Privacy Proofs for OpenDP: Impute Uniform Float Transformation

Grace Tian

Summer 2021

Contents

1	Algorithm Implementation	1
	1.1 Code in Rust]
	1.2 Pseudo Code in Python	
2	Proof	2

1 Algorithm Implementation

1.1 Code in Rust

The current OpenDP library contains the make_impute_uniform_float function implementing the impute uniform float function. This is defined in lines 14-30 of the file impute.rs in the Git repository https://github.com/opendp/opendp/blob/21-impute/rust/opendp/src/trans/impute.rs#L14-L30

(grace) Since there's arithmetic, it seems like we have to take into account rounding that might get added into the stability?

1.2 Pseudo Code in Python

Preconditions

To ensure the correctness of the output, we require the following preconditions:

• User-specified types:

- Variables lower and upper must be of type T
- Type T must have traits float, SampleUniform, Clone, Sub(Output=T), Mul, Add, and InherentNull.

Postconditions

• A Transformation is returned (i.e., if a Transformation cannot be returned successfully, then an error should be returned).

```
def make_impute_uniform_float(lower : T, upper : T):
      input_domain = VectorDomain(InherentNullDomain(AllDomain(T)));
      output_domain = VectorDomain(AllDomain(T))
3
      input_metric = SymmetricDistance()
4
      output_metric = SymmetricDistance()
5
6
      def Relation(d_in: u32, d_out: u32) -> bool:
          return d_out >= d_in*1
      # should input to function include inherent null?
      def function(data : Vec(T)) -> Vec(T):
11
          return list(map(Uniform(lower, upper), data))
      let stability_relation = (d_in <= d_out);</pre>
14
      return Transformation(input_domain, output_domain, function,
16
      input_metric, output_metric, stability_relation)
      # TODO replace with return row_by_row_fallible
```

2 Proof

Theorem 2.1. For every setting of the input parameters (lower, upper) to make_impute_uniform_float such that the given preconditions hold, the transformation returned by make_impute_uniform_float has the following properties:

- 1. (Appropriate output domain). If vector v is in the input_domain, then function(v) is in the output_domain.
- 2. (Domain-Metric Compatibility). The domain input_domain matches one of the possible domains listed in the definition of input_metric, and likewise output_domain matches one of the possible domains listed in the definition of output_metric.
- 3. (Stability Guarantee). For every pair of elements v, w in $input_domain$ and for every pair (d_in, d_out) , where d_in is of the associated type for $input_metric$ and d_out is the associated type for $output_metric$, if v, w are d_{in} -close under $input_metric$ and $Relation(d_in, d_out) = True$, then function(v), function(w) are d_{out} -close under $output_metric$.
- *Proof.* 1. (Appropriate output domain). In the case of make_impute_uniform_float, this corresponds to showing that for every vector v of elements of type InherentNullDomain(T)

(grace) T? or InherentNullDomain???, function(v) is a vector of elements of type T.

(grace) TODO We show the type signature + nullity works

- 2. (Domain-metric compatibility). The Symmetric distance is both the input_metric and output_metric. Symmetric distance is compatible with VectorDomain(T) for any generic type T, as stated in "List of definitions used in the pseudocode". The theorem holds because for make_impute_constant, the input domain is VectorDomain(InherentNullDomain(AllDomain(T))) and the output domain is VectorDomain(AllDomain(T)).
- 3. (Stability guarantee).

To show the stability guarantee, it suffices to show that make_impute_uniform_float is a valid transformation. This is defined in Definition 5.2 in the list of definitions used in the proofs.

We assume that vectors v, w are d_{in} -close, and that $d_{in} \leq d_{out}$ because Relation (d_{in}, d_{out}) = True. By the histogram notation, this means that

$$d_{Sym}(v, w) = ||h_v - h_w||_1 = \sum_{z} |h_v(z) - h_w(z)| \le d_{in}.$$

Recall that the $make_impute_uniform_float$ transformation only changes the null values in the vectors v and w.

In make_impute_uniform_float, we sample from random variable each time the vector entry is null. If there are k nulls in v and k' nulls in w, note that $|k-k'| \leq d$ _in. This corresponds to random variables $R = (R_1, \ldots R_k)$ in f(v) that replaces the k nulls in x and $R' = (R'_1, \ldots R'_k)$ in f(w) that replaces the k' nulls in w. The elements of R and R' are are defined over Uniform(lower, upper).

We define a coupling (r, r') of random variables R and R' as follows. We set the ith element in r and r' both equal to ith element in R. Otherwise, if the ith element doesn't exist, we leave it unchanged.

This is a valid coupling since r has the same marginal distribution as R and r' has the same marginal distribution as R'. The transformations resulting from this coupling is $d_{\text{out-close}}$.

With this coupling, the symmetric distance of the replaced null values stays bounded by the symmetric distance of its original null subset: $d_{sym}(r, r') \leq d_{sym}(v_{null}, w_{null})$.

The remaining non-null values in v and z stay the same after the transformation, so the transformations resulting from this coupling $f_r(v)$ and $f_{r'}(w)$ are d_{out} -close:

$$d_{sym}(\mathbf{f}_r(v), \mathbf{f}_{r'}(w)) = \sum_{z} \left| h_{\mathbf{f}_r(v)}(z) - h_{\mathbf{f}_{r'}(w)}(z) \right|$$
$$\leq \sum_{z} |h_v(z) - h_w(z)| \leq d_{in} \leq d_{out}$$

Therefore make_impute_uniform_float is a valid transformation and thus the stability guarantee holds.