}\ Y$
and $div1: DP\text{-divergence}_Q\ p\ q\ \varepsilon \leq ereal\ \delta$
and $div2: DP\text{-divergence}_Q\ p'\ q'\ \varepsilon' \leq ereal\ \delta'$
and $[arith]: 0 \leq \varepsilon\ 0 \leq \varepsilon'$
shows $DP\text{-divergence}_Q\ (p \otimes_{Q_{mes}} p')\ (q \otimes_{Q_{mes}} q')\ (\varepsilon + \varepsilon') \leq ereal\ (\delta + \delta')$
 $\langle proof \rangle$

corollary *DP-qbs-divergence-strength*:
assumes $[qbs]: p \in monadP\text{-qbs}\ X\ q \in monadP\text{-qbs}\ X\ x \in qbs\text{-space}\ Y$
and $dp: DP\text{-divergence}_Q\ p\ q\ \varepsilon \leq ereal\ \delta$
and $[simp]: 0 \leq \varepsilon$
shows $DP\text{-divergence}_Q\ (return\text{-qbs}\ Y\ x \otimes_{Q_{mes}} p)\ (return\text{-qbs}\ Y\ x \otimes_{Q_{mes}} q)$
 $\varepsilon \leq ereal\ \delta$
 $\langle proof \rangle$

1.2 Differential Privacy using QBS

definition $DP\text{-}qbs$ ($differential'\text{-}privacy_Q$) **where**

$DP\text{-}qbs\text{-}qbs\text{-}L:differential\text{-}privacy_Q\ M \equiv differential\text{-}privacy\ (\lambda x. qbs\text{-}l\ (M\ x))$

lemma $DP\text{-}qbs\text{-}def$:

$differential\text{-}privacy_Q\ M\ adj\ \varepsilon\ \delta \longleftrightarrow$
 $(\forall (d1, d2) \in adj. DP\text{-}inequality_Q\ (M\ d1)\ (M\ d2)\ \varepsilon\ \delta \wedge DP\text{-}inequality_Q\ (M\ d2)$
 $(M\ d1)\ \varepsilon\ \delta)$
 $\langle proof \rangle$

lemma $DP\text{-}qbs\text{-}adj\text{-}sym$:

assumes $sym\ adj$
shows $differential\text{-}privacy_Q\ M\ adj\ \varepsilon\ \delta \longleftrightarrow (\forall (d1, d2) \in adj. DP\text{-}inequality_Q$
 $(M\ d1)\ (M\ d2)\ \varepsilon\ \delta)$
 $\langle proof \rangle$

lemma $pure\text{-}DP\text{-}qbs\text{-}comp$:

assumes $adj \subseteq qbs\text{-}space\ X \times qbs\text{-}space\ X$
and $adj' \subseteq qbs\text{-}space\ X \times qbs\text{-}space\ X$
and $differential\text{-}privacy_Q\ M\ adj\ \varepsilon\ 0$
and $differential\text{-}privacy_Q\ M\ adj'\ \varepsilon'\ 0$
and $M \in X \rightarrow_Q monadP\text{-}qbs\ Y$
shows $differential\text{-}privacy_Q\ M\ (adj\ O\ adj')\ (\varepsilon + \varepsilon')\ 0$
 $\langle proof \rangle$

lemma $pure\text{-}DP\text{-}qbs\text{-}trans\text{-}k$:

assumes $adj \subseteq qbs\text{-}space\ X \times qbs\text{-}space\ X$
and $differential\text{-}privacy_Q\ M\ adj\ \varepsilon\ 0$
and $M \in X \rightarrow_Q monadP\text{-}qbs\ Y$
shows $differential\text{-}privacy_Q\ M\ (adj \rightsquigarrow k)\ (k * \varepsilon)\ 0$
 $\langle proof \rangle$

proposition $DP\text{-}qbs\text{-}postprocessing$:

assumes $\varepsilon \geq 0$
and $differential\text{-}privacy_Q\ M\ adj\ \varepsilon\ \delta$
and $[qbs, measurable]: M \in X \rightarrow_Q monadP\text{-}qbs\ Y$
and $[qbs, measurable]: N \in Y \rightarrow_Q monadP\text{-}qbs\ Z$
and $adj \subseteq qbs\text{-}space\ X \times qbs\text{-}space\ X$
shows $differential\text{-}privacy_Q\ (\lambda x. M\ x \gg N)\ adj\ \varepsilon\ \delta$
 $\langle proof \rangle$

corollary $DP\text{-}qbs\text{-}postprocessing\text{-}return$:

assumes $\varepsilon \geq 0$
and $differential\text{-}privacy_Q\ M\ adj\ \varepsilon\ \delta$
and $M \in X \rightarrow_Q monadP\text{-}qbs\ Y$
and $N \in Y \rightarrow_Q Z$
and $adj \subseteq qbs\text{-}space\ X \times qbs\text{-}space\ X$
shows $differential\text{-}privacy_Q\ (\lambda x. M\ x \gg (\lambda y. return\text{-}qbs\ Z\ (N\ y)))\ adj\ \varepsilon\ \delta$
 $\langle proof \rangle$

lemma *DP-qbs-preprocessing*:

assumes $\varepsilon \geq 0$
 and *differential-privacy*_Q $M \text{ adj } \varepsilon \delta$
 and *[measurable]*: $f \in X' \rightarrow_Q X$
 and $\forall (x,y) \in \text{adj}'. ((f\ x), (f\ y)) \in \text{adj}$
 and $\text{adj} \subseteq \text{qbs-space } X \times \text{qbs-space } X$
 and $\text{adj}' \subseteq \text{qbs-space } X' \times \text{qbs-space } X'$
 shows *differential-privacy*_Q $(M \circ f) \text{ adj}' \varepsilon \delta$
 $\langle \text{proof} \rangle$

proposition *DP-qbs-bind-adaptive*:

assumes $\varepsilon \geq 0$ and $\varepsilon' \geq 0$
 and *[qbs]*: $M \in X \rightarrow_Q \text{monadP-qbs } Y$
 and *differential-privacy*_Q $M \text{ adj } \varepsilon \delta$
 and *[qbs]*: $N \in X \Rightarrow_Q Y \Rightarrow_Q \text{monadP-qbs } Z$
 and $\bigwedge y. y \in \text{qbs-space } Y \Rightarrow \text{differential-privacy}_Q (\lambda x. N\ x\ y) \text{ adj } \varepsilon' \delta'$
 and $\text{adj} \subseteq \text{qbs-space } X \times \text{qbs-space } X$
 shows *differential-privacy*_Q $(\lambda x. M\ x \gg N\ x) \text{ adj } (\varepsilon + \varepsilon') (\delta + \delta')$
 $\langle \text{proof} \rangle$

proposition *DP-qbs-bind-pair*:

assumes $\varepsilon \geq 0$ $\varepsilon' \geq 0$
 and *[qbs]*: $M \in X \rightarrow_Q \text{monadP-qbs } Y$
 and *differential-privacy*_Q $M \text{ adj } \varepsilon \delta$
 and *[qbs]*: $N \in X \rightarrow_Q \text{monadP-qbs } Z$
 and *differential-privacy*_Q $N \text{ adj } \varepsilon' \delta'$
 and $\text{adj} \subseteq \text{qbs-space } X \times \text{qbs-space } X$
 shows *differential-privacy*_Q $(\lambda x. M\ x \gg (\lambda y. N\ x \gg (\lambda z. \text{return-qbs } (Y \otimes_Q Z) (y,z)))) \text{ adj } (\varepsilon + \varepsilon') (\delta + \delta')$
 $\langle \text{proof} \rangle$

end

2 Examples

theory *DP-QBS-Examples*

imports *DP-QBS*

Differential-Privacy.Differential-Privacy-Randomized-Response

begin

lemma *qbs-space-list-qbs-borel*[*qbs*]: $\bigwedge r. r \in \text{qbs-space } (\text{list-qbs borel}_Q)$

and *qbs-space-list-qbs-count-space*[*qbs*]: $\bigwedge i. r \in \text{qbs-space } (\text{list-qbs } (\text{count-space}_Q (UNIV :: - :: \text{countable})))$

$\langle \text{proof} \rangle$

2.1 Randomized Response

lemma *qbs-morphism-RR-mechanism*[qbs]: *qbs-pmf* $\circ$ *RR-mechanism* $e \in \text{count-space}_Q$
 $\text{UNIV} \rightarrow_Q \text{monadP-qbs } (\text{count-space}_Q \text{ UNIV})$
 $\langle \text{proof} \rangle$

lemma *qbs-DP-RR-mechanism*:

assumes $[\text{arith}]: \varepsilon \geq 0$

shows $\text{DP-divergence}_Q (\text{RR-mechanism } \varepsilon x) (\text{RR-mechanism } \varepsilon y) \varepsilon = 0$

$\langle \text{proof} \rangle$

2.2 Laplace Distribution in QBS

lemma *qbs-morphism-laplace-density*[qbs]: *laplace-density* $\in \text{borel}_Q \Rightarrow_Q \text{borel}_Q \Rightarrow_Q$
 $\text{borel}_Q \Rightarrow_Q \text{borel}_Q$
 $\langle \text{proof} \rangle$

definition *qbs-Lap-mechanism* (*Lap'-mechanism*_Q) **where**

*Lap-mechanism*_Q $\equiv \lambda e x. \text{ if } e \leq 0 \text{ then return-qbs borel}_Q x \text{ else density-qbs lborel}_Q$
 $(\text{laplace-density } e x)$

lemma *qbs-morphism-Lap-mechanism*[qbs]: *Lap-mechanism*_Q $\in \text{borel}_Q \rightarrow_Q \text{borel}_Q$
 $\Rightarrow_Q \text{monadP-qbs borel}_Q$
 $\langle \text{proof} \rangle$

lemma *qbs-l-Lap-mechanism*: *qbs-l* (*Lap-mechanism*_Q $e r$) = *Lap-dist* $e r$
 $\langle \text{proof} \rangle$

lemma *qbs-Lap-mechanism-qbs-l-inverse*: *Lap-mechanism*_Q $e x = \text{qbs-l-inverse } (\text{Lap-dist } e x)$
 $\langle \text{proof} \rangle$

proposition *qbs-DP-Lap-mechanism*:

assumes $\varepsilon > 0$ **and** $|x - y| \leq r$

shows $\text{DP-divergence}_Q (\text{Lap-mechanism}_Q (1 / \varepsilon) x) (\text{Lap-mechanism}_Q (1 / \varepsilon) y) (r * \varepsilon) = 0$
 $\langle \text{proof} \rangle$

2.3 Naive Report Noisy Max Mechanism

primrec *qbs-NaiveRNM* :: *real* $\Rightarrow$ *real list* $\Rightarrow$ *real qbs-measure* **where**

qbs-NaiveRNM $\varepsilon [] = \text{return-qbs borel } 0$ |

qbs-NaiveRNM $\varepsilon (x \# xs) =$

(*case xs of*

Nil $\Rightarrow \text{Lap-mechanism}_Q (1 / \varepsilon) x$ |

y # ys $\Rightarrow \text{do } \{x1 \leftarrow \text{Lap-mechanism}_Q (1 / \varepsilon) x; x2 \leftarrow \text{qbs-NaiveRNM } \varepsilon xs;$
 $\text{return-qbs borel } (\max x1 x2)\}$)

lemma *qbs-morphism-NaiveRNM*[qbs]: *qbs-NaiveRNM* $\in \text{borel}_Q \Rightarrow_Q \text{list-qbs borel}$
 $\Rightarrow_Q \text{monadP-qbs borel}_Q$

$\langle proof \rangle$

theorem *qbs-DP-NaiveRNM'*:

assumes *pos*[arith,simp]: $\varepsilon > 0$

and *length* *xs* = *n* **and** *length* *ys* = *n*

and *adj*: $(\sum i < n. |nth\ xs\ i - nth\ ys\ i|) \leq r$

shows *DP-divergence_Q* (*qbs-NaiveRNM* ε *xs*) (*qbs-NaiveRNM* ε *ys*) (*r* * ε) = 0

$\langle proof \rangle$

definition *adj-naive-RNM* :: *real* $\Rightarrow$ (*real list* $\times$ *real list*) **set** **where**

adj-naive-RNM *r* $\equiv \{(xs,ys). \text{length } xs = \text{length } ys \wedge (\sum i < \text{length } xs. |nth\ xs\ i - nth\ ys\ i|) \leq r\}$

theorem *qbs-DP-NaiveRNM*:

assumes *pos*: $\varepsilon > 0$

shows *differential-privacy_Q* (*qbs-NaiveRNM* ε) (*adj-naive-RNM* *r*) (*r* * ε) 0

$\langle proof \rangle$

end

References

- [1] T. Sato and S. Katsumata. Divergences on monads for relational program logics. *Mathematical Structures in Computer Science*, 33(45):427–485, 2023.